Bidirectional Light Transport

Today

- Path tracing challenges
- Path integral form of the rendering equation
- Light tracing
- Bidirectional path tracing
- Photon mapping



















Path Tracing, 32 samples / pixel



Sun vs. Earth Solid Angles





149.6B m

	Solid Angle
Sun f/Earth	6.87e-5 sr
Earth f/sun	5.70e-9 sr
1 m^2 on Earth f/sun	4.47e-17 sr

r=6.371M m

 \bigcirc

% of Sphere 0.00547% 0.000000454% 0.0000000000003555%

Rendering Equation: Directional Form

$$L(\mathbf{p},\omega) = L_e(\mathbf{p},\omega) + \int_{H^2} f_r(\mathbf{p},\omega_i \to \omega) L(\mathbf{p},\omega_i \to \omega) L(\mathbf{p},\omega$$

Can equivalently write as an integral over surface area of objects in the scene by applying

$$\mathrm{d}\omega = \frac{\cos\theta}{r^2}\mathrm{d}A$$

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$d(tr(\mathbf{p},\omega_i),-\omega_i)\cos\theta_i\,\mathrm{d}\omega_i)$

Rendering Equation: Area Form

$$L(\mathbf{p}' \to \mathbf{p}) = L_e(\mathbf{p}' \to \mathbf{p}) + \int_A f_r(\mathbf{p}'' \to \mathbf{p}' \to \mathbf{p}) L(\mathbf{p}'' \to \mathbf{p})$$

Geometric coupling term $G(p'' \leftrightarrow p') = \frac{\cos \theta' \cos \theta'' V(p'' \leftrightarrow p')}{||p'' - p'||^2}$ Binary visibility function

Binary visibility function $V(p \leftrightarrow p')$

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$p') G(p'' \leftrightarrow p') dp''$



Recursive Expansion

$$L(\mathbf{p}' \to \mathbf{p}) = L_e(\mathbf{p}' \to \mathbf{p}) + \int_A f_r(\mathbf{p}'' \to \mathbf{p}' \to \mathbf{p}) L(\mathbf{p}'' \to \mathbf{p}') \mathbf{0}$$

$$L(\mathbf{p}' \to \mathbf{p}) = L_e(\mathbf{p}' \to \mathbf{p}) + \int_A f_r(\mathbf{p}'' \to \mathbf{p}' \to \mathbf{p}) \\ \begin{bmatrix} L_e(\mathbf{p}'' \to \mathbf{p}') + \int_A f_r(\mathbf{p}''' \to \mathbf{p}'' \to \mathbf{p}') \\ G(\mathbf{p}'' \leftrightarrow \mathbf{p}') d\mathbf{p}'' \end{bmatrix}$$

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$G(\mathbf{p}'' \leftrightarrow \mathbf{p}') \,\mathrm{d}\mathbf{p}''$

$L(\mathbf{p}''' \to \mathbf{p}'') G(\mathbf{p}''' \leftrightarrow \mathbf{p}'') d\mathbf{p}'''$

Recursive Expansion

$$L(\mathbf{p}' \to \mathbf{p}) = L_e(\mathbf{p}' \to \mathbf{p}) + \int_A f_r(\mathbf{p}'' \to \mathbf{p}' \to \mathbf{p}) \\ \begin{bmatrix} L_e(\mathbf{p}'' \to \mathbf{p}') + \int_A f_r(\mathbf{p}''' \to \mathbf{p}'' \to \mathbf{p}') \\ G(\mathbf{p}'' \leftrightarrow \mathbf{p}') d\mathbf{p}'' \end{bmatrix}$$

$$L(\mathbf{p}' \to \mathbf{p}) = L_e(\mathbf{p}' \to \mathbf{p}) + \int_A f_r(\mathbf{p}'' \to \mathbf{p}' \to \mathbf{p}) L_e(\mathbf{p}'' \to \mathbf{p}') G(\mathbf{p}'' \leftrightarrow \mathbf{p}') d\mathbf{p}'' + \int_A \int_A f_r(\mathbf{p}'' \to \mathbf{p}' \to \mathbf{p}) G(\mathbf{p}'' \leftrightarrow \mathbf{p}') f_r(\mathbf{p}''' \to \mathbf{p}'' \to \mathbf{p})$$

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$L(\mathbf{p}''' \to \mathbf{p}'') G(\mathbf{p}''' \leftrightarrow \mathbf{p}'') d\mathbf{p}'''$

$p') G(p''' \leftrightarrow p'') L(p''' \rightarrow p'') dp'' dp'''$

Rendering Equation: Sum Over Paths

$$L(\mathbf{p}_1 \to \mathbf{p}) = L_e(\mathbf{p}_1 \to \mathbf{p}) + \sum_i \left[\int_{A^i} T_i(\bar{\mathbf{p}}) L_e \right]$$
$$T_n(\bar{\mathbf{p}}) = \prod_i^n f_r(\mathbf{p}_{i+1} \to \mathbf{p}_i \to \mathbf{p}_{i-1}) G(\mathbf{p}_{i+1} \to \mathbf{p}_i)$$



