Quasi-Monte Carlo Light Transport Simulation by Efficient Ray Tracing

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Abstract

Photorealistic image synthesis can be described by a path integral. This integral is numerically approximated by summing up contributions of transport paths that connect light sources and sensors like e.g. a camera or the eye. The paths are trajectories of Markov processes, whose edges are straight lines along rays of light and whose vertices are light scattering events.

The goal of this thesis was to accelerate the simulation of light transport, find new algorithms and data structures to efficiently trace rays, and to better approximate the distribution of light by simultaneously simulating an ensemble of paths instead of single trajectories, using quasi-Monte-Carlo methods. The main results of this thesis are contributions to both computer science and mathematics.

We first present new data structures and heuristics that feature a smaller memory footprint at improved numerical precision. In addition it is possible to ray trace even massive scenes in a strictly limited, a priori fixed, memory block using rapid construction techniques that allow to rebuild the complete data structure at interactive frame rates. All efforts were combined in a unified framework that further allows one to build the acceleration hierarchy using an on demand policy and optionally balance the construction time versus the ray intersection time.

Besides finding faster ray tracing algorithms, the total number of rays to be shot was reduced by mathematical means. By simplifying complicated mathematical schemes in a non-obvious way, the time complexity of the quasi-Monte Carlo simulation process was reduced. When concentrating on the fact that the underlying Fredholm integral equation is in fact of low dimensional structure if not solved by the Neumann series, the resulting algorithms are simpler and in addition much more efficient.

The combination of these new techniques allows photorealistic image synthesis in almost realtime. The results are demonstrated by several academic and industrial applications.
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"Any sufficiently advanced technology is indistinguishable from magic" - Arthur C. Clarke
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Chapter 1

Introduction

Ray tracing [Gla89, Shi00] is the core technology of global illumination simulation for photorealistic image synthesis. While graphics hardware cannot efficiently handle all effects of global illumination [Hai06], software ray tracing additionally can compete easily with high end graphics hardware image synthesis for massive geometry [WK06]. Although known for long [App68, Whi80], only recently "realtime" ray tracing became available and even approximating global illumination effects is feasible at interactive framerates on moderately sized multicore architectures.

Image synthesis itself consists of connecting light sources and pixels by light transport paths and summing up their contributions. Vertices along these transport paths are found by tracing straight rays from one point of interaction to the next one. Many other direct simulation methods in scientific computing rely on tracing particles along straight lines, too, and usually a considerable part of the total computation time is spent on ray tracing.

The time for tracing a large number of rays can be dramatically reduced by constructing an auxiliary acceleration data structure that allows for the efficient exclusion of large portions of the scene to be intersected, instead of intersecting each ray with all objects in a scene. Compared to immediate mode rendering on a rasterizer, however, the construction time and memory footprint of the acceleration data structures were slow.

This work will focus on implementations of ray tracing as a backtracking algorithm that searches trees or directed acyclic graphs, because these schemes adapt best to general geometry. Constructing such a hierarchical data structure is easily formulated as a recursive procedure whose construction time is amortized when tracing many rays.

![Whitted style ray tracing](image1.png) ![Physically based shader feat. measured BRDFs](image2.png)

Figure 1.1: While classical ray tracing techniques were already able to cope with reflection and refraction effects, only the full simulation of light transport can result in photorealistic quality.

While full hardware support for rasterization techniques is available on low end personal computers and
even portables, a comparable solution, especially the capability to cope with dynamic scenes, was missing
in the context of ray tracing. Due to drawbacks of previously existing ray tracing implementations, such
as the unpredictable memory usage and immense pre-processing times to construct the additional data
structure along with numerical precision problems, the wide usage of ray tracing and suitable hardware is still
delayed. Note that although ray tracing systems on current multicore architectures (and even on the Cell
[MFT05, MMN06] and GPU [FS05, HSHH07, PGSS07, GPSS07] platforms) claim to offer full scalability, as
ray tracing is generally believed to be a highly parallel algorithm, this is only true for the ray traversal phase
and using simple light transport simulation techniques such as path tracing (see Section 1.1.1). With the
upcoming availability of massive multicore architectures and its proposed use to efficiently handle software
ray tracing, new problems, such as a simple, but also fully scalable hierarchy construction, arise.

At the same time, the need to overcome limitations of crude and complicated rasterization approxima-
tions [Hai06], while still offering comparable performance to GPUs, is demanded by the professional
rendering industry [CCC87, DH05, CFLB06] as well as the growing field of entertainment and games
[Oud99, SPD+04, Hur05, FGDM06, Bik07]. While many global illumination effects can be pre-computed
by ray tracing based techniques, only walkthroughs of static scenes using the additional data stored on
the surfaces (e.g. lightmaps/pre-baked lighting or photonmaps [Jen96]) are possible, and approximations
(e.g. ambient occlusion [Lan02, IKSZ03] or deep shadow maps [LV00]) are needed to support fully dynamic
scenes or unpredictable user interaction with the geometry. Still, all pre-computation will become very dif-
icult or even fail for highly specular and refractive materials. Although it is possible to combine the best of
ray tracing and rasterization for some scenarios [Kel97, Sti06], the advantage of a completely ray tracing
based implementation allows for unmatched picture quality and freedom in material design and lighting.

Figure 1.2: Modeling/rendering/compositing-pipeline as used in product design and the movie industry. The
scene is usually modeled by free form surfaces, triangulated, and linked to different material properties.
By approximating real world lighting by a high dynamic range environment map, the mesh can be illuminated
and composited into existing background shots.

1.1 Light Transport

Ray tracing based image synthesis has come a long way. Starting with the simulation of simple materials,
Whitted style ray tracing [Whi80] was introduced to cope with perfectly refractive and reflective material
properties. Distributed ray tracing [CPC84, Coo86a, SW92] incorporates the direct simulation of glossy
surfaces and camera lenses, making a "real" integration step necessary as the probability distribution function
of materials cannot be described solely by Dirac distributions anymore, resulting in the need for multiple
samples per scattering event on the surfaces to properly approximate the illumination.

By introducing the concept of integral equations to computer graphics [Kaj86], a mathematical framework
of the underlying light transport was presented. In the following a short overview of the evaluation, along
with possible models to describe material properties will be presented. For a full introduction, see [Kaj90,
Arv93, Gla95].

The change of the radiance $L$ in a point $x$ along a unit direction $\omega$ is

$$\frac{\partial}{\partial \omega} L(x, \omega) = L(x, \omega) + \sigma_s(x) \int_{S^2} f_p(\omega, x, \omega') L(x, \omega') d\omega' - \sigma_t(x) L(x, \omega), \quad (1.1)$$

where the first term $L(x, \omega)$ on the righthand side is the volume emission. The second term describes the
increase due to inscattering from all directions of the unit sphere $S^2$, where $f_p$ is the phase function accounting
for the angular distribution and \( \sigma_s \) is the scattering coefficient depending on the location \( x \). In order to solve the integro-differential equation, boundary conditions are required. These are given by the Rendering Equation
\[
L(x, \omega) = L_{e;\partial V}(x, \omega) + \int_{S^2} f_s(\omega, x, \omega') L(x, \omega') |\cos \theta_x| d\sigma(\omega')
\] (1.2)
governing light transport on the surface \( \partial V \) [Kaj86]. Here, \( f_s \) is the Bidirectional Scattering Distribution Function (BSDF), which describes the scattering behavior at \( x \in \partial V \), and \( \cos \theta_x \) is the cosine of the angle between direction \( -\omega' \) and the surface normal in \( x \). \( L_{e;\partial V} \) is the emission on the boundary.

Usually, Equation 1.1 is integrated along straight light rays to the next surface interaction point \( x_S = h(x, -\omega) \), which is found by a ray casting function \( h \). This approach yields a Fredholm integral equation of the second kind, which can be efficiently handled by Monte Carlo techniques.

For the scope of this thesis we focus on vacuum transport, i.e. \( \alpha = s = 0 \), which results in
\[
\frac{\partial}{\partial \omega} L(x, \omega) = 0
\]
Consequently, the radiance does not change along straight rays of light in vacuum, and the boundary condition in Equation 1.2 completely describes light transport.

This formulation is the most commonly used in current rendering systems as it allows for a reasonably fast implementation using ray tracing, while still incorporating the most important physical effects.

### 1.1.1 Path Integral Formulation

The previously presented local description of light transport can be recursively expanded to obtain an integral over the abstract space of complete transport paths [Vea97, PKK00, Kol04a], allowing for the evaluation of the integral by using Markov chain simulation (see Section 3.4).

The path space \( P \) can be modeled as the union of spaces containing paths of a specific finite length, i.e.
\[
P := \bigcup_{k \in \mathbb{N}} P_k \text{ where } P_k := \{ \bar{x} = x_0 \ldots x_k : x_i \in \mathbb{R}^3 \}.
\]

For each of these spaces a product measure \( \mu_k \) is used, defined for a set \( M_k \subseteq P_k \) by
\[
\mu_k(M_k) := \int_{M_k} d\lambda(x_0) d\lambda(x_1) \cdots d\lambda(x_k).
\]
of the corresponding Lebesgue measure on the surface, \( d\lambda(x) \).

The measure \( \mu \) of a set \( M \subseteq P \) then is the natural expansion of those disjoint spaces’ measures
\[
\mu(M) := \sum_{k \in \mathbb{N}} \mu_k(M \cap P_k).
\]
In this context, the sensor response \( I_j \) of pixel \( j \) can be expressed as an integral over \( P \),
\[
I_j = \int_P f_j(\bar{x}) d\mu(\bar{x}),
\] (1.3)
where \( f_j \) is called the Measurement Contribution Function. We now write Equation 1.2 as Three Point Form [Vea97]:
\[
L(y \rightarrow z) = L_\epsilon(y \rightarrow z) + \int_{\mathbb{R}^3} L(x \rightarrow y) f(x \rightarrow y \rightarrow z) G(x \leftrightarrow y) V(x \leftrightarrow y) d\lambda(x).
\] (1.4)
where the following abbreviations are used:

- **Three point scattering function**
  \[
f(x \rightarrow y \rightarrow z) := f_s(\bar{y} \hat{z}, y, \bar{x} \hat{y})
\] (1.5)

- **Source radiance distribution**
  \[
  L_\epsilon(x \rightarrow y) := L_{\epsilon;\partial V}(x, \bar{x} \hat{y})
\] (1.6)
\( G(x \leftrightarrow y) := \frac{|\cos \theta_x| |\cos \theta_y|}{||y - x||^2} \)  \hspace{1cm} (1.7)

with \( \theta_x \) and \( \theta_y \) the angles between \( \vec{x} \vec{y} \) and the surface normals at the respective points.

- Visibility function (evaluated through ray tracing)
\[
V(x \leftrightarrow y) = \begin{cases} 
1 & \text{if } ||y - x|| \leq ||h(x, \vec{x} \vec{y}) - x|| \\
0 & \text{otherwise}
\end{cases}
\]

Finally, we need a function \( \psi_j(x_{k-1} \rightarrow x_k) \) to describe the sensor response of a pixel \( j \) to radiance arriving from \( x_{k-1} \) at the point \( x_k \) on the image plane. Along with the recursive expansion of the three point form (see Equation 1.4) this yields the measurement contribution function for a path \( \vec{x} = x_0 \ldots x_k \) beginning on a light source and ending on the image plane:

\[
f_j(\vec{x}) := L_e(x_0 \rightarrow x_1)G(x_0 \leftrightarrow x_1)V(x_0 \leftrightarrow x_1) \\
\prod_{l=1}^{k-1} (f(x_{l-1} \rightarrow x_l \rightarrow x_{l+1})G(x_l \leftrightarrow x_{l+1})V(x_l \leftrightarrow x_{l+1})) \\
\cdot \psi_j(x_{k-1} \rightarrow x_k) \hspace{1cm} (1.8)
\]

**Numerical Algorithms**

Although the formulation seems to be fairly complicated, possible algorithms to evaluate the presented path integral are as simple as summing up all contributions of transport paths that connect light sources and sensors.

- Direct simulations in the form of Light and Path Tracing: The Light Tracing technique simply emits photons from the light sources and traces particle paths/scattering events until (hopefully) an intersection with the camera lens is found. Path Tracing inverts this technique as it starts the paths from the camera lens to find an intersection with the light sources. Both techniques can be refined by trying to directly connect to the camera (or light source respectively) at each scattering event.

- Bidirectional Path Tracing [LV93, VG94, VG95, LW96, Ve97] and its special cases Instant Radiosity [Kel97] and Robust Global Illumination [KK04, Kol04a, Raa05] will start paths from both the camera lens and the light sources and search for "valuable" connections of the subpaths or simply connect all vertices of the subpaths with non-zero contribution (see Figure 1.3).

- Metropolis Light Transport [VG97, Ve97, PKK00, KSKAC02, Sei06] is the most advanced technique of them all, as it can smear "valuable" paths to allow for an efficient simulation of extremely difficult light transport problems (e.g. light transmitted through narrow geometry like a key hole) to perfectly adapt to the actual distribution of light.

In a mathematical context the stochastic creation of the transport paths corresponds to the simulation of Markov chains.

Still, finding all of the important light paths that actually deliver significant contributions to the final image is a difficult task [Ve97, Kol04a]. See Section 4.4.1 for more details on an actual implementation.

### 1.1.2 Scattering

To describe material properties and define transition probabilities on the surfaces for the Markov chain simulation one must model the Bidirectional Scattering Distribution Function (BSDF). This work will concentrate on a subset of BSDFs, the Bidirectional Reflectance Distribution Function (BRDF) [NRH+77, LW94], which does not cope with translucent materials. A BRDF is defined as
1.1. Light Transport

Figure 1.3: Sampling transport path space by bidirectional path tracing. Trajectories from the eye and the light sources are generated by Markov chains (see Section 3.4) and connected to determine the transported amount of light (illustration provided by [Rze05]).

\[
f_r(\omega_o, x, \omega) := \frac{dL_o(\omega_o)}{L_i(\omega) \cos \theta} d\sigma(\omega)
\]

and describes the amount of radiance from incoming direction \(\omega_i\) to be scattered in direction \(\omega_o\) at point \(x\). \(dL_o(\omega_o)\) is the differential amount of radiance reflected in the outgoing direction and \(L_i(\omega)\) the amount of incoming radiance. \(\cos \theta\) describes the angle between the incoming direction \(\omega_i\) and the surface normal.

A physically plausible BRDF fulfills at least two constraints: It must respect the reciprocal nature of the reflection of light (Helmholtz-reciprocity), i.e.

\[
f_r(\omega_o, x, \omega) \equiv f_r(-\omega_i, x, -\omega_o).
\]

Furthermore, it must not amplify the incoming energy by guaranteeing

\[
p(x, \omega) = \int_{S^2} f_r(\omega_o, x, \omega) |\cos \theta_o| d\omega_o \leq 1, \forall x, \omega,
\]

i.e. not reflecting more light than incident.

**Analytical Reflection Models**

To describe \(f_r\), a vast amount of analytical models were created to cope with existing real life materials. While early models such as the Phong model [LW94, LFTG97] (based on a shading model proposed in [Pho75]) were mainly based on empirical experiments of simple materials (such as plastic) and the efficient computational evaluation, currently used models as the Ashikhmin-Shirley model [AS00] allow one to model a large variety of materials at the cost of increased computation time. Some models even use microfacet theory to model the BRDF by defining a distribution of position and orientation of a set of small microfacets on the actual surface to simulate a micro structure (e.g. to immitate the appearance of brushed metal). One
classical example is the Cook-Torrance [CT82] model. A more recent example was proposed in [APS00], and generally all half-vector based BRDF models are somehow related to microfacet theory.

While analytical models allow to describe a material within few parameters and mostly offer an easy way to employ importance sampling [Sob91, Vea97] by inversion techniques, the biggest problem is the insufficient approximation of more complex real life materials as can be seen in [Ash06] and the amendments of [NDM05]. Note that common analytical models are of low dimensional structure (usually 1-dimensional for isotropic and 2-dimensional for anisotropic materials) due to symmetry assumptions, while the BRDF function is in fact 4-dimensional.

Another negative aspect found in many available models is the problem of lobes intersecting the surface. If the incoming direction is close to parallel to the surface, the resulting lobe (of specular materials) will only partially be situated above the surface, resulting in rays penetrating the surface. Possible "solutions" are the mirroring of these rays on the surface plane (which clearly changes the appearance of the BRDF model) or assuming a zero contribution and discarding these rays. While this does not sound like much of a problem, this actually means a loss of energy for some of the incoming directions and can be clearly spotted in visual comparisons. See [Rze05] for more details and, more important, solutions for such problems. [EBJ+06] proposes an additional solution based on the halfway vector disc, however, sacrificing reciprocity.

Measured Reflection

By using a Gonioreflectometer or similar devices, it is possible to measure the BRDF in a region (ideally one point) of a real life material. While the technique is rather advanced by now, problems for highly specular/glossy materials are still present. Available datasets can be received from MERL in isotropic [MPBM03] and anisotropic [NDM05] form for non-commercial usage (see www.merl.com/brdf/ for details). Cornell university also offers BRDF datasets, some of them even featuring 31 or 65 wavelength samples (but at a decreased overall resolution) [oCG]. The BRDFs used in this work were kindly provided by the Spheron VR AG.

While measured BRDFs usually represent a full 2-dimensional (4-dimensional in the anisotropic case) approximation of existing material BRDFs, there are also some drawbacks. Especially anisotropic datasets can become exceptionally large (several 100MB are not uncommon), while isotropic datasets can fit in less than 10MB if the underlying material features only fair complexity. However, an accompanying runtime interpolation of the BRDF data should be considered to avoid artifacts.

While mapping to existing analytical models seems to be an option to save memory and discard the interpolation, it mostly fails to exactly reproduce subtle (but still important) details of the function [Ash06, NDM05].

Other possibilities are the transformation/projection into other bases (see e.g. [Mat03]) or the factorization of the data [LRR04], which additionally allows for importance sampling.

Related to BRDFs is the Bidirectional Texturing Function (BTF) [MMS05]. It holds the BRDF over a set of points, similar to a (high-dimensional) texture that can be assigned to a surface. Note that this obviously increases the size of a single dataset again, often demanding a decrease of the local resolution of the BRDF to be able to keep multiple BTFS in memory at the same time.

One additional drawback is the missing user interaction when working with measured BRDF data. It is not possible to tweak an existing BRDF in the fashion of an analytical model that often features intuitive parameters to slightly change the look of a material. However, a simple combination of existing BRDFs is presented in [Mat03]. While omitting physical correctness, [Bjo06] proposes the "drawing" of BRDFs to allow full artistic freedom. A physically correct implementation of artistically designed BRDFs can be found in [CPK06]. Another possibility of editing BTFS was presented in [KBD07].

Note that although the general concern in rendering is currently focused on fast ray tracing (as this was long considered to be the bottleneck of every renderer), a modern rendering system working with measured BRDFs can easily spend $\frac{1}{3}$ of the total time evaluating the BRDFs (including the transformation of ray directions and importance sampling). While it seems to be faster to use analytical models instead, a complex model like [AS00] will take far more cycles to evaluate than a properly implemented measured BRDF library.
1.2 Structure of the Thesis

The thesis first separately treats ray tracing and the simulation of Markov chains. Then the results are combined in applications.

Chapter 2 concentrates on efficient ray tracing. To accelerate ray tracing, an additional data structure decreases the number of necessary intersection calculations with the objects of a scene dramatically. The chapter presents a new data structure, accompanied by new construction schemes, new partitioning heuristics, and a numerically stable implementation that allows for a simple, yet exceptionally fast construction phase and competitive ray traversal performance. A unified framework merges ray traversal and in-place construction in an a priori fixed memory block and proposes new algorithms to fully amortize the time to build the data structure.

Chapter 3 uses ray tracing to simulate paths of light for photorealistic image synthesis. By introducing the principle of simulating many paths simultaneously, i.e. an ensemble of particles, quasi-Monte Carlo Markov chain simulation becomes more efficient. By sorting the states upon scattering events at the surfaces, a more uniform quasi-Monte Carlo sampling decreases the "variance" of the results. An analysis of a classical algorithm reveals a simpler and more efficient simulation scheme.

Chapter 4 demonstrates the improvements that can be achieved in computer graphics by using the research results of the previous two chapters.
Chapter 2
Ray Tracing

This chapter will present several ray tracing engine approaches, culminating in a unified kernel, that will pick the best of all techniques while keeping the implementation effort low and still being flexible enough to inherit refinements necessary to efficiently support a diverge set of scenarios like fully dynamic scenes, massive geometry, or strict memory limitations.

First, an overview over currently popular triangle intersection algorithms is provided and analyzed in terms of performance and numerical robustness, along with proposals to further increase performance and decrease the memory footprint. Secondly, a new acceleration data structure, the bounding interval hierarchy, featuring exceptionally low memory footprint and construction time is presented. Third, different heuristics to construct spatial and object list partitioning hierarchies are presented, culminating in new proposals for \(kd\)-trees and bounding volume hierarchies. Finally, a new algorithm that can construct hierarchies in-place (using no temporary memory at all) and working in a priori fixed memory limitations is proposed. The previous findings are then combined into a unified framework that allows to easily incorporate several techniques necessary to support fully dynamic or massive scenes at maximum performance and minimum memory footprint. By using a tight coupling of ray traversal and hierarchy construction, new algorithms to further decrease the total time to image are provided. The chapter is accompanied by new findings to overcome several numerical problems and general implementation hurdles.

Accelerated Ray Tracing

When tracing huge numbers of rays for image synthesis it is obviously not efficient to intersect each ray with all objects in the scene in order to find the closest point of intersection for each ray. It amortizes to construct additional data structures that allow one to exclude most of the objects from actual intersection testing.

To obtain such a data structure either the space containing the objects (see Section 2.2) or the list of objects (see Section 2.3) can be partitioned. While it is obvious to have a hybrid (see Section 2.5), the two pragmatic approaches are briefly sketched in Figure 2.1.

Top-Down Construction of Hierarchies

As the usage of uniform and regular structures is commonly inferior to hierarchies under practical concerns, the steps of a general recursive procedure to construct spatial hierarchies in a top-down manner are as follows:

**Termination:** The termination condition is a combination of controlling memory consumption and considering the amortization of appending more nodes to the tree to reduce the total amount of ray-triangle intersections later on. Usually, one clause terminates the tree by depth, which only results in an exponential bound on the number of nodes and thus memory. Another common clause avoids to further subdivide the tree if not enough references are provided. This clause saves memory and balances node versus leaf processing time.
Figure 2.1: Taxonomy of partitioning schemes. While non hierarchical data structures allow for rapid constructions, a hierarchical scheme better adopts to the geometry. Partitioning the object list results in linear memory usage, while spatial partitioning copes better with complex surface structures. Note that all commonly used data structures utilize axis aligned bounding boxes for culling.

Split selection: A heuristic is used to split the current volume element into two or more elements. Both the efficiency of the construction and the efficiency of ray tracing later on are heavily influenced by how the split is selected. Heuristics range from simple criteria as splitting axis-aligned volumes in the middle along the longest side, over global criteria [WK06] to more costly methods like the surface area heuristic (SAH) [GS87, Hav01, Wal04, WBS06] (Section 2.4 will present more details).

Classification: For each object referenced in the object list it is determined which of the new volumes it intersects.

Node creation: For each new child a list of references to objects is assembled according to the classification in the previous step. The case of concern is usually the reference duplication caused by objects that are assigned to more than one child, as it is difficult to a priori predict their number, an issue that will be addressed in Section 2.5.1.

Post-processing: After recursively processing the children, a post-processing step can perform several tree optimizations, such as reordering or quantization. Examples are found in [Kel98, App. A] and [Wal04].

2.1 Triangle Boundary Representation

The core routine needed in any ray tracing engine is the intersection test of a ray with the geometric primitives in the scene. As the boundary of the objects in a scene is usually modeled using different descriptions for the geometry such as free form surfaces (especially NURBS), voxelized data-sets, or even fractal geometry, the inner loop of a generalized ray tracer cannot be optimized easily due to its high complexity. Restricting to approximations using a simple single primitive, however, allows for a tight inner loop that can be optimized to exploit the used hardware. As triangles allow for an intersection test with small computational effort and triangulation algorithms are available for most of the relevant complex primitives, the presented system sticks to using triangles as its basic scene elements.

Although this is the way to go for almost all current realtime ray tracing implementations (and even most of the big off-line rendering-packages) it must be noted that free form surfaces might be feasible for future ray tracing implementations, as modern hardware platforms tend to increase the raw processing power while at the same time also increasing latencies for memory fetches. As such, the triangulation of free form surfaces to multiple smaller triangles increases the total amount of memory used for the geometry (and
2.1. Triangle Boundary Representation

Thus memory traffic and cache pollution during ray tracing, but also decreases the number of operations needed to compute the intersection with a ray segment, resulting in idle cores waiting for new geometric data instead of doing useful work.

2.1.1 Survey of Ray-Triangle Intersection Tests

Current ray tracing systems rely on an acceleration structure, thus minimizing the necessary intersection tests with triangles to usually less than a dozen (ideally just one) primitives per ray. It must be noted that the strategy of measuring ray-triangle intersection performance must always be seen in this context, especially when thinking about early-out conditions that can skip non intersected triangles as early as possible to save a large part of unnecessary calculations.

Another important issue can be the numerical precision of the chosen triangle test, as it is fairly common that rays will slip "through" the edges of connected triangles due to numerical issues, especially when using algorithms working in (32-bit) floating point precision. Though this seems to be not much of an issue in the context of computer graphics as the resulting artifacts mostly vanish with increasing sample numbers, the problems for simulations can be immense (think of water in a triangulated bottle using a particle method for simulation).

Further it must be pointed out that all presented algorithms allow to be transformed into each other, so several implementation variations of the tests and their combinations are possible (see [KS06] for more details). In the following subsections only the Plücker and barycentric coordinate test are presented in great detail (as they are amongst the most popular at the moment and the remaining algorithms are easily transformed into these), for an excessive description of the mentioned (and even more) ray-triangle tests, see [LAM05]. Another nice result is given in [BP02] where some possibilities are presented to optimize single ray tracing in the context of SIMD execution.

Plücker Coordinates

The basic idea of the Plücker test [Jon00] is to store the ray coordinates (starting point and direction) and the triangle edges in Plücker coordinates. This transformation allows for an easy intersection test by testing the orientation of the edges towards the ray for equality, along with the ability to solve the equations for the three edges in parallel, thus allowing to use SIMD execution even for single rays. However, computing the intersection distance requires a separate step to solve the ray-triangle-plane intersection equation.

Conversion of a line through 3-dimensional-points $X$ and $Y$ to the Plücker coordinates 6-dimensional-vector $L$ is defined by

$$L = (L^x, L^+) := (X \times Y, X - Y).$$
The orientation of two lines $L_1$ and $L_2$ is given by

$$\alpha = L_1 \times L_2 := L_1^+ L_2^- + L_2^+ L_1^-.$$ 

Defining $a, b, c$ to be the edges and $r$ the ray (all converted to Plücker representation), testing for intersection of the line through $r$ with the triangle defined by $a, b, c$ uses

$$\mu_1 := (r \cdot a > 0) \land (r \cdot b > 0) \land (r \cdot c > 0).$$

Triangles that must be two sided (e.g., transparent or non watertight meshes) require a second check to test if all orientations are negative:

$$\mu_2 := (r \cdot a < 0) \land (r \cdot b < 0) \land (r \cdot c < 0).$$

Note that the case in which all orientations/inner products deliver zero is a ray being coplanar to the triangle (and thus no intersection).

![Figure 2.3: Illustration of the Plücker test: $a, b, c$ are the lines defined by the triangle edges, $r_1$ intersects the triangle, whereas $r_2$ does not. Red and blue arrows indicate the orientation of the edges towards the corresponding ray.](image)

If $\mu_1$ (or one of the conditions $\mu_{1,2}$ respectively) is fulfilled, the last step is to solve the ray-triangle-plane equation to obtain the distance of the ray’s original starting point $r_{fr}$ with respect to the length of the original ray direction $r_{dir}$ to the point of intersection on the triangle-plane $P$ (given through a point $P_p$ of the triangle and the normal of the triangle $P_n$):

$$\lambda := \frac{(P_p - r_{fr}) \cdot P_n}{r_{dir} \cdot P_n}.$$ 

Optionally, the necessary data that must be stored for each triangle can be minimized by using

$$P_d := P_{p_a}P_{p_b}"
2.1. Triangle Boundary Representation

scaling $P_n$ and $P_d$ value, so the $i$'th component of $P_n$ is constant). This results in a total of 21 floats.

As demonstrated by [CA97], the redundant memory increase can be reduced by storing per-edge-information rather than per-triangle-information. Neighboring triangles also share same edges, so a triangle can be represented by indices pointing to the three edge-vectors (one integer index vs. six floats each). Additionally the authors implemented a caching system that tries to save computed information about the edge-orientations towards the current ray for neighboring triangles (one bit for the sign bit of the inner product of that edge plus 31 bits for an ID of the current ray resulting in one additional integer variable per face). Unfortunately, the benefit of this solution in the context of using an additional acceleration data structure for ray tracing is not convincing, though this scheme could still be exploited to increase numerical precision for the triangle intersections (see Section 2.1.3).

A slightly differing implementation of the test (mainly optimized for 4xSIMD hardware) has been presented in [Ben06]: Instead of storing transformed triangle data, the necessary transformations to Plücker coordinates are amortized by intersecting several rays at once. As the rays of the bundle additionally feature the same origin and the tested triangle must be oriented clockwise towards the rays, a large part of the calculations must only be computed once per bundle and triangle, while the rest can be done using SIMD for the multiple rays. A more general implementation allowing for both clockwise and counter clockwise orientations and/or differing ray origins is possible but at the same time less efficient.

To summarize, the Plücker test seems to be especially useful when considering single rays only and wanting to use SIMD optimizations, although this comes at the price of increased memory usage. The second option is to use ray bundles for SIMD execution and not doing any pre-calculation at all as proposed in [Ben06], though the ray properties must be restricted to allow for a fast implementation.

Two features that are generally seen as advantages of the Plücker coordinate test, namely numerical precision (only signs are checked in the first tests) and the possibility for fast early-outs, did not hold in the context of this work’s comparisons. The amount of rays slipping through closed geometry is actually the highest of all tests and the delayed ray-triangle-plane intersection test is of little use in the context of using an additional acceleration structure, as will be explained at the end of this section. Interestingly enough the precision problem did not arise as distinct for the [Ben06] variant, a phenomenon also described in detail at the end of this section.

Barycentric Coordinates

Different from the Plücker coordinate test, this algorithm first tests for an intersection of the ray segment with the plane of the triangle resulting in the distance $\lambda$. If this initial test is successful, the resulting point of intersection $H$ is used to determine whether it lies inside the triangle (thus an intersection) or outside (no intersection).

By assuming that the areas of the three sub-triangles $A_0 := \text{Area}(H, P_{p_1}, P_{p_2})$, $A_1 := \text{Area}(H, P_{p_2}, P_{p_3})$, $A_2 := \text{Area}(H, P_{p_3}, P_{p_1})$ must sum up to the overall triangle area $A_t := \text{Area}(P_{p_1}, P_{p_2}, P_{p_3})$ if $H$ is located inside the triangle, one can formulate the following equations:

$$A_t = A_0 + A_1 + A_2, \quad A_0 = u A_t, \quad A_1 = v A_t, \quad \text{and} \quad A_2 = w A_t$$

with

$$0 \leq u, v, w \leq 1, \quad \text{and} \quad u + v + w = 1.$$ 

Defining that $a := P_{p_1} - P_{p_3}$, $b := P_{p_2} - P_{p_3}$, $c := P_{p_3} - P_{p_2}$, $d := H - P_{p_3}$, $e := H - P_{p_1}$ and $f := H - P_{p_2}$, one can calculate the barycentric coordinates $u, v, w$ by solving:

$$u = \frac{u A_t}{A_t} = \frac{A_0}{A_t} = \frac{\frac{1}{2} ||b \times f||}{\frac{1}{2} ||a \times b||}.$$ 

$$v = \frac{v A_t}{A_t} = \frac{A_1}{A_t} = \frac{\frac{1}{2} ||c \times d||}{\frac{1}{2} ||a \times b||}, \quad \text{and}$$

$$w = \frac{w A_t}{A_t} = \frac{A_2}{A_t} = \frac{\frac{1}{2} ||a \times e||}{\frac{1}{2} ||a \times b||}.$$
By verifying that \( u + v + w = 1 \), \( H \) is valid or not. Note that the equations can be simplified, since the constant \( \frac{1}{A_f} \) can be multiplied with \( a, b, c \) in a pre-calculation step.

An interesting simplification of this algorithm by [Bad90] is the projection of the triangle onto a principal plane and testing the barycentric coordinates in the simple 2-dimensional case. Since a random pick of the projection case could lead to numerical instability, the projection plane should be chosen as orthogonal as possible with respect to the triangle normal. Projecting the point of intersection \( H \) and the edge-vectors \( a, b, c \) onto this plane (by simply “discarding” the \( i \)th component), the equations simplify to their 2-dimensional case with \( d, b', c' \) being the projected edges and \( H' \) the projected point of intersection along with the resulting sub-triangle edges \( d', e', f' \) and the corresponding areas \( A_0, A_1, A_2, A_3 \). The variables \( j, k \) of the edges represent the vector-component-indices not being discarded during the projection.

\[
\begin{align*}
u &= u \frac{A_1}{A_f} = \frac{A_0}{A_f} = \frac{1}{2} \|(b'_j, b'_k, 0) \times (f'_j, f'_k, 0)\| = \frac{\|b'_j f'_k - b'_k f'_j\|}{\|d'_j b'_k - d'_k b'_j\|} \\
v &= v \frac{A_2}{A_f} = \frac{A_1}{A_f} = \frac{1}{2} \|(c'_j, c'_k, 0) \times (d'_j, d'_k, 0)\| = \frac{\|c'_j d'_k - c'_k d'_j\|}{\|d'_j b'_k - d'_k b'_j\|} \\
w &= w \frac{A_3}{A_f} = \frac{A_2}{A_f} = \frac{1}{2} \|(d'_j, d'_k, 0) \times (b'_j, b'_k, 0)\| = \frac{\|d'_j e'_k - d'_k e'_j\|}{\|d'_j b'_k - d'_k b'_j\|} \\
\end{align*}
\]

Since we are generally only interested in two of the three barycentric coordinates (since the third is implicitly given by \( u + v + w = 1 \)), the calculation of \( w \) can be discarded. Note though that \( u + v + w = 1 \) can only hold if \( H \) lies inside the triangle. To get the missing case of \( H \) being outside, a closer look at the algorithm reveals that this can only be the case if the signs of \( (b'_j f'_k - b'_k f'_j), (c'_j d'_k - c'_k d'_j) \) are not equal to the sign of \( (d'_j b'_k - d'_k b'_j) \). A geometrical interpretation: One or both edges \( d, f \) of the sub-triangles \( (b, f, e), (c, d, f) \) are in reverse direction compared to the clockwise orientation of the edges \( a, b, c \) defining the main triangle \( P \) and thus have to point outside the triangle. This leads to the following simplification:

\[
\begin{align*}
u &= \frac{b'_j f'_k - b'_k f'_j}{d'_j b'_k - d'_k b'_j} \\
v &= \frac{c'_j d'_k - c'_k d'_j}{d'_j b'_k - d'_k b'_j}.
\end{align*}
\]
By testing for \((u \geq 0), (v \geq 0)\) and \((u + v \leq 1)\) to check for a valid point of intersection \(H^t\) a complete intersection test is obtained, while \(1/(a_1b'_2 - b'_1d_2)\) can once again be pre-multiplied onto \(b'_1, c'_1\) to save some more multiplications and divisions, while also providing a reduced memory footprint.

This leads to an overall data usage of three floats for the plane normal, three floats for \(P_{p0}\) (or \(P_{pJ}\) plus \(P_{p0}\) discarding the \(i\)th component) and two floats each for the two projected edge vectors \((3 + 3 + 2 + 2 = 10\) floats plus an additional byte needed to store the chosen projection case making a total of 41 bytes). This can further be decreased by scaling the normal such that the \(i\)th component is always one (and thus must not be stored explicitly) and a different storage of the two bits used for the projection case (which will be shown later on), bringing the memory usage down to nine floats again. Note that the resulting \(u, v\)-coordinates can be used later on for interpolations of texture-coordinates, vertex-normals or similar information.

