### Monte Carlo Ray Tracing & Irradiance Caching & Photon Mapping

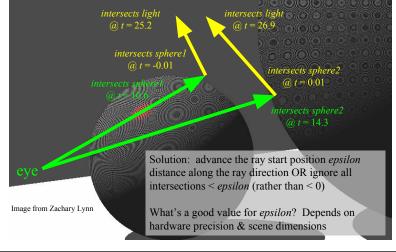
#### Announcements: Quiz

- On Friday (3/10), in class
- One 8.5x11 sheet of notes allowed
- Sample quiz (from a previous year) on website
- Focus on "reading comprehension" and material for Homeworks 0, 1, & 2

#### Announcements: Final Projects

- Everyone should post two ideas for a final project on LMS ("due" Monday 3/20 @ 11:59pm)
- Connect with potential teammates (teams of 2 strongly recommended)
- Start finding & reading background papers
- Proposal & summary of background research are due April 3<sup>rd</sup>
- See webpage for details on brainstorming post, proposal, and overall project requirements





#### Reading for Today

• "The Rendering Equation", Kajiya, SIGGRAPH 1986



#### Reading for Today

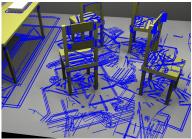
"Implicit Visibility and Antiradiance for Interactive

Global Illumination"
Dachsbacher,
Stamminger,
Drettakis, and
Durand
Siggraph 2007



#### Reading for Today

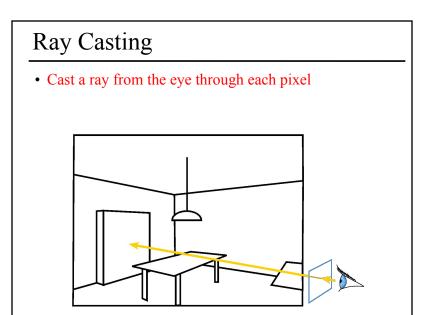


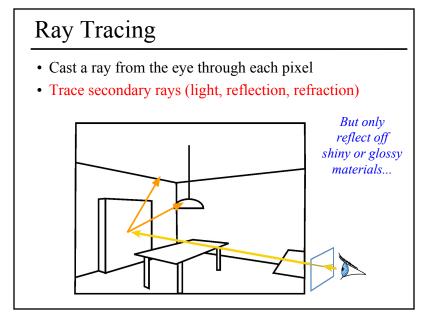


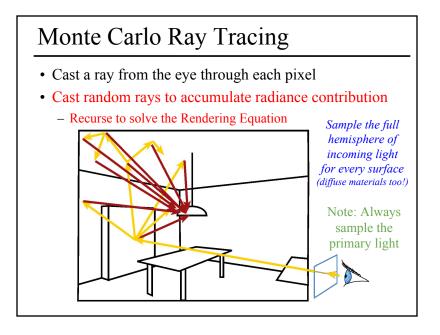
"Fast and Accurate Hierarchical Radiosity Using Global Visibility" Durand, Drettakis, & Puech 1999

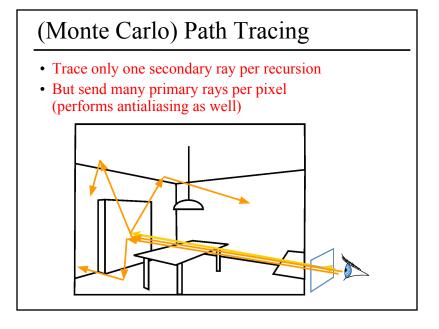
#### Today

- Ray Casting vs. Ray Tracing vs.
   Monte-Carlo Ray Tracing vs. Path Tracing
- Irradiance Caching
- Photon Mapping
- Ray Grammar
- Monte-Carlo Integration
- Importance Sampling

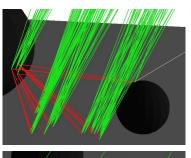








#### Ray Tracing vs. Path Tracing



- 2 bounces
- 5 glossy samples
- 5 shadow samples

How many rays cast per pixel?

