

Towards Path Tracing

Pixel Filtering and Multidimensional Monte Carlo Integration



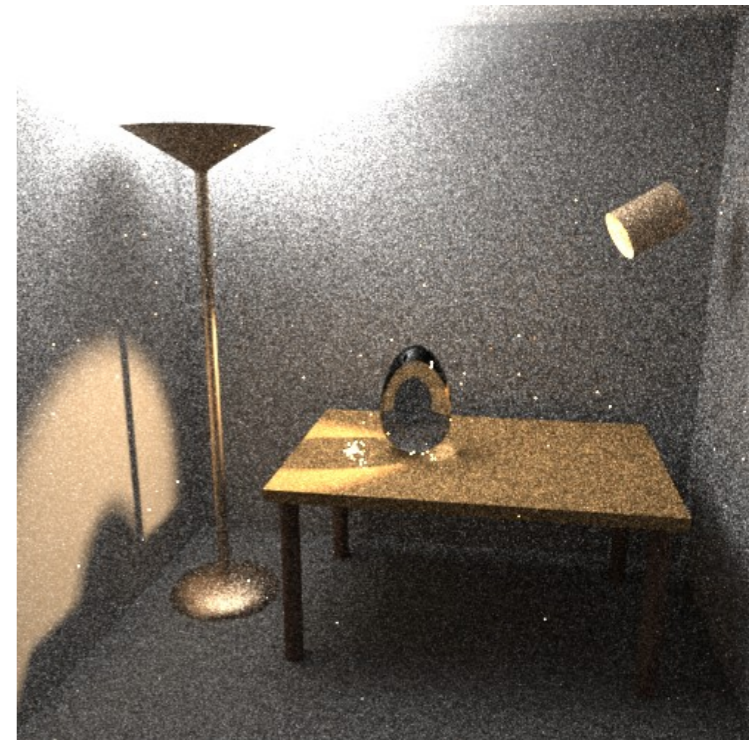
Aalto CS-E5520 Spring 2022
Jaakko Lehtinen

Today

- Path Tracing

- Intro: nested vs. multidimensional integrals and pixel filtering
- Recursive sampling of rendering equation using Monte Carlo
- Direct light sampling

- Bells and whistles



NVIDIA Marbles at Night



Monte Carlo Integration

$$\int_S f(x) dx = E\left\{\frac{f(x)}{p(x)}\right\}_p$$

- Distribute samples in integration domain S according to probability density function $p(x)$
- Then integral equals the expected value of $f(x)/p(x)$

Let's Go Back to Pixel Filtering

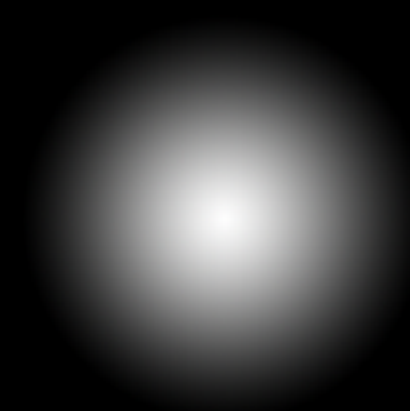
- Remember antialiasing theory from C3100?
- To reduce aliasing, we should ideally...?

Let's Go Back to Pixel Filtering

- Remember antialiasing theory from C3100?
- To reduce aliasing, we should ideally...
 1. Low-pass filter the radiance on the image plane before sampling (convolve continuous radiance function + prefilter)
 2. Then sample the low-pass filtered radiance at pixel centers
- But we found this was impossible so we turned to supersampling (average many samples in pixel)
 - There is a “proper” way to look at that as well, and here it is..
- (And separate tricks for textures)
 - MIP-maps





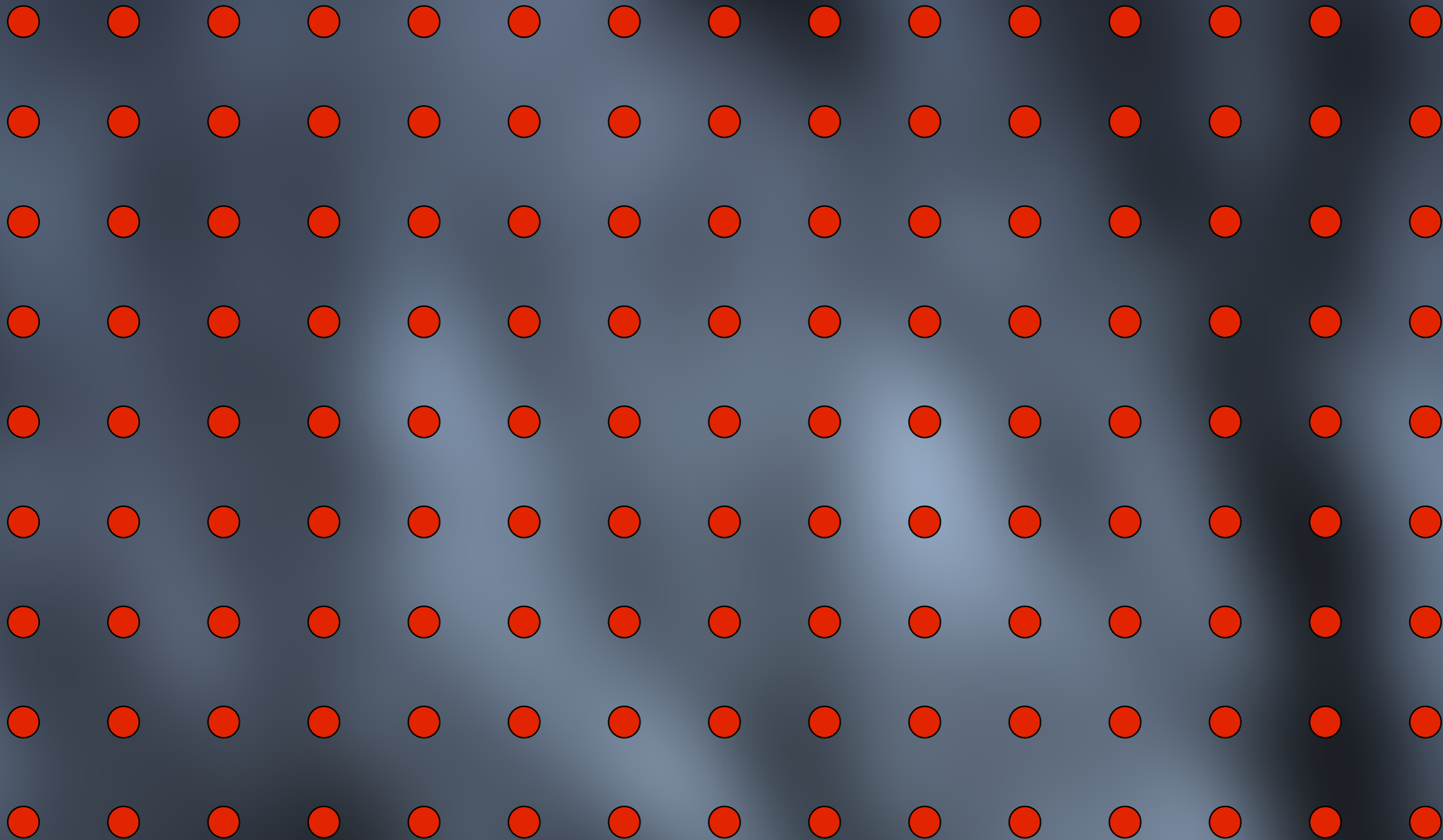


Filter $f(x-x_j, y-y_j)$ centered at pixel (x_j, y_j)



Filter $f(x-x_j, y-y_j)$ centered at pixel (x_j, y_j)
times the underlying signal

Low-pass filtered continuous image
(convolution of f and input image; we can
actually never compute this exactly)



Samples at pixel centers

Samples evaluate convolution result at pixel centers


$$\int \text{d}x \text{d}y$$

i.e., value for pixel at (x_j, y_j) is the integral of the filter times the underlying signal

Pixel Filtering

- Prefilter convolution and sampling can be combined:

$$I_j = \int_{\text{screen}} f(x - x_j, y - y_j) L(x, y) dx dy$$

- I_j is the (discrete) intensity/radiance value of j th pixel
- Here x_j, y_j are the center of pixel j , f is the *pixel filter*
 - *Yes, it's just a weighted average*

Filter Normalization

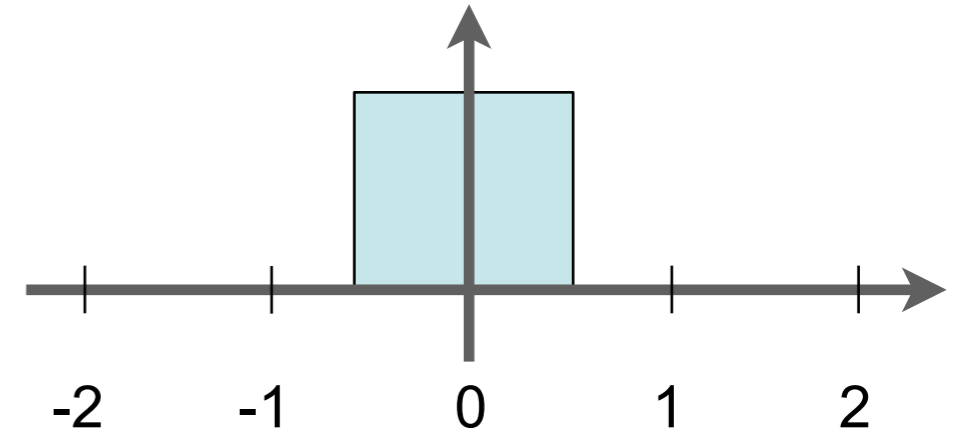
- In practice, we don't care about normalizing the filters analytically, but do it numerically instead

$$I_j = \frac{\int_{\text{screen}} f(x - x_j, y - y_j) L(x, y) dx dy}{\int_{\text{screen}} f(x - x_j, y - y_j) dx dy}$$

