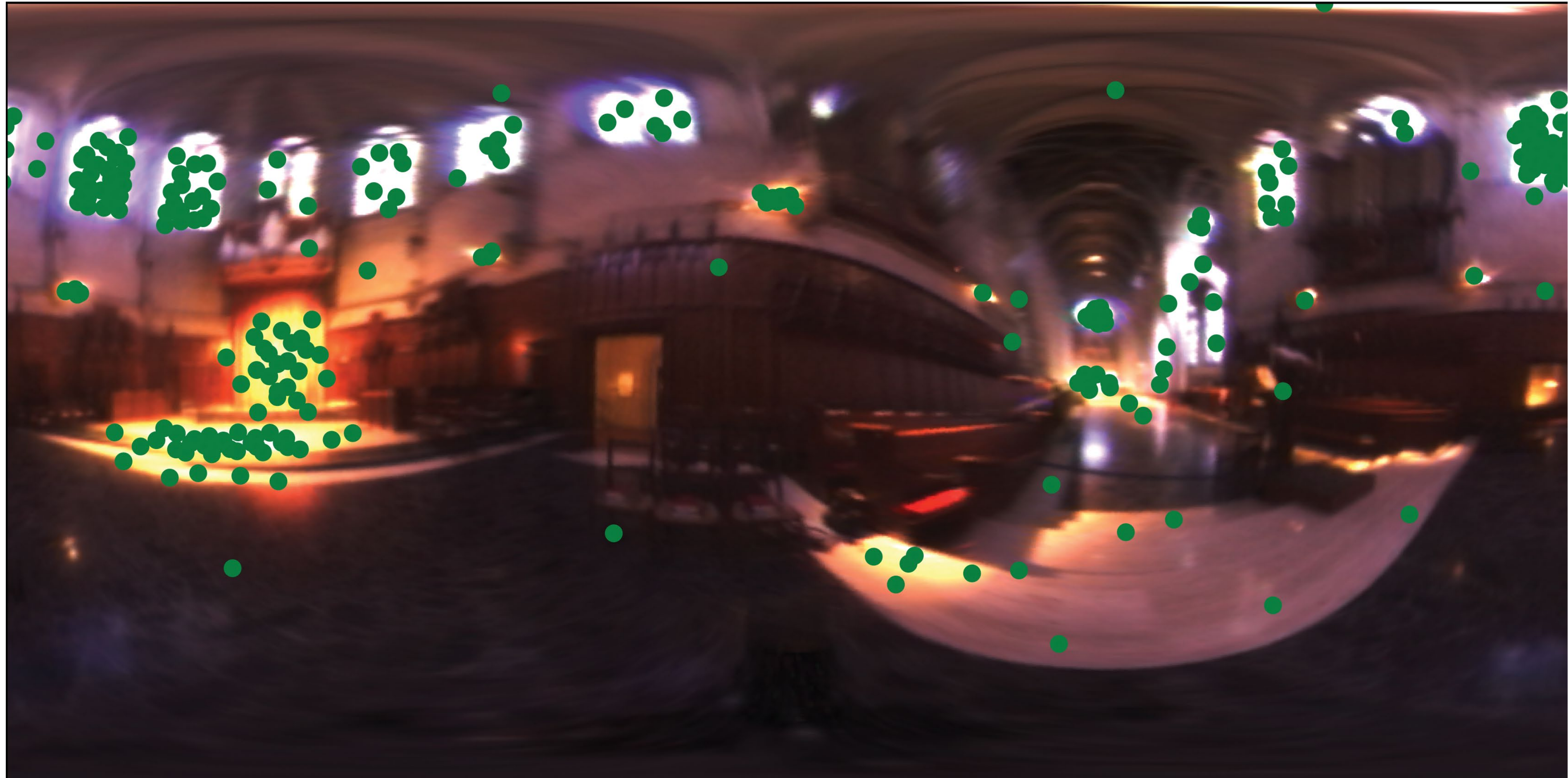


# Direct illumination



15-468, 15-668, 15-868  
Physics-based Rendering  
Spring 2024, Lecture 10

# Course announcements

- Programming assignment 2 posted, due Friday 2/23 at 23:59.
  - How many of you have looked at/started/finished it?
  - Any questions?
- Remember to vote for the rendering competition of programming assignment 1!

# Overview of today's lecture

- Importance sampling the reflectance equation.
- BRDF importance sampling.
- Direct versus indirect illumination.
- Different forms of the reflectance equation.
- Environment lighting.
- Light sources.
- Mixture sampling.
- Multiple importance sampling.

# Slide credits

Most of these slides were directly adapted from:

- Wojciech Jarosz (Dartmouth).

# Reflection equation

---

$$L_r(\mathbf{x}, \vec{\omega}_r) = \int_{H^2} f_r(\mathbf{x}, \vec{\omega}_i, \vec{\omega}_r) L_i(\mathbf{x}, \vec{\omega}_i) \cos \theta_i d\vec{\omega}_i$$

What terms can we importance sample?

- BRDF
- incident radiance
- cosine term

# Reflection equation

---

$$L_r(\mathbf{x}, \vec{\omega}_r) = \int_{H^2} f_r(\mathbf{x}, \vec{\omega}_i, \vec{\omega}_r) L_i(\mathbf{x}, \vec{\omega}_i) \cos \theta_i d\vec{\omega}_i$$

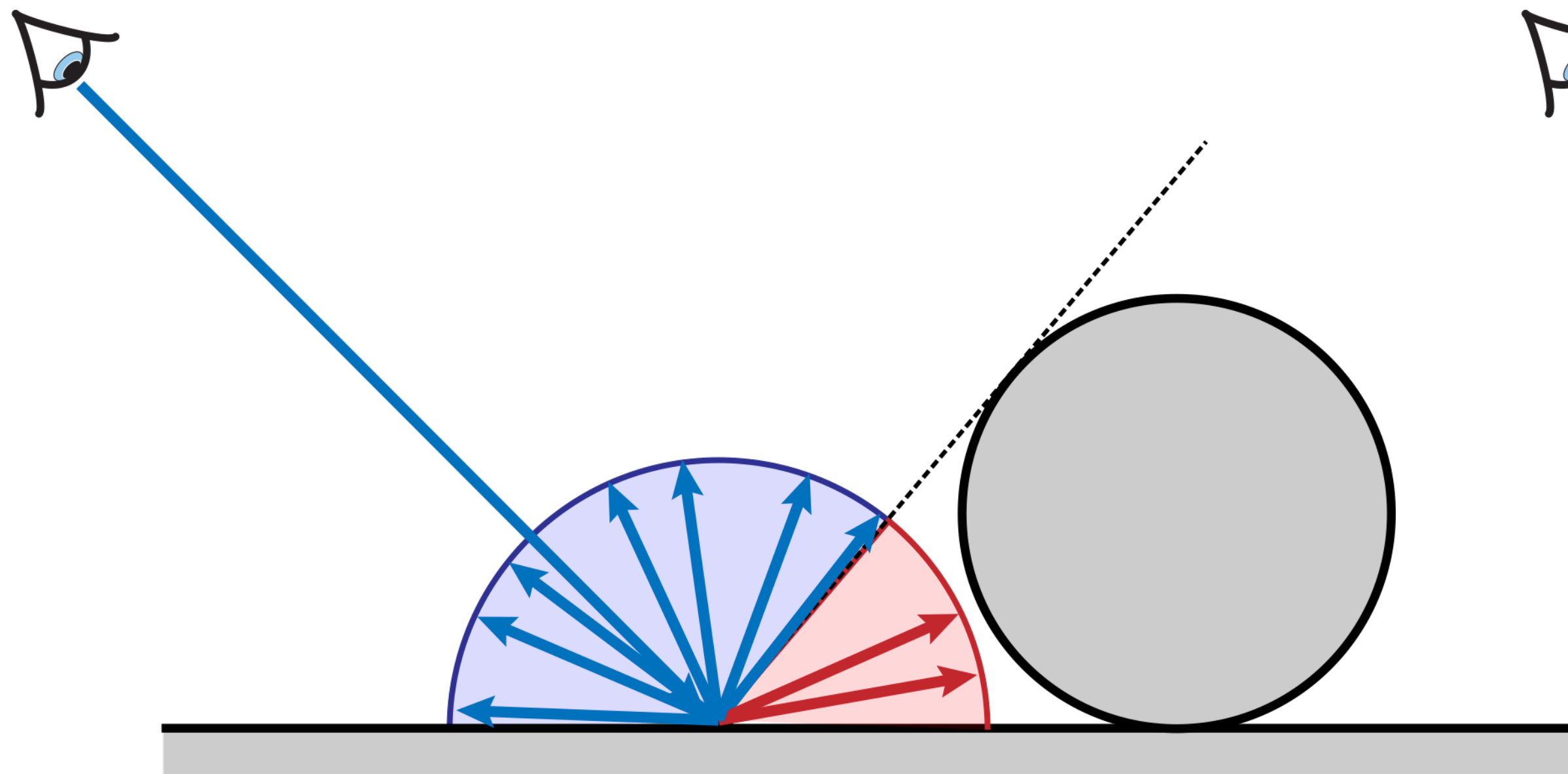
What terms can we importance sample?

