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# CS482: Monte Carlo Ray Tracing:

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Sung-Eui Yoon  
(윤성의)

<http://sglab.kaist.ac.kr/~sungeui/ICG>

**KAIST**



# Class Objectives

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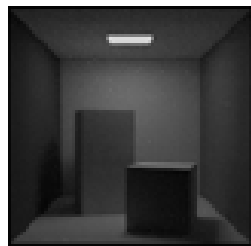
- **Understand a basic structure of Monte Carlo ray tracing**
  - **Russian roulette for its termination**
  - **Path tracing**

# Rendering Equation

$$L(x \rightarrow \Theta) = L_e(x \rightarrow \Theta) + \int_{\Omega_x} f_r(\Psi \leftrightarrow \Theta) \cdot L(x \leftarrow \Psi) \cdot \cos(\Psi, n_x) \cdot d\omega_\Psi$$

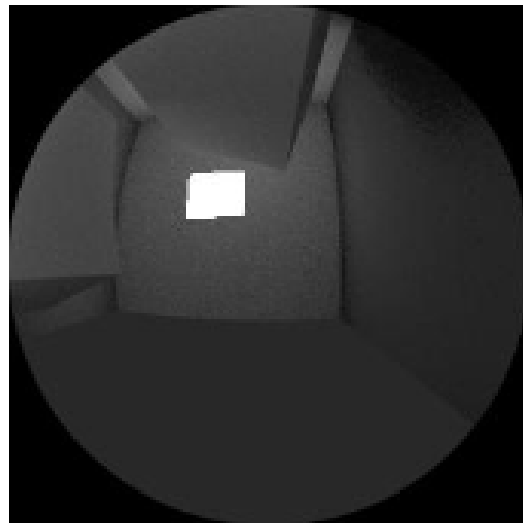


Value we want



$$= L_e + \int_{\Omega_x} \text{[Hemisphere Image]} \cdot f_r \cdot \cos$$

function to integrate over all incoming directions over the hemisphere around x



# How to compute?

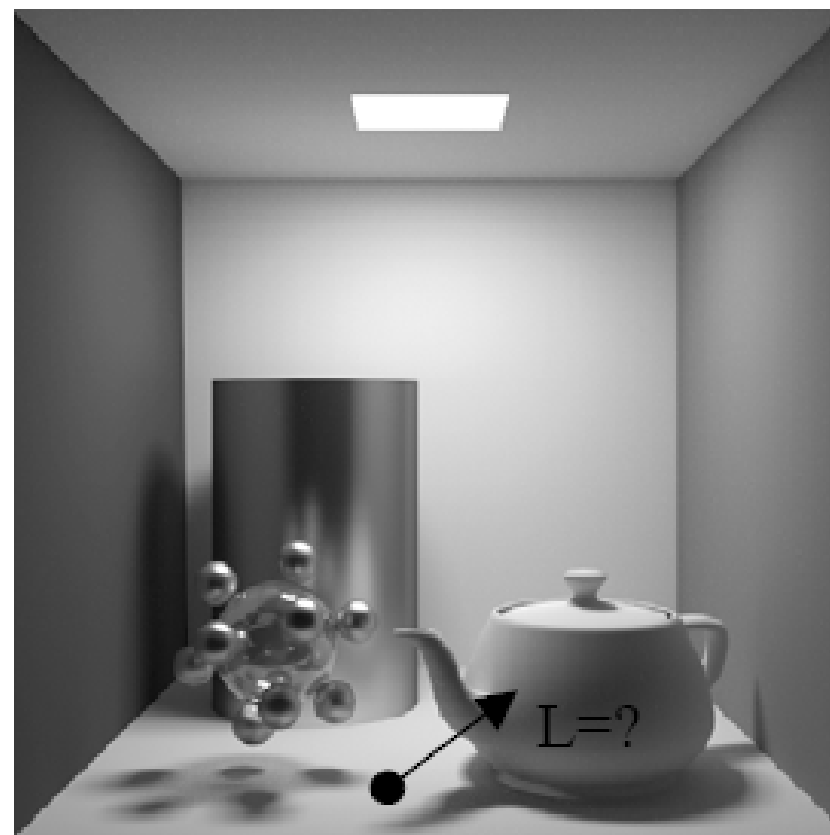
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$$L(x \rightarrow \Theta) = ?$$

Check for  $L_e(x \rightarrow \Theta)$

Now add  $L_r(x \rightarrow \Theta) =$

$$\int_{\Omega_x} f_r(\Psi \leftrightarrow \Theta) \cdot L(x \leftarrow \Psi) \cdot \cos(\Psi, n_x) \cdot d\omega_\Psi$$



# How to compute?

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- Use Monte Carlo
- Generate random directions on hemisphere  $\Omega_x$  using pdf  $p(\Psi)$

$$L(x \rightarrow \Theta) = \int_{\Omega_x} f_r(\Psi \leftrightarrow \Theta) \cdot L(x \leftarrow \Psi) \cdot \cos(\Psi, n_x) \cdot d\omega_\Psi$$

$$\langle L(x \rightarrow \Theta) \rangle = \frac{1}{N} \sum_{i=1}^N \frac{f_r(\Psi_i \leftrightarrow \Theta) \cdot L(x \leftarrow \Psi_i) \cdot \cos(\Psi_i, n_x)}{p(\Psi_i)}$$

# How to compute?

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Generate random directions  $\Psi_i$

$$\langle L \rangle = \frac{1}{N} \sum_{i=1}^N \frac{f_r(\dots) \cdot L(x \leftarrow \Psi_i) \cdot \cos(\dots)}{p(\Psi_i)}$$

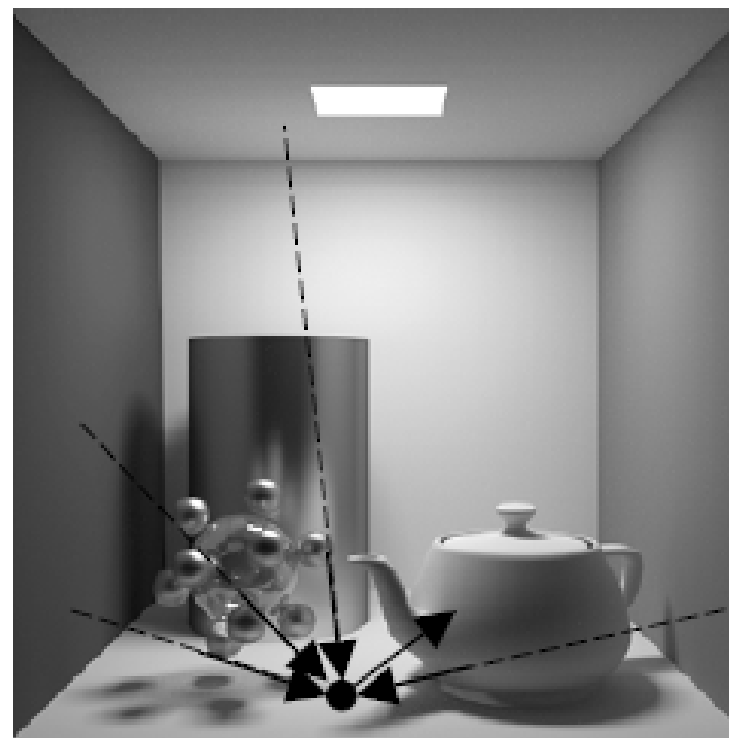
- evaluate brdf
- evaluate cosine term
- evaluate  $L(x \leftarrow \Psi_i)$



# How to compute?

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- evaluate  $L(x \leftarrow \Psi_i)$ ?
- Radiance is invariant along straight paths
- $vp(x, \Psi_i) =$  first visible point
- $L(x \leftarrow \Psi_i) = L(vp(x, \Psi_i) \rightarrow \Psi_i)$



# How to compute? Recursion ...

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- Recursion ....
- Each additional bounce adds one more level of indirect light
- Handles ALL light transport
- “Stochastic Ray Tracing”

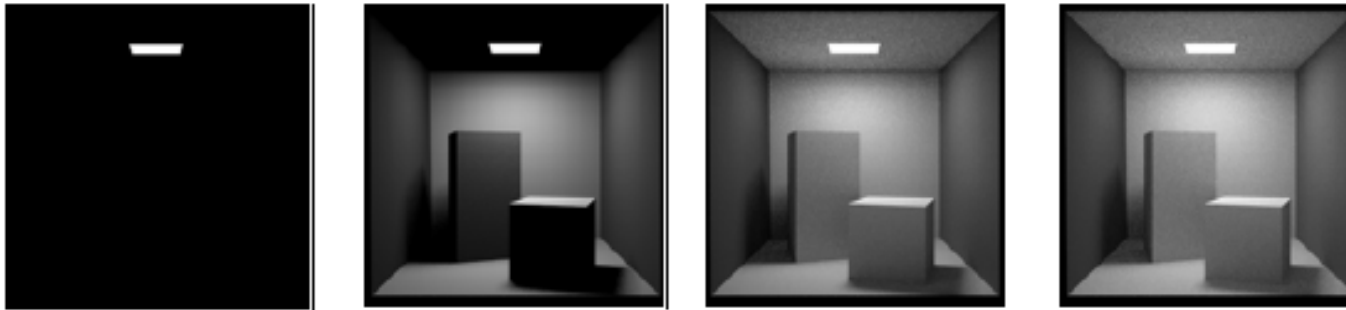




# When to end recursion?

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From kavita's slides

- **Contributions of further light bounces become less significant**
  - Max recursion
  - Some threshold for radiance value
- **If we just ignore them, estimators will be biased**