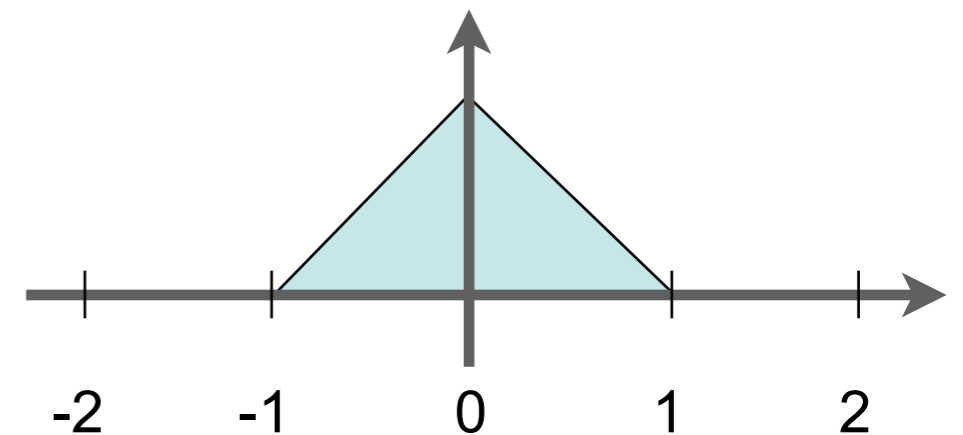
- Intuitive: when we evaluate the above using MC, we sum the “filter weights” from each sample and divide by the sum in the end
 - Note that $1/N$ cancels out as it's both above and below
 - **IMPORTANT** do it this way; don't rely on $\int f(x, y) = 1$

Common Pixel Filters, 1D profiles

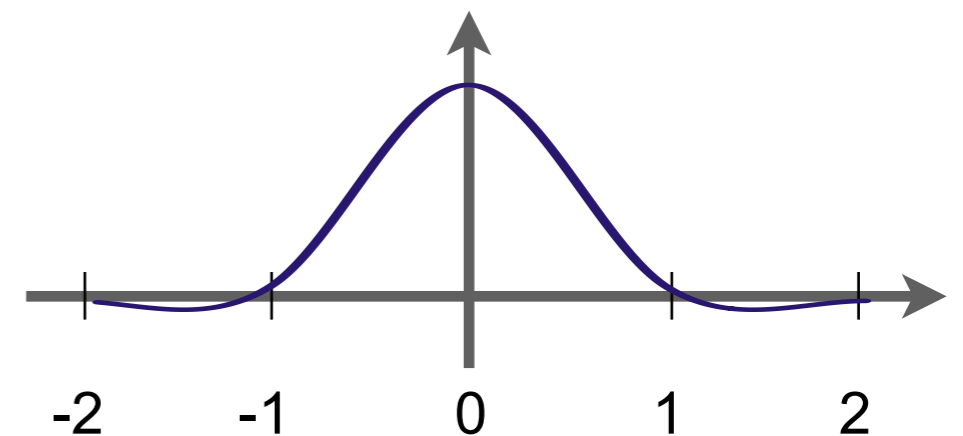
$$f_{\text{box}}(x) = \begin{cases} 1, & -0.5 \leq x \leq 0.5 \\ 0, & \text{otherwise} \end{cases}$$



$$f_{\text{tent}}(x) = \begin{cases} x + 1, & -1 \leq x \leq 0 \\ 1 - x, & 0 \leq x \leq 1 \\ 0, & \text{otherwise} \end{cases}$$



$$f_{\text{M-N}}(x) = \frac{1}{6} \begin{cases} 7|x|^3 - 12|x|^2 + \frac{16}{3} & |x| < 1 \\ -\frac{7}{3}|x|^3 + 12|x|^2 - 20|x| + \frac{32}{3} & 1 \leq |x| \leq 2 \\ 0, & \text{otherwise} \end{cases}$$



Mitchell-Netravali filter with A=1/3, B=1/3

Extension to 2D

- “Tensor product” or “separable” filters are constructed from the 1D filters by multiplication

$$f(x, y) = f(x)f(y)$$

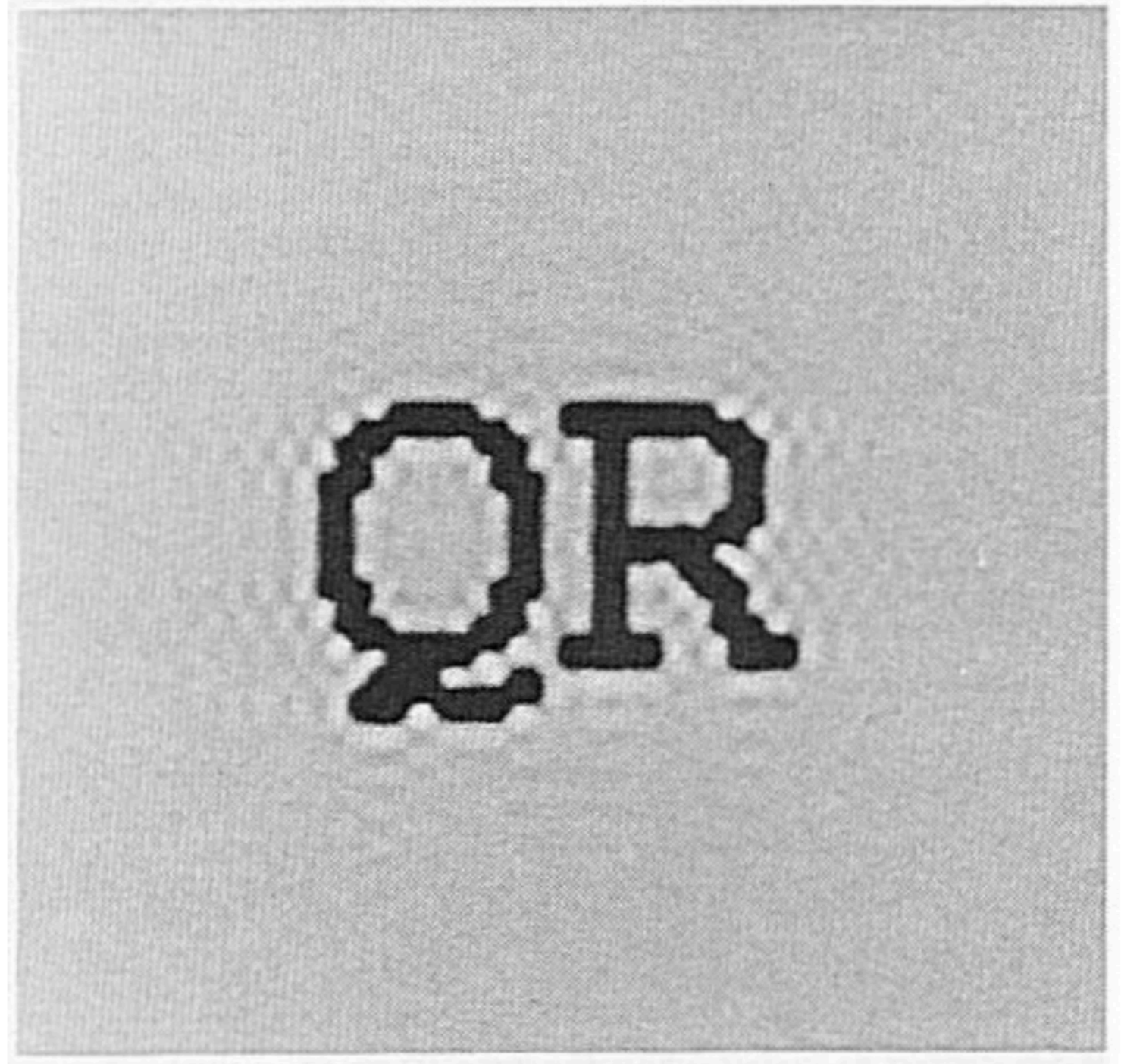
- You can also use a non-separable pyramid as a 2D filter, but there seems to be little point
- OK, one more: Gaussian

$$f_{\text{Gaussian}}^{\sigma}(x) = \exp\left\{-\frac{x^2}{2\sigma^2}\right\}$$

- Notes: sigma controls width; not normalized to unit integral!
- Never drops to zero. We usually cut the filter at 3*sigma or so.

Sinc is really not good

- From Mitchell & Netravali



Down to Business: AO

- What if each value of the original image is an integral?
- In assignment 1 you compute, for each primary hit P

$$\int_{\Omega} V(P, \omega) \cos \theta \, d\omega$$

using Monte Carlo integration

- V is a function that is 1 if the ray of a certain length is unblocked, 0 if it is blocked