- BRDF
- incident radiance
- cosine term

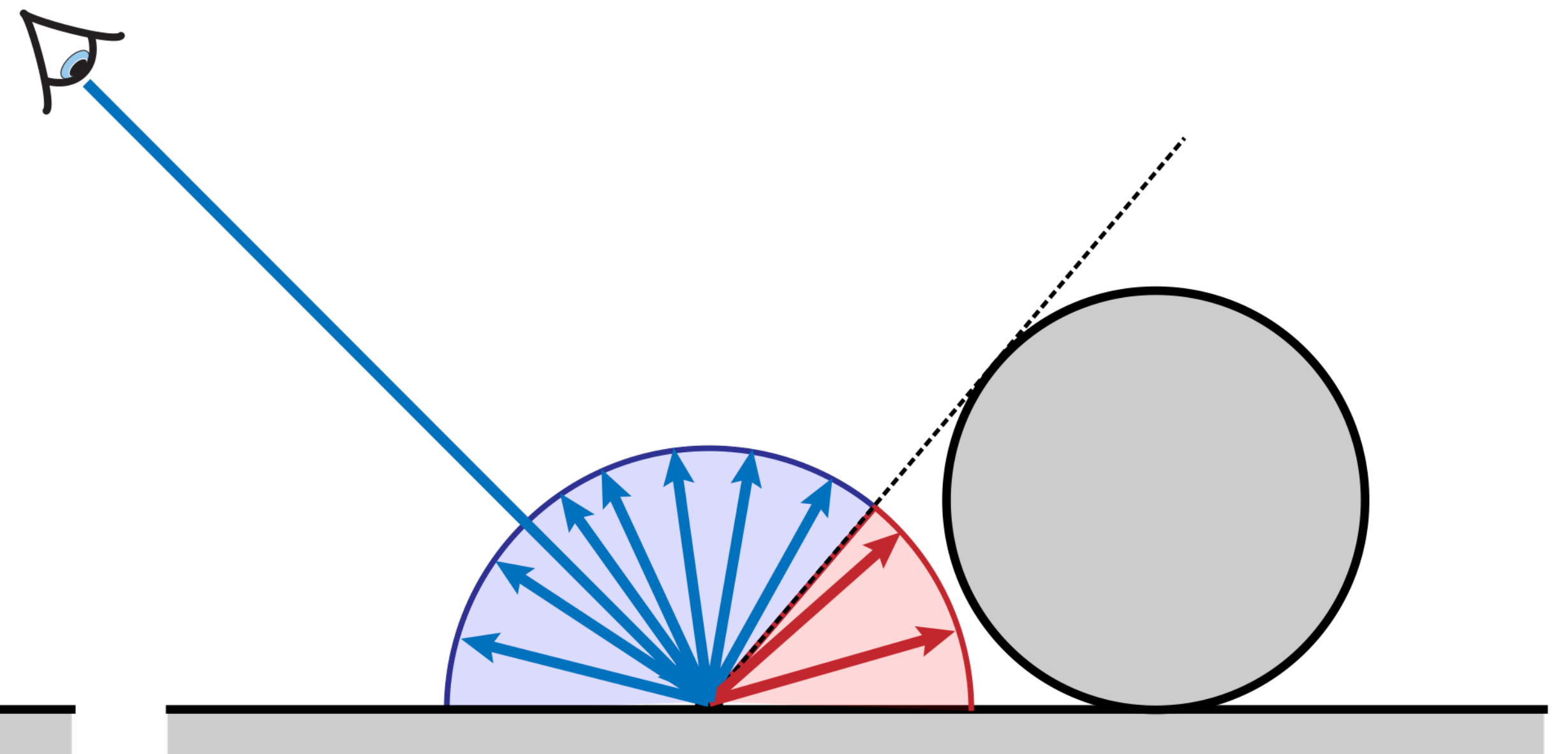
# This is what we did for ambient occlusion

---

Uniform hemispherical sampling



Cosine-weighted importance sampling



# Reflection equation

---

$$L_r(\mathbf{x}, \vec{\omega}_r) = \int_{H^2} f_r(\mathbf{x}, \vec{\omega}_i, \vec{\omega}_r) L_i(\mathbf{x}, \vec{\omega}_i) \cos \theta_i d\vec{\omega}_i$$

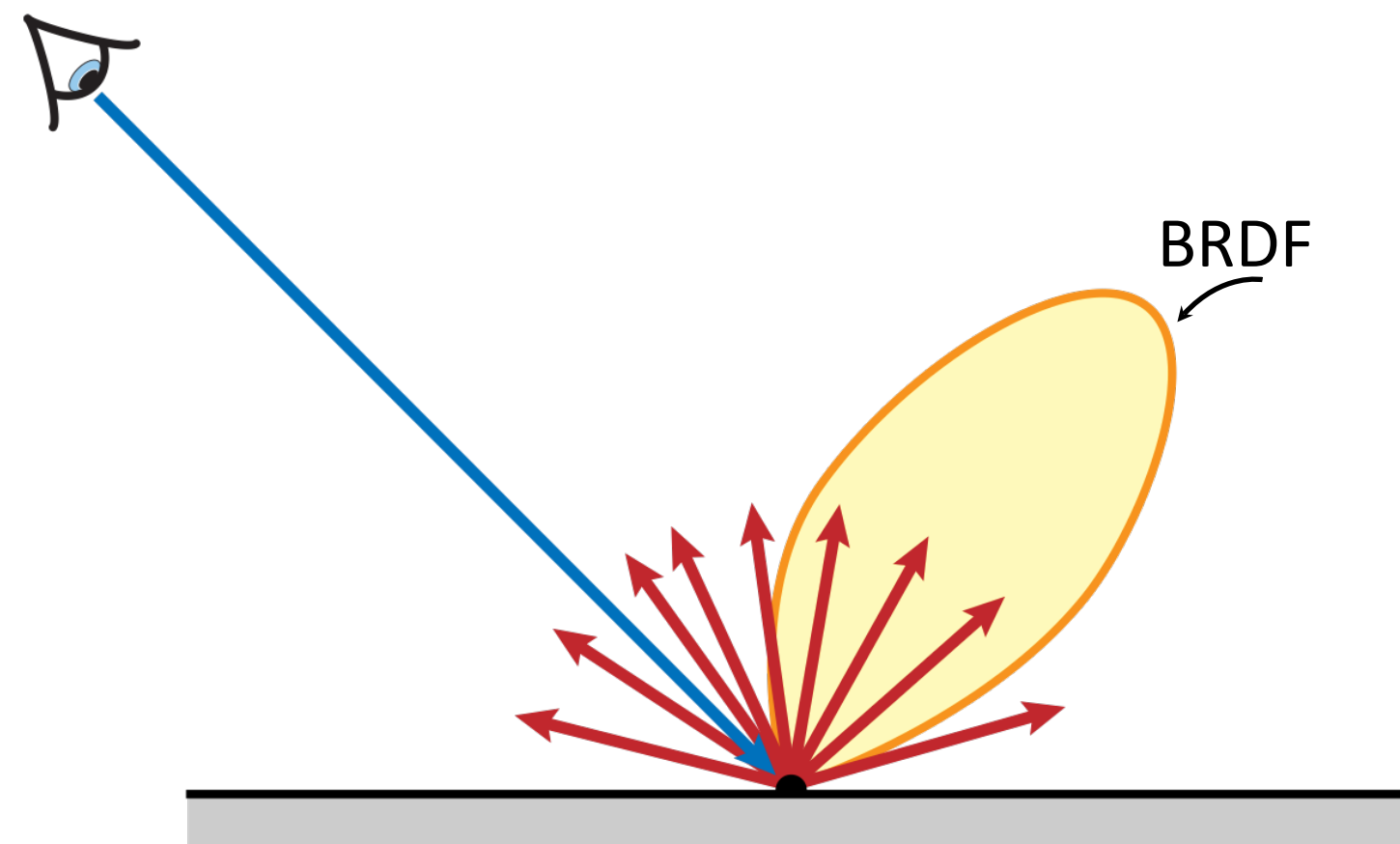
What terms can we importance sample?

