
CS482: Monte Carlo Ray Tracing:

Sung-Eui Yoon
(윤성익)

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

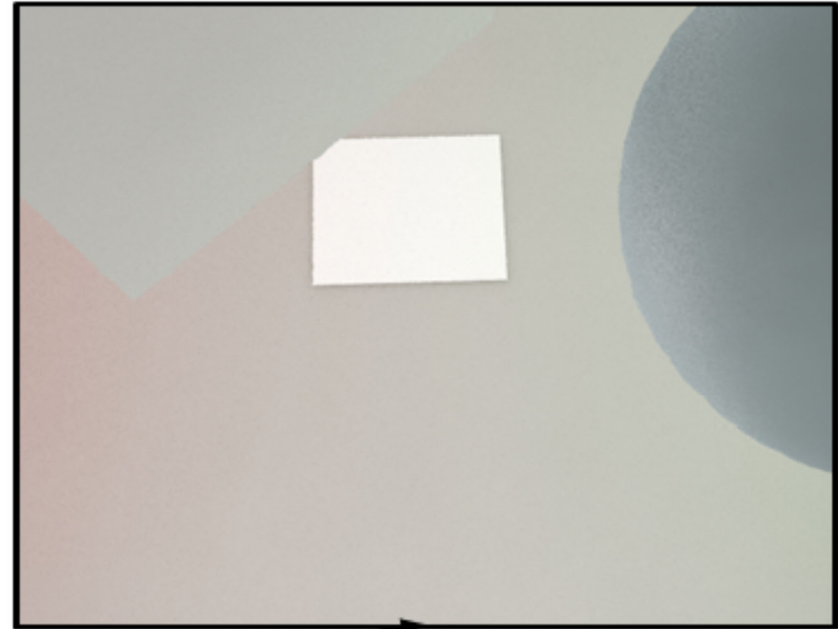
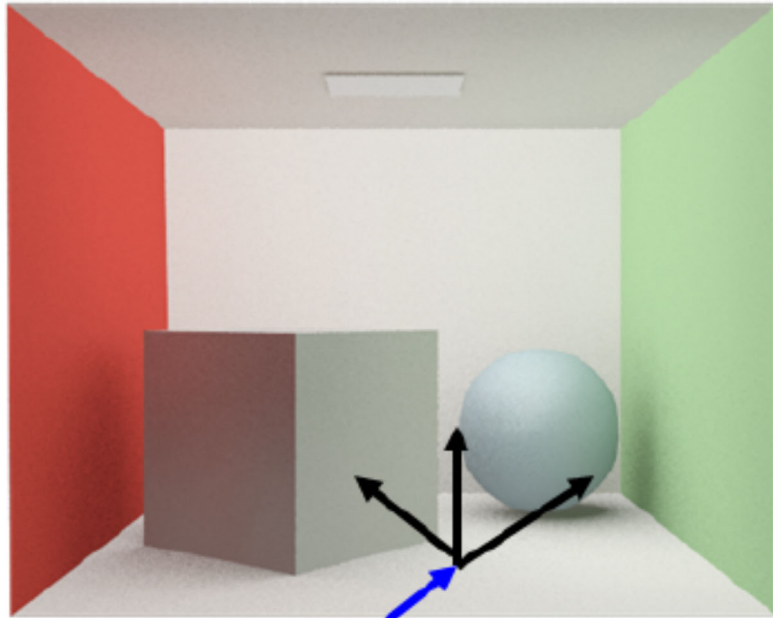
KAIST

The KAIST logo consists of the word "KAIST" in a bold, blue, sans-serif font. Below the text is a light blue, horizontal oval shape that serves as a shadow or base for the letters.