# Russian Roulette

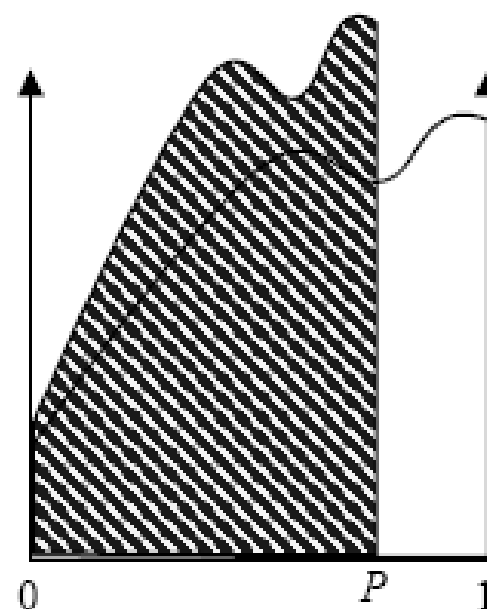
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Integral

$$I = \int_0^1 f(x) dx = \int_0^1 \frac{f(x)}{P} P dx = \int_0^P \frac{f(y/P)}{P} dy$$

Estimator

$$\langle I_{\text{roulette}} \rangle = \begin{cases} \frac{f(x_i)}{P} & \text{if } x_i \leq P, \\ 0 & \text{if } x_i > P. \end{cases}$$



Variance

$$\sigma_{\text{roulette}} > \sigma$$

# Russian Roulette

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- **Pick absorption probability,  $\alpha = 1-P$** 
  - Recursion is terminated
- **$1-\alpha$ , i.e.,  $P$ , is commonly to be equal to the reflectance of the material of the surface**
  - Darker surface absorbs more paths

# Algorithm so far

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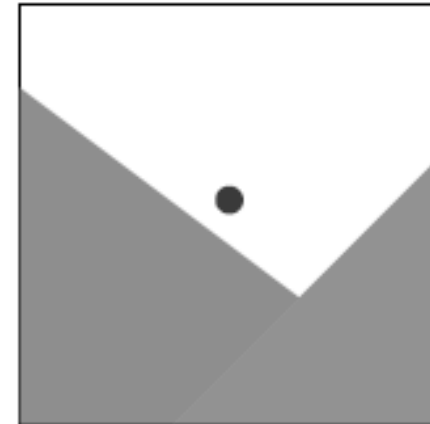
- **Shoot primary rays through each pixel**
- **Shoot indirect rays, sampled over hemisphere**
- **Terminate recursion using Russian Roulette**

# Pixel Anti-Aliasing

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- **Compute radiance only at the center of pixel**
  - Produce jaggies
- **We want to evaluate using MC**
- **Simple box filter**
  - The averaging method



# Stochastic Ray Tracing

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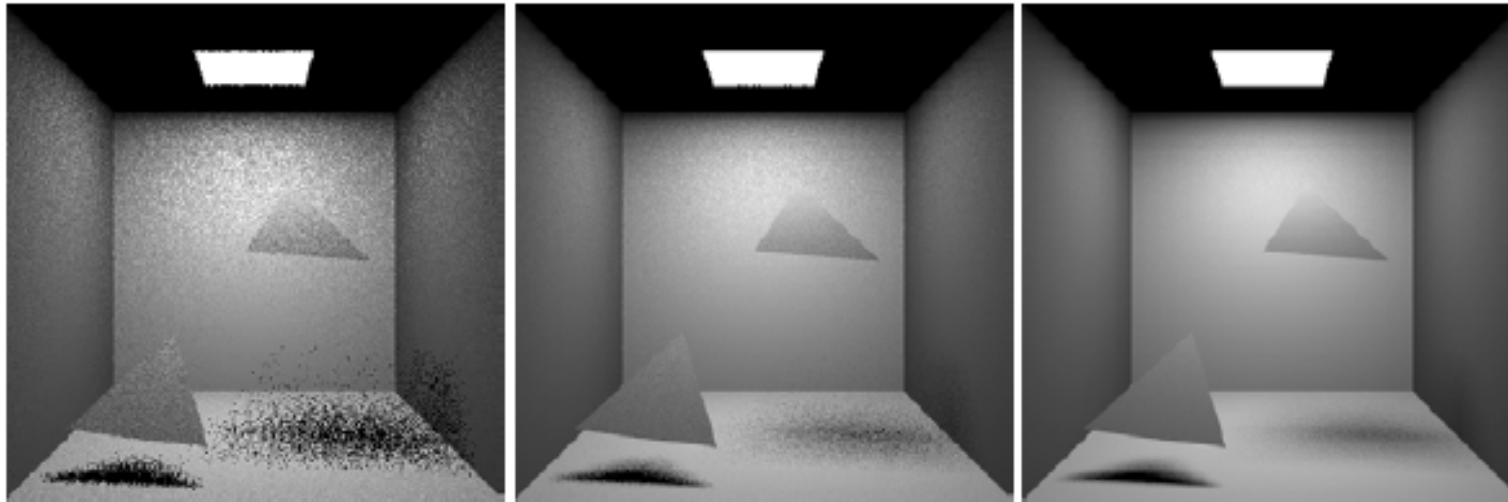
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- **Parameters**
  - **Num. of starting ray per pixel**
  - **Num. of random rays for each surface point (branching factor)**
- **Path tracing**
  - **Branching factor = 1**

# Path Tracing

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1 ray / pixel

10 rays / pixel

100 rays / pixel

From kavita's slides

- **Pixel sampling + light source sampling folded into one method**

# Algorithm so far

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- **Shoot primary rays through each pixel**
- **Shoot indirect rays, sampled over hemisphere**
  - **Path tracing shoots only 1 indirect ray**
- **Terminate recursion using Russian Roulette**



# Performance

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- **Want better quality with smaller # of samples**
  - **Fewer samples/better performance**
  - **Quasi Monte Carlo: well-distributed samples**
  - **Adaptive sampling**

# Some Example

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**Uniform sampling  
(64 samples per pixel)**

**Adaptive sampling**

**Reference**

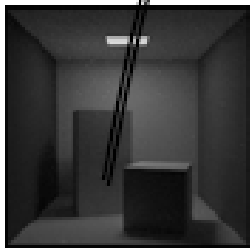
# Importance Sampling

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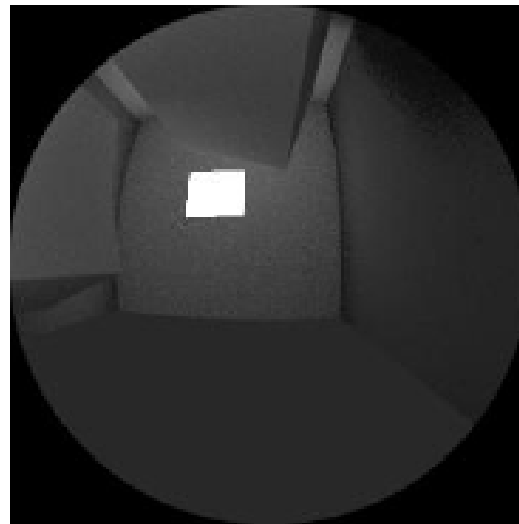
$$L(x \rightarrow \Theta) = L_e(x \rightarrow \Theta) + \int_{\Omega_x} f_r(\Psi \leftrightarrow \Theta) \cdot L(x \leftarrow \Psi) \cdot \cos(\Psi, n_x) \cdot d\omega_\Psi$$



Radiance from light sources + radiance from other surfaces



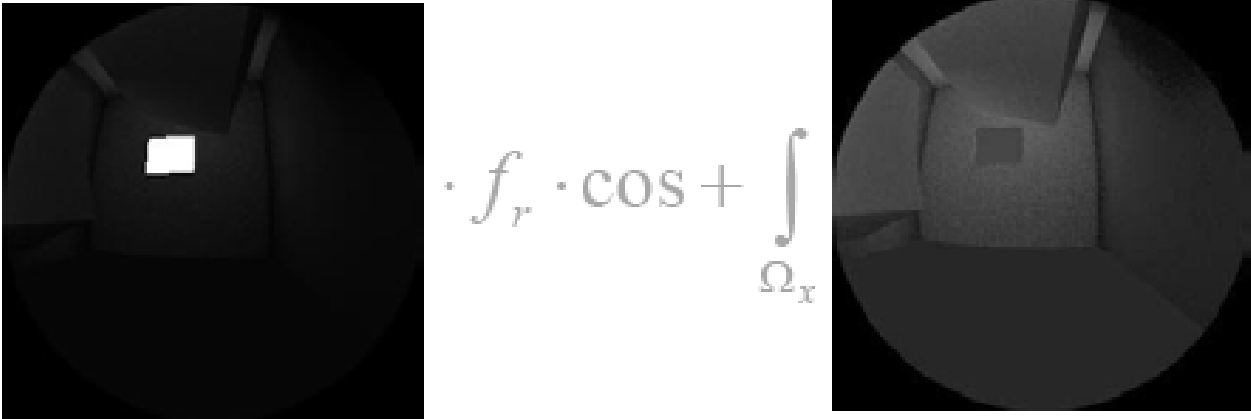
$$= L_e + \int_{\Omega_x} \cdot f_r \cdot \cos$$



# Importance Sampling

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$$L(x \rightarrow \Theta) = L_e + L_{direct} + L_{indirect}$$