- BRDF
- incident radiance
- cosine term

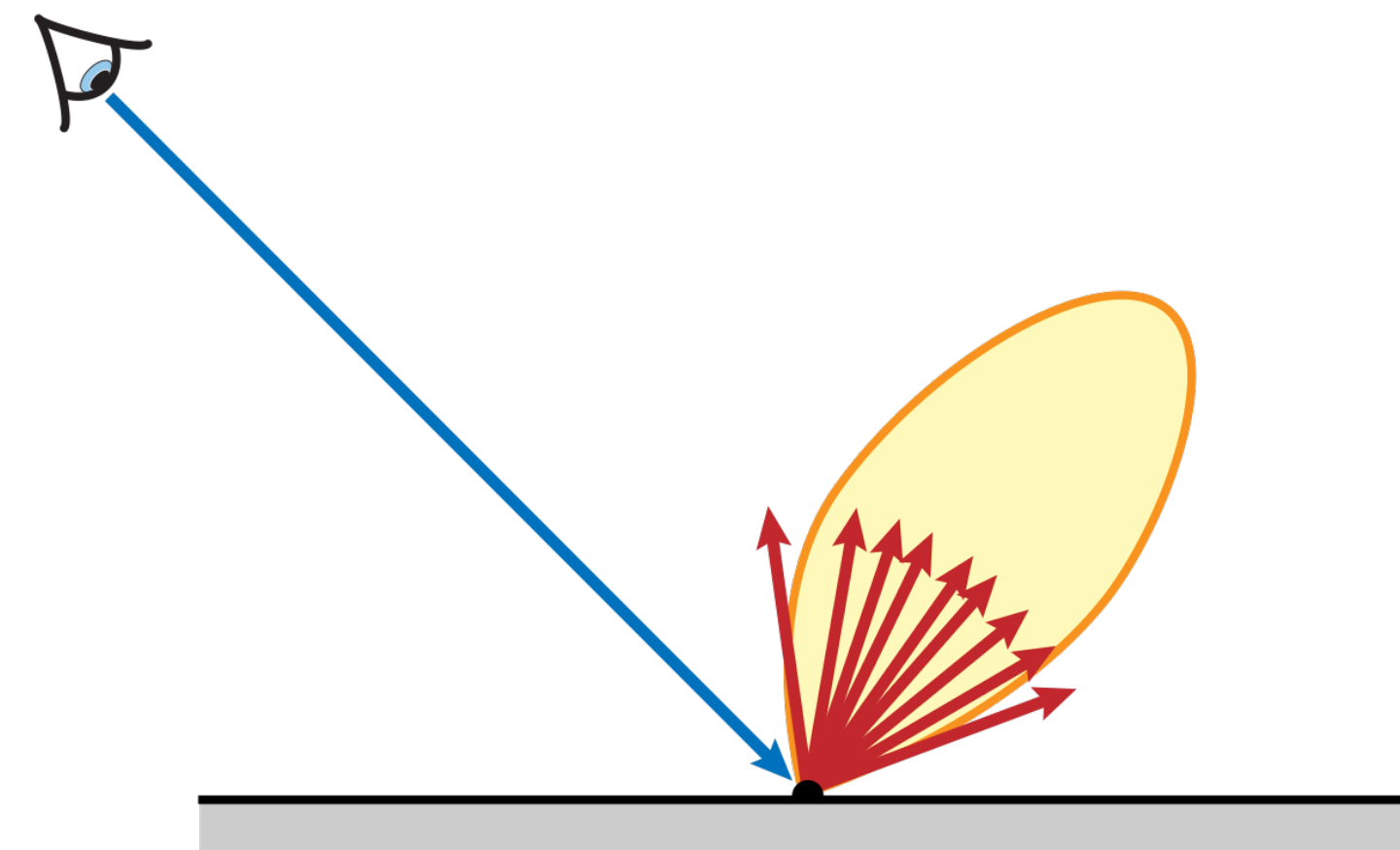


# Importance Sampling the BRDF

Cosine-weighted  
importance sampling

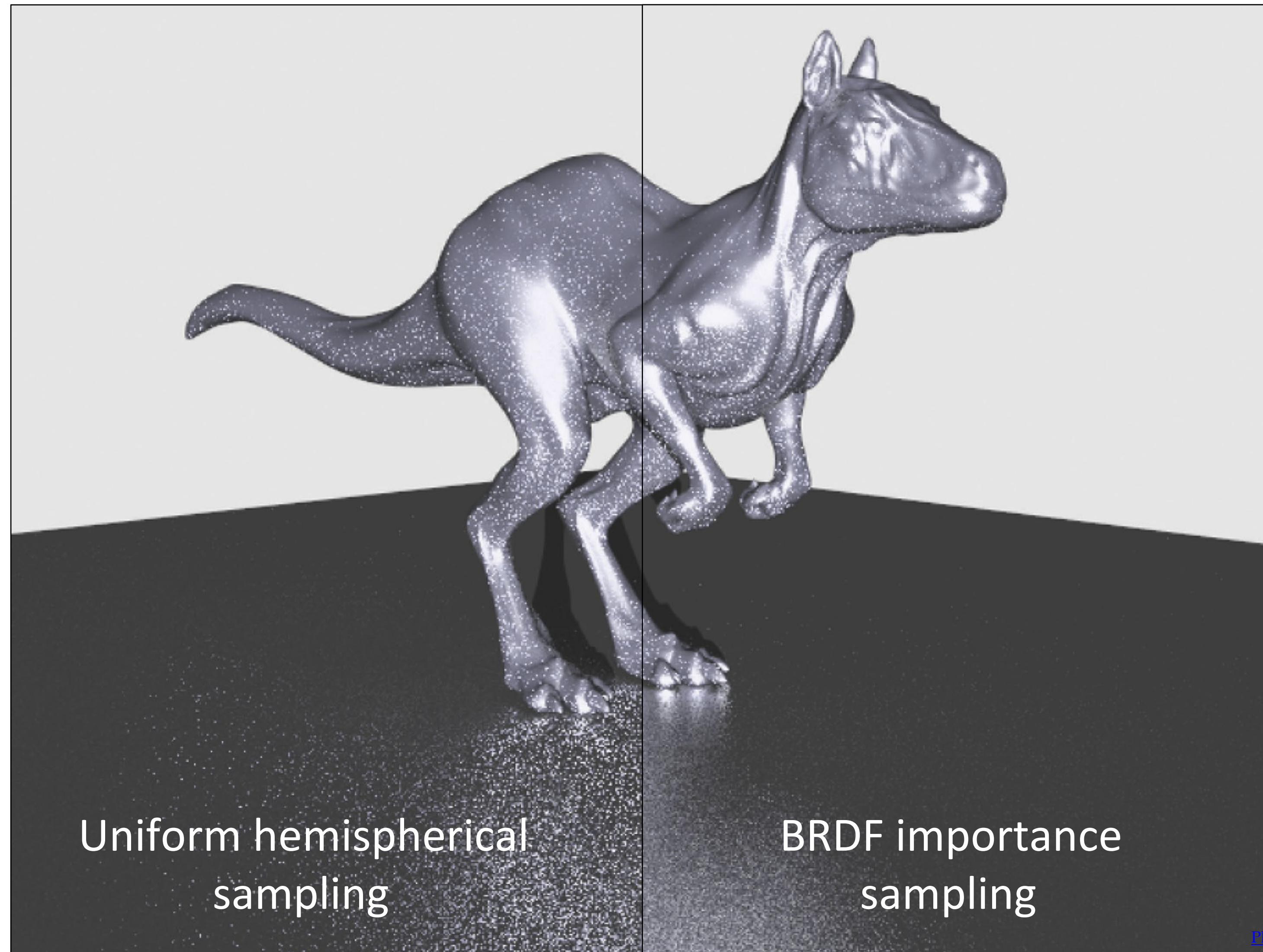


BRDF importance  
sampling



$$p(\vec{\omega}_i) \propto f(\mathbf{x}, \vec{\omega}_i, \vec{\omega}_r)$$

# Importance Sampling the BRDF

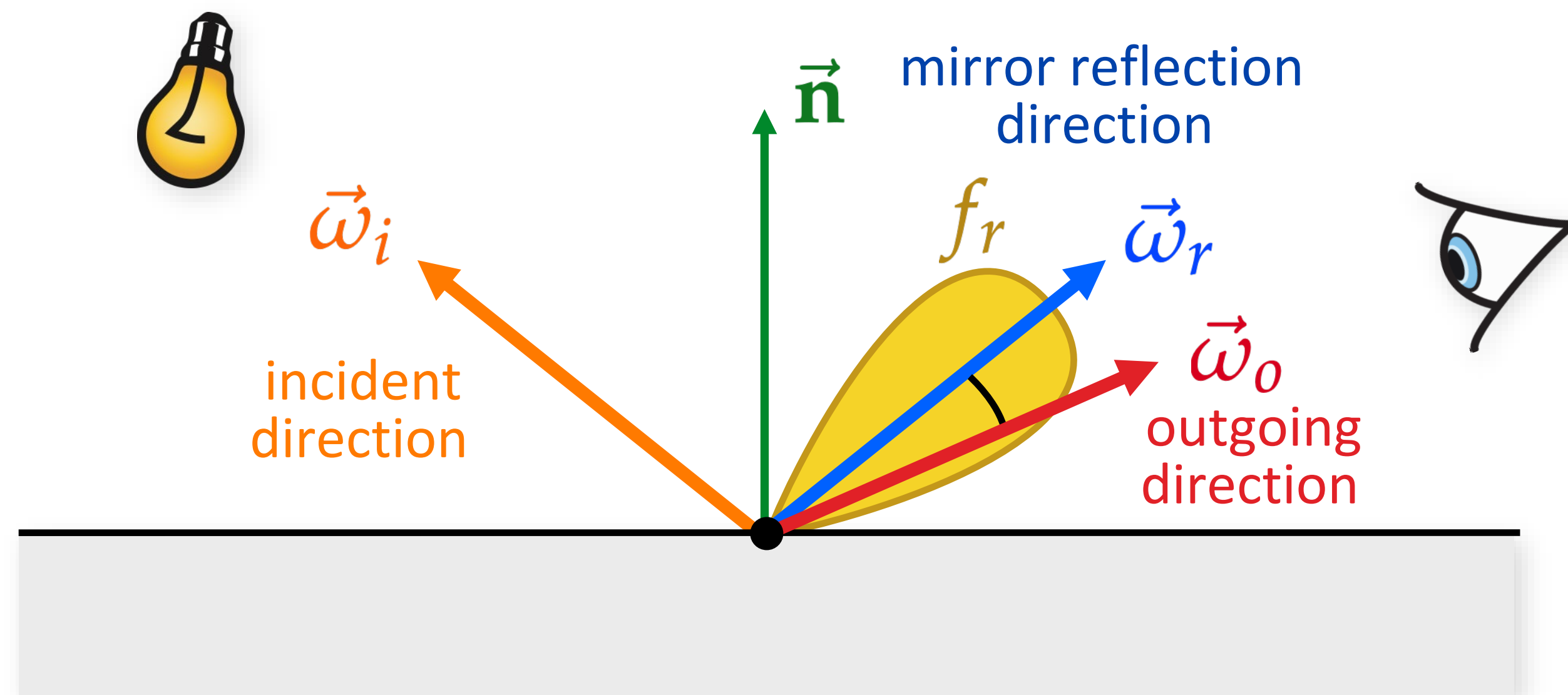


# Phong BRDF

Normalized exponentiated cosine lobe:

$$f_r(\vec{\omega}_o, \vec{\omega}_i) = \frac{e + 2}{2\pi} (\vec{\omega}_r \cdot \vec{\omega}_o)^e$$

$$\vec{\omega}_r = (2\vec{n}(\vec{n} \cdot \vec{\omega}_i) - \vec{\omega}_i)$$



# Phong BRDF

---

Normalized exponentiated cosine lobe:

$$f_r(\vec{\omega}_o, \vec{\omega}_i) = \frac{e + 2}{2\pi} (\vec{\omega}_r \cdot \vec{\omega}_o)^e$$