A large number of implementation variants (for example, reducing the memory footprint by quantization or omitting the division for the ray-plane test completely) can be found in [Wäc04].

**Arenberg Test**

The triangle data as used by [Are88] is defined by twelve floats (one of the original vertices plus nine additional floats of pre-transformed data). One of the positive aspects of the algorithm is the delay of the division for the ray-plane intersection, as it is actually the last operation of the test and only needed to test if the plane intersection is inside the active ray segment. Note that in this optimized variant it is only possible to test for clockwise (towards the ray) oriented triangles. When trying to optimize the amount of data used and allowing for both clockwise and counter clockwise oriented triangles the algorithm actually transforms into a variant of the Badouel test.

Another variant of this algorithm is proposed in [SWW+04, WSS05]. The authors first transform the triangle to be aligned to the Cartesian coordinate system to save some calculations, but then must use the inverse transform for the ray during intersection. While the original Arenberg test is numerically rather stable, this variant suffers from tremendous precision problems due to the excessive transformations that must be used (a problem inherent to instancing (see Section 2.1.3), as it is used in the context of these papers).

**Chirkov Test**

Chirkov’s test [Chi05] needs to store only nine float values (when optimized, plus additional two bits for the projection axis of the triangle plane, thus comparable to Badouel), but is defined to work on origin and endpoint of a ray. Thus infinitely long rays are problematic, though under practical concerns this is not much of a problem as the ray can always be clipped to the scene’s bounding box first to determine a stable endpoint. On the bright side, the test can omit divisions completely to just deliver a boolean style ray-triangle intersection, however, in the context of ray tracing multiple objects the division has to be performed nevertheless to calculate the distance along the ray to determine the closest primitive and/or the point of intersection to continue ray tracing with second generation rays.

**Segura/Feito Test**

A rather different approach as proposed by [YNF90] (and based on a quite similar theory also published in [O’R98]), and refined by [SF01] in a more efficient form, uses the signed volumes of three tetrahedrons formed by the triangle vertices (thus only nine floats of data are used by this test) and the ray origin and endpoint. In fact this is actually closely related to the Plücker coordinate test [Eri07]. It was first proposed on the knowledge of an error-free 4x4 matrix determinant calculation [Yam85, OR005], so the intersection computation was designed to be exceptionally robust as it only uses numerically stable computed signs to determine if an intersection takes place. The point of intersection itself (along with the distance along the ray) has to be calculated by solving the classic ray-plane equation after the intersection test, which made the implementation in the context of this work’s ray tracing system too slow to be of practical use. Still, the algorithm is very interesting if only a boolean intersection test is needed and additionally can be implemented to be completely error free.
Kensler/Shirley Test (Möller/Trumbore Test)

[KS06] proposed to find an optimal ray-triangle intersection test by using genetic/evolutionary algorithms to determine a test that uses either a minimum number of instructions or optimizes the runtime of the algorithm (when used in the context of ray bundles). The found algorithm is related to the Möller/Trumbore test [MT97] as Shirley states on his web-blog, but was found to be rather slow (compared to the other presented algorithms in this section) in the context of using it in a real ray tracing system using an additional acceleration data structure. Thus the published implementation was successively hand-tuned, which resulted in a variant of the original barycentric coordinate test.

This again states the importance of the context in which the ray triangle test is used, as even different acceleration data structures (and the "quality" of the heuristic used to construct the data structure) can dramatically influence the amount of triangles forwarded to the triangle intersection loop, as well as the relative probability of early-outs and the actual fully calculated ray-triangle intersections.

2.1.2 Investigation of Numerical Precision and Performance

The following figures demonstrate some representative scenarios for the different tested triangle tests, omitting the Segura/Feito test (full precision possible, although performance not satisfying in the context of using an additional data structure) and Kensler/Shirley test (performance not satisfying in the context of using an additional data structure), while the original Plücker coordinate test was only included due to its current popularity:

- **Arenberg (CCW):** Optimized Arenberg test (thus only usable for single sided, counter clockwise oriented triangles), working on pre-transformed data (12 floats)
- **Chirkov:** Standard Chirkov test, working on pre-transformed data (9 floats)
- **Plücker:** Standard Plücker coordinate test, working on pre-transformed data (21 floats)
- **Org. Data Plücker (CCW):** Optimized Plücker coordinate test as proposed in [Ben06] (thus only usable for single sided, counter clockwise oriented triangles), working on the original triangle data (9 floats)
- **Badouel:** Standard Badouel test, using the scaling of the i'th component of the normal to one and working on pre-transformed data (9 floats)
- **Org. Data Barycentric:** Variant of the standard barycentric coordinate test, working on the original triangle data (9 floats)

Ray tracing performance is represented as percentage (100 being the fastest algorithm, the Arenberg test) of the total ray-hierarchy intersection time (not just the actual triangle tests). The specific scenes and views (and thus numerical errors) were picked as representatives for a large amount of all tested scenes and scales. Note that the test proposed by [Ben06] (Org. Data Plücker (CCW)) can be additionally optimized for ray bundles featuring the same ray origin at the cost of flexibility. The results for the numerical precision were equal for both kd-tree and BIH (see Section 2.3.1) data structures (as expected when using stable construction and traversal implementations) while the overall performance was averaged for both cases (note that the relative results were quite similar for both scenarios). As the illustrated missing intersections were difficult to spot in print, the pictures were slightly post-processed to better spot the holes in the geometry due to numerical precision issues.

What is actually astonishing is the performance of the rather uncommon Arenberg test along with its exceptionally robust numerical results. The only argument against using it is unfortunately the increased amount of data that needs to be kept in memory and caches. If this is not an issue in the used ray tracing environment, it can be seen as the best performing test of them all (not only for the presented scenes, but for all tested scenes during the progress of this work), at least when working in static scenes where the additional pre-transformation step can be amortized over several rendered frames.
2.1. Triangle Boundary Representation

Figure 2.5: Triangle precision and performance for the Stanford Bunny. The numbers represent the frames per second, given as percentages relative to the Arenberg test.

Figure 2.6: Triangle precision and performance for the Stanford Happy Buddha. The numbers represent the frames per second, given as percentages relative to the Arenberg test.
Figure 2.7: Triangle precision and performance for Shirley's "Scene 6". The numbers represent the frames per second, given as percentages relative to the Arenberg test. The visualization of the Chirkov test intersection errors had to be filtered differently, due to false intersections found behind the current viewpoint, as the camera was placed rather close to scene geometry on purpose.

The Badouel test performs rather well (as expected, as it is the choice in a large number of realtime ray tracing systems at the moment), but features an unexpected high number of numerical errors. This behavior is usually not recognized, as it is said to be a rather robust test due to the used projection and doing floating point calculation work on similar scales/exponents most of the time. The actual reason is directly related to the 2-dimensional projection of the triangles: As long as the projection axis remains the same for neighboring triangles, all rays will use related kind of data (same components of the original unprojected triangle data), thus "sharing" the same kind of rounding errors. At the switch to a new projection axis, rays close in space (as it is the case for sampling neighboring pixels for example) will perform vastly differing calculations, so all precision problems will suddenly become distinct in these cases. As there is a higher probability that neighboring triangles produce different results for their edges during the barycentric u,v-test, some floating point cells "between the triangles" will actually never be hit by these rays.

The Plücker coordinate test (as it was originally proposed) features an unusual high number of numerical errors, a problem directly inherent to the "wild" mixture of supposedly differing ranges of floating point values during the calculations. This poses additional problems as soon as the scene is scaled to a larger extent and/or the ray origin is located far away from the object(s) in the scene. So although the same edge information is shared by neighboring triangles, the calculations with the ray data expose numerical problems. Also note that the Plücker coordinate calculations should actually be replaced by equivalent scalar triple products (which are closely related to the Segura/Feito test), as they can be calculated much faster and additionally robust if needed [Eri07].

Benthin's optimized variant proves to be more robust, as the optimization also delivers more floats on the same scale (floating point exponents that are likely to be similar) as well as featuring a pre-calculation step that includes the ray origin. This drastically helps to decrease the probability of missed intersections for all ray bundles originating from the same location, as the same calculations on similar data (up to a certain point) are executed, thus resulting in equal floating point errors. So the chance of rays slipping through edges is decreased (as neighboring triangles obviously feature shared edge data). As soon as the ray origins differ (as it is the case in the context of a distribution ray tracer, or more generally in global illumination simulations), similar problems as with the original version of the Plücker test arise, though. Still, it might
provide some of the best results (performance and error-wise) when using it in the context of shared origin bundle tracing to accelerate primary rays and shadow rays to point-light sources.

The Chirkov test, which is able to place the division at the very end of the implementation, loses this advantage in the context of an acceleration data structure. The probability to actually intersect a triangle is drastically increased in this scenario as the ray is mostly already clipped to the bounding box of the triangle (or even smaller when using spatial partitioning), so the faster early-out for the edge tests does not amortize. Additionally, divisions were made a lot faster (latency and throughput-wise) in the last years, resulting in less demand for optimizing them by any needs.

Finally, the test that was actually found to be the most attractive in a general and practical context, is the variant of the original 3-dimensional barycentric coordinate test. It is rather fast, works on the original vertex-data (which is very useful for dynamic scenes or an on demand acceleration data structure build, as will be seen in Section 2.5) and features only a small amount of numerical errors, a fact related again to preferring floating point calculations that work on a similar exponent.

2.1.3 Improving Ray Tracing Precision

Using floating point precision calculations [Gol91, Daw] in rendering software and especially ray tracing implementations can lead to all kinds of problems [WPO96, Eri05], as could already be seen in the last section. In this section more solutions to general numerical precision problems encountered during ray tracing are presented (unfortunately not for all of them), avoiding the common epsilon solutions if possible.

Geometry and Surface Normals (Terminator Problem)

A common hurdle when using interpolated vertex normals, bump-, or normal-maps is the terminator problem [WPO96]. As it only changes the shading calculations of a mesh, but not the underlying geometry, all calculations involving ray tracing or related approximations on the GPU, like shadow volumes (e.g. connecting to light sources or sampling the hemisphere) will introduce a bias. This is due to some rays being distributed underneath the geometric surface, resulting in blocky artifacts at shadow edges on smooth geometry (see Figure 2.8). Adding a small epsilon offset to avoid the self intersection problem results in missed intersections in detailed regions though (see Figure 2.9). While there exist solutions to blend between shading normal and geometric normal (or optionally their shading results) for problematic regions (a work around available in mental ray and blender) this only pushes the self intersection problems slightly away, but still cannot fully avoid them without also dramatically changing the shading calculations (and thus the physical simulation) of otherwise non-problematic regions.

Figure 2.8: Left: A low triangle count to approximate smooth surfaces results in the terminator problem, visible as blocky artifacts in regions between shadowed and non-shadowed regions. Right: Increasing the triangle count cannot fully avoid the problem but decrease the overall error.

Unfortunately, the only true solution is to use a better approximation of the real surface (demanding more primitives and thus decreasing the overall approximation error) or to use surfaces that feature a higher degree of continuity such as free form surfaces, avoiding the need for interpolating vertex normals [Dam05], and to use displacement-maps (which provide real geometry) instead of bump- or normal-maps.
Surface Acne

Using path tracing techniques or even simple Whitted style ray tracing, second generation rays will start on the previously intersected primitives, thus demanding a small offset to avoid self intersection with the same primitive again, a very common problem known as surface acne [WPO96]. Common practice is to choose a global (user-defined) epsilon value that defines an offset from the primitive to start the next generation of rays. While this epsilon may work well for some parts of the scene, some parts may still feature self intersections (as some points will not be changed at all by the offset, due to the limited floating point resolution for large exponents) or miss intersections in detailed (see Figure 2.9) or concave (see Figure 2.10) regions due to a too large epsilon value. A more robust solution to the problem was already presented in [AM90], but as the implementation is rather slow it must be rated as completely impractical for realtime solutions.

![Figure 2.9: Choosing an epsilon offset for the ray to avoid self intersections that works for the left situation can result in missed intersections in the right scenario.](image)

![Figure 2.10: While a pessimistic epsilon along the normal can avoid self intersections, a false intersection may still be detected for concave geometry: a) corners will become too dark due to false intersections with neighboring primitives, though b) and c) will not be affected by the problem as the offset avoids self intersections for all primitives (but one may still see dark lines on the primitive edges in some difficult scenes). d) An offset being too large results in rays starting from beneath the surface. The only stable solution is to find an optimal epsilon for all primitives near-by, which, however, makes the approach impractical.](image)

The following implementation uses iterations of the intersection along with a fully automatic local adaptation of the offset to the logarithmic floating point number scale by exploiting integer arithmetic and avoiding the need to fine-tune a global epsilon for each rendered scene.

- The procedure starts as usual with determining the point of intersection with the primitive.
- This point is then used to shoot a new ray reusing the direction vector of the original ray, resulting in a refined intersection point. This iteration smooths numerical problems resulting from rays starting far away from the actual point of intersection (so the components feature highly differing floating point exponents).
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- The last step, adding a locally fixed offset to avoid a self intersection, offers several possible implementations:
  - Using a weighted offset in direction of the outgoing ray,
  - using a (negative) weighted offset in the direction of the incoming ray,
  - using a weighted offset in direction of the normal,
  - using discrete versions of the previous proposals depending only on the signs, or
  - using a discrete offset only along the dominant component of the normal

While offsetting the point along the outgoing or incoming ray exposes the need for a too large offset due to the special case of rays being almost parallel to the surface (see Figure 2.9), offsetting along the normal always offers the maximum distance from the surface. Further, an offset solely based on the signs of the normal results in a simple and very efficient implementation, but the weighted variant being more stable in general (however it demands an additional costly float to int conversion per axis). The last variant, offsetting only along the dominant component of the normal, is the fastest of them all and very stable on the average, but demands a too large offset for all primitives with about equally large normal components.

Note that the signs of the offset must depend on the side of the primitive the new ray starts from, thus rays that handle refractive or transparent effects must be treated different from rays handling reflective and diffuse paths. In a user-controlled ray tracing system featuring programmable shaders, it may be valuable to automatically calculate two points of intersection (one in front and one behind the primitive, relative to the ray that originally intersected the primitive).

```
// Calculate Point of Intersection
Point endPoint = ray.o + (ray.d * t);

// Refine Point (based on newly calculated "t")
t = RayPlaneIntersect(endPoint, ray.d, tri_index);
if (t < 0.0)
{
    backHitPoint = endPoint;
    frontHitPoint = endPoint + (ray.d * t);
}
else
{
    backHitPoint = endPoint + (ray.d * t);
    frontHitPoint = endPoint;
}

// Offset Final Points by Integer Arithmetic
if (n * ray.d > 0.0)
{
    backHitPoint.Offset(n);
    n = -n;
    frontHitPoint.Offset(n);
    backSide = false;
}
else
{
    backHitPoint.Offset(-n);
    frontHitPoint.Offset(n);
    backSide = true;
}
```

Listing 2.1: Refine the point of intersection and offset by using integer arithmetic afterwards.
void Point::Offset(Vector &d) { //Vector d should be normalized
    //Point component close to zero?
    if ((uintCast(x)&0x7FFFFFFF) < uintCast(PT_SMALL))
        x += d.dx*PT_SMALL;
    else
        uintCast(x) += (int)(PT_EPSF[uintCast(x)>>31]*d.dx);
        //Faster solution, but also demanding a larger offset:
        //uintCast(x) += PT_EPSF[(uintCast(x)^uintCast(d.dx))>>31];

    //...same for y, z
}

Listing 2.2: Offset point by integer arithmetic, weighted by vector d. PT_EPSI/F stores the offset (positive and negative) that is addressed by the sign of the component.

Note that point components that are very close to zero must be handled separately due to the fact that the following ray-plane intersection of the new ray features direction components with floating point values of up to 1. As this largely differs from the rather small point component’s exponent, the fixed integer epsilon is not able to handle these immense numerical errors that arise during the ray-plane calculations, demanding a larger, but also fixed, floating point offset for these special cases.

While it seems possible to choose the value of the integer epsilon depending on the number of floating point operations involved in the ray-plane intersection code and determining the worst case, the resulting value is totally impractical and so only practical experiments can estimate a useful value. This work’s implementation uses a value of 32, which is very small but never resulted in any visible errors.

Still, there is one drawback of the method for some scenarios: Comparing the performance of this implementation with a simple globally fixed floating point value epsilon in the context of a path tracer, most scenes were found to feature close to no performance decrease, while some suffered a speed hit of few percent. The reasons are found in the nature of the used axis aligned data acceleration structure. All of the almost axis aligned primitives will offset the point of intersection into the same leaf again due to the very small locally adaptive epsilon value, while the fixed, and rather pessimistic global floating point epsilon will offset the points directly into the next neighboring leaf. Thus the more precise solution needs to traverse at least one more leaf, also resulting in an increased number of primitive intersections. Counting the number of shadow-ray-primitive intersections for extreme scenes (e.g. a just slightly rotated “Conference Room” or Shirley’s ”Scene 6”, see Figure 4.6), the overall amount rises by a factor of 5-10! While this seems like a lot of additional calculations, one must consider that most shadow rays in an axis aligned, and so rather simple, scene will not need to do a single shadow-ray-primitive intersection at all, as rays are originating from within an empty leaf (neighboring the starting primitive) and end inside an empty leaf (neighboring the light source), without intersecting anything in between. By slightly rotating such a scene, the leaf’s axis aligned bounding boxes will no longer tightly approximate the included surfaces and rays starting from these primitives will always need to traverse the same leaf again, even though the probability of an actual intersection is minimal. A pessimistic floating point based epsilon of course avoids this problem for such extreme (and also rather theoretical) scenarios, but at the cost of correctness of the image which will become visible in concave regions.

While the presented locally adaptive solution is able to handle the large majority of scenes with a ridiculously small local epsilon, the problem shown in Figure 2.10 still holds and cannot be resolved differently than using surfaces with higher continuity and computing intersections in full floating point precision [Dam05].

**Missed Triangle Intersections**

While the problem of missed triangle intersections has already been addressed in Sections 2.1.1 and 2.1.2, and some solutions to handle one part of the problem, namely missed or false intersections in the ray-plane intersection calculations, have been proposed in the previous subsections, rays may still slip through closed meshes in the vertices and along edges of neighboring triangles, and as such, this problem not only depends on the triangle test alone, but upon the whole design of the ray tracing kernel.
2.1. Triangle Boundary Representation

Obviously, a switch to error free floating point calculation [Knu81] can help to resolve the precision problems of the $u, v$ test, but a (moderately) fast implementation must need hardware support for FMA and ADD3 operations (where FMA is at least announced to be included in the SSE5 instruction set). Using interval-arithmetic [Moo66] will demand a redesign of large parts of the ray tracing system and additionally suffers from performance problems due to the increased number of calculations and an immense amount of rounding mode switches of the FPU which harms throughput on most common hardware platforms. While a simple switch to double precision calculation for the primitive intersection test already helps, it only pushes the problem away. Though it can be seen as an adequate "temporary solution", especially as modern x86-CPU's can perform double precision calculations at equal speeds as single precision operations (or even faster, as verified in experiments using double precision ray-triangle intersections), some hardware platforms do not support the double type at full speed yet (e.g. Cell and GPU) or become less efficient when utilizing SIMD operations (e.g. x86 platforms only feature 2xSIMD support for doubles).

One rather fast workaround for missed edge and vertex intersections is the slight scaling of primitives (in the sense of a locally adaptive scaling as seen in the previous subsection) to force an overlap of neighboring triangles, thus fully resolving missed intersections, but at the price of false intersections happening for neighboring triangles and on the silhouettes of each mesh. Two variants are feasible: Scaling before the hierarchy construction will resolve additional numerical problems of the ray traversal for (partially) axis aligned primitives and/or corner-vertices (see Figure 2.11), while scaling afterwards does not harm performance of the ray traversal due to an increased number of primitive replications (when using spatial partitioning) or the increased bounding volume size (when using object list partitioning). Unfortunately, this simple extension will not work when using a vertex/index-array representation (see Section 2.1.4) of the meshes.

A stable solution to force a robust ray-scene intersection for primitive edges was proposed in [Eri05]: By using consistent ray-edge orientations and thus sharing calculations for neighboring triangles (in a way similar to a refined Plucker coordinate test that works on multiple triangles at once), it is impossible for a ray to slip through neighboring triangles. The only remaining possible error is a false index of the intersected primitive (e.g. left instead of right triangle plane). However, a fast implementation in the context of ray tracing has yet to be found, as the additional data structure complicates the calculation of multiple triangle edges at once, or at least demands storing additional edge information. Note that [Eri05] additionally proposes to optionally use fat rays to avoid intersections if the edge testing seems unpractical, though the intersection performance is greatly reduced for this implementation.

Another solution can be found in the 2-dimensional projection of triangles onto ray-aligned planes, in a way similar to the beam tracing algorithm found in [ORM07], sharing calculations (and more important: floating point calculation errors) between neighboring rays. Note that this proposal still needs more research to be fully applicable.

Additional hurdles that can lead to missed intersections:

- Primitives should be tested using the full ray segment (especially when utilizing spatial partitioning (see Section 2.1.4)) instead of limiting the ray interval to the intersection with the bounding box of the current node, one reason being very thin nodes in the hierarchy (e.g. consisting solely of axis aligned geometry).
- Using a pre-calculated ray direction inverse $\frac{1}{\mathbf{r}}$, one should force the consistent use of either the ray direction or its inverse for all intersection tests and traversal decisions to avoid hazards with all (nearly) axis aligned geometry.
• Traversal decisions and the ray-triangle test should correctly handle results with value \(\infty\) and NaN, although it can be faster to fully avoid these case beforehand (see following subsection).

• Either the ray-triangle test should be able to handle degenerated and very small triangles (which is mainly an issue for all triangle tests that use pre-transformed data) or the hierarchy construction must sort these triangles out beforehand, which can result in small holes in the geometry though.

Instancing

While the use of object instances is rather popular, especially as it can help to animate simple objects in a complex scene in realtime [Wai04, 9.2], the numerical precision is drastically reduced by the additional forward and backward transformations of the ray and its intersection results. An additional variant that uses a pre-transformation of the triangles to save operations in a hardware implementation was presented in [SWW+04, WSS05], but introduces additional errors due to the implicit second transformation that must be done per triangle.

Note that visible errors will in fact only evolve for neighboring objects that need a seamless fit, however this is already the case for simple meshes like a car model. In such a setting it is extremely difficult to propose a stable, but still efficient solution. Those objects that do not overlap will not pose a problem at all, but still demand special care for second generation rays to avoid self intersections.

Performance Issues

While a numerically stable algorithm is necessary for flawless simulations, a carefully designed implementation should also consider possible performance issues or missing features on the used hardware. Although modern x86-CPU still offer IEEE 754 conformity, some platforms like the Cell or GPU only support a subset of IEEE 754, and even on x86, compilers allow to break IEEE 754 by using special instructions (like SSE and its successors) that allow for faster calculations.

Even more important, some calculations can result in an unexpected performance loss, with total ray tracing performance dropping by a factor of 10 or even 100, and thus is actually not only a theoretical issue! Slowdowns can happen in calculations involving NaN, \(\infty\) or denormalized numbers. While NaN and \(\infty\) are not longer an issue on the very latest generations of x86 CPUs (like the Intel Core 2 Duo or current AMD Athlon and Opteron generations) and can be handled without performance losses, denormalized numbers are still a problem as can be seen in Figure 2.12. Note that the current AMD Athlon architecture only has problems when loading and storing denormalized numbers, a fact seemingly related to all calculations supposedly being computed in higher precision internally and the denormalization only happening during the truncation to the standard 32- or 64-bit values. Thus, many compilers offer a switch to automatically truncate denormalized numbers to zero.

<table>
<thead>
<tr>
<th>Identifier</th>
<th>Cycles/op</th>
<th>Slowdown</th>
</tr>
</thead>
<tbody>
<tr>
<td>adding den to nrm</td>
<td>159,735</td>
<td>79.514</td>
</tr>
<tr>
<td>max den to nrm</td>
<td>124,555</td>
<td>61.994</td>
</tr>
<tr>
<td>mul den to nrm</td>
<td>159,912</td>
<td>79.601</td>
</tr>
<tr>
<td>div den to nrm</td>
<td>160,742</td>
<td>31.716</td>
</tr>
<tr>
<td>multiply to underflow</td>
<td>268,187</td>
<td>148.992</td>
</tr>
</tbody>
</table>

Figure 2.12: Denormalized number tests using SSE2 at double precision, measured by Thierry Berger-Perrin on a Core 2 Duo (using a variant of a CPU measuring tool released by Cygnus Software). The slowdown is measured against calculations using standard floating point numbers.

Note that all implementations that still use x87 compatible (and thus fully compliant IEEE 754) FPU code will suffer from tremendous performance issues for all denormalized numbers, \(\infty\) and NaN. In addition, performance of most standard x87 opcodes are significantly slower when compared to SSE operations (e.g. three times slower multiplication and six times slower division). This is related to all x87 calculations running internally on a 80-bit FPU, a feature that most likely is not supported directly in hardware anymore.
2.1. Triangle Boundary Representation

Another important fact of modern x86 architectures is the improved throughput (and latency) of divisions and square-root calculations that can even omit the usage of reciprocal approximations (available as specialized SSE opcodes) in some implementations.

Also, sine and cosine, amongst other trigonometric functions, can be automatically translated by the compiler into an array of SSE opcodes instead and so do not demand traditional x87 code anymore, thus further improving the calculation and throughput speed.

2.1.4 General Optimizations

After providing a short abstract overview and comparison of ray-triangle intersection algorithms and insights on possible numerical issues, this section presents more detailed implementation details and practical aspects on the ray-triangle intersections.

Bounding Volume Optimization

Similar to the use of bounding volumes for the culling of primitives to optimize rasterization on the GPU, it is helpful to limit the ray-triangle-plane range check. By clipping the ray to the currently active segment \([\max(r_{near, node_{near}}, r_{far}), \min(r_{far}, node_{far})]\), which describes the intersection with the actual bounding volume, the ray is limited by the actual intersections with the currently visited leaf node. This is useful for every triangle test that needs to calculate the distance along the ray first (i.e. barycentric coordinate tests) as it decreases the probability of needing to do an additional \(u, v\)-test. When using this simple trick, the relevance of an early-out due to an early \(u, v\)-test is further decreased, especially if there is a large amount of triangles stored in the leaf (see Figure 2.13). Note that this trick may harm the performance of spatial partitioning schemes such as the kd-tree, as primitives can extent over multiple leaf nodes. As a result, the same triangle will be tested multiple times (as no mail-boxing [Wal04] will work in that case) and additionally increases the probability of missing triangles during the hierarchy traversal due to numerical precision issues, especially in the case of triangles being aligned to the principal planes. So in the case of spatial partitioning it is more stable to use the full ray segment \([r_{near}, r_{far}]\) (not even clipped to the actual scene bounding box!) to avoid missed intersections.

A more general approach that builds upon this idea was just recently presented in [Res07] to better balance construction and ray traversal time. It proposes a dynamic construction of shafts for each visited leaf to allow for a fast culling of multiple rays against the triangles.

Figure 2.13: Limiting the ray-triangle test to the currently active ray-segment (intersection of the leaf volume with the ray) increases the probability of an early-out for the ray-triangle-plane test.
Reducing Branches

Apart from the standard way to avoid branches, namely the masking of results (which proves to be especially useful on SIMD capable hardware), this section will present some elegant algorithmic and arithmetic simplifications that allow to save a large portion of the rather unpredictable branches during ray-triangle tests, especially as these are able to drastically decrease performance by flushing the long pipelines of modern hardware architectures.

All tests that only need to check signs of results (e.g. barycentric coordinates, Segura/Feito or Plücker coordinates) can be optimized by using integer arithmetic instead of using floating point comparisons, as the sign of a floating point number is always stored in the upper bit. This will be demonstrated for the barycentric $u, v$ test:

```c
if( ((u < 0.0) || (v < 0.0) || (u + v > 1.0))
    return;
```

can be simply replaced by testing the combined signs of $u$ OR $v$:

```c
if( (intCast(u) || intCast(v)) < 0 || (u + v > 1.0))
    return;
```

In the case of testing multiple signs (e.g. Segura/Feito or Plücker coordinates) this simple combination of tests will help when using branchless SIMD masking as it allows to only do the masking of the result once.

A more advanced trick is to use unsigned integer arithmetic for certain tasks. In the case of the barycentric test, it would be possible to replace $u + v > 1$ by the almost equivalent $\text{intCast}(u+v) > 0x3F800000$ (almost, as it will not handle NaN the same way as will be shown later on). Doing the same test using unsigned integer arithmetic $\text{uintCast}(u+v) > 0x3F800000$, the case of both $u < 0$ and $v < 0$ is additionally tested for free, as then the sign of $u + v$ will be set ($0x80000000$ or greater in the unsigned integer representation). Actually this single test will already handle a large fraction of rays missing the triangle as can be seen in Figure 2.14.

![Figure 2.14: Triangle projected to the barycentric coordinate space. Triangle vertex $p_0$ equals $(u, v) = (0, 0)$.](image)

The test to discard the remaining cases can then be inserted afterwards by testing either $u \neq v < 0$, $(\text{intCast}(u) || \text{intCast}(v)) < 0$ or $(\text{intCast}(u) \ XOR \ \text{intCast}(v)) < 0$ (whatever is fastest).

Still, the test must use two separate conditions for testing $u, v$. Taking a closer look at the tests reveals the following situation: If scaling both $u$ and $v$ by two (which can be achieved in the pre-calculation step of
the barycentric test for free), the test \texttt{uintCast(u+v) > 0x3F800000} can be replaced by \texttt{uintCast(u+v) > 0x40000000}. This does not seem like much, but it actually allows to combine this test with \texttt{(intCast(u)|intCast(v)) < 0} by just using \texttt{(uintCast(u+v)|uintCast(u)|uintCast(v)) > 0x40000000}. In the context of a hardware implementation, this trick is rewarding as it means that no floating point comparison at all has to be implemented and a simple bit-test on positions \texttt{0x40000000} and \texttt{0x80000000} is enough to know that a ray actually missed the triangle. A fact that also saves a lot of operations when implementing a branchless SIMD version of the barycentric triangle test using masking.

\begin{verbatim}
if((uintCast(u+v)|uintCast(u)|uintCast(v)) > 0x40000000)
    return;
\end{verbatim}

As mentioned earlier the special case of NaN won't be handled exactly the same as before (depending on the test using \texttt{>}) to exit the test, or \texttt{<} to continue), but this does not pose a problem here. As the case of \texttt{u} or \texttt{v} being NaN should mean that the ray missed the triangle, the presented tests will always handle this special case correctly.

Directly related to the previous arithmetic simplifications, the ray segment-triangle plane range check of the intersection distance \texttt{t} along the ray to the segment \([0, \text{\texttt{Far}}]\) can also be simplified from

\begin{verbatim}
if(!(t >= 0.0) | (t > rayFar))
    return;
\end{verbatim}

(the \texttt{NOT}-operation being used to handle the case of \texttt{t} being NaN, as comparisons with NaN always result in false) to

\begin{verbatim}
if((uintCast(t) > uintCast(rayFar))
    return;
\end{verbatim}

\section*{Reducing the Memory Footprint}

As stated earlier, triangle tests that perform a projection onto a principal plane (i.e. the Badouel test or its variants) must store the axis of the projection for each triangle, resulting in two bits of additional data (so at least one byte if stored in a brute force manner). \cite{Wal04} proposed a padding of the data from nine floats (plus these two bits) to 48 bytes, mainly due to cache-line alignment issues (two triangles combined are aligned to 32-byte, four to 64-byte boundaries). Although a performance increase was reported by this alignment, the results of this thesis’ experiments were always slower when increasing the memory footprint of the triangle data (no matter how much). This may be a result of the increased cache usage due to the bloated data-set.

In the following some possible variants are presented to hide the two bits in other data, thus keeping the transformed data at nine floats (as it is the case with the original data-set of three triangle vertices).

- The first (and most elegant) variant was proposed by \cite{Naw07} and uses the projected normal vector of the triangle plane to store the bits. Remember that the third component was set to a constant (usually one, to be able to discard multiplications with this component) and that this component is the dominant value of the normal, thus leaving the remaining components \(|n_u|, |n_v| \leq 1\). In the corresponding integer representation, both values have their \texttt{0x40000000}-bit set to constant zero. Thus it is possible to abuse these bits to store the projection axis bits. The only problem of this neat trick is that extracting the bits and \texttt{AND}-ing the normal components bits back to zero costs a lot of rendering performance (usually 5-10 percent, depending on the overall number of triangles tested per ray).

- The second variant proposed in \cite{Wäc04} makes use of an additional degree of freedom in the barycentric coordinate test. When transforming the triangle-data, the selection of the vertex \(p_0\) to compute the
edge-data for is actually not determined. Thus it can be selected such that the (projected) edge-vectors will feature both positive u or both positive v components (see Figure 2.15), leaving the sign-bits of the respective floating point values at constant zero. Again, the problem is extracting the bits and \text{AND}-ing the edge components bits back to zero, resulting in a similar performance drop as the previous variant.

- Another variant using an additional degree of freedom involves the usage of the accompanied data acceleration structure: As triangles are referenced in the leaves of the tree, it is possible to keep these references sorted by their respective projection axis. This variant however needs to store three counters in the leaves (one for each principal axis) instead of the standard single reference-counter. In practice this does not pose much of a problem as the counter is usually four bytes long, due to alignment issues [Wal04], so it is possible to split this counter into three smaller counters, each eight or even ten/eleven bits long. This is usually enough for real life scenes, though it is theoretically possible to have some leaves featuring more than 255 or 1023/2047 references. One example being brute force tesselated cones (resulting in a lot of thin triangles, all sharing the same vertex) or more generally speaking, vertices featuring an extremely high valence.

- The last presented variant is also the fastest of them all and uses the alignment of the triangle references stored in the leaves of the acceleration data structure. When using pointers, the lowest two (or three) bits of the pointers can be forced to be constant zero (due to the references being four (or eight) byte wide). Using indices for the references, the upper two (or three) bits can be forced to be constant zero (the only exception being 32-bit indices used on 64-bit hardware). Then the projection case can be stored in these two unused bits.

![Figure 2.15: Possible selections for the triangle vertex $p_0$ to compute (projected) edge vectors featuring positive $u$ or $v$ components.](image)

An additional problem can be the lost signs of the normal vector, using the optimized Badouel test. As the dominant component was re-scaled to one in the transformation process, the signs of all components have been flipped if this component was negative before. This can become a problem if the original orientation (and not just direction) of the geometric normal is needed in the shading computations (e.g. to handle refractive materials). In this case it is possible to reconstruct the original sign of the component by the cross-product of the edge-vectors.
2.2. Acceleration by Spatial Partitioning

Amortizing Pre-Computation

In the context of dynamic scenes all primitive tests that use transformed data must actually be rated differently: Does it amortize to transform the primitives to be used for one frame only? As the primitive might only be intersected with few rays or even no ray at all, the pre-computation might result in more memory traffic and operations than it actually saved for the ray tracing step itself. An additional problem arises if some of the original data (i.e. the triangle vertices) is needed during shading computations or an on demand tree construction, as some triangle tests need costly computations (or even some additional data) to calculate the backward transformation to receive the original vertices again, demanding the original data to be saved along with the transformed data. In the context of games (which need to share memory with additional textures, etc.) or in massive geometry scenarios this is clearly not possible.

Compact Vertex/Index-Array Representation

Although most realtime ray tracing implementations currently used prefer to store triangle data on a per triangle basis (e.g. to allow for pre-transformations like the presented Badouel test), it can be worth to use the common vertex/index-array representation of meshes that separates vertex data from triangle connectivity, thus omitting the multiple storage of vertices that are shared amongst triangles. In practice (rule of thumb) this can save up to half the amount of memory used for a triangulated scene (of course this depends on the exact scene, leaving a large gap between a scene consisting of almost randomly scattered unconnected triangles, and closed triangle meshes with an extremely high vertex valence (such as a triangulated cone)). In this work’s experiments, the use of the vertex/index representation resulted in almost no performance loss at all (on the average), as the reduced cache thrashing (vertices in the CPU cache can be reused for neighboring triangles during ray tracing) seems to outbalance the additional indirection due to the index-list. Further, omitting a pre-transformation leaves more freedom for the on demand build of the data acceleration structure (see Section 2.5.5) and the possibility of an easier update of the tree (see Section 2.5.6).

