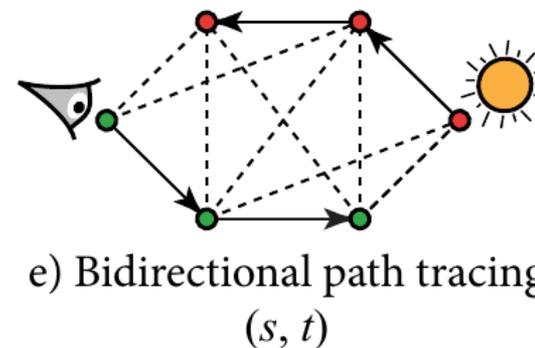
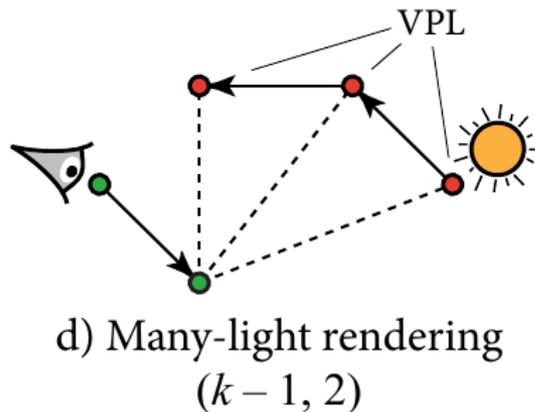
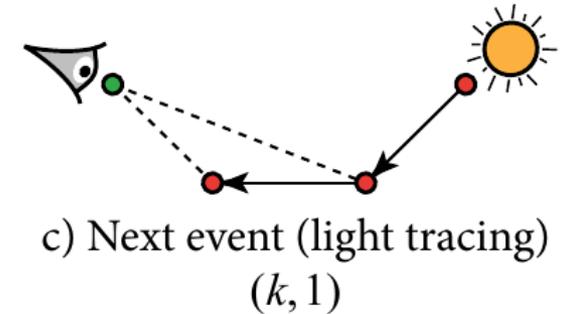
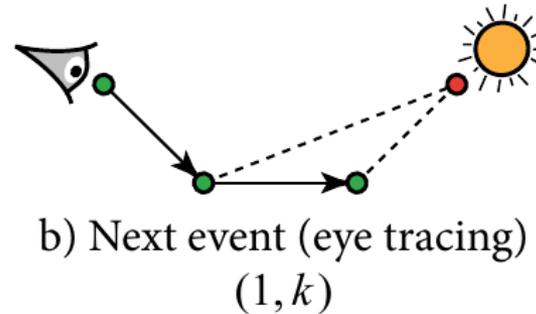
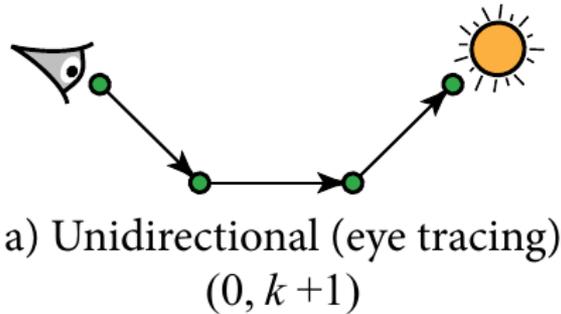

Realistic Image Synthesis

Bidirectional Path Tracing & Reciprocity

Philipp Slusallek
Karol Myszkowski
Gurprit Singh

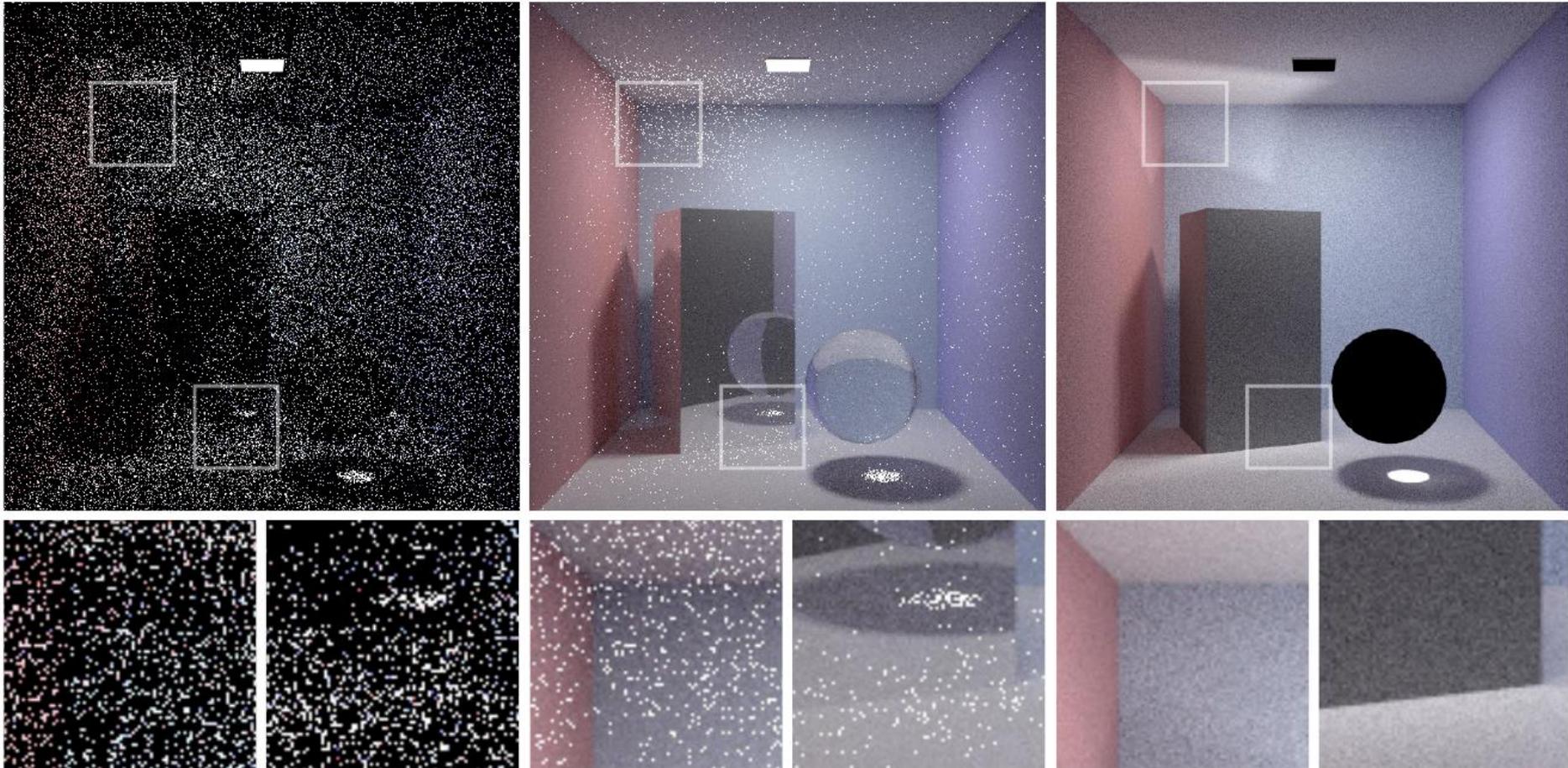
Path Sampling Techniques

● light subpath vertex
● eye subpath vertex



- **Different techniques of sampling paths from both sides**
 - Numbers in parenthesis are # of vertices traced from light/camera, resp.
 - See later, for many light methods