$$\vec{\omega}_r = (2\vec{n}(\vec{n} \cdot \vec{\omega}_i) - \vec{\omega}_i)$$

## Interpretation

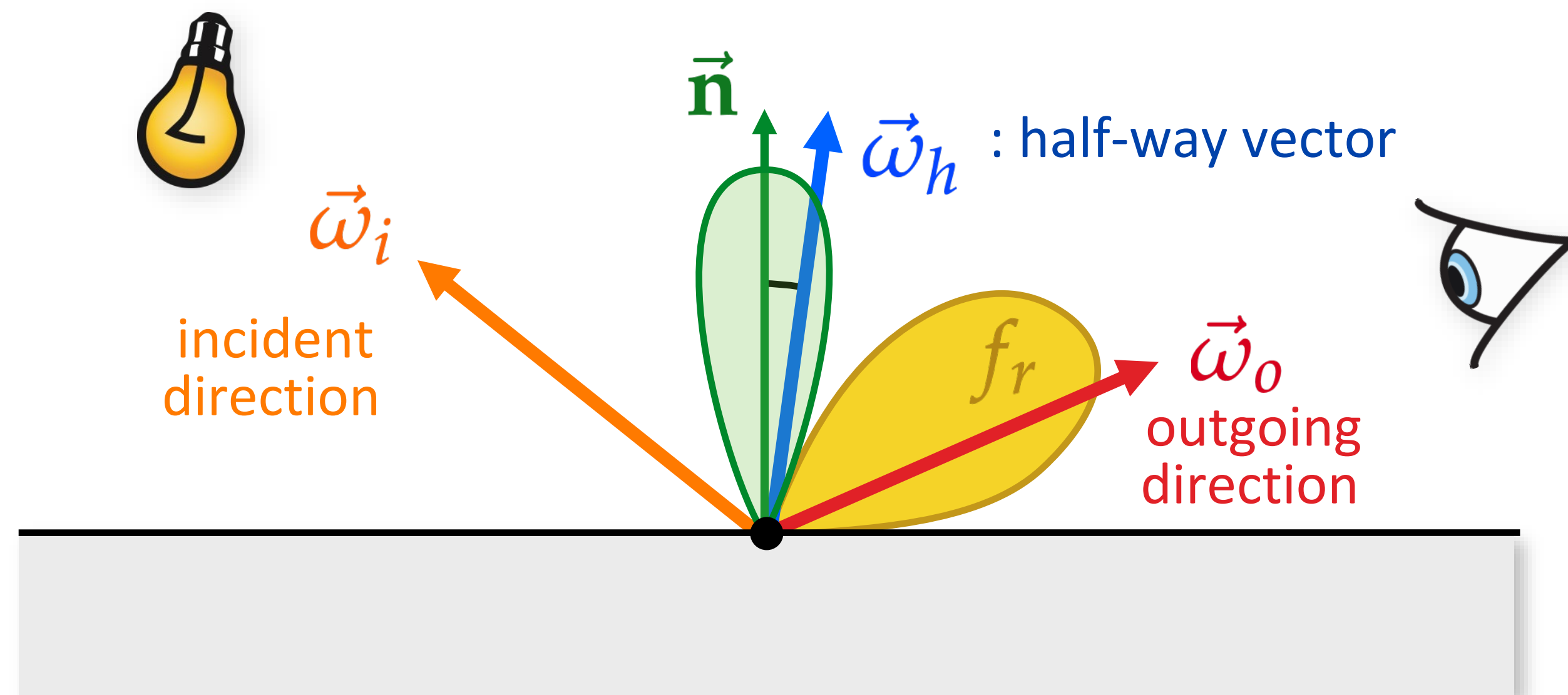
- randomize reflection rays in a lobe about mirror direction
- perfect mirror reflection of a blurred light

# Blinn-Phong BRDF

Randomize normals instead of reflection directions

$$f_r(\vec{\omega}_o, \vec{\omega}_i) = \frac{e + 2}{2\pi} (\vec{\omega}_h \cdot \vec{n})^e$$

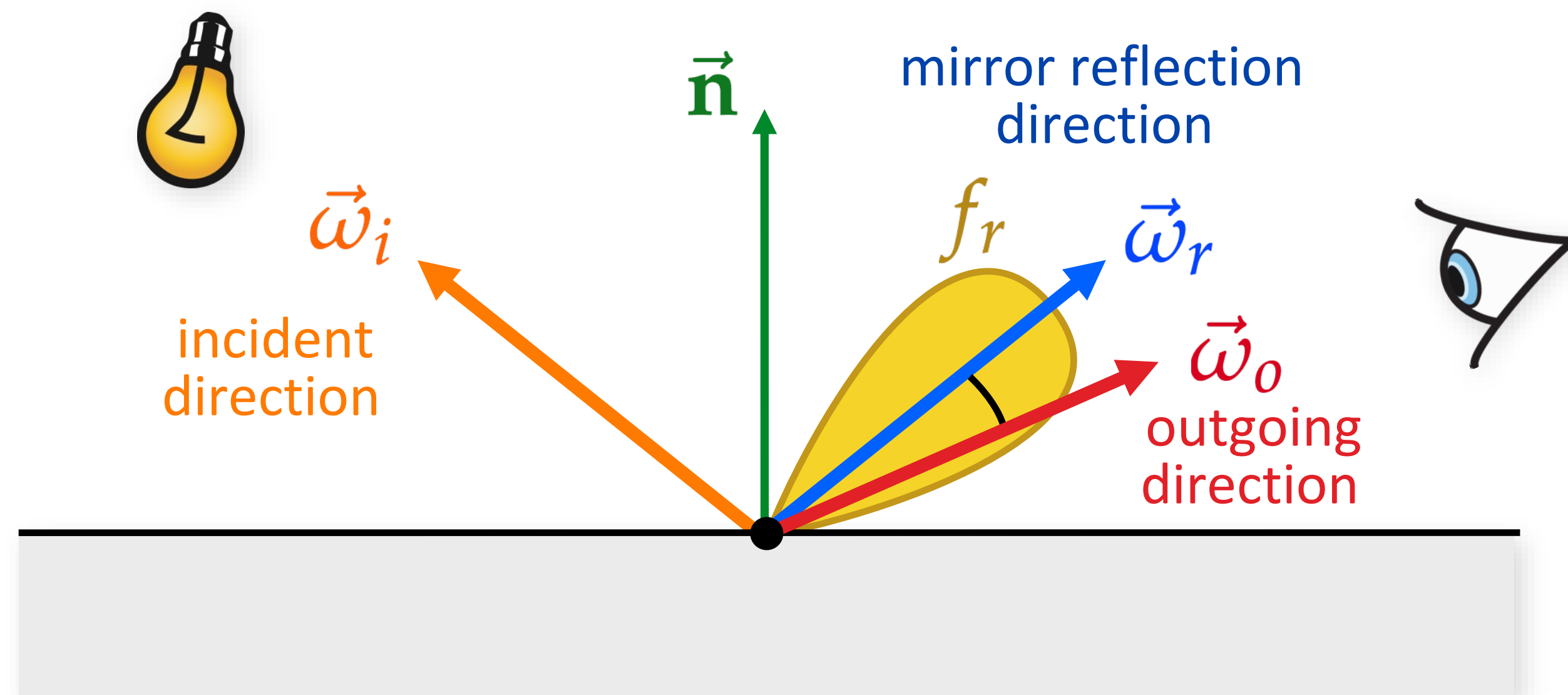
$$\vec{\omega}_h = \frac{\vec{\omega}_i + \vec{\omega}_o}{\|\vec{\omega}_i + \vec{\omega}_o\|}$$



# Phong BRDF

$$f_r(\vec{\omega}_o, \vec{\omega}_i) = \frac{e + 2}{2\pi} (\vec{\omega}_r \cdot \vec{\omega}_o)^e$$

$$\vec{\omega}_r = (2\vec{n}(\vec{n} \cdot \vec{\omega}_i) - \vec{\omega}_i)$$



# Importance Sampling the BRDF

---

## Recipe:

1. Express the desired distribution in a convenient coordinate system
  - requires computing the Jacobian
2. Compute marginal and conditional 1D PDFs
3. Sample 1D PDFs using the inversion method

# Sampling the Blinn-Phong BRDF

---

$$f_r(\vec{\omega}_o, \vec{\omega}_i) = \frac{e + 2}{2\pi} (\vec{\omega}_h \cdot \vec{\mathbf{n}})^e$$

Mirror reflection from random micro-normal

General recipe:

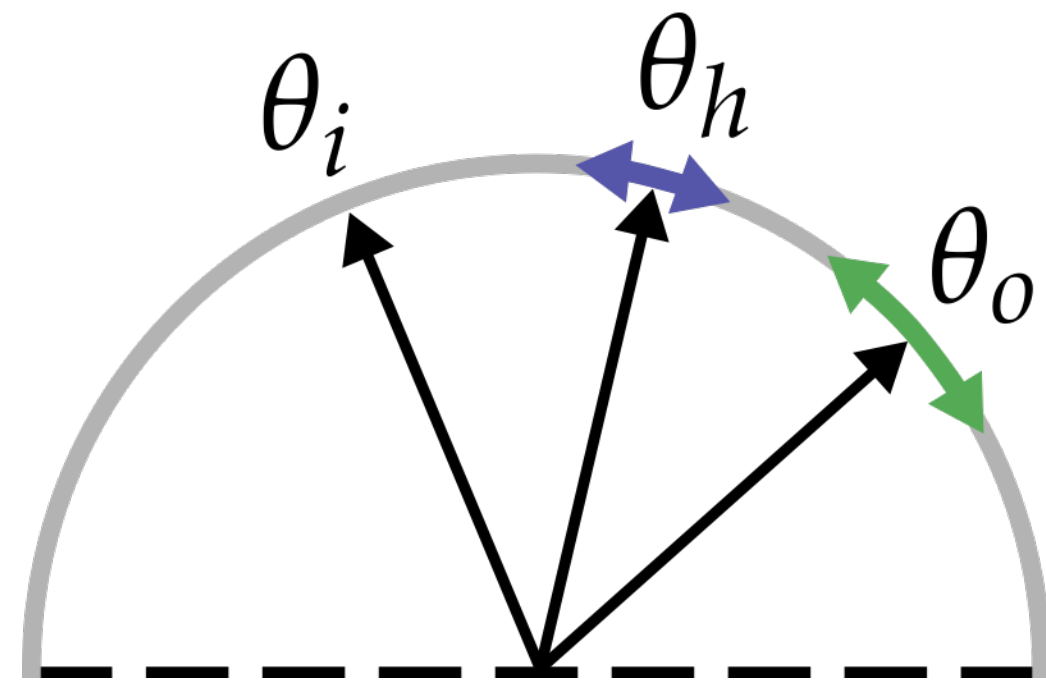
- randomly generate a  $\omega_h$ , with PDF proportional to  $\cos^e$
- reflect incident direction  $\omega_i$  about  $\omega_h$  to obtain  $\omega_o$
- convert  $\text{PDF}(\omega_h)$  to  $\text{PDF}(\omega_o)$  (change-of-variable)