- 1 main ray + 5 shadow rays +
- 5 glossy rays + 5x5 shadow rays +
- 5\*5 glossy rays + 5x5x5 shadow rays
- = 186 rays



How many 3 bounce paths can we trace per pixel for the same cost?

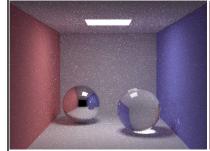
186 rays / 8 ray casts per path =  $\sim$ 23 paths

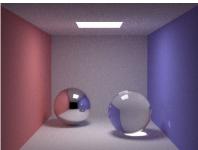
Which will probably have less error?

#### Questions?

10 paths/pixel

#### 100 paths/pixel





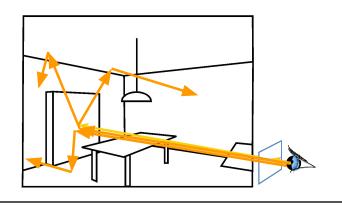
Images from Henrik Wann Jensen

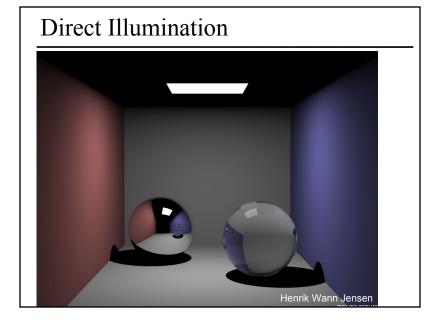
#### Today

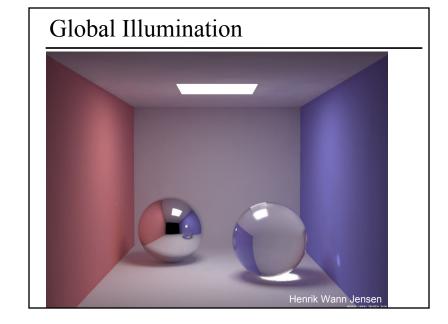
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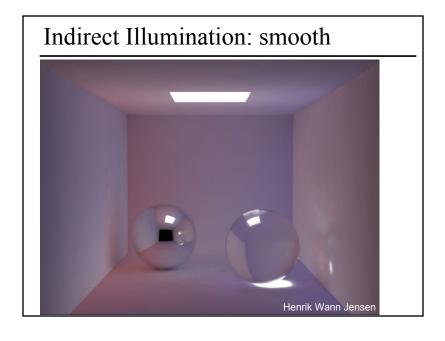
#### Path Tracing is costly

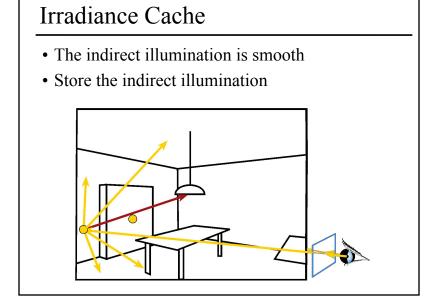
• Needs tons of rays per pixel





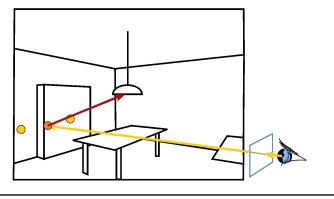






#### Irradiance Cache

- Interpolate nearby cached values
- But do full calculation for direct lighting



## Irradiance Cache

Henrik Wann Jensen

#### Questions?

- Why do we need "good" random numbers?
  - With a fixed random sequence, we see the structure in the error

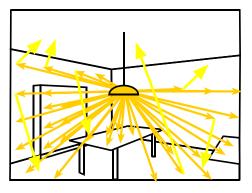


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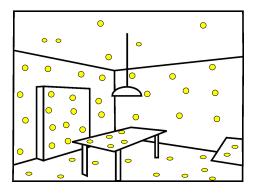
#### **Photon Mapping**