Results from Different Techniques



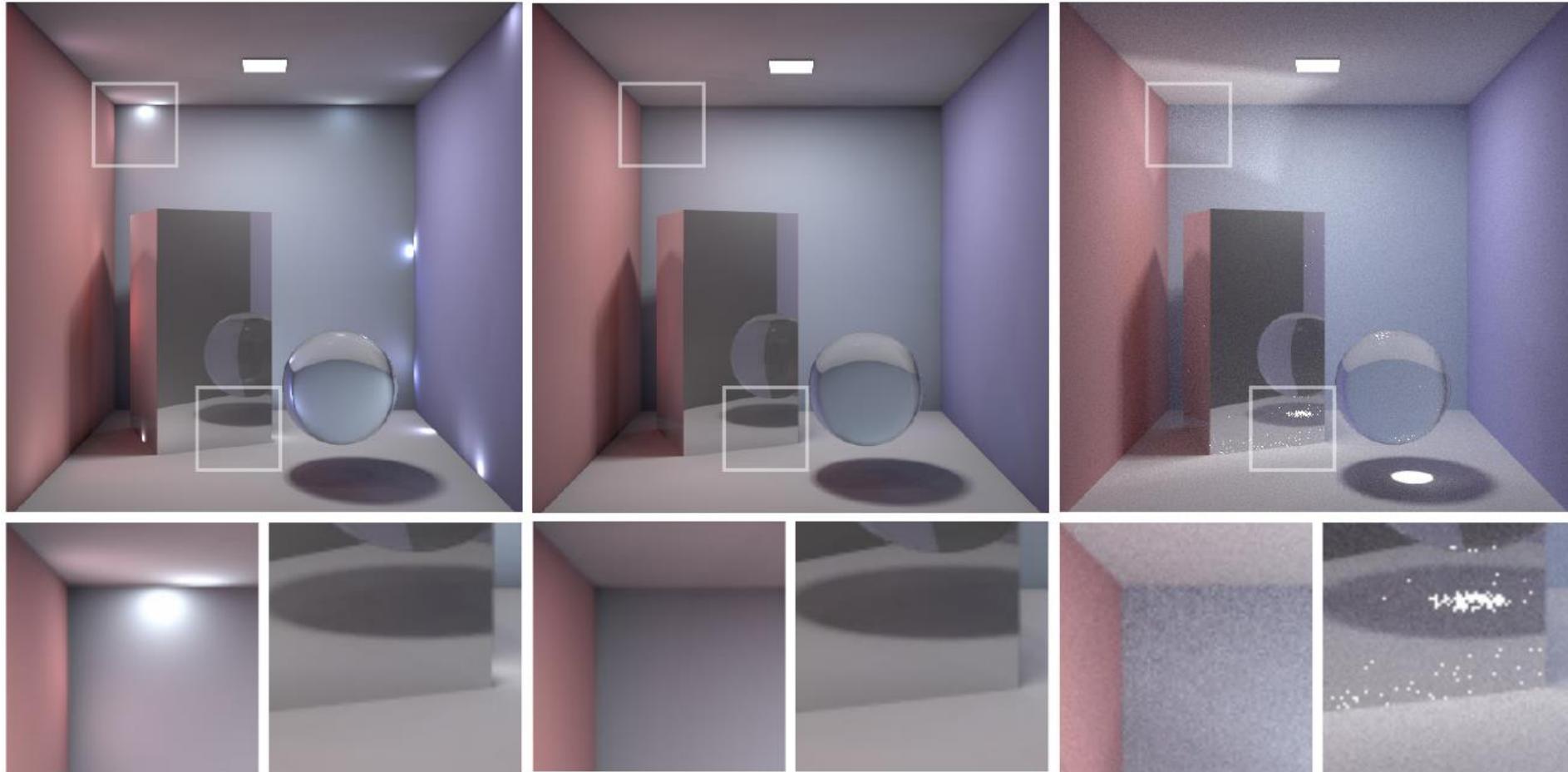
a) Unidirectional (eye tracing)

b) Unidirectional + next event

c) Next event (light tracing)

- **Results from tracing 40 paths per pixel**

Results from Different Techniques



d) Instant radiosity

e) Instant radiosity (clamped)

f) Bidirectional path tracing

- **Results from tracing 40 paths per pixel**

- f): „Problem of insufficient techniques“ for sampling SDS paths

BIDIRECTIONAL PATH TRACING

Light & Path Tracing

- **Problem:**

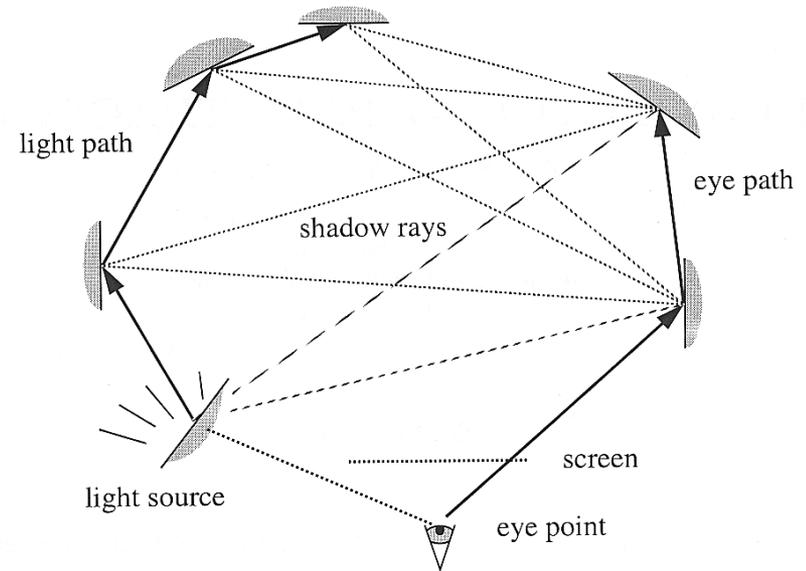
- Probability of hitting the camera from the light sources is almost zero
- Probability of hitting the light source is often also very small
 - Next Event Estimator: Try to find direct connections
 - Non optimal (e.g. on mirror surface)
 - Ignores secondary light sources (e.g. via mirror, at caustics)

- **Approaches:**

- Bidirectional Path Tracing
 - Combination of eye and light paths
 - Weighted MC sampling for best results
 - Includes Vertex Connection and Merging (VCM, later)
- Metropolis-Sampling [Veach '1997] (see later)
 - Random variation and mutations of bidirectional paths
 - Very well suited for very complex light paths
 - Unbiased but complex algorithms
 - Uneven convergence

Bidirectional Path-Tracing

- **Idea: Combine Paths from Both Sides**
 - Generate path from the light sources and the camera
 - Connect paths deterministically (every pair of two hit points)
 - Different probabilities of generating paths
 - Compute weighted sum of contributions
- **References:**
 - **Lafortune et al.**, Bidirectional Path-Tracing, [CompuGraphics`93]
 - **Veach, Guibas**, Bidirectional Estimators for LightTransport, [EGRW`94, Siggraph`95]