If a certain pre-transformation of the triangle data proves to be faster in a certain implementation, it is also possible to use a caching mechanism to combine the best from both worlds: When visiting a leaf, the contained triangles are pre-transformed and the data is stored in a cache for other rays visiting the same leaf later (or in the case of a spatial partitioning scheme, when the same triangle is intersected again).

It is also possible to store the index-list directly into the leafs of an acceleration data structure to save one indirection. This is easy in the case of leafs guaranteed to feature only one triangle (as it can simply be hardwired into tree construction and traversal when using this fixed leaf size) or by using the three references (to each vertex of a triangle) as a substitute for the usually used single one (to reference a whole triangle in memory). This is actually the way to go for bounding volume hierarchies as it decreases the overall memory used (due to the completely missing triangle reference list in the leafs), but can harm all spatial partitioning schemes due to the tripled amount of data needed per triangle reference (whether this outweighs the missing triangle reference list is strongly depending on the used scene).

2.2 Acceleration by Spatial Partitioning

This section will describe some everyday problems that arise using spatial partitioning data structures and present efficient solutions in the context of using a kd-tree for acceleration. For more details see [Wän04] that presents various implementation details, data structure layouts, and low level optimization possibilities.

But first, a small introduction of spatial partitioning schemes: The space containing the objects is partitioned into disjoint volume elements. Efficiency is obtained by enumerating the volumes intersected by a ray and then testing the objects in the volumes for intersection. A major disadvantage of space partitions is that objects are often referenced more than once, since they may have non-empty intersections with more than one of the volume elements. This results in larger memory footprint and requires a mailbox mechanism to avoid performance losses when intersecting one ray with the same geometric object multiple times.

Regular Grids The space is partitioned into a raster of identical rectangular axis-aligned volumes. This regular structure allows for simple algorithms that enumerate the volume elements along a given ray.
As each volume contains a list of the objects that it intersects, only their objects are intersected with the ray.

The memory footprint of the acceleration data structure cannot be determined in advance, because objects can intersect multiple volume elements and thus requires dynamic memory management. The data structure is constructed by rasterizing the objects. This requires variants of an object-volume element intersection routine, which is numerically unreliable due to the finite precision of floating point arithmetic. Rasterizing the bounding box of the object is numerically stable, but increases the memory footprint due to unnecessary replications of the references.

The efficiency of this straightforward approach severely suffers from traversing empty volume elements, as explored for massive scenes and the teapot-in-a-stadium scenario. A solution to this problem is found in hierarchical grids that allow empty space to be traversed faster while still having a moderate number of objects per volume element. Switching between the levels of the hierarchy, however, is expensive and can be achieved more efficiently by other spatially adaptive schemes.

Hierarchical Space Partition
One example of a hierarchical data structure is the binary space partition. The general idea is to adaptively subdivide space by using arbitrary planes, which allows one to overcome the efficiency issues of regular grids caused by empty volumes elements. In polygonal scenes an obvious choice is to use the planes determined by the polygons themselves. However, it is not known how to do this in an optimal way and randomized algorithms are expected to yield trees of quadratic size in the number of objects in the scene [MR95, Sec.9.7]. Even when restricting the construction to only a tiny subset of all possible split-planes, the construction could take several hours or even days for common scenes [KM07].

kd-trees restrict binary space partitions to using only planes that are perpendicular to the canonical axes. Since all normals of the subdivision planes coincide with one of the canonical axes unit vectors, scalar products and object-volume element intersection tests become more efficient and numerically robust than for the original binary space partitions. Still, the decision of how a partitioning plane intersects an object remains a numerical issue. Along with heuristics for subdivision, kd-trees have been used very successfully for accelerating ray tracing [Kel98, Hav01, Wal04, RSH05, Ben06].

As with all spatial partitioning schemes, objects can reside in more than one volume element. Although the number of multiple references can be effectively reduced by allowing only partitioning planes through the vertices of the objects, the number of references cannot efficiently be bounded a priori. Consequently, memory management becomes an issue during the construction of the hierarchy. Heuristics used for memory estimation and allocation only hold for common scenes of low complexity, but can be way too pessimistic for others, or even worse, can result in various reallocations if the lists need to grow during the construction phase. This results in performance losses by memory fragmentation.

2.2.1 Sorting Primitives into a kd-Tree
A general problem of spatial partitioning is the determination of which primitives intersect which voxels of the partitioning scheme. While several algorithms exist to test for an intersection of an axis aligned box and a triangle (see [AM01, Ebe01] for efficient implementations), a less general implementation can achieve better performance when optimized for a specific spatial partitioning scheme. A kd-tree implementation can rely on the following assumptions to design an efficient triangle-box test:

- The primitive is guaranteed to (at least partially) intersect the current node and thus must intersect at least one of its children.
- Thus, if the primitive's bounding box is completely situated on one side of the splitting plane, it is guaranteed to exactly intersect one child.
- If the primitive is situated in the split-plane, a heuristic can choose in which of the children it will be sorted into.

Based on these assumptions and the insights of [Kol04b], the following optimized test is proposed:
2.2. Acceleration by Spatial Partitioning

- Test the primitive if it is aligned with the split plane (see Section 2.4.2).
- Test the primitive’s bounding box for an intersection with the split-plane. If no intersection is found, sort the primitive into the one child it intersects.
- Otherwise, the primitive has to be intersected with the rectangular resulting from the intersection of the split-plane and the node bounding box.

  - Determine the vertex situated alone on one side of the split-plane and find the intersections \( a_1, a_2 \) of the two triangle edges (connected in that vertex) with the split-plane (see Figure 2.16).
  - Test the bounding rectangle of \( a_1, a_2 \) for an overlap with the node-split-plane rectangular (see Figure 2.17). If no overlap is found, the triangle is only sorted into one child.
  - Otherwise, calculate a vector \( n \) orthogonal to the line through \( a_1, a_2 \) (see Figure 2.16). The signs of the components of \( n \) are then used to index the two corners \( B_1, B_2 \) of the rectangular that feature the largest distance from the line through \( a_1, a_2 \). By testing the signs of \((a_1 - B_1) \cdot n \) and \((a_2 - B_2) \cdot n \) for equality it can be determined if both corners are on different sides of the line and thus if the line intersects the rectangle, resulting in both children being intersected by the primitive.

![Figure 2.16: Triangle overlapping the split-plane.](image1.png)

Dist is used to calculate the \( a_1, a_2 \) intersection points, \( n \) to determine the two corners of the rectangular featuring the largest distance to the line through \( a_1, a_2 \).

![Figure 2.17: First, test \( a_1, a_2 \) to be completely on one side, outside of the rectangle. Then test the line through \( a_1, a_2 \) against the rectangle.](image2.png)

Note, that the test features the highest possible numerical precision by postponing numerically unstable operations as possible, with the only critical operations being the calculation of the 2-dimensional intersection points \( a_1, a_2 \) and the signed distance test of the bounding rectangle corners with respect to the line through \( a_1, a_2 \). Additionally, the test can be designed to use just a minimum of branches, thus increasing throughput on modern hardware architectures.
**Faster, Approximate Construction**

It is of course also possible to omit the detailed triangle test and just test the primitive's bounding box for an intersection with the split-plane and sort the primitive into both children without further testing. Note that this can only work well for the upper levels of a hierarchy and/or scenes consisting solely of small triangles. Otherwise a lot of leaves will inherit primitives that are actually not intersecting these leaves, resulting in a lot of unnecessary ray-triangle intersection tests later on. In fact this introduces a problem that is usually only inherent to the nature of a BVH (see Section 2.4.2). Optionally, it is possible to use the approximate sorting only for some levels of the tree and an exact sorting for the rest. While this increases the quality of the trees, it also introduces the need for a full box-triangle intersection test again, as the assumptions made to formulate a faster test as in Section 2.2.1 do not hold anymore, thus also decreasing the construction speed. Note that the resulting tree is still of lower quality compared to the full exact solution, as the partitioning heuristics will also be working on a temporary approximate solution.

**Exact, but Memory Intense Construction for Void Area Clipping**

Additionally it is possible to formulate a computationally slightly simplified exact version, however resulting in a dramatic increase of temporary memory. Instead of duplicating primitive references during the hierarchy construction, the actual primitives themselves are replicated into a temporary buffer by clipping overlapping primitives on the split-plane. This allows to use a fairly simple primitive sorting (the approximate solution found in the previous section is then exact), but at the price of clipping the newly created triangles (which somehow resembles the exact solution presented earlier in this section) and an increased overall memory footprint. Note that the duplicated triangles can be removed from the hierarchy in a post-processing step by keeping an index to the original primitive for each replicated triangle, thus removing all temporary memory and resulting in a memory layout using only references to primitives again.

Still, numerical precision is greatly reduced in comparison to the previous solutions, as the clipping process results in an increased amount of numerically unstable operations, due to the newly created triangles being used afterwards to again create new triangles, recursively over various levels of the hierarchy, thus spreading previous calculation errors.

The interesting fact about this implementation actually is not the computational speed (which is rather disappointing for scenes that do not fit into the caches anymore due to the increased memory usage), but rather the ability to easily determine empty volume cut off splits (see Section 2.4.2). These would otherwise require an immense computational effort by clipping all triangles to the full node bounding box to determine the "real" bounding box of the primitives in order to propose efficient split-planes for the cut off of the surrounding void area.

**2.2.2 Managing Temporary Memory**

Spatial partitioning schemes need to handle objects that overlap volume elements. For example, the recursive \( kd \)-tree construction needs a vast amount of temporary data to be placed on the stack for the continuation later on, due to the replication of primitive references. The straightforward method of handling this problem is to allocate new arrays consisting of the references for each child of a node.

A variant of the quicksort algorithm can alleviate these inefficiencies and makes in-place sorting available for \( kd \)-trees. The procedure requires a second global array of object references that is used to keep the objects that are classified both left and right, similar to a self-managed stack. Testing with a large number of scenes revealed that the size of this array can be chosen by a default value (a length equal to the number of objects is far more than one will ever need in 99 percent of the cases). However, because the real length of the array cannot be predicted and the worst case memory usage being quadratic in the number of objects, it might be necessary to reallocate memory. The procedure is illustrated in Figures 2.18 and 2.19.
2.2. Acceleration by Spatial Partitioning

Figure 2.18: \(kd\)-tree in-place sorting of object indices (upper array) and storing on the global stack (lower array) during tree construction. From left to right, from top to bottom: In this illustration, sorting the unordered array first reveals a left element which stays in place. The second element must go to the right and is exchanged with an unsorted element from the right end of the array. In the third case an object has to be sorted into both volumes and, therefore, is moved to the stack. After sorting all objects, the stacked elements are copied to the middle segment of the index array and the tree construction continues with the left child.

Figure 2.19: Restoring stacked object indices after a stack pop operation.
2.3 Acceleration by Object List Partitioning

When partitioning a list of objects, each object remains referenced at most once and it becomes possible to a priori predict memory requirements. In addition each object is intersected at most once with a ray and consequently mailboxes (that can accelerate spatial partitioning schemes) become redundant. As an unavoidable consequence the volumes enclosing groups of objects often cannot be disjoint.

Bounding Volume Hierarchies have been introduced in [RW80, KK86] and are often used in industry since memory requirements can be bounded a priori linearly by the number of objects. However, limits of the classic approaches were explored in the first ray tracing chip [Hal99] and efficient software implementations [GM03, WBS06] that remained inferior in performance compared to kd-trees.

Implementing bounding volume hierarchies does not require object-plane intersection routines (see Section 2.2.1). As a consequence they are simpler to implement than spatial partitioning schemes. Using axis-aligned rectangular bounding volumes avoids any numerical stability issues during construction as only minimum/maximum operations are used.

There exist heuristics for both the bottom-up and top-down construction, which have been summarized in [WBS06]. The usual heuristic is to minimize the overall volume of all volume elements along with a tight packaging of the included primitives [WHG84]. These optimization procedures are prohibitively slow and in fact it is not clear what the most efficient construction algorithm is.

Severe performance penalties stem from the fact that opposite to binary space partitions the volume elements are not ordered. Usually all child nodes have to be intersected with a ray, because an early pruning was impossible due to a lack of a spatial order. This problem has already been identified in [FP93].

Another reason is found in the packaging of the objects into voxels that do not allow to efficiently discard empty volume around non-axis aligned geometry, as the minimum volume of a leaf is limited by the included primitives, thus resulting in an increased number of primitive intersections for rays that just slightly miss geometry on the silhouettes of an object. More details on this problem will be listed in Section 2.4.2.

Note that a uniform and regular structure similar to spatial grids is yet unexplored for object list partitioning as it is difficult to efficiently handle large, overlapping objects. One step into this direction will be presented in Section 2.5.7 though.

While the principles of implementing bounding volume hierarchies to accelerate ray tracing are well understood, only little research has been carried out concerning the efficient construction of the acceleration data structure and advances for massive geometry and/or dynamic geometry were rather moderate until recently.

This section proposes a hierarchical acceleration data structure for ray tracing that can be constructed much more efficiently than previous approaches. The procedure is so fast that interactive ray tracing of dynamic scenes becomes available on even mono-processor systems (see Section 4.2).

The basic underlying idea is to save calculations and data memory for BVHs by decreasing the number of stored clipping planes per node. As fast construction methods are mostly based upon top-down construction, the partitioning of a node results in most changes happening on one axis only, thus making the storage of additional clip planes for the remaining axes unnecessary, a behavior spottable in the upper levels of a hierarchy especially.

When sticking to partitioning on solely one axis, the traversal can be implemented to resemble the most efficient kd-tree implementations, while memory control and numerical precision of BVHs still hold, as will be demonstrated in this section where a new and efficient data structure will be presented.

2.3.1 Bounding Interval Hierarchy (BIH)

So unlike classic bounding volume hierarchies [KK86, GM03], which store a full axis aligned bounding box for each child, the idea of the bounding interval hierarchy is to only store two parallel planes perpendicular to either one of $x$, $y$, and $z$-axis. Given a bounding box and the axis, the left child $L$ results from replacing
the maximum value along that axis by the first plane. In an analogous way the right child \( R \) results from replacing the minimum value by the second plane (see Figure 2.20). Resulting zero volumes are used to represent empty children.

![Figure 2.20](Image)

Figure 2.20: Geometric primitives overlapping the splitting plane of a \( kd \)-tree have to be referenced in both child volume elements in a), while the bounding interval hierarchy in b) assigns an object to either the left \( L \) or right \( R \) child. Traversing a node of the \( kd \)-tree in c) requires to distinguish the four cases \( L, R, LR, \) or \( RL \), whereas for the bounding interval hierarchy in d) the additional fifth case of traversing empty volume has to be considered. Optionally, the cases e) and f) can be introduced by allowing splitting planes to tighten the node volume instead of partitioning it.

```c
typedef struct {
    long Index; //lowest bits: axis (00,01,10) or leaf (11)
    union {
        long Items; //leaf only
        float Clip[2]; //internal node only
    }
} BIH_Node;
```

The inner nodes of the tree are described by the two clipping planes and a pointer to a pair of children. As this sums up to 12 bytes (or 16 when using 64-bit hardware) in total, all nodes are aligned on four-byte boundaries. This allows one to use the lower two bits of the children-pointer to indicate the axis (00: x, 01: y, 10: z) or a leaf (case 11). Leaf nodes consist of a 32-/64-bit-pointer to the referenced objects and their overall number. The overhead of four bytes when using 32-bit-pointers in the leaves (as they only use eight bytes out of the node data structure) can be resolved by a careful implementation.
Note that a slightly faster implementation (about five percent) can be achieved by storing the encoding of a node to be a leaf or an internal node in the sign bit of the integer index, but as such also decreases the maximum number of addressable nodes.

Furthermore, it proves valuable to introduce an additional encoding for the clip planes to allow for an efficient cut off of empty volume (also see Section 2.4.2) on both sides of the included primitives of a node at once. By using the clip planes to tighten the volume instead of partitioning it (see Figure 2.20e/f), the overall memory usage can be decreased along with receiving faster traversal times (see Figure 2.21). Note that the node in this case obviously only features one single child instead of the commonly used two.

<table>
<thead>
<tr>
<th>Scene</th>
<th>Memory Internal Nodes</th>
<th>Memory Leafs</th>
<th>Trace Time</th>
</tr>
</thead>
<tbody>
<tr>
<td>UNC Power Plant</td>
<td>92</td>
<td>100</td>
<td>86</td>
</tr>
<tr>
<td>Happy Buddha</td>
<td>92</td>
<td>100</td>
<td>93</td>
</tr>
<tr>
<td>TBP Kitchen</td>
<td>93</td>
<td>100</td>
<td>91</td>
</tr>
<tr>
<td>Dragon</td>
<td>92</td>
<td>100</td>
<td>95</td>
</tr>
<tr>
<td>Car 1</td>
<td>91</td>
<td>99</td>
<td>93</td>
</tr>
<tr>
<td>Thai Statue</td>
<td>91</td>
<td>100</td>
<td>92</td>
</tr>
</tbody>
</table>

Figure 2.21: Ray tracing time and memory footprint differences (in percent) when using the additional special case encoding to use two clipping planes at once to discard empty volume around primitives. While the memory footprint of internal nodes and the trace time clearly decrease, the amount of memory for the leaf nodes is just slightly smaller due to the packing of the leaves: empty space cut off is most of the time only valuable for one side on each axis (if at all).

Note that the BIH is in fact rather similar to the earlier proposed R-trees [Gut84] and the s-kD [OSDM87, GG98]. Also later released publications (that were developed in parallel, independent from the BIH though), like the loose kd [EGMM07], the DE-tree [ZU06] and the h-tree [HHS06] are based on similar ideas. Additionally, [WMS06] proposes a hardware implementation based on a data structure similar to a variant of the BIH (using four clipping planes per node though). Still, the following sections will extend the basic BIH data structure even further and propose new and faster construction schemes, partitioning heuristics and general optimizations that were not possible before.

**In-Place Construction**

Because the bounding interval hierarchy is an object partitioning scheme, all object sorting can be done in-place and no temporary memory management is required. The recursive construction procedure only requires two pointers to the left and right objects in the index array similar to a quicksort implementation.

The implementation can then either compute the array straight forward from left to the right, which is very simple (but can result in an "unnecessary" scrambling of the primitive references), or use standard optimizations of the quicksort algorithm like computing the array from left and right "simultaneously" until elements are found on both sides that need to be exchanged. This not only reduces the number of necessary memory transfers, but also tries to keep the original input array unscrambled. The total construction time is decreased by an average of about ten percent over the straight forward solution.

Note that the in-place sorting holds for any kind of bounding volume hierarchy and can additionally be extended to be used for spatial partitioning schemes as was shown in Section 2.2.2.

Additionally, it is also possible to directly exchange the primitive data during the sorting step instead of using a reference list. This allows to discard the additional reference array for the leaves, as the leaf can directly refer to the primitive array. Although this should help memory transfers (as triangles of the same leaf will be stored in same cache lines or at least following memory addresses) and saves the additional indirection, the total improvement for the tracing time averages at two-three percent only. This comes at the cost of an increased hierarchy construction time (bigger memory blocks that need to be moved) and the scrambled order of the primitives, which is unfortunate if additional data like texture coordinates and materials is needed for rendering, as it demands to reintroduce an inverse reference list.
2.3. Acceleration by Object List Partitioning

**Numerical Precision**

The bounding interval hierarchy construction only uses bounding box information and minimum/maximum operations in the canonical coordinate system. As such the construction procedure is numerically unconditionally robust. In contrary, spatial partitioning schemes require object plane intersection routines that rely on operations that suffer from floating point imprecisions. In order to make the algorithms completely stable, tolerances must be added. As a consequence performance suffers. See Section 2.1.3 for details.

**Memory Complexity**

Since the construction algorithm matches split-planes to object bounding boxes, the number of inner nodes in the hierarchy is bounded by six times the number of objects. In the case of polygons, this number is bounded by the minimum of the previous bound and three times the number of vertices. Note that the number of object references in the leaves always exactly matches the number of objects in the scene.

Due to multiple object references, the latter bound is not available for spatial partitioning schemes as for example the $k$d-tree. The problem becomes apparent where objects feature a locally high number of overlapping bounding boxes or almost random distribution: In the BART museum (see Figure 2.23) the number of replicated triangle indices is about 400 times (peak value) the number of triangles, which also results in a 40 times higher number of nodes than in the bounding interval hierarchy. This problem is intrinsic to $k$d-trees, as the first top level splits already duplicate a lot of triangle references, an effect which is continued during recursion. This behavior will be investigated in Section 2.5.3 in detail.

**Variants of the Data Structure**

In contrary to standard bounding volume hierarchies, the number of clip planes in a BIH node is completely flexible. Thus two additional variants are proposed and rated against the original dual plane implementation.

The first variant tries to decrease the traversal complexity while increasing the construction flexibility by reducing the number of clip planes to one. While this resembles a $k$d implementation even more, it has been found that the resulting trees will actually be very similar to the dual plane variant, regarding the number of total split-planes and their positions. This is due to the fact, that the only case where the single plane variant case can save one of the otherwise used two split-planes, is an empty volume cut off on only one side. In the other cases, the single plane always leaves one of the children at the original size of the father node, so heuristics based on the volume of a node (see Section 2.4) will propose a split on the same axis again, resulting in exactly the same two clip planes as if directly using the dual plane variant. As such, this variant can be seen as unpractical as it only increases the total number of pointers in the tree, while just slightly decreasing the total amount of split planes. Additionally the ray traversal is slowed down due to increased memory bandwidth, additional overhead due to the increased number of branches in the tree plus more stack operations due to one child being always the size of the father.

The second variant increases the number of split-planes to four. This can be seen as the one-dimensional variant of a classical three-dimensional axis aligned bounding volume hierarchy. The additional two planes are used to automatically clip empty volume surrounding the included primitives (note that this variant was also proposed in [WMS06], independent from this work's research). While this variant decreases some of the overhead used for empty volume splits that are necessary for the performance of the dual plane variant, it at the same time increases the amount of unused clip planes. Especially on the top levels of the hierarchy the additional two clip planes do not pay off, same when coming down to the object level of a scene, as the triangle distribution will then be rather dense and useful empty volume cuts can not happen every split. Thus the increased overhead for memory transfers and intersection operations during the ray traversal make this variant slightly slower, for some scenes even dramatically slower than the dual variant, at an increased total memory footprint.

2.3.2 Ray Hierarchy Intersection

Intersecting a ray with the bounding interval hierarchy binary tree is similar to traversing a bounding volume hierarchy. However, since the children are spatially ordered in the new data structure this can be implemented
much more efficiently, since it is possible to directly access the child that is closer to the ray origin by the
sign of the ray direction. The traversal thus becomes almost identical to that of a kd-tree, as illustrated
in Figure 2.20. Recently, the advantage of ordering the children of general BVHs along one axis was also
demonstrated in [WBS06], though the implementation needs to store additional data inside the nodes to be
able to retrieve the ordering during traversal.

In analogy to bounding volume hierarchies, it is also possible to not intersect any child at all if the valid
ray segment is between two non-overlapping children (see Figure 2.20d). Handling this additional case is
even beneficial, because it implicitly skips empty leafs. Consequently, empty leafs can never be accessed and
therefore do not need to be stored.

Opposite to spatial partitions, the volume elements of a bounding interval hierarchy can overlap and
consequently the recursive traversal cannot stop as soon as an intersection is found. It is always necessary to
test all remaining volume elements on the stack for closer intersections. However, as soon as an intersection
is found, branches of the hierarchy can be pruned if they represent volume elements further away than the
current intersection.

Note that an additional performance increase can be gained by implementing a specialized ray-hierarchy
intersection for shadow rays. As these rays can skip further traversal and intersection calculations as soon
as an intersection with any primitive has been found, the costly popping of the complete stack can also
be discarded, an optimization which is usually only possible for spatial partitioning. Consequently, the
optimization of shadow ray intersections in a complete global illumination simulation results in an overall 5
percent performance increase on the average using the BIH, while not providing any measurable gain at all
for the kd version.

2.3.3 Reducing the Memory Footprint

A major problem for current and especially upcoming and future hardware is the latency of memory fetches,
so ever-growing caches are included to hide latencies at least for recurring data. To better utilize the
rather small caches (when compared to the total size of the geometry and/or hierarchy), the two following
subsections will concentrate on further reducing the memory footprint of the acceleration data structure.

Quantization

A straight forward solution to compress floating point data is an additional (post-process) quantization step
at the loss of numerical precision, however, in the case of BVHs, a quantization of the split-planes can be
achieved that still provides full precision for the ray tracing step. Along the lines of [Ter01] that uses BVH
quantization to 16 bits for collision detection and the recently released [CSE06] which uses quantization to
16 or 8 bits in a k-ary BVH for ray tracing, a comparable implementation will be presented for the BIH.

When quantizing the clip planes of a BIH, the left child’s plane has to be rounded to the next larger value,
the right child’s plane to the next smaller value (note that this can also be implemented without switching
rounding modes by adding constants and always truncating). This slightly increases the volume of the child
bounding volumes, thus preventing rays of missing triangle intersections due to quantization errors later on.
The presented implementation uses a quantization to 16 bits per float to reduce the memory footprint for
each node from 12 to 8 bytes, thus also achieving a better cache line alignment:

\[
\begin{align*}
\text{newsplit}[0] &= (\text{split}[0] - \text{worldBoxMin.x}) \times 65536 / (\text{worldBoxMax.x} - \text{worldBoxMin.x}); \\
\text{newsplit}[1] &= (\text{split}[1] - \text{worldBoxMin.x}) \times 65536 / (\text{worldBoxMax.x} - \text{worldBoxMin.x}); \\
\end{align*}
\]

While the inverse mapping is straight forward to compute during the ray tracing phase, it can be optimized
to use exactly the same number of floating point operations as the non-quantized version (see Listing 2.3).

Still, the code must use two 16-bit-integer to float conversions per traversal, an operation that can
easily harm performance over the non-quantized traversal. Also note that the pre-calculation must use six
additional multiplications and three subtractions per ray (which, however, easily amortizes over the complete
traversal phase).
2.3. Acceleration by Object List Partitioning

Listing 2.3: Using the quantized data in the ray-hierarchy intersection implementation.

Testing with the Stanford models Happy Buddha, Thai Statue and (the extremely detailed) Lucy, an average of 10 percent total ray tracing performance loss is experienced (compared to the original BIH implementation). While this seems to stem from the slightly larger bounding boxes, the amount of total traversal steps only slightly increases by an average of 1-2 percent. At the same time the number of triangle intersections rises from 2-3 percent (Happy Buddha) up to 15 percent (Lucy). This is directly related to the limited amount of partitioning possible when using an uniform quantization, thus leading to an increased number of primitives per leaf. Still, this rather good and unexpected initial performance, even for these extremely detailed scenes, can be further optimized.

When analyzing the converted floating point value in detail, it can be seen that only 19 of the 32 bits are actually used (the others are constant when discarding the converted integer numbers 0 and 1). But a simple binary shift of the data, Or-ing the constant bits to it, and loading into a floating point register afterwards is unfortunately not enough as the 19 bits obviously do not fit into the 16-bit-data that is available.

Instead it is possible to load the 16-bit-data directly into the mantissa of a floating point number, Or-ing an exponent to it and scale it afterwards so it matches the range of 0...65535 again. Moving the scaling to the pre-calculation part and using 0x40000000 (= 2.0) as exponent bits results in the following code snippet:

Listing 2.3: Using the quantized data in the ray-hierarchy intersection implementation.

To summarize, the two low-level optimizations (either using integer arithmetic or a SSE2 optimized 16-bit-int to float cast) of the quantized BIH brings the total ray tracing performance at least on par with the non-quantized version, thus outperforming or at least matching the results found in [CSE06] while saving 33 percent of the total BIH node memory. As the second variant is even more hardware friendly (as no int to float cast needed), it could become useful for future hardware implementations.

However, the main problem of such regular quantization is found in its missing adaptivity, thus introducing
the exact same problems to ray tracing as utilizing uniform grids as data acceleration structure. At the lower levels of the quantized hierarchy it is not possible to further partition a node, as the resulting quantized children will feature the same bounding planes as the father node again. Thus, see [Mah05] for more details on a hierarchical quantization of BVHs, though even using this hierarchical encoding, the published statistics show a demand for at least 8 bits per quantized float to not drastically decrease the overall ray tracing performance. Still, this relative encoding of quantized floating point numbers could also allow to easily include fixed point integer arithmetic for ray tracing [HK07, Han07] or even a full precision integer triangle test [Wäc04].

Compression

Instead of using a simple difference encoding with a fixed bit budget per float for BVHs as proposed in [Mah05], it should also be possible to use more advanced predictors and/or varying bit counts to allow for even less cache usage per node. Especially if an additional incorporation of the original primitive data into the compressed tree is used, the overall memory footprint for geometry and hierarchy could be drastically reduced. While this is still an open topic for research in the context of ray tracing, [ILS05, LI06, RKB06] deliver results for efficient mesh compression implementations that are at the same time computationally simple, thus a runtime mesh decompression for ray tracing seems feasible. First experiments using the bounding interval hierarchy and an additional simple compression based on prediction and RLE were significantly slower due to the overhead of the decompression step, but could still pay off in a full hardware implementation, using a separate decompression unit that can work in parallel to the traversal and intersection unit(s). Additionally, scenes that are not fitting into main memory and must use disk swapping (see Section 4.3) already can profit from the optional compression step to save roughly half of the disk reads and writes (mileage will vary as the ratio obviously depends on the compression scheme used and the underlying geometry data).

2.3.4 Performance

Certainly the simplicity, numerical robustness, and predictable memory footprint make bounding volume hierarchies most attractive for accelerating ray tracing [KK86, GM03]. However, performance was generally below what is obtained using \(k\)-d-trees. At the price of hardly predictable memory requirements and numerical issues during the construction of the spatial acceleration data structure, realtime performance is obtained [Wal04, Bik05, Ben06] for static and moderately dynamics scenes.

Both principle approaches of either dividing space or object lists suffered from construction routines that were far from realtime and additionally use greedy algorithms, which certainly is another disadvantage. The most successful concept is the surface area heuristic [GS87, Hav01, Wal04, WS06] (also see Section 2.4.1). As it requires a fair amount of analysis of scene geometry and twiddling, the construction for a complex mesh can easily be in the range of minutes to even days [WS01, Wal04, WDS04, WH06, SBB06]. Although a lot of research (especially on the faster construction by approximating the cost function) has been achieved (see e.g. [WH06, PGSS06, SSK07]), the resulting hierarchies are either of moderate quality [SSK07] or still take a lot of time to be constructed [WH06].

While the construction times amortize for static scenes, very moderate dynamics [Wal04], or deformables only [WS06], they are much more difficult to amortize in fully dynamic settings. Attempts to deal with fully dynamic scenes so far use regular grids [RSH00, WIK06] with all their disadvantages and are only efficient for scenes of moderate complexity.

Comparing the BIH to other ray tracing implementations by measuring ray-hierarchy intersection performance and total time to image (including the data structure construction) for common test scenes, Figure 2.22 presents results for static scenarios, while Figure 2.23 features fully dynamic scenes, demanding a rebuild of the data structure on a per frame basis.

Note that the measurements only feature an absolutely straight forward BIH implementation (using the simple global heuristic (see Section 2.4.2) for partitioning), not incorporating an additional empty volume cut off (see Section 2.4.2), approximate sorting (see Section 2.5.7), advanced heuristics (see Section 2.4.2) or SIMD opcodes/multiple threads to accelerate the hierarchy construction (as e.g. demonstrated by [PGSS06]). The only optimization done for the measurements seen in Figure 2.22 is the inclusion of 2x2
SIMD optimized ray bundles. Still, the performance improvements of the BIH over other data structures are astonishing and can even outperform the kd-tree for most practical scenes featuring detailed geometry. The construction speed to build the hierarchy is still unmatched as of today, while further refinements are still possible that will be presented in 2.5.

2.4 Partitioning Heuristics

As mentioned in the previous sections the choice of the right partitioning heuristic to build the data acceleration structure is crucial for both construction and ray tracing time. Furthermore, there is an increasing demand for fully dynamic scenes (as it is possible to a certain extent in modern game and especially demo-scene graphics engines). Therefore the merging of these normally separated steps is necessary to minimize the total time to image, but this also poses new problems for the partitioning heuristic, as will be seen later.

This section will take a closer look at some of the popular partitioning heuristics in the context of both object list and spatial partitioning using binary trees. In a second step some rather simple (and thus computationally fast) but still very efficient heuristics are proposed, concentrating on top-down hierarchy constructions which are preferred over bottom-up heuristics in current ray tracing system implementations due to their generally faster implementations. For even more information regarding "optimal" partitioning heuristics see [Hav01, HKRS02, BG03, Wal04]. For a general overview of currently used acceleration structures and the inherent partitioning heuristics, see [WMG+07].

2.4.1 Survey of Partitioning Heuristics

Perfect Volume Split

The most simplistic split-plane proposal for spatial partitioning, closely related to the partitioning using a hierarchical grid or octree, is the perfect volume split as it splits the volume of the current node in two equally sized halves. The only degree of freedom is the selection of the split-plane axis, with some possible heuristics being the round robin tactic, a randomly picked axis or using the axis featuring the longest extend (with the last heuristic usually delivering the best performance).

As the heuristic has no relation at all to the distribution of the primitives in a scene (as it never touches any triangle at all to determine the position of the splitting planes), the same problems as for hierarchical grids arise: Large spaces of empty volume are only discarded by accident (as it's always possible for some lonely objects to be scattered in a vast potion of empty space) combined with a huge amount of duplicated references (which is directly related to an increased probability for numerical errors, see Section 2.2.1), especially around vertices that feature a high valence. Still, the regularity of this partitioning can be exploited to reduce the memory footprint of a kd-tree, as one can completely omit to store split-plane positions.

Perfect Vertex Split

This (greedy) heuristic refines the previous (global) heuristic, as it uses the split-plane proposal of the perfect volume split to search for the primitive of the current node that features the smallest distance to it. This primitive is then used to determine the split-plane that will actually be used for partitioning. In the case of triangles, it is possible to either choose the triangle's vertex or bounding box edge closest to the split-plane proposal. Even considering only one vertex of each triangle as possible candidate can be an option if the mesh is closed.

By using this additional pass over the primitives, the heuristic relates the split-plane proposal to the geometry, resulting in an increased probability of empty volume being discarded (as the split-planes adapt themselves to the edges of separated objects) and less reference replications (as usually at least one primitive less is duplicated per split).

The heuristics efficiency can be further increased by considering all intersections of the triangles with the current node's bounding box to allow for more or better empty volume cut offs, though this operation is very costly in general (see Section 2.2.1).
Figure 2.22: Comparison of the bounding interval hierarchy and different state-of-the-art kd-tree implementations, using a very simple shader and 2x2 (SSE accelerated) ray bundles (640x480 pixels, measured on a Pentium 4HT 2.8 GHz, whereas [WH06] performance data were measured on a faster Opteron 2.6 GHz). The varying number of triangles results from the VRML converter that we used to generate input for InView. Since the free version of InView is limited to $10^6$ triangles, we removed invisible triangles from larger models for InView. Time to image measures the total rendering time for one picture, thus including (on demand) tree construction, ray tracing, and shading. The bounding interval hierarchy is clearly superior in memory and total time to image.