Read PBRTv3 14.1



# Half-direction transform

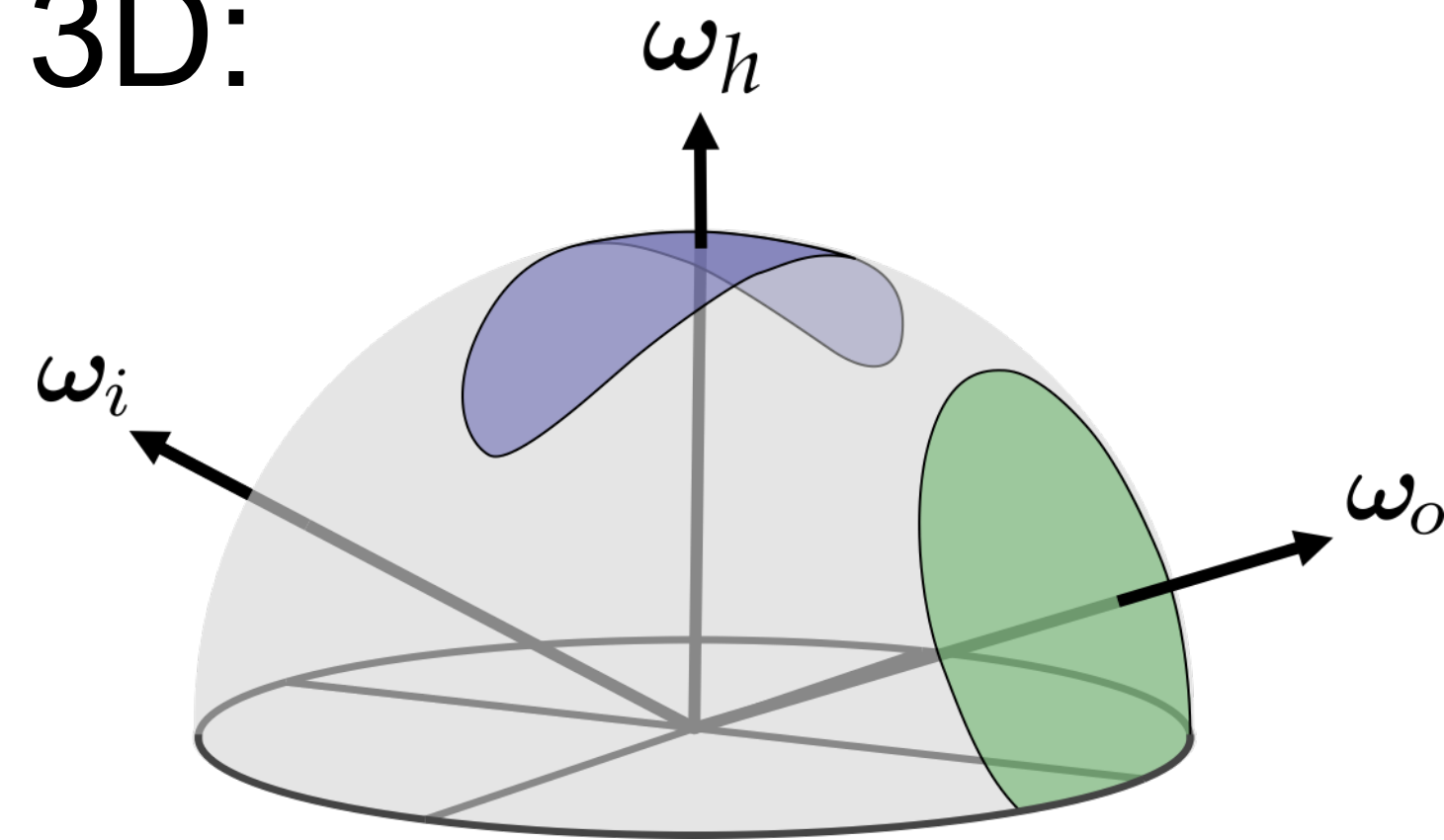
2D:



$$\theta_h := \frac{\theta_i + \theta_o}{2}$$

$$\frac{d\theta_h}{d\theta_o} = ?$$

3D:



$$\omega_h := \frac{\omega_i + \omega_o}{\|\omega_i + \omega_o\|}$$

$$\frac{d\omega_h}{d\omega_o} =$$

# Reflection equation

---

$$L_r(\mathbf{x}, \vec{\omega}_r) = \int_{H^2} f_r(\mathbf{x}, \vec{\omega}_i, \vec{\omega}_r) L_i(\mathbf{x}, \vec{\omega}_i) \cos \theta_i d\vec{\omega}_i$$

What terms can we importance sample?

- BRDF
- incident radiance
- cosine term

# Direct vs. Indirect illumination

# Direct vs. Indirect Illumination

---

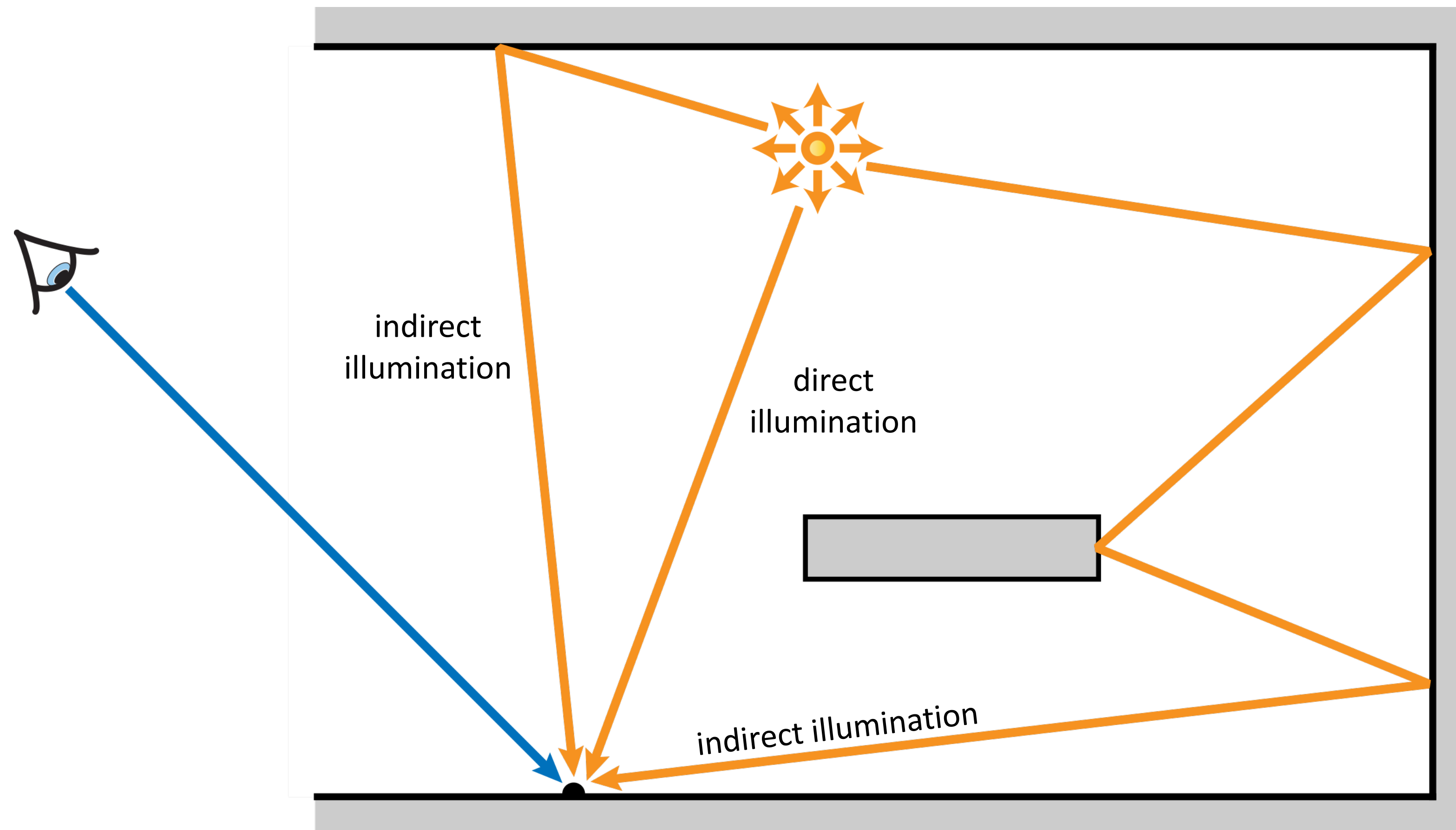
Where does  $L_i$   
“come from”?

$$L_r(\mathbf{x}, \vec{\omega}_r) = \int_{H^2} f_r(\mathbf{x}, \vec{\omega}_i, \vec{\omega}_r) L_i(\mathbf{x}, \vec{\omega}_i) \cos \theta_i \, d\vec{\omega}_i$$