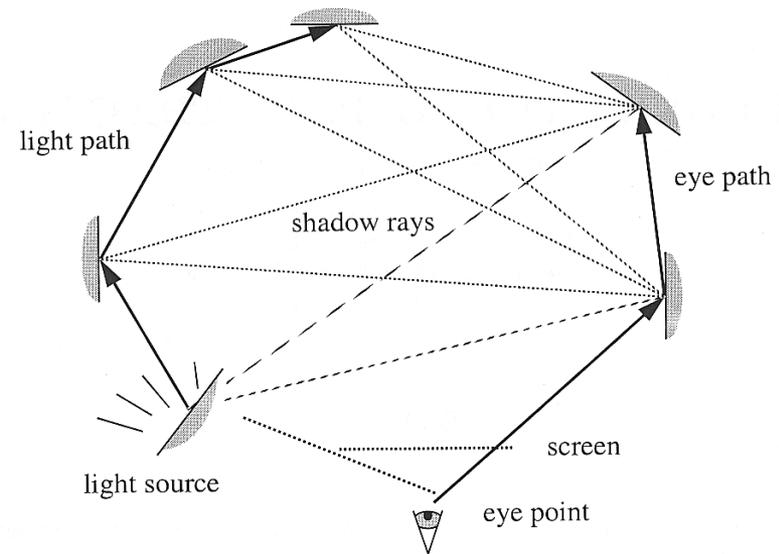


Solving the Rendering Equation

- **Von Neumann Expansion of Measurement Equation**

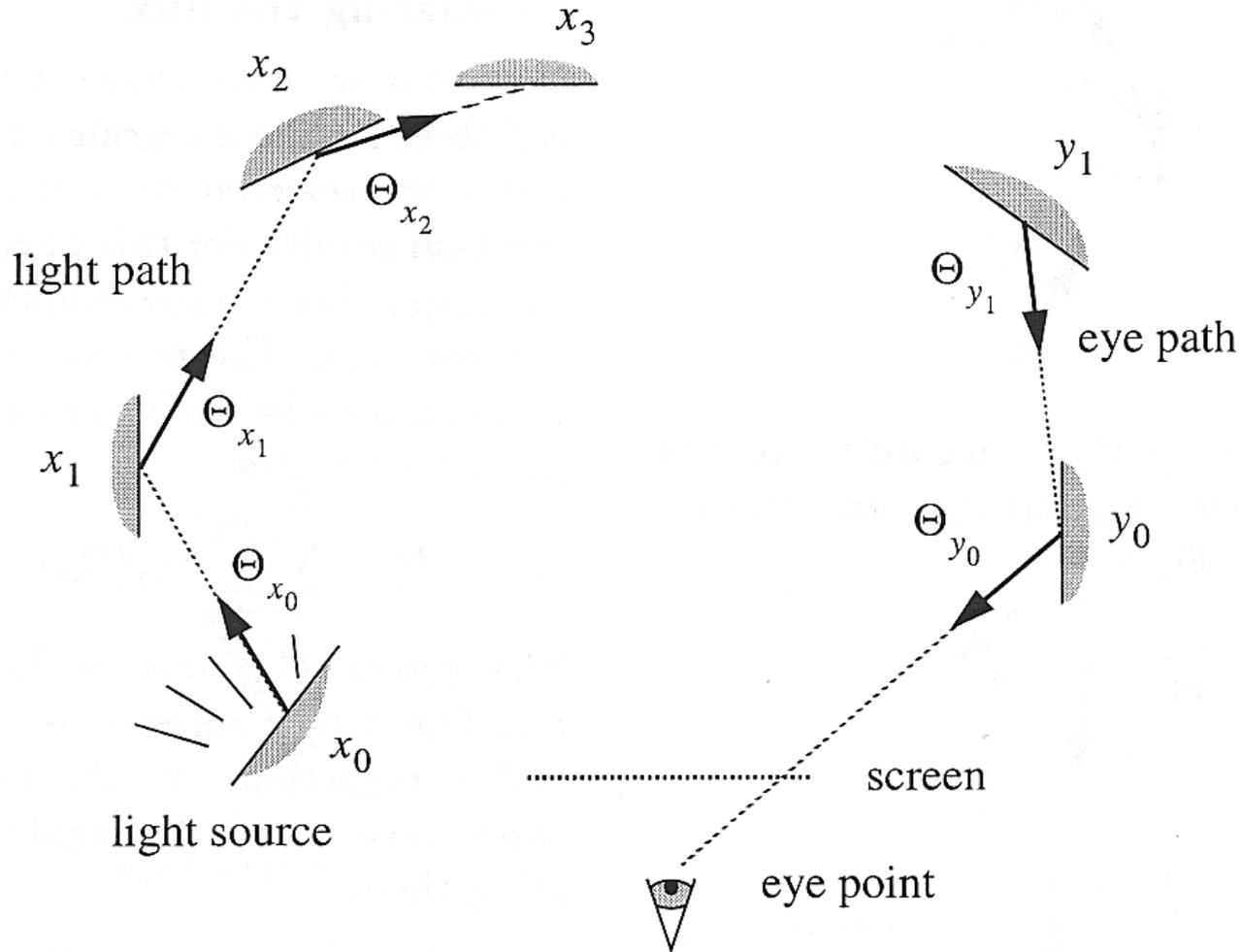
$$I_p = \int_{S \times S} L_e(x \rightarrow x') G(x \rightarrow x') W_p(x \rightarrow x') dA(x) dA(x') +$$
$$+ \int_{S \times S \times S} L_e(x \rightarrow x') G(x \rightarrow x') f_r(x \rightarrow x' \rightarrow x'') G(x' \rightarrow x'') W_p(x' \rightarrow x'')$$
$$\rightarrow x'') dA(x) dA(x') dA(x'') + \dots \quad \text{with } G(x, y) = \frac{\cos\theta_x \cos\theta_y}{\|x - y\|^2}$$

- Independent estimation of all paths with fixed lengths
- Bidirectional generation of paths
- Weighted MC integration for each term (MIS)
- More efficient by reusing costly paths (i.e. visibility samples) multiple times
- Typically: One pair of paths per pixel sample



Bidirectional Path-Tracing

- Notation



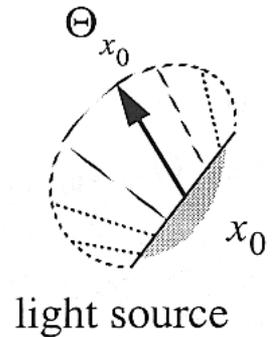
Bidirectional Path-Tracing

- **Generating Light Paths**

- On the light source

$$p(x, \Theta_x) = \frac{L_e(x, \Theta_x) \|\Theta_x \cdot N_x\|}{\Phi}$$

$$\Phi = \int_A \int_{\Omega_+} L_e(x, \Theta_x) \|\Theta_x \cdot N_x\| d\Theta_x dA_x$$