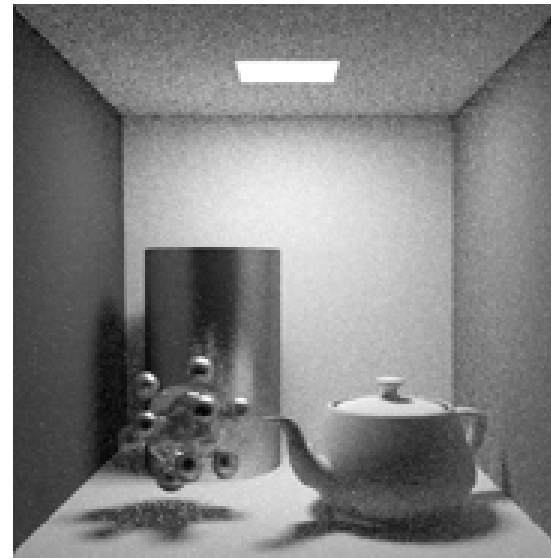
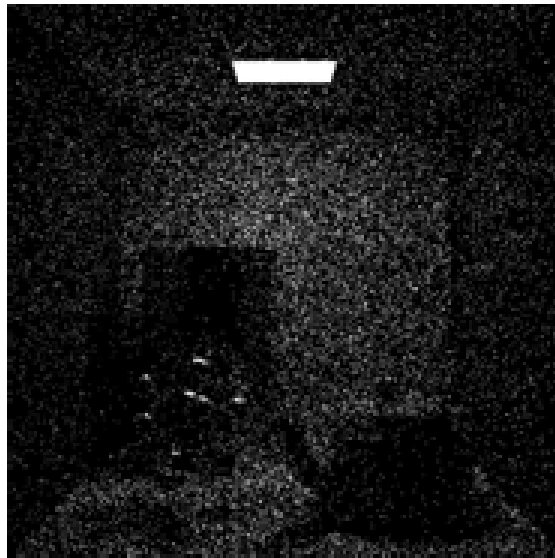
$$= L_e + \int_{\Omega_x} \text{img}_1 \cdot f_r \cdot \cos + \int_{\Omega_x} \text{img}_2 \cdot f_r \cdot \cos$$


- So ... sample direct and indirect with separate MC integration

# Comparison

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From kavita's slides

- **With and without considering direct illumination**
  - **16 samples / pixel**

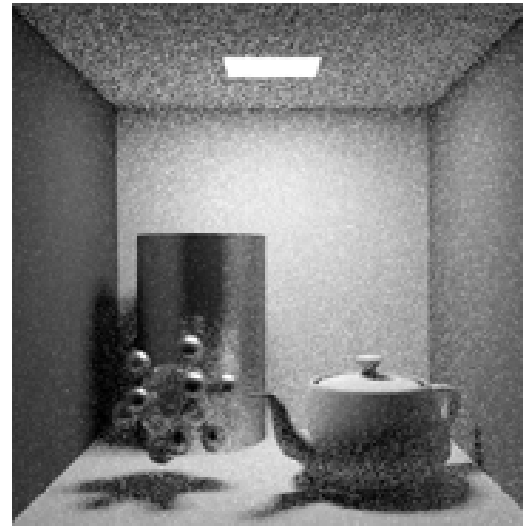
# Rays per pixel

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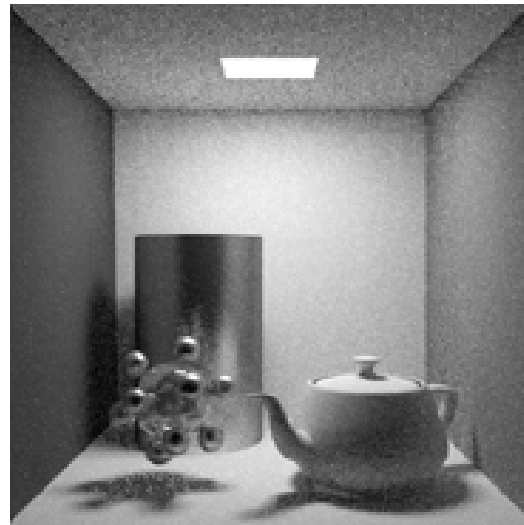
1 sample/  
pixel



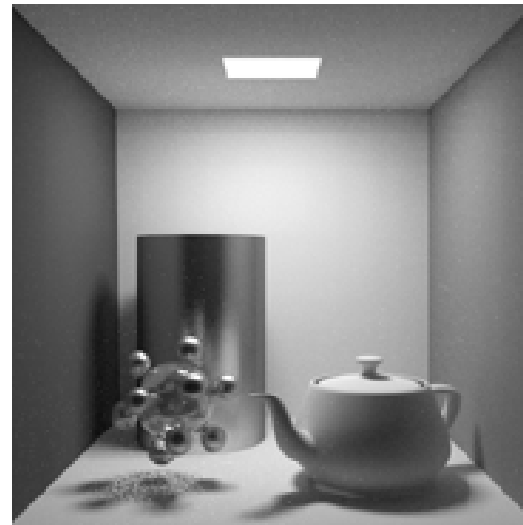
4 samples/  
pixel



16 samples/  
pixel



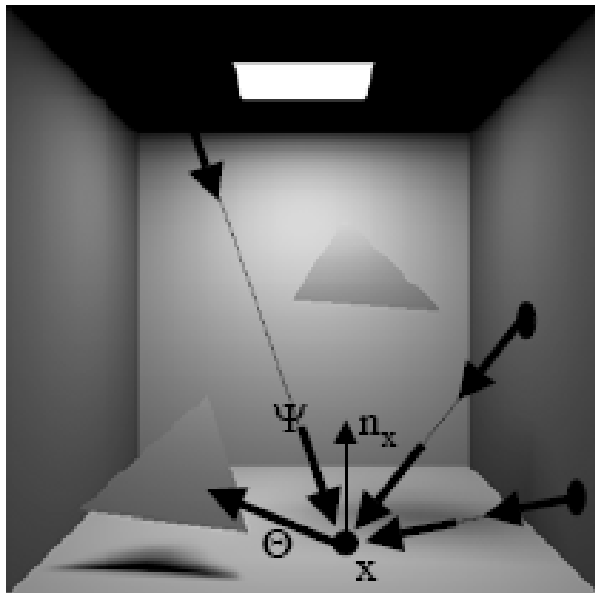
256 samples/  
pixel



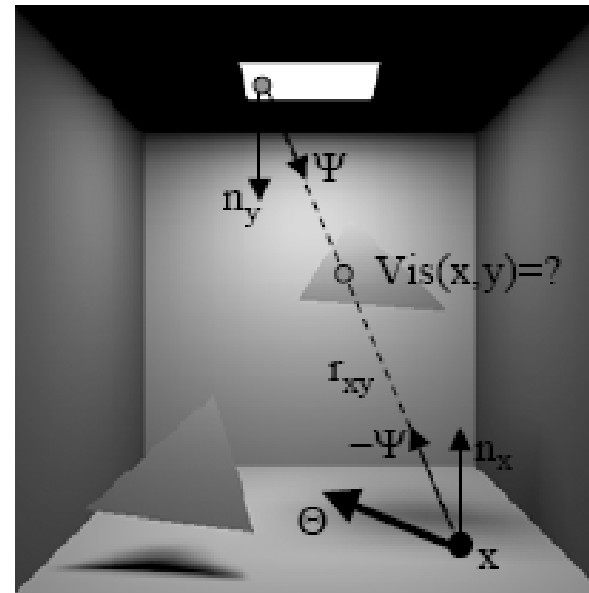
# Direct Illumination

$$L(x \rightarrow \Theta) = \int_{A_{source}} f_r(x, -\Psi \leftrightarrow \Theta) \cdot L(y \rightarrow \Psi) \cdot G(x, y) \cdot dA_y$$

$$G(x, y) = \frac{\cos(n_x, \Theta) \cos(n_y, \Psi) Vis(x, y)}{r_{xy}^2}$$



hemisphere integration



area integration

# Estimator for direct lighting

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- Pick a point on the light's surface with pdf

$$p(y)$$

- For N samples, direct light at point x is:

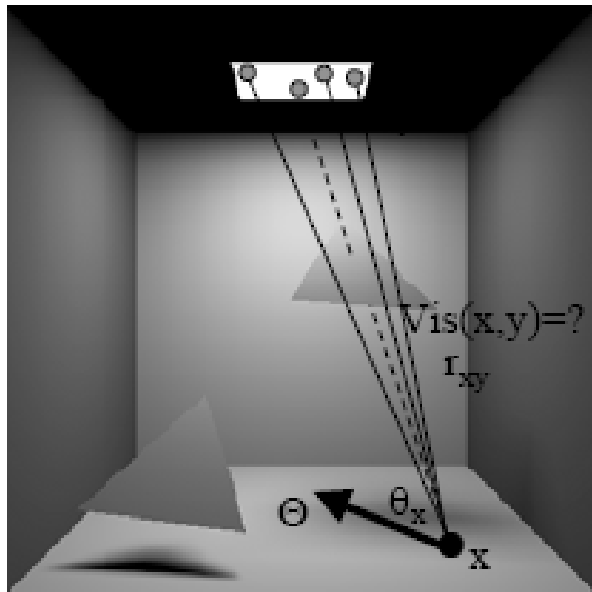
$$E(x) = \frac{1}{N} \sum_{i=1}^N \frac{f_r L_{source} \frac{\cos \theta_x \cos \theta_{\bar{y}_i}}{r_{x\bar{y}_i}^2} Vis(x, \bar{y}_i)}{p(\bar{y}_i)}$$



# Generating direct paths

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- Pick surface points  $y_i$  on light source
- Evaluate direct illumination integral



$$\langle L(x \rightarrow \Theta) \rangle = \frac{1}{N} \sum_{i=1}^N \frac{f_r(\dots)L(\dots)G(x, y_i)}{p(y_i)}$$