Let's Combine Pixel Filter with AO

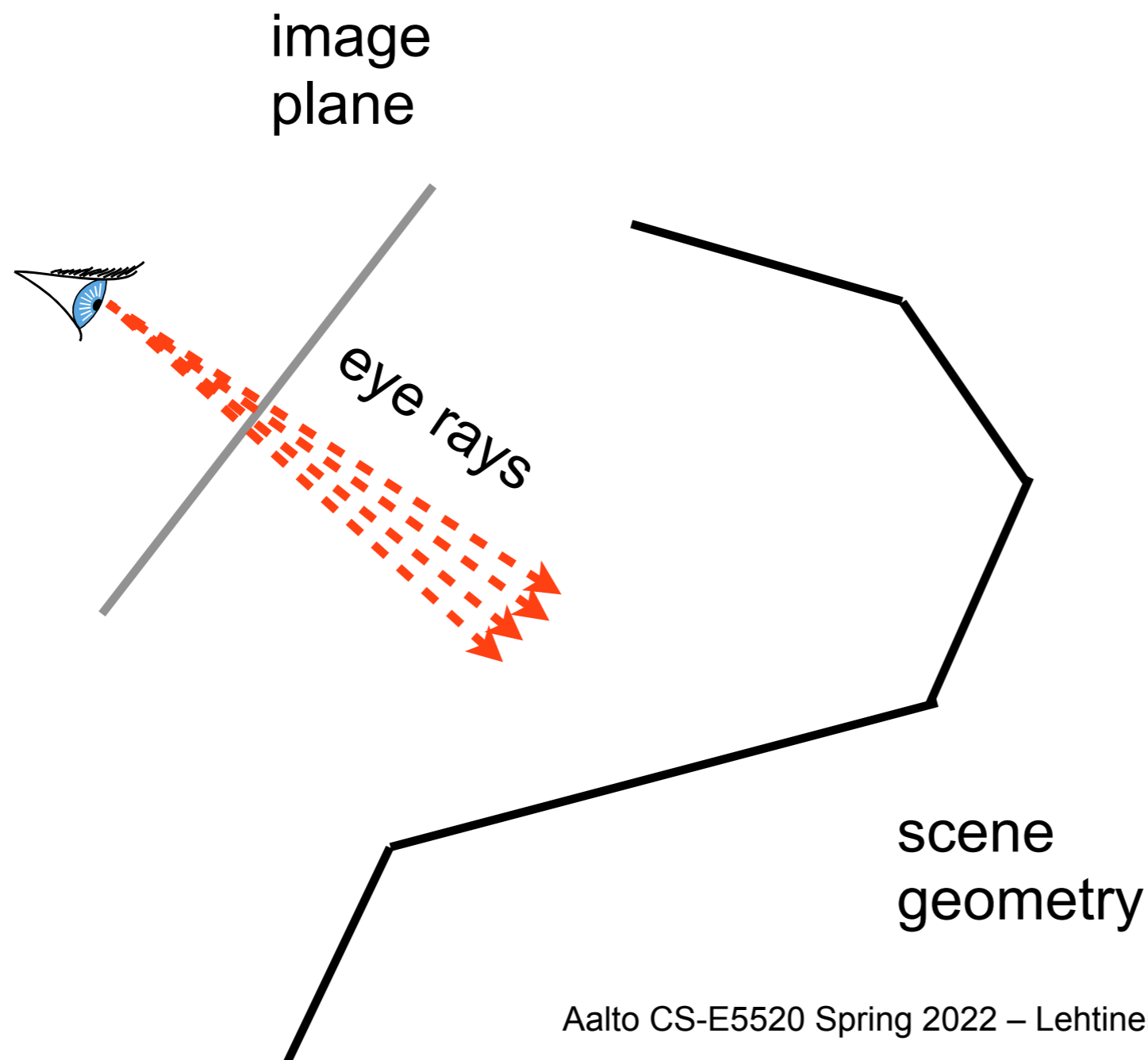
- Each pixel value given by

$$I_j = \int_{\text{screen}} f(x - x_j, y - y_j) \left(\int_{\Omega} V(P(x, y), \omega) \cos \theta d\omega \right) dx dy$$

- (Normalization not shown)
- Two nested 2D integrals
 - Outer one over the screen (2D)
 - Inner one over the hemisphere at the point P hit by ray through image coordinages x,y
 - Again, 2D (hemisphere)

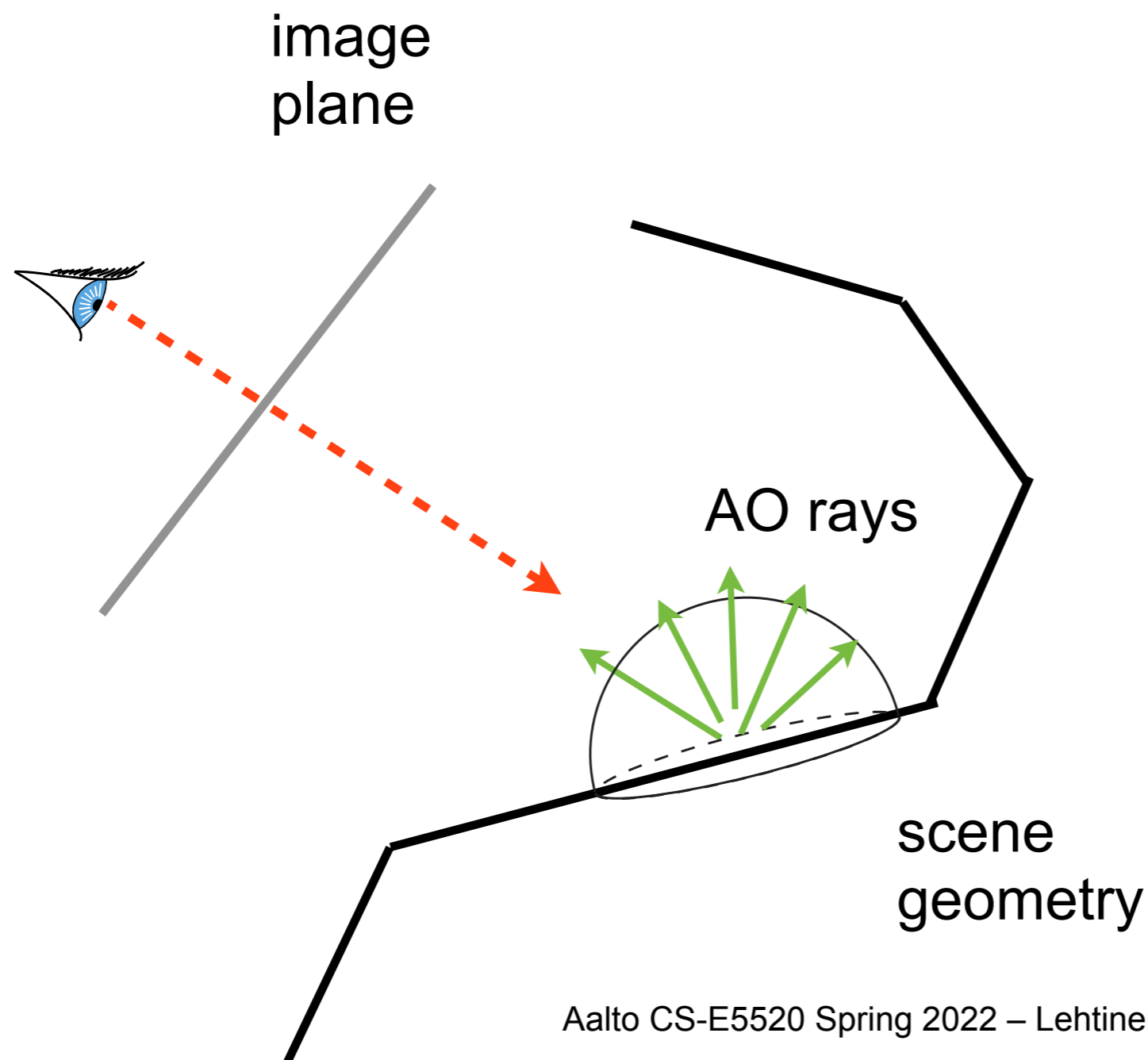
Outer Integral

$$I_j = \int_{\text{screen}} f(x - x_j, y - y_j) \left(\int_{\Omega} V(P(x, y), \omega) \cos \theta d\omega \right) dx dy$$



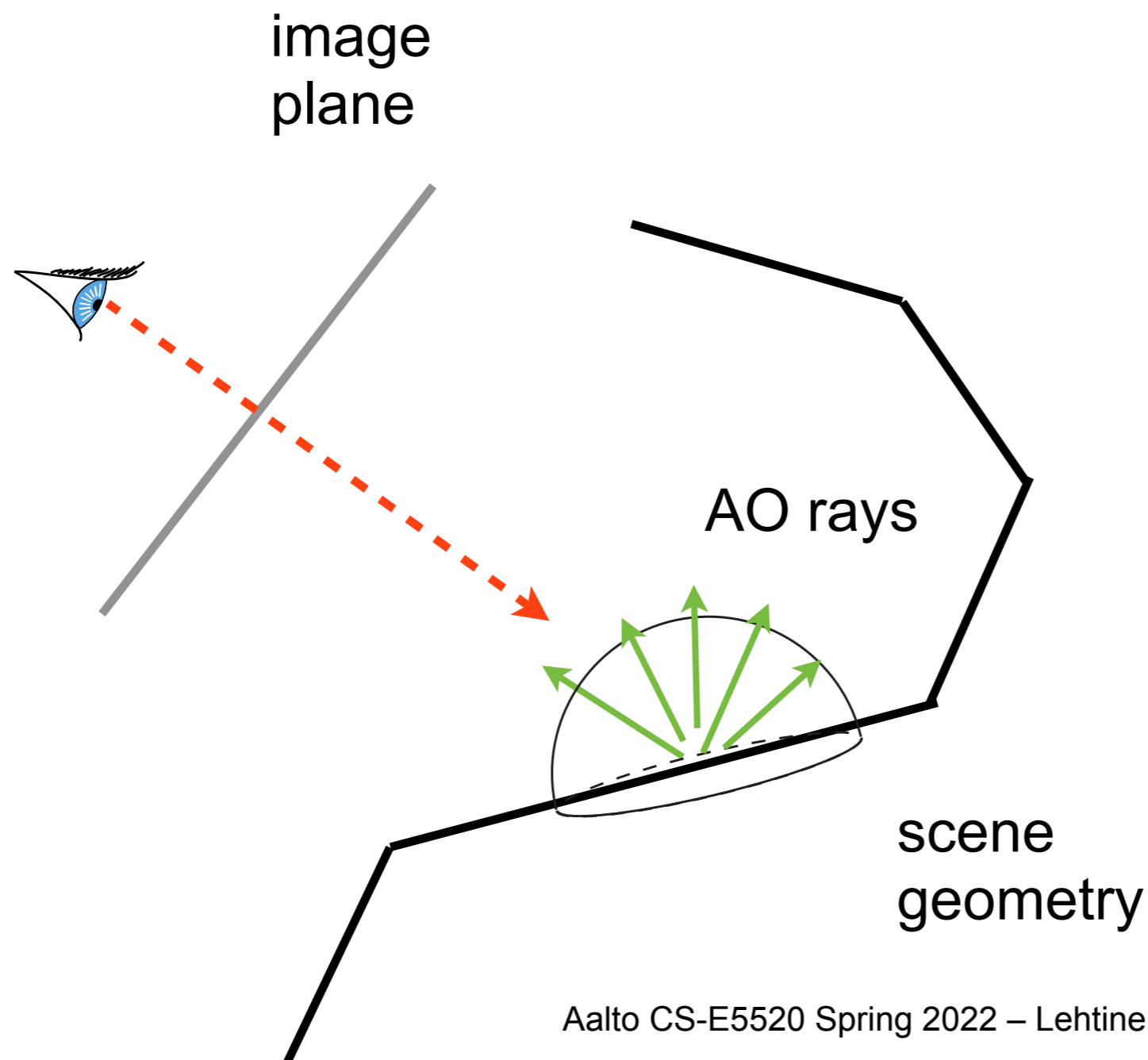
Inner Integral, for each eye ray

$$I_j = \int_{\text{screen}} f(x - x_j, y - y_j) \left(\int_{\Omega} V(P(x, y), \omega) \cos \theta d\omega \right) dx dy$$



