# 15 Years of Instant Radiosity

Alexander Keller



### Outline

#### **Light Transport Simulation**

- Using many point light sources
- The accumulation buffer
- Light path vertices as point light sources
- Instant radiosity

#### **Path Space Partitioning**

The weak singularity

#### **Consistent Generation of Light Transport Paths**

- Monte Carlo and quasi-Monte Carlo integration
- **15 Years of Instant Radiosity** 
  - Hindsights



### Using many point light sources

accumulation of images illuminated by a point light source





rendered by the rasterizer with shadow maps

#### The accumulation buffer





#### The accumulation buffer

-





#### The accumulation buffer







#### The accumulation buffer

\* \* \* \* \* \*





#### The accumulation buffer

\* \* \* \* \* \*





#### The accumulation buffer

. . . . . . . . . .





#### The accumulation buffer

. . . . . . . . . . . . .





#### The accumulation buffer

\* \* \* \* \* \* \* \* \* \* \* \* \* \* \*





#### The accumulation buffer





#### The accumulation buffer





#### The accumulation buffer





### Light path vertices as point light sources

• trace light paths and store vertices  $(x_j, L_j(x_j \rightarrow \cdot))_{i=1}^M$  as point lights





### Light path vertices as point light sources

• trace light paths and store vertices  $(x_j, L_j(x_j \rightarrow \cdot))_{i=1}^M$  as point lights



in fact equivalent to generating a tiny photon map



Light path vertices as point light sources

trace camera paths





### Light path vertices as point light sources

trace shadow rays to illuminate camera paths by points lights





### Light path vertices as point light sources

trace shadow rays to illuminate camera paths by points lights



$$L(y,z) \approx L_e(y,z) + \sum_{j=1}^M L_j(x_j \to y) f_r(x_j,y,z) \frac{\cos \theta_{x_j} \cos \theta_y}{|x_j - y|^2}$$

#### Instant radiosity





### Instant radiosity

-





### Instant radiosity







### Instant radiosity

. . . . . . .





### Instant radiosity

\* \* \* \* \* \*





### Instant radiosity

. . . . . . . . .





### Instant radiosity

\* \* \* \* \* \*





### Instant radiosity





### Instant radiosity





### Instant radiosity





### Instant radiosity





### Instant radiosity





### Instant radiosity





### Instant radiosity





### Instant radiosity





### Instant radiosity





### Instant radiosity




## Instant radiosity





## Instant radiosity





## Instant radiosity





## Instant radiosity





## Instant radiosity





## Instant radiosity





#### The weak singularity

at that time, hardware avoided overmodulation of geometric term

$$G(x_j, y) := V(x_j, y) rac{\cos heta_{x_j} \cos heta_y}{|x_j - y|^2}$$

for small distances  $|x_j - y|^2$  and visibility  $V(x_j, y) = 1$  by clipping



#### The weak singularity

at that time, hardware avoided overmodulation of geometric term

$$G(x_j, y) := V(x_j, y) rac{\cos heta_{x_j} \cos heta_y}{|x_j - y|^2}$$

for small distances  $|x_j - y|^2$  and visibility  $V(x_j, y) = 1$  by clipping

$$L_d(y,z) = \int_{\text{supp } L_e} L_e(x,y) f_r(x,y,z) G(x,y) \, dx$$



#### The weak singularity

at that time, hardware avoided overmodulation of geometric term

$$G(x_j, y) := V(x_j, y) rac{\cos heta_{x_j} \cos heta_y}{|x_j - y|^2}$$

for small distances  $|x_j - y|^2$  and visibility  $V(x_j, y) = 1$  by clipping

$$L_d(y,z) = \int_{\operatorname{supp} L_e} L_e(x,y) f_r(x,y,z) G(x,y) dx$$
  

$$\approx \int_{\operatorname{supp} L_e} L_e(x,y) f_r(x,y,z) \min\{G(x,y),b\} dx$$