Class Objectives (Ch. 15)

- **Understand a basic structure of Monte Carlo ray tracing**
 - Russian roulette for its termination
 - Path tracing
- **Last time:**
 - Monte Carlo integration: sampling approach for solving the rendering equation
 - Estimator and its variance

Rendering Equation

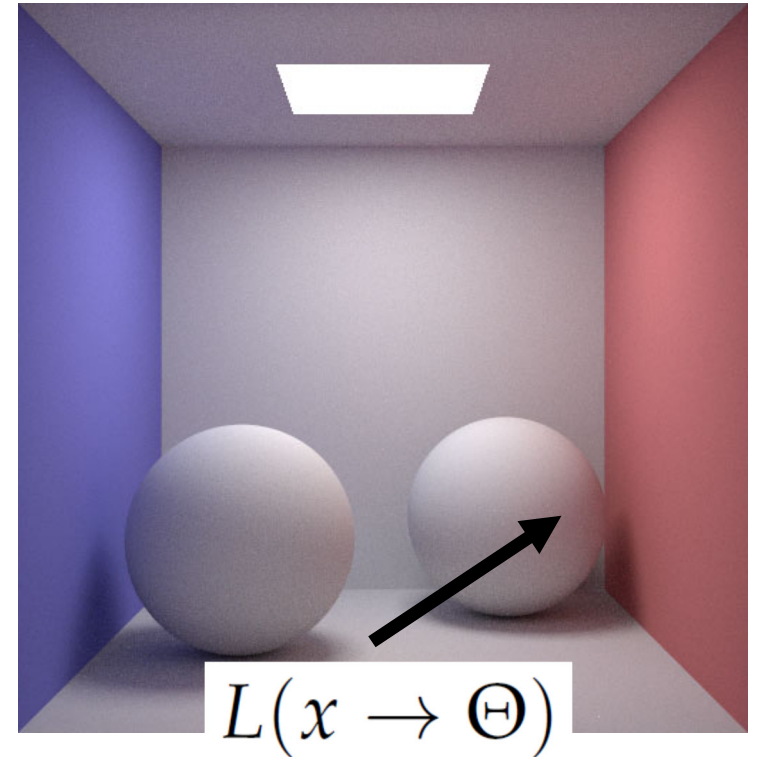


Incoming radiance on the hemisphere

$$L_r(x \rightarrow \Theta) = \int_{\Psi} L(x \leftarrow \Psi) f_r(x, \Psi \rightarrow \Theta) \cos \theta_x dw_{\Psi}$$

Evaluation

- **To compute** $L(x \rightarrow \Theta)$:
 - **Check** $L_e(x \rightarrow \Theta)$
 - **Evaluate** $L_r(x \rightarrow \Theta)$



$$L_r(x \rightarrow \Theta) = \int_{\Psi} L(x \leftarrow \Psi) f_r(x, \Psi \rightarrow \Theta) \cos \theta_x dw_{\Psi}$$

Evaluation

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

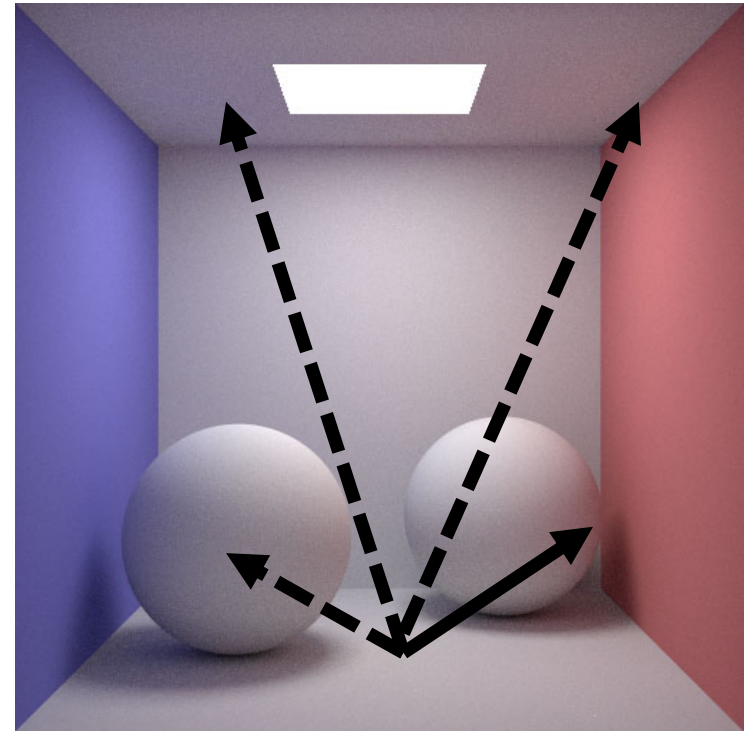
$$L_r(x \rightarrow \Theta) = \int_{\Psi} L(x \leftarrow \Psi) f_r(x, \Psi \rightarrow \Theta) \cos \theta_x d\omega_{\Psi}$$

