

## Lecture 13: Photon Mapping and Environment Maps

**CS 6620, Spring 2009**

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### Regexps to classify paths

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- “Looking” directly at a light source  
– **LE**
- Seeing a diffuse surface  
– **LDE**
- Seeing a diffuse surface reflected in a mirror  
– **LDSE**

- 
- Bidirectional path tracing is...?
    - $L (G|D|S)^* (G|D|S)^* E$
    - $= L (G|D|S)^* E$

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## Light Tracing

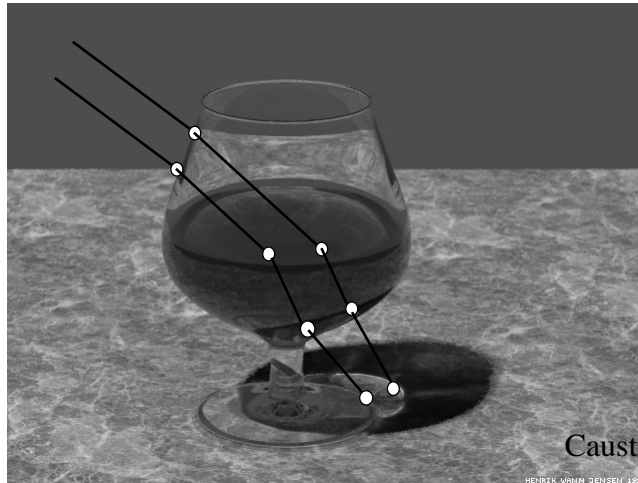
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- Some sub-paths are best sampled from the light
  - Where light is focused – *caustics*
  - Where most of the light's energy goes to a small region of the world
- Several variants

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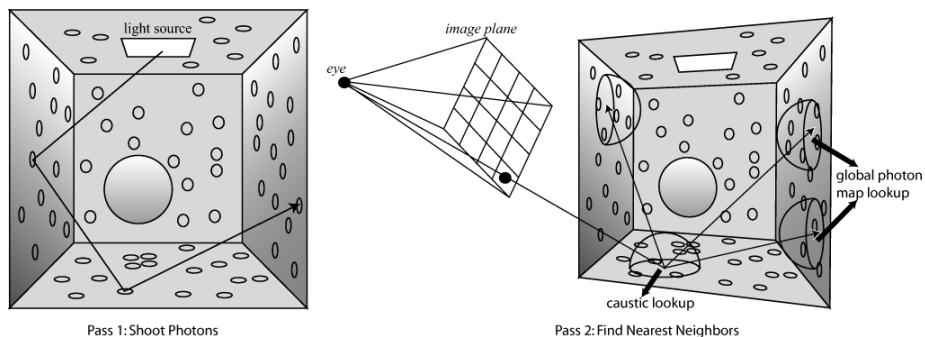
## Photon Map

- Build on irradiance caching
- Use bidirectional ray tracing



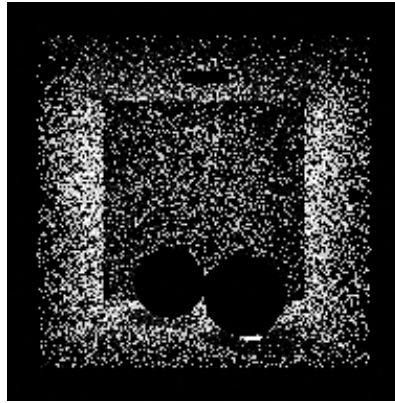
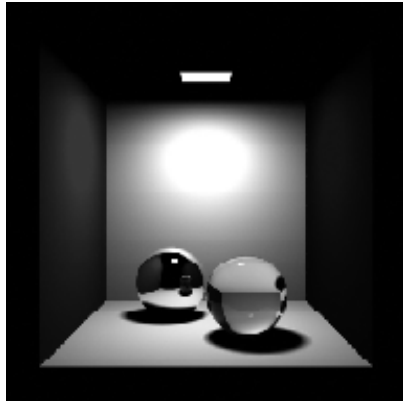
## Photon Map