- Preprocess: cast rays from light sources
  - independent of viewpoint



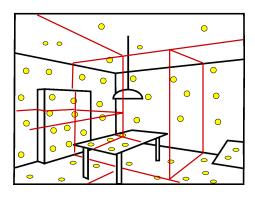
#### **Photon Mapping**

- Store photons
  - position + light power + incoming direction



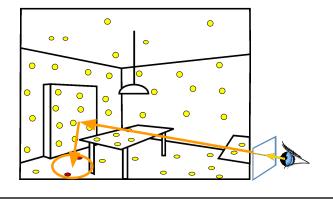
#### Storing the Photon Map

- Efficiently store photons for fast access
- Use hierarchical spatial structure (kd-tree)

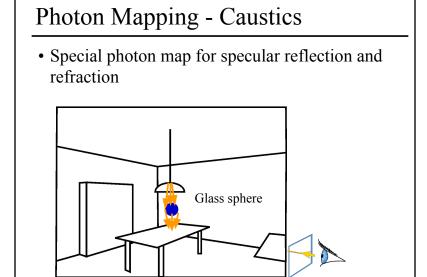


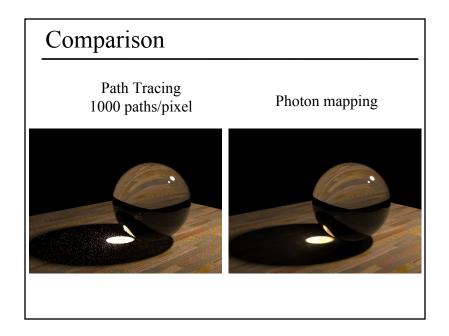
#### Rendering with Photon Map

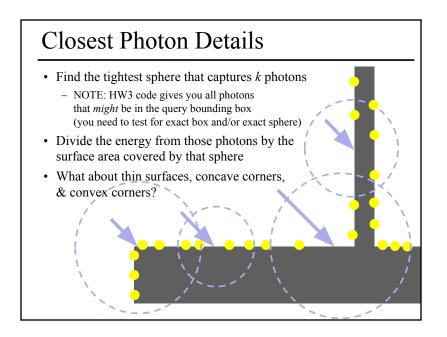
- Cast primary rays
- For secondary rays: reconstruct irradiance using k closest photons
  Combine with irradiance caching and other techniques



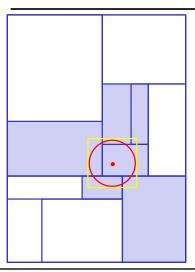
# Photon Map Results







#### HW3: Photons in the k-d tree



- You start with query point & radius (red)
- You give the KDTree::CollectPhotonsInBox function a bounding box (yellow)
- The algorithm finds all k-d tree cells that overlap with bounding box (blue)
- The function returns all photons in those cells
- You need to discard all photons not in your original query radius

#### Readings for Tuesday after break:

• "Rendering Caustics on Non-Lambertian Surfaces", Henrik Wann Jensen, *Graphics Interface* 1996.



• "Global Illumination using Photon Maps", Henrik Wann Jensen, *Rendering Techniques* 1996.





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#### Ray Grammar

• Classify local interaction:

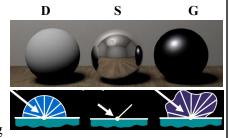
E = eye

L = light

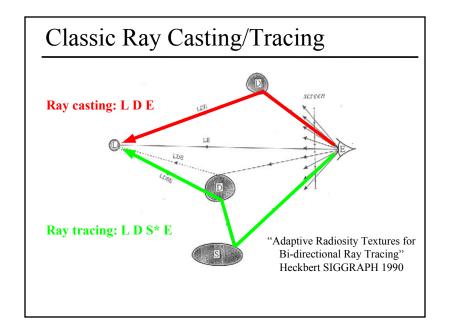
S = perfect specular reflection or refraction