$$\hat{L}_r(x \rightarrow \Theta) = \frac{1}{N} \sum_{i=1}^N \frac{L(x \leftarrow \Psi_i) f_r(x, \Psi_i \rightarrow \Theta) \cos \theta_x}{p(\Psi_i)}$$

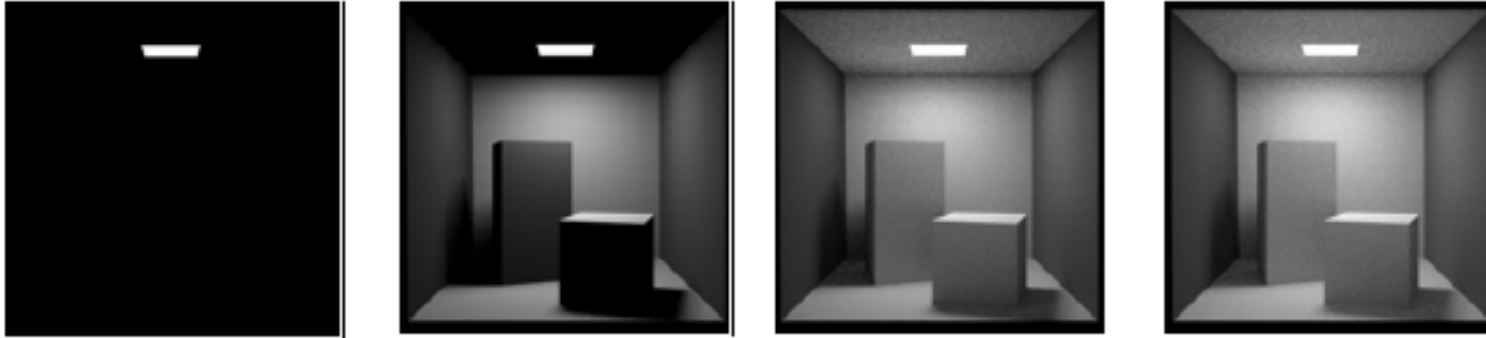
- How about $L(x \leftarrow \Psi_i)$?

Evaluation

- How about $L(x \leftarrow \Psi_i)$?
- Perform ray casting backward
- Compute radiance from those visible points to x
 - Assume reciprocity
- Recursively perform the process
 - Each additional bounce supports one more indirect illumination



When to end recursion?



