

# Instant Radiosity

by

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# *Outline of Talk*

- Global Illumination and Radiosity
- Basic Techniques:
  - Quasi-Monte Carlo Integration
  - Quasi-Random Walk
  - Accumulation Buffer
- The Instant Radiosity Algorithm
- Extensions, Conclusion and Future Research

# *The Global Illumination Problem*

- Given by  $(S, f_r, L_e, \Psi)$ 
  - reflectance distribution  $f_r(\vec{\omega}_i, y, \vec{\omega}_r)$  on the surface  $S$
  - source radiance  $L_e(y, \vec{\omega}_r)$  from  $y \in S$  into direction  $\vec{\omega}_r$
  - flux detector  $\Psi(y, \vec{\omega})$

- Seeking for

$$\langle L, \Psi \rangle := \int_S \int_{\Omega} L(y, \vec{\omega}) \Psi(y, \vec{\omega}) \cos \theta d\omega dy$$

where

$$L(y, \vec{\omega}_r) = L_e(y, \vec{\omega}_r) + \underbrace{\int_{\Omega} f_r(\vec{\omega}_i, y, \vec{\omega}_r) L(h(y, \vec{\omega}_i), -\vec{\omega}_i) \cos \theta_i d\omega_i}_{(T_{f_r} L)(y, \vec{\omega}_r)}$$

– first point  $h(y, \vec{\omega}_i) \in S$  hit by the ray from  $y$  into direction  $\vec{\omega}_i$

- Special case: Diffuse environments (radiosity setting)

$$f_r = f_d(y) := \frac{\rho_d(y)}{\pi}$$

# State of the Art

- Galerkin method with kernel and solution discretization

$$L_i = L_{e,i} + \rho_i \sum_{j=1}^K L_j f_{ij}$$

- Random walk approach with only solution discretization

$$L(y) \approx \sum_{k=1}^K \langle L, \Psi_k \rangle \Psi_k(y)$$

- Bidirectional path tracing

$$\langle L, \Psi_P \rangle = \sum_{j=0}^{\infty} \sum_{n=0}^j w_{jn} \langle T_{f_r}^n L_e, T_{f_r}^{*(j-n)} \Psi_P \rangle$$

# *The new Approach*

- Goals:
  - very fast and simple algorithm
  - avoid memory overhead introduced by discretizations
  - allow textures, cyclic scene graphs, dynamic environments
- Means:
  - low discrepancy sampling (quasi-Monte Carlo integration)
  - standard graphics hardware and rendering pipelines

# Quasi-Monte Carlo Integration

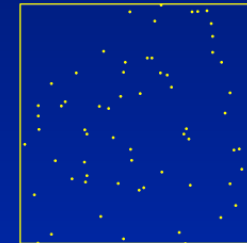
- Uniformly distributed points  $P_N = \{x_0, \dots, x_{N-1}\} \subset [0, 1)^s$

$$\int_{[0,1)^s} f(x) dx \approx \frac{1}{N} \sum_{i=0}^{N-1} f(x_i)$$

- Convergence rate

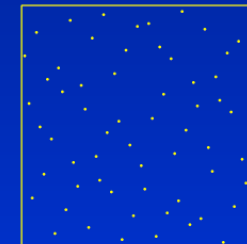
– for random sampling

$$\mathcal{O}\left(\frac{1}{\sqrt{N}}\right)$$



– for low discrepancy sampling

$$\mathcal{O}\left(\frac{\log^s N}{N}\right)$$



# Low Discrepancy Sample Points

- Radical inversion of  $i$  in base  $b$

$$\Phi_b(i) := \sum_{j=0}^{\infty} a_j(i) b^{-j-1} \Leftrightarrow i = \sum_{j=0}^{\infty} a_j(i) b^j$$

- Halton sequence

$$x_i = \left( \Phi_{b_1}(i), \dots, \Phi_{b_s}(i) \right)$$

- Hammersley point set

$$x_i = \left( \frac{i}{N}, \Phi_{b_1}(i), \dots, \Phi_{b_{s-1}}(i) \right)$$



# The Quasi-Random Walk

- Projection onto  $(\Psi_k)_{k=1}^K$  spanning  $S = \text{supp } \sum_{k=1}^K \Psi_k$