# PDF for sampling light

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- Uniform

$$p(y) = \frac{1}{Area_{source}}$$

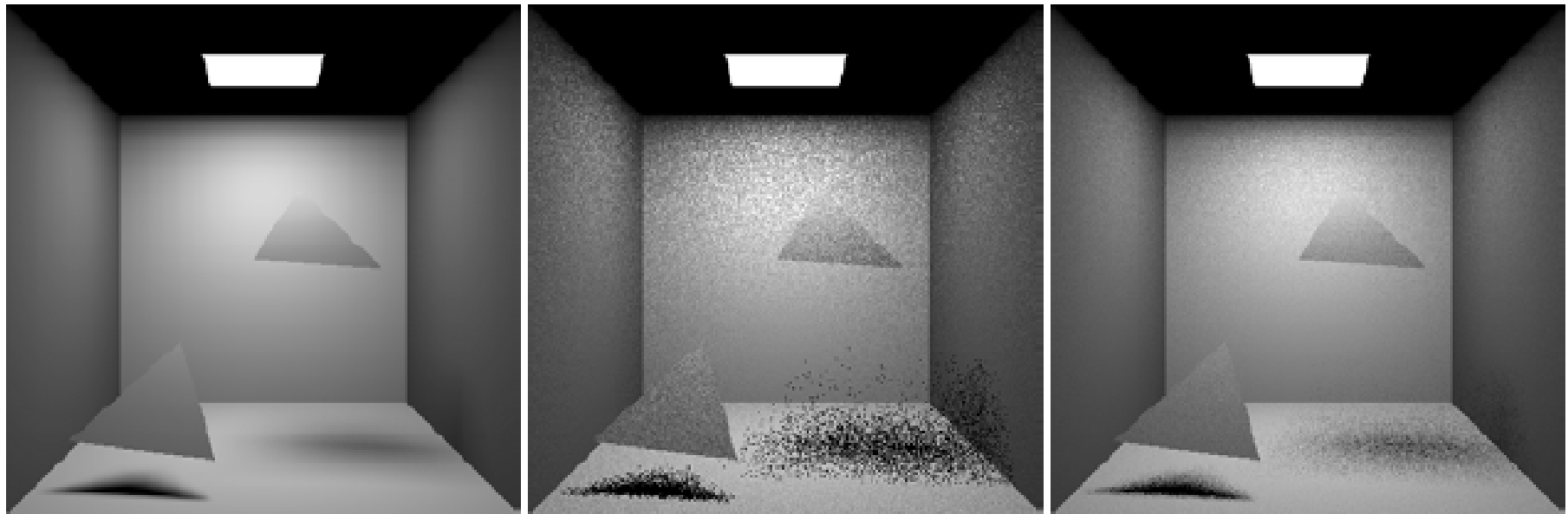
- Pick a point uniformly over light's area
  - Can stratify samples

- Estimator:

$$E(x) = \frac{Area_{source}}{N} \sum_{i=1}^N f_r L_{source} \frac{\cos \theta_x \cos \theta_{\bar{y}_i}}{r_{x\bar{y}_i}^2} Vis(x, \bar{y}_i)$$

# More points ...

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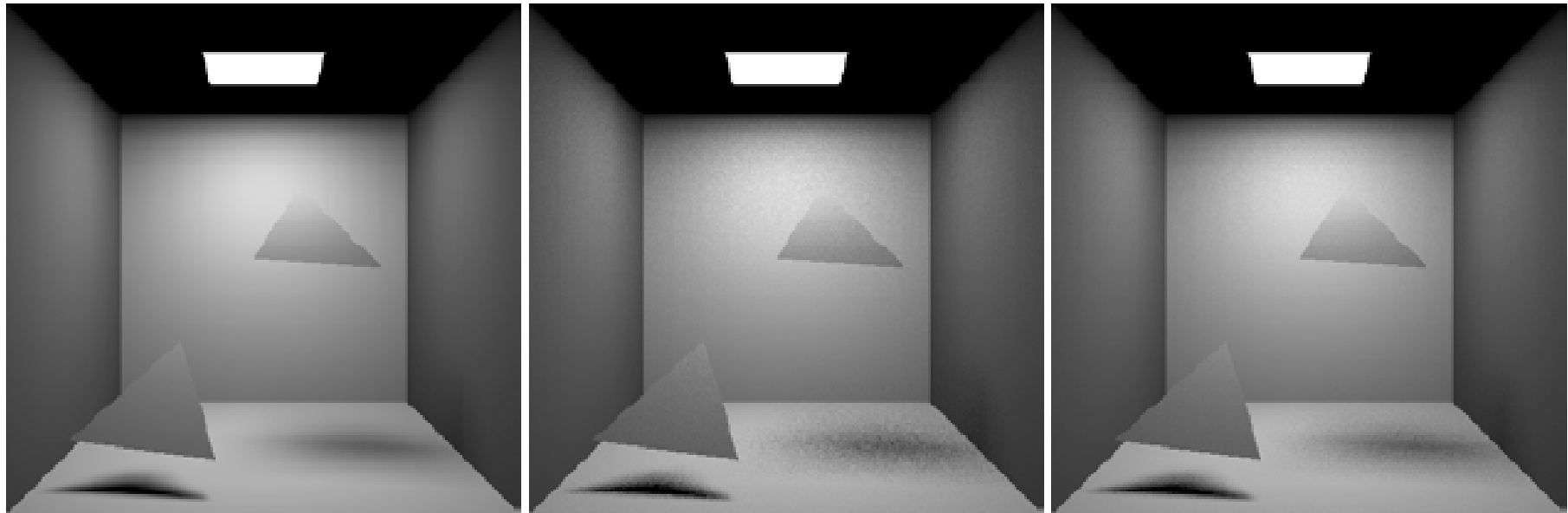
1 shadow ray

9 shadow rays

$$E(x) = \frac{Area_{source}}{N} \sum_{i=1}^N f_r L_{source} \frac{\cos \theta_x \cos \theta_{\bar{y}_i}}{r_{x\bar{y}_i}^2} Vis(x, \bar{y}_i)$$

# Even more points ...

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36 shadow rays

100 shadow rays

$$E(x) = \frac{Area_{source}}{N} \sum_{i=1}^N f_r L_{source} \frac{\cos \theta_x \cos \theta_{\bar{y}_i}}{r_{x\bar{y}_i}^2} Vis(x, \bar{y}_i)$$

# Different pdfs

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- Uniform

$$p(y) = \frac{1}{\text{Area}_{\text{source}}}$$

- Solid angle sampling

- Removes cosine and distance from integrand
- Better when significant foreshortening

$$E(x) = \frac{1}{N} \sum_{i=1}^N \frac{f_r L_{\text{source}} \frac{\cos \theta_x \cos \theta_{\bar{y}_i}}{r_{x\bar{y}_i}^2} \text{Vis}(x, \bar{y}_i)}{p(\bar{y}_i)}$$

# Parameters

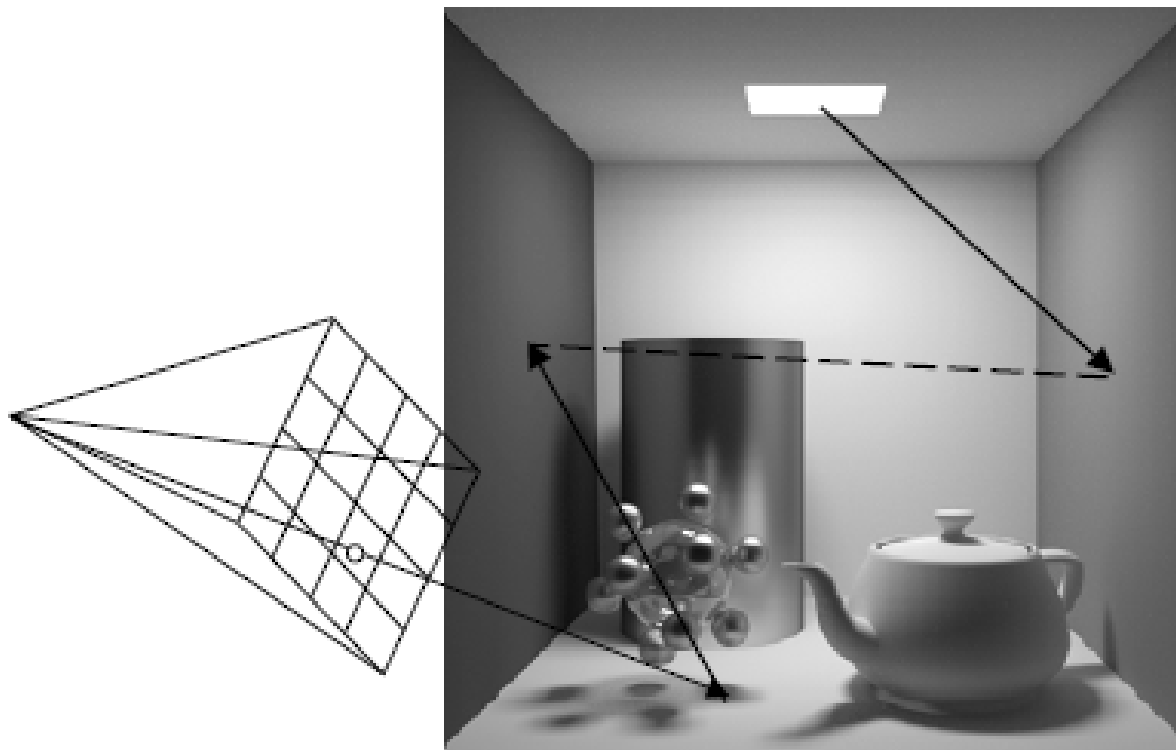
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- How to distribute paths within light source?
  - Uniform
  - Solid angle
  - What about light distribution?
- How many paths (“shadow-rays”)?
  - Total?
  - Per light source? (~intensity, importance, ...)