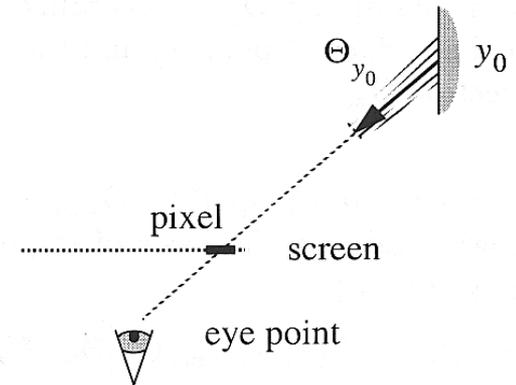


- **Generating Eye Paths**

- On the eye/camera (via point in the scene)

$$p(y, \Theta_y) = \frac{g(y, \Theta_y) W(y, \Theta_y) \|\Theta_y \cdot N_y\|}{G}$$

$$G = \int_A \int_{\Omega_+} g(y, \Theta_y) W(y, \Theta_y) \|\Theta_y \cdot N_y\| d\Theta_y dA_y$$



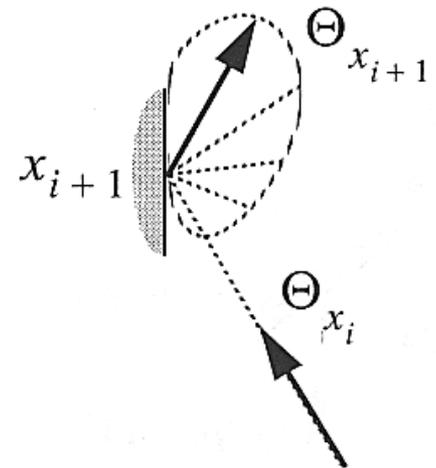
- $g()$: 1, if point is visible in this direction

Bidirectional Path-Tracing

- **Extension of Paths at Hit Points**

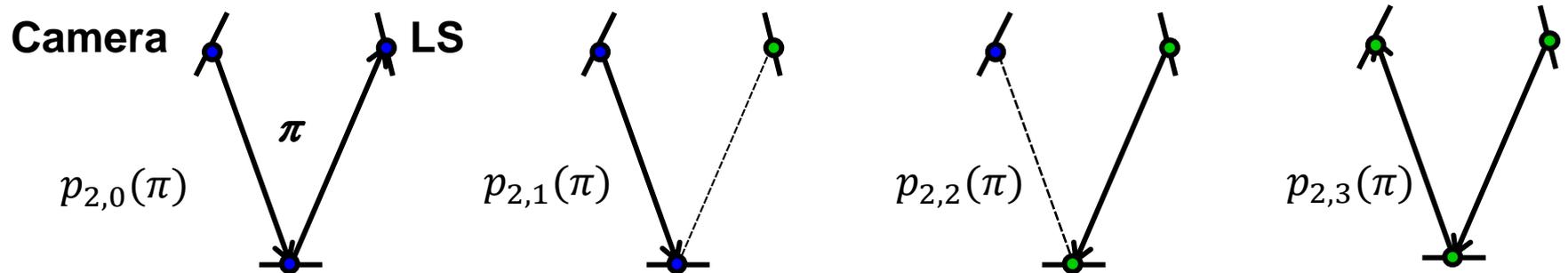
- Identical for both directions
 - Reciprocity of BRDF under reflection
- Use whatever BRDF-sampling technique suits best
 - But must be a joint probability (conditioned on the previous point)
 - This does include uniform probability on any surface
 - (But not a point generated from some other point, e.g. due to occl.)
 - E.g.

$$p(\Theta) = f_r(\Theta_{x_i}, x_{i+1}, \Theta) \parallel \Theta_{x_{i+1}} \cdot N_{x_{i+1}} \mid$$



Bidirectional Probabilities

- **Probabilities of Paths π in Bidirectional Path Tracing**
 - Different locations of *vertex connections* (see VCM later)
 - k : length of paths (# of transports or segments)
 - m : # of vertices generated from light source
 - 0: None
 - 1: Vertex on light source
 - 2: Vertex on light source and directional sample
 - Etc.
 - Similar for paths from the eye
 - $p_{k,m}(\pi)$: Probability to choose path π with method (k,m)

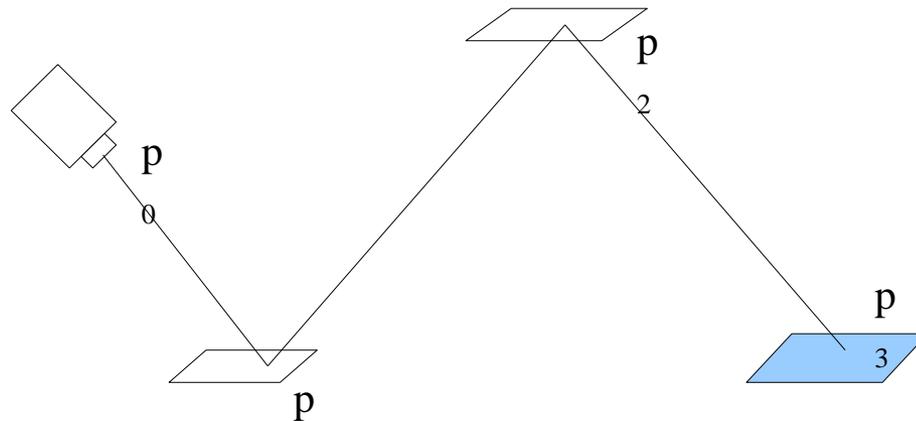


Mathematical Formulation

- **Rendering Equation with Area Parametrization**

$$L(p_1 \rightarrow p_0) = L_e(p_1 \rightarrow p_0) + \int_A L_e(p_2 \rightarrow p_1) f(p_2 \rightarrow p_1 \rightarrow p_0) G(p_2 \rightarrow p_1) dA(p_2) + \int_A \int_A L_e(p_3 \rightarrow p_2) f(p_3 \rightarrow p_2 \rightarrow p_1) G(p_3 \rightarrow p_2) f(p_2 \rightarrow p_1 \rightarrow p_0) G(p_2 \rightarrow p_1) dA(p_3) dA(p_2) + \dots$$

with $G(p_2 \rightarrow p_1) = V(p_1, p_2) \frac{\cos(\theta_{p_1}) \cos(\theta_{p_2})}{|p_2 - p_1|^2}$



Mathematical Formulation

- **Path Formulation**

- π_i : Path of length i $L(p_1 \rightarrow p_0) = \sum_{i=1}^{\infty} L(\pi_i(p_1, p_0)) = \sum_{i=1}^{\infty} L(\pi_i)$

$$L(\pi_i) = \underbrace{\int_A \int_A \cdots \int_A}_{i-1} L_e(p_i \rightarrow p_{i-1}) \left(\prod_{j=1}^{i-1} G(p_{j+1} \rightarrow p_j) f(p_{j+1} \rightarrow p_j \rightarrow p_{j-1}) \right) dA(p_2) \cdots dA(p_i)$$

- **Connection Throughput $T(\pi)$ of a path π**

$$T(\pi_i) = \prod_{j=1}^{i-1} G(p_{j+1} \rightarrow p_j) f(p_{j+1} \rightarrow p_j \rightarrow p_{j-1})$$

$$L(\pi_i) = \underbrace{\int_A \int_A \cdots \int_A}_{i-1} L_e(p_i \rightarrow p_{i-1}) T(\pi_i) dA(p_2) \cdots dA(p_i)$$

- **With Measurement**

$$I = \int_A \int_{A_{\text{pixel}}} L(p_1 \rightarrow p_0) G(p_1 \rightarrow p_0) W(p_1 \rightarrow p_0) dA(p_0) dA(p_1)$$

$$I = \sum_i \underbrace{\int_A \int_A \cdots \int_A}_{i+1} L_e(p_i \rightarrow p_{i-1}) T(\pi_i) G(p_1 \rightarrow p_0) W(p_1 \rightarrow p_0) dA(p_0) \cdots dA(p_i)$$

Mathematical Formulation

- **Path Tracing with Russian Roulette**

$$L(p_1 \rightarrow p_0) = \sum_{i=1}^{\infty} L(\pi_i) = L(\pi_1) + \frac{1}{1-q_2} \sum_{i=2}^{\infty} L(\pi_i)$$

- And similar for higher path lengths

- **How to choose the probabilities of sample points**

- Whatever works, from wherever (!!!), e.g.