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



## Pass 1: shoot photons

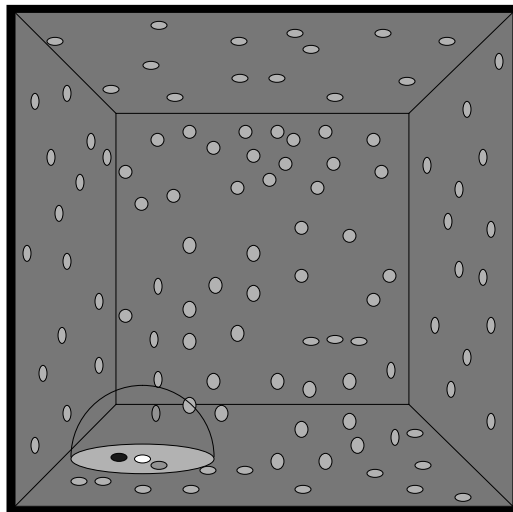
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## Pass 2: viewing ray (naive)

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- Search for  $N$  closest photons
- Assume these photons hit the point we're interested in
- Compute estimate

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## Indirect Visualization

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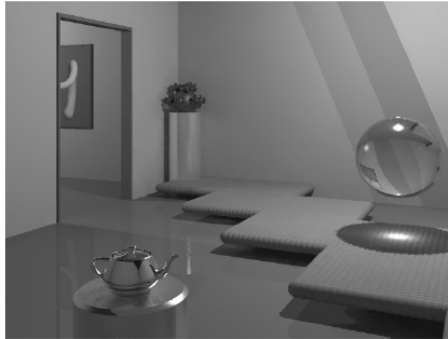


Figure 3: The Museum scene



Figure 4: Direct visualization of the global photon map in the Museum scene

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## Photon Map

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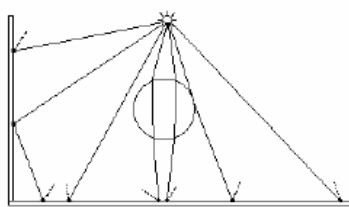
- Break lighting into several components
  - $LDS^*E$  and  $LS^*E$  paths (direct + specular)
  - $LS+DS^*E$  paths (caustic)
  - $L(S|D)*DDS^*E$  paths (indirect)
- Use a different technique for each component
  - Note that  $S^*E$  part is common – always start by tracing rays from the eye through specular bounces to a diffuse surface (or a light)

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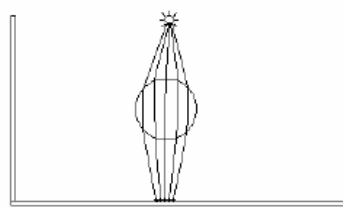
## 2 Photon Maps

- Global Photon Map

- Low resolution
  - Captures low frequency variation in shading
- Will not be viewed directly
- Not good enough for caustics and specular



Global photon map



Caustic photon map

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### Reflection Equation Decomposition