# Direct vs. Indirect Illumination

Where does  $L_i$   
“come from”?

$$L_r(\mathbf{x}, \vec{\omega}_r) = \int_{H^2} f_r(\mathbf{x}, \vec{\omega}_i, \vec{\omega}_r) L_i(\mathbf{x}, \vec{\omega}_i) \cos \theta_i d\vec{\omega}_i$$



# Direct vs. Indirect Illumination

---

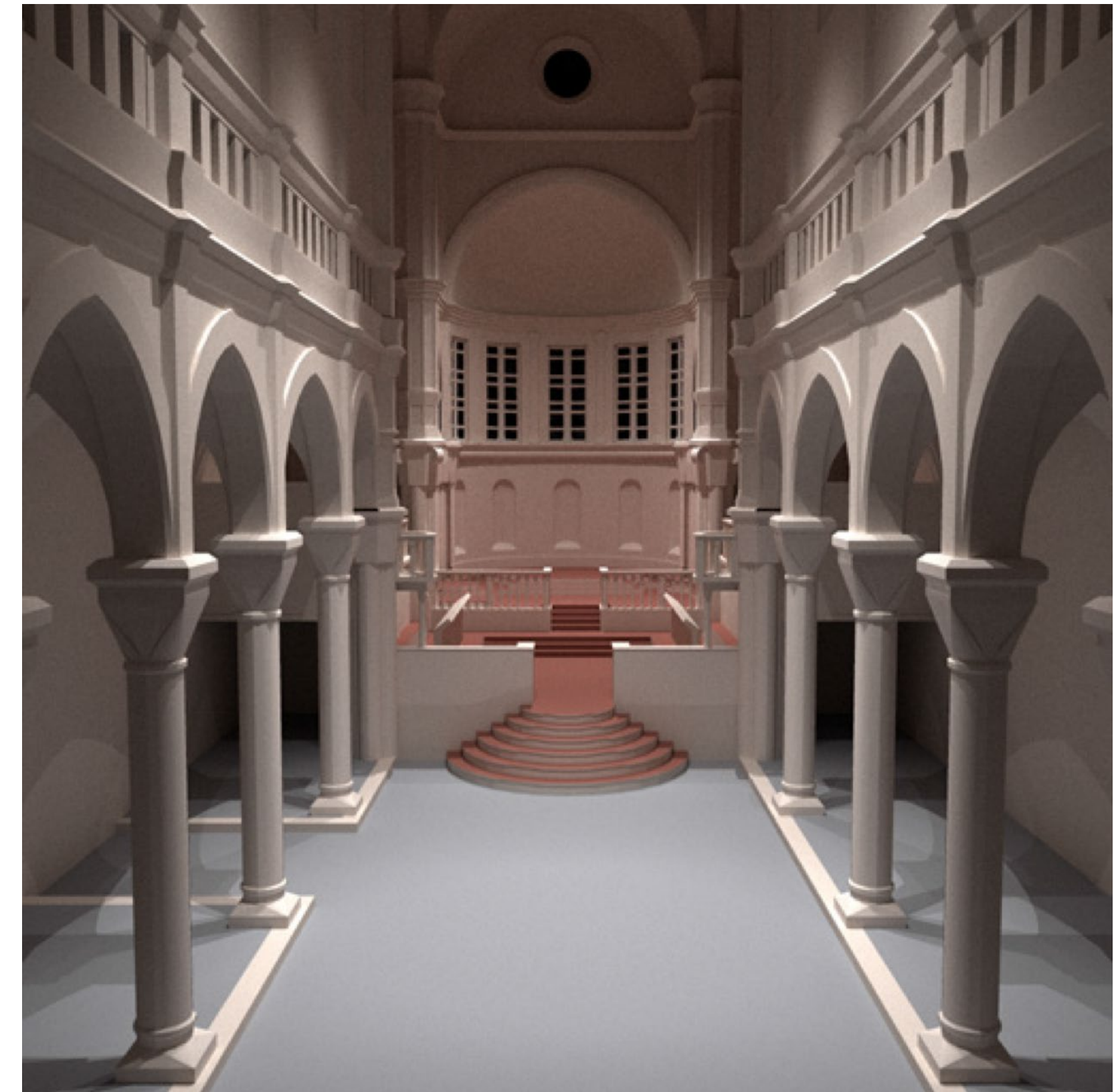
Direct illumination



Indirect illumination



Direct + indirect illumination



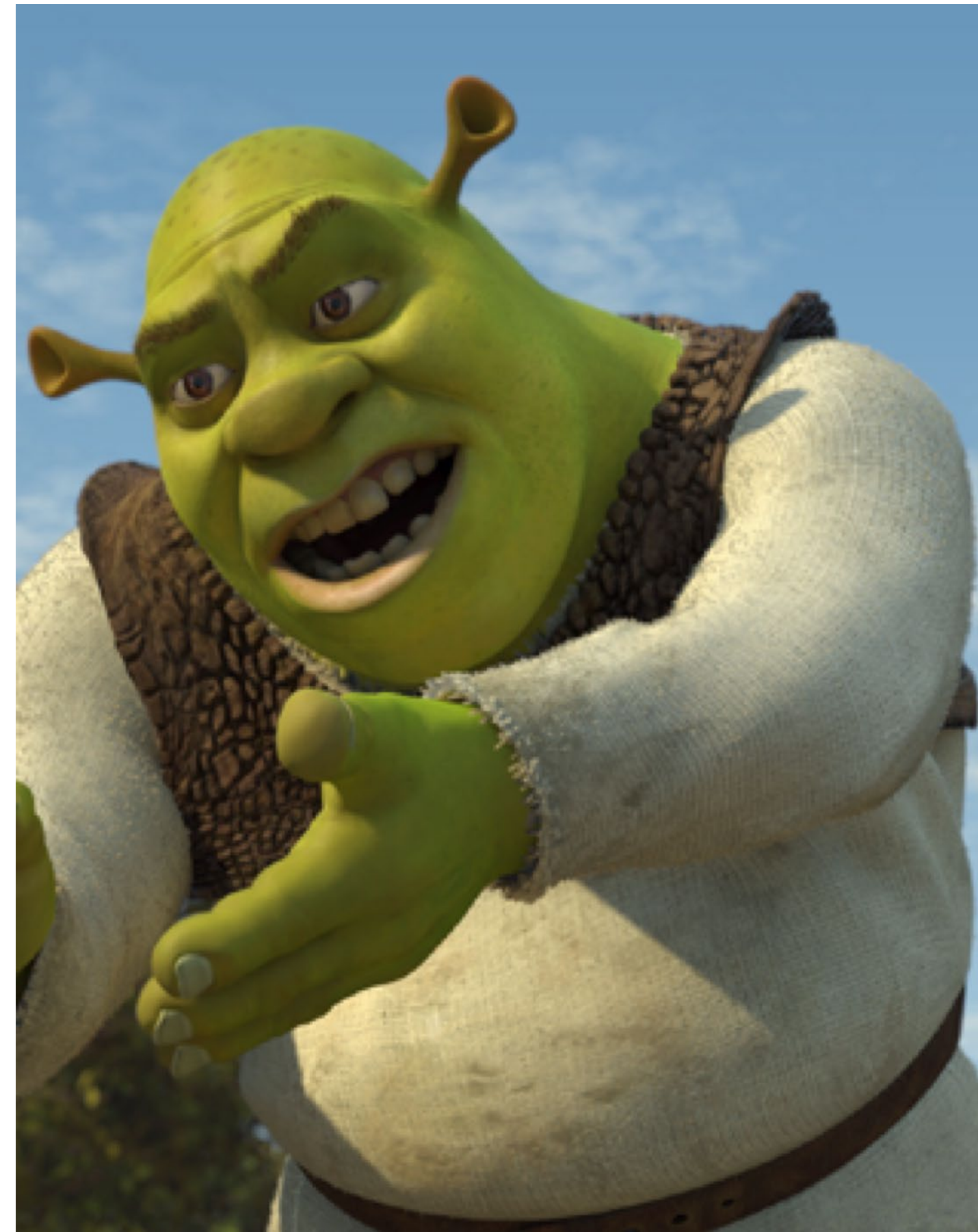
# Direct vs. Indirect Illumination

---

Direct illumination only



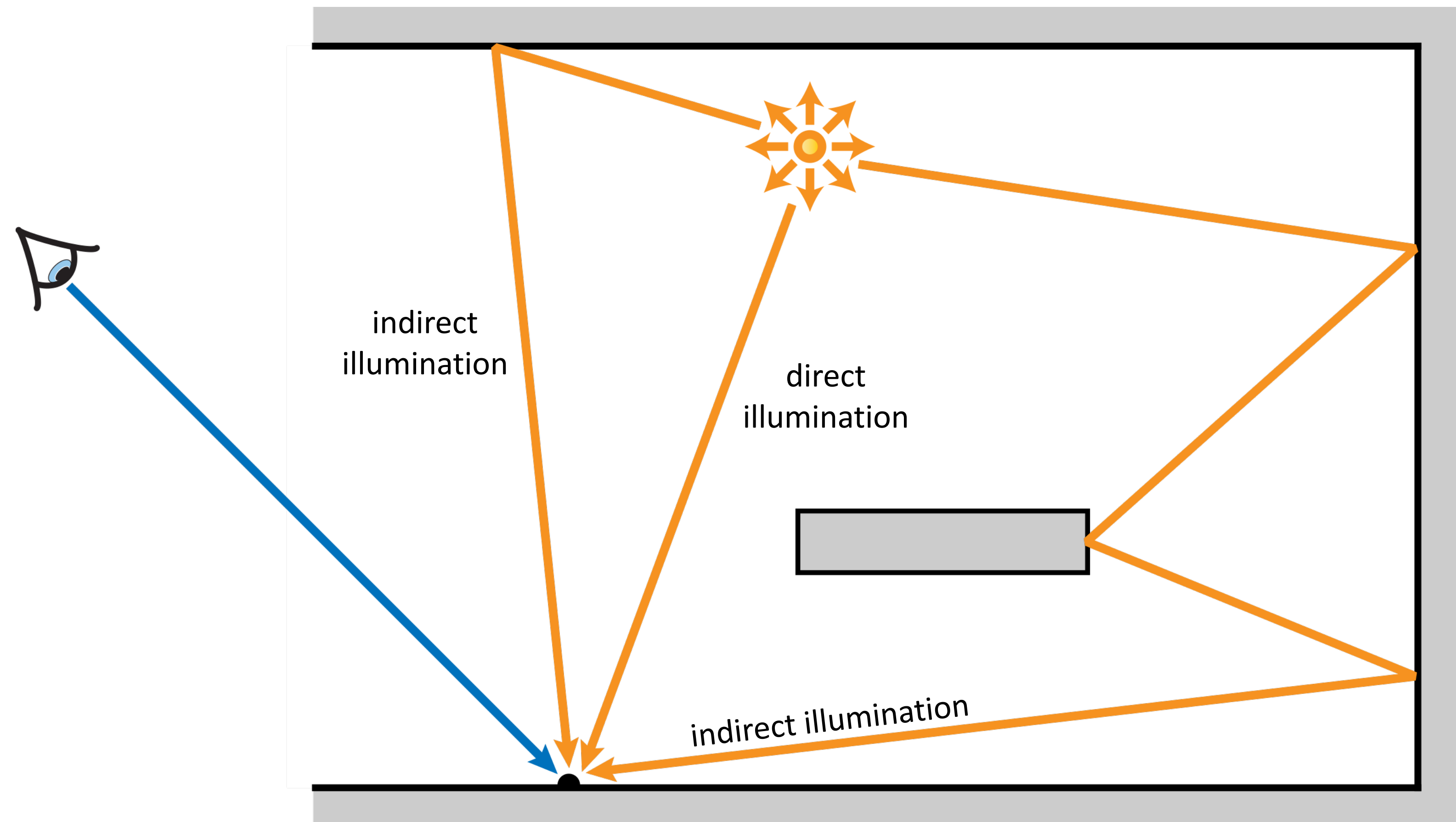