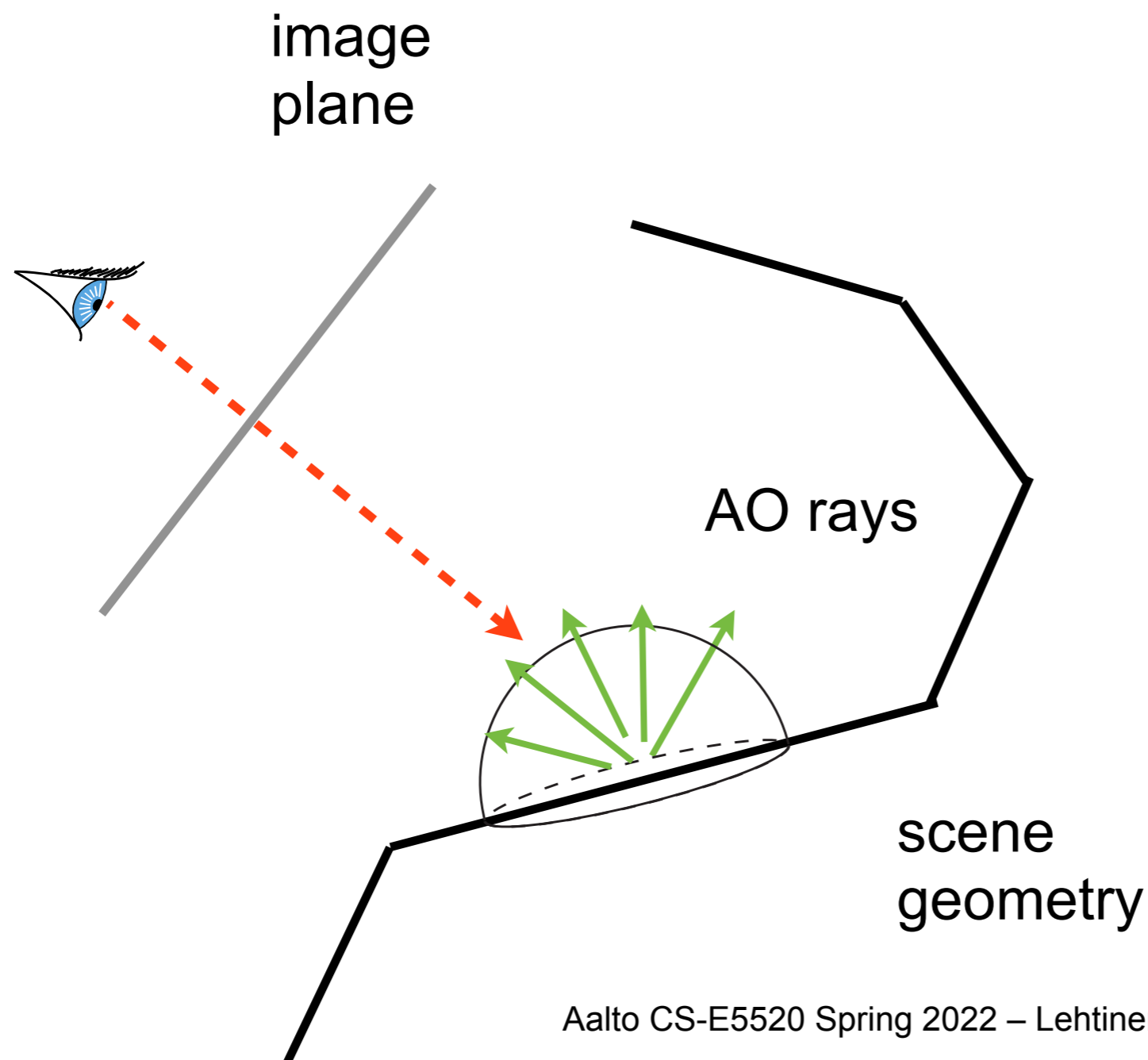
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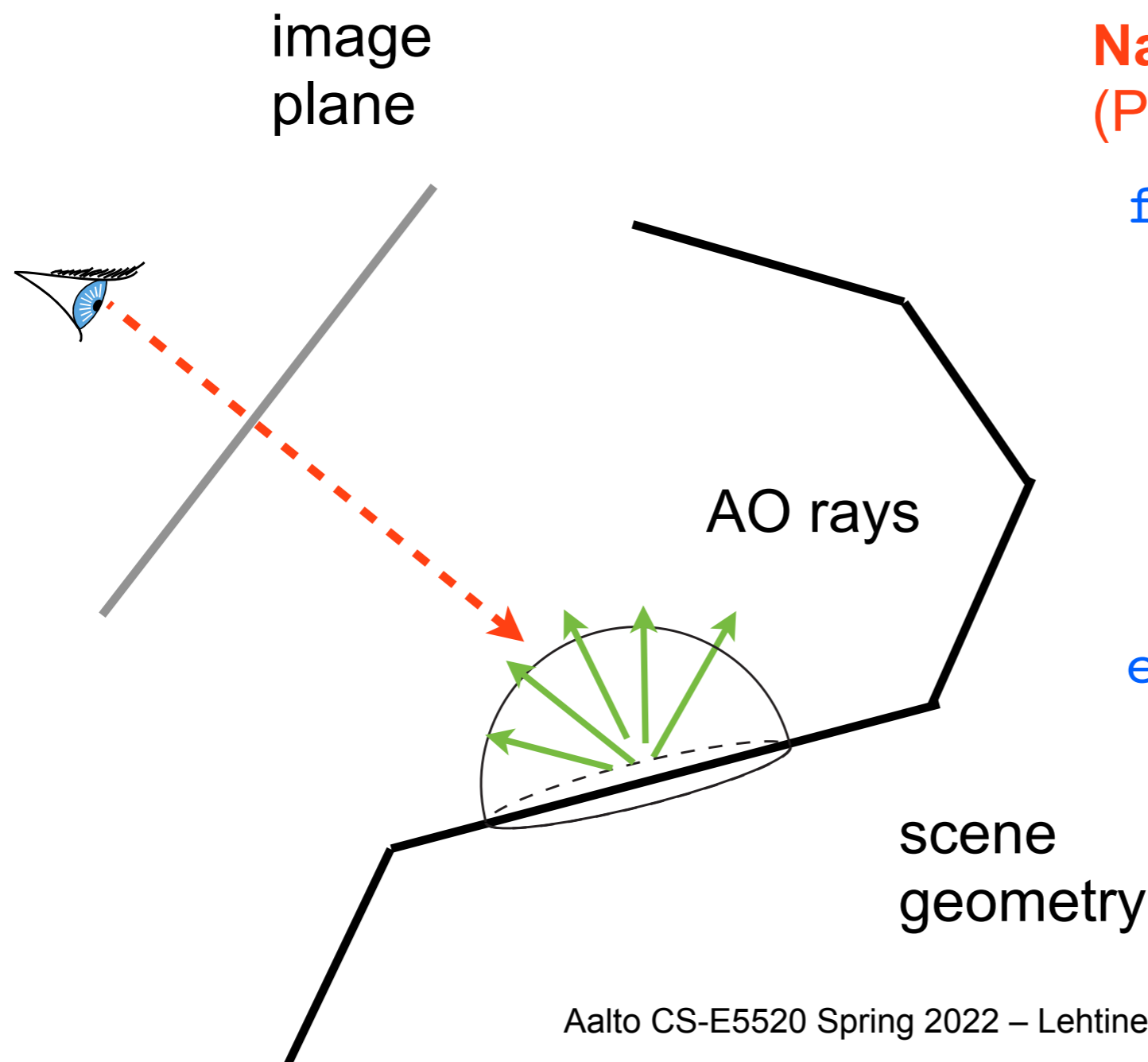
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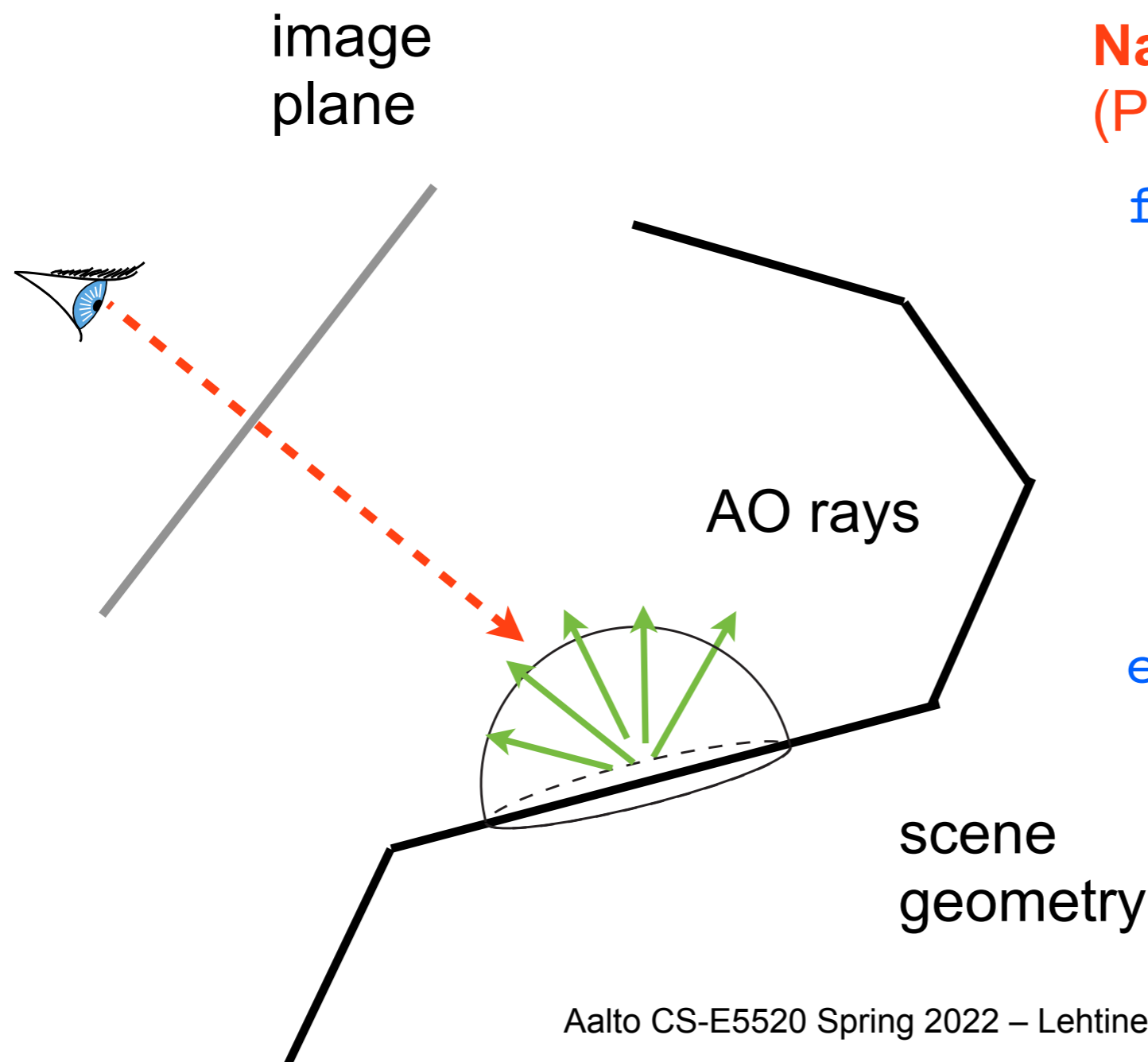