# Bidirectional Path Tracing

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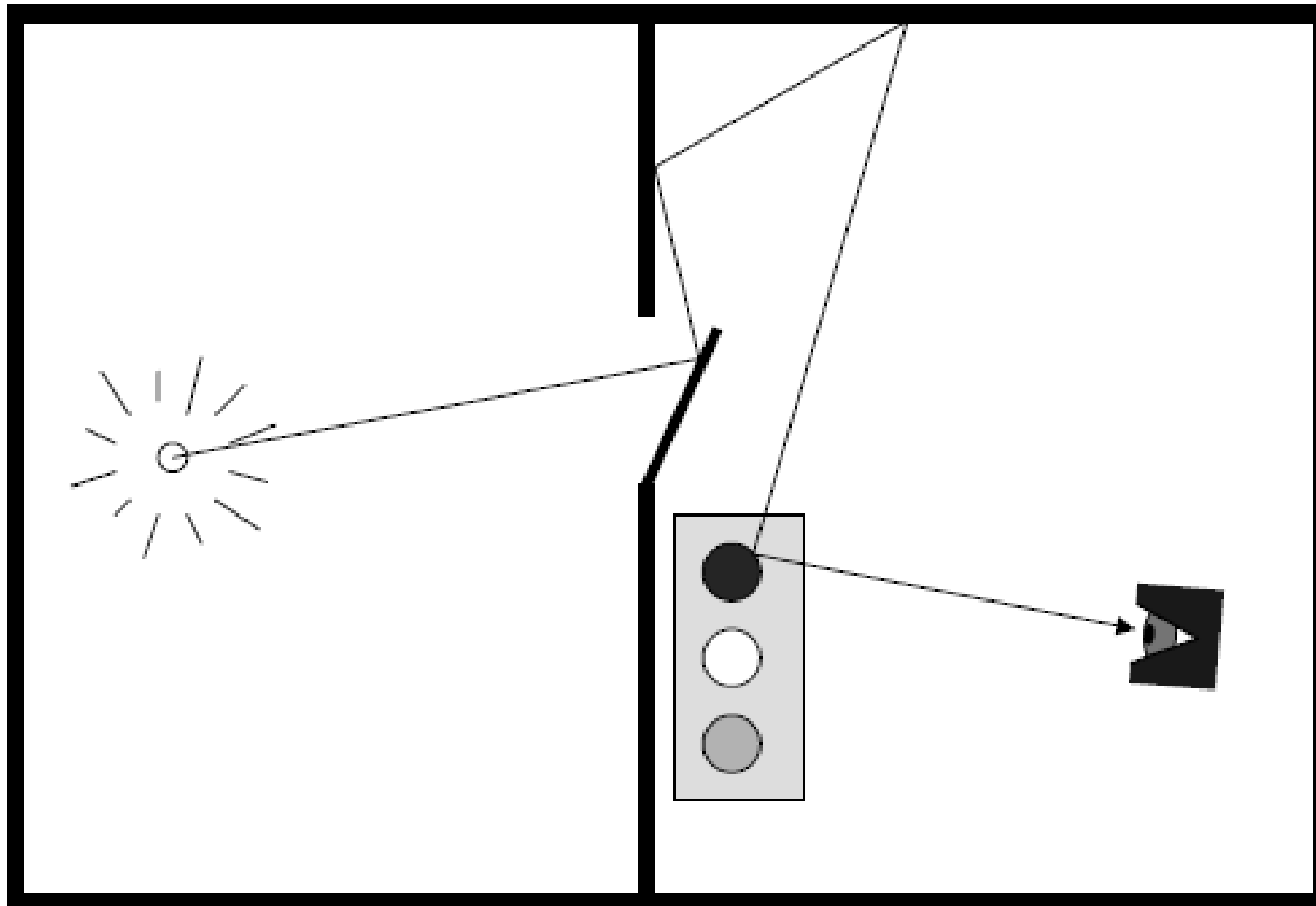
- Or paths generated from both camera and source at the same time ...!



- Connect endpoints to compute final contribution

# Metropolis

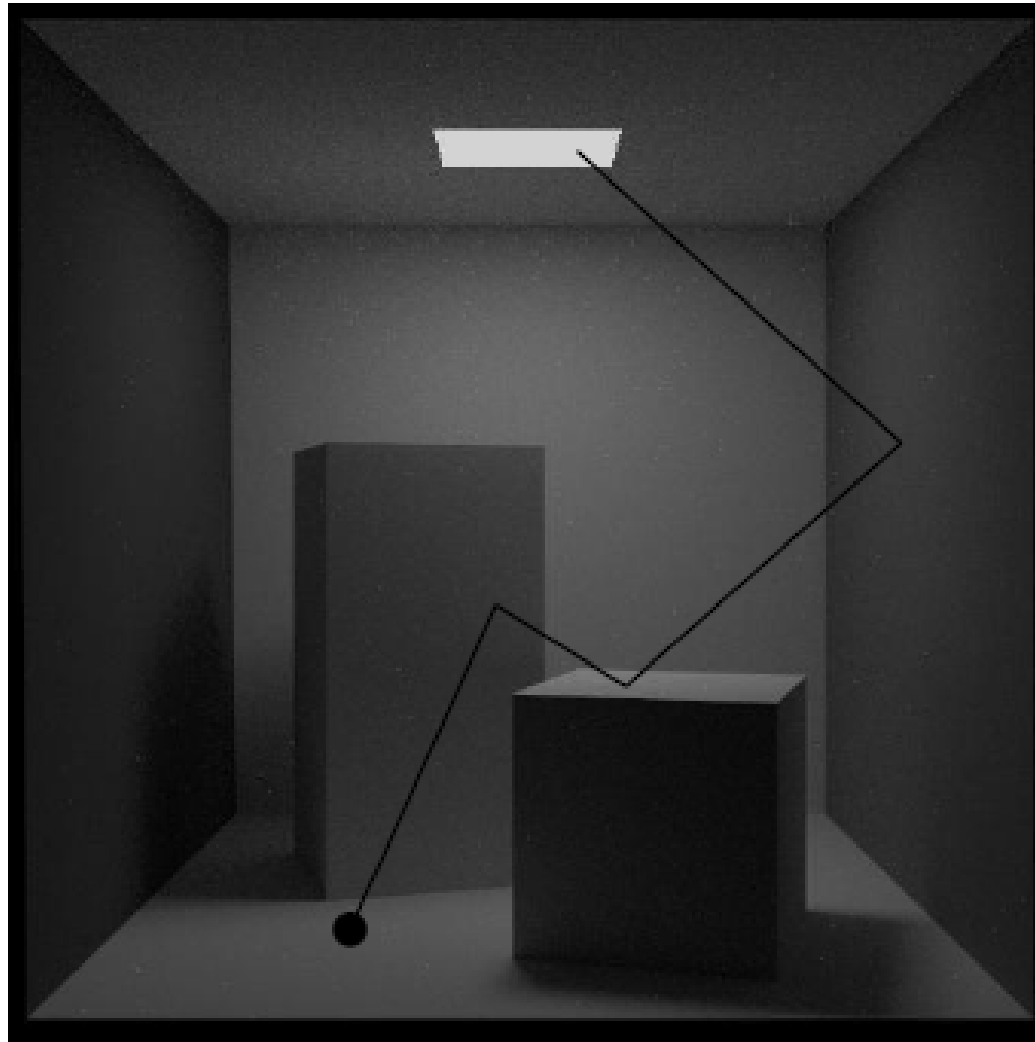
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# Metropolis

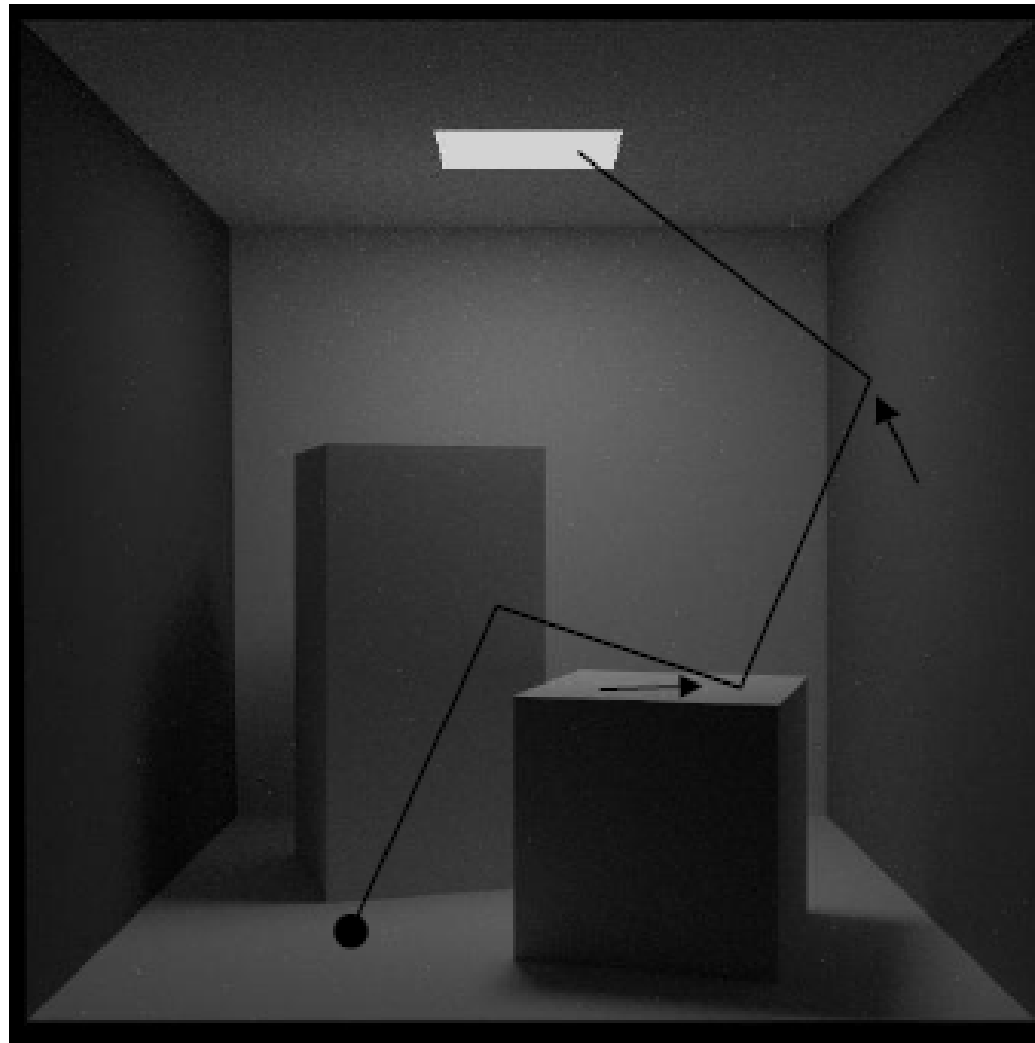
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valid path