- Area (uniform):

$$P_A(p_i) = \frac{1}{\sum_j A_j}$$

- Solid angle, depending on direction from previous sample:

$$P_A = P_\omega \frac{\cos \theta_i}{r^2}$$

- Any other joint probability that integrates to one over all surfaces and is non-zero where there could be a contribution
- Must be a conditional probability

- **Splitting of BRDFs or Emissions**

- Make sure all paths are accounted for !
- Make sure no path is counted multiple times, either !

Example

- Light tracing (one eye ray, 1st generation only)



Example

- **Standard MC Path Tracing (same number of paths)**



Example

• Contribution of Different Paths

[Not shown: direct connection from eye to light]

One reflection

One step from the eye (plus direct connection to light)



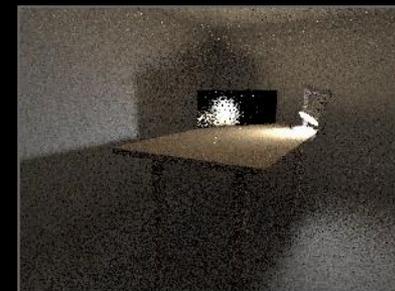
Two reflections

l: Two steps from the eye
r: One step from the eye, one step from light source



Three reflections

l: Three steps from the eye
m: two steps from the eye, one from light source
r: one step from the eye, two from the light sources



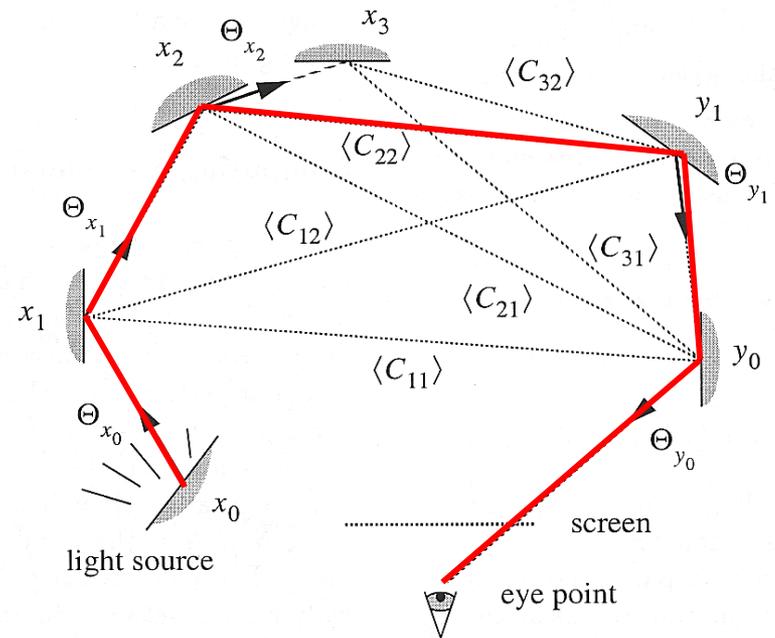
Bidirectional Path-Tracing

- **Combination of Estimators**

- Every option of generating a specific path π defines its own estimator with given $p_{k,m}(\pi)$
- Weighted MC sampling provides new combined estimator of a bi-directionally generated path

$$\bar{C} = \sum_{i=1}^{N_e} \sum_{j=1}^{N_l} w_{ij} \langle C_{ij} \rangle$$

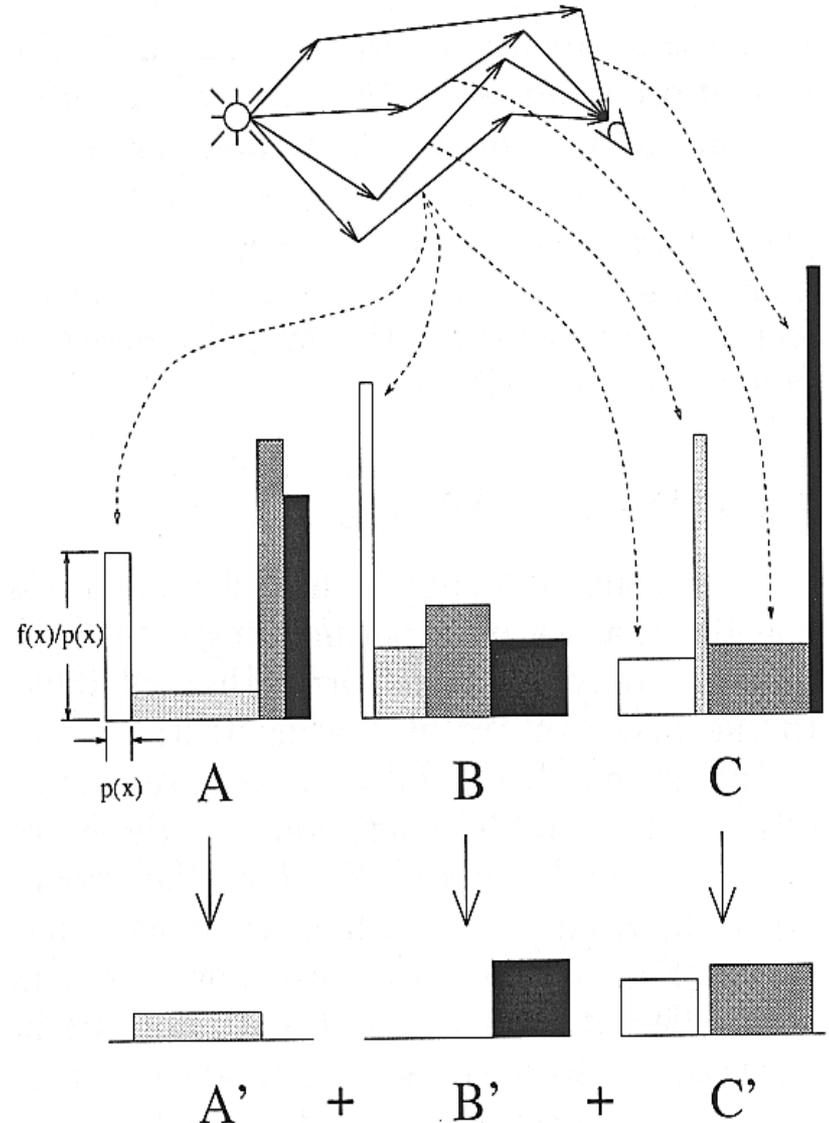
- N_e : # reflections on eye paths
- N_l : # reflections on light paths
- w_{ij} : weights for combination