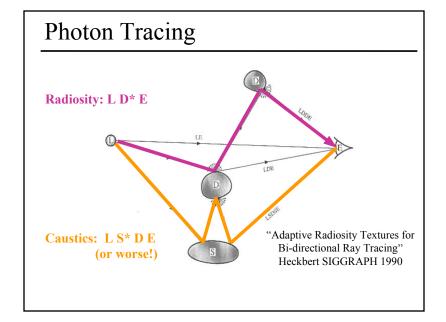
G = glossy scattering

D = diffuse scattering



From Dutre et al.'s slides





#### Questions?

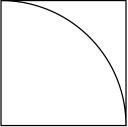


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  - Probabilities and Variance
  - Analysis of Monte-Carlo Integration
- Importance Sampling

#### Monte-Carlo Computation of $\pi$

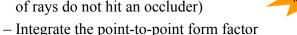
- Take a random point (x,y) in unit square
- Test if it is inside the ½ disc
  - $\text{ Is } x^2 + y^2 < 1?$
- Probability of being inside disc?
  - area of ¼ unit circle / area of unit square = π /4

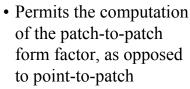


- $\pi \approx 4$  \* number inside disc / total number
- The error depends on the number or trials

#### Use MC to calculate Form Factor

- Cast *n* rays between the two patches
  - Compute visibility (what fraction of rays do not hit an occluder)







### ch posed

#### Convergence & Error

- Let's compute 0.5 by flipping a coin:
  - 1 flip: 0 or 1
    - $\rightarrow$  average error = 0.5
  - -2 flips: 0, 0.5, 0.5 or 1
    - $\rightarrow$  average error = 0. 25
  - 4 flips: 0 (\*1),0.25 (\*4), 0.5 (\*6), 0.75(\*4), 1(\*1)
    - $\rightarrow$  average error = 0.1875
- Unfortunately, doubling the number of samples does not double accuracy

#### Review of (Discrete) Probability

- Random variable can take discrete values x<sub>i</sub>
- Probability  $p_i$  for each  $x_i$

$$0 < p_i < 1, \sum p_i = 1$$

- Expected value  $E(x) = \sum_{i=1}^{n} p_i x_i$
- Expected value of function of random variable
  - $f(x_i)$  is also a random variable

$$E[f(x)] = \sum_{i=1}^{n} p_i f(x_i)$$

#### Variance & Standard Deviation

- Variance  $\sigma^2$ : deviation from expected value
- Expected value of square difference

$$\sigma^2 = E[(x - E[x])^2] = \sum_i (x_i - E[x])^2 p_i$$

• Also

$$\sigma^2 = E[x^2] - (E[x])^2$$

• Standard deviation  $\sigma$ : square root of variance (notion of error, RMS)

#### Monte Carlo Integration

- Turn integral into finite sum
- Use *n* random samples
- As *n* increases...
  - Expected value remains the same
  - Variance decreases by n
  - Standard deviation (error) decreases by  $\frac{1}{\sqrt{n}}$
- Thus, converges with  $\frac{1}{\sqrt{n}}$

#### Advantages of MC Integration

- Few restrictions on the integrand
  - Doesn't need to be continuous, smooth, ...
  - Only need to be able to evaluate at a point
- Extends to high-dimensional problems
  - Same convergence
- Conceptually straightforward
- Efficient for solving at just a few points

#### Disadvantages of MC Integration

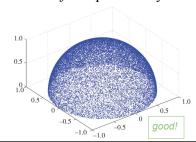
- Noisy
- Slow convergence
- Good implementation is hard
  - Debugging code
  - Debugging math
  - Choosing appropriate techniques
- Punctual technique, no notion of smoothness of function (e.g., between neighboring pixels)

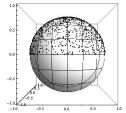
#### Today

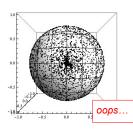
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  - Stratified Sampling
  - Importance Sampling

#### Domains of Integration

- Pixel, lens (Euclidean 2D domain)
- Time (1D)
- Hemisphere: Work needed to ensure *uniform* probability



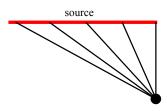




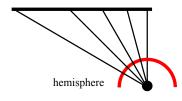
#### Example: Light Source

- We can integrate over surface or over angle
- But we must be careful to get probabilities and integration measure right!