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$e(\mathbf{p}_{i+1} \to \mathbf{p}_i) \,\mathrm{dp}_{i+1} \dots \,\mathrm{dp}_2 \bigg|$

 $_{-1} \leftrightarrow \mathbf{p}_i$

Light Tracing



Splat to the image at each vertex

Wojciech Jarosz

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Path Tracing, 32 samples / pixel



Light Tracing, 32 samples / pixel



Bidirectional Path Tracing (BDPT)

Start paths from camera and lights Connect path vertices with shadow rays Can handle difficult light sampling situations more robustly

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Bidirectional Path Tracing (BDPT)



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Bidirectional Path Tracing (BDPT)

- Provides a family of approaches for generating paths of length n
 - I camera, n-1 light
 - 2 camera, n-2 light

Key: apply multiple importance sampling to reweighs path contributions

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Path Tracing, 32 samples / pixel



Light Tracing, 32 samples / pixel



BDPT, 32 samples / pixel







Bidirectional Path Tracing

[Veach and Guibas 1995]

Path Tracing

Path Pyramid

















Challenges Remain





Challenges Remain

Reference



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Challenges Remain



Wojciech Jarosz

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Photon Mapping

Trace particles from lights

- Interpolate *nearby* samples when computing reflected radiance
- Key ideas:
 - Particle histories give scene radiance distribution
 - Illumination information close to the point being shaded is generally applicable



Photon Distribution

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Radiance Estimate

Density estimate:

- Take a small spherical neighborhood containing k photons
- Count the power of the photons
- Compute the enclosed surface area
- Power / Area = density



Radiance Estimate

$$L(\mathbf{p}, \omega_o) = \int_{H^2} f_r(\omega_i \to \omega_o) L_i(p, \omega_i)$$
$$= \int_{H^2} f_r(\omega_i \to \omega_o) \frac{\mathrm{d}^2 \Phi(\mathbf{p})}{\mathrm{d}\omega_i \cos \theta}$$
$$= \int_{H^2} f_r(\omega_i \to \omega_o) \frac{\mathrm{d}^2 \Phi(\mathbf{p}, \theta)}{\mathrm{d}A}$$
$$\approx \sum_i^N f_r(\omega_i^{-1}) = \omega_o^{-1} \frac{\Delta \Phi(\mathbf{p}, \theta)}{\pi r^2}$$

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 $(z) \cos \theta_i \mathrm{d}\omega_i$ $\frac{(\mathbf{p},\omega_i)}{\mathrm{vs}\,\theta_i\,\mathrm{d}A}\,\cos\theta_i\,\mathrm{d}\omega_i$ $\omega_i)$

 $,\omega[i])$

Unbiased vs. Consistent Estimators



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Unbiased vs. Consistent Estimators

Graphics interpretation:

- Consistent: image approaches correct solution as some parameter is increased
- Unbiased: produces correct answer on average

Value of biased estimators: May have lower variance May look better (less noise)

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Radiance Estimate

$$L(\mathbf{p}, \omega_o) = \int_{H^2} f_r(\omega_i \to \omega_o) L_i(p, \omega_i)$$
$$= \int_{H^2} f_r(\omega_i \to \omega_o) \frac{\mathrm{d}^2 \Phi(\mathbf{p})}{\mathrm{d}\omega_i \cos \theta}$$
$$= \int_{H^2} f_r(\omega_i \to \omega_o) \frac{\mathrm{d}^2 \Phi(\mathbf{p}, \theta)}{\mathrm{d}A}$$
$$\approx \sum_i^N f_r(\omega[i] \to \omega_o) \frac{\Delta \Phi(\mathbf{p}, \theta)}{\pi r^2}$$

Biased, but consistent

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 $(z) \cos \theta_i \mathrm{d}\omega_i$ $\frac{\mathbf{p}, \omega_i}{\mathbf{s}\,\theta_i\,\mathrm{d}A}\,\cos\theta_i\,\mathrm{d}\omega_i$ $\omega_i)$

 $,\omega[i])$



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Zhu et al., 2020. Deep Kernel Destiny Estimation for Photon Mapping.

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Standard Photon Mapping

- Store all photons in memory
- Trace path from camera, lookup photons at each path vertex ("pull")
- Memory required ~ # photons



Progressive Photon Mapping

- Store visible points in memory
- Trace one photon at a time, splat to relevant points ("push")
- Memory required ~ # visible points

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Progressive Photon Mapping

- 1. Trace rays from camera, find a VisiblePoint at each pixel
- 2. Store VisiblePoints in a 3D spatial structure that allows fast lookups
- 3. Trace photons from lights; at intersections search for nearby VisiblePoints and contribute illumination
- 4. Decrease acceptance radius as more photons contribute

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struct VisiblePoint { Point3f p; Vector3f wo; const BSDF *bsdf; Spectrum beta; };



SPPM, 1 Iteration

Contraction of the second

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SPPM, 4 Iterations

10

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SPPM, 32 Iterations

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1.5

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SPPM, 256 Iterations



SPPM, 4096 Iterations





BDPT, 4096 spp



SPPM, 4096 Iterations