Combination of Estimators

- **Example:**

- Four paths between LS and eye
- Weighted with three estimators
 - A, B, C
- Selection with maximum heuristics
 - Choose $p_X(\pi)$ maximum
- Area of rectangles is constant across A, B, C
 - $f/p * p$
- Width corresponds to $p_X(\pi)$



Implementation

Example: Maximums Heuristics

$S = 0$

$P = \text{GenerateBiDirPaths}()$

for $\text{light_segs} = 0$ to $P.\text{max_light_segments}$

for $\text{eye_segs} = 0$ to $P.\text{max_eye_segments}$

$SP = \text{ChooseSubPath}(P, \text{eye_segs}, \text{light_segs})$

// Compute best estimator (Max-Heuristics)

$p = 0$; $\text{segments} = \text{eye_segs} + \text{light_segs}$;

// Iterate over different estimators:

// assuming j segments generated

// from camera

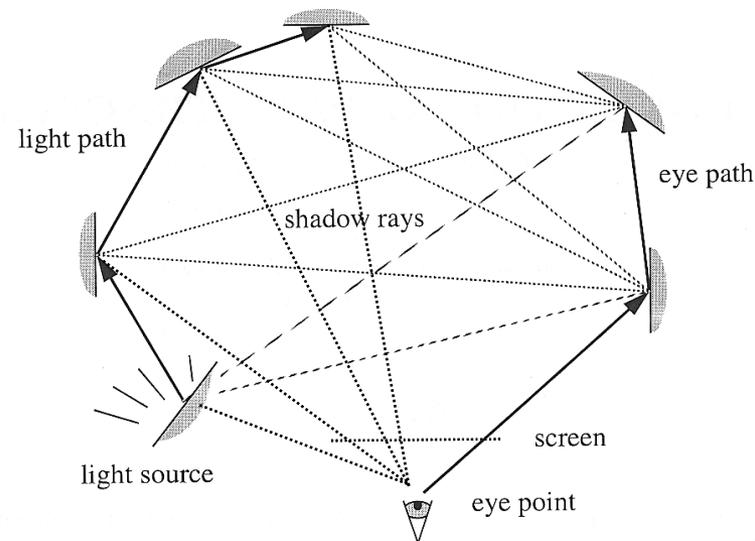
for $\text{estimator} = 0$ to segments

$p_t = \text{Probability}(SP, \text{estimator})$

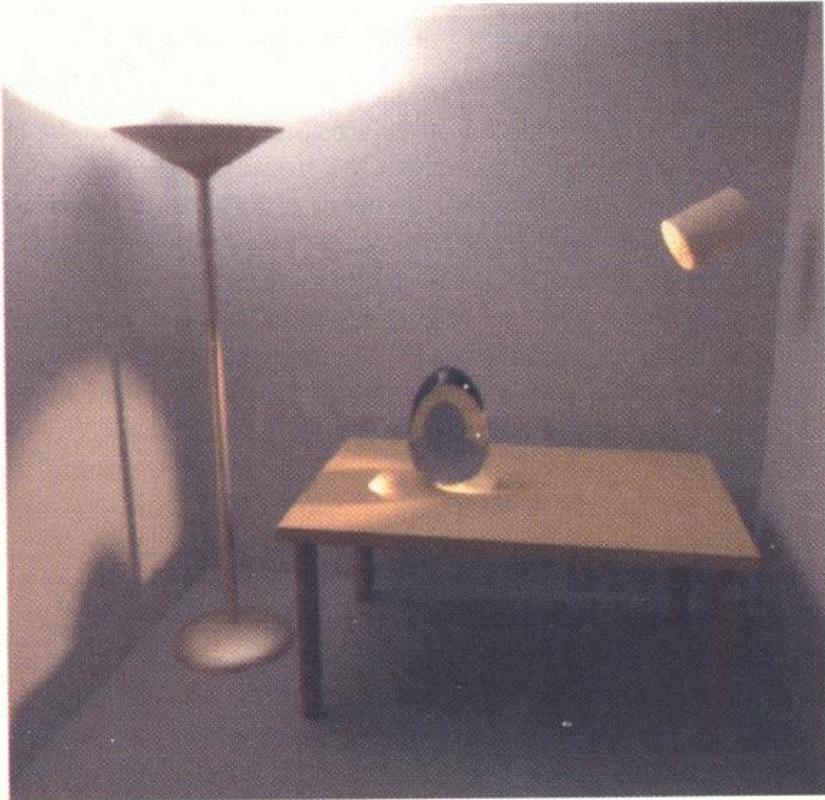
if $(p_t > p)$ $p = p_t$

$S = S + SP.f/p$

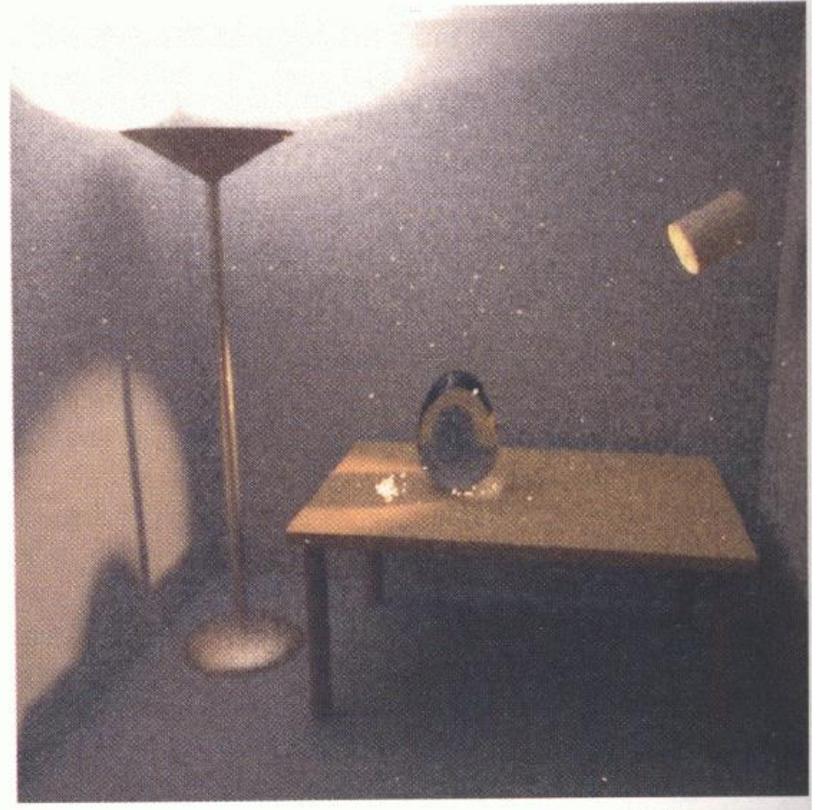
return S



Example



Bidirectional Path Tracing

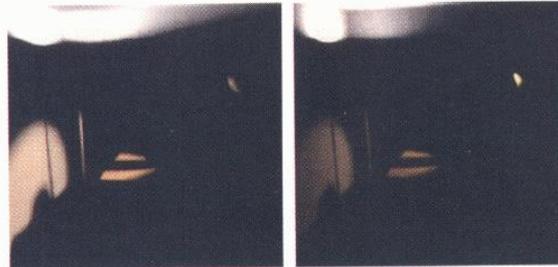


Path Tracing

Contributions of Different Paths

More camera segments

$p_{2,x}$

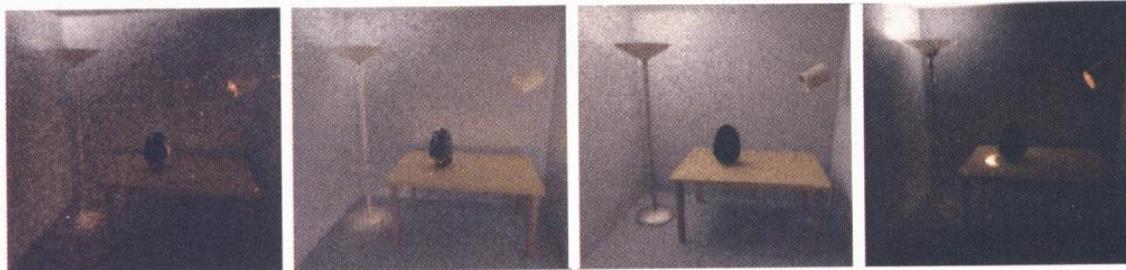


More light segments
(right: n-1)

$p_{3,x}$



$p_{4,x}$



$p_{5,x}$



Comparison w/ Path Tracing