# Metropolis

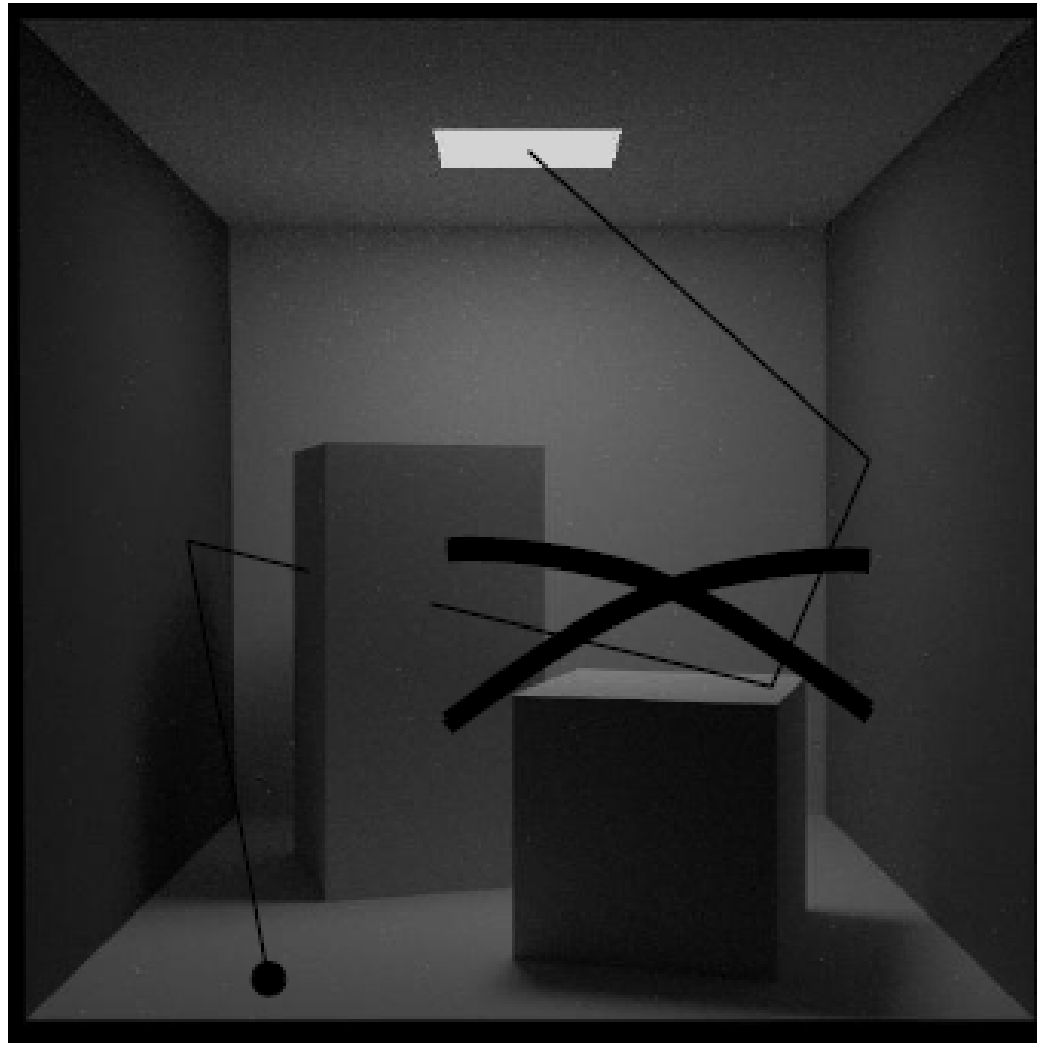
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small  
perturbations

# Metropolis

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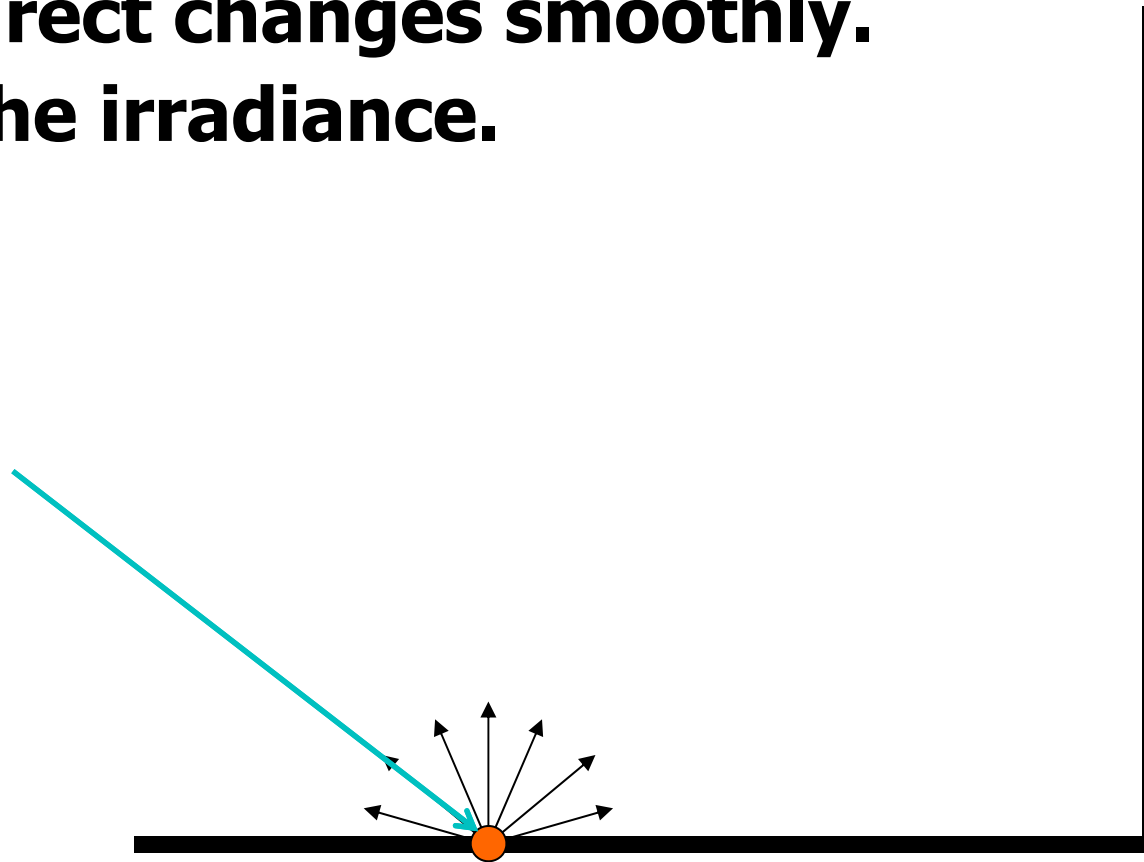
Accept  
mutations  
based on  
energy  
transport

# Biased Methods: Irradiance Caching

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- **Indirect changes smoothly.**
- **Cache irradiance.**

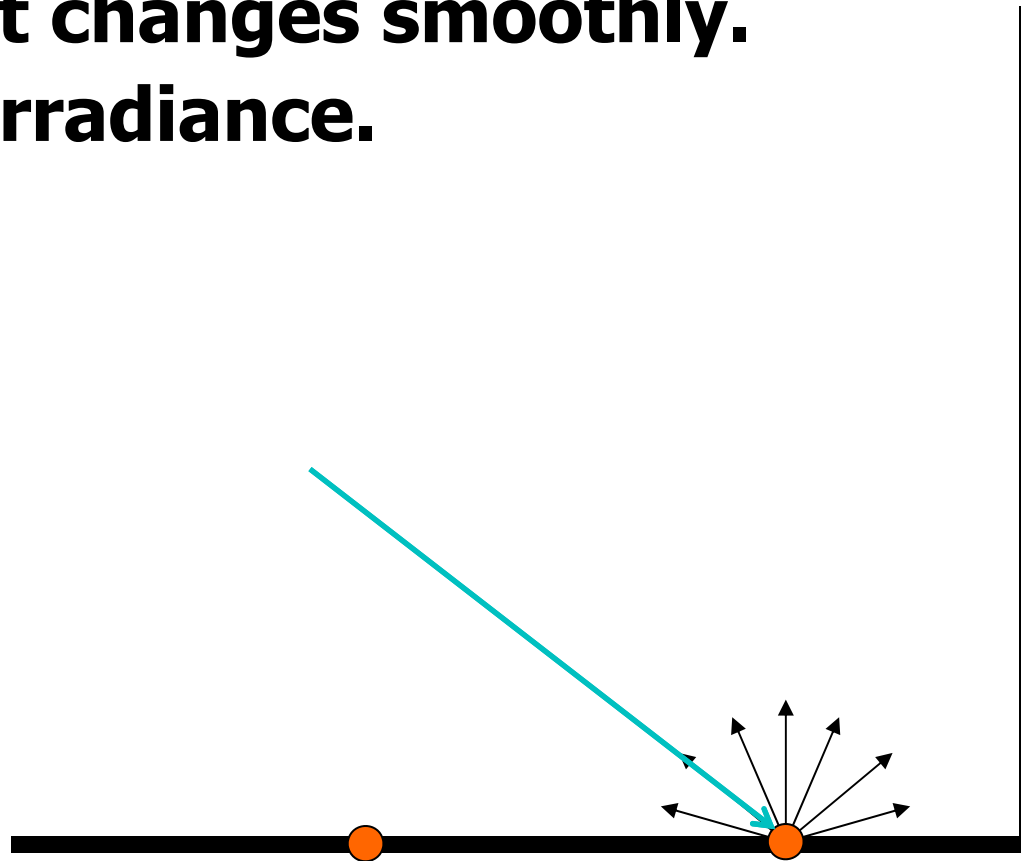


# Irradiance Caching

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- **Indirect changes smoothly.**
- **Cache irradiance.**