Direct + Indirect illumination



Images courtesy of PDI/DreamWorks

# Importance Sampling Incident Radiance

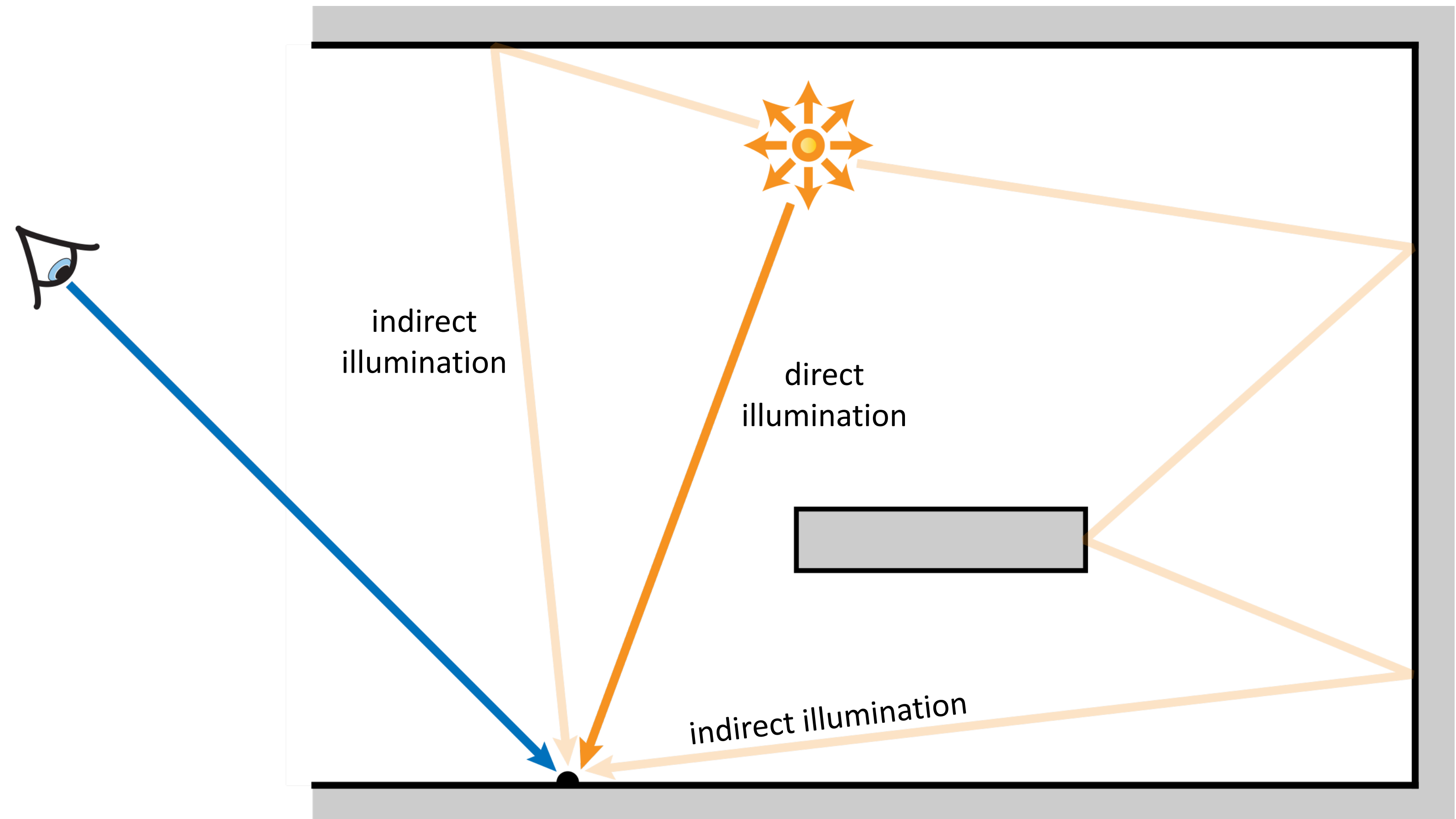
Generally impossible, but...





# Importance Sampling Incident Radiance

Generally impossible, but possible if we assume only direct illumination

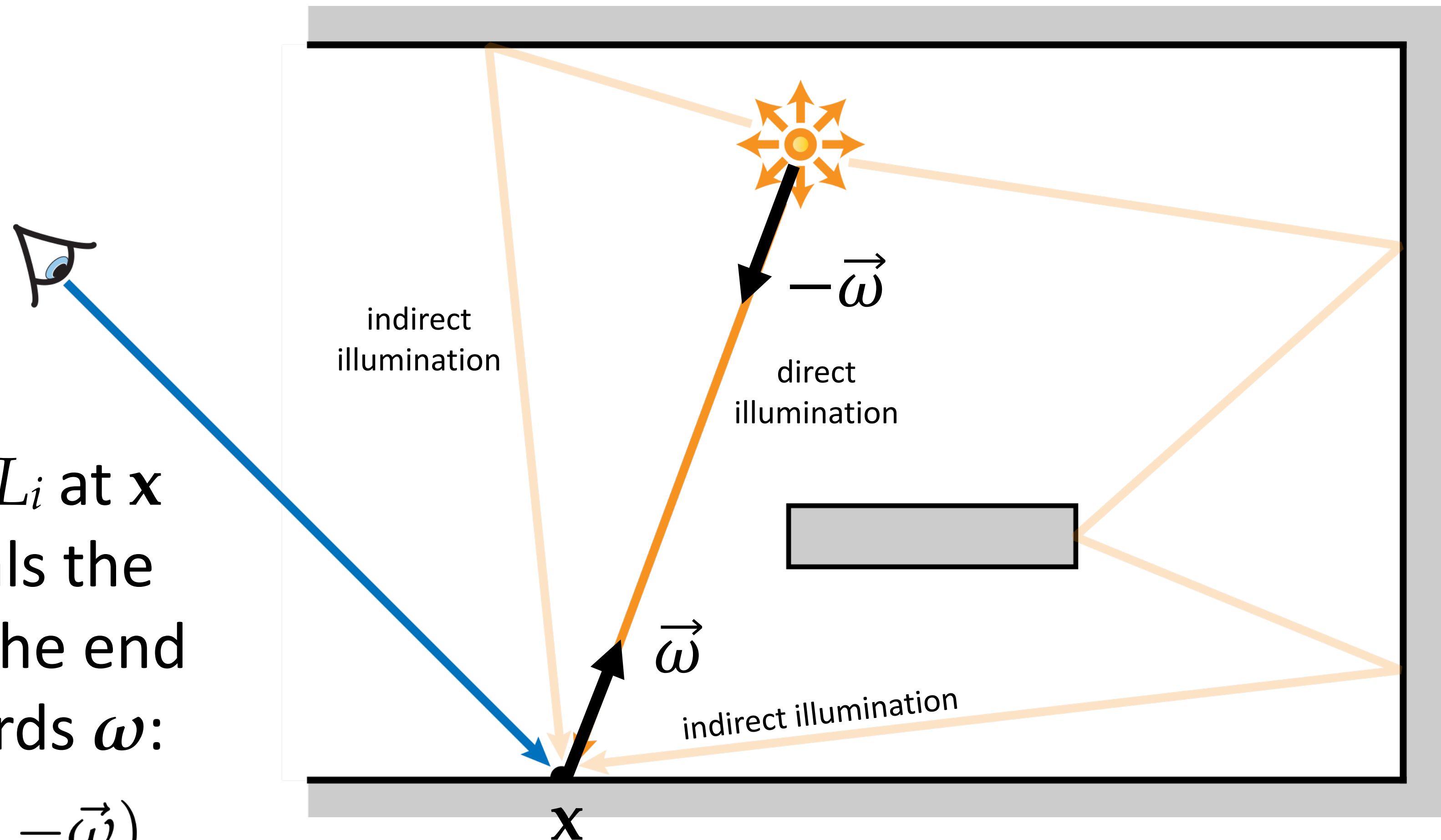


# Direct Illumination

$$L_r(\mathbf{x}, \vec{\omega}_r) = f_r \int_{H^2} f_{ni}(\mathbf{x}, \vec{\omega}_i) L_e(r(\mathbf{x}, \vec{\omega}_i), -\vec{\omega}_i) \cos\theta_i d\omega_i$$