<table>
<thead>
<tr>
<th>Model</th>
<th>InView</th>
<th>WH06</th>
<th>kd</th>
<th>BIH</th>
<th>on demand</th>
</tr>
</thead>
<tbody>
<tr>
<td>Shirley Scene 6</td>
<td>1,380</td>
<td>n.a.</td>
<td>804</td>
<td>dto.</td>
<td>dto.</td>
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<td>dto.</td>
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<td>Acc. Data memory</td>
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<td>n.a.</td>
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</tr>
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<td>11.17</td>
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<td>n.a.</td>
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<td>83</td>
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</tr>
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<td>863k</td>
<td>871,414</td>
<td>n.a.</td>
<td>804</td>
</tr>
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<td>dto.</td>
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<td>13,466,176</td>
<td>5,175,936</td>
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<td>5.92</td>
<td>5.98</td>
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</tr>
<tr>
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<td>23,900</td>
<td>3,106</td>
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</tr>
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<td>1,087,716</td>
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<td>1,837</td>
<td>705</td>
</tr>
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</tr>
<tr>
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</tr>
<tr>
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<td>dto.</td>
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<td>165</td>
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<td>dto.</td>
<td>dto.</td>
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<td>3,707,292</td>
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<td>n.a.</td>
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<td>n.a.</td>
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<td>Scene</td>
<td>kd fps (primary rays only)</td>
<td>BIH fps (average 4 rays per pixel)</td>
<td>Rendering time for complete animation (msec)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>------------------------</td>
<td>---------------------------</td>
<td>-----------------------------------</td>
<td>---------------------------------------------</td>
<td></td>
<td></td>
</tr>
<tr>
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<td>3.26</td>
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<tr>
<td></td>
<td>3.34</td>
<td>3.07</td>
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<tr>
<td>BART Kitchen (110,561 triangles)</td>
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<td>1.41</td>
<td>530,561</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>1.96</td>
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<tr>
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<td>2.17</td>
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</tr>
<tr>
<td>BART Robots (71,708 triangles)</td>
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<td>1.41</td>
<td>530,561</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>1.49</td>
<td>1.49</td>
<td>537,974</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Utah Fairy Forest (174,117 triangles)</td>
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<td>1.79</td>
<td>26,780</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>1.95</td>
<td>1.95</td>
<td>11,695</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Figure 2.23: Comparison using dynamic environments (640x480 pixels, Pentium 4HT 2.8 GHz). The complete data structure is rebuilt per frame from scratch. The museum is traced using a) simple shading and b) full shading (using an average of 4 rays per pixel), where in both cases only single rays are traced. The remaining BART scenes are rendered with the simple shader, while the Utah Fairy Forest uses full shading.
Perfect Object Median Split (Leftish Tree)

The last presented simple heuristic concentrates on a perfect object list splitting resulting in leftish (or also called balanced) trees, similar to classical search tree partitioning, and was used for a long time as the standard partitioning scheme to accelerate ray tracing. However, with increasing scene complexity this simple heuristic was found to be of little use (especially when used in combination with spatial partitioning) and additionally features slow construction times due to the necessary median element search for every internal node of the tree. The main reason why this usually good performing heuristic (in the context of databases and searching algorithms) fails when used to accelerate ray tracing is found in the actual nonuniform distribution of objects in real life scenes, especially for all so called teapot-in-a-stadium scenarios. The leftish tree construction assumes that every primitive of the tree has the same probability to be intersected by a ray, while in fact surface area, volume of the surrounding bounding box and local primitive density are more important. Also, ray tracing is nowadays implemented as a backtracking algorithm to find the closest intersection along a ray, so the empty volume gap in-between objects is usually of great interest, as it is usually wanted to step over empty gaps along the ray in few steps (ideally one visited empty leaf if spatial partitioning is used) and finding an intersection directly in the next visited leaf. As the median split does not take empty volume into account (as it is designed to deliver a fast result for a point search only), a resulting problem can be seen in Figure 2.24. The split-plane will only cut off empty space by accident, as long as the distribution of primitives in the children is not equal. This results in fragmented empty volume with most leaves of the resulting hierarchy being large, but very thin slices of empty volume featuring one (or more, depending on tree depth) triangle on one of their sides. During ray tracing the ray has to traverse an extreme amount of leaves to step over empty volume resulting in a drastic performance decrease. Experiments delivered that some rays, especially those just slightly missing the objects of a scene (e.g. on silhouettes), have to visit up to 10,000 leaves for complex meshes due to the volume fragmentation, resulting in some scenes with factor 50-100 increased rendering times compared to the perfect vertex split heuristic. Note, that the same problems hold for the BIH and its variants that use single and quadruple partitioning planes (although the last variant can automatically cut off more empty volume on the lower levels and is thus significantly faster, especially when using simple scenes like 3-dimensional scanning data) along with every BVH implementation in general.

Figure 2.24: General problem of the perfect object median partitioning (using the BIH as an example): Unequal distribution (illustrated by using slightly differing triangulated spheres in this simple example scene) results in a continuous splitting of the objects instead of removing empty volume, which is crucial for ray tracing performance.

An additional problem that needs to be addressed is the search for the median element when using a kd-tree, as it is difficult to predict the median due to the inherent reference replication. [Cu96] proposes a solution using two quicksorts of the minima and maxima of the bounding boxes of the primitives to determine the median element. Although the author claims that this solution is proven correct, this can only be seen as an approximate solution, as finding an "optimal" median element must involve an exact clipping of the triangles to the node bounding box. Nevertheless, this approximation works rather well, especially for the higher levels of the hierarchy build.

Still, leftish trees also feature some nice properties that can be exploited. As shown in [Wä04], leftish trees can be used to build ray tracing acceleration structures that do not need to store pointers for their children explicitly (as the index can be calculated out of the current node's index) and additionally any stack operation during ray traversal can be omitted, as a back-pointer to the parent of a node is implicitly defined by the data structure itself. These properties might prove valuable in the context of certain hardware architectures like GPUs, where several publications (e.g. [Pur04, Les04, FS05, HSHH07, PGSS07]) propose
2.4. Partitioning Heuristics

short/fixed size stack or even stack-less ray tracing variants. Note that leftist trees also feature (by design) the smallest memory footprint possible for hierarchies, as the total amount of nodes and empty leafs is minimized.

**Surface Area Heuristic**

Presented only for the sake of completeness, a complete discussion of this widely used heuristic [GS87, Hav01, WH06, WB06] is out of context for the purpose of this work, as the main weakness of the heuristic is found in extremely long precalculation times [Wal04, Ben06] that only deliver minimal improvements (if any at all) for most common scenes (compared to the partitioning heuristics developed in this thesis, see Section 2.4.2 and Figure 2.45). Although there are various publications dealing with decreasing the precalculation times, either by using additional memory to store temporary information (as e.g. [WH06]) and/or decreasing the amount of tested split-plane proposals by quantization (as e.g. [PGSS06, SSK07]), it can still be seen as the most complicated heuristic (especially when using the additional runtime optimization tricks) that also features the by far longest construction times. Also note, that decreasing the amount of used proposals to approximate the cost-function will result in a decreased quality of the thus approximated SAH hierarchy [SSK07]. Nevertheless, the resulting data structures of the originally proposed, fully computed SAH are still amongst the most efficient (especially for complicated scenes) when only considering pure ray tracing time (and thus completely omitting the necessary construction time), so it can pay off to use full SAH trees in the context of static scene walkthroughs, where precalculation times are not too much of an issue.

It must be pointed out that the SAH is commonly seen as an "optimal" heuristic when used for building ray tracing acceleration data structures, while in fact a lot of assumptions were necessary to formulate the heuristic that clearly do not hold for practical ray tracing: The distribution of the rays is assumed to be uniform over the whole ray space (while a heuristic based on the real ray distribution as in Section 2.4.2 should generally lead to better ray tracing performance). Further, these rays are assumed to not intersect anything on their path to the node (which omits the fact that ray tracing is a backtracking algorithm that needs to find the closest intersection by inspecting multiple leafs along the ray) and the time to traverse a node or intersect a primitive is fixed to a constant (while it is obvious that e.g. top level nodes that will remain in the caches can be traversed a lot faster, along with a lot of local issues depending on the primitive distribution and overlap). Additionally, building a non-greedy SAH hierarchy is also non-trivial and as such most of the time greedy implementations are utilized.

2.4.2 New and Improved Heuristics

As seen in the previous section each of the presented simple heuristics has certain useful properties:

- **Perfect Volume Split**: Places split-planes independent of the underlying geometry.
- **Perfect Vertex Split**: Generally the overall most efficient simple heuristic, considering the total time to image.
- **Perfect Object Median Split**: Effectively limits the total amount of used memory.

Thus a solution that tries to combine the best from these simple heuristics should generally be outperforming the stand-alone solutions. Starting with the perfect vertex split as the base algorithm, it must be noticed that hierarchies resulting from this heuristic generally feature a lot more nodes than trees constructed using the SAH. Simply decreasing the maximum tree depth or increasing the maximum number of primitives that can be stored in a leaf (both classic, but way to simple construction termination criteria, see Section 2.5) unfortunately results in proportional performance decreases. So one idea is to locally limit the memory by using the perfect object median split on the lower levels of the tree instead. Finally, the third heuristic (perfect volume split) will be of use on the lowest levels of the tree, as soon as no more regular vertex (and thus also object median) splits can be found, in order to voxelize big nodes that still consist of many primitives.

So the first step is integrating the object median into the regular vertex split heuristic to automatically terminate the tree construction much earlier than before. [Mül04] proposed to use a ray tracing system
consisting of a top-level BVH along with low-level regular grids for the local objects of a scene, while these "objects" are actually determined by testing the local distribution of the primitives. If this distribution is rather uniform, the tree construction switches from the BVH to use a regular grid instead. In his work, that unfortunately uses rather abstract and simple scenes as testing scenarios, this proved to be a well performing heuristic when compared to eleven other heuristics. Even more interesting, the just recently released [LYM07] uses a top-level BVH and a triangle-strip like storage of triangles on the lower levels that is organized in an object median tree. The results are rather promising, especially for massive scenes consisting of several millions of triangles. Additionally [MB90] hinted that an "optimal" split (in the sense of using the SAH) must always be found in between the perfect volume and perfect object median split.

The key explanation why such heuristics can work is actually the distribution of triangles in a scene. While the upper levels of the hierarchy must first of all try to separate the objects of a scene (and thus maximizing the empty volume gaps between them, see Section 2.4.1), the lower levels can assume that the triangle distribution is rather uniform (teapot-in-a-stadium problem), thus an object median tree or regular grid being an obvious choice for a low-level tree. Additionally, the object median split can keep the memory usage low for locally highly detailed objects, thus preventing a portion of cache thrashing that is inherent to some rays.

The problem that remains is the detection of when to make the switch to the object median split. The heuristic proposed in [Mül04] is clearly too computationally expensive. A simpler heuristic is to use the current tree depth and/or number of primitives in a node instead to determine when to make the switch. Unfortunately, both proposals only depend upon the scene's triangle count, where actually the switch should depend upon the local distribution. In reality this simple heuristic works rather well though, even when testing very different scenes (see Figure 2.25). While a tree depth of 28 seems like the best trade-off between trace time and node memory used for the massive UNC Power Plant scene, it is 23 for the multiple times smaller Quake II game geometry. As several ideas to propose a simplistic estimator for a locally adaptive switch did not work out (otherwise a more advanced method like the one in [Mül04] must be used), a fixed tree depth of 25 was found to be a rather good trade-off for all kinds of tested scenes (see Figures 2.26 and 2.27).

![Figure 2.25: Different tree depths to switch from the perfect vertex split to object median splits compared to the ray tracing time (and additionally the node memory in the left graph). Left: UNC Power Plant (12,748,510 triangles) using single rays (trace time averaged for multiple views, thus time scale omitted). The longest tracing time takes about 1.8 the time of the shortest tracing time. The largest node memory used takes about 2.1 the size of the smallest amount of node memory. Right: Quake II (10k-30k triangles, depending on number of enemies and items) using 8x8 shafts and 2x2 rays as refinement if shaft tracing fails (trace time averaged for a whole timdemo, thus time scale omitted). Here the longest tracing time takes only about 1.14 the time of the shortest tracing time.](image)

An additional small refinement to the combination of these two heuristics can be made by abusing a heuristic popular for the construction of BSP trees: By using large triangles as split-planes, as the probability of hitting exactly those are rather high (as they can be considered ray blockers), the traversal steps for some rays are drastically decreased. While this heuristic only partially holds with increasing scene complexity (even for architectural scenes), the basic idea of the herein used variant is to only assume blockers to exist on the
2.4. Partitioning Heuristics

lowest levels of the tree. Thus, only primitives whose plane-normal is more or less aligned with the split-plane axis are considered for the perfect vertex split and/or object median search. Note the amount of memory that can be saved while providing almost constant ray tracing times (see Figures 2.26 and 2.27).

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<th>Perfect Vertex</th>
<th>plus Object Median</th>
<th>plus Aligned Triangles</th>
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<tr>
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<tr>
<td></td>
<td>100</td>
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</table>

Figure 2.26: Comparison of the number of triangle references in the leafs (upper value) and the total number of nodes (lower value) (best equals 100 percent). Reducing the number of split plane candidates to triangles whose normal is aligned with the split axis (though only on the lower levels of the tree) results in a clear reduction of the total memory used.

<table>
<thead>
<tr>
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<th>Perfect Vertex</th>
<th>plus Object Median</th>
<th>plus Aligned Triangles</th>
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</thead>
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<td>108</td>
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<tr>
<td></td>
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<tr>
<td>Leafless Maple Tree (1.26M)</td>
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Figure 2.27: Time to trace single rays (multiple numbers reflect different camera positions) (best equals 100 percent). While the memory consumption (see Figure 2.26) is greatly decreased, rendering performance remains mostly constant.

Note that the combination of the presented heuristics did not prove successful for the BIH and its variants, a fact related to the natural memory limit inherent to bounding volume hierarchies. By additionally limiting the memory by using median splits not much memory can be saved, but instead only the picking of better suited split planes is avoided. So the performance drop is directly related to the amount of memory saved.

The second step consists of including the perfect volume split heuristic into the combination of the heuristics. As such, a volume split can only be useful as soon as no more vertices are located inside the current node, or more generally speaking, if the vertex nearest to the middle of the node volume is too far away from the middle to still deliver a "valuable" split-plane. While the latter case is rather difficult to detect and rate (as there are two cases that must be distinguished: is the vertex too far away because there is a lot of empty space in-between two objects (so a vertex aligned split would prove useful) or is it too far
away because the node mostly consists of a lot of long triangles that occupy the whole node (in which a real volume split would be useful?), the case of no more vertices found inside a node is easily detectable as in that case a leaf needs to be created. If additionally the number of primitives in the node is rather high or the volume or surface area rather large (e.g. when related to the total scene size), then a volume split can help to further voxelize the problematic node. Such nodes mostly feature lengthy and/or non-axis-aligned primitives that still have a lot of empty space around them (one example being a 45 degree rotated pipe consisting of hundreds of thin, but long triangles, such as the cranes standing in front of the UNC Power Plant or the huge pipe system inside the building), a problematic scenario even for the SAH (see Figure 7.8 in [Wål04]). Further subdivisions by utilizing pure volume splits allow to approximate the primitives by a large amount of small voxels. While this does not seem like a wise idea as all rays that will intersect such primitives will have to do much more traversal steps than usual to get to the actual leafs that contain the primitive, all rays that actually miss the primitives will profit from the newly created empty leafs, as only these leafs must be traversed instead of intersecting all primitives (which is especially important if many primitives were located inside the original node).

Figure 2.28: Additional volume splits when used on the chairs of the BART Kitchen, with blue corresponding to the number of triangle intersections. Left: No volume splits, middle: if still more then 15 primitives are in a node, right: if still more then 5 primitives are in a node.

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Figure 2.29: Relative ray tracing times (100 percent being slowest) and memory footprint (100 percent being best) for the BART Kitchen (see Figure 2.28) when varying the amount of necessary primitives in a node to allow for further perfect volume splits. A value of 10-15 seems to be a good trade-off.

Note that this additional heuristic is obviously useless for bounding volume hierarchies as they cannot feature volume splits by design. In that case, two possible solutions can be thought of:

- Simply split triangles, either in a pre-process, so all triangles in a scene are more or less of the same size, or, only when needed (see above criteria). A neat trick that can help to increase performance is to not split triangles in a classical way, but to resemble the axis aligned nature of the acceleration data structure by just allowing splits along the principal planes (see Figure 2.32). This can be of great help to reduce the overlap of the newly created children of the hierarchy.

Additionally, the newly created triangles can share data amongst them, as long as a ray-triangle intersection test is used that demands for a pre-transformation step: As the triangle-plane will always be equal for all created triangles, the normal can be shared, as well as the edge-vectors of the newly created triangles. Note, however, that an actual implementation that could make use of shared data needs to be designed radically different.
2.4. Partitioning Heuristics

Figure 2.30: Additional volume splits when used for the UNC Power Plant, with blue corresponding to the number of triangle intersections. *Left:* No volume splits, *middle:* if still more then 25 primitives are in a node, *right:* if still more then 15 primitives are in a node.

<table>
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Figure 2.31: Relative ray tracing times (100 percent being slowest) and memory footprint (100 percent being best) for the UNC Power Plant (see Figure 2.30) when varying the amount of necessary primitives in a node to allow for further perfect volume splits. A value of 20-25 seems to be a good trade-off.

Figure 2.32: Different subdivision-possibilities. a) Traditional 1 to 4 refinement. b) Traditional 1 to 2 refinement (repeated). c) Axis-aligned 1 to 2 (repeated), but needing an additional detection if the triangle already features two edges aligned to the principal axes. If this is the case then continue with: d) Axis-aligned variant of a.
• It is also possible to simply ignore the fact that bounding volume hierarchies cannot employ duplicated face references by design, thus implicitly creating a kind of hybrid data structure (see Section 2.5.4) that combines object list with spatial partitioning.

Figure 2.33: One problem of axis aligned hierarchies (especially when only using split-planes aligned to vertices): All rays that just slightly miss the silhouette of a complex object will visit a large amount of leafs, resulting in multiple triangle tests along silhouettes. Such a case is not uncommon and even moderately complex scenes can feature some rays that need to test hundreds of triangles. Main problem of such rays is not the time it takes to trace (as they represent just a very small portion of the total amount of rays), but rather the cache thrashing they induce: A normal Pentium IV CPU features 8KB of L1 cache, which can be completely flushed by the data that is fetched by just one single ray!

An additional variant presented in [Kal05] uses a 3-dimensional BSP tree at the lower levels of regular axis aligned hierarchies to help encapsulate long and thin triangles. Though the results are promising, it is also accompanied by a overcomplicated tree build (how to sort triangles into these convex polyhedra?) and numerical precision issues, both for construction and traversal. The additional degree of freedom on how to optimally pick the 3-dimensional split-plane is an unsolved problem, especially as simple approximations like only using triangle-planes as proposals will not deliver optimal results on the lower levels of the tree.

One way to tackle these problems is found in [KM07]. Only few split-plane proposals are allowed (the normals of the proposed planes are determined before tree-construction, so only a point on the plane along with the index of the plane-normal must be selected by a heuristic). The published results show a decreased number of triangle intersections and traversals, along with a decreased memory footprint, due to the better approximation possibilities of the nodes of the underlying scene geometry. The downside is an extremely slow tree construction, due to the complex sorting of the triangles into the tree and the more difficult split selection procedure.

Rapid Split Proposals by the Global Heuristic

As pointed out in the previous section, the combination of the three simple heuristics did not work out for bounding volume hierarchies, so this section proposes a new efficient heuristic, originally designed for the BIH and its variants, but which was also found to perform better than other heuristics when used to construct general axis aligned BVHs [WMG+07, Sec.6.4.2].

The key to the performance of the new heuristic is the efficient construction. Assuming a split-plane to be given, the algorithm is fairly simple: Each object is classified either left or right depending on the overlap (see Section 2.4.2 for details). Then the two partitioning plane values of the child nodes are determined by the maximum and minimum coordinate of the left and right objects, respectively.

What remains is to determine the split-planes. Unlike previous approaches [GS87, Hav01, WH06, WBS06], a non-greedy heuristic is used that is cheap to evaluate, because it does not explicitly analyze the objects to be ray traced (similar to the simple heuristic presented in Section 2.4.1).

As illustrated in Figure 2.34a, the heuristic uses candidate planes resulting from hierarchically subdividing the axis-aligned scene bounding box along the longest side in the middle. Note that all candidates in fact form a regular grid. If a candidate plane is outside the bounding box of a volume element to subdivide, the recursion continues with candidates from the half, where the volume element resides (see Figure 2.34b).
The object list is recursively partitioned and bounding boxes are always aligned to object bounding boxes: If a split-plane candidate separates objects without overlap of the new child nodes, the resulting split-planes implicitly become tightly fitted to the objects on the left and right thus maximizing empty space (see Figure 2.20d). Although the recursion automatically terminates when only one object is left, it is possible to define a number of objects for which a recursion is still efficient. Thus trees become flatter, memory requirements are reduced and the traversal cost is balanced against the cost of primitive intersections. A better (locally adaptive) heuristic will be presented in Section 2.5 though.

It is important to note that the split-plane candidates themselves are not adapted to actual bounding boxes of the inner nodes, but are solely determined by the global bounding box of the scene. This is the actual difference to previous approaches, which results in an adaptive approximation of the global regular grid as close as possible and node bounding boxes as cubic as possible throughout the hierarchy.

Another reason for the high quality of the hierarchy: During construction, a large portion of the nodes will be partitioned into overlapping children. By neglecting the fact that no more useful split proposals can be found inside these overlapping regions, as it is the case when using the traditional heuristic (middle of the current node’s bounding box), some split-plane proposals will not be able to partition the current node at all, thus resulting in a too early termination. The global heuristic avoids exactly these cases too, as it does not adapt itself to the real bounding box of a newly created child node.

It is worth to mention that building kd-trees based on the global heuristic results in ray tracing times almost similar to the otherwise highly optimized kd-tree implementation for most scenes (at increased memory usage though). Actually, this behavior is not unexpected as the global heuristic will only resemble a mixture of the perfect volume split and perfect vertex split heuristic when used for spatial partitioning schemes.

**Impact of Void Area**

An additional criterion that has been found useful in various ray tracing implementations is to discard or pack empty volume as early as possible in the hierarchy construction [HKRS02]. The theoretical background can be found in [WHG84], where the authors propose the term of *void area* to describe the difference in the projected areas of the bounding volume and the bounded primitives, where the bounding volume with the minimal void area consists of the primitives themselves. It is shown that the minimization of void area automatically leads to fewer intersection calculations necessary, a fact that also can be spotted in practical experiments using non-axis-aligned hierarchies [KM07]. To efficiently discard empty space around primitives in a node, Section 2.3.1 proposes the use of an additional encoding for BIH nodes to cut off empty volume on both sides at once. To detect these cases where it pays off to actually encapsulate primitives into a tighter volume instead of partitioning the node, the loop that sorts primitives to the left and right (during a
partitioning step) additionally calculates the real bounding interval of the primitives in a node simultaneously. Afterwards, the primitive interval is checked against the size of the node interval to determine if the node interval should be tightened to better fit the primitives. Experiments delivered a value of at least ten percent space that should be empty to insert these special nodes. While it is obvious that an adaptive value (e.g. depending on the node volume or surface area in respect to the total scene size) should deliver far better results, no simple criteria were found that resulted in an overall better ray tracing performance. Note that the same procedure proves to be quite difficult for spatial partitioning hierarchies like the kd-tree, as a lot more analysis of the geometry in a node has to be made to find optimal clipping planes to tighten a node, due to primitives that only partially intersect a node. A solution (at increased memory usage though) can be found in Section 2.2.1.

Another generally useful heuristic refinement that is closely related to the topic was first described in [HKRS02]: Primitives being part of the split-plane (e.g. in axis aligned architectural scenes) may demand special treatment. Instead of sorting the primitives into both children (which should rather be resolved by a stable ray traversal) or a random assignment, it is proposed to sort the primitives into the smaller volume, thus decreasing the probability of it being intersected by the traversed rays. While this seems like an obvious extension, it must be noted that the assumption of the smaller volume featuring less traversed rays did not hold for all experiments of this work. Thus, the heuristic was extended to only change the default sorting of these primitives if it results in one node being completely empty and such an empty leaf can be created.

Dealing with Bad Split Proposals

Although the presented heuristics tend to be pretty robust, bad split proposals can occur due to the simplistic nature inherent to the presented algorithms. One case for the BIH and its variants being a partitioning that results in one child being empty and the other child holding all triangles, but having the same volume as the father node. This can happen if the node mainly consists of large triangles spanning (almost) the full volume (also see the following subsection). If this happens, no more splits on this axis are allowed (or a leaf is created if the same already happened on the two remaining axes). Optionally one can also use a counter for each axis to be able to continue with different splits later on. In that case, giving each axis the chance to retry five times has been found to provide a rather well performing trade-off. Apart from being completely unnecessary anyway, increasing the number of possible retries further can lead to a quadratic behavior of the runtime for the tree construction and should thus be omitted.

Additionally, some heuristics can profit from testing the split plane proposal to the perfect volume split proposal of the current node box. If the distance amongst the proposals is too large (related to the total length of the box on this axis), so resulting in one child to feature a much smaller volume than the other one, the heuristic can be switched to temporarily use the perfect volume split. Note, however, that this cannot be used successfully for the global heuristic.

Sorting Primitives into a BVH

As noted earlier, a top-down tree construction usually picks a split plane proposal and, in the case of utilizing a BVH, needs a second heuristic to decide what triangles will be sorted into the left or right child of the current node. While it seems obvious to just use the center of the primitive or of the bounding box (which directly relates to the amount of overlap with the estimate of the child volumes), first sorting all objects that overlap the potential splitting plane into a separate left-and-right block (like it is also needed in spatial partitioning) can prove useful. After scanning all objects, it is decided whether all objects in that block are appended completely to either the left or right block. Compared to single decisions for each object as described in [WK06] the overlap is minimized much more. Consequently, empty volume is maximized and in turn the overall performance improves (see Table 2.35).

Note that this heuristic actually resembles the behavior of a ternary tree (see Figure 2.36), in which the large and/or overlapping triangles will be sorted into the middle child of a node. The new heuristic just delays the construction of this (now optional) third child by one partitioning step and thus even improves the results that can be achieved by using a ternary hierarchy. Experimental results running a highly optimized ternary hierarchy showed that the third child actually is of little use for a large percentage of the nodes' partitioning, thus making the extra cases that must be handled during ray traversal and construction obsolete.
<table>
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<tbody>
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</tr>
<tr>
<td>Car 3</td>
<td>88</td>
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<td>Interior 09</td>
<td>87</td>
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<td>Sponza</td>
<td>99</td>
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<tr>
<td>Quake II</td>
<td>92</td>
</tr>
<tr>
<td>Happy Buddha</td>
<td>100</td>
</tr>
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</table>

Figure 2.35: Trace time in percentage when comparing the new sorting heuristic (one decision for all primitives in the left-and-right block) against the commonly used (separate decisions for every primitive based on the overlap). The Happy Buddha shows almost no difference at all, as this mesh solely consists of small triangles, thus making both heuristics perform almost the same decisions. All other scenes that feature a mixture of small and large triangles profit from the simple extension, with the exception of the Sponza and Quake II scene on the other hand, as they mainly consist of large triangles only, a general problem for BVHs (see Section 2.4.2) that cannot be fixed by only using a different sorting heuristic.

Figure 2.36: Illustration of a ternary tree partitioning.
A second option which is especially popular for BSP trees is the optional storage of primitives in inner nodes (e.g., as they extend the whole volume). While it can work wonders for some simple (and especially axis aligned) architectural scenes, as large ray blockers (e.g., wall, floor and ceiling) will be stored directly inside internal nodes, thus leading to immediate intersections with the ray(s). Note that this has not proven to be successful for more complex scenes, especially as the number of triangles that would need to be stored into one single node can become fairly high. However, see [EGMM07] for a possible and very interesting implementation.

BIH and Large Primitives

One might argue that the bounding interval hierarchy (and also its variants) performance suffers when encountering a mixture of small and large geometric elements. While this is partially true, it is also true for spatial partitioning schemes.

In this situation a kd-tree subdivides the scene by inserting more splitting planes. This results in deeper trees, a duplication of object references, and an overall increased memory footprint. Deeper trees increase traversal steps and the hierarchy construction timings.

The performance problem of bounding interval hierarchies in such a scenario can be spotted by the example of the BART robots (see Figure 2.23). The scene is made up of large triangles for the streets and houses, but also features a lot of finer geometry like signs or the walking robots. As the large triangles cause large overlapping volumes in the hierarchy, an early pruning of the tree becomes impossible and more triangles per ray have to be tested.

Another simple example of an unfortunate mixture of large and small triangles, that is exceptionally hard to handle for simple heuristics, would be a pipe (triangulated by large but also very thin triangles) featuring a detailed valve (see the simplified situation in Figure 2.37). To detect the (seemingly obvious) partitioning of these objects is actually very difficult: Sorting by overlap will most certainly result in some of the large triangles going to the left, some to the right child. While the previously presented left-and-right sorting heuristic offers at least a 50 percent chance to successfully partition these objects, this heuristic can be even further improved to handle such special cases: In the case of one side being completely empty and the left-and-right section non-empty, the overlapping primitives may be forced to go to the empty section if the extent is large enough. This allows the hierarchy construction to insert further nodes later on that will either cut off more empty space or even achieve a better partitioning of both separated objects (see Figure 2.38).

Still, a classic workaround in a rendering system is to subdivide large objects beforehand. In order to moderately increase memory, the objects should be divided by planes perpendicular to the canonical axis (see
2.4. Partitioning Heuristics

<table>
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<tr>
<td>Happy Buddha</td>
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</table>

Figure 2.38: Trace time in percentage when comparing to the updated left-and-right sorting heuristic. Scenes featuring large triangles once again profit from the refined, but still simple heuristic update.

Figure 2.32). While the memory consumption now increases similar to the kd-tree, it is still possible to a priori determine memory consumption.

**View Dependence**

By merging the tree construction with the ray tracing phase (one step of it being the easy inclusion of an on demand tree construction as will be demonstrated in Section 2.5.5), a general increase in performance can be expected. In the context of non-static scenes the hierarchy can then be optimized for the real distribution of the used rays, resulting in better performance for both construction and tracing time [Kal05]. A simple trick to optimize the hierarchy for a specific set of rays is to minimize the number of leafs that need to be visited by the primary rays, which can be achieved by forcing the first split-planes to be aligned to the camera (see Figures 2.39 and 2.40).

Figure 2.39: Forcing the first split-planes to be aligned to the camera can be achieved easily for the global heuristic, as only the initial bounding box for the split-plane candidates needs to be adapted.

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<td>Garage</td>
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Figure 2.40: General decrease in ray tracing time due to first splits being aligned to the camera. The Garage scene is the only mesh that is slightly harmed by accident (less efficient partitioning due to obviously differing split-plane proposals, as the heuristic is independent of the underlying geometry).

An obvious extension to this simple scheme is to choose split-plane proposals in a more general way, depending on all the actually used ray positions and directions, although more research on this open topic will be necessary to propose a generally usable on demand partitioning heuristic.
2.5 A Unified Acceleration Framework

As there are still problems inherent to the actual construction process of acceleration data structures, especially in the practical implementation of common algorithms and partitioning schemes, this section proposes a new in-place construction scheme working on an a priori limited memory block. In the following, the framework will even allow for a simple inclusion of advanced ray tracing features such as an on demand construction and support for highly massive scenes that demand disk swapping functionality on consumer hardware. The new method additionally allows to formulate new algorithms that further decrease the total time to image by inspecting and finally merging ray (bundle) traversal, hierarchy construction, and partitioning heuristic.

Limited Memory

While the efficiency of the ray tracing algorithm heavily depends on how the search structures are built, memory management during construction is an important, but often neglected issue. While hierarchically partitioning the list of objects [WK06, WMS06, WBS06] allows one to predict the memory footprint, these techniques based on bounding volumes can suffer from inefficiencies caused by large objects and the convexity of the applied bounding volumes [WK06, Wal04, Sec.7.3.3] (also see Section 2.4.2). Partitioning space [Kel98, RSH00, Hav01, Wal04, RSH05, Ben06, WIK+06] alleviates these problems, however, objects now can be referenced multiple times as they may intersect multiple elements of a spatial hierarchy and it had been infeasible to efficiently predict this multiplicity and thus the memory footprint in a general way. While standard implementations use growing memory pools or a self managed scheduler of all available memory, some applications (like games and industrial rendering packages) demand complete memory control and may need to force a limitation of the usable memory due to other data like texture-maps and the geometry itself or, in the case of special hardware (video game consoles, GPUs or upcoming ray tracing hardware), because there is no possibility of using virtual memory as a fall back solution. While it is of course also possible to limit memory in a standard construction process of the hierarchy, a depth first constructed tree will spend the available memory unevenly, resulting in extremely fat leaves for some branches of the tree. Constructing the tree breadth first (by using an optional priority queue) resolves this issue, but demands large portions of temporary memory, due to the replication of primitive references that need to be kept on a stack (see Section 2.2.2).

Automatic Termination Criterion

Another criterion for the quality of the ray tracing acceleration structure is the truncation of branches of the tree. While the standard procedure involves a maximum tree depth and a minimum number of primitives per leaf to avoid excessive splitting, resulting in too many traversal steps later on, [HB02] proposes a maximum tree depth \(d_{\text{max}} := k_1 \log_2(N) + k_2\) and experimentally fixing \(k_1\) and \(k_2\), but this global criterion cannot adapt itself to locally changing behavior of the geometry (as it is the case for all teapot-in-a-stadium problems). The same problem happens for the SAH (see Section 2.4.1), as it features global parameters that control the automatic termination. Also see the closely related proposal in [SF91].

While the partitioning heuristic presented in Section 2.4.2 also features an automatic termination criteria by building a leftish tree at the deeper levels of the hierarchy to limit the number of traversals, it is still unclear on how to efficiently propose a locally based switch to the median split heuristic.

2.5.1 Constructing Hierarchies in an a priori fixed Memory Block

In the following a solution to the memory footprint prediction problem of spatial partition hierarchies is presented, which in addition can avoid memory fragmentation completely and can work completely in-place, using no additional temporary memory. As a result of this new algorithm, a fully automatic, locally adaptive termination criterion that can be used in addition to previously used heuristics is proposed.

Termination by Bounding Available Memory

In fact only two modifications of the construction procedure found in Section 2 are sufficient to construct a tree in an a priori fixed piece of memory.
First, the argument list of the construction procedure is extended by passing along a continuous piece of memory along with its size. Instead of terminating the hierarchy by controlling the maximum depth of the tree, a leaf node is constructed, if the reference lists resulting from the classification step plus the size of the new tree nodes for the children do not fit into the given memory block anymore.

Second, we predict the memory consumption of the new sub-trees in the sorting step and accordingly schedule the available memory to the new children when continuing recursively. We therefore compute a prediction $p \in [0, 1]$, which schedules the fraction of memory for the left child. The remainder of the memory is scheduled for the right child. Note that for branches with more than two children we need a prediction $p_i \in [0, 1]$ for each child $i$ with the condition that all $p_i$ sum up to one, so in fact any convex combination will do.

Contrary to the classical clause, which terminated the tree by one global depth parameter, the depth now is implicitly controlled by the scheduled memory. The scheduling allows to locally adapt the depth, which is superior to the previous exponential bound. Scheduling in addition replaces the second classical termination clause, which limited the number of items worth a further tree subdivision. The same effect now can be obtained by just providing a smaller memory block upon the initial call of the construction procedure.

The procedure succeeds as long as there is sufficient memory to store at least one node along with the list of references to all objects in the scene. This is a reasonable assumption as exactly this data must be provided upon each procedure call anyhow.

The construction algorithm now can be implemented to run on limited memory and in-place (see the illustration for a binary tree in Figure 2.41). Instead of single memory allocations for nodes and lists, the memory management reduces to one block allocation before the initial call of the construction routine.

Implementation

The new concept is easily verified by just implementing the termination condition using any existing hierarchy construction implementation at hand. Although the memory allocation remains untouched in that case, the memory footprint then implicitly becomes controlled by the new termination criterion.

A more efficient implementation still follows the outline of Sections 2 and 2.5.1, however, memory management can be simplified: Before calling the construction routine a memory block is allocated in which a root node along with the object references is stored in sequential order (similar to Figure 2.41a). Inside the routine the classification uses the memory layout as illustrated in Figure 2.41b: For the next unclassified object we decide, whether it belongs to the left, right, or both children in the subtree. The first case just requires to increment the $l$ variable. For the second case the last element of the left-and-right block is moved to the front to make space for the new element on the right. In addition the last unclassified element is moved to the vacancy left by the element classified right. The last case requires to replicate a reference: The current element is moved to the front of the left-and-right-block and again the last unclassified element needs to be moved to the just created vacant memory position. If not enough memory is available for replication, the routine has to create a leaf node by copying back the right block and left-and-right block.

The creation of the children will only be performed, if there was sufficient memory for two nodes and the total number of references including the replicated ones. According to the memory layout in Figure 2.41c, some elements from the left list have to be moved to its end to make space for the nodes and the left-and-right block has to be copied to the end of the left list. Based on the classification we can use the proportional heuristic (see Equation 2.1) to compute the offset $m_0 + [p \cdot (m_e - m_0)]$ of memory scheduled for the right subtree. Then the memory block of the left-and-right with the only-right items has to be moved to this offset.