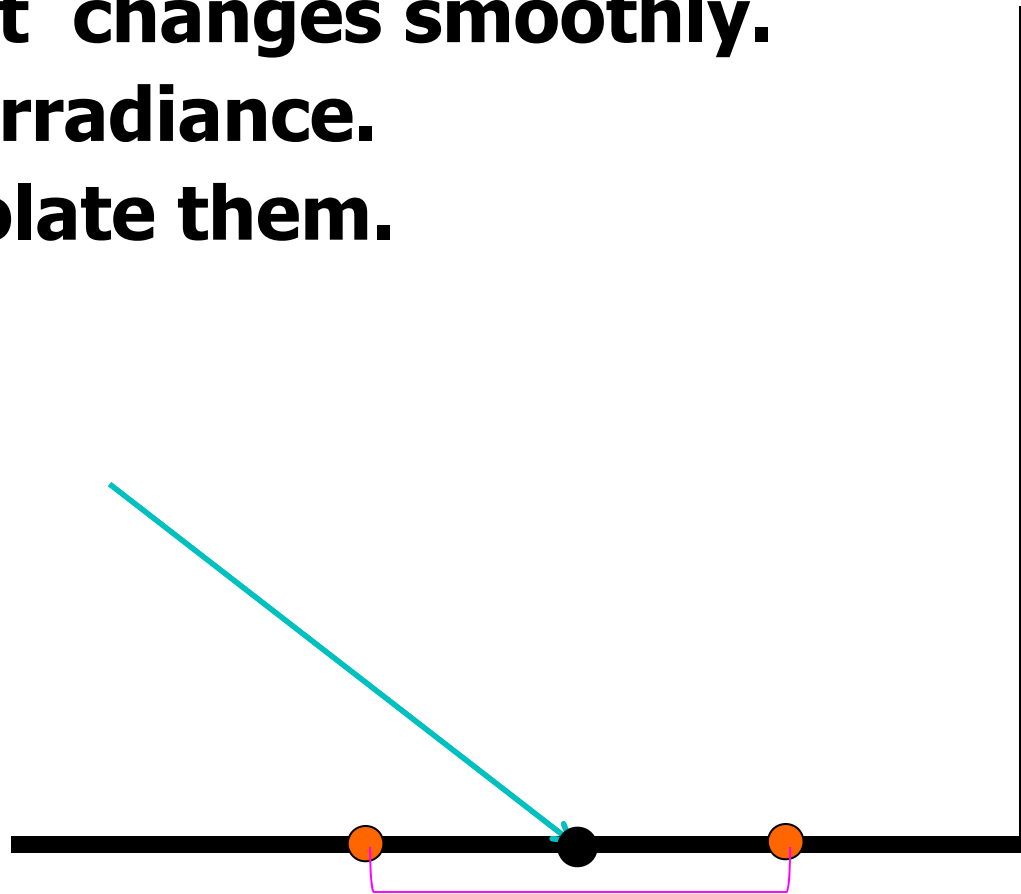


# Irradiance Caching

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- **Indirect changes smoothly.**
- **Cache irradiance.**
- **Interpolate them.**



# Biased Method: Photon Mapping

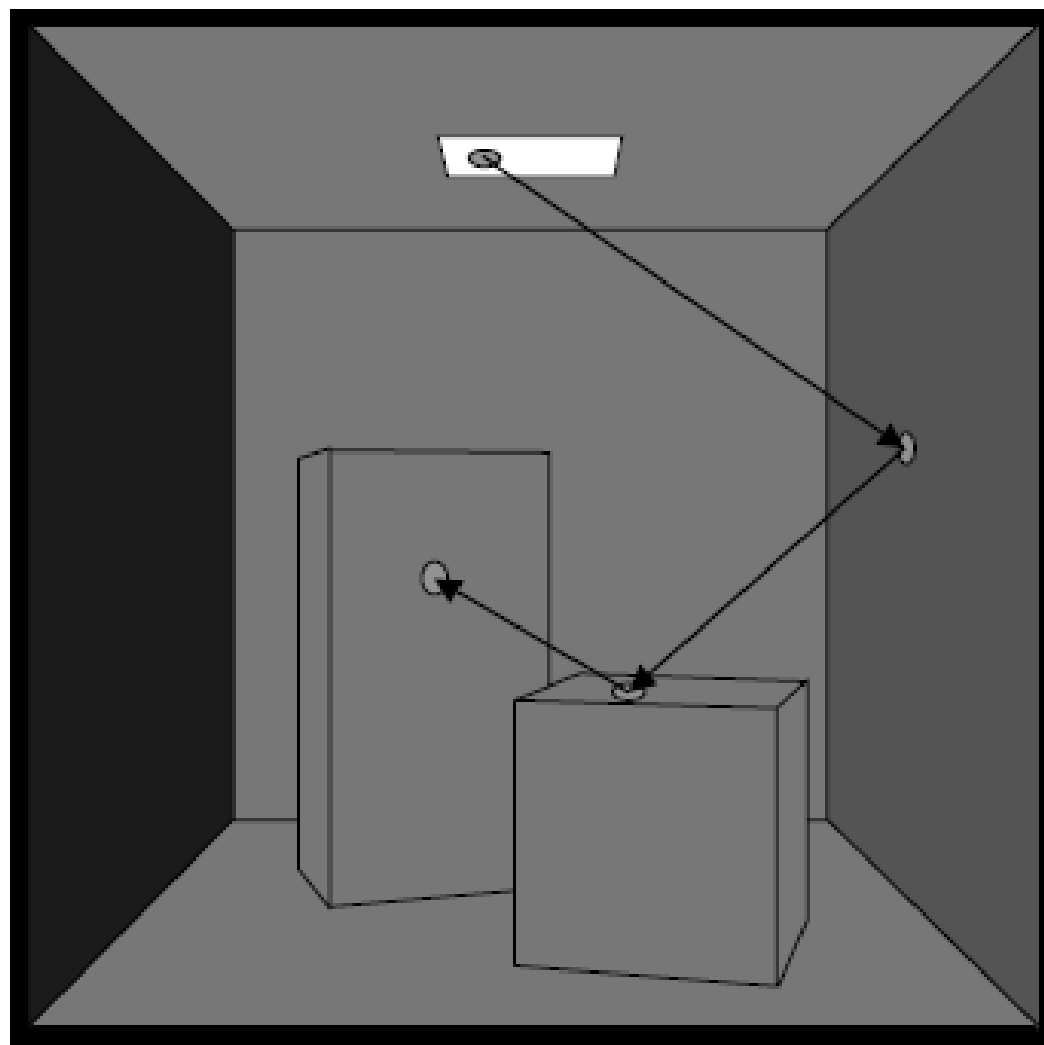
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- **2 passes:**
  - **Shoot “photons” (light-rays) and record any hit-points**
  - **Shoot viewing rays and collect information from stored photons**

# Pass 1: shoot photons

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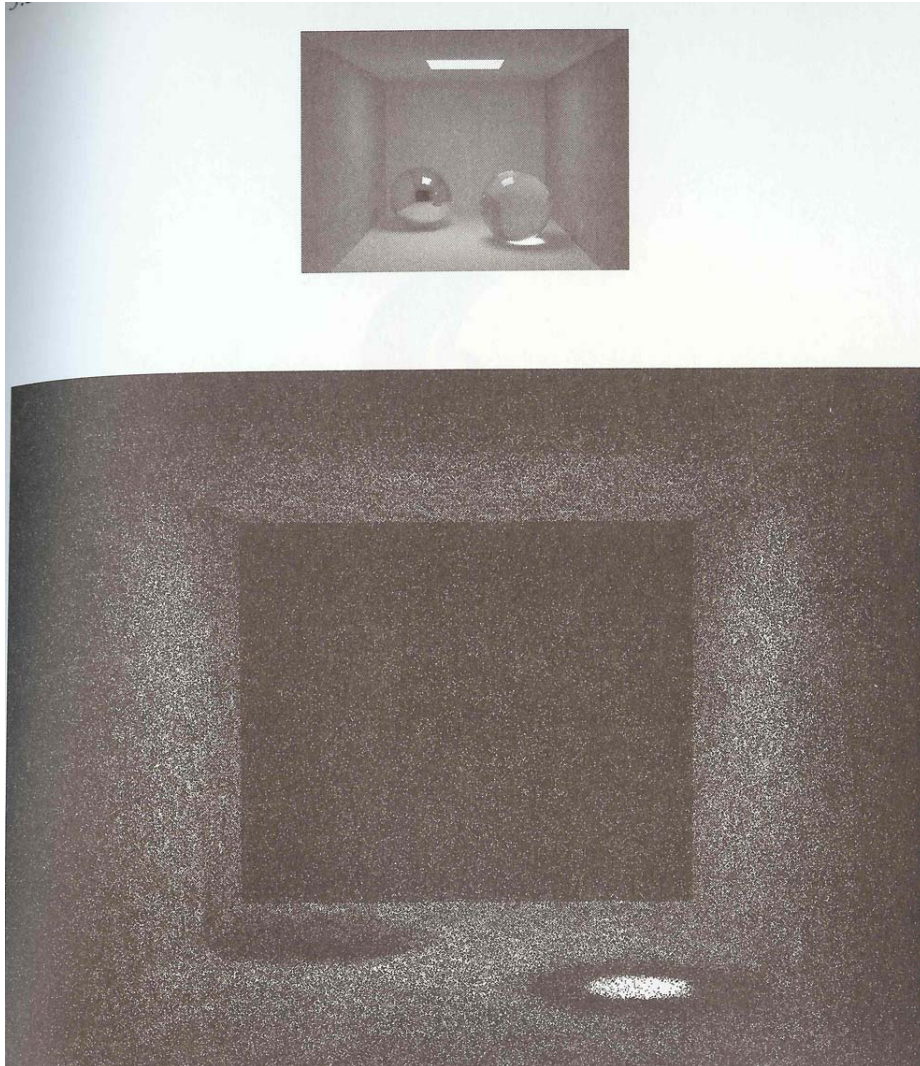
- Light path generated using MC techniques and Russian Roulette
- Store:
  - position
  - incoming direction
  - color
  - ...