Sampling the source uniformly



Sampling the hemisphere uniformly

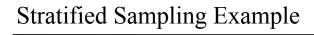


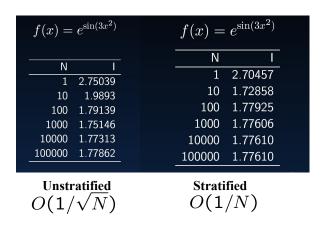
#### Stratified Sampling

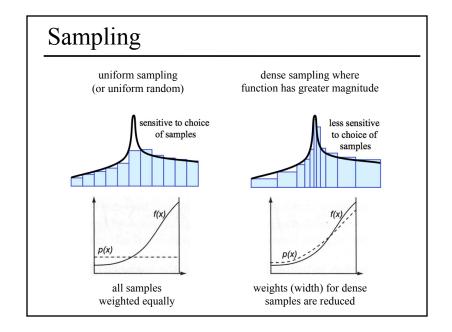
- With uniform sampling, we can get unlucky
  - E.g. all samples in a corner
- To prevent it, subdivide domain  $\Omega$  into non-overlapping regions  $\Omega_i$ 
  - Each region is called a stratum
- Take one random samples per  $\Omega_{_{i}}$









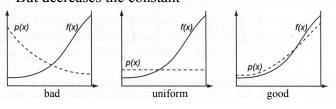


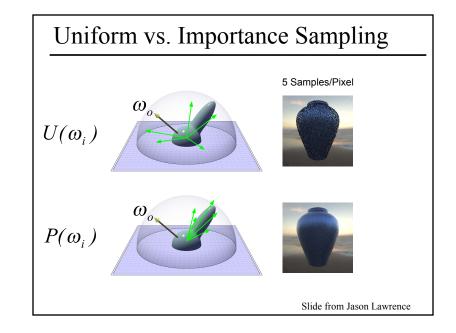
#### Importance Sampling

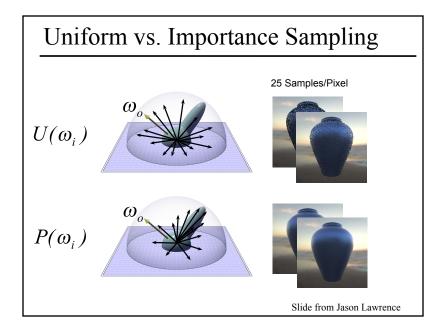
$$\langle I \rangle = \frac{1}{N} \sum_{i=1}^{N} \frac{f(x_i)}{p(x_i)}$$

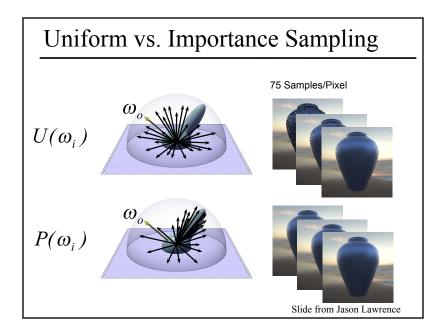
Slide from Henrik Wann Jensen

- Choose p wisely to reduce variance
  - Want to use a p that resembles f
  - Does not change convergence rate (still sqrt)
  - But decreases the constant









#### **Bidirectional Path Tracing**

• "A Theoretical Framework for Physically Based Rendering", Lafortune and Willems, Computer Graphics Forum, 1994.





Figure B: An indirectly illuminated scene rendered using path tracing and bidirectional path tracing respectively. The latter method results in visibly less noisefor the same amount of work.

#### Questions?



Naïve sampling strategy

Optimal sampling strategy

Veach & Guibas "Optimally Combining Sampling Techniques for Monte Carlo Rendering" SIGGRAPH 95