Using this memory layout, the optimization of sharing the left-and-right block of two leaf nodes [Kel98] comes in handy and allows for one additional partitioning level in the tree without the cost of additional memory by directly using the overlapping reference blocks.

While the above description treated $kd$-tree construction, it applies as well to bounding volume hierarchies by just omitting the part that allows for reference replication. It is obvious, too, how to generalize the algorithm for $m$-ary trees.

Taking a closer look reveals memory fragmentation, unless the memory scheduled for a subtree is completely used. This is trivially avoided by recognizing the prediction $p$ as an upper bound on the memory
a) Upon procedure call

<table>
<thead>
<tr>
<th>$\downarrow m_b$</th>
<th>$\downarrow m_b + n$</th>
<th>$\downarrow m_e$</th>
</tr>
</thead>
<tbody>
<tr>
<td>object references</td>
<td>free memory</td>
<td></td>
</tr>
<tr>
<td>·</td>
<td></td>
<td>·</td>
</tr>
</tbody>
</table>

b) Indices during in-place sorting

<table>
<thead>
<tr>
<th>$\downarrow m_b$</th>
<th>$\downarrow m_b + l$</th>
<th>$\downarrow m_b + n$</th>
<th>$\downarrow m_b + lr$</th>
<th>$\downarrow m_b + r$</th>
<th>$\downarrow m_e$</th>
</tr>
</thead>
<tbody>
<tr>
<td>·</td>
<td>left</td>
<td>unclassified</td>
<td>free memory</td>
<td>left and right</td>
<td>right</td>
</tr>
</tbody>
</table>


c) Creation of children

<table>
<thead>
<tr>
<th>$\downarrow m_b$</th>
<th>$\downarrow m'_b$</th>
<th>$\downarrow m'_b + \lfloor p \cdot (m_e - m'_b) \rfloor$</th>
<th>$\downarrow m_e$</th>
</tr>
</thead>
<tbody>
<tr>
<td>·</td>
<td>2 nodes</td>
<td>memory for left child</td>
<td>memory for right child</td>
</tr>
</tbody>
</table>


d) Alternative memory layout in pre-order

| ·                       | ·                       | left subtree                                             | ·                       | right subtree                | ·                |

Figure 2.41: Illustration of a contiguous block of memory for a binary kd-tree. a) Upon a call the construction procedure is provided the memory block starting at $m_b$ and ending at $m_e$ along with the number $n$ of objects. b) In order to enable in-place sorting the array is partitioned into regions for objects either on the left or right, or in both children, the unclassified items to be processed, and the remaining free memory area. See the text for the complete algorithm description. c) If children are created, the memory for two nodes is taken from the provided memory block and the remainder starting at $m'_b$ is scheduled for the two children according to the prediction $p \in [0, 1]$. d) A memory layout in pre-order improves cache performance, because less memory must be accessed.
scheduled to the left child and just linearly writing the tree data structure into the memory. This in addition increases data coherency which is beneficial with modern processor cache architectures. A practical solution is to simply move the primitive references of the still to be constructed subtree to fill the previously unused memory and as such the free memory at the end of the array is maximized again.

Proceeding that way may seem unfair, as the rightmost branch potentially receives the most memory, however, this can be compensated by modifying $p$. A simple improvement is to change the order, in which the children are built recursively. This can be done randomly or by changing the order if the tree depth is odd. Another possibility is to delay the construction of the larger child, in the hope of the smaller child not using all of its memory.

### Alternative Memory Layout

The classic technique of storing the children as an array (see Figure 2.41c) allows one to use only one pointer. Storing the tree in pre-order (see Figure 2.41d) even more memory can be saved though: A pointer followed by its subtree points to the next subtree [Smi98, GM03]. While leaf nodes are directly followed by references to objects or the objects themselves, an inner node is followed by e.g. the splitting plane offset. This saves one level of indirections and results in more cache coherent memory access, but actually did not perform noticeably different from the classic layout due to a slightly more complicated traversal. In addition it does not allow for the last level optimization possible in $kd$-trees (sharing of references in neighboring leaves).

Especially for a hardware design it would be beneficial to have two separate memory areas with each a separate memory controller for the inner and leaf nodes, respectively. The ray tracing implementation in the Intel Performance Primitives uses such a layout: First the tree nodes are enumerated, which are followed by the object references. With our algorithm it is also possible to schedule a chunk of nodes in order to increase cache performance. Ideally this would be in van der Emde Boas layout (see Section 2.6.3), however, a penalty is paid, if nodes must remain empty.

#### 2.5.2 Scheduling Available Memory by Prediction

In order to predict the ratios $p_i$, we first take a look at bounding volume hierarchies [WBS06, WK06, WMS06], which recursively partition the list of objects and thus lack reference replication. Built to full depth, the number of inner nodes in the tree plus one is equal to the total number of objects, which is true for any subtree, too. As a consequence the relative frequency

$$p_i = \frac{e_i}{\sum_{j=1}^{m} e_j}$$

(2.1)

determined by the count $e_i$ of objects to be sorted into the $i$-th child of an $m$-ary node, respectively, exactly predicts the fraction of memory to be scheduled for the $i$-th subtree.

If the tree is not built to full depth, the prediction remains optimal in the sense that all subtrees will receive memory proportional to their number of objects. This means that all leaves will store about equal numbers of objects as they become about uniformly pruned and subtrees containing more objects will be deeper. To summarize, the new termination criterion along with proportional prediction allows one to build efficient bounding volume hierarchies in a predetermined contiguous block of memory.

### Prediction in the Presence of Reference Replication

There are situations, where bounding volume hierarchies suffer from severe performance penalties: For example these are encountered when bounding boxes of objects expose big regions of overlap, when axis-aligned boxes are used to bound non-convex objects, or when empty space cannot be separated as e.g. for diagonal pipes [Wal04, Sec.7.3.3]. Then, instead of subdividing object lists, usually spatial partition schemes are employed.

Considering hierarchies such as e.g. $kd$-trees [SS92, Kel98, Wal04, Ben06] that potentially replicate references to objects that intersect splitting planes, we still can use the prediction in Equation 2.1 to schedule the memory for the children. This simple prediction overcomes the static termination by tree depth
as it does not just cut the deepest branches of the tree but uniformly bounds memory proportional to the number of references in a subtree as mentioned before.

However, the prediction relies on local information only and cannot predict the reference replication in the lower levels of the tree. Hence there may be a discrepancy between the prediction and the optimal subtree memory footprint. As a consequence some subtrees can be pruned by insufficient memory, while others cannot use the scheduled memory. Although this effect vanishes as more memory becomes available, it strongly implies that reference duplication should be minimized by e.g. preferring split-planes aligned to the bounding boxes of the objects [RSH05] in order to allow for potentially deeper trees and consequently more efficient culling.

Our experiments indicate that the aforementioned degradations in performance are hardly noticeable. This is strongly supported by [WH06]: The $O(n \log n)$ behavior allows to draw the conclusion that the observed average case reference replication is at most linear in the number $n_t$ of objects, which is a perfect justification of the proportional heuristic.

**Optimal Scheduling**

However, from theory [MR95] a worst case behavior of $O(n_t^2)$ is known and thus, although maybe rare, situations must exist, where the prediction will not work well: The number of replicated references is proportional to the surface area of the object. Long, thin objects will create references proportional to the length of their longest side, while more extended objects will create references quadratic in the longest side. As long as objects are small with respect to the total scene size, this effect will not be noticeable as it happens only on some levels of the hierarchy. For a Mikado/Jackstraws-game like scene, however, the effect will be quite noticeable (see e.g. the BART Museum scene as mentioned in Section 2.3.1). Note that this is true for any ray tracing algorithm.

Another situation, where proportional prediction will fail locally is a prediction of $p = 0.5$, where all objects sorted to the left are randomly placed, whereas the objects on the right describe regular structures. While the right subtree will have minimal reference replication, the left one will have a lot, however, both subtrees were given the same amount of memory. Situations like this are not theoretical: Trees around a building are a setting, where the prediction can be not optimal locally. A similar scenario are highly detailed objects (like 3D scans) in an otherwise rather crude modeled (architectural) environment.

Possible worst case scenarios for the proportional prediction (see Equation 2.1) were characterized. However, a vast set of experiments did not show any failures of the proportional heuristic, and in fact any probability density, relative frequency, or measure of importance $p \in [0, 1]$ can be chosen instead.

While it seems obvious to include the SAH (see Section 2.4.1) as an estimator, one must note that the SAH tries to estimate costs for finding ray intersections, not predicting the total memory usage of a hierarchy or subtree. Instead an "optimal" prediction must estimate the local complexity or overlap of the underlying geometry of a node. As such, estimators including the surface area, volume and (projected) lengths of the included primitives were tested. While the results of these predictors (and also their combinations) were rather disappointing, simply summing up the projected lengths of the children primitives onto the current axis and averaging with the proportional prediction resulted in a slight speed increase (about 10 percent on the average) over using only the proportional prediction. Actually, the problem of estimating the total "optimal" memory needed for a subtree is once again coupled to estimating the local distribution and overlap of the included primitives and as such a computationally intense problem as the prediction must differentiate between a rather linear or rather quadratic behavior of reference replication happening in the complete subtrees.

An additional possibility to optimally estimate the amount of memory for each subtree in the context of an on demand construction (see Section 2.5.5) for dynamic scenes is to reuse intersection information from previous frames. By marking each primitive that was previously intersected, then counting the flagged primitives for each child during construction and using these numbers instead of the total number of primitives to estimate $p$, the construction will adapt itself to previously intersected geometry. Note that this not only works when intersecting exactly the same geometry in the new frame again, but due to the natural locality of neighboring triangles during the sorting process, implicitly includes nearby geometry. As this prediction may fail for rapidly changing scenes or fast camera movements, it is still possible to average the estimator
2.5. A Unified Acceleration Framework

with the standard proportional prediction to be able to better adapt to newly visible geometry.

2.5.3 Time and Memory Complexity

The construction procedure actually coincides with a quicksort. While it is easy to see that finding the pivot element in fact corresponds to selecting a splitting plane, it is rather unconventional that some elements are generated by reference duplication during sorting. However, their maximum number is bounded by the a priori fixed memory footprint.

The analysis of the quicksort algorithm is readily available in standard textbooks on algorithms and data structures and states an average $O(n \log n)$ runtime in the number of $n$ of object references to be sorted. This matches the findings in [WH06]. The actual runtime depends on how the split is selected and on how the references are ordered. Of course there exist geometric configurations, which violate the assumptions of [WH06] and cause the worst case runtime of $O(n^2)$ as can be seen in the initial segment behavior of the graphs documented in [WH06] and the experiments using the BART Museum scene as mentioned in Section 2.3.1. This is in accordance with the theoretical results in [MR95]. By limiting the maximum amount of reference replications (as it is implicitly done by providing the fixed memory block of size $m$), the runtime of the construction can now actually be forced to always run on an average of $O(m \log m)$ instead.

The question is now how to choose the size $m$ of the memory footprint. It must be sufficiently large to store at least one node along with all object references. But there are more choices and considerations:

1. Providing memory $m = \alpha \cdot n$, proportional to the number of objects to be ray traced is more reasonable. The factor $\alpha > 1$ then represents the amount of allowed reference replication (see also Figure 2.45). This is the most practical choice and was actually motivated by the findings in [WH06], because an $O(n \log n)$ quicksort can only touch $O(n)$ memory.

2. Choosing $m = \beta \cdot n_r$ proportional to the number $n_r$ of rays to be shot, which in some path tracing algorithms is proportional to the number of pixels, exposes an interesting relation to the classical Z-buffer used for rasterization: The memory footprint is linear in the number of pixels, however, only if the memory required to store the scene objects does not exceed this limit. As will be illustrated in Section 2.5.8 on numerical experiments (see also Figure 2.46), it becomes feasible to amortize the hierarchy construction time by controlling it by the number of rays shot. An extreme example is a single ray, where no hierarchy needs to be built and the triangles are just tested in linear order.

3. Providing the maximum available memory reduces the risk of penalties from bad memory predictions. However, building the tree adaptively then can result in subtrees incoherently scattered over the whole memory block. Even worse, the construction of $kd$-trees can fill arbitrary amounts of memory due to reference replication. Consequently this last option is a bad choice.

Besides these obvious sizes and combinations thereof, it is convincing that there exist other useful methods. Note that ray tracing complexity is not only determined by the cost of constructing the hierarchy; in fact the complexity later on is ruled by the backtracking search, as indicated by the second item of the above enumeration.

Improving Amortization and Prediction

An improved balancing of the cost of building the acceleration data structure and traversing it by rays can be achieved by building the data structure deeper, where actually more rays traverse. Therefore, an on demand procedure can be modified: Instead of building the hierarchy, where the ray traverses, only a certain number of levels of the hierarchy are built. When a next ray intersects the same parts of the hierarchy, the hierarchy is extended again by a certain number of levels. The more rays share the same parts of the hierarchy, the better the building cost becomes amortized, so if only one ray passes through a certain region of space, building the hierarchy does not pay off and automatically is omitted. The number of levels to be built at once should be higher in regions of higher ray density. Thus the number of additional ray object intersections can be reduced.
As the hierarchy is not fully built, but objects of the parts under construction become already intersected, marking these objects or arranging them in a priority queue by a least-recently used strategy allows for improving memory prediction as well as split-plane selection heuristics. It is reasonable to assume that unmarked objects or objects at the end of a priority queue will be intersected less probably. Thus they could be separated by selecting suited split-planes and only marked or least recently intersected objects could be considered to select split-planes. On the other hand memory scheduling can be adapted to count only marked or more recently used objects instead of all objects. Similar information can be inferred by using intersection information from a previous frame.

To justify the assumptions, the early rays must be distributed proportional to their final density in space. This is approximately achieved by first computing an image in a low resolution and then increasing resolution (even in a multi-resolution fashion). Another possibility is to trace a full ensemble of rays simultaneously to improve the predictions; this will be discussed in more detail in Section 2.6.2.

2.5.4 Acceleration by Hybrid Partitioning

As now both approaches of partitioning space (e.g. kd-trees) and partitioning object lists (e.g. bounding volume hierarchies) fit a unified framework, it becomes straightforward to use a hybrid approach to optimize performance. The split selection then tries to first divide the list of objects unless this is inefficient and reference replication by spatial partition becomes unavoidable. The corresponding ray traversal must distinguish between nodes with and without reference replication.

Otherwise it is also possible to allow a BVH implementation to simply replicate primitive references by setting the left and right child clip planes to the same value and sorting overlapping triangles into both children. While the ray traversal can still be used unchanged, the construction procedure must switch to the kd implementation as some primitives will now only partially intersect the node volume (see Section 2.2.1).

The problem of when to switch from BVH to kd or vice versa is topic of future research, as it demands a lot of analysis of the underlying geometry, a fact which makes dynamic scenes impossible right now. Some simple heuristics can include the overlap of the BVH children or the depth of the tree, while the latter depends too much on a particular scene type, as some scenes like architectural meshes may demand kd style splits on the top levels of the tree to split large triangles, while others such as 3D scans or other highly tesselated geometry may demand BVH splits on the top levels and kd splits only to cut off empty volume of single, rather big triangles at the bottom levels.

2.5.5 Construction on Demand

Construction time and memory footprint of the acceleration data structure can be further optimized by constructing it only where rays traverse, i.e. where geometry is "visible". The on demand construction removes the classic separation of traversal and construction routines and as such is beneficial for all kinds of dynamic scenes [RSH00, WBS03a, LAM03, SWW+04] and/or large scenes that feature a high depth complexity, thus blocking a big portion of the total geometry. While traditional approaches (e.g. [DGP04, SMD+06]) tend to be rather impractical due to the immense amount of temporary data and as such are rarely used in industrial rendering packages, the scheme presented in this section allows to omit temporary memory completely while at the same time feature an efficient on demand construction. Since all object sorting is done in-place, only a flag is required to mark nodes that have not yet been subdivided. Upon traversal of a ray, the subdivision routine is called if the flag is set.

Therefore, all non-normalized nodes have to store the temporary information necessary to proceed the construction procedure upon the next visit, into the free memory section at the end of the current block. If the scheduled free memory does not allow to store this additional information due to the locally limited free memory, or if all objects contained in it fit into the caches (e.g. L1- or L2-cache), the regular recursive construction routine is called to build the complete subtree fitting in that memory block.

While the implementation is rather simple and elegant, it only saves computation time, but leaves unused memory regions, as the complete memory block is allocated once before. These empty gaps can additionally harm performance due to decreased locality resulting in cache thrashing.
However, if used in a fully dynamic environment, the information of old frames and the previously intersected geometry can be reused to get a tighter memory block by reducing the existing empty gaps and additionally limiting the total memory usage of the new hierarchy (see Section 2.5.2 for details).

In a static environment though, the limitation of memory and avoiding empty gaps in the on-demand construction at the same time is a much harder problem. The implementation in that case must fall back to use a classical separation of one node array and one reference list for the leafs that both grow dynamically during the build process. As such, a complete in-place construction is no longer possible unless the temporary information to continue the construction of unfinished subtrees fits into the standard node-data. So using the BIH and a simple partitioning heuristic, a simple on-demand construction is still possible.

By using the proposed simple on-demand extension to the ray tracing framework it is possible to render a fully detailed Boeing 777 mesh at 1280 x 1024 resolution in 3-9 minutes (depending on camera position) from scratch on a single core of an Opteron 875 2.2 GHz machine with 32GB RAM. Compared to previous approaches the system only uses a fraction of memory (see Figure 4.4). The total rendering time of one view thus matches the time that InTrace's InView/OpenRT 1.4 uses to just build the acceleration data structure of the more than 350 times smaller "Conference Room".

For a comparison of the raw performance of the on-demand build versus a classical preprocess full hierarchy construction in a stress test, as the rendering demands to almost construct the full tree anyways, see Figure 2.42. Note, that apart from the decreased total time to image, the merging of ray traversal and hierarchy construction also offers new possibilities to further increase performance by formulating locally adaptive construction algorithms, as in Section 2.5.3.

<table>
<thead>
<tr>
<th>Car 2 (542,108 triangles)</th>
<th>kd</th>
<th>BIH</th>
<th>on demand</th>
</tr>
</thead>
<tbody>
<tr>
<td>Triangle memory</td>
<td>19,515,888</td>
<td>dto.</td>
<td>dto.</td>
</tr>
<tr>
<td>Acc. Data memory</td>
<td>23,756,320</td>
<td>8,807,636</td>
<td>8,132,108</td>
</tr>
<tr>
<td>fps (1st Pass)</td>
<td>0.46</td>
<td>0.44</td>
<td>n.a.</td>
</tr>
<tr>
<td>Time to image (msec)</td>
<td>4,595</td>
<td>2,944</td>
<td>2,830</td>
</tr>
</tbody>
</table>

Figure 2.42: Comparison of the on-demand technique and our state-of-the-art kd-tree and BIH implementations, using advanced shaders that trace single rays only (640x480 pixels, Pentium 4HT 2.8 GHz). Time to image measures the total rendering time for one picture, thus including (on demand) tree construction, ray tracing, and shading. This is a stress test for the on-demand construction, because the global illumination computations requires to construct almost the full tree as is visible in the total amount of memory used.

2.5.6 Partially Updating and Rebuilding the Hierarchy

A rather standard extension to support deformable (i.e., constant topology of the underlying geometry) scenes, is updating an existing data structure instead of rebuilding it from scratch [WBS06]. This is especially easy for all kinds of BVHs, as a simple bottom-up update starting in all leafs that feature changed primitives is straightforward, updating the bounding volume of each father node out of the merged children bounding volumes, without needing to change pointers. As soon as the bounding volume is not changed from the updates anymore, the subtree update is finished. It’s also possible to stop as soon as the updated bounding volume is smaller, at the cost of a decreased quality of the hierarchy though. Note that the update process is of course also possible for the BIH and its variants.

Using the proposed in-place hierarchy build, the update process can be refined and extended to also work with spatial partitioning, like KD-trees. By searching for the root node of the subtree that includes all changed primitives using a bottom-up process, the underlying subtree can be completely rebuild afterwards. If the root of this subtree is found to be on the upper levels of the tree (thus demanding most of the whole tree to be rebuild), the procedure should restart searching for multiple, but far smaller subtrees. This results in better hierarchies than just updating the planes of the tree, while no additional temporary memory or a complicated implementation is needed, as the already existing construction code can be used without changes. If a structure of the dynamic primitives is known beforehand, as it is the case for transformations of complete objects like an avatar or a car, the splitting into multiple subtrees that must be updated or
rebuild is greatly simplified. By having such additional object information, it is also possible to add additional triangles to an object (as long as all primitive indices will still fit into the subtree memory), as the subtree can be fully rebuild and does not depend on a fixed number of primitive indices.

The rebuild procedure will work well for all kinds of local changes, like moving objects in an otherwise static scene, but will fail if locally only few triangles are animated, but simultaneously in all parts of the scene, thus resulting in a complete rebuild of the whole hierarchy. Otherwise such a scenario can only be handled efficiently by updating the nodes.

2.5.7 Improving Time Complexity by Approximate Sorting

As the hierarchy construction in fact is a sorting algorithm with a structure identical to quicksort and consequently runs in $O(n \log n)$ on the average (see Section 2.5.3), using a bucket sorting preprocess similar to [FP93], the constant of the order can be decreased. Therefore the objects are processed in three steps:

1. The resolution of the regular grid is determined by the ratio of the scene bounding box divided by the average object size (also see the just recently published [ISP07]). As a spatially uniform distribution of objects is highly unlikely, this number is scaled down by a user parameter (we used the factor $\frac{1}{6}$ for the measurements in Figure 2.44) in order to avoid too many empty cells. Each grid cell consists of a counter that is initialized to zero.

2. One point of every object (e.g. one corner of its bounding box) is used to increment the counter in the grid cell containing that point. In the end the sum of all counters equals the number of objects. Now the counters are transformed to offsets by replacing each counter by the sum of all previous counters.

3. A global object index array is allocated and using the same point of every object, the objects now can be sorted into the buckets using the offsets from the previous step. For each bucket the bounding box of the objects it contains is computed on the fly.

Sorting the bounding boxes instead of the contained objects speeds up construction by a factor of usually two to three. If a volume element during hierarchy construction consists of one container only, the container is replaced by the objects within. The resulting trees are very similar in rendering performance and size (see Figure 2.44).

This simple algorithm partitions a scene in linear time using a limited amount of memory. Even the index array can be processed in chunks. While it removes the need for a conservative 3D-rasterization of the triangles into a regular grid, the resulting partitioning suffers from similar problems as every object list partitioning (see Section 2.4.2).

Optionaly, a pre-process may combine neighboring, sparsely populated voxels or free the list of all empty voxels to reduce the amount of containers for the following hierarchy construction.
### 2.5. A Unified Acceleration Framework

<table>
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<th>BIH</th>
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</thead>
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<tr>
<td>Triangles</td>
<td>12,748,510</td>
<td>n.a.</td>
</tr>
<tr>
<td>fps</td>
<td>79%</td>
<td>100%</td>
</tr>
<tr>
<td>Construction (msec)</td>
<td>11,609</td>
<td>20,282</td>
</tr>
</tbody>
</table>

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**Figure 2.44:** Comparison of the bounding interval hierarchy with/without using the bucket sort preprocess (640x480, Pentium 4HT 2.8 GHz) to numbers taken from [WH06], where a faster Opteron 2.6 GHz processor was used. Frames per second (fps) are given relative to the "pure" bounding interval hierarchy, as several camera positions’ fps were averaged.

### Visualization of Massive Geometry

Current data sets used in industrial applications and production rendering consist of massive amounts of geometry, which usually range from hundreds of megabytes to several gigabytes of raw data. Although it was demonstrated (see Section 2.5.5) that the small memory footprint of the new construction allows massive scenes to be efficiently ray traced by simple means there still may be situations where the data does not fit into the main memory and a special handling of these massive scenes is necessary [KK86, PKGH97].

Along the lines of [WSB01], a minimal memory footprint renderer was implemented which is able to render pictures of the Boeing 777 (349,569,456 triangles amounting 12,584,500,416 bytes) using only 50 MB of RAM. If more RAM is available (1 GB were assumed for the measurements) it is possible to render a picture from scratch in less than an hour even on a standard consumer desktop PC (see Figure 4.5) with the option to render further pictures (using the preprocessed, cached data on the hard disk) and different views in only a fraction of this time.

To achieve the minimal memory usage, the previously proposed preprocessing step was used to sort the triangles into buckets and store these on the hard disk. For the rendering step, a top level bounding interval hierarchy is created out of the buckets (without needing to touch any triangle). Each bucket that is intersected by a ray creates its own tree using the on demand policy. The bucket’s triangles and the acceleration data structure are kept in a cache of either dynamic (able to grow until no more RAM is available) or fixed user-defined size. The bucket with the largest number of triangles defines the maximum memory footprint. Note that this results for free from the bucket sorting preprocess.

In this scenario the processing speed is dominated by the speed of the hard disks, although the implementation uses software caching along with an automatic global reordering of the voxels using a peano-curve [Mor66] to exploit hardware caching mechanisms when writing/reading the reordered primitives to/from disk. As mentioned in Section 2.3.3, an accompanying transformation and compression of the triangle data can further reduce disk transfers (note that this was not yet used in the presented measurements). Additionally, as the tree construction algorithm is so fast, that if parts of the acceleration data structure have to be flushed, they are just thrown away and rebuilt on demand, so no additional hard disk traffic for the acceleration structure is created during the rendering process.

When a scene description does not fit into main memory, one can also just rely on virtual memory mechanisms of the operating system to manage the memory block [WDS04]. In order to reduce page thrashing, the upper part of the inner nodes of the spatial hierarchy can be stored in a part of memory that permanently remains in main memory [WSB01, DH05, WK06]. As the subtrees memory consumption can be fixed during the construction it is also exceptionally simple to prefetch needed subtrees.

In addition, two separate memory blocks for the upper part and lower parts of the hierarchy were considered. The new termination criterion can efficiently fit the upper part of a hierarchy into a priori fixed first memory block. The lower parts are built on demand in the second block of memory. The least recently used
parts of the hierarchy become flushed, if memory is not sufficient. This procedure then somewhat resembles multi-pass algorithms on rasterization hardware.

A further possible refinement is including the usage of prefetching (see Section 2.6.1) to amortize the latency of hard disk reads, especially when using multiple threads.

**Highly Parallel, Completely Linear, Bottom-Up Construction**

Instead of using the bounding boxes as containers for a regular top down hierarchy construction, it is also possible to use the ordered boxes of the preprocessing bucket sort step directly for a bottom-up approach. As the boxes were created based on a regular grid, all containers still feature an implicit spatial ordering. This ordering can be exploited for a crude bottom-up construction:

- The original resolution of the grid used for the bucket sort is stored in three counters (note that these will be of different size for all non cubic scenes). Utilize a round robin approach to pick one counter that also must be non-zero. Divide the counter by 2 and use the axis assigned to this counter. This keeps volumes as cubic as possible during the hierarchy construction.

- Loop over all boxes (divided by 2): Merge two directly neighboring boxes along that axis and create a BIH node if both are non empty. If only one is empty, decide if an additional BIH node for empty space cut off is useful.

- Continue until only one box (the root) is left.

This hierarchical partitioning scheme does not need to sort or even touch any primitives except for the initial bucket-sorting phase, resulting in a guaranteed linear construction time $O(n + m)$, where $m$ is the resolution of the grid. Additionally, the construction can be trivially parallelized, as each thread can work on a different part of the current set of boxes due to the implicit spatial ordering. Note, however, that the number of maximally possible threads gets halved after each step, a natural limit when constructing binary hierarchies. Also note that if some of the leafs/original boxes still feature too many primitives, a regular top-down process can still be done for each of these leafs separately afterwards or even better using the on demand policy.

Obviously, the resulting tree is not of the same quality as using one of the presented top-down approaches, e.g. using the global heuristic (see Section 2.4.2), as it omits one degree of freedom: the individual selection of the splitting axis for each node is discarded in favor of a global decision. While the split proposals will still match (most of the time) on a more global scale, local decisions can and will differ. Comparing to the results recorded in Figure 2.44, the time to trace rays is mostly equal, hinting that the simple bottom-up heuristic is comparable to the global heuristic for most "well-behaved" scenes. Though, the time to construct the tree is reduced by only 20-30 percent. This can be explained by comparing the complexities of the constructions: While the top-down approach using the additional bucket sort features a $O(n + l \log l)$ complexity, with $l$ being the amount of non-empty containers, the bottom-up approach features $O(n + m)$, with $m$ being the resolution of the full grid. So, as the ratio of non-empty containers rises, the completely linear hierarchy construction will get on par with the non linear construction scheme which is a problem for all sparsely populated scenes. Another possible limitation of the runtime might be memory bandwidth problems, as reported in [Wal07].

Note that the merging of the boxes can be done in-place (offering perfect streaming capabilities) in the already available memory block allocated for the bucket sorting, but still a dynamic allocation of nodes for the resulting BIH structure must be used.

### 2.5.8 Numerical Experiments and Results

It is important to note that the new termination criterion is not a new algorithm to build hierarchies, it is much more a way to efficiently control memory consumption and allowing for a simple in-place construction. Also note that providing sufficient memory results in hierarchies identical to the ones built using the classic criteria.
Stanford Happy Buddha, \( n_t = 1,087,716 \) triangles.

TBP Kitchen, \( n_t = 181,755 \) triangles.

Car 1, \( n_t = 312,888 \) triangles.

Figure 2.45: The graphs illustrate the asymptotic behavior of the new termination criterion. For each of the three test scenes the memory ranges from \( \alpha = 5 \ldots 64 \) bytes times the number \( n_t \) of triangles. The left graph shows how the ray tracing time (excluding construction time) for rendering 512 x 512 pixel images depends on the size of the memory available for the acceleration data structure. As expected an exponential decay is observed. The right graph shows how limiting the memory first behaves linear in the number of triangles and as the trees become deeper asymptotically blends to an \( n \log n \) complexity. Since the bounding interval hierarchy (BIH) construction does not replicate references, the BIH curves become clipped once the tree is built to the maximally possible depth. Note that using the SAH efficiently requires additional memory linear in the number \( n_t \) of triangles, whereas the graphs only consider the memory used for the actual acceleration data structure.
The new termination criterion was verified using the $kd$-tree and the bounding interval hierarchy. Measurements were done using one processor core on a Core2Duo Laptop running at 1.66GHz.

For the first set of numerical experiments three scenes have been used: A car scene represents a currently very attractive domain for ray tracing, a kitchen that has a mixture of small and big, partially overlapping triangles, and the Stanford Happy Buddha as a very simple scene. As split-plane heuristics different and current widely used schemes for the measurements were used: fast, but yet efficient heuristics for $kd$ (nearest vertex to the middle split on the longest axis of the node bounding box) and BIH (global heuristic including the refinement from Section 2.5.4) as well as two different $O(n \log n)$ SAH implementations for $kd$ (using only 20 SAH candidates on each axis) and BIH (featuring the full SAH computations). The BIH SAH implementation variant requires additional precomputation time and memory for the presorted lists (16 bytes times the number of triangles in the scene) to achieve $O(n \log n)$ runtime. Yet, the additional memory consumption is omitted in the graphs, otherwise all BIH SAH plots must be thought of shifted to the right.

In Figure 2.46 we illustrate the behavior of our new termination criterion on the size of the provided memory block. Providing more memory, the hierarchies can be built deeper and as expected the time spent for ray tracing decreases. Looking precisely, it can be seen that for the Buddha ray traced with a $kd$-tree too much memory allows for too deep trees that in turn become inefficient. The construction time asymptotically behaves like $O(n \log n)$ on the average. The beginning segments of the construction time curve, however, are linear in the number of triangles as not enough memory is provided to build deep hierarchies. Contrary to the $kd$-trees the BIH curves are clipped, because no reference replication is possible.

In the second set of experiments the Stanford Thai statue was used to investigate how the construction time can be amortized over rendering time, i.e. how the total time needed to render a frame from scratch can be minimized. From Figure 2.46 it becomes obvious that this minimum depends on both the number of rays and the memory block provided. In other words: The quality of the hierarchy needs to be traded for its construction time. While the "valley" of the minimum is clearly visible for the $kd$-tree in Figure 2.46, it is less visible for the BIH (though similar), because the BIH can be constructed much faster due to the lack of checks for reference replication (see Section 2.2.1). The shape of the graphs in Figure 2.46 is easily explained: With increasing resolution first the construction time dominates before the ray tracing time takes over. The more rays are shot the deeper, i.e. better, hierarchies pay off. This depth is controlled by the new termination criterion and thus by the size of the available memory.

The above observation together with the new termination criterion is especially useful in a dynamic setting, where each frame must be computed from scratch. Here the frame rate is easily maximized by measuring the total time to image and then increasing or decreasing the size of the provided memory block from frame to frame. This allows for automatically determining a close to optimal memory block size for offline animation rendering, too.

This section introduced a new termination criterion and memory scheduling heuristic that allows to construct an efficient ray tracing acceleration data structure in an a priori fixed memory block. The resulting simplified memory management is beneficial for both software and especially hardware implementations.

Although strong arguments for the proportional memory scheduling heuristic were given, still other heuristics could be developed and explored that must be based on the local distribution and overlap of the primitives. In addition the further exploration of locally adaptive heuristics, taking the actual ray distribution into account as proposed in Sections 2.4.2 and 2.5.3 seems very promising due to experiments.

The new construction concept allows to finally merge spatial and object list partitioning in one simple unified framework, enabling a trivial inclusion of hybrid partitioning schemes.

By providing simple implementations for an on demand construction and a completely linear hierarchy construction, that are both useful for the visualization of massive geometry, unmatched performance is received for rendering images from scratch. These timings can be even further reduced by using locally adaptive predictions or even simpler by just limiting the a priori fixed memory block, thus balancing the construction time against the actual tracing time.

The principle to terminate hierarchies by memory consumption applies to other ray tracing acceleration schemes, too, as for example hierarchical grids [JW89, CDP95, KS97, WIK+06], octrees, or even ray classification [AK87]. It is straightforward to apply the scheme to classic BSP trees (see e.g. [FvDFH96] or current games) in order to determine visibility and to point clouds (e.g. photon mapping [Jen01]) for faster range searching.
Figure 2.46: Illustration of the total time to image in dependence on the provided memory size and number of rays (256^2 ... 1024^2) for the Stanford Thai Statue (n_t = 10,000,000 triangles). For both approaches the minimal time clearly depends on both the number of rays and provided memory.
Finally it should be interesting to explore the new scheme in the field of collision detection and occlusion culling with graphics hardware, where a first step was already presented in [SWK07].

2.6 Efficient Implementation

Hierarchy traversal is generally implemented as a backtracking algorithm. This section presents more details that should be considered upon an actual implementation and proposes concepts that are worthwhile to be considered for future ray tracing hardware.

2.6.1 Hierarchy Traversal

As this work mainly concentrates on the efficient construction of the acceleration data structures, it will not treat efficient traversal implementations in detail, which, however, can be found in [Wal04], [Wäc04] and [Ben06]. A more stable implementation working completely in 3 dimensions instead of using distances along the ray can be found in [HKBv97].

Ray Initialization

Each data acceleration structure that does not store complete bounding boxes per node (note that these are available implicitly when using a grid or octree) needs a proper initialization routine of the active ray segment by intersecting the scene bounding box (a robust algorithm can be found in [WBMS05] and further optimized variants in [Wäc04]), to allow for an optimized ray traversal implementation. Additionally, the ray traversal loop itself also has to update the active ray segment to avoid extra traversal steps.

If the initialization is "optimized" by omitting the ray/bounding box test, the ray will traverse nodes that actually do not intersect the ray (as in Figure 2.47), as the traversal determines intersections with the children solely by comparing the plane intersection(s) to the active ray segment. As such, unnecessary nodes will be traversed or put onto the stack due to plane intersections found along the original ray which are actually situated outside the actual node volume. Note, however, that secondary rays with fixed length (e.g. shadow rays) can omit the initial bounding box test completely.

Figure 2.47: Without clipping, A will be traversed, as an optimized traversal-implementation must assume completely valid ray segments to be able to decrease the number of calculations.

Ray Traversal Loop

When working with spatial partitioning, rays not intersecting any primitives at all will empty the stack completely. Thus a test before every pop operation is needed, to see if there are still nodes on the stack. If the scene is guaranteed to be closed, this test can theoretically be completely omitted. Note though that it is still possible (due to numerical errors, see Section 2.1.3) that some rays slip through the edges of the surrounding primitives and thus crash the implementation. A better solution is to use a sentinel as the first entry of the stack array to guarantee a stable completion of the traversal. In addition, this optimization can also be used for object list partitioning, too.