# Stored Photons

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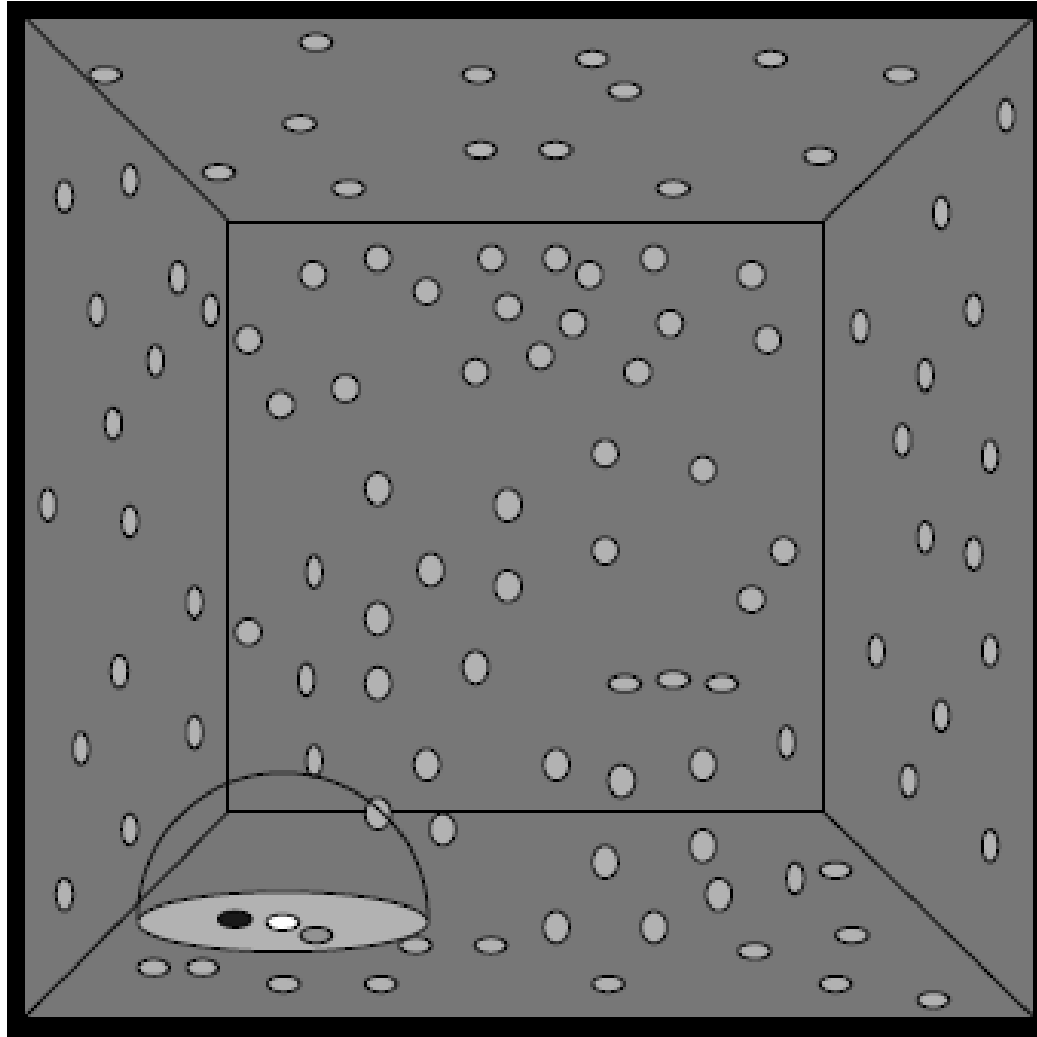
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**Generate a few  
hundreds of  
thousands of  
photons**

# Pass 2: viewing ray

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- Search for  $N$  closest photons (+check normal)
- Assume these photons hit the point we're interested in
- Compute average radiance

# Result

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**350K photons  
for the caustic  
map**

# Result

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**350K photons  
for the caustic  
map**

# Class Objectives were:

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- **Understand a basic structure of Monte Carlo ray tracing**
  - **Russian roulette for its termination**
  - **Path tracing**

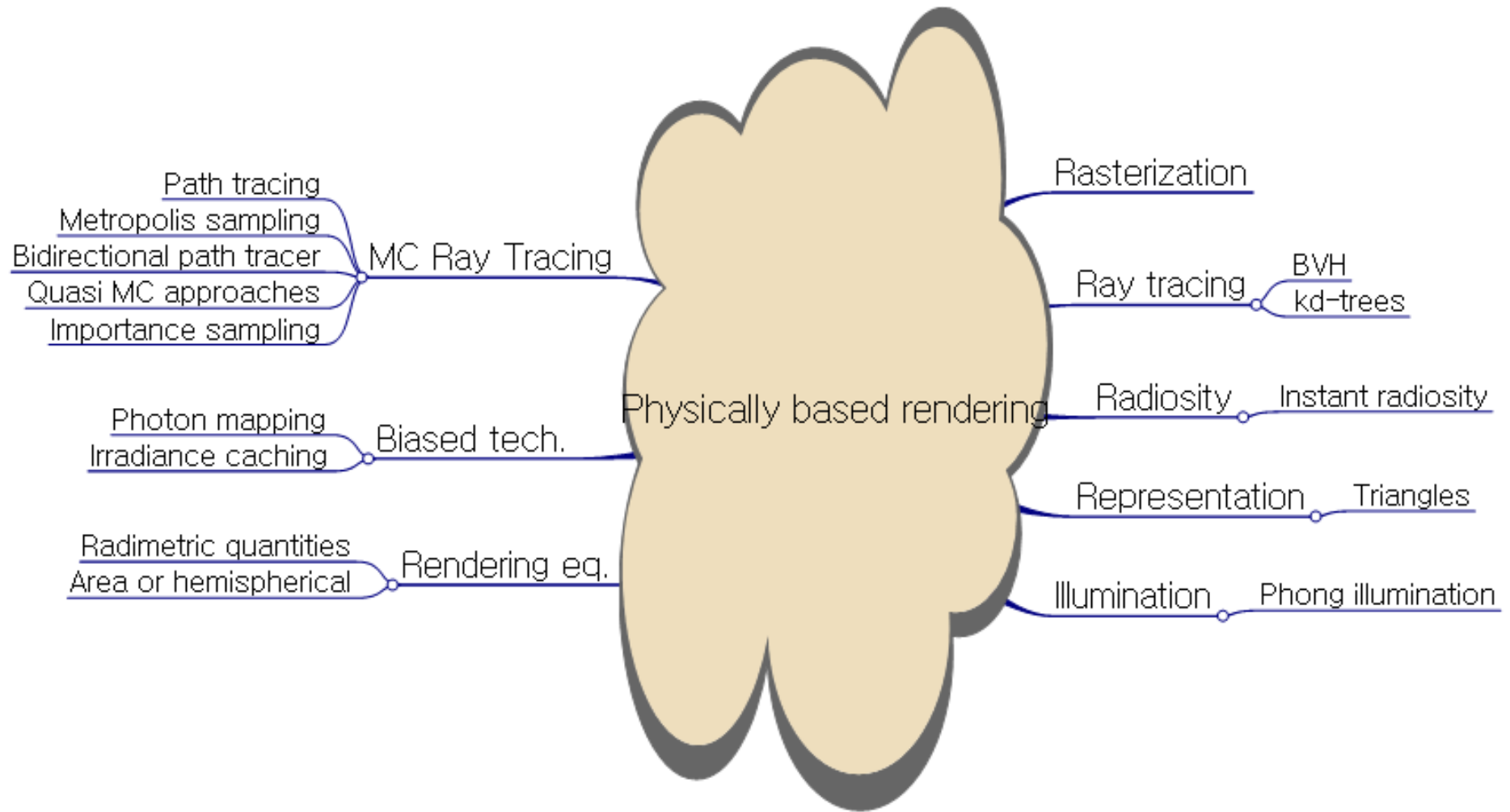
# Summary

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- **Two basic building blocks**
- **Radiometry**
- **Rendering equation**
- **MC integration**
- **MC ray tracing**
  - **Unbiased methods**
  - **Biased methods**

# Summary



# Next Time...

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- **Instant radiosity**



# Homework

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- **Go over the next lecture slides before the class**
- **Watch 2 SIG/I3D/HPG videos and submit your summaries every Tue. class**
  - **Just one paragraph for each summary**

## **Example:**

**Title: XXX XXXX XXXX**

**Abstract: this video is about accelerating the performance of ray tracing. To achieve its goal, they design a new technique for reordering rays, since by doing so, they can improve the ray coherence and thus improve the overall performance.**

# Any Questions?

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- **Submit four times in Sep./Oct.**
- **Come up with one question on what we have discussed in the class and submit at the end of the class**
  - **1 for typical questions**
  - **2 for questions that have some thoughts or surprise me**