Naive MC implementation:
(PDFs, accumulation not shown)

```
for i=1 to #eyerays
  pick (x,y)
  P=castray(x,y)
  for j=1 to #aorays
    // shoot rays from P
    // etc
  end
end
```

Inner Integral, for each eye ray

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Although you do this in assn1, it makes little sense

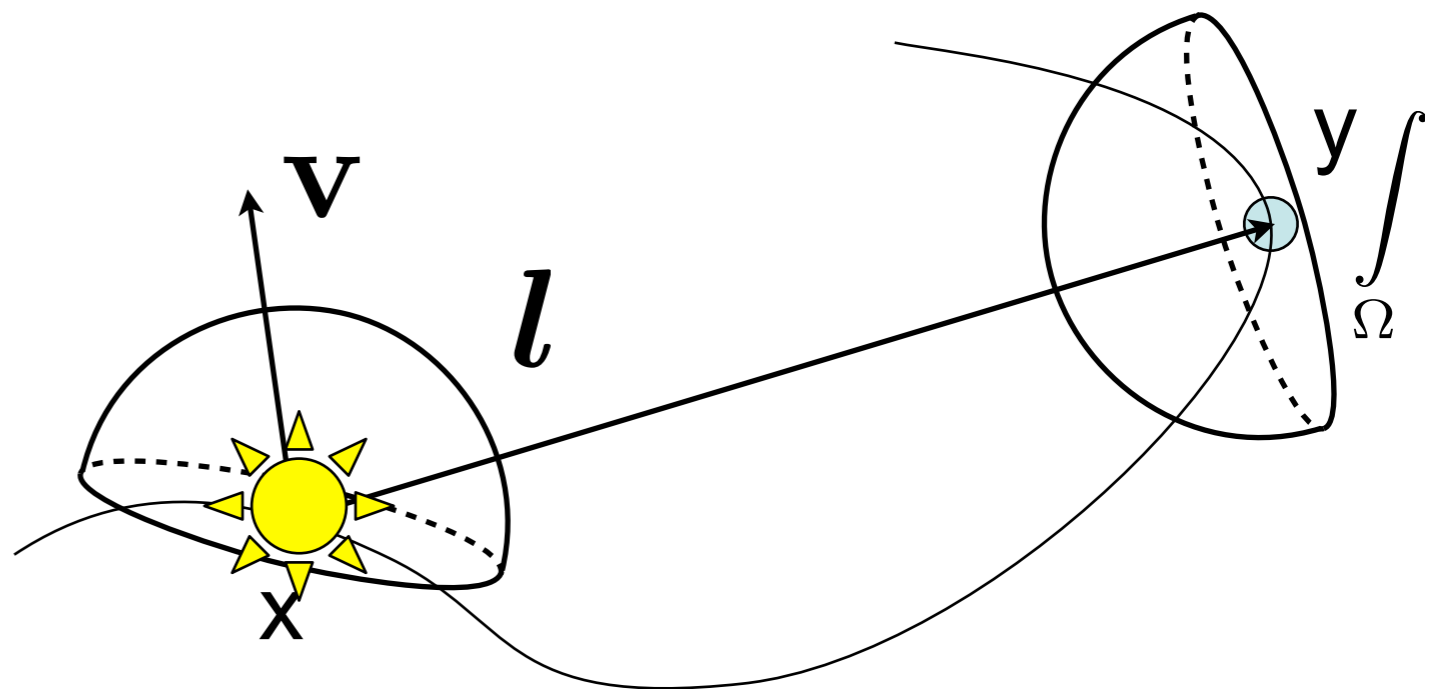
Problems

- Difficult to control number of rays cast in the pixel
 - You have two knobs to tweak
- What if we had even further integrals...?

Recap: Rendering Equation

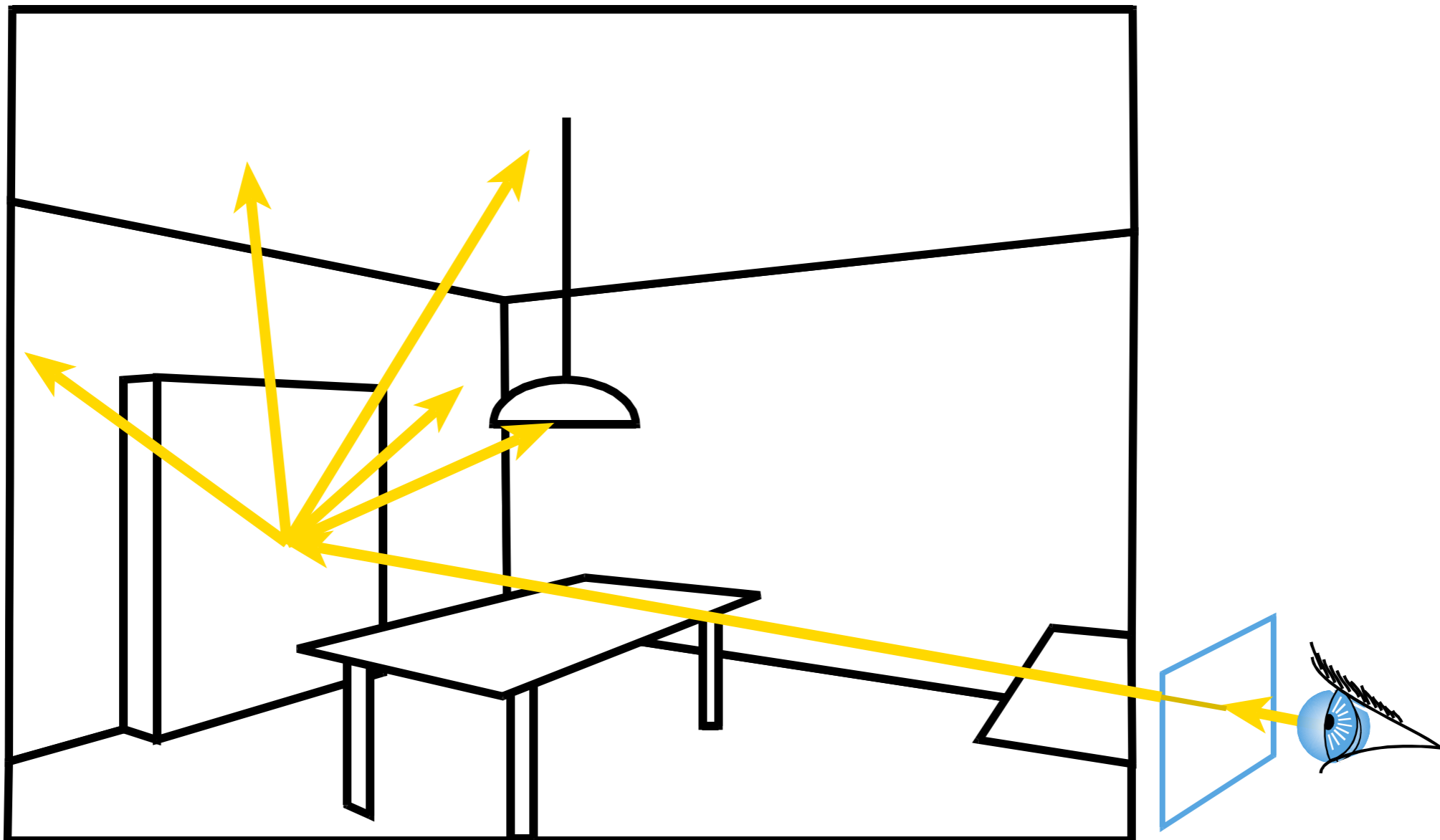
$$L(x \rightarrow \mathbf{v}) = \int_{\Omega} L(x \leftarrow \mathbf{l}) f_r(x, \mathbf{l} \rightarrow \mathbf{v}) \cos \theta \, d\mathbf{l} + E(x \rightarrow \mathbf{v})$$

**to know incoming radiance,
must know outgoing radiance
elsewhere => recursion!**



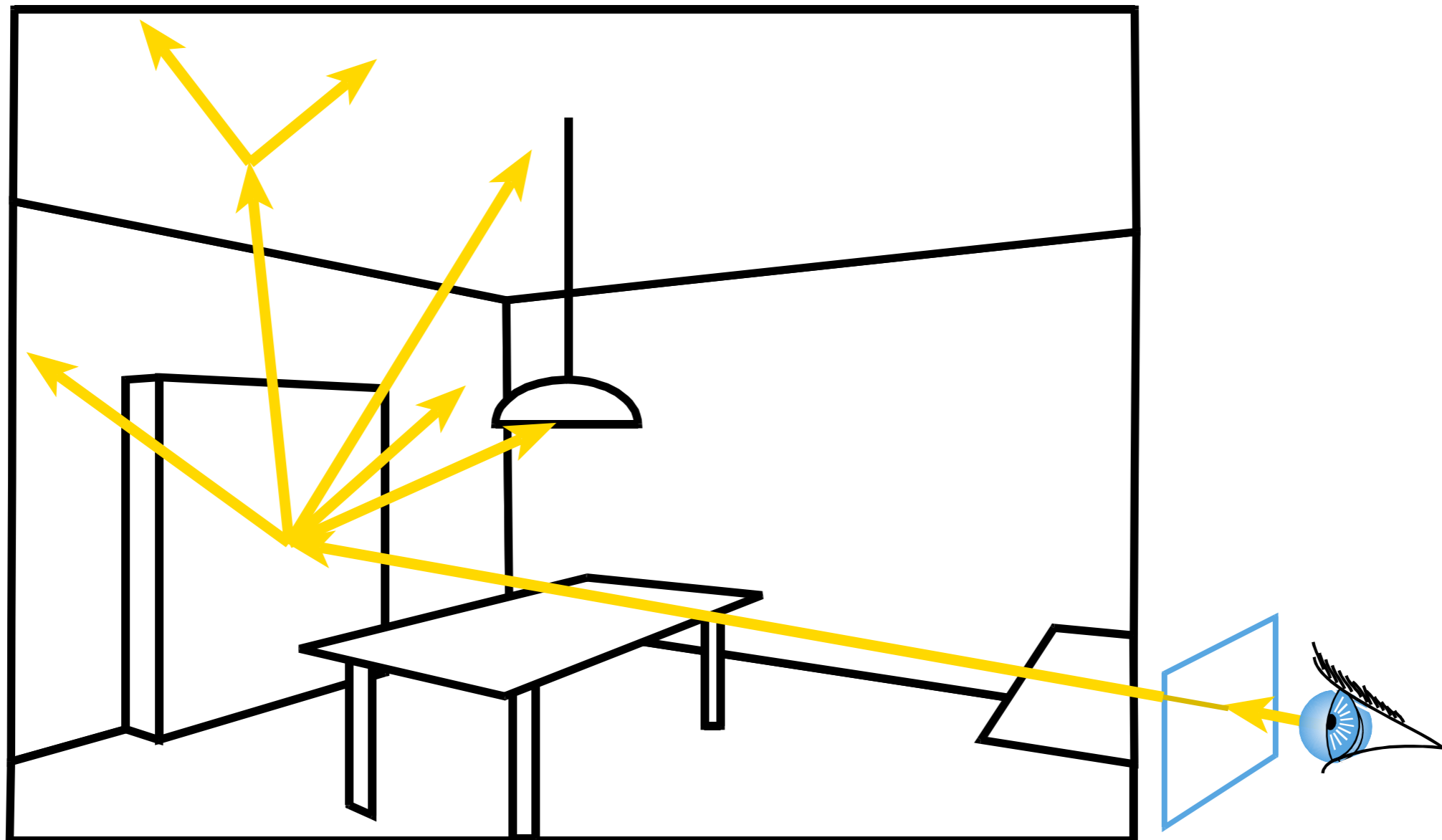
“Monte-Carlo Ray Tracing”