2.6. Efficient Implementation

While the standard way for binary hierarchies to save memory is the neighboring storage of the child nodes, so one pointer or index can be omitted, the traversal implementation can also make use of this alignment. In the case of $2^n$ sized nodes (e.g. 8 bytes as used in the 32-bit kd-tree), the lowest bits of the node index can directly be used to address the child node by a simple OR-operation with the sign of the ray component matching the current split axis.

Note that this can be excessively optimized when having specialized implementations for each ray octant, for example when using SIMD ray tracing (the more general solution found in [Res06] might be better suited for a compact ray tracing implementation though).

Prefetching

Modern CPUs allow the manual prefetch of data into the caches by offering specialized opcodes. As such, several experiments were made to include simple prefetching mechanisms into the ray traversal and/or ray-primitive intersection implementation. A general problem of manual prefetching is to avoid fetching an unnecessarily high number of later unused data, while still fetching additional useful data that the automatic hardware prefetcher would have missed, so the increased number of opcodes in the traversal code is amortized. As classical ray tracing can generally be seen as a rather incoherent and thus unpredictable algorithm, the only rather useful nodes that can be fetched are the ones that are put onto the stack as these are very likely to be visited later on (for object list partitioning these are even guaranteed to be visited). Unfortunately, this rather promising optimization already resulted in a slight overall speed decrease during experiments.

Another approach is to prefetch primitives when visiting the leafs of the data structure. As leafs only store references to the primitives, an early fetch of the actual (not necessarily neighboring in memory) primitive data should reduce latencies. Unfortunately, this approach did not result in notable differences either.

Variations of the presented implementations and additional excessive speculative prefetching (e.g. by inserting special prefetch nodes into the hierarchy that keep information about the underlying subtree) resulted in even more performance loss. One explanation can be found in the coherency of successive ray traversals. As a large portion of the upper level hierarchy is reused amongst rays, the caches will keep this information available and as such, prefetches cannot help here. For the lower levels of the tree, the data can be assumed to be stored cache friendly, as spatially neighboring leafs and nodes are most of the time also close in memory (due to the construction implementation) and so should be automatically fetched by the hardware prediction or are located in previously loaded cache-lines. While a prefetch of the primitive data should at least theoretically result in a performance increase due to the additional indirection, real life scenes mostly feature some kind of implicit spatial ordering. Thus neighboring triangles are likely to be already loaded into the caches. When working on permuted triangle data, prefetching can have positive effects on the ray tracing performance though.

2.6.2 Tracing Sets of Rays

Aside of amortizing the construction cost of the acceleration data structure (see Section 2.5.3), there are situations, where the additional cost of not only tracing single rays but sets of rays can amortize, too, if rays are coherent. While primary rays easily can be grouped in coherent sets, it becomes already difficult to get the same benefits for sets of shadow rays from point light sources [RSH05, WIK+06]. Refracted, reflected, or rays from random walk simulations are even more difficult. This also demands a change in paradigms, as shaders can no longer be using traditional recursive implementations, but must be able to be delayed to wait for the result of intersection calculations due to tracing. A better way is to fix the design of shaders to only allow for fully iterative computations, thus simplifying the program flow.

Using SIMD

As almost all current available hardware architectures feature some kind of SIMD units (mostly 2x or 4x wide), with even wider SIMD units proposed for upcoming hardware, the obvious way to receive more performance on the same hardware is the optimization of the ray traversal for SIMD. While it is common to implement a SIMD parallelization by tracing and intersecting $n$ rays at once [HKRS02, Wal04, Wäc04, Bik05, Ben06], renderings using a large number of secondary rays (e.g. for global illumination simulation) might prefer a
SIMD optimized single ray implementation [BP02] due to a loss of efficiency for diverging rays that are traced as one bundle [Mån07, BEL+07]. Note that increasing $n$ to be larger than the supported SIMD word width only is interesting for primary rays, and only to a certain extent, due to a certain loss in coherency on the lower levels of the hierarchy, resulting in too much overhead managing the single SIMD packages and masking too many rays, thus wasting computational power on later unused results. See [WGBK07] for statistics on the efficiency using various package sizes of SIMD style ray tracing.

**Bounding Interval Hierarchy**

Nevertheless, given a set of coherent rays, the approaches to trace sets of rays benefit from the new data structure proposed in Section 2.3.1. This is certainly due to the reduced memory bandwidth and increased cache coherency resulting from the small memory footprint and the fact that the bounding interval hierarchy always is flatter than the kd-tree. Furthermore the volume elements appear generally larger than the corresponding volumes of a kd-tree, which relaxes the conditions on coherency. Our experiments reveal that the speedup ratio from single ray to 2x2-ray-bundle-tracing (see previous subsection) is slightly higher for the bounding interval hierarchy as compared to a kd-tree.

Also note that it is possible to incorporate the optimized BVH traversal for a set of rays as found in [WBS06] into the BIH by using a traditional axis aligned bounding box hierarchy traversal and keeping full bounding boxes on the stack. The BIH is then used to only update the bounding boxes, resulting in less memory transfers as when using the complete six floats stored in a BVH.

A similar transition can be done for the Multi-Level Ray Tracing Algorithm proposed in [RSH05], although the practical usage of this algorithm is highly dubious, due to reported real life implementations gaining only a 20-30 percent performance increase instead of magnitudes (for few scenes though) that are stated in the publication.

**Reordering**

As divergent rays, especially when allowing user-defined shaders or high-dimensional simulations such as particle tracing, will fail to be traced as ray bundles, an obvious idea is the resorting and repackaging of rays into coherent sets. While even excessive resorting strategies deliver only mild coherency improvements [Mån07], this work's ray tracing system uses a straightforward ray direction octant based reordered package tracing (along with an optional spatial sorting, see Section 3.4.3 for details). By collecting a fixed number of rays per octant (1024 has proven to be a good trade-off) before starting the trace, the implementation can also increase cache coherence, as the data used in the shaders (e.g., texture maps) cannot interfere with the geometry and data structure.

Note that of course tracing coherent SIMD packages directly in the shader, instead of storing and regrouping single rays, is always preferable as it increases the performance, but cannot be achieved for complex shaders and/or all secondary rays. For more information on grouping and repacking and especially the efficiency of different schemes, see [Mån07, BEL+07].

**Sorting into the Hierarchy**

While reordering seems like a good idea to increase coherence amongst neighboring rays, the sorting algorithm itself is difficult to formulate as it has been unclear which sorting criteria of a set of rays lead to a coherent bundle.

To get the best coherency that is theoretically possible out of a given set of rays, this section proposes the combination and merge of the reordering with the ray traversal itself. Note that it does not pose any assumptions on the ray input at all, so completely unordered or divergent rays do not result in problems. The algorithm in detail:

- The rays are passed in an (unordered) array, along with an additional array of references $0..N - 1$ for each ray.
2.6. Efficient Implementation

- All rays are then initialized and clipped to the scene bounding box. The references are sorted in-place into two groups: rays that intersect the bounding box (left side of the array) and rays that miss it (right side).

- All rays of the first group are passed on to the ray traversal, starting at the root node:
  - It is tested whether the node is a leaf:
    - If not, all rays are intersected with the plane(s) and accordingly sorted into three groups: Left only, Right only or Both (or None for a BVH) (see Section 2.2.2 for an in-place implementation).
    - The traversal continues with one of the children (e.g. the one with the majority of rays passing through it, or depending on the dominant direction of the rays) and storing the ray references that need to visit both sides on a stack (see Section 2.2.2) for later use.
    - If the node is a leaf, all rays are tested against the primitives in the leaf. When using spatial partitioning, an additional step is needed: Rays that can stop the traversal and rays that did not find an intersection yet need to be separated and the traversal continues for the active rays by popping a node and its assigned references from the stack (see Section 2.2.2).

- The procedure can stop if all rays have found an intersection or if the stack is empty.

While this implementation is using the exact same amount of primitive intersections and overall number of ray traversals as using a standard single ray traversal, the proposed solution increases the coherency to a maximum, as all rays that visit similar branches of the tree are automatically regrouped and separated from the rest of the rays. Each node and each primitive has to be loaded exactly once and so the active rays can be linearly streamed as they all work on the same data. Although no experiments have been made using an optimized SIMD implementation, the presented algorithm perfectly fits a streaming model.

The algorithm can be interpreted in several ways. It can be seen as a sorting of the rays into the hierarchy, similar to the hierarchy construction itself, when sorting primitives into the nodes. It is also a repackaging of the rays, similar to SIMD regrouping. In addition it is possible to look at it as a ray tracing variant of Warnock's area subdivision algorithm [War69].

While the implementation is still unoptimized, first results are already promising: The overall framerate is just slightly decreased over the standard (but highly optimized) ray traversal. Hardware architectures that favor a streaming implementation (such as Cell or GPU) might further profit from this algorithm, especially as the rather chaotic standard ray traversal with unpredictable branches and memory fetches is reordered to a fully linear and streamable implementation working on fixed data. While the working set (the ray array and the references) can also be kept as compact as needed (the Cell SPEs e.g. only feature very limited local memory), the memory fetches needed for the hierarchy are additionally fully predictable (expressed by the nodes on the stack) and can easily be transferred during the computation of the current traversal step. This could also be exploited for massive scenes that do not fit into memory and as such directly results in less disk swapping.

A further bonus is the implicit spatial sorting of the intersection results. By passing intersection results to the shaders in the order they were computed, neighboring intersections will be passed directly afterwards, increasing the probability of sharing the same shaders and/or textures, thus enabling the use of SIMD and increasing cache coherence. The benefits are obviously increased with a growing number of rays passed to the procedure (ideally all rays that will be intersected with the scene). Note that this automatic sorting of ray tracing and shading indeed perfectly fits the idea of simultaneously simulating Markov chains (see Section 3.4), but that the actual implementation demands more research and experiments.

Along the lines of [RSH05] an additional entry point search for all rays in the hierarchy can be done after initializing and clipping the rays to the scene bounding box. To avoid loosing performance for all cases where a deep entry point cannot be found, the implementation only uses an axis aligned box (instead of the usually employed frustums) to encapsulate the rays, which in turn can be constructed very fast, compared to the rest of the traversal procedure, so not finding a suitable entry point comes at almost no cost. Still, the performance can increase by up to 20 percent, but usually remains equal, especially when used for simple scenes.
Additionally it is also possible to more tightly encapsulate the rays into a frustum or shaft for the entry point search [RSH05] and to clip this shaft during the ray traversal itself to the geometry [ORM07] which should allow for even fewer operations needed during the traversal step. Note, however, that this limits the otherwise available generality of the algorithm, as it demands that rays feature some kind of coherence to find a tight hull surrounding all rays.

Along the proposals given in Section 2.5, it should further be interesting to tightly merge the in-place tree construction with the proposed sorting of rays into the hierarchy, resulting in an optimal amortization of the tree construction time and additionally allows for a self-adopting partitioning heuristic.

Note that this section is related to an idea independently proposed in [WGBK07], but which can rely only on theoretical results, though. Still, the tables published there are worth a read.

Shaft Tracing

A very popular technique, especially to accelerate the tracing of primary rays, is the packaging of a set of rays into a shaft by calculating a more or less tight hull around the rays and using the intersection results of this hull as a generalization for the inherited rays. A general overview is given by [BWS06], while practical results and implementations can be found in [AM00, HB00, DHS04, RSH05] and new and rather elegant algorithms in [ORM07, RAH07, Res07].

As including shafts into a generally usable rendering system is rather difficult, due to user defined shaders and the resulting loss in coherence (see Section 2.6.2), this work’s implementation relies on the work of [DHS04] (using a corner ray technique) to only accelerate primary rays. Note, however, that shaft culling techniques become problematic anyhow for diverging sets of rays as it naturally happens over the distance and secondary rays, which is a very active topic in research.

When tracing corner rays only to accelerate the shaft tracing, the traversal has to stop (or fall back to regular ray tracing) as soon as the rays take different decisions. As this is a very conservative algorithm, the results are rather disappointing, especially when using it in highly detailed scenes, as becomes visible in the results of [DHS04].

Some results are presented in Figure 2.48, using the Quake II game (see Section 4.2 for the implementation scheme). The pictures show successful shaft traversing encoded in blue. Note, however, that the shafts are relatively small, related to the overall screen size and Quake II levels are of very low detail. Using highly detailed scenes like the Happy Buddha and moderate resolutions, the number of complete successful shafts is greatly reduced. Some reasons are illustrated in Figure 2.49.

Another, but completely different approach: If using dynamic scenes, the data acceleration structure construction can be changed to use the shaft to sort the primitives into, instead of the complete scene bounding box. So split-planes will be selected depending on the structure of the inherited rays and the primitives, to find intersections more efficiently during a combined shaft traversal/construction phase. As such, primitives outside of the shaft can be discarded rather quickly as they will not be intersected by the shaft anyway. Note, however, that this can only work for either a limited amount of shafts or moderately complex scenes, as it demands to sort all primitives into each shaft separately (thus featuring the same problems as rasterization hardware). The sorting performance can be increased though, by using a crude acceleration hierarchy for the primitives and/or rays.

Figure 2.48: Successful shaft traversing for the Quake II game (encoded in blue).
2.6. Efficient Implementation

Figure 2.49: Problems of the corner ray shaft tracing implementation: Ray signs will always differ if the spread of the shaft is rather large (a). In the other case, corner rays will take different traversal decisions (b,c). This can only be solved by a specialized tree construction that tries to encapsulate as much empty space as possible in one single piece (d).

2.6.3 Cache Oblivious Design

Optimally using existing hardware caches and memory bandwidth is one of the keys to performance on modern hardware [Wal02, Eri03], especially for ray tracing implementations [Wal04]. Using a cache oblivious design, e.g. cache oblivious hierarchies [BDFC00, OS02, BFJ02, Dem02], the applied algorithms are vastly independent of the actual cache size and the structure of the underlying cache hierarchy.

One popular reordering of hierarchies is the van der Emde Boas layout [BDFC00], a variation of the original algorithm that was actually proposed to accelerate the search for values in a specialized tree. The reordering results in a hierarchy of hierarchies, and as such, nodes of each individual subtree are stored close in memory. As it is likely for ray tracing to visit neighboring nodes during traversal (especially on the lower levels of the tree), implicitly fetching neighboring nodes, as these are stored in the same cache-line, can increase performance.

The actual problem implementing the van der Emde Boas layout for non balanced trees is the ordered on-the-fly storage of the nodes, and as such it is generally used as an optional post-processing step to increase performance. Using the in-place construction proposed in Section 2.5.1, the implementation can be easily tweaked to become at least cache friendlier than before. By allocating more than two nodes at once out of the free memory block, subtree nodes are direct successors in memory. A useful allocation size can be 128 bytes of memory (as e.g. the P4 fetches 2 * 64 bytes), enough memory for 16 kd-nodes (16 * 8 = 128 bytes). The construction implementation can then use this block to store the nodes of a complete three level subtree. As such, the block can only be filled by a maximum of 1 + 2 + 4 + 8 = 15 nodes, on the lower levels this will even feature less nodes as the subtrees will not be balanced anymore. Note that actually implementing this scheme requires some subtrees to hold 16 nodes, some only 14 nodes, due to the requirement that the two children of a node need to be stored directly next to each other.

An additional refinement of the implementation allows to omit the storage of pointers on a per node basis (thus each node only stores the split-plane(s)), and favor a storage per complete subtree. This can be combined with a new form of the traversal, intersecting n nodes of the tree "simultaneously", resulting in m children that are put onto the stack. Note that still using cache line alignment (31 nodes for a four level subtree and one pointer to the 15 child subtrees, 31 * 4 + 4 = 128 bytes) must result in a large amount of unused memory, due to the original tree being unbalanced. While this could prove valuable to resolve latencies, as the amount of calculations per memory fetch is increased, the overall amount of loaded and computed nodes will increase, too.

Another possible post-processing step is the reordering of references (and even the triangles themselves) in the leafs to support hardware prefetching mechanisms. One possible extension is the complete re-clustering of neighboring nodes/leafs [YM05, YLPM05].
2.6.4 Multi Threaded Hierarchy Construction

Although ray tracing is generally seen as a highly parallel algorithm, this is only true for static scenes and when omitting the construction phase completely. While this is subject to change (in fact a lot of current research is dedicated on the parallel hierarchy construction on SMP architectures [IWRP06, IWP07, SSK07, Wal07]), most approaches either demand a lot of temporary memory or cannot offer full scalability for multiple CPUs (although many publications claim this). This will pose additional problems for upcoming multicore architectures, especially as the cores must share available memory bandwidth, a problem already visible on current multicore CPUs [Wal07]. Note that a SMP capable bottom-up algorithm featuring exceptionally low bandwidth and offering perfect streaming capability has already been proposed in Section 2.5.7 and so this section will concentrate on the parallelization of a classical top-down construction.

If using an on demand construction policy (see Section 2.5.5), a quick SMP implementation is astoundingly easy using the presented framework: If a thread visits an unfinished node, the node is marked as on construction and is build by the thread. If other threads visit the node they will be stopped until the construction is complete. Note that this can be achieved by only inserting few OpenMP pragmas into the source and is rather efficient considering the low implementation effort. Of course, due to practical reasons such as the limited number of atomic operations that can be achieved per second, the implementation is far from optimal. Also, the first levels of the hierarchy construction will only include few threads that can actually do work in parallel, due to the blocking of nodes currently under construction. But without these nodes it is not possible to further descend the tree, thus leaving all other threads idle.

This can either be solved by using a bucketsort pre-processing step (see Section 2.5.7), which can be rather trivially parallelized (in analogy to [IWRP06]), and continue with the on demand implementation only for the lower levels, thus decreasing the probability of blocking other threads. Or, by additionally implementing a parallelized sorting step of the primitives during the construction of one single node, which however is difficult to balance.

A variation of the implementation is the use of an additional worklist that holds nodes that still need to be constructed and each idle thread taking one of the list. This resolves the need of using locks or atomic operations during the traversal as this is now solely needed for the queue implementation where it does not pose performance problems anymore. Note that this can also be used for a classical static scene hierarchy construction.

Offering full multi-host scalability is even harder to achieve and thus is out of scope of this work's ray tracing implementation. Though practical solutions exist to accelerate off-line rendering or static walkthroughs, interactive and dynamic scenes still pose problems for memory bandwidth and communication overhead amongst the hosts.

2.6.5 Ray Tracing Hardware

Based on the recent findings in realtime ray tracing several ray tracing chips [Hal99, WSS05, WMS06] have been designed. While the architectures can efficiently ray trace and shade bundles of rays, they can be easily improved by the presented approaches: The bounding interval hierarchy has a smaller memory footprint and the proposed simple in-place construction scheme using the global heuristic should make a full frame-per-frame hardware construction of the data structure possible as the algorithms only use simple operations and therefore the unified framework is a very good candidate for hardware implementation. Optionally, the highly parallel bottom-up approach presented in Section 2.5.7 can be used to get an approximation of the global heuristic based hierarchy construction in even linear time.

Furthermore it could be interesting to add specialized (de)compression or (de)quantization units (see Section 2.3.3) to reduce memory bandwidth, similar to texture compression schemes that are already used in rasterization hardware.

Maximizing the coherency of all traversal and intersection calculations by using the ray-hierarchy sorting proposed in Section 2.6.2 will additionally help to reduce cache thrashing and is also able to work on a very limited amount of local memory.
2.7 Summary

The work inferred a unified, hybrid framework for accelerating ray tracing using hierarchical auxiliary data structures for both spatial and object list partitioning schemes. The construction is a quicksort algorithm that consequently runs in average $O(n \log n)$ (even faster when using the on demand policy), achieved by a simplified memory management that uses an a priori allocated block of memory of fixed size. The overall memory consumption is considerably reduced by the in-place construction and the proposed bounding interval hierarchy. The work further derived new heuristics for building high quality hierarchies two orders of magnitude faster than the previous state of the art, which allows to ray trace fully dynamic geometry and consequently the new algorithms are already widely applied in science and industry.

While the surface area heuristic is mainly used to optimize the ray tracing of static scenes (as a full computation is too expensive for dynamic scenes), this thesis' heuristics are not only faster to evaluate, but can often outperform the SAH. In the future it will be interesting to continue research and understand the remaining benefits of the SAH.

While ray bundles are generally problematic for all secondary rays (i.e. global illumination simulations), assembling ray sets and exploiting coherence by efficiently computing transient data structures is a remaining challenge.
Chapter 3

Quasi-Monte Carlo Simulation of Markov Chains

While the previous chapter described efficient ray tracing algorithms, this chapter will concentrate on the actual selection and generation of the path space samples to achieve a "variance" reduction of the results. After surveying the basics of Monte Carlo and quasi-Monte Carlo integration, some general problems regarding the simulation of Markov chains using MC and QMC methods are discussed in detail. Then, new algorithms to overcome these problems are proposed along with embedding the findings into existing rendering solutions. By tightly coupling the sampling with the evaluation step (using ray tracing), the efficiency of light transport simulations can be increased.

A very brief History of Quasi-Monte Carlo Methods in Computer Graphics

While Monte Carlo integration [Sob94] has been in use for simulations since the 1950’s (developed by Ulam, von Neumann and Metropolis for neutron collisions) and even in the context of computer graphics since [App68], a growing interest in deterministic, or later called quasi-Monte Carlo methods [Shi91, Nie92] just started in the late 1990’s. While the Monte Carlo method relies on unpredictable, independent, and uniform random numbers to derive the probabilistic error bound, quasi-Monte Carlo methods exploit the fact that in practical applications the uniform distribution alone is crucial for obtaining good performance, thus avoiding locally higher or lower densities. Although the use of these fully deterministic, and thus correlated, quasi-Monte Carlo samples introduces a bias into the calculations, the method is consistent as it converges to the correct solution for Riemann-integrable functions [Hla71] and generalized for the space of square integrable functions $L^2$ with the Lebesgue measure [Nie03].

In the context of computer graphics, [Whi80] proposed the randomization of primary rays. [LRU85, Coo86b, Kaj86] extended the idea to sample higher dimensions. While random samples were mainly used to map aliasing artifacts (a problem inherent to simple deterministic techniques such as Cartesian grids) to noise, Shirley delivered more insight on the nature of when and how to use Monte Carlo integration in computer graphics [SW92, Shi94, SWZ96, Shi00]. The works of [Laf96, Vea97] provide further applications to extended path tracing techniques (see Section 1.1.1).

Although fully deterministic techniques like stratified or Latin hypercube sampling and also randomized versions such as jittered sampling were already common in use, a broad application of quasi-Monte Carlo techniques along with mathematical insight was first provided by [Shi91, HK94a, HK94b, Kel95, Kel98] (see [Owe03, Kel06] for an extended overview) along with the validation in a practical environment by the mental ray renderer [GH97].

3.1 Monte Carlo Integration

To evaluate the path integral formulation of the light transport derived in Section 1.1.1 we need to approximate $s$-dimensional integrals by the means of Monte Carlo integration: we approximate the integral of a
function \( f : [0,1]^s \rightarrow \mathbb{R} \)
\[
\int_{[0,1]^s} f(x) \, dx \approx \frac{1}{n} \sum_{i=0}^{n-1} f(x_i)
\]
it is possible to evaluate the integrand at uniformly distributed and independent random sample positions \( x_i \). The integrand thus is interpreted and estimated as an average. The big advantage of the Monte Carlo method is its simplicity and linear complexity.

### 3.1.1 Random Number and Pseudo Random Number Generation

The samples \( x_i \) are independent realizations of a uniformly distributed random variable on the \( s \)-dimensional unit cube. In addition real random numbers must be unpredictable, which is more important in the context of cryptography though.

As it is currently difficult or expensive (e.g. id Quantique provides a quantum based measurement hardware solution, Havege a computationally intense software package) to have real random number generators on standard computer hardware (a fact that may change with future PC architectures), deterministic algorithms must be employed to generate pseudo random samples featuring the above properties when a fast generation is necessary. This actually makes the random numbers predictable and the independence of the samples only an approximate property. The most commonly used generator is the Mersenne Twister, which has recently been extended and SIMD-optimized, delivering high quality, yet ultra fast pseudo random numbers [SM07].

### 3.1.2 Error Bound

As the Monte Carlo method is also called the method of statistical trials, the basic idea is to construct a random variable whose expectation is the desired integral. Then \( n \) independent realizations of the random variable are averaged in order to estimate the integral value. With the high probability
\[
\mu \left( \left\{ \left| \int_{[0,1]^s} f(x) \, dx - \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) \right| < \frac{3\sigma(f)}{\sqrt{n}} \right\} \right) \approx 0.997
\]
the error of the integral estimation is bounded by three times the standard deviation \( \sigma(f) \) of the integrand \( f \) divided by the square root of the number \( n \) of observations. The proof uses basic ingredients of probability theory, namely the law of large numbers and independence. This stochastic error bound can be derived by assuming that \( f \) is square integrable, i.e. \( f \in L^2([0,1]^s) \), which guarantees the variance \( \sigma^2(f) \) to be finite. Additionally, it must be possible to point sample the integrand to evaluate at positions \( x_i \) except for sets of zero measure. Note that \( f \) can always be designed to be of finite energy in the case of light transport simulation as it is physically motivated.

The convergence rate indicates that by increasing the number of samples by two orders of magnitude, one digit in precision can be gained, which is not very fast. However, the error bound is completely independent of smoothness properties and especially the dimension \( s \) so unlike standard tensor product quadrature there is no curse of dimension.

### 3.1.3 Time Complexity

The time complexity of Monte Carlo algorithms is defined as
\[
t_s \cdot \sigma^2(f) = t_s \cdot \mathbb{E}(f - \mathbb{E}f)^2.
\]
where \( \mathbb{E} \) is the expectation. Usually the focus is on reducing the variance \( \sigma^2(f) \) (for surveys on variance reduction see [KW86, Vea97]), while in fact the time \( t_s \) to generate a sample must be considered, too.

A common technique is to stratify the domain of integration in order to increase the uniformity of the samples. Stratification can easily be proven to be never worse than the original Monte Carlo method [Nie92] and is especially popular in computer graphics [CPC84, Shi90, KK02b] as it is exceptionally cheap to evaluate and efficiently reduces noise in images. However, most implementations are affected by the curse of dimension, as they require to draw \( n = N^s \) samples for a full uniform stratification.
3.2 Concepts of Uniformity

Using Monte Carlo and quasi-Monte Carlo integration, the distribution of a chosen point set (and even its subsets, as will become clear in Section 3.4.1)

\[ P_n := \{x_0, \ldots, x_{n-1}\} \subset [0,1)^d \]

is directly related to the quality of the obtained result [Shi91]. Thus various constructions and error bounds have been proposed to determine the quality of sequences and point sets and measure their distribution.

### 3.2.1 Maximum Minimum Distance

Utilizing samples based on the spectrum of blue noise has a long tradition in computer graphics, because the otherwise disturbing aliasing artifacts are mapped to noise, which perfectly matches the characteristics of the human visual system. One possible construction is using the Poisson disk random process [Coo86b] (see [WCE07] for a survey and efficient implementation) or apply an iterative Lloyd relaxation to random points. Note that such point sets are difficult to generate in an incremental way.

The quality of such point sets can be measured by using the mutual maximum minimum distance amongst the points on the torus. This actually provides an isotropic and shift invariant measurement of the uniformity of a point set. As found in [DK07] for rank-1 lattices (see Section 3.3.3), maximizing the mutual minimum distance

\[ d_{\text{min}}(P_n) := \min_{0 \leq i < n} \min_{i < j < n} \|x_j - x_i\|_T \]

the regions of the Voronoi-diagram of the rank-1 lattice approximate spheres as much as possible, resulting in approximately hexagonal regions (in two dimensions). Note that \(\|\cdot\|_T\) denotes the Euclidean distance on the unit torus.

### 3.2.2 Discrepancy

One measure of the deviation from a uniform distribution is the discrepancy [Nie92], with one form of discrepancy measure being the star discrepancy:

\[ D^*(P_n) := \sup_{A=\prod_{j=0}^{s-1} [0, a_j) \subseteq [0,1)^s} \left| \int_{[0,1)^s} \chi_A(x) dx - \frac{1}{n} \sum_{i=0}^{n-1} \chi_A(x_i) \right| \]

which measures the uniformity of the point set \(P_n\) by employing the maximum integration error of a point set when integrating axis aligned bounded boxes anchored in the origin. If \(D^*\) vanishes for \(n \to \infty\) the point set is uniform.

While uniform random points only achieve a discrepancy of roughly speaking \(O\left(\frac{1}{\sqrt{n}}\right)\), the lower bound of discrepancy is about \(O\left(\frac{1}{n}\right).\) The deterministic low discrepancy point sequences in the following Sections 3.3.1-3.3.3 in fact reach this lower bound and thus are much better uniformly distributed than random numbers can be.

### 3.2.3 \((\mathcal{M}, \mu)\)-Uniformity

The notion of \((\mathcal{M}, \mu)\)-uniformity relates to the stratification properties of a point set: Let \((X, \mathcal{B}, \mu)\) be an arbitrary probability space, i.e. \(X\) is an arbitrary nonempty set, \(\mathcal{B}\) is a \(\sigma\)-algebra of subsets of \(X\), and \(\mu\) is a probability measure defined on \(\mathcal{B}\), and let \(\mathcal{M}\) be a nonempty subset of \(\mathcal{B}\). A point set \(P_n\) of \(n\) elements of \(X\) is called \((\mathcal{M}, \mu)\)-uniform if

\[ \sum_{i=0}^{n-1} \chi_M(x_i) = \mu(M) \cdot n \quad \text{for all } M \in \mathcal{M}, \]

where \(\chi_M(x) = 1\) if \(x \in M\), zero otherwise.
Examples of \((\mathcal{M}, \mu)\)-uniform point sets mentioned in [Nie03] are the Cartesian product midpoint rule and radical inversion (see Section 3.3.1) based points. In addition, rank-1 lattices (see Section 3.3.3) are \((\mathcal{M}, \mu)\)-uniform, too: The Voronoi-diagram of a lattice partitions the unit cube into \(n\) sets of identical shape and volume \(\frac{1}{n}\) [Kel06]. This underlines that for \((\mathcal{M}, \mu)\)-uniformity all \(\mu(M)\) must have the same denominator \(n\).

### 3.3 Quasi-Monte Carlo Integration

As Monte Carlo simulation must employ deterministic pseudo random number generation on current hardware, and the averaging of the samples for integration makes the independence property imperceptible anyway, omitting independence allows one to increase uniformity. Using such deterministic samples in Monte Carlo integration has been named quasi-Monte Carlo.

#### 3.3.1 Point Sequences by Radical Inversion

Classical quasi-Monte Carlo sequences like Halton and Sobol' are radical inverse [HW64] based constructions. The calculation of the radical inverse \(\phi_b(i)\) of an integer \(i \in \mathbb{N}_0\) relies on its representation \((a_j)_{j=0}^{\infty}\) in base \(b\). By mirroring the digits \(a_j\) at the decimal point, the radical inverse (or van der Corput sequence) in base \(b\) is obtained:

\[
\phi_b^i : \mathbb{N}_0 \rightarrow \mathbb{Q} \cap [0, 1)
\]

\[
i = \sum_{j=0}^{\infty} a_j(i) \cdot b^j \quad \mapsto \quad \phi_b(i) := \sum_{j=0}^{\infty} a_j(i) \cdot b^{-j-1}
\]

#### Fauve Scrambling

Using the radical inverse, a full uniform stratification is only obtained when drawing an amount of samples that is a multiple of the base \(b\). As this decreases the discrepancy, Fauve [Fau92] proposed a simple remedy by a deterministic scrambling of the van der Corput sequence. The radical inverse is replaced by the scrambled radical inverse

\[
\phi_b^i : \mathbb{N}_0 \rightarrow \mathbb{Q} \cap [0, 1)
\]

\[
i = \sum_{|l|=0}^{\infty} a_l(i) \cdot b^l \quad \mapsto \quad \sum_{|l|=0}^{\infty} \pi_b(a_l(i)) \cdot b^{-l-1}
\]

yielding a scrambled van der Corput sequence. The scrambling permutations \(\pi_b\) applied to the digits \(a_l(i)\) are determined by a recursion starting with the identity \(\pi_2 = (0, 1)\). If \(b\) is odd, \(\pi_b\) is constructed from \(\pi_{b-1}\) by incrementing each value \(\geq \frac{b-1}{2}\) and inserting \(\frac{b-1}{2}\) in the middle. Otherwise \(\pi_b\) is constructed by concatenating \(2\pi_{\frac{b}{2}}\) and \(2\pi_{\frac{b}{2}} + 1\). This algorithm yields

\[
\begin{align*}
\pi_2 &= (0, 1) \\
\pi_3 &= (0, 1, 2) \\
\pi_4 &= (0, 2, 1, 3) \\
\pi_5 &= (0, 3, 2, 1, 4) \\
\pi_6 &= (0, 2, 4, 1, 3, 5) \\
\pi_7 &= (0, 2, 5, 3, 1, 4, 6) \\
\pi_8 &= (0, 4, 2, 6, 1, 5, 3, 7) \\
&\vdots
\end{align*}
\]

The scrambling actually increases the distance of subsequent points and breaks the otherwise linear enumeration (also see Figure 3.4 for a visual comparison).
3.3. Quasi-Monte Carlo Integration

Halton Sequence

Unlike pseudo random numbers, the van der Corput sequence can by default not be used for high dimensional integration: For example, generating 2-dimensional samples out of $\frac{1}{2}^2$, all samples will be $\frac{1}{2}$ in the first dimension (all even indices $i$), the components of the second dimension (odd indices) $> \frac{1}{2}$.

Therefore, Halton introduced

$$x_i := (\Phi_{b_1}(i), \ldots, \Phi_{b_s}(i))$$

where $b_i$ is the $i$-th prime number.

A problem when using the radical inverse can be spotted when picking low dimensional projections of the Halton sequence. While usually not perceptible, these low dimensional correlations often interfere with the integrands in computer graphics. For example, correlation can slow down convergence in a sequence of two-dimensional scattering events as used in Markov chain simulations. In this case, scrambling improves the uniformity (even of subsets as can be seen in Figure 3.4). In addition the minimum distance of samples is increased, which indicates an increased uniformity, while keeping its discrepancy properties.

3.3.2 Properties and Implementation of $(t, s)$-Sequences in Base $b$

One disadvantage of the Halton sequence is its demand for inexact floating point calculations: most prime fractions cannot be exactly represented in the floating point precision format. Additionally, decomposing $i$ into $a_j$ is expensive for varying $b$. Therefore, a construction working within only one base for all dimensions should be preferred.

$(t, m, s)$-Nets in Base $b$

For two integers $0 \leq t \leq m$, a finite point set of $b^m$ points in $s$ dimensions is called a $(t, m, s)$-net in base $b$, if every elementary interval

$$E := \prod_{j=1}^{s} \left[ a_j b_j, a_j + 1 \right[ b_j \right) \subseteq I^s$$

of volume $\lambda_s(E) = \prod_{j=1}^{s} \frac{1}{b_j} = \frac{1}{b^{s-1} b} = b^{t-m}$ contains exactly $b^t$ points. The star discrepancy for such sequences is

$$D^*(P_n) \leq B(s, b) b^t \frac{\log^{s-1} n}{n} + O \left( b^t \frac{\log^{s-2} n}{n} \right).$$

Smaller $t$ values result in better nets, as $t$ is the quality parameter controlling the stratification properties of the net, which are best for $t = 0$, because then every elementary interval (see Figure 3.1 for an example) contains exactly one point. Thus the pattern is both Latin hypercube sampling and stratified in the sense of [CPC84]. Note that for base $2$, $(0, m, s)$-nets can only exist up to dimension $s = 3$ [Nie92, Cor. 4.21].