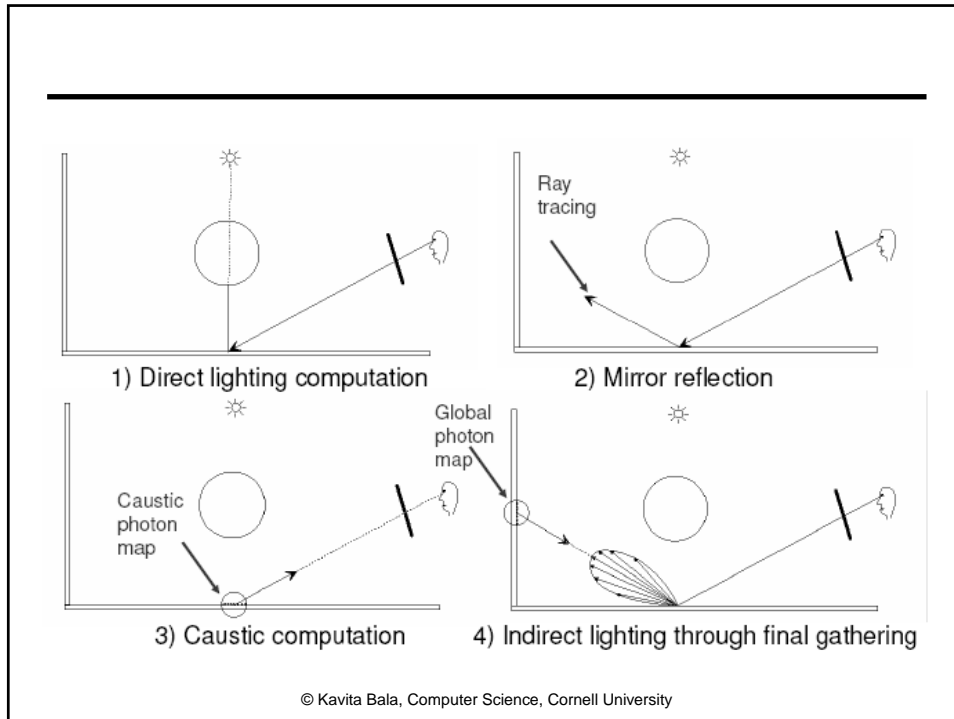
$$f_r(\underline{\omega}_i, \underline{x}, \underline{\omega}_o) = f_{r,spec}(\underline{\omega}_i, \underline{x}, \underline{\omega}_o) + f_{r,diffuse}(\underline{\omega}_i, \underline{x}, \underline{\omega}_o)$$

$$L(\underline{x}, \underline{\omega}_i) = L_{direct}(\underline{x}, \underline{\omega}_i) + L_{caustic}(\underline{x}, \underline{\omega}_i) + L_{indirect}(\underline{x}, \underline{\omega}_i)$$

$$L(\underline{x}, \underline{\omega}_o) = \int_{\Omega_+} f_r(\underline{\omega}_i, \underline{x}, \underline{\omega}_o) L(\underline{x}, \underline{\omega}_i) \cos \theta_i d\omega_i =$$

- 1)  $\int_{\Omega_+} f_r(\underline{\omega}_i, \underline{x}, \underline{\omega}_o) L_{direct}(\underline{x}, \underline{\omega}_i) \cos \theta_i d\omega_i +$
- 2)  $\int_{\Omega_+} f_{r,spec}(\underline{\omega}_i, \underline{x}, \underline{\omega}_o) (L_{caustic}(\underline{x}, \underline{\omega}_i) + L_{indirect}(\underline{x}, \underline{\omega}_i)) \cos \theta_i d\omega_i +$
- 3)  $\int_{\Omega_+} f_{r,diffuse}(\underline{\omega}_i, \underline{x}, \underline{\omega}_o) L_{caustic}(\underline{x}, \underline{\omega}_i) \cos \theta_i d\omega_i +$
- 4)  $\int_{\Omega_+} f_{r,diffuse}(\underline{\omega}_i, \underline{x}, \underline{\omega}_o) L_{indirect}(\underline{x}, \underline{\omega}_i) \cos \theta_i d\omega_i$

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## Questions

- What is in the map?
- How to compute it?
- Store in what data structure?
- How to use it?

## Pass 1

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- Shoot some number of photons
  - Magic parameter
  - Photons carry flux
- Sample from the light based on light source
- Photon hits surface
  - Pure specular: reflect
  - Diffuse, glossy: Deposit and continue
- How to continue?

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## Diffuse photon interaction

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- Material with reflectivity  $d$ ;  $p = d$
- $u = \text{random}()$
- If ( $u < p$ )
  - Reflect photon with power  $\phi$
- Else
  - Absorb photon

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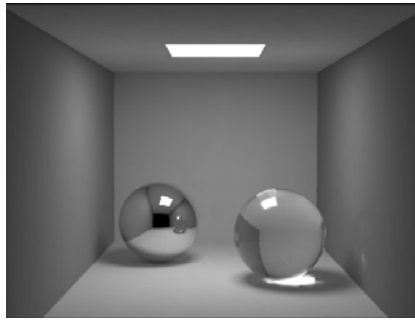
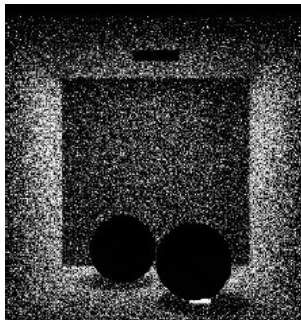
## Glossy Interaction