$$L(y) \approx \sum_{k=1}^K \langle L, \Psi_k \rangle \Psi_k(y)$$

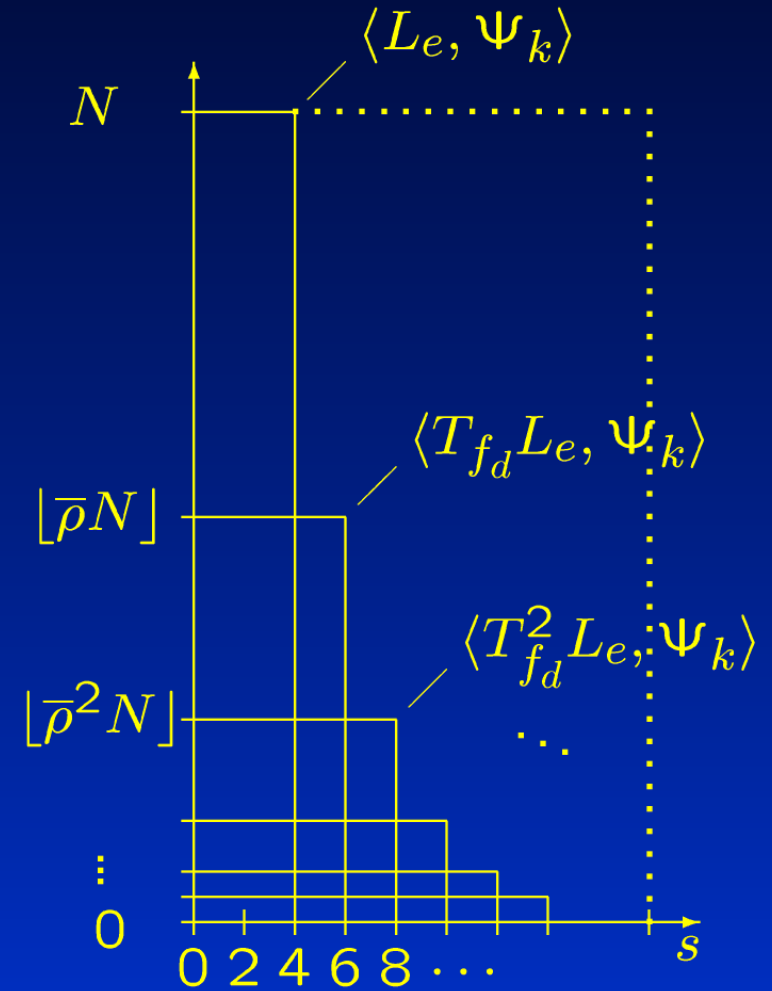
- $\bar{\rho} = \frac{\sum_{k=1}^K |A_k| \rho_k^d}{\sum_{k=1}^K |A_k|} \approx \|T_{f_d}\| < 1 \Rightarrow$  convergent Neumann series

$$\Rightarrow \langle L, \Psi_k \rangle = \underbrace{\langle L_e, \Psi_k \rangle}_N + \underbrace{\langle T_{f_d} L_e, \Psi_k \rangle}_{[\bar{\rho}N]} + \underbrace{\langle T_{f_d}^2 L_e, \Psi_k \rangle}_{[\bar{\rho}^2N]} + \dots$$

```

Start = End = N; Reflections = 0;
while(End > 0)
{
    Start *=  $\bar{\rho}$ ;
    for(i = (int) Start; i < End; i++)
    {
        Sample position  $y$ 
        for(j = 0; j <= Reflections; j++)
        {
            Record contribution to  $\Psi_k$  in  $y$ 
            Sample direction  $\vec{\omega}$ 
             $y = h(y, \vec{\omega});$ 
            Attenuate particle
        }
    }
    Reflections++;
    End = (int) Start;
}

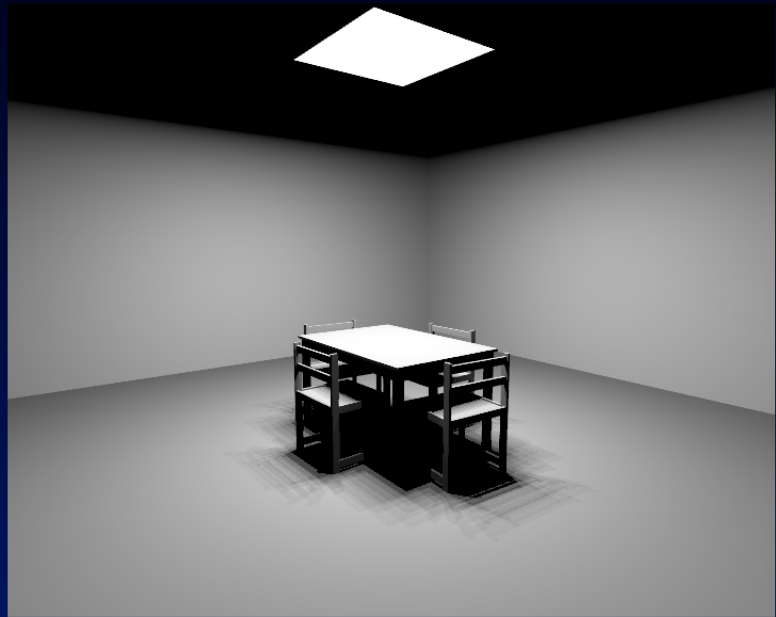
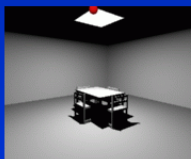
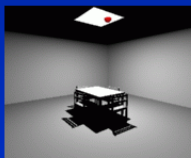
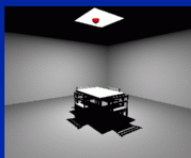
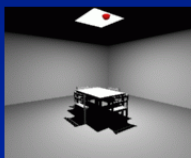
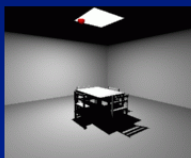
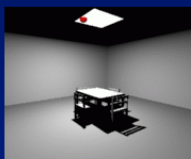
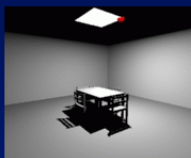
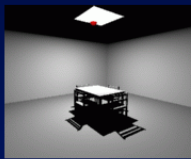
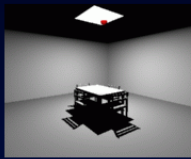
```