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**

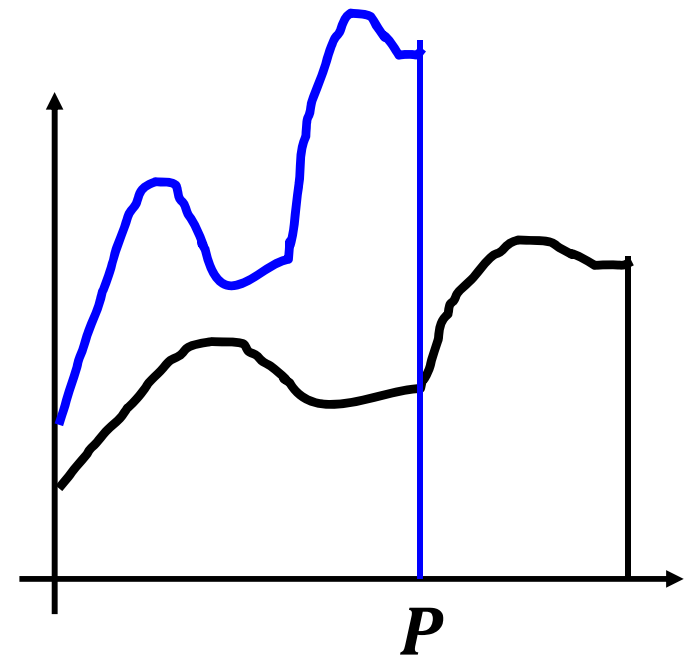
Russian Roulette

- **Integral: Substitute $y = Px$**

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

- **Estimator**

$$\hat{I}_{\text{roulette}} = \begin{cases} \frac{f(x_i)}{P} & \text{if } x_i \leq P, \\ 0 & \text{if } x_i > P. \end{cases}$$



- **Variance?**

Russian Roulette

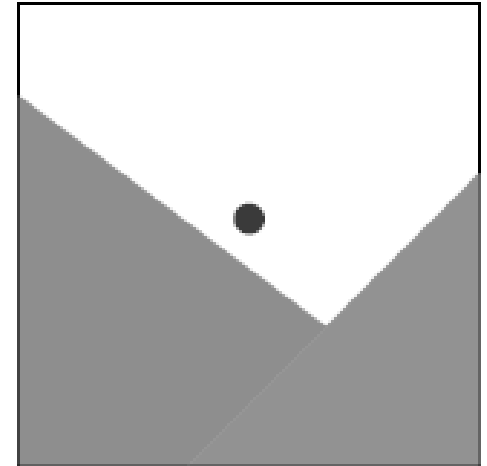
- **Pick absorption probability, $\alpha = 1-P$**
 - **Recursion is terminated**
- **$1-\alpha$, i.e., P , is commonly to be equal to the reflectance of the material of the surface**
 - **Darker surface absorbs more paths**

Algorithm so far

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

Pixel Anti-Aliasing

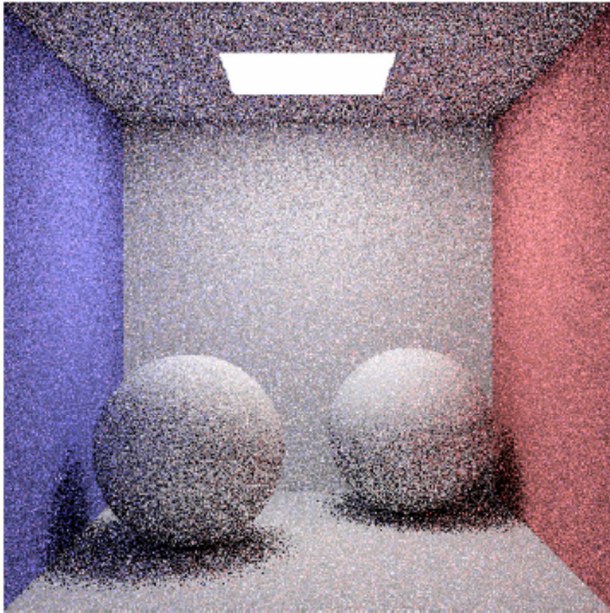
- **Compute radiance only at the center of pixel**
 - Produce jaggies
- **We want to evaluate using MC**
- **Simple box filter**
 - The averaging method



Stochastic Ray Tracing

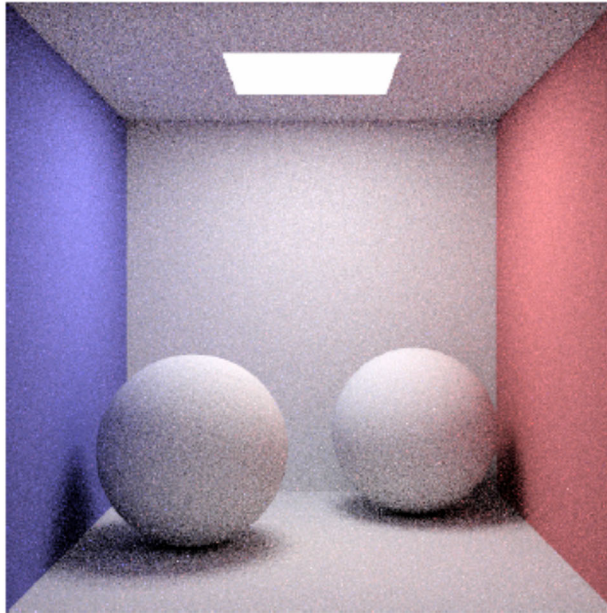
- **Parameters**
 - **Num. of starting ray per pixel**
 - **Num. of random rays for each surface point (branching factor)**
- **Path tracing**
 - **Branching factor = 1**