#### The weak singularity

at that time, hardware avoided overmodulation of geometric term

$$G(x_j, y) := V(x_j, y) rac{\cos heta_{x_j} \cos heta_y}{|x_j - y|^2}$$

for small distances  $|x_j - y|^2$  and visibility  $V(x_j, y) = 1$  by clipping

$$L_{d}(y,z) = \int_{\sup L_{e}} L_{e}(x,y) f_{r}(x,y,z) G(x,y) dx$$
  
= 
$$\int_{\sup L_{e}} L_{e}(x,y) f_{r}(x,y,z) \min\{G(x,y),b\} dx$$
  
+ 
$$\int_{\sup L_{e}} L_{e}(x,y) f_{r}(x,y,z) \max\{G(x,y) - b, 0\} dx$$



#### The weak singularity

at that time, hardware avoided overmodulation of geometric term

$$G(x_j, y) := V(x_j, y) rac{\cos heta_{x_j} \cos heta_y}{|x_j - y|^2}$$

for small distances  $|x_j - y|^2$  and visibility  $V(x_j, y) = 1$  by clipping

$$L_d(y,z) = \int_{\operatorname{supp} L_e} L_e(x,y) f_r(x,y,z) G(x,y) dx$$
  
= 
$$\int_{\operatorname{supp} L_e} L_e(x,y) f_r(x,y,z) \min\{G(x,y),b\} dx$$
  
+ 
$$\int_{\mathscr{S}^2(y)} L_e(t,y) f_r(t,y,z) \frac{\max\{G(h(y,\omega),y) - b,0\}}{G(h(y,\omega),y)} \cos \theta_y d\omega$$



## The weak singularity

$$L_d(y,z) \approx \frac{|\operatorname{supp} L_e|}{n} \sum_{j=0}^{n-1} L_e(x_j,y) f_r(x_j,y,z) \min\{G(x_j,y),b\}$$





## The weak singularity

$$L_{d}(y,z) \approx \frac{|\operatorname{supp} L_{e}|}{n} \sum_{j=0}^{n-1} L_{e}(x_{j},y) f_{r}(x_{j},y,z) \min\{G(x_{j},y),b\}$$



# 

## The weak singularity

$$L_{d}(y,z) \approx \frac{|\operatorname{supp} L_{e}|}{n} \sum_{j=0}^{n-1} L_{e}(x_{j},y) f_{r}(x_{j},y,z) \min\{G(x_{j},y),b\} + \frac{\pi}{n'} \sum_{j=0}^{n'-1} L_{e}(h(y,\omega_{j}),y) f_{r}(h(y,\omega_{j}),y,z) \frac{\max\{G(h(y,\omega_{j}),y)-b,0\}}{G(h(y,\omega_{j}),y)}$$



# 

## The weak singularity

$$L_{d}(y,z) \approx \frac{|\operatorname{supp} L_{e}|}{n} \sum_{j=0}^{n-1} L_{e}(x_{j},y) f_{r}(x_{j},y,z) \min\{G(x_{j},y),b\} + \frac{\pi}{n'} \sum_{j=0}^{n'-1} L_{e}(h(y,\omega_{j}),y) f_{r}(h(y,\omega_{j}),y,z) \frac{\max\{G(h(y,\omega_{j}),y)-b,0\}}{G(h(y,\omega_{j}),y)}$$



## The weak singularity

biased result of only bounding the weak singularity





## The weak singularity

consistent simulation





## Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\mathbb{P}\left(\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) - \int_{[0,1)^s} f(x) dx = 0\right) = 1$$



## Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\mathbb{P}\left(\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) - \int_{[0,1)^s} f(x) dx = 0\right) = 1$$





## Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\mathbb{P}\left(\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) - \int_{[0,1)^s} f(x) dx = 0\right) = 1$$





## Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\mathbb{P}\left(\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) - \int_{[0,1)^s} f(x) dx = 0\right) = 1$$





## Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\mathbb{P}\left(\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) - \int_{[0,1)^s} f(x) dx = 0\right) = 1$$





## Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\mathbb{P}\left(\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) - \int_{[0,1)^s} f(x) dx = 0\right) = 1$$





## Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\mathbb{P}\left(\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) - \int_{[0,1)^s} f(x) dx = 0\right) = 1$$





## Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\mathbb{P}\left(\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) - \int_{[0,1)^s} f(x) dx = 0\right) = 1$$





## Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\mathbb{P}\left(\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) - \int_{[0,1)^s} f(x) dx = 0\right) = 1$$





## Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\mathbb{P}\left(\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) - \int_{[0,1)^s} f(x) dx = 0\right) = 1$$





## Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\mathbb{P}\left(\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) - \int_{[0,1)^s} f(x) dx = 0\right) = 1$$





## Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\mathbb{P}\left(\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) - \int_{[0,1)^s} f(x) dx = 0\right) = 1$$





## Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\mathbb{P}\left(\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) - \int_{[0,1)^s} f(x) dx = 0\right) = 1$$





## Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\mathbb{P}\left(\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) - \int_{[0,1)^s} f(x) dx = 0\right) = 1$$





## Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\mathbb{P}\left(\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) - \int_{[0,1)^s} f(x) dx = 0\right) = 1$$





## Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\mathbb{P}\left(\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) - \int_{[0,1)^s} f(x) dx = 0\right) = 1$$





## Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\mathbb{P}\left(\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) - \int_{[0,1)^s} f(x) dx = 0\right) = 1$$





## Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\mathbb{P}\left(\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) - \int_{[0,1)^s} f(x) dx = 0\right) = 1$$





## Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\mathbb{P}\left(\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) - \int_{[0,1)^s} f(x) dx = 0\right) = 1$$




#### Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\mathbb{P}\left(\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) - \int_{[0,1)^s} f(x) dx = 0\right) = 1$$





#### Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\mathbb{P}\left(\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) - \int_{[0,1)^s} f(x) dx = 0\right) = 1$$





#### Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) = \int_{[0,1)^s} f(x) dx$$





### Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) = \int_{[0,1)^s} f(x) dx$$





### Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) = \int_{[0,1)^s} f(x) dx$$





#### Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\mathbb{P}\left(\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) - \int_{[0,1)^s} f(x) dx = 0\right) = 1$$

- example: anti-aliasing a zone plate at 4 samples per pixel
  - · aliasing independent of numerical integration scheme





### Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\mathbb{P}\left(\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) - \int_{[0,1)^s} f(x) dx = 0\right) = 1$$

- example: emission and scattering using pseudo-random sampling





### Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) = \int_{[0,1)^s} f(x) dx$$

- example: emission and scattering using low discrepancy sequences





### Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) = \int_{[0,1)^s} f(x) dx$$

- example: sampling the cosine weighted hemisphere (diffuse BRDF)





### Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) = \int_{[0,1)^s} f(x) dx$$

- example: sampling the cosine weighted hemisphere (diffuse BRDF)





### Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) = \int_{[0,1)^s} f(x) dx$$

- example: sampling the cosine weighted hemisphere (diffuse BRDF)





#### Monte Carlo and quasi-Monte Carlo integration

consistency matters most

$$\lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} f(x_i) = \int_{[0,1)^s} f(x) dx$$

#### You still don't trust deterministic consistent sampling ?

come to the course

Advanced (Quasi-) Monte Carlo Methods for Image Synthesis Thursday, 9 August 9:00 AM - 12:15 PM — LACC - Room 406AB



# **15 Years of Instant Radiosity**

#### Hindsights

- improved accumulation buffer
  - added indirect illumination by using many point lights
    - · unlucky formulation of "operator norm estimation"
  - introduced quasi-Monte Carlo methods for light transport simulation
    - much more advanced by today
- Iimitations of rasterization
  - generating the light paths using the rasterizer is tricky
  - reflection and refraction are not feasible in a general way

#### Acknowledgements

Peter Schröder