# The Accumulation Buffer

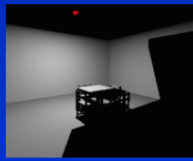
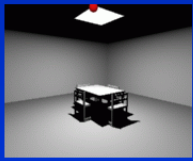
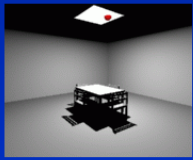
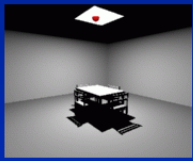
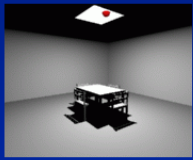
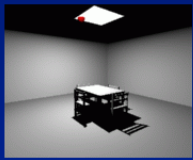
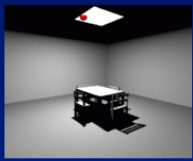
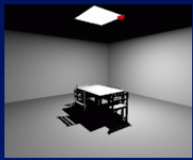
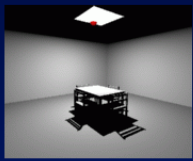
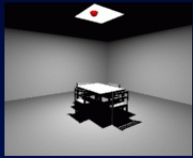
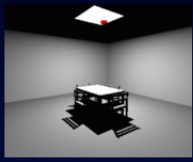
- Direct illumination by extended light sources

```
for(int i = 0; i < N; i++)  
{  
    Sample position  $y$   
     $L = L_e(y) * \text{supp } L_e$ ;  
    glRenderShadowedScene( $L, y$ );  
    glAccum(GL_ACCUM,  $\frac{1}{N}$ );  
}  
glAccum(GL_RETURN, 1.0);
```



# The Instant Radiosity Algorithm

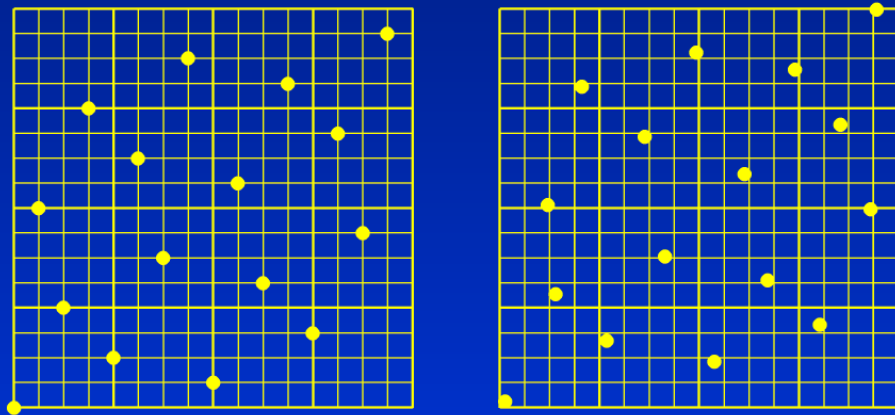
```
Start = End = N; Reflections = 0;
while(End > 0)
{
    Start *=  $\bar{\rho}$ ;
    for(int i = (int) Start; i < End; i++)
    {
        Sample position  $y$ 
         $L = L_e(y) * \text{supp } L_e$ ;
        for(int j = 0; j <= Reflections; j++)
        {
            glRenderShadowedScene ( $\frac{N}{\lceil \rho N \rceil} L, y$ );
            glAccum(GL_ACCUM,  $\frac{1}{N}$ );
            Sample direction  $\vec{\omega}$ 
             $y = h(y, \vec{\omega})$ ;
             $L *= f_d(y)$ ;
        }
    }
    Reflections++; End = (int) Start;
}
glAccum(GL_RETURN, 1.0);
```