Path Tracing

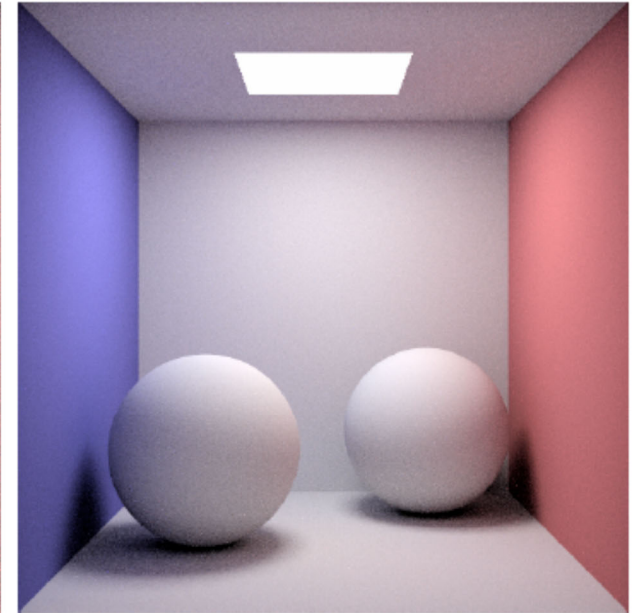


1 spp

(samples per pixel)