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- Discrete pdf  $q_1, q_2, q_3$   $p_1 + p_2 + p_3 = 1$   
 $p_1 = \rho_d$   
 $p_2 = \rho_s$

$$PhotonPath = \left\{ \begin{array}{l} diffuse \mid u < p_1 \\ specular \mid p_1 < u < p_1 + p_2 \\ absorb \mid p_1 + p_2 < u < 1 \end{array} \right\}$$

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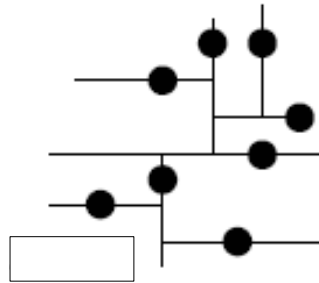


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## Efficiency

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- Want  $k$  nearest photons
  - Use *Left-balanced kd-tree*
  - Heap representation to avoid pointing to sub-trees



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## Kd-tree

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- Compactness
  - Location of photon
  - Flux
  - Incoming direction

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## Search algorithm

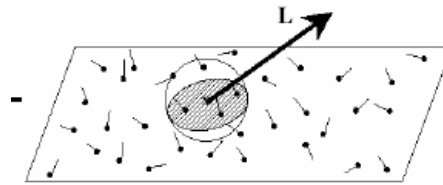
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- Want k nearest photons
- Or want photons within some distance d
  
- Start at root of kd-tree: check left first or right first based on location of x
  - Check other side if not enough on this side
  
- Keep heap of photons (based on distance from query point)
  - Can check root to see which is farthest photon
  - If find a new photon, remove the farthest one

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## Estimate

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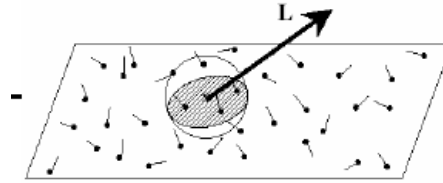
$$L_r(x \rightarrow \Theta) = \int f_r(x, \Theta \leftrightarrow \Psi) L_i(x \leftarrow \Psi) \cdot (n \cdot \Psi) \cdot d\omega_\Psi$$

$$L_i(x \leftarrow \Psi) = \frac{d^2 P_i(x, \Psi)}{dA_i(n \cdot \Psi) d\omega_\Psi}$$

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## Estimate

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$$L_r(x \rightarrow \Theta) = \int f_r(x, \Theta \leftrightarrow \Psi) \frac{d^2 P_i(x, \Psi)}{dA_i(n, \Psi) d\omega_\Psi} \cdot (n \cdot \Psi) \cdot d\omega_\Psi$$

$$L_r(x \rightarrow \Theta) = \int f_r(x, \Theta \leftrightarrow \Psi) \frac{d^2 P_i(x, \Psi)}{dA_i}$$

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## Estimate

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- Estimation from map:
  - Find nearest neighbors (ignore those with very dissimilar normal)
  - Compute estimate of outgoing radiance

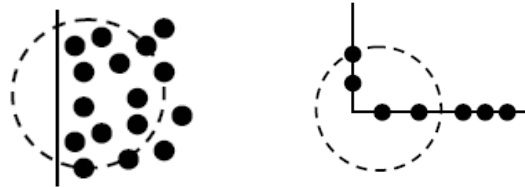
$$L_r(x \rightarrow \Theta) \approx \sum_{p=1}^n f_r(x, \Theta \leftrightarrow \Psi_p) \frac{\Delta P_p(x, \Psi_p)}{\Delta A}$$

$$L_r(x \rightarrow \Theta) \approx \frac{1}{\pi r^2} \sum_{p=1}^n f_r(x, \Theta \leftrightarrow \Psi_p) \Delta P_p(x, \Psi_p)$$

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## Is the area right?