***Conference room Image***

# Extensions

- Specular effects
  - last reflection: Full BRDF by hardware
  - virtual light sources
  - multipass techniques
- Jittered low discrepancy sampling





# *Radiosity in Dynamic Environments*

- Fixed length quasi-random walk

$$l_{max} := \left\lceil -\frac{\log N}{\log \bar{\rho}} \right\rceil$$

– frame rate divided by  $l_{max} \Rightarrow$  (close to) realtime

- Display with motion blur

– superimpose the last  $N$  composite frames with weight  $\frac{1}{N}$

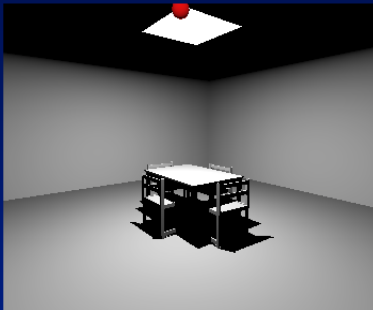
$t$

$$\langle L_e, \Psi_{mn} \rangle + T_{mn} L_e$$

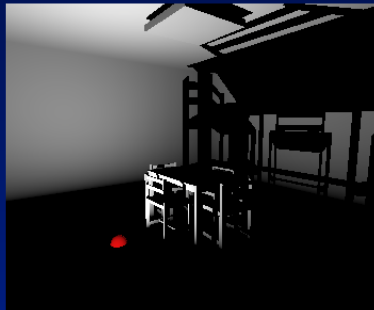
$$T_{mn} T_{fd} L_e$$

$$T_{mn} T_{fd}^2 L_e$$

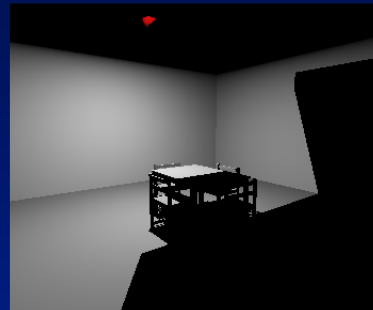
$i$



+



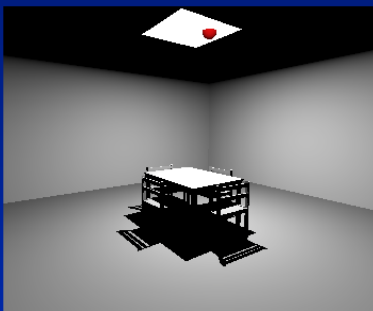
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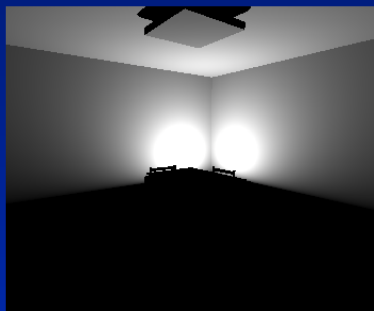
=



$i + 1$



+



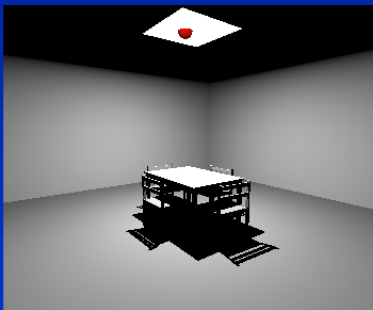
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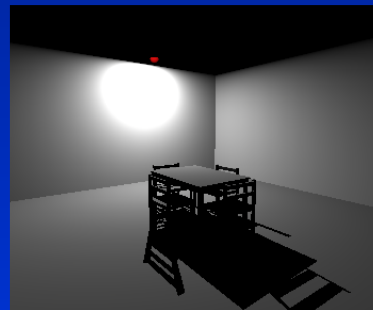
$i + 2$



+



+



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## ***Conclusion***

- Fast and simple global illumination algorithm
  - quasi-Monte Carlo integration
  - existing graphics hardware
- Compatible with standard graphics APIs

# *Future Research*

- Hardware implementation
  - shadow maps
  - specular effects
- Volume scattering
- Adjoint transport operator
- Dynamic environments (AR and VR)