4 spp



16 spp

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

Algorithm so far

- 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

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

Some Example



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



Adaptive sampling

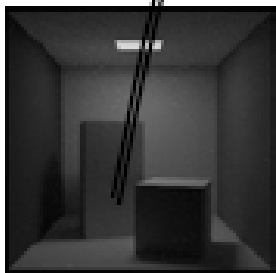


Reference

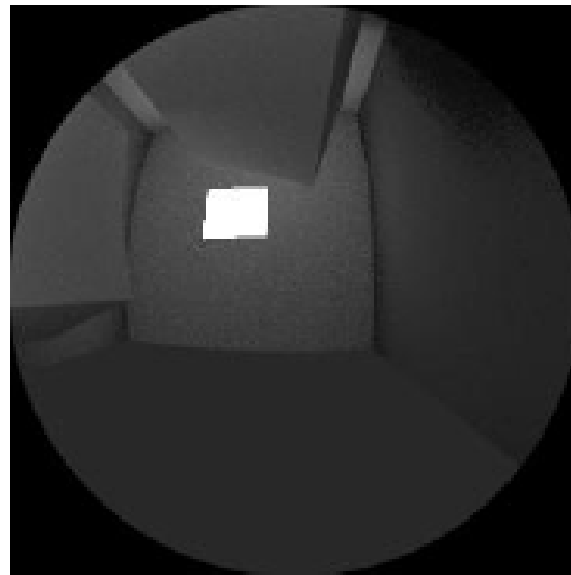
Importance Sampling

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

Radiance from light sources + radiance from other surfaces

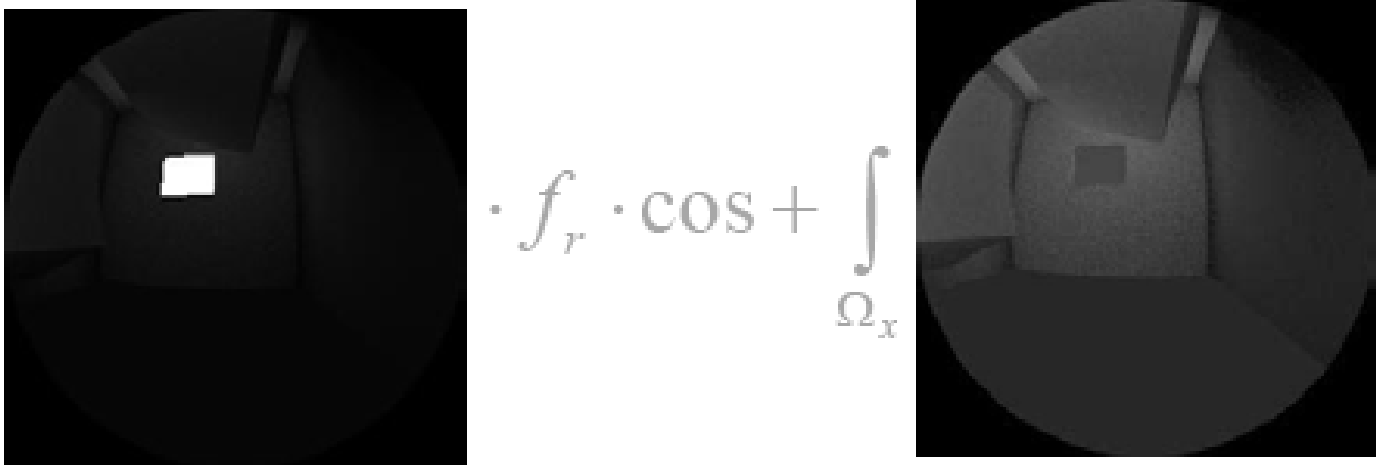


$$= L_e + \int_{\Omega_x} \cdot f_r \cdot \cos$$



Importance Sampling

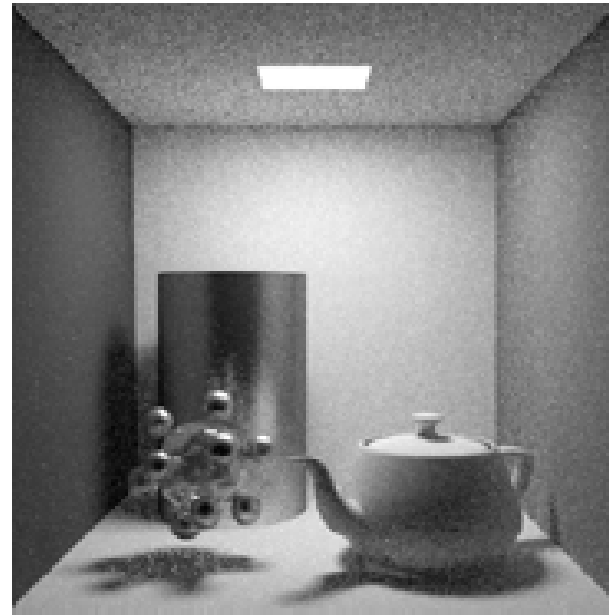
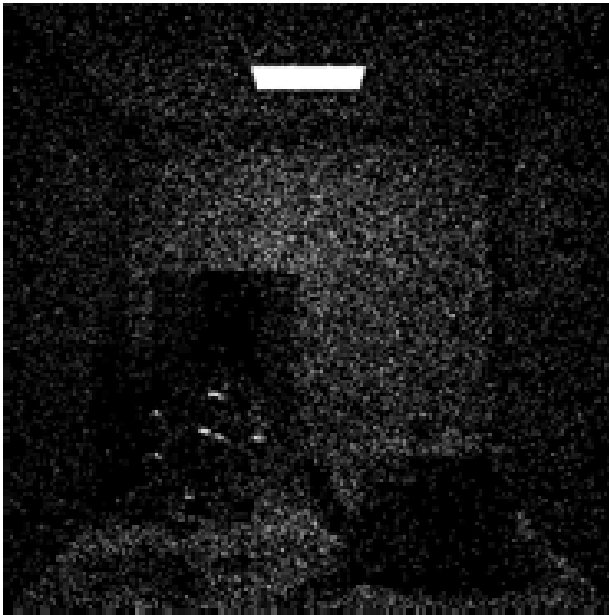
$$L(x \rightarrow \ominus) = L_e + L_{direct} + L_{indirect}$$