![Figure 3.1: All elementary volumes of a $(0, 3, 2)$-net in base $b = 2$.](image)

$(t, s)$-Sequences in Base $b$

For $t \geq 0$, an infinite point sequence is called a $(t, s)$-sequence in base $b$, if for all $k \geq 0$ and $m \geq t$, the vectors $x_k b_1^{\cdot}, \ldots, x_k b_s^{\cdot} \in I^t$ form a $(t, m, s)$-net [Nie92], so the point sets match each other perfectly. Note, however, that $(0, s)$-sequences can only exist for $b \geq s$ [Nie92].
For a \((t, s)\)-sequence in base \(b\) the following star discrepancy is received

\[
D^*(P_n) \leq C(s, b) b^{n \log^2 n} + O \left( b^t \log^{s-1} n \right)
\]

Adding the component \(\frac{1}{n}\) to a \((t, s)\)-sequence yields a \((t, m, s+1)\)-net, one example being the van der Corput sequences \(((0,1)\)-sequence in base \(b\)). Adding the component \(\frac{1}{n}\) with \(n = b^m\) yields a \((0, m, 2)\)-net e.g. Hammersley point set for \(s = 2\) and \(n = 2^m\) points. While this adds one dimension "for free", it also restricts the maximum and minimum number of samples, as these will be obviously fixed to \(b^m\), which will not allow for incremental sampling.

**Digital \((t, m, s)\)-Nets and \((t, s)\)-Sequences**

One construction of \((t, s)\)-sequences is based on the use of generator matrices [Nie92] as it also allows for an efficient implementation, especially if working in base 2:

An \(s\)-dimensional point set \(P_n := \{x_0, \ldots, x_{n-1}\}\) with \(n = 2^m\) and \(x_i := (x_i^{(1)}, \ldots, x_i^{(s)}) \in [0,1)^s\) is called a \(C_1, \ldots, C_s\)-generated point set in base 2 and dimension \(s\), if \(C_1, \ldots, C_s \in \mathbb{F}_2^{m \times m}\) with

\[
x_i^{(j)} = \left( \frac{1}{2} \ldots \frac{1}{2^m} \right) \begin{bmatrix} d_b(i) \\ \vdots \\ d_{m-1}(i) \end{bmatrix}, \text{ where } i = \sum_{k=0}^{m-1} d_k(i)2^k
\]

and the matrix-vector product in brackets being performed in \(\mathbb{F}_2\). The matrices \(C_j\) for \(j = 1, \ldots, s\) are called generators of the point set \(P_n\), more precisely, the matrix \(C_j\) is the generator of the \(j\)-th coordinate of the point set.

Some examples of digital \((t, s)\)-sequences are the van der Corput, Sobol', Faure, Niederreiter, and Niederreiter-Xing sequences [Nie92, NX96].

While the construction using generator matrices already delivers a vast amount of \((t, m, s)\)-nets and \((t, s)\)-sequences, [GHSK07] propose using additional permutations, resulting in \((t, m, s)\)-nets featuring a larger maximum minimum distance. Note that the additional permutations of course result in a further increase of the total search space for the construction. In addition, for some of the new nets the permutation has to be stored explicitly, which is a disadvantage.

While \((t, s)\)-sequences will always feature a uniform distribution, a full stratification of the domain can only be guaranteed when using \(n = b^m\) points. This can even be a problem for similar constructions like Halton that must use large valued \(b\) for high-dimensional problems as it increases the number \(n\) of points needed for a full stratification (\(b^m\) points are needed to form a sub-net that uniformly stratifies the unit cube) and will also result in a worse distribution when picking only small sub-sets of the whole sequence, which is implicitly the case for all Markov chain simulations (see Section 3.4). So the problem of full stratification implicitly reintroduces the curse of dimension [KK02b].

**Efficient Implementation**

Computations in base 2 allow for an exact representation of the coordinates in the IEEE floating point standard, as long as the mantissa holds enough bits (23 in the case of using 32-bit floating point, 53 for double precision). Using 32-bit floats allows to distinguish only \(2^{32}\) samples, meaning that the same samples will be computed after less than 8.5 million samples.

Interpreting unsigned integers as bit vectors, i.e. elements of \(\mathbb{F}_2^{32}\) with \(\mathbb{F}_2 = \mathbb{Z}/2\mathbb{Z}\), allows one to use standard bitwise operations for vector operations on \(\mathbb{F}_2^{32}\). The matrix-vector multiplication \(C_j(d_b(i), \ldots, d_{m-1}(i))\), which has to be performed in \(\mathbb{F}_2\), then becomes very efficient by exploiting the fact that addition corresponds to XOR in \(\mathbb{F}_2\) (see Listing 3.1).

See [KK02b] for additional optimized implementation variants of the radical inverse, the second component of Sobol' points and Larcher-Pillichshammer points. See [GHSK07] for an optimized implementation using additional permutations. Note that even using 32-bit integers and mapping to double precision floats, the amount of possible samples may still be insufficient for large resolutions, thus industrial rendering implementations should rely on the full 53-bits of double precision.
3.3. Quasi-Monte Carlo Integration

```c
unsigned int x_i(unsigned int i) {
    unsigned int result = 0;
    for(unsigned int k = 0; i; i >>= 1, k++)
        if(i & 1) // only if corresponding bit set
            result ^= C_j[k]; // vector addition of (k+1)-th leftmost column of C_j
    return result;
}
```

Listing 3.1: Efficient implementation in base 2.

Performing the mapping to floating point representation in $[0,1)$ can be achieved by either dividing the integer by the largest representable number plus one $(\text{double})\text{result}/(\text{double})(1ULL << m)$ or by directly mapping the bit vector into the corresponding mantissa bits and correcting the range afterwards:

```c
double map(unsigned int result) {
    double tmp;
    uint Cast(tmp) = 0x3FF000000000000ULL | (((unsigned long long)result)<<20);
    return tmp-1.0;
}
```

Note that the latter variant allows for a simpler hardware implementation, but can be a bit slower on current x86 hardware when compared to just multiplying the result by $1.0/(1ULL << m)$.

While it is possible to further optimize the implementation by using a gray-code permutation of a sequence [PTVF92, BFN92], such that only one of the input bits $d_k(i)$ changes for neighboring indices, the user will be limited by the fixed order that requires to draw successive samples from the sequence. Additionally, the local distribution of subsets of the samples will be different, a fact that will become apparent in simulations, as will be explained in Section 3.4.2.

A better possibility is the parallel computation of bits using SIMD opcodes. By masking the XOR matrix operation with the bits of the index, the previously needed conditional branches can be avoided and it is possible to (theoretically) compute all bits in parallel as illustrated in Listing 3.2.

While drawing QMC samples usually has been recognized as computationally slow as compared to pseudo random number generation, this optimized implementation is on par with the recently presented and also SIMD based, refined variant of the Mersenne Twister [SM07].

3.3.3 Rank-1 Lattice Sequences

As compared to $(t,s)$-sequences and the Halton sequence, there exists an even simpler algorithm to generate highly uniform point sets:

$$x_i: \mathbb{N}_0 \rightarrow \mathbb{Q}^s \cap [0,1)^s$$

$$i \mapsto \phi_b(i) \cdot g \mod [0,1)^s$$

generates rank-1 lattice sequences from a suitable generator vector $g \in \mathbb{N}^s$. For $n=b^m$ this results in a full rank-1 lattice, since $(\phi_b(i))_{i=0}^{b^m-1} = (0, \ldots, \frac{b^m-1}{b^m})$. Other than Cartesian grids, which suffer from a bad discrepancy of $O\left(\frac{1}{n} \right)$, rank-1 lattice sequences of low discrepancy exist. Note that rank-1 lattices exist for any $n$ and $s$ [Nie92, SJ94].

One example for a rank-1 lattice $x_i := \frac{i}{n} \cdot g \mod [0,1)^s$ that features a simple construction for $s = 2$ is the Fibonacci lattice with generator vector

$$g = (1, F_{k-1})$$

at $n = F_k$, $k \geq 2$
\section*{Chapter 3. Quasi-Monte Carlo Simulation of Markov Chains}

\begin{verbatim}
__m128 LUT[16];

void precalcLookup() {
    for(unsigned int c = 0; c < 16; c++)
    {
        unsigned int i[4] = {-(c&1),-(c>>1)&1,-(c>>2)&1,-(c>>3)};
        LUT[c] = _mm_set_ps(floatCast(i[3]), floatCast(i[2]),
                            floatCast(i[1]), floatCast(i[0]));
    }
}

double x_iSSE(unsigned int i) {
    __m128 result = _mm_xor_ps(
        _mm_xor_ps(_mm_and_ps(LUT[i&15], matrix[0]),
                    _mm_and_ps(LUT[(i>>4)&15], matrix[1])),
        _mm_xor_ps(_mm_and_ps(LUT[(i>>8)&15], matrix[2]),
                    _mm_and_ps(LUT[(i>>12)&15], matrix[3]))),
        __m128 tmp = _mm_xor_ps(result, _mm_movehl_ps(result, result));
    unsigned int tmp2 = (((unsigned int*)&tmp)[0] ^ (((unsigned int*)&tmp)[1]));
    return (double)tmp2 * (1.0/(1ULL << 32));
}
\end{verbatim}

Listing 3.2: Fast generation of digital nets and sequences using SIMD opcodes. precalcLookup generates a bitmask (0 or 0xFFFFFFFF) for every bit of numbers 0, \ldots, 15 in a precomputation step. These are stored in pairs of 4 bitmasks each in the array LUT. x_iSSE then decomposes the incoming index i into 8 pairs of 4 bits each. By using the previously generated look-up table, the corresponding matrix entries are in fact only XORed onto the temporary result if the stored bitmask is set (0xFFFFFFFF). Using a hierarchically organized XORs for the partial results, the compiler can then decide the best execution order to get the final result. Finally, the single 32-bit integer is mapped to the range of [0, 1).

\begin{figure}[h]
\centering
\includegraphics[width=0.3\textwidth]{fibonacci_lattice.png}
\caption{The $n = 34$ points of the Fibonacci lattice $L_{34,21}$ with generator vector $\mathbf{g} = (1, 21)$.
\label{fig:fig323}}
\end{figure}

where the Fibonacci numbers are defined by

\[ F_k := F_{k-1} + F_{k-2} \text{ with } F_1 := F_2 := 1 \]

While the computation is exceptionally fast, the pattern and enumeration of rank-1 lattices is difficult to
3.3. Quasi-Monte Carlo Integration

Handle in the context of Markov chain simulations (see Section 3.4). The regularity of the samples will result in distinct artifact patterns which will only vanish for a large number of samples. Additionally, adaptive sampling techniques are even harder to include as lattices only feature a fixed number \( n \) of samples by default. Also the search for an optimal generator vector \( \mathbf{g} \) (in the sense of maximizing the minimum distance of the samples) must be done for every number of samples \( n \) and the dimension \( s \) of a problem. This is even more problematic for rank-1 lattice sequences as the choice of the right \( \phi_\mathbf{g} \) is an additional degree of freedom.

Another open problem is how to achieve properties similar to \((t, s)\)-sequences, in the sense that small subsets are again lattices featuring high uniformity.

3.3.4 Deterministic Error Bounds

Using the deterministic low discrepancy point sets of the previous Section 3.2 allows to proof error bounds for quasi-Monte Carlo integration. Unlike the Monte Carlo case which features a probabilistic bound, the quasi-Monte Carlo point sets allow to formulate deterministic upper bounds. These error bounds obviously depend on the integrand function classes. From all possibilities, only those are mentioned that are important in the computer graphics setting.

**Error Bound using Maximum Minimum Distance**

Given the radius \( r(n, g) \) of a rank-1 lattice (the smallest circumcircle of the fundamental Voronoi cell), \( f \) being a Lipschitz function periodic on \([0, 1]^s\) with Lipschitz constant \( L \) and \( P_n = \{x_0, \ldots, x_{n-1}\} \) a rank-1 lattice, the integration error

\[
\left| \int_{[0,1]^s} f(x) dx - \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) \right| \leq L \cdot r(n, g)
\]

is bounded by the Lipschitz constant multiplied by the maximum minimum distance. As a result, the quality of the integration will depend on the well behavior of the function (expressed by the Lipschitz constant), and the uniform distribution of the samples.

It is interesting to note that comparing constructions like Hammersley, Sobol', and Larcher Pillichshammer, a smaller discrepancy often also features a larger maximum minimum distance. The same can be spotted comparing Halton against the Faure permutated Halton sequence [Kel06]. Also see [GHSK07] to find an application for \((t, m, s)\)-Nets.

**Error Bound using Spectral Properties**

One drawback of using the maximum minimum distance error bound is its implicit curse of dimension, which is intrinsic to the radius \( r(n, g) \). Actually, a much better error bound can be formulated if the function class is further restricted by demanding a certain decay of the Fourier coefficients of the integrand [SJ94].

**Error Bound using Discrepancy**

The Koksma-Hlawka-inequality

\[
\left| \int_{[0,1]^s} f(x) dx - \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) \right| \leq V(f)D^*(P_n)
\]

bounds the integration error by a product of the variation \( V(f) \) of the integrand in the sense of Hardy and Krause [Nie92] and the star discrepancy of the samples. Note that the function class of functions of bounded variation is much smaller than \( L^2 \).

The star discrepancy is an anisotropic error bound as it is based on axis aligned boxes anchored in the origin. Rotating the point set changes discrepancy and even shifting the point set on the torus (see Section 3.3.5) can change discrepancy.
Rotating a function instead illustrates the potentially catastrophic behavior of the error bound: The variation of
\[ f(x, y) = \begin{cases} 1 & y > x \\ 0 & \text{else} \end{cases} \quad (3.2) \]
is infinite, however, using the Hammersley point set (see Section 3.3.2) it can be shown that the error is in fact between \( \frac{1}{2\pi n} \) (using an even amount of samples) and \( \frac{1}{\sqrt{n}} \) (for an uneven number).

Note that discontinuities that are not aligned to the canonical axes are quite common in the real world.

**Error Bound using \((\mathcal{M}, \mu)\)-Uniformity**

Let \((X, \mathcal{B}, \mu)\) be an arbitrary probability space (see Section 3.2.3) and let \(\mathcal{M} = \{M_1, \ldots, M_k\}\) be a partition of \(X\) with \(M_j \in \mathcal{B}\) for \(1 \leq j \leq k\). Then for any \((\mathcal{M}, \mu)\)-uniform point set \(P = \{x_0, \ldots, x_{n-1}\}\) and any bounded \(\mu\)-integrable function \(f\) on \(X\) we have
\[
\left| \int_X f(x) \, d\mu(x) - \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) \right| \leq \sum_{j=1}^{k} \mu(M_j) \left( \sup_{x \in M_j} f(x) - \inf_{x \in M_j} f(x) \right).
\]

Note that this works for any function \(f \in L^2([0, 1]^s)\), but unlike other QMC error bounds, does not feature a distinct separation of function and sample properties.

With \((X, \mathcal{B}, \mu) = ([0, 1]^s, \mathcal{B}, \lambda_s)\) the theorem is applicable in the setting of computer graphics. For the example function in Equation 3.2 the deterministic error \(O(n^{-2})\) bound can be obtained by selecting an \((\mathcal{M}, \mu)\)-uniform point set with \(k = n\). The difference of the supremum and the infimum can only be one in the \(O(n^2)\) sets of the partition, which are crossed by the discontinuity otherwise it must be zero. With \(\mu(M_j) = \frac{1}{n}\) the bound is obtained. This argument is similar to the argument used in the Numerical Recipes book [PTVF92], but, what is more important, does not use probabilistic arguments for a deterministic algorithm. However, note that this only holds for two dimensions.

### 3.3.5 Randomized Quasi-Monte Carlo Integration

As was seen in the previous sections, QMC patterns suffer from a rather pessimistic or difficult error bound analysis. To be able to use the original MC probabilistic error bound, it is possible to randomize any deterministic point set and obtain an unbiased estimator. In the following, different practical randomizations will be presented that can also be used in the context of padded replications sampling [KK02b] (also see Section 3.4.2). Using low discrepancy point sets with these randomizations results in a variance reduction technique for Monte Carlo integration. Note that the randomizations are random bijections.

**Cranley Patterson Rotation**

A rather trivial technique is to randomly shift [CP76] deterministic points
\[
x_i \mapsto x_i + \xi \mod 1,
\]
where \(\xi = (\xi^{(1)}, \ldots, \xi^{(s)}) \in [0, 1]^s\) is a vector of \(s\) independent realizations of uniform random numbers on the unit interval.

While the shift can change the discrepancy measure, it leaves the maximum minimum distance unaffected.

**Scrambling**

The earlier presented, deterministic Faure scrambling is in fact only one subset of the more general (but also more computationally intense) random scrambling [Owe95] that can be used to randomize \((t, m, s)\)-nets and \((t, s)\)-sequences. The algorithm first subdivides the domain \(P\) on one axis into \(b\) equally sized volumes \(H_1, H_2, \ldots, H_b\). Then these volumes are randomly permuted and the process continues recursively with the \(H_b\) and on each axis. As the computation on finite precision numbers terminates the algorithm automatically.
3.4 Simultaneous Simulation of Markov Chains

The scrambling becomes practical for all digital constructions (see Section 3.3.2). Note that scrambling does not affect the quality \( t \) parameter.

A simplified but sufficient randomization for digital constructions in base 2 can be found in [FK02] and only needs to use one additional XOR operation on the integer bit-vector to achieve a scrambling of base 2 sequences.

3.4 Simultaneous Simulation of Markov Chains

A Markov chain describes a memoryless stochastic process, where the transition probabilities do not depend on the history of the process. For example [Sha48] modeled the English language by a Markov chain, where the author computed the relative frequencies of one word following another from an English book. Using these transition probabilities it is possible to generate seemingly English sentences. Many more evolution processes, like e.g. the Brownian motion or particle trajectories, can be described as Markov chains.

Properties of processes can be estimated by averaging the contributions of multiple Markov chains. Instead of simulating each trajectory independently, it has been found that simultaneously simulating Markov chains using correlated samples and sorting the ensemble of states after each transition step can notably improve convergence.

In a light transport setting, the underlying integral equation can be reformulated as a path integral [Vea97] (also see Section 1.1.1). Sampling path space (see Figure 1.3) corresponds to simulating Markov chains, where the paths are established by ray tracing and scattering. The initial distribution is determined by the emission characteristics of the light sources and the transition probabilities are given by bidirectional reflectance distribution functions (see Section 1.1.2) on the surface.

Markov chain simulation using deterministic samples was indistinguishable from using Monte Carlo methods in a practical context as the resulting convergence rates were similar or even worse as soon as the dimension of the problem setting rises. The usage of quasi-Monte Carlo methods in Markov chain simulation could not be successfully applied as a result.

The idea of improving the simultaneous simulation of Markov chains with quasi-Monte Carlo methods by an intermediate sorting step was originally introduced by Lécot in a series of papers dealing with the Boltzmann equation [Léc89a, Léc89b, Léc91, LC98] and later on the heat equation [KL99]. This idea was then used and refined to solve the heat equation on a grid by Morokoff and Caflisch [MC93] and recently extended by L’Écuyer, Turin, Demers et al. [LL02, LT04a, LT04b, DLT05, LTT05, LDT06, LTT06, HLL07, LDT07] to incorporate randomized versions of the algorithm and splitting for rare event simulation. Independent research, but in a way related to the above approaches, was conducted by [BNN+98] in the field of computer graphics.

In the following the deterministic version of the scheme will be analyzed and simplified. For the derivation the algorithm from [LT04b] is used. The results give practical insight, when and why the scheme is superior to approaches without sorting and how to implement it.

3.4.1 Analysis of a Deterministic Algorithm in One Dimension

The algorithm presented in [LT04b] simultaneously simulates Markov chains on a discrete state space \( E \) with an initial distribution \( \mu := (\mu_i)_{i \in E} \) and a transition matrix \( P := (p_{ij})_{i,j \in E} \). Using a \( (t, 2) \)-sequence \( (x_i)_{i \in \mathbb{N}} \) in base \( b \) [Nie92], \( N = b^m \) chains are simulated in parallel, where \( \mathbf{X}_{n,i} \) is the state of chain \( i \) at time step \( n \). Furthermore, the algorithm requires that for \( m > t \) the \( N = b^m \) subsequent components \( x_{n+1} \) form a \((0, m, 1)\)-net in base \( b \). As shown in [LT04b] the algorithm converges if

\[
\forall k \in E : \left| \sum_{n=1}^{k-1} p_{n+1,n} - \sum_{n=1}^{k-1} p_{n,k} \right| \leq 1 \tag{3.3}
\]

holds.
Simplification of the Algorithm

Now the Sobol' sequence \( x_t = (x_{t,1}, x_{t,2}) = (\Phi_2(t), \Phi_3(t)) \in [0, 1]^2 \) will be considered, which is a \((0, 2)\)-sequence in base \( b = 2 \) and fulfills the assumptions required for the convergence condition to hold. For the definition of the original Sobol' sequence see [Sob67], while a simple code example for \((\Phi_2(t), \Phi_3(t))\) is found in [KK02b].

The simulation itself starts at time step \( n = 0 \), initializing state \( X_{0,[N_{N_0}]} \) for \( 0 \leq t < N \) using \( x_{t,2} \) for the realization of \( \mu \). The algorithm then continues by sorting the states (this will be discussed in detail in Section 3.4.3) and continues the chains by computing \( X_{n,[N_{N_0}+n_{N_0}]} \) using \( x_{l+n,N_0} \) for \( 0 \leq t < N \) to realize transitions according to \( P \). The index

\[
\sigma(l) := [N \cdot x_{(i+n\cdot N_0)}]
\]

for selecting the next state for transition in fact uses the van der Corput sequence \( \Phi_2 \) in base 2, which is a \((0, 1)\)-sequence and thus a sequence of \((0, m, 1)\)-nets (see Section 3.3.2). For example choosing \( m = 3 > t = 0 \) results in \( N = 2^3 = 8 \) and

\[
([8 \cdot \Phi_2(l + n \cdot 8)])_{l=0} = \{0, 4, 2, 6, 1, 5, 3, 7\}.
\]

for \( n \in N_0 \). Hence all indices used during the different timesteps \( n \) are in fact identical for all \( m \).

Assuming uniform probabilities \( p_{ij} = \frac{1}{|P|} \) the convergence theorem still applies, but more important, stable sorting does not change the state order. It thus follows that in fact the index permutation can be chosen as identity without touching the convergence conditions. The same applies for selecting the initial states \( X_{0,i} \) and it results the simplified, but equivalent algorithm

- \( n := 0 \)
- initialize \( X_{0,j} \) using \( x_{t,2} \) for \( 0 \leq t < 2^m \)
- loop
  - sort state vector using a suitable order
  - \( n := n + 1 \)
  - continue chain by computing \( X_{n,i} \) using \( \Phi_3(l + n \cdot 2^m) \) for \( 0 \leq l < 2^m \)

using only the second component of the \((0, 2)\)-sequence \( x_i \).

When and Why it Works

The improved convergence of the scheme, which has been observed in many applications (see the references at the beginning of Section 3.4) must be caused by the structure of the samples \( \Phi_2(l + n \cdot 2^m) \) used to realize the transitions of \( X_{n,i} \) according to \( P \). This can be understood by decomposing the radical inverse (see also [Kel06])

\[
\Phi_3(l + n \cdot 2^m) = \Phi_3(l) + \frac{1}{2^m} \Phi_3(n).
\]

which reveals an implicit stratification: \( \Phi_3(l) \) is an offset with spacing \( \frac{1}{2^m} \) depending on the state number \( l \), while the shift \( \Phi_3(n) \) is identical for all the interval at timestep \( n \).

Here the low dimensional setting allows for a misleading interpretation of the samples being a shifted lattice or stratified samples, as the entirety of the \( \Phi_3(l) \) for \( l = 0, \ldots, 2^m - 1 \) in fact must be an \((0, m, 1)\)-net and thus an equidistant set of samples.

However, the good performance stems from the property that \( \Phi_3(l) \) is a \((t, s)\)-sequence and thus a sequence of \((t, m', s)\)-nets for any \( m' \) with \( t \leq m' \leq m \). This means that \( b^m \) states, that are similar in
state space and, therefore, subsequent by order after sorting, will sample their transition by a \((t, m', s')\)-net, which guarantees for good discrete density approximation. The maximum improvement would be obtained if all \(2^m\) chains were in the same state (see Figure 3.3). The more the states of the chains are separated in state space, the smaller the performance improvements will be.

![Figure 3.3](image)

Figure 3.3: Uniform sampling of the hemisphere originating in a single point, adding six successive samples at a time. Note the varying local uniform distributions when comparing random sampling (top row) versus the Halton sequence (bottom row).

### 3.4.2 New, Simplified Algorithm in \(s\) Dimensions

Using a \((t, s)\)-sequence in base \(b\), which is a sequence of \((t, m, s)\)-nets, the scheme also works in \(s\) dimensions: Markov chains, whose states are similar after sorting are guaranteed to sample the transition probability by low discrepancy samples. The simplified algorithm in \(s\) dimensions now looks like:

- \(n := 0\)
- initialize \(X_{0,1}\) using quasi-Monte Carlo points \(x_i\)
- loop
  - sort state vector using a suitable order
  - \(n := n + 1\)
  - continue chain by computing \(X_{n,j}\) using subsequent samples \(x_i\) from a \((t, s)\)-sequence

Some simulations require trajectory splitting in order to capture certain local subtle effects. While this has already been addressed in [DLT05, LDT06, LDT07], it in fact can be achieved in a simpler way by just drawing more samples out of the \((t, s)\)-sequence for states to be split.

This is a consequence of the fact that it is not even necessary to simultaneously simulate exactly \(b^m\) chains. It is only important to draw subsequent samples from the \((t, s)\)-sequence and to minimize the number \(b^m\) of points in the subsequent \((t, m, s)\)-nets in order to enable the maximal performance gain. The choice of the \((t, s)\)-sequence, however, is restricted by the condition that \((0, s)\)-sequences only exist for \(b \geq s\) and that \(m > t\) [Nie92]. Note that other radical inversion based points sets like the Halton sequence or its scrambled variants fulfill properties similar to \((t, s)\)-sequences [Mat98] and will result in similar performance gains (see Sections 3.3.2 and 3.3.2).

**Randomization**

While there exists no general proof for convergence of the deterministic algorithm in higher dimensions yet, the algorithm becomes unbiased by freshly randomizing the quasi-Monte Carlo points in each time step \(n\).
Since this is in fact an instance of padded replications sampling as introduced in [KK02a, KK02b] the argument for unbiasedness becomes simpler than in [LLT06]. Randomization, however, deserves special attention.

The most efficient implementation along the lines of [KK02b] consists of choosing a \((t, s)\)-sequence in base \(b = 2\), from which subsequent samples are drawn, which are XOR-ed by an \(s\)-dimensional random vector. This random vector is freshly drawn after each transition step. However, as random scrambling changes the order in which the points are enumerated, the local properties of the sequences of \((t, m, s)\)-nets are changed, too.

This observation can be taken as an explanation for some of the effects seen in [LLT05]: Sobol’ and Korobov points used in the array-(R)QMC simulation are worse up to an order of magnitude in variance reduction than their transformed (Gray-code for Sobol’, Baker transform for Korobov) counterparts. This can be explained by the structure of the points. The sequence of \((t, m, s)\)-nets extracted from the Sobol’ sequence is locally worse than its Gray-code variant. The same goes for the Korobov lattice and its transformed variant.

While this observation did not hold for the shown experiments (comparing the standard Halton sequence and its Faure scrambled variant in Figure 3.13), this may be due to the effect only becoming apparent for a really large amount of simultaneously simulated chains (also see [LLT05], where the variance reduction increases superproportionally for the scrambled variants). As the average distance of neighboring states will be decreased for an increasing number of chains, the local uniform distribution of the used QMC samples is becoming more and more important as the problem will locally be reduced to uniformly sampling a hemisphere (see Figures 3.5 and 3.3). Figure 3.4 illustrates the change in local distribution for a subset of the Halton sequence and scrambled variants. Note that the maximum minimum distance of successive samples is increased when using Faure scrambling, while the standard Halton sequence features the worst possible distribution for successive samples. In addition, maximizing the minimum distance of successive samples over all possible subsets, the best local uniformity should be received.

![Figure 3.4: Comparing successive samples drawn from the Halton sequences in base 7 and 31 and its Faure (see Section 3.3.1) and randomly scrambled variants. Note that the maximum minimum distance of successive samples is increased for the Faure scrambled variant.](image)

Using scrambling techniques might be avoidable by only using sequences and point sets based on small \(b\). As for \((t, s)\) sequences, the definition of \((t, m, s)\) nets already expresses that it is advantageous to use a small base \(b\) as it will result in a smaller \(t\) parameter for a decreasing amount of total points \(b^m\), thus resulting in a full stratification for successive samples of a \((t, s)\) sequence down to a level of ideally only \(b\) points in one subset. However, scrambling will still help to decorrelate low dimensional projections of high
3.4. Simultaneous Simulation of Markov Chains

dimensional point sets (see Section 3.3.1).

This finding additionally allows to optimize the implementation of the QMC samplers, as concentrating on base 2 sequences such as Sobol' will result in very efficient code (see Section 3.3.2).

![Figure 3.5: Two neighboring states and uniformly sampling their corresponding hemispheres (e.g. when using splitting techniques). As the distance of the states decreases, the uniform distribution of the combined sample sets should still be as uniform as possible. Note that the illustration can only show the projected samples, as plotting the 3-dimensional ray directions (as in Figure 3.6) does not allow to spot uniform distributions as easily.](image1)

![Figure 3.6: Using sorted QMC for higher dimensions of the Markov chain simulation will result in the same effect as the illustrated sampling of the initial 4-dimensional particle distribution of a scene light source by QMC instead of MC: Samples will be distributed more uniformly, resulting in better stratification properties.](image2)

3.4.3 Sorting Strategies

In order to have the states as closely together as possible, they have to be enumerated in an order such that the sum of the distances of neighboring states is minimal. This in fact relates to the traveling salesman problem\(^1\), where for a given set of cities and the costs of traveling from one city to another city, the cheapest round trip is sought that visits each city exactly once and then returns to the starting city.

Our problem is very similar except that it is not necessary to return from the last state of the route to the first state. Some techniques are already available to efficiently calculate approximate, but close to optimal, solutions for the traveling salesman problem [DMC91, Rei94]. However, runtimes of these algorithms are not acceptable in MC and QMC simulations, as the calculation of the distance matrix alone exhibits an $O(N^2)$ complexity, while the algorithm should be kept as close as possible to the $O(N)$ complexity of classic Monte Carlo methods.

In the following some possible orders to achieve fast sorting for high-dimensional state spaces are discussed.

---

\(^1\)This problem already was investigated by Euler in the early 18th century and unfortunately can be shown to be NP-hard [Kar72]. For a historical survey on the problem see Lawler et al. [LLKS85].
Norm of State

The average complexity of quicksort is $O(N \log N)$, but for certain scenarios even $O(N)$ algorithms exist, like e.g. radixsort, which, however, requires additional temporary memory. In order to use one-dimensional sorting algorithms, the multi-dimensional state must be reduced to one dimension. Amongst many choices often some norm $\|X_n\|$ is used to define an order on the state space. However, similar norms do not necessarily indicate proximity in state space. A simple example for this is similar energy of particles in a transport simulation that are located far away in space (as it is the case in symmetrical scenes like Shirley’s “Scene 6”, see Figure 4.6).

Spatial Hierarchy

Figure 3.7: Sorting the states into the leafs of a spatial hierarchy defines an order of proximity by traversing the hierarchy in in-order.

A second possibility to enable multidimensional sorting is to use a spatial hierarchy to define an order on the states [Wie03]. Efficient data structures for this purpose are the BSP-tree [SBGS69, Abri95], its specialized axis aligned subset, the $kd$-tree [Ben75], or bounding volume hierarchies [RW80, KK86, WK06]. The construction of such binary hierarchies is simple: The space is recursively subdivided using planes selected by some heuristic (see Section 2.4). The construction runs in $O(N \log N)$ on the average. Traversing the hierarchy in in-order enumerates the leaves in an order of proximity. This traversal even becomes trivial, if the tree is left-balanced and in consequence can be stored in an array.

If a spatial hierarchy must be used anyway, for example to accelerate ray tracing (as it is the case in this work’s implementation), there is no additional construction time for the hierarchy. The particles are then stored as linked lists attached to the leaves of the hierarchy (see Figure 3.7). Unfortunately the quality of this order is strongly determined by the quality of the spatial hierarchy used for simulation, which is especially problematic if the number of leafs in the hierarchy is much smaller than the number of chains $N$ as this results in several states being mapped to the same leaf.

Still, experiments using an already present ray tracing hierarchy ($kd$-tree or BIH) demonstrated that the results can be comparable to using an additional bucket sort (see Section 3.4.3), as soon as the leafs of the hierarchy are small enough to only keep an average of 10 to 20 or less particles at a time. While this is obviously only the case for detailed geometry, it will allow for a complete merge of the usually separated ray tracing and sampling step at almost no additional costs.

Bucket Sorting and Space Filling Curves

In order to guarantee linear time complexity, bucket sorting can be used. In the $s$-dimensional extension [HLL07] of the simple algorithm sketched in Section 3.4.1, multidimensional states were sorted into buckets by the first dimension of the state, then the states of each bucket were sorted into buckets according to the second dimension, and so forth. This procedure works well, but has the problem that states close in state space can still be separated by the sorting procedure. In addition, a stratification of each dimension has to be used, which induces the curse of dimension in the number of Markov chains to be simulated simultaneously.

Therefore the state space is divided into equal voxels, which serve as buckets. The bucket of each state is found by truncating the state coordinates according to the resolution of the voxel grid. Note that this
applies for continuous as well as discrete state spaces. Enumerating the voxels by proximity yields the desired order on the states and can be done in linear time in the number of voxels.

Orders that enumerate the voxels by proximity are given by space filling curves \[\text{[Sag94]}\] like e.g. the Peano curve \[\text{[Mor66]}\], Hilbert curve, or H-indexing. These curves guarantee every voxel to be visited exactly once along with an overall path length that is relatively short. For problems with large geometry, which is the case in industrial simulations, this can be even one of the few possibilities to generate fast and memory efficient approximate solutions to the traveling salesman problem \[\text{[Rei94]}\]. However, these curves are rather costly to evaluate, need to be tabulated to be efficient, or are not available for higher dimensions.

Fortunately, the Z-curve, also known as Lebesgue-curve or Morton order, avoids these problems. Given integer coordinates of a bucket in multidimensional space, its one dimensional Z-curve index is simply calculated by bitwise interleaving the coordinate values (see Figure 3.8). This is very easy and fast to compute for any dimension and problem size, and requires no additional memory. Unfortunately, the results are not as good as for example the Hilbert-curve in a global context. However, the average case partitioning quality and average/worst case logarithmic index ranges are comparably good \[\text{[Wie03]}\]. Problems can arise in highly symmetrical scenes like Shirley’s “Scene 6” (see Figure 4.6) used for our numerical experiments: States on the walls parallel to the \((x, y)\)-plane will be sorted very well, but the states located on the other two walls parallel to the \((y, z)\)-plane will be visited by the curve in an alternating manner, which can lead to correlation artifacts in some scenarios.

![Figure 3.8: The Z-curve in two dimensions for \(2 \times 2\), \(4 \times 4\), and \(16 \times 16\) buckets. With the origin \((0, 0)\) top left the point marked by \(\times\) has the integer coordinates \((3, 4)\), which corresponds to \((011, 100)\)\(_2\) in the binary system. Its binary Z-curve index \(100101\)\(_2\) is computed by bitwise interleaving the binary coordinates.](image)

While the Z-curve offers an elegant and very efficient solution to enumerating the buckets, special care has to be taken when using the sorted array along with a QMC sampling technique. As the Z-curve is based on base 2 calculations, any QMC sequence that is working in the same base will result in correlations between neighboring voxels. Using real life scenarios though, this correlation will not be as distinct anymore as it is very unlikely that each and every voxel will feature exactly one sample. Still, avoiding the same base can be achieved trivially either using a different base for QMC sampling or defining a curve similar to the Morton ordering, but in base \(b > 2\).

### On-the-Fly Sorting and Sampling

As the additional sorting step requires the temporary storage of states and thus introduces an implicit delay of a recursive Markov chain implementation (which becomes an issue when evaluating user-written recursive shaders in a rendering system), an on-the-fly solution might be preferable. Given the position of a state, it is possible to determine its Z-curve index in a fixed size regular grid by using an inverse mapping from the \(s\)-dimensional coordinates to the 1-dimensional index.