The incident radiance  $L_i$  at  $\mathbf{x}$  from direction  $\omega$  equals the *emitted* radiance  $L_e$  at the end of the ray from  $\mathbf{x}$  towards  $\omega$ :

$$L_i(\mathbf{x}, \vec{\omega}) = L_e(r(\mathbf{x}, \vec{\omega}), -\vec{\omega})$$



# Direct Illumination

---

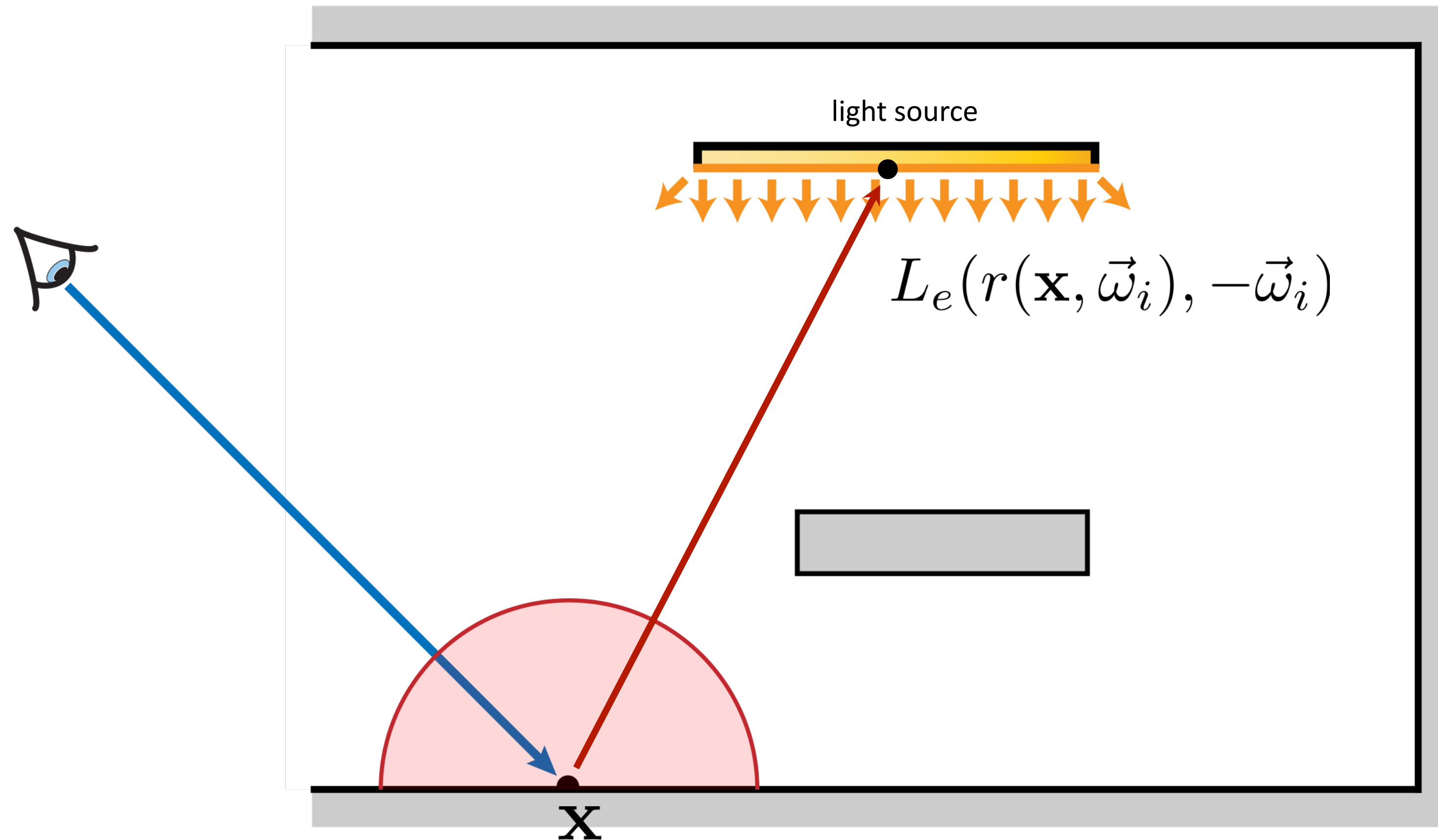
$$L_r(\mathbf{x}, \vec{\omega}_r) = \int_{H^2} f_r(\mathbf{x}, \vec{\omega}_i, \vec{\omega}_r) L_e(r(\mathbf{x}, \vec{\omega}_i), -\vec{\omega}_i) \cos \theta_i \, d\vec{\omega}_i$$

How can we estimate the integral?

$$\langle L_r(\mathbf{x}, \vec{\omega}_r)^N \rangle = \frac{1}{N} \sum_{k=1}^N \frac{f_r(\mathbf{x}, \vec{\omega}_{i,k}, \vec{\omega}_r) L_e(r(\mathbf{x}, \vec{\omega}_{i,k}), -\vec{\omega}_{i,k}) \cos \theta_{i,k}}{p_\Omega(\vec{\omega}_{i,k})}$$

# Direct Illumination

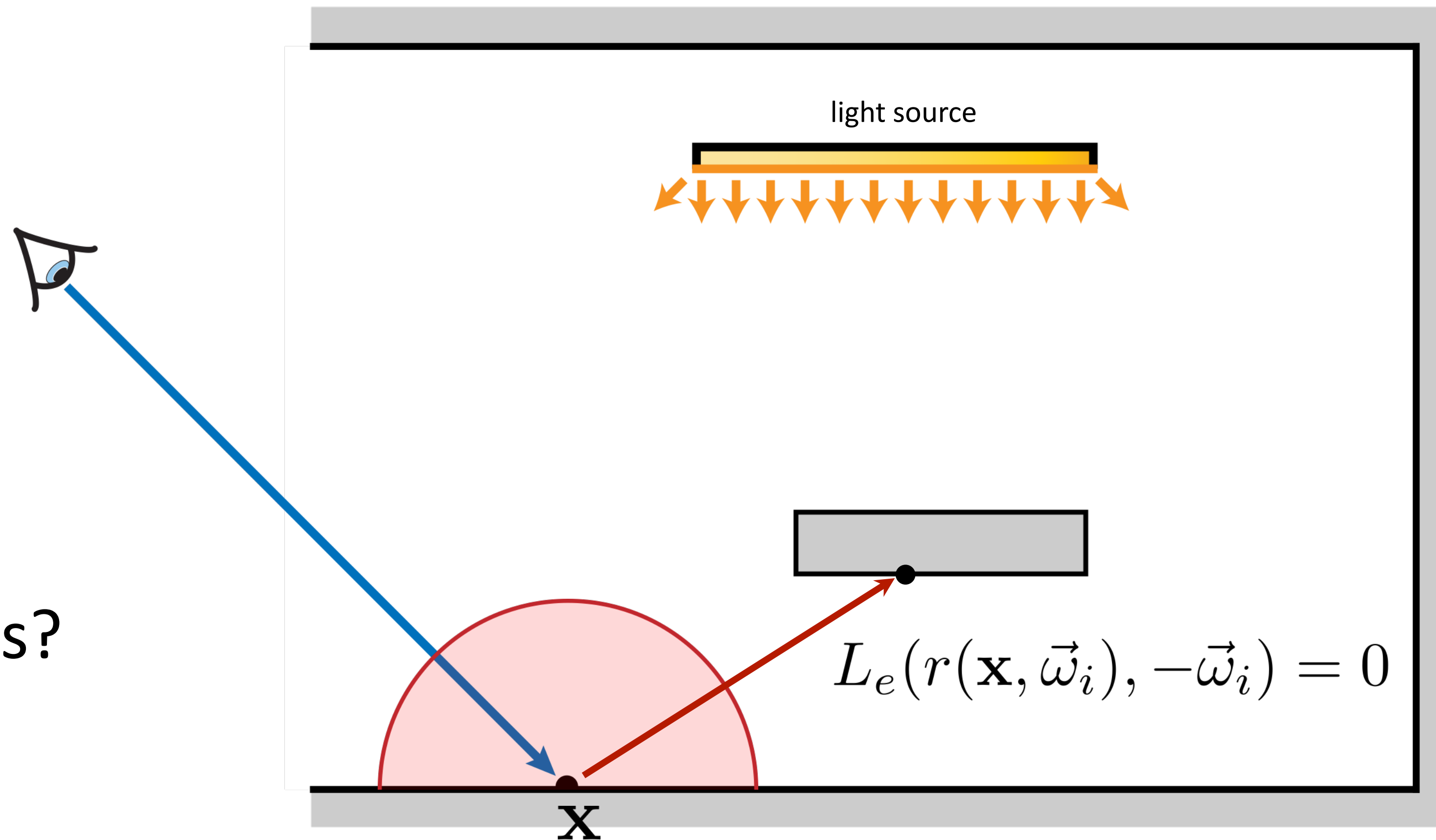
$$L_r(\mathbf{x}, \vec{\omega}_r) = \int_{H^2} f_r(\mathbf{x}, \vec{\omega}_i, \vec{\omega}_r) L_e(r(\mathbf{x}, \vec{\omega}_i), -\vec{\omega}_i) \cos \theta_i d\vec{\omega}_i$$



# Direct Illumination