- **Brute Force Method**

- Only use $p_{n,0}$ method to generate paths
 - No points sampled from light source
- Highly inefficient:
 - Probability of hitting the light is almost zero
 - Especially for point lights :-)

- **Path Tracing with Direct Lighting Optimization**

- Next Event Estimation
- Use $p_{n,0}$ and $p_{n,1}$ paths only
 - Path from the eye/camera plus direct connection to point sampled on light source

NON-SYMMETRIC SCATTERING IN LIGHT TRANSPORT ALGORITHMS

Use of Shading Normals

- **Shading Normals**

- It is common to shade with respect to arbitrary normals
 - E.g. specified as normals at each triangle vertex
- Allow many neat tricks
 - Smooth surface even though real surface is tessellated
 - Bump mapping

- **Problem**

- Use of shading normals θ' is generally not energy conserving

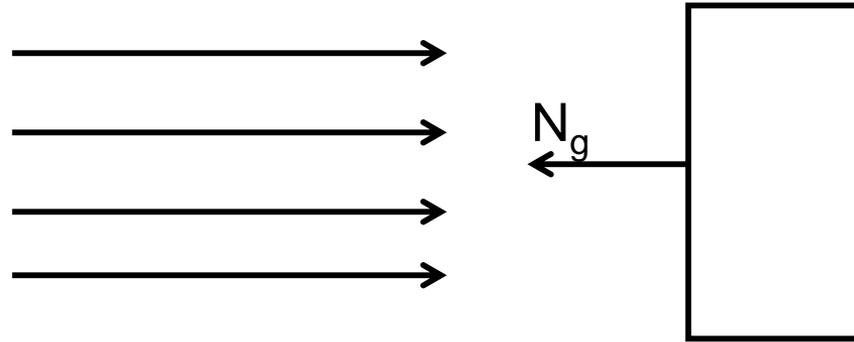
$$\begin{aligned} L_r &= \int_{\Omega_+} f_r(\omega_o, x, \omega_i) \cos\theta_i d\omega_i \\ &= \int_{\Omega_+} f_r'(\omega_o, x, \omega_i) \underbrace{\frac{\cos\theta_i'}{\cos\theta_i}}_{\text{can be arbitrarily large}} \cos\theta_i d\omega_i \end{aligned}$$

- Can “generate” energy

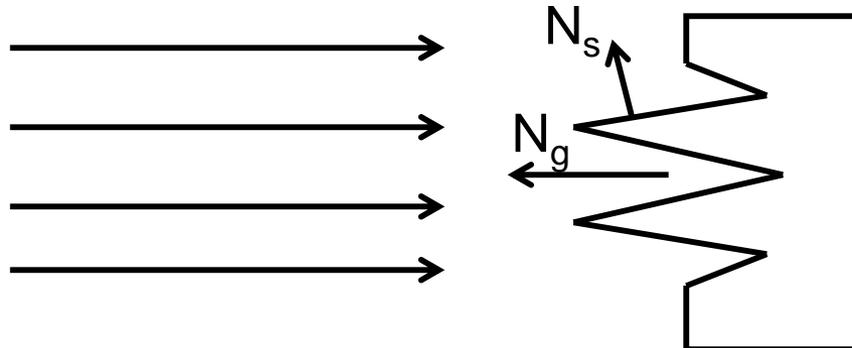
Use of Shading Normals

- **Energy “Generator”**

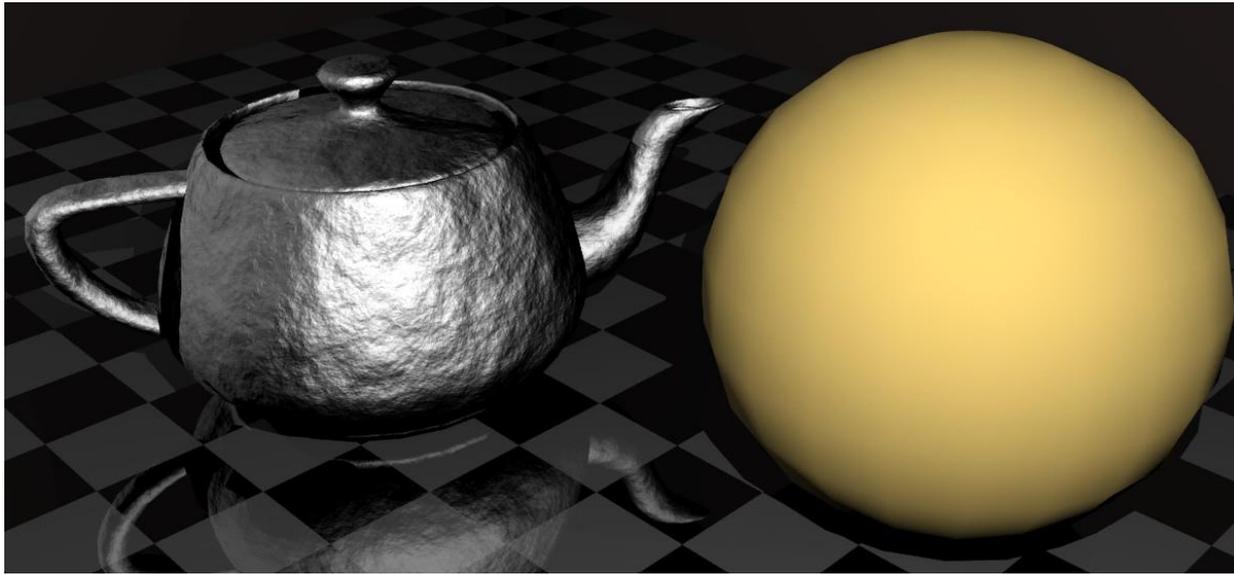
- Light is received by an apparently small surface