- Cast a ray from the eye through each pixel
- Cast N random rays from the hit point to evaluate hemispherical integral using random sampling



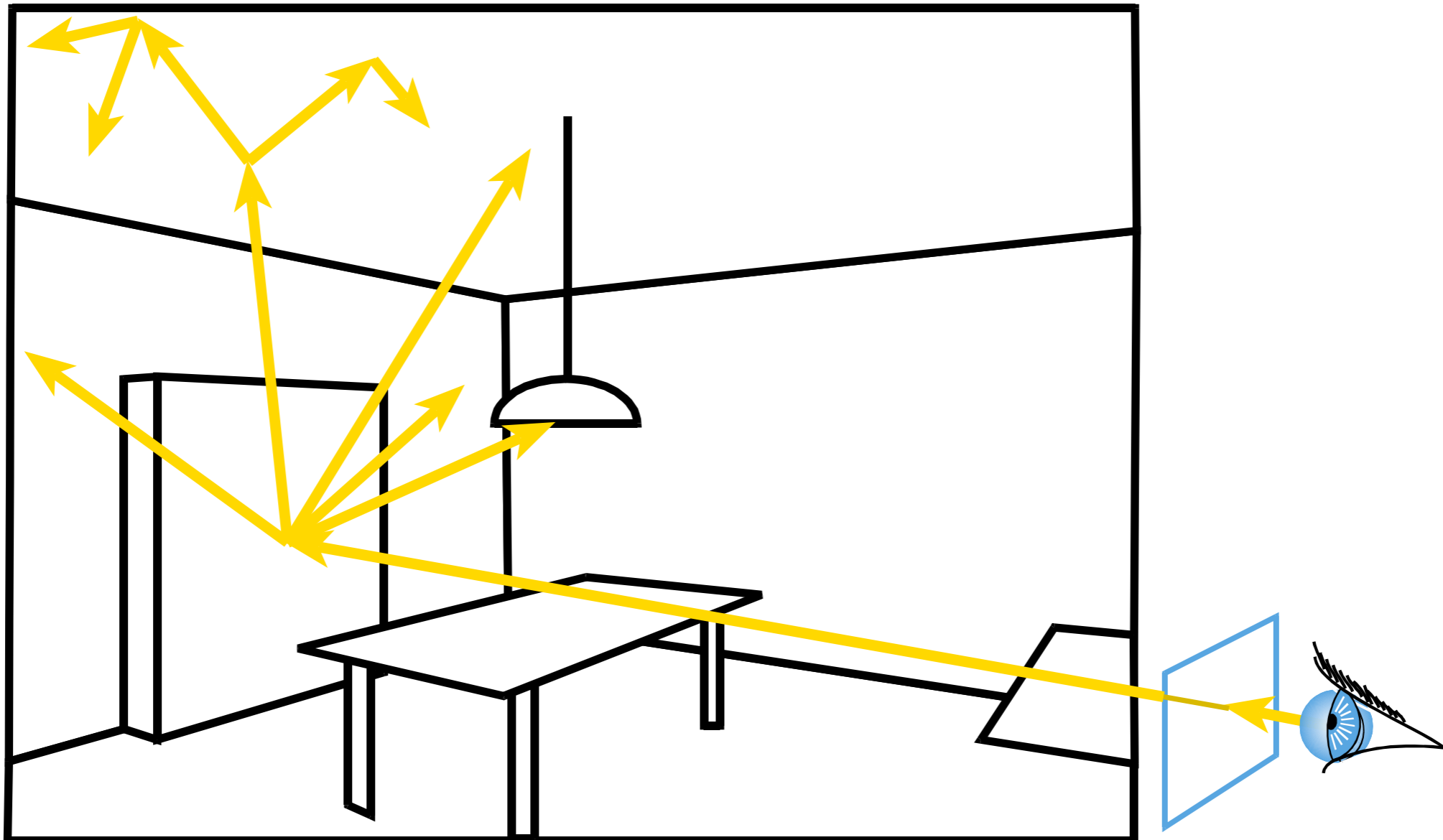
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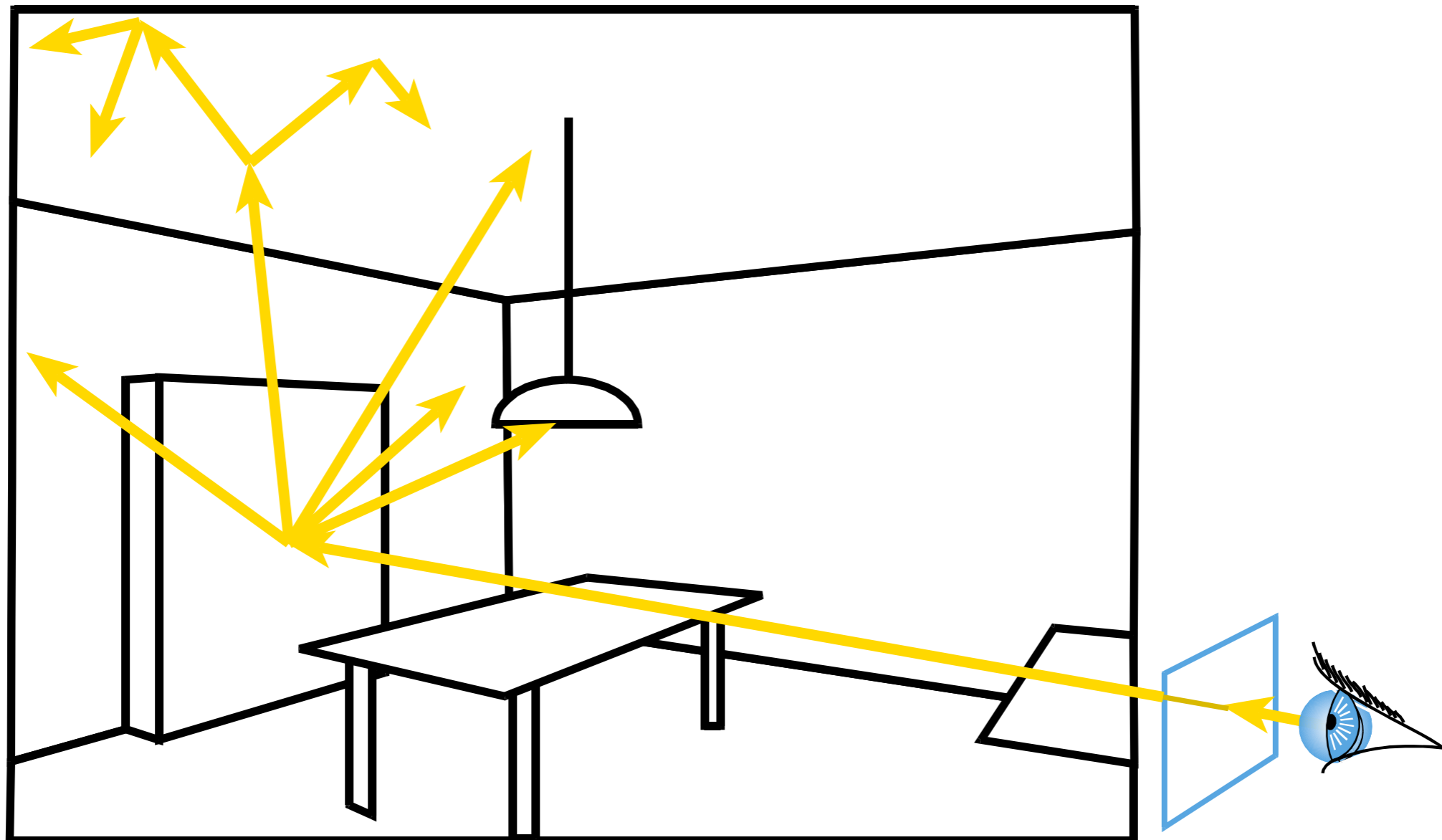
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“Monte-Carlo Ray Tracing”

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Combinatorial explosion!