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## Filtering

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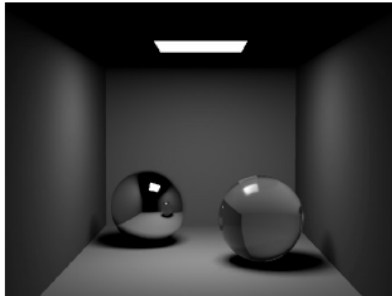
- Visualizing maps as is create noisy images
  - Need EXTREMELY large amount of photons
- Filtering techniques can be used with different type of kernels
  - Apply a Gaussian (or other kernels) on the photon map results
    - Particularly used for caustics

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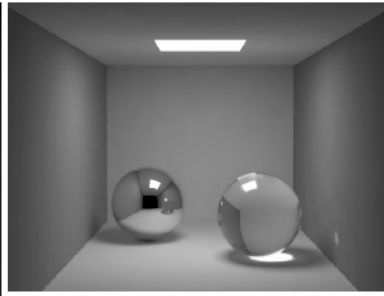
## Results

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200,000 and 50,000 photons in the global and caustic maps



Ray traced image  
(direct lighting only)



Full global illumination

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## Results

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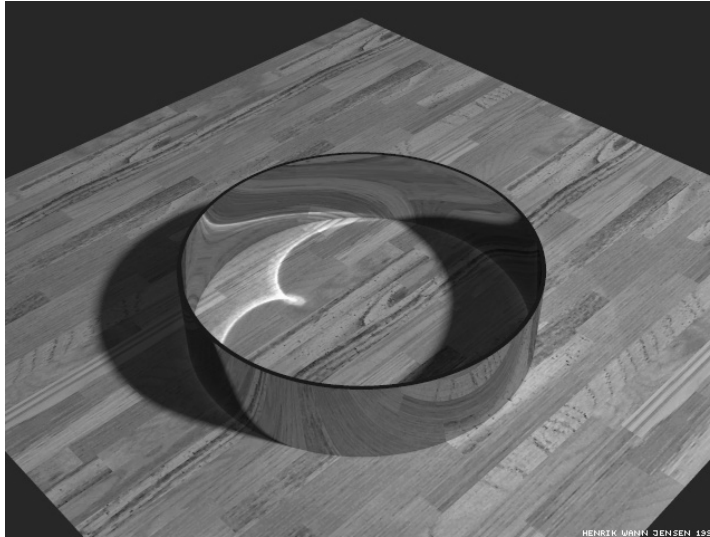
500,000 photons in both the global and caustic maps,  
100 nearest photons used in the radiance estimate



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## Photon Map Results

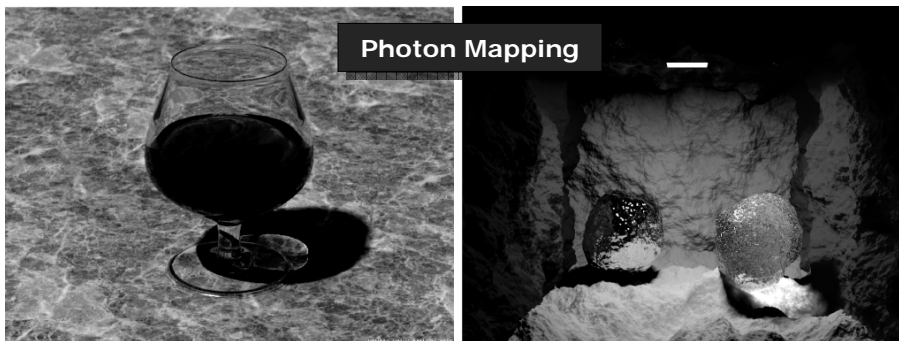
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## Showcase pictures

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- Geometry-independent

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# Examples

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# Photon Map Failure

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## With Better Sampling

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## Photon Map Summary

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- Combines various techniques depending on the situation
- Produces very good results reasonably efficiently
- Geometric problems can still cause bleeding
- Sampling from the light can cause problems if most of the light's power fails to reach the image

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