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


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



Comparison



From kavita's slides

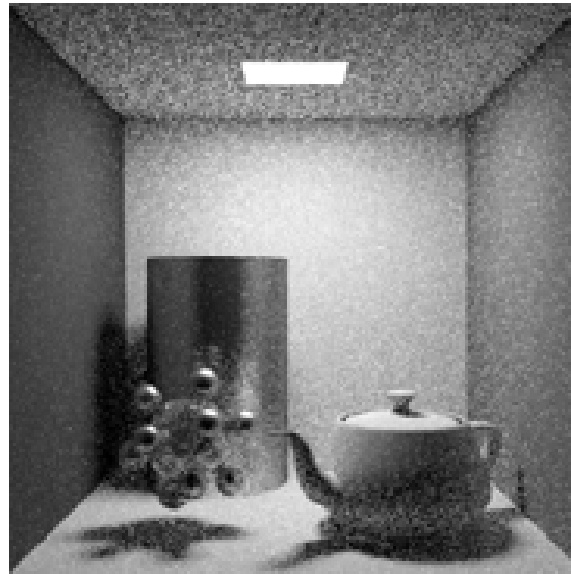
- **With and without considering direct illumination**
 - **16 samples / pixel**

Rays per pixel

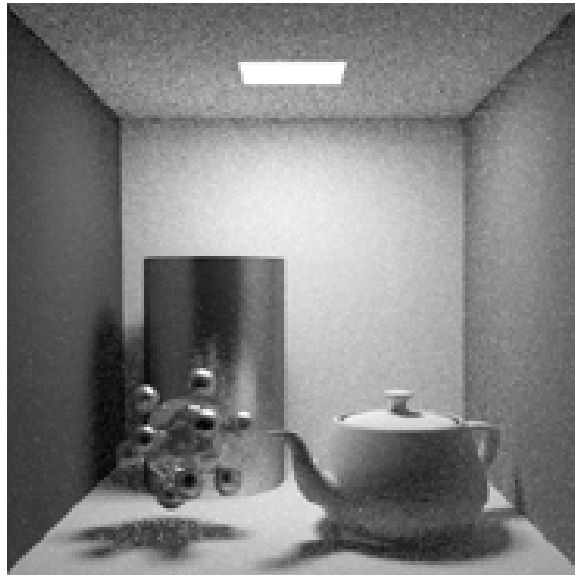
1 sample/
pixel



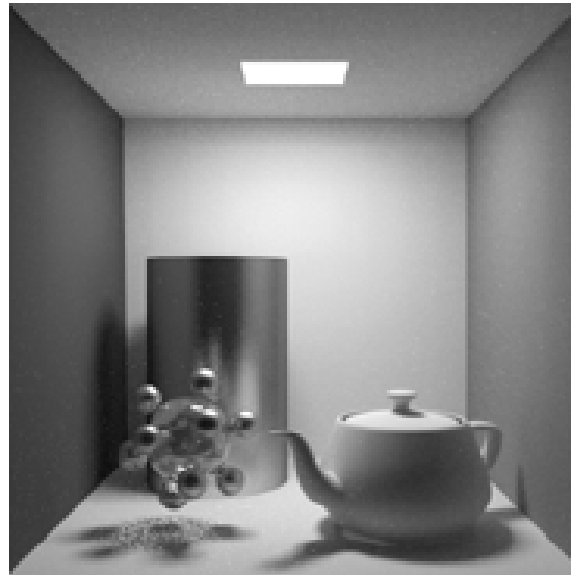
4 samples/
pixel



16 samples/
pixel



256 samples/
pixel



Class Objectives were:

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Next Time...

- **Acceleration techniques for global illumination methods**

Homework

- **Go over the next lecture slides before the class**
- **Watch 2 SIG/CVPR/ISMAR videos and submit your summaries every Mon. class**
 - **Just one paragraph for each summary**
 - **Any top-tier conf (e.g., ICRA) is okay**

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?

- **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**