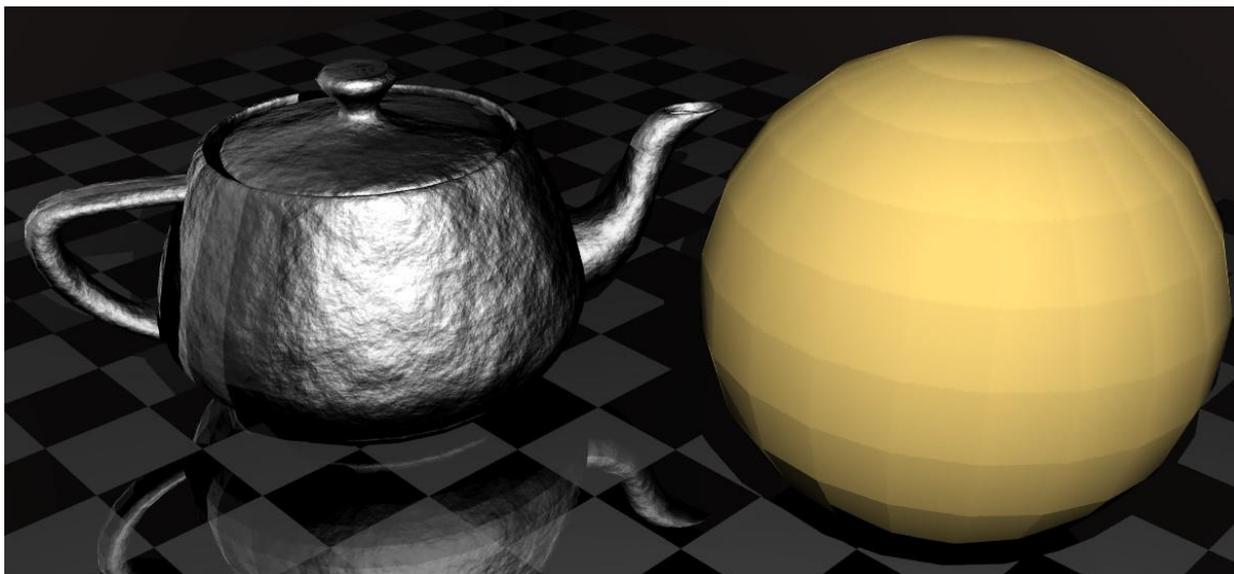
- And emitted from an apparently much larger one



Use of Shading Normals



Correct results



Wrong results

Use of Shading Normals

- **Solution**

- Unfortunately there seems to be no good solution to the problem
- Except not using shading normals :-(

Power versus Radiance

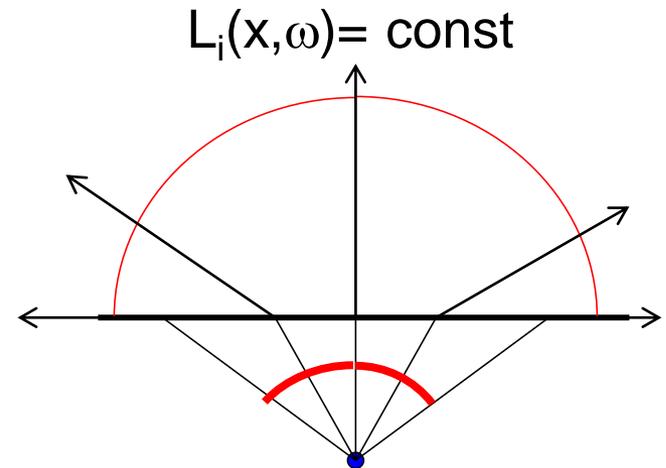
- **Light tracing and Refraction**

- Distribution of “photons” carrying a certain energy/power
- Power/energy does not change when photon is refracted

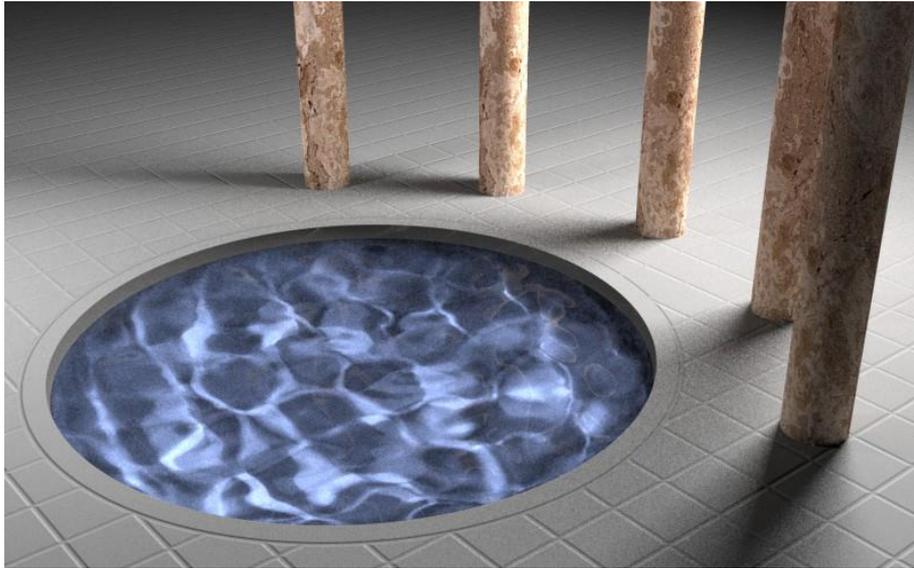
- **Ray Tracing and Refraction**

- Consider
 - uniform illumination
 - a point below a refracting surface
- If no light is absorbed at the surface then the same power comes through a smaller solid angle
→ increased radiance

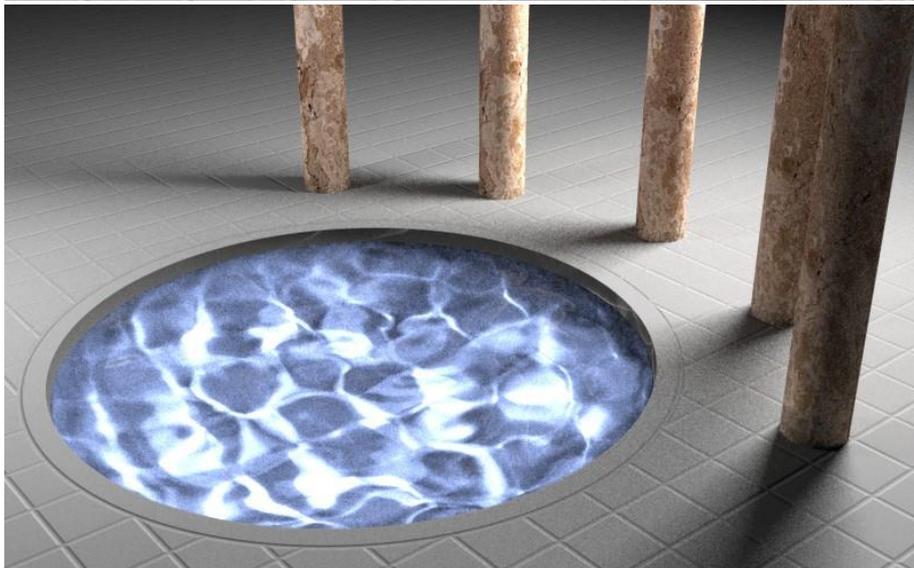
$$L_t = \frac{\eta_t^2}{\eta_i^2} L_i$$



Power versus Radiance



Correct image rendered with particle tracing



Incorrect image rendered assuming the BRDF is symmetric also for refraction