Combinatorial Explosion

- Sample indirect illumination with 100 rays
- Each ray results in N more rays.. grows exponentially
- For $N=100$
 - 1 eye ray
 - 100 indirect rays at primary hit
 - 10 000 indirect rays at the secondary hits
 - 1 000 000 at the tertiary hits
 - You get the picture

Back to AO: Better Way

- Rather than 2D x 2D, one integral over 4D domain:

$$I_j = \int_{\text{screen} \times \Omega} g(x, y, \omega) dx dy d\omega$$

with integrand

$$g(x, y, \omega) = f(x - x_j, y - y_j) V(P(x, y), \omega) \cos \theta$$

Back to AO: Better Way

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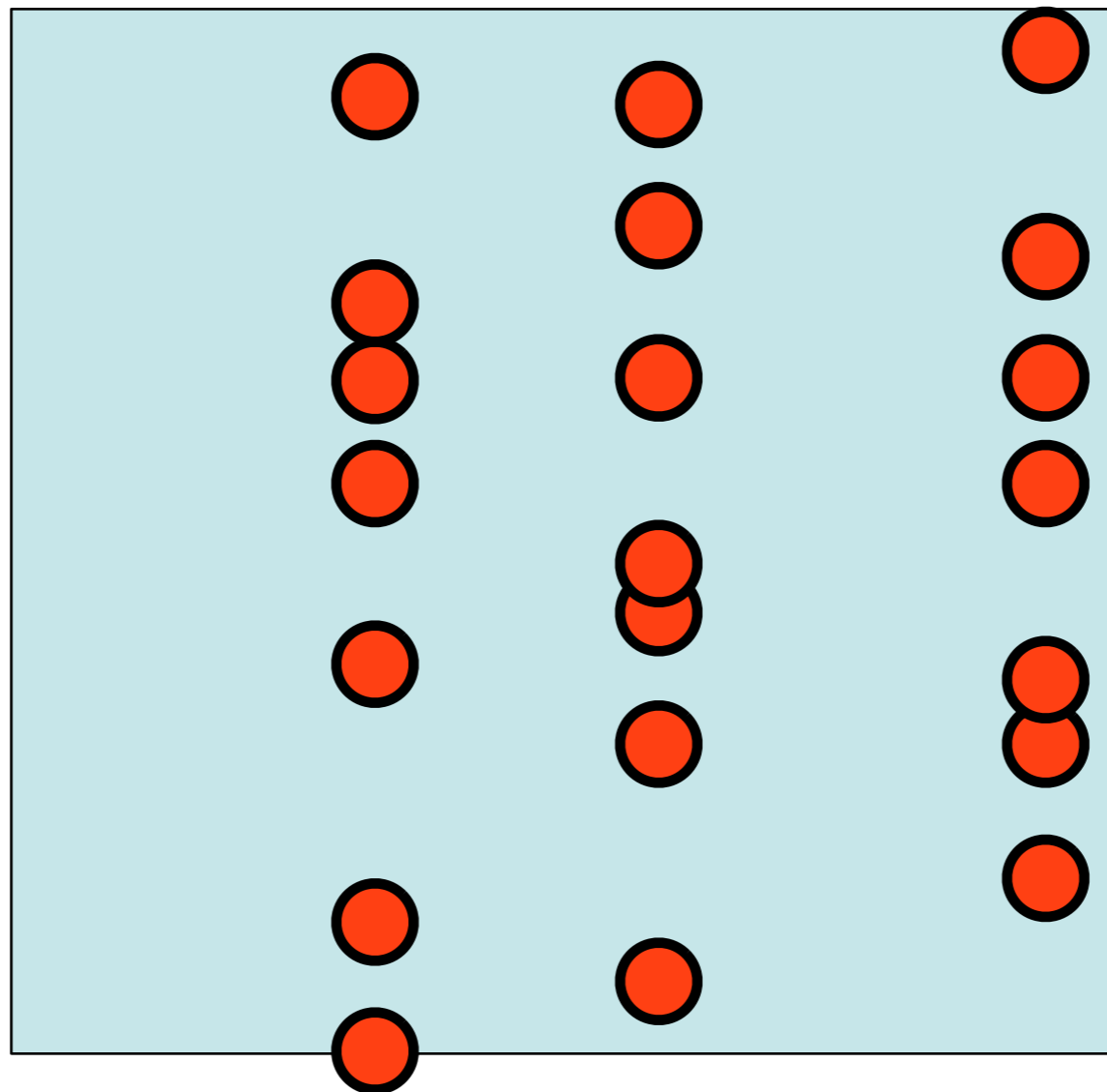
with integrand

$$g(x, y, \omega) = f(x - x_j, y - y_j) V(P(x, y), \omega) \cos \theta$$

- This is strictly equivalent; just another point of view
– *Think of 1D vs. 2D integrals*

Nested 1D + 1D, naive

$$\int \left(\int f(x, y) dy \right) dx$$

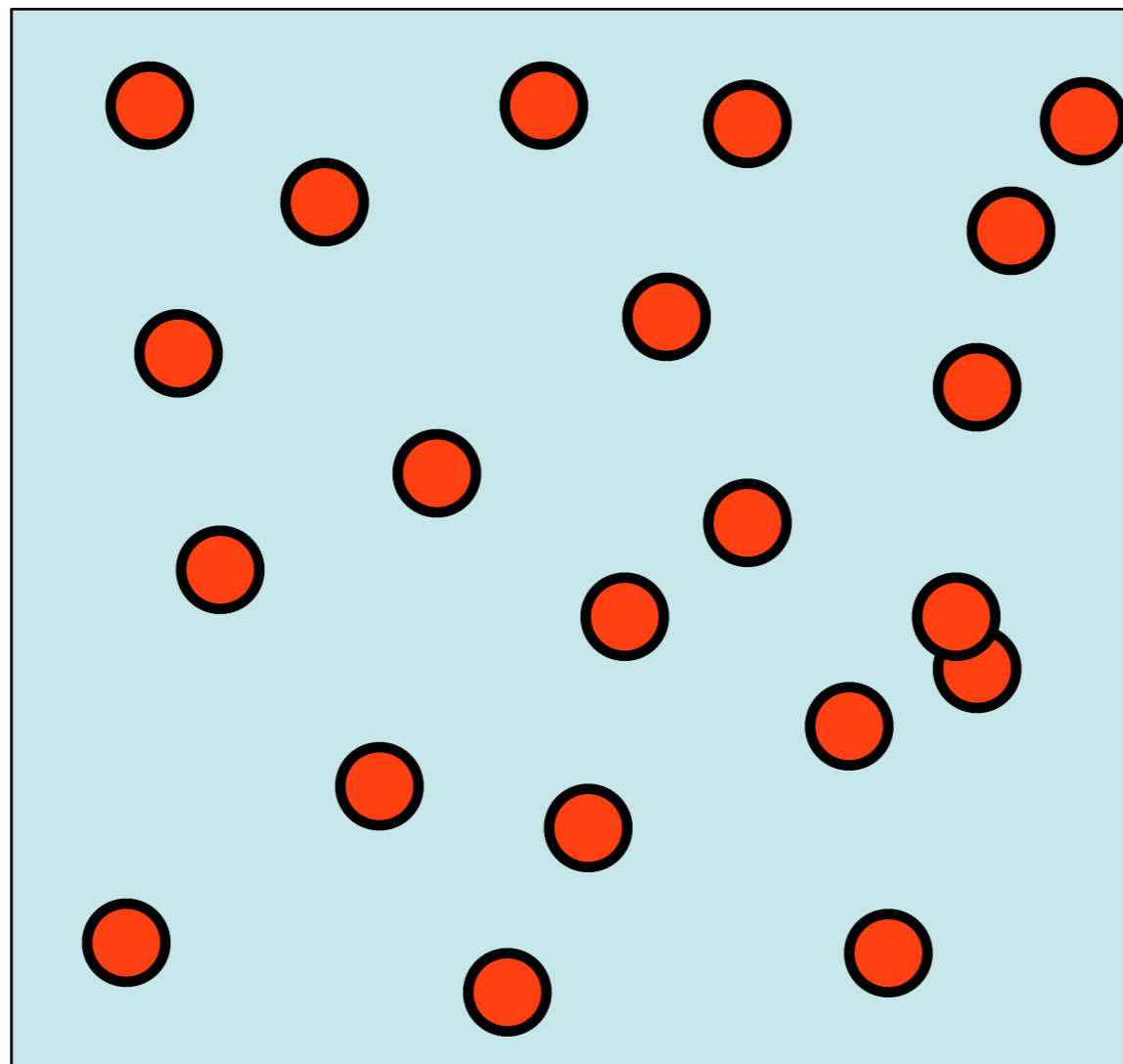


First pick x ,
then pick a
bunch of y s

Repeat

Nested 1D + 1D, treat as 2D

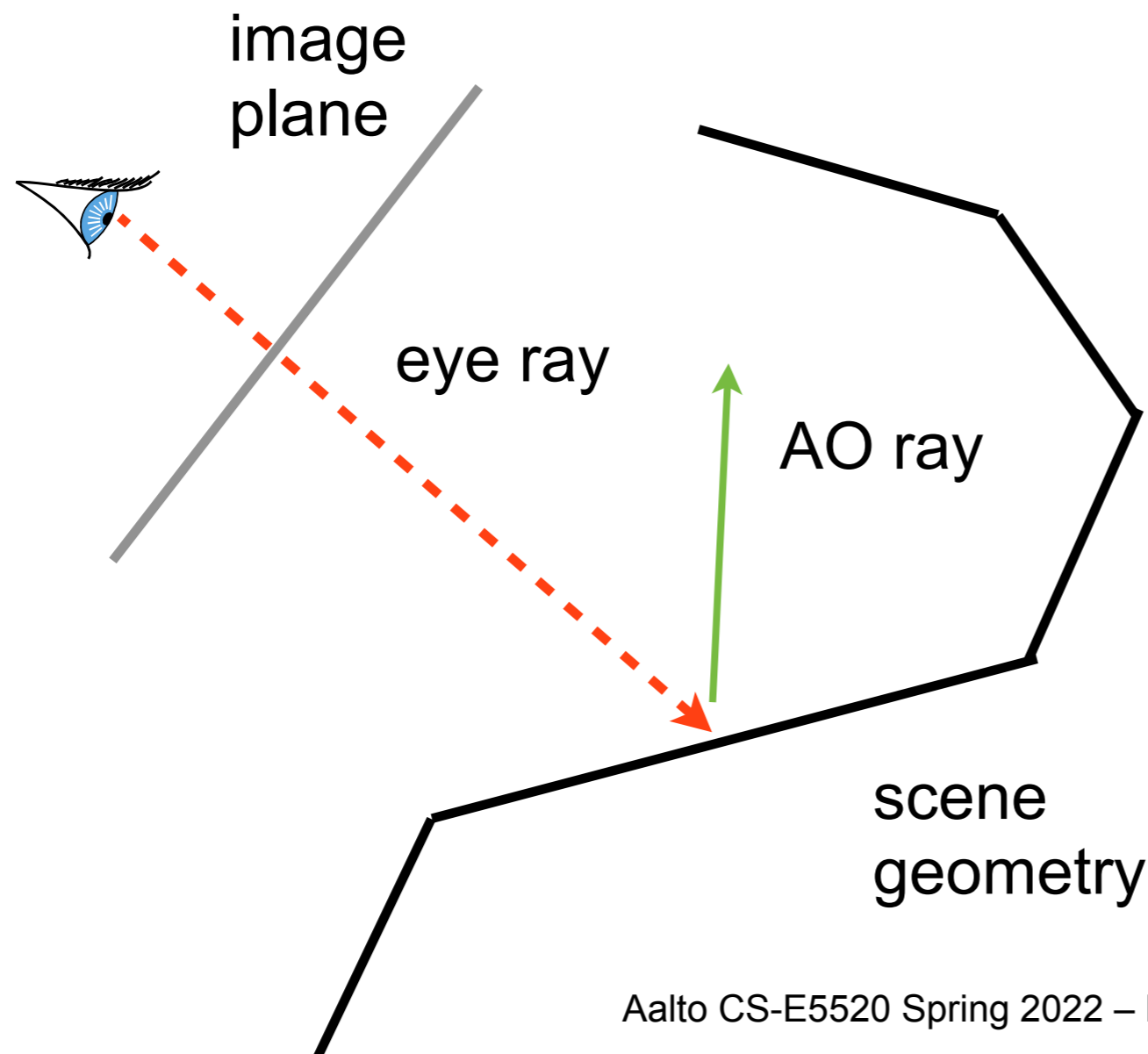
$$\int \left(\int f(x, y) dy \right) dx = \int \int f(x, y) dx dy$$