By taking this index as an estimate for the position in a sorted array, the index can directly be used to draw a sample out of a QMC sequence. Unfortunately, this technique is not even consistent as it cannot be guaranteed that all samples of the QMC sequence will be used over time due to the probability of states' positions in space (which can simply be 0 for some voxels). Additionally, using voxel sizes that will not match the distribution of the states will result in strong correlation artifacts (see Figure 3.9). While it is possible to tweak the implementation, e.g. by using an additional memory block to store counters for each voxel so
each state is guaranteed to receive a unique sample, it still cannot be implemented to result in consistent behavior.

Figure 3.9: Choosing voxel cells too large results in correlation artifacts in this simple Cornell Box variant, since neighboring states reuse the same sample-points of the chosen sequence over and over again.

However, the presented algorithm can be used in simulations where each cell is guaranteed to feature at least one particle. It may also be possible to achieve an on-the-fly simulation in the context of merging the particle transport itself with the sampling, as this would allow for a direct estimation of local state distributions and remove the need for an explicit sorting step.

### 3.4.4 Application to Light Transport Simulation

For numerical evidence, the algorithm developed in the previous sections is applied to light transport simulation for synthesizing realistic images. To solve the path integral, one can think of two basic strategies, which are either using high dimensional low discrepancy points or padding low dimensional low discrepancy points [KK02b]. The latter approach fits our findings in Section 3.4.2, where high dimensional events are composed as subsequent transitions of a Markov chain. As measured in [KK02a] the difference to using high dimensional sequences for Markov chain simulations is next to none, but using padded sequences is computationally faster and requires less implementation effort (see Section 3.3.2). It is also simpler for practitioners in the rendering industry.

In addition the low dimensional approach allows for much better results, because the stratification properties of \((t,s)\)-sequences or the Halton sequence and its scrambled variants are much better for small dimensions (see Section 3.4.2).

Figure 3.10: Photon trajectories are started from the light sources. Upon hitting a surface after tracing a ray, the bidirectional reflectance distribution function is sampled to determine a direction of scattering to continue the path.
Fredholm or Volterra?

The integral equation underlying light transport can be considered either as a Fredholm or Volterra integral equation, which matters for implementation.

\[
L_r(x, \omega) = \int_{S^2(x)} f_r(\omega, x, \omega') L(x, \omega') (n(x) \cdot \omega') d\sigma(\omega')
\]

is the radiance reflected off a surface in point \(x\) in direction \(\omega\), where the domain \(S^2(x)\) of the integral operator is the hemisphere aligned to the normal \(n(x)\) in \(x\) (see the illustration in Figures 3.10 and 3.11). \(f_r\) is the bidirectional reflectance distribution function describing the optical surface properties and \(L\) is the incident radiance. Using this integral operator results in a Volterra integral equation of the second kind, as the integration domain depends on \(x\).

\[
L_r(x, \omega) = \int_{S^2} f_r(\omega, x, \omega') L(x, \omega') \max\{-n(x) \cdot \omega', 0\} d\sigma(\omega')
\]

on the other hand results in a Fredholm integral equation of the second kind, as we are integrating over all directions of the unit sphere \(S^2\) independent of \(x\).

Using the approach of generating global directions [SKFNC97] and rejecting directions with negative scalar product with respect to the surface normal \(n(x)\) is computationally attractive (for low dimensional point sets or MC sampling), while the first approach requires to generate directions in the hemisphere that have to be transformed into the local surface frame, which is more expensive.Mappings from the unit square to \(S^2\) or \(S^2(x)\) are found in [SC94, Rus98, Shi00, GHB + 05] with some positive and negative aspects illustrated in Figure 3.11.

![Diagram](image)

Figure 3.11: One problem of considering light transport as a Volterra integral equation: Spatially neighboring states that are located on surfaces with different orientations will demand an additional transformation from the unit coordinate system to the local coordinate system, thus potentially destroying uniformity properties induced by the additional sorting step. Two possible transformations are: a) Generating samples on the unit sphere and mirroring samples that are located beneath the surface (which is computationally fast, but loses some of the uniformity properties). b) Generating samples directly on the unit hemisphere and a transformation to the local coordinate system (which is computationally slower and loses all of the global uniformity properties). c) considers light transport as a Fredholm integral equation and uses rejection sampling. This was found to offer the best trade-off between efficient implementation and keeping full global uniformity of the samples.

An even more important argument for generating global directions is related to our algorithmic approach (see Section 3.4.2): By sorting it can happen that two close by surface points with different surface normals (e.g. in a corner) use subsequent samples of a \((t, s)\)-sequence to generate scattering directions. Generating global directions now works fine, whereas generating directions in the two different local frames using subsequent samples will destroy the low discrepancy properties. These discontinuities become clearly visible. Using global directions, however, does not allow for importance sampling according to \(f_r\) or the cosine term, which often is a major disadvantage and deserves further investigation.

Note that it may also help to incorporate surface properties into the sorting process, so states on differently oriented surfaces (but close to each other) will automatically be separated. This can be achieved by including the surface normal into the sorting step, increasing the sorted dimensions to six.
Numerical Evidence

Following the arguments in Section 3.4.2, we use the Halton sequence with permutations by Faure \cite{Kel06} randomized by a Cranley-Patterson-rotation \cite{CP76} in order to have unbiased error estimates. For the sorting step, the Z-curve order (see Section 3.4.3) worked best in our setting and was used for the following experiments. Note that it was numerically verified that omitting the randomization has no notable effect on the precision of the results.

In our numerical experiments we compared four approaches to simulate Markov chains:

\textbf{MC:} Uniform random numbers generated by the Mersenne Twister \cite{SM07} were used for classical Monte Carlo sampling.

\textbf{RQMC:} Used the high-dimensional Halton sequence with permutations by Faure randomized by a Cranley-Patterson rotation, where pairs of components were used to sample the two dimensional emission and scattering events.

\textbf{lo-dim RQMC:} Used the two-dimensional Halton sequence randomized by a Cranley-Patterson rotation. The Z-curve was used to enumerate the bucket-sorted states.

\textbf{hi-dim RQMC:} Used the high-dimensional Halton sequence with permutations by Faure randomized by a Cranley-Patterson rotation. The Z-curve was used to enumerate the bucket-sorted states.

In a first experiment the robust global illumination algorithm \cite{KK04, Kol04a} was used to compute the path integrals. The resulting graphs are depicted in Figure 3.12 and display the RMS error to a master solution and the variance averaged over the whole image as well as the pixel-based variance. The numbers were obtained by averaging 10 independent runs for a varying number of Markov chains. The measured numbers only convince for the simple test scene. In the complicated cases even no performance gain over Monte Carlo sampling can be measured, because the number of independent runs is too small and more experiments were not possible due to excessive runtimes. However, the visual error tells a dramatically different story as can be seen in Figure 3.14, where a clear superiority of the new algorithm in even very difficult settings becomes obvious. This case is not an exception, but can be observed for many test cases. It only emphasizes that standard error measures are not appropriate error measures for visual quality, which is a known but unsolved problem in computer graphics.

Figure 3.13 shows measurements for a very difficult light transport problem, where we directly traced photons from the light sources and connected their final path segment to the camera (one technique of bidirectional path tracing \cite{Vea97}). Opposite to the above measurements only one number of simultaneously simulated Markov chains is considered. Now a sufficient amount of experiments was computationally feasible and the superiority of the new algorithm became clearly visible.

Visual Evidence

Aside of the visual improvements in Figure 3.14, the sorting scheme was tested in the context of car visualization (see Section 4.5). Some visual results are presented in Figures 3.15 and 3.16.

3.5 Summary

The algorithms to simultaneously simulate Markov chains were successfully simplified and intuition was provided when and why sorting the states can improve convergence, i.e. reduce the visible noise in images. In addition the algorithm is no longer bounded by the curse of dimension. The restriction to homogenous Markov chains is easily removed by using transition probabilities $P \equiv P_t$ that are time dependent.

As global directions are used, i.e. an integration over product of spheres, it is also interesting to establish connections to recent research in that direction \cite{KS05}. However, an intrinsic problem of that approach is that importance sampling of BRDFs cannot be incorporated.

The numerical experiments also revealed that not all $(t,s)$-sequences and radical inversion based point sequences are equally good. This deserves further characterization and certainly is related to scrambling
3.5. Summary

a) Test scene "Shirley’s Scene 6".

b) Test scene "Invisible date".

Figure 3.12: RMS error, global, and per pixel variance for an a) simple and b) more complicated light transport setting, considering only indirect illumination.
Chapter 3. Quasi-Monte Carlo Simulation of Markov Chains

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Figure 3.13: Schematic view of the labyrinth test scene, where floor and roof have been removed for illustration. The camera is situated in the room at the bottom, while a light source is located on the other end of the connecting tunnel. The graphs show the Box-and-Whisker plots and the average amount (marked by ×) of the total radiance received by the camera for 1048576 simulated light paths for 50 independent realizations using each technique. The lower graph enlarges the interesting part of the upper graph. The table finally displays the deviation from a master solution using the $L_1$- and $L_2$-norm (variance) for the total radiance received by the camera and received by each pixel (256 × 256 pixels resolution) averaged over the 50 independent realizations. In this difficult setting the new method (RQMCS (low dimensional Halton, randomized by Cranley Patterson rotation) and RQMCFS (high dimensional, Faure scrambled Halton, randomized by Cranley Patterson rotation)) is clearly superior. Note that RQMCFS, due to single but distinct outliers in one realization, features a slightly increased per pixel variance.

<table>
<thead>
<tr>
<th>Method</th>
<th>$L_1$ per camera</th>
<th>$L_2$ per camera</th>
<th>$L_1$ per pixel</th>
<th>$L_2$ per pixel</th>
</tr>
</thead>
<tbody>
<tr>
<td>Monte Carlo</td>
<td>0.0026524</td>
<td>0.02368870</td>
<td>0.202377</td>
<td>463.397</td>
</tr>
<tr>
<td>RQMC</td>
<td>0.0407531</td>
<td>0.00773514</td>
<td>0.162835</td>
<td>62.0994</td>
</tr>
<tr>
<td>RQMCS</td>
<td>0.0247592</td>
<td>0.00178907</td>
<td>0.138360</td>
<td>9.79994</td>
</tr>
<tr>
<td>RQMCFS</td>
<td>0.0283253</td>
<td>0.00180906</td>
<td>0.152618</td>
<td>32.4620</td>
</tr>
</tbody>
</table>
3.5. Summary

Figure 3.14: Visual comparison for the test scene "Invisible Date" using 300 chains for simulation. The only lightsource is not visible as it is placed on the ceiling of the neighboring room. Due to the better distribution of the photons randomized quasi-Monte Carlo (RQMC) outperforms Monte Carlo (MC) visually, as can be seen by the reduced shadow artifacts. RQMC with sorting (RQMCS, using $256^3$ voxels for the bucket sort) is even superior as more photons made it into the second room and even the back of the door is lit very well.

permutations (see Equation 3.1 and Section 3.3.5). Related to this, the algorithm would be even simpler, if rank-1 lattice sequences could be applied. The constructions so far, however, lack the properties of $(r,s)$-sequences that are required for the improved performance.
Figure 3.15: Results of using sorted QMC Markov chain simulation in car visualization (while using only a very moderate number of overall samples): The top row uses standard QMC techniques, while the lower row uses the proposed additional sorting step. Left: Even in the full unfiltered view, outliers are more distinct using standard QMC. Middle: By enhancing the low frequency noise, QMC can be found slightly superior over sorted QMC, as the sorting increases uniformity of the samples, which directly results in neighboring samples that will sample different parts of the hemisphere. Right: While this is "unfortunate" for the low frequencies, enhancing the high frequency noise reveals the superiority of the new scheme as the overall estimate of the pixel integral will be of higher quality due to the increased uniformity of the samples. So the relative frequency of resulting outliers is decreased.
Figure 3.16: Increasing the overall number of samples (while additionally removing the clearcoating of the car and slightly increasing the BRDF values to keep the noise distinct for the human eye) results in a clear superiority of the new scheme also for low frequency noise. **Left:** Full view. **Right:** Slightly sharpened detail region view.
Chapter 4

Applications and Results

This chapter illustrates the benefits of the research results in practical applications. Note that although the application scenario is computer graphics, the new algorithms can easily be applied to other domains of scientific computing, especially transport simulation.

4.1 Compact Code: A 4KB Demo

The ray tracing kernel based on the BIH can be implemented in such an efficient manner that it is possible to use it in a just 4KB sized standalone executable (see Figure 4.1), without having to do further size optimizations that would otherwise decrease rendering performance, limit flexibility, increase memory footprint, or decrease numerical precision. As such, the 4KBs include the almost unchanged (on demand) ray tracing kernel that was used to ray trace all other BIH pictures and statistics in this work (except for the massive geometry, see Section 4.3, that of course needs additional disk swapping functionality).

Figure 4.1: Different design stages of the 4KB executable after geometry and camera position were finalized.

Note that the 4KBs additionally include a procedural generation of the geometry and the fog, two shaders (one of them approximating global illumination), picture post-processing, simple memory management, windows setup and blitting of the final picture to this window. Note that no additional libraries or data except for the standard Windows-API to open the window and handle keyboard input are utilized. The executable was released at the procedural graphics compo at the Buenzli demoparty. For an introduction to the world of the demoscene, see [Sch02].

This demonstrates the simplicity of the proposed implementation (and the included partitioning heuristics as presented in Section 2.4), while the performance is outperforming, or at least competitive with, existing (and far more complex) ray tracing engines (see provided statistics in Section 2.3.4).

4.2 Realtime Ray Tracing

To get an impression of the real life needs that games demand from a rendering engine, the Quake II game engine (originally developed and designed by id software) was used to test the ray tracing implementation
with. As the complete original source code of the game is available under the GPL, this work's ray tracing kernel and some custom made shaders were incorporated into the game as an optional renderer.

Figure 4.2: Difference shots of the standard Quake II rendering (using per-pixel lighting though) and the improved ray tracing version: Shadows and glossy reflections clearly enhance realism.

Two possibilities were thought of to design the plugin: Either use the existing separation of dynamic objects and static geometry to accelerate the hierarchy construction and rendering (as non-visible geometry can be optionally pre-culled by the Quake II engine), or think of all triangles as an unordered triangle soup, thus allowing unmatched flexibility and stress testing the acceleration data structure implementation. By choosing the second option, the actual port of the engine from using a traditional rasterizing approach to realtime ray tracing is exceptionally easy, as the implementation can work on a per frame basis and as such can be fully separated from the rest of the engine:

1. All triangles of the level and the inherited objects and enemies are collected in a single array, along with textures and other material information like transparency.
2. The ray tracing kernel constructs the complete acceleration structure for the scene. Note that using an on demand policy is of course possible, but was found unnecessary due to the fast construction.
3. Rays are shot by using 8x8 shafts (see Section 2.6.2) and optionally split into 2x2 ray bundles if the applied corner ray technique to trace a shaft fails (see Figure 2.48).
4. The 2x2 packets are immediately textured and shaded, and optional secondary rays are shot to include refractions, glossy reflections and point light shadows.
5. The picture is blitted and additional 2-dimensional effects like particles, HUD, and menus are rendered on top.

Note that unlike other released implementations of ray traced Quake "engines" (which in fact are only advanced level viewers), the presented ray traced version is based on the complete game code, thus the inclusion of Quake II mods, demos or even (internet) multi-player support does not pose any problem at all.

Figure 4.3: Closeups of glossy reflections and refraction. Additionally, the left picture shows a Quake II mod (XxX) running in action.

The implementation proves the superiority of the used hierarchy construction algorithm(s), as it hardly influences the performance at all (for both kd-tree and the BIH), even when using a non-parallelized version on a multi-core system. The achievable frame-rate mainly depends on the overall number of rays shot (up to four per camera sample) and especially the shading calculations (involving interpolating multiple texture maps and the full per-pixel lighting for all light sources). Note that this is subject to change with modern
games as the amount of (dynamic) geometry used in "next generation" titles is clearly not feasible for current ray tracing implementations, even when using specialized algorithms as presented in Sections 2.5.3 and 2.5.7.

At a resolution of 640x480, using 2x2 ray bundles only, the average performance is 11.6 fps (using the "timedemo" command on a Core2Duo 1.66GHz) using up to 4 rays per pixel for the whole screen. Utilizing 8x8 shafts (see Section 2.6.2), the performance can be further increased.

### 4.3 Interactive Visualization of Massive Scenes

Using the implementation presented in Section 2.5.7, it is possible to render pictures of a fully detailed Boeing 777 mesh out of a completely unsorted triangle soup in 2-8 minutes from scratch on a single CPU (see Table 4.4). Actually, the data structure setup times that made ray tracing infamous are indeed thrashed and not only static walkthroughs of massive models, but interactive changes of the data-sets become feasible.

#### Boeing View 1

<table>
<thead>
<tr>
<th>BIH mem.</th>
<th>View 1</th>
<th>View 2</th>
<th>View 3</th>
<th>View 4</th>
<th>View 5</th>
<th>View 6</th>
</tr>
</thead>
<tbody>
<tr>
<td>326,447,848</td>
<td>12,748,120</td>
<td>15,471,692</td>
<td>259,261,404</td>
<td>50,963,768</td>
<td>324,602,460</td>
<td></td>
</tr>
<tr>
<td>fps</td>
<td>0.26</td>
<td>0.13</td>
<td>0.13</td>
<td>0.38</td>
<td>0.11</td>
<td>0.34</td>
</tr>
<tr>
<td>Total</td>
<td>8 min.</td>
<td>133 sec.</td>
<td>153 sec.</td>
<td>270 sec.</td>
<td>118 sec.</td>
<td>252 sec.</td>
</tr>
</tbody>
</table>

Figure 4.4: Total rendering times (1280x1024 pixels, single core of an Opteron 875 2.2 GHz 32GB) including on demand tree construction for the Boeing 777 data set (349,569,456 triangles amounting 12,584,500,416 bytes) and overall memory usage for the BIH. Reading the triangle data from hard disk is omitted, since it heavily depends on the hard disks used (usually 40-90 seconds loading time).

Additionally, it is possible to limit the memory to a user chosen value and the implementation will fall back to include disk swapping (see Figure 4.5). Note that the results not even use the proposed compression scheme (see Section 2.3.3) to further reduce the necessary hard disk swapping.

#### Boeing View 1

<table>
<thead>
<tr>
<th>Peak memory</th>
<th>View 1</th>
<th>View 2</th>
<th>View 3</th>
<th>View 4</th>
<th>View 5</th>
<th>View 6</th>
</tr>
</thead>
<tbody>
<tr>
<td>1,075,418,112</td>
<td>1,054,035,968</td>
<td>1,267,494,912</td>
<td>1,078,779,904</td>
<td>1,100,248,448</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Rendering time</td>
<td>19 sec.</td>
<td>6 min. 46 sec.</td>
<td>6 min. 46 sec.</td>
<td>1 min. 57 sec.</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Figure 4.5: Rendering times (1280x1024 pixels, Pentium 4HT 2.8 GHz 2GB RAM, 349,569,456 triangles), including low level on demand tree construction and loading all necessary triangle groups from disk. The top level bucket sort preprocess, done once for all views, takes additional 53 minutes but only uses a peak 737 MBytes of RAM. The cache sizes for the preprocess and rendering were chosen to be suitable for any consumer machine offering at least 1 GB of RAM. More RAM allows for even faster rendering times (see previous figure), whereas the preprocessing step is mainly limited by the (slow and cheap) hard disk.

Moderately complex scenes (see statistics in Figure 2.44) can even be constructed in realtime using a multicore system and one of the proposed parallelization schemes (see Section 2.6.4).

### 4.4 Realtime Photorealistic Image Synthesis

This section will demonstrate the tight coupling of accelerated ray tracing and numerical algorithms to solve the light transport equation (see 1.2) by using the path integral formulation of Section 1.1.1. All efficient algorithms generate path segments from the light sources and the camera and try to connect both subpaths in order to find contributing paths. By applying the results of the previous chapter, the strategy is to first
generate an ensemble of paths starting from the light sources and connect them with an array of paths starting from the camera. The goal is to deliver a high quality image in close to realtime.

4.4.1 Path Generation for the Robust Global Illumination Algorithm

The general class to solve light transport is bidirectional path tracing (see Section 1.1.1). The simplest unbiased algorithm of this class certainly is the Robust Global Illumination (RGI) algorithm [KK04, Kol04a, Raa05], whose path generation will be explained in the following, and later refined to be used in combination with the simultaneous Markov chain simulation and sorted quasi-Monte Carlo of Section 3.4.

Algorithm

RGI is similar to the Instant Radiosity algorithm [Kel97], but in contrast provides finite variance at unbiased results.

- Starting from the light sources (partial) transport paths are generated and continued using a classical random walk technique
  - Upon interaction with a material, a point light source is stored for each particle in order to be able to connect with transport paths starting from the camera later on
  - New directions are sampled for each particle and the energy of a particle is changed to accompany the measurement contribution function and the probability density of the path
  - Ray tracing is used to continue the paths
- The simulation stops as all paths have been terminated by Russian roulette [AK90]
- In a second pass, the point light sources are connected with camera paths to form complete transport paths

Numerically Robust Handling of the Weak Singularity

As the connection of a camera path with the point light sources requires to evaluate the geometric term (see Section 1.1.1, Equation 1.7) \( G(x \leftrightarrow y) = \frac{\cos \theta_x \| y - x \|}{\| x - y \|} \), resulting in \( G(x \leftrightarrow y) = \infty \) for points \( x \) and \( y \) close in space, a classic solution to prevent overmodulation due to the weak singularity, is to clip the function:

\[
G'(x \leftrightarrow y) := \begin{cases} 
G(x \leftrightarrow y) & \text{if } G(x \leftrightarrow y) < b \\
0 & \text{otherwise}
\end{cases}
\]

As this introduces a bias into the simulation that will result in an overall too dark image (especially for all concave geometry). The bias of the bounded version \( L'(y \rightarrow z) \) is

\[
L(y \rightarrow z) - L'(y \rightarrow z) = \int_{\partial \Sigma} L(x \rightarrow y) f(x \rightarrow y \rightarrow z) \max\{G(x \leftrightarrow y) - b, 0\} V(x \leftrightarrow y) d\lambda(x)
\]

\[
= \int_{\partial \Sigma} L(x \rightarrow y) f(x \rightarrow y \rightarrow z) \frac{\max\{G(x \leftrightarrow y) - b, 0\}}{G(x \leftrightarrow y)} G(x \leftrightarrow y) V(x \leftrightarrow y) d\lambda(x)
\]

\[
= \int_{\partial \Sigma} L(h(y, -\omega) \rightarrow y) f(h(y, -\omega) \rightarrow y \rightarrow z) \frac{\max\{G(h(y, -\omega) \leftrightarrow y) - b, 0\}}{G(h(y, -\omega) \leftrightarrow y)} |\cos \theta_x| d\sigma(\omega).
\]

[KK04] proposed the usage of a correction term for cases

\[
G(x \leftrightarrow y) \geq b \iff \frac{|\cos \theta_x| |\cos \theta_y|}{\| y - x \|^2} \geq b \iff \| y - x \| \leq \sqrt{\frac{|\cos \theta_x| |\cos \theta_y|}{b}}.
\]
and evaluate by integration over the hemisphere instead of area. The actual implementation is fairly simple as the rays that sample the hemisphere will just recursively call the existing shader on an interaction with the intersected geometry. Note that these rays can actually be bound by length $l_{\text{max}} = \sqrt{\cos \theta + y}$, thus avoiding a lot of calculations for the bias compensation as a large fraction of the shot rays actually ends in empty space. In fact this is similar to terminating the camera path by using russian roulette. Still, the correction term will introduce additional noise to the image in regions featuring concave geometry, a common problem for brute force hemisphere sampling techniques. However, it results in a fully unbiased algorithm if using MC or RQMC sampling techniques.

4.4.2 Simultaneous Simulation of Markov Chains

As shown in the previous chapter, the noise level can be reduced by applying the simultaneous simulation of Markov chains and sorting, similar to the retrieved results in Figures 3.15 and 3.16. By using sorted QMC for the random walks, the local distribution of the point light sources is becoming more uniform, thus decreasing the "variance" of the results over standard QMC and MC techniques (see Figure 3.12).

Sampling the image plane does not demand an explicit sorting step, as it is possible to directly enumerate the screen samples using the Z-curve ordering (see Section 3.4.3). To handle the scattering events for the initial random walks and the bias compensation step, the implementation utilizes a wavefront based simulation. Upon an interaction with the surfaces, the particles are resorted and subsequent QMC samples are drawn to sample the hemisphere.

Increasing Performance by Omitting Shadow Rays

To make global illumination calculations possible in close to realtime on current hardware, additional tricks must be utilized. Apart from using a hybrid solution that calculates the random walks on the CPU and the shading on optional graphics hardware [Kel97, Kel98, SIMP06b], it is also possible to just use a subset of the light sources for shading, depending on a coarsely sampled estimation of the contribution to the final image [WBS03b, GWS05]. Note that the latter can be made unbiased by randomly selecting additional light sources from the whole set. Advanced techniques that incorporate more advanced estimators can be found in [SIMP06a, SIP07].

Although the mentioned algorithms can already discard a large portion of the complete set of light sources, it is possible to further decrease the number of locally used light sources by utilizing local estimators. Related to proposals found in [War91a, Woo93, HDG99], the set of light sources is sorted by the contribution they add to a single sample of the image. So instead of connecting the sample to the complete set of light sources, only $x$ point lights with the largest contributions are taken into account. The introduced bias can optionally be removed by adding additional randomly picked light sources. Note that discarding $\frac{3}{4}$ of the shadow rays will usually still result in comparable images but at a fraction of the rendering time (see Figure 4.6).

4.4.3 Photon Mapping by Trajectory Splitting in the RGI Algorithm

All bidirectional path tracing algorithms suffer the problem of insufficient techniques [KK02a, Kol04a], i.e. paths that are difficult to generate (see Figure 4.12). To support the generation and evaluation of such light transport paths, it is possible to add biased techniques, such as photon maps [Arv86, Jen96, KW00] to the RGI implementation. Unlike the original algorithm though, separate photon maps are generated at each vertex of the initial random walk paths by splitting the path into $N$ new subpaths. The new photons are then used to only simulate scattering along highly glossy and refractive surfaces, as the more diffuse effects are perfectly captured by the RGI algorithm already, resulting in a drastic decrease of memory usage and construction time of the smaller photon map. Incorporating the sorted quasi-Monte Carlo method is actually fairly easy (see Section 3.4.2), as one can simply draw more samples out of the utilized $(t, s)$-sequence for the splitted subpaths. Optionally, it is also possible to use a stratified point set (such as rank-1 lattices as in Section 3.3.3) for the sampling of the hemisphere at path-vertices to be split (which in fact is very similar to the principle of using shadow maps), which further enhances the local uniformity of path space sampling.
Figure 4.6: Full Robust Global Illumination versus a biased implementation discarding $\frac{3}{4}$ of the shadow rays. Render time is decreased to 28 (Conference Room) / 26 (Shirley's "Scene 6") percent at the price of a slight overestimation of the light contribution in some detailed areas.

In addition, an optimized variant could be thought of that uses importance sampling to prefer regions of the hemisphere where highly glossy and refractive surfaces are likely to be expected.

Figure 4.7: Robust global illumination and additional photon mapping to add missing contributions transmitted by glossy and refractive surfaces. Light is emitted from four different colored area light sources on the ceiling. Adding the photon map to the implementation increases render-time by just 6.2 percent.

Although the inclusion of photon maps is rather easy and produces nice results in little additional time (see Figure 4.7), it must be noted that using only few paths for the initial random walk (like it is the case for realtime and interactive visualizations) results in distinct artifacts in the caustics, an effect once again similar to shadow maps, which also suffers from undersampling in the form of aliasing and thus shadowing artifacts. Still, accumulating over multiple frames is possible and the caustic artifacts will vanish.
4.5 Interactive Car Visualization

While it is possible to easily SIMD optimize shaders and even the whole framework for a fixed function rendering engine like Quake II, this is clearly not possible anymore for user defined recursive shaders. While there has been a proposal on automatic recompiling of shaders to incorporate SIMD tracing and partially SIMD shading recently [PBBR07], the freedom for shader writers is clearly decreased, thus making it impractical for professional off-line rendering packages.

Also note, that rendering times are not relying by 80-95 percent on ray tracing anymore [Whi80], but are almost on par with the involved shading calculations, especially when using complex analytic BRDF models like the Ashikhmin-Shirley model [AS00] or measured BRDF data (see Section 1.1.2), which are additionally hard to optimize using SIMD opcodes. This ratio is shifted even more when incorporating currently popular shaft tracing techniques to accelerate primary rays and some of the secondary effects.

Figure 4.8: High quality car visualizing using global illumination approximations and measured BRDFs.

While high quality, straight forward solutions to visualize a car lit from a HDRI environment map take several minutes (see Figure 4.9) to get a more or less noise and artifact free picture, a custom shader, designed especially for car visualization, can deliver comparable results within seconds.

The implementation is based on path tracing, but uses adaptive splitting to reduce noise, and ambient occlusion [Lan02, IKSZ03] to approximate global illumination effects inside the car and for the shadowing on the geometrically not existing ground.

By using the Halton points to uniformly distribute samples over the whole screen [Kel06] (accompanied by a Z-curve ordering (see Section 3.4.3) of the samples to increase coherency), aliasing can be reduced more efficiently but at the price of additional memory necessary for tabulation. Additionally, the simulation features multi-pass rendering utilizing QMC sequences for sampling to deliver a fast realtime preview. On no user interaction, the image is automatically refined without discarding the samples of the preview.

Optionally the simulation can use sorted QMC sampling (see Section 3.4.2), however, it must be noted that the additional sorting step only becomes valuable when using multiple samples per pixel. Otherwise the performance decrease due to the additional sorting step outweighs the perceptual noise reduction as seen in Section 3.4.4.

Estimating the pixel value by using a box filter to integrate the samples can optionally be replaced by a higher order approximation of the Gauss filter, incorporating tone mapping [RSSF02, Eq.3.4] and an optional bloom filter to allow bright spots in the picture to distribute energy onto surrounding pixels. Note that the last steps should obviously be omitted if storing HDR pictures. While bloom and glow filters are commonly used in current game engines to increase "realism", most photorealistic renderings will look worse. The reason can be found in the psychological effect of the uncanny valley [Mor70]: While slightly realistic renderings can often be tweaked to look more realistic by simple filter operations, a full physically based simulation demands high quality, locally adaptive tone mapping [RSSF02, LFUS06] to preserve realism. With the (upcoming?) broad usage of high dynamic range displays this problem will solve itself in the future, though.

An additional problem of using floating point based BRDFs, HDRIs, and image buffers for the simulation is the drastically increased memory usage. While a block based quantization similar to texture compression found on graphics hardware is an option, it will at the same time increase rendering time as the decompression must be done in software. A simple way to cut down memory requirements to $\frac{1}{3}$ without removing too much precision for floating point colors is the use of the R G B E format [War91b]. The simple scheme did not reduce the rendering performance at all, as the additional operations are outweighed by reduced cache thrashing and less memory transfers.
While using a simple Phong based BRDF model in combination with a Fresnel approximation [Sch94, SHSL97] seems like a valid option to decrease the memory usage, the price paid for the image quality disqualifies this solution for commercial usage.

Figure 4.9: Full solution using path tracing and measured BRDFs only. Instead of seconds, the render time here was several minutes, with some minor artifacts still remaining, even when using a factorization [LRR04] to importance sample the BRDFs.

Note that a common problem for physically correct simulations that must use HDRI environment maps to light a scene is the wrong energy of bright spots (like the sun) in the map. As the dynamic range of the measurements is limited by the currently available hardware, bright spots are implicitly darkened and as such the overall energy is reduced. This results in scene geometry looking "too dark" when compared to the surrounding environment map. This can only be "solved" by breaking with the rule of physically correctness and thus energy conserving BRDFs, so the materials will distribute more energy into the camera.

4.5.1 Actually, Nobody can Render a Car, yet

Although the presented application already delivers convincing results, a true simulation poses more demands on the underlying geometry and the rendering methods.

**Free Form Surfaces:** To avoid artifacts in reflection, refraction and shadowing (see Section 2.1.3), full floating point precision ray tracing of the original CAD data set is necessary. One step in that direction was presented in [Dam05, DK06].

**Coating and Sparkles:** The coating of a car is usually varying in structure and thickness along with additional microfacet effects that become visible upon close inspection. The simulation of such effects can hardly be simulated by only using additional procedural bump/normal maps (see Figure 4.10).

Figure 4.10: Almost perfectly smooth car coating versus a (fake) simulation of a real surface with variations in smoothness and coating.
Some car coatings even contain particles that must be either fully simulated by micro-geometry [Rze05] or more commonly approximated by simple techniques [GCG+05].

An even bigger problem is the simulation of spectral variations in the coating (ranging from subtle color variations to a full flip-flop effect), as these demand an increase in the amount of samples due to the multiple wavelength simulation and increased memory usage due to bloated BRDF data.

**Polarization:** A subtle, but necessary effect to simulate is the change of polarization of light. This will become visible especially on the reflections on the car glass. Simulating polarization generally demands more samples and results in vastly increased computation time [WTU+04].

**Exponential Ray Tree:** Visualizing the headlight of a car (see Figure 4.11) will result in an exponential increase in the number of rays due to the nature of the shading tree that induces several rays on each interaction with a material. By omitting such splitting techniques, additional noise will be introduced to the image.

![Figure 4.11: Different recursion depths used for the path tracing, visualizing the simple headlight of the car.](image)

The ray tree (and thus the overall computation time) ranges from a single ray for reflection (~1 minute), to 2 more rays for reflection and refraction (~3 minutes), to 4 more rays for reflection and refraction (~8 minutes) and finally $1 + 2 + 4 + 8 = 15$ rays for reflection and refraction (~25 minutes), resulting in an exponential explosion of computation time.

**Fast Global Illumination:** Simulating transport paths that pass through the car glass, bounce in the inside, back through the car glass and into the camera lens (see Figure 4.12), is a very difficult setting for every sampling based renderer. The only algorithms capable to visualize correct and unbiased images are full path tracing techniques, resulting in extremely long render times, as these paths are difficult to generate [Vea97, Kol04a]. While photon mapping can decrease the rendering time, it also introduces a bias into the calculations and features additional memory usage. In addition, precision problems due to ignoring topology (especially at corners) are introduced [Kel03].

Most of these problems are not solved in a practical environment, so actually nobody can render a car, yet.
Figure 4.12: A difficult scenario for every bidirectional path tracer: Light transport paths containing scattering events on highly specular and refractive surfaces.
Chapter 5

Conclusion

The results of this thesis are already widely applied by industry (including the mental images RealityServer and future product lines) and the academic community (see e.g. [WMG+07, Wal07]) and in addition are also recognized by hobbyists (e.g. Quadrille/Radius, Indigo, Sunflow, Lucille, XFRT, xiphos, RenderSpud, Gangsta). Additionally, the bounding interval hierarchy is supposedly included in the public available IBM ray tracing demo running on the cell platform [MNM06] and in the Blender package, while [MKK07] uses it to better support the interactive simulation of wireless networks, and the Arauna renderer [Bik07] for education and state of the art visualization of fully dynamic objects in a game environment. Along its unbeaten simplicity and robustness it also gave rise to a lot of research, especially because of its rapid construction using the proposed global heuristic.

The presented unified ray tracing framework allows to render massive geometry within a strictly limited memory block at unmatched performance, while the same framework is able to effectively visualize fully dynamic scenes, such as games. The tight coupling of construction, partitioning heuristic and ray traversal even allows to further decrease the total time to image by introducing a better amortization of the construction time of the hierarchy. As demonstrated, upcoming ray tracing hardware will also profit from the findings, as the proposed algorithms are exceptionally simple to implement and can optionally run in a priori fixed memory blocks.

In the same spirit, the simplified quasi-Monte Carlo Markov Chain simulation algorithm features a close to trivial implementation, but notable convergence improvements and is already included in the mental images product line in fusion with the unified ray tracing framework. The research shed light on the effects of scrambling and the enumeration order when implementing \((t,s)\)-sequences. This certainly is a good entry point for better understanding the relation of \((t,s)\)-sequences and lattice-sequences and finding better point sequences for simulation purposes.

The work successfully applied the principle of understanding and identifying remaining degrees of freedom. Further research results are to be expected by closely integrating simulation and visualization, i.e. ray tracing, acceleration data structures, shading and dynamics, to further simplify the algorithms and improve performance.

"No... no more scientists, no more laboratories, no more experiments. I thought you'd be able to understand that. No more!" - Sayer of the Law, The Island of Dr. Moreau
Bibliography