Draw 2D
samples (x,y)
from 2D pdf

Visually: One sample is Two Rays

$$I_j = \int_{\text{screen} \times \Omega} g(x, y, \omega) dx dy d\omega$$

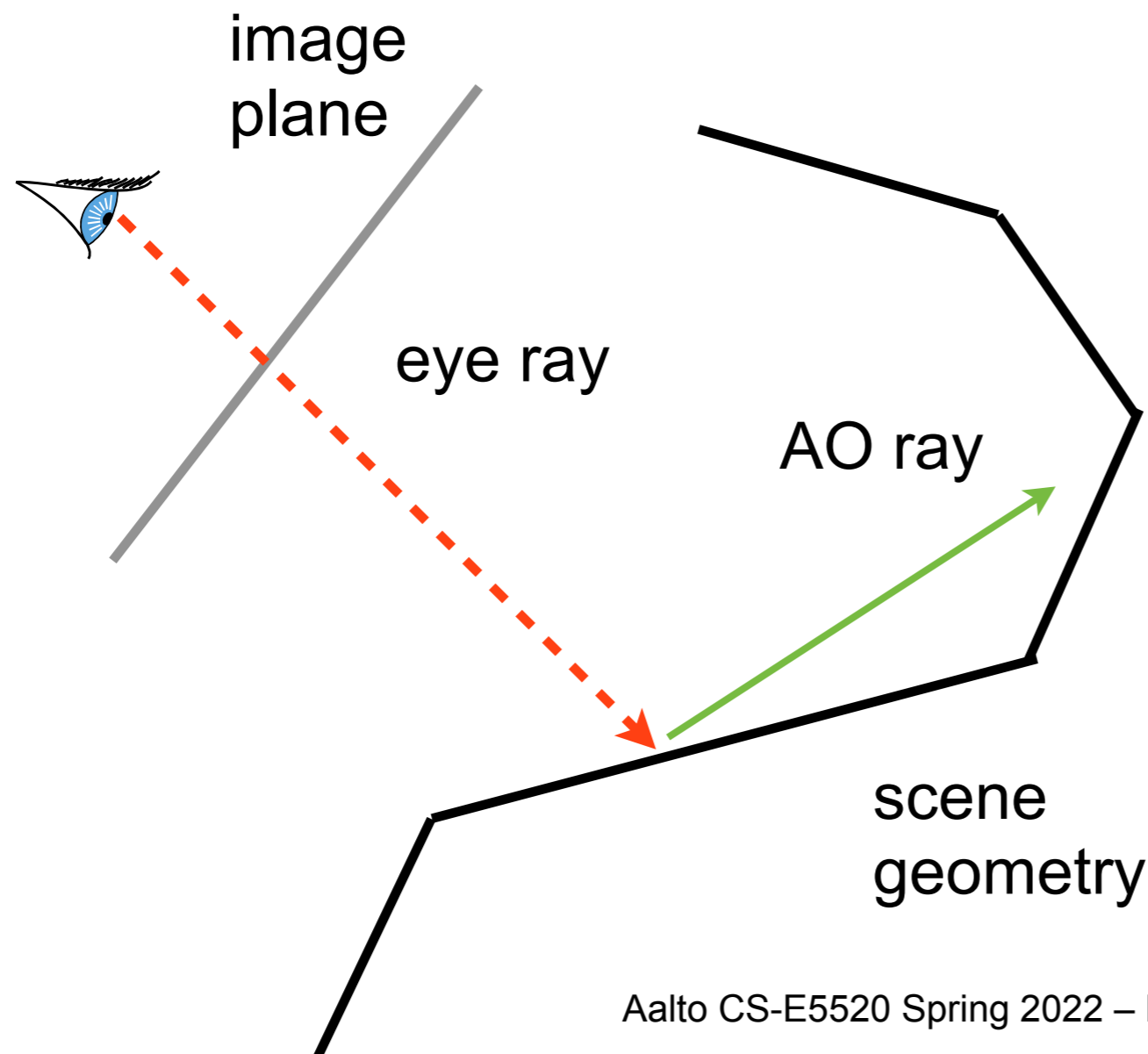


Better MC implementation:

```
res = 0
for i=1 to #samples
  pick sample (x,y,w_out)
  pdf=p(x,y)*p(w_out)
  P=castray(x,y)
  V=castray(P,w_out)
  res += g(x,y,w_out)/pdf
end
res = res/#samples
```

Visually: One sample is Two Rays

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Implementation Details

- Naturally, if your pixel filters overlap, you use the same samples for updating all the pixels with nonzero filter responses

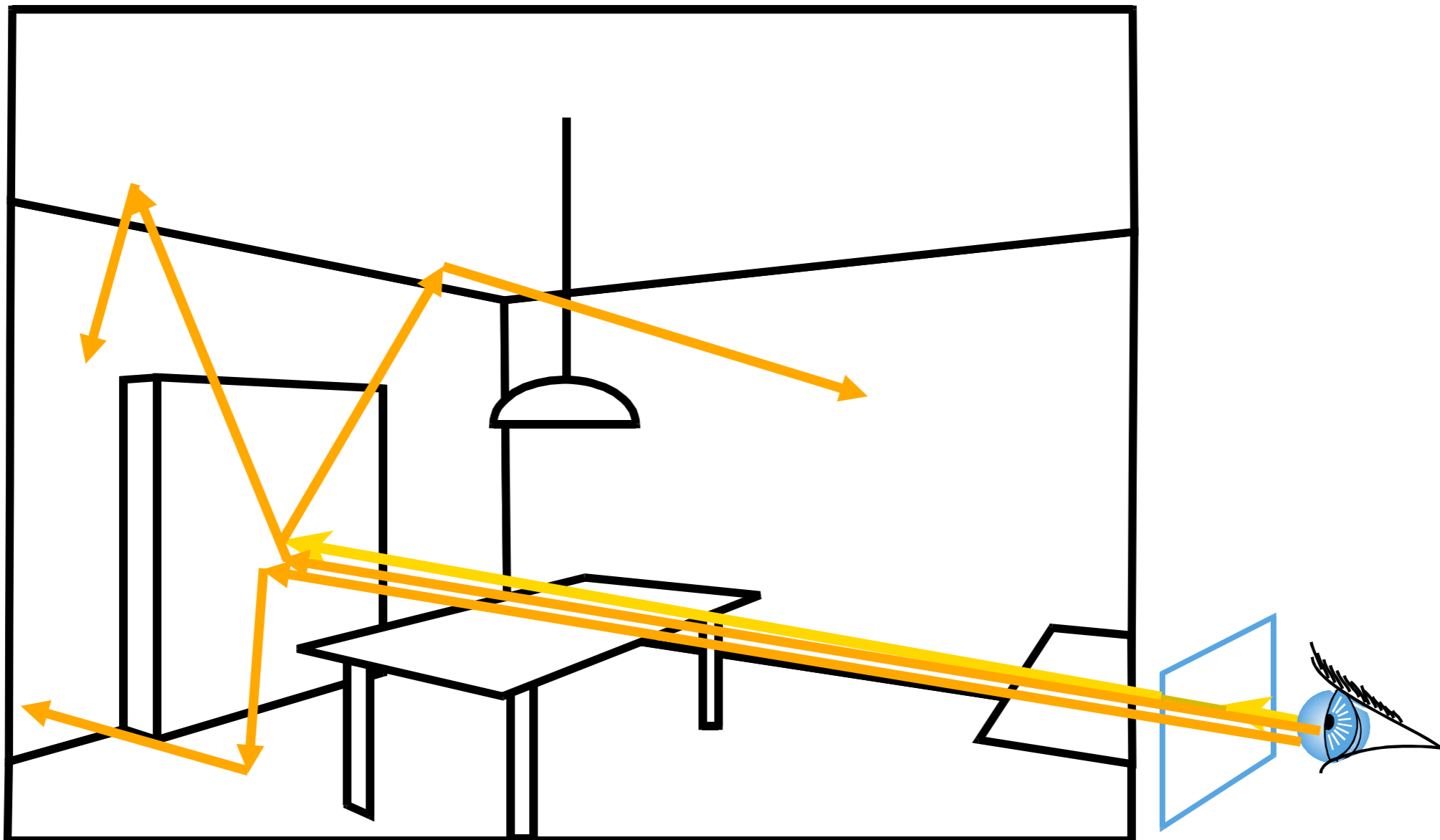
```
res[k] = weight[k] = 0 for all pixels k
for each pixel k
  for i=1 to #samplesperpixel
    pick sample (x,y,omega) // e.g. 4D Sobol'
    pdf=p(x,y)*p(omega) // usually p(x,y) == 1
    P=castray(x,y) // find primary hit
    V=castray(P,omega).length()>D // evaluate AO shadow term
    for each pixel j where f_j(x,y) is nonzero
      res[j] += f_j(x,y)*cos(theta)*V/pdf
      weight[j] += f_j(x,y)/p(x,y)
    end
  end
end
end
res[k] = res[k]/weight[k]
```

Filter of j th pixel

$$f_j(x, y) = f(x - x_j, y - y_j)$$

Monte Carlo Path Tracing

- Trace only one secondary ray per recursion
 - Otherwise number of rays explodes!
- But send many primary rays per pixel (antialiasing)



Monte Carlo Path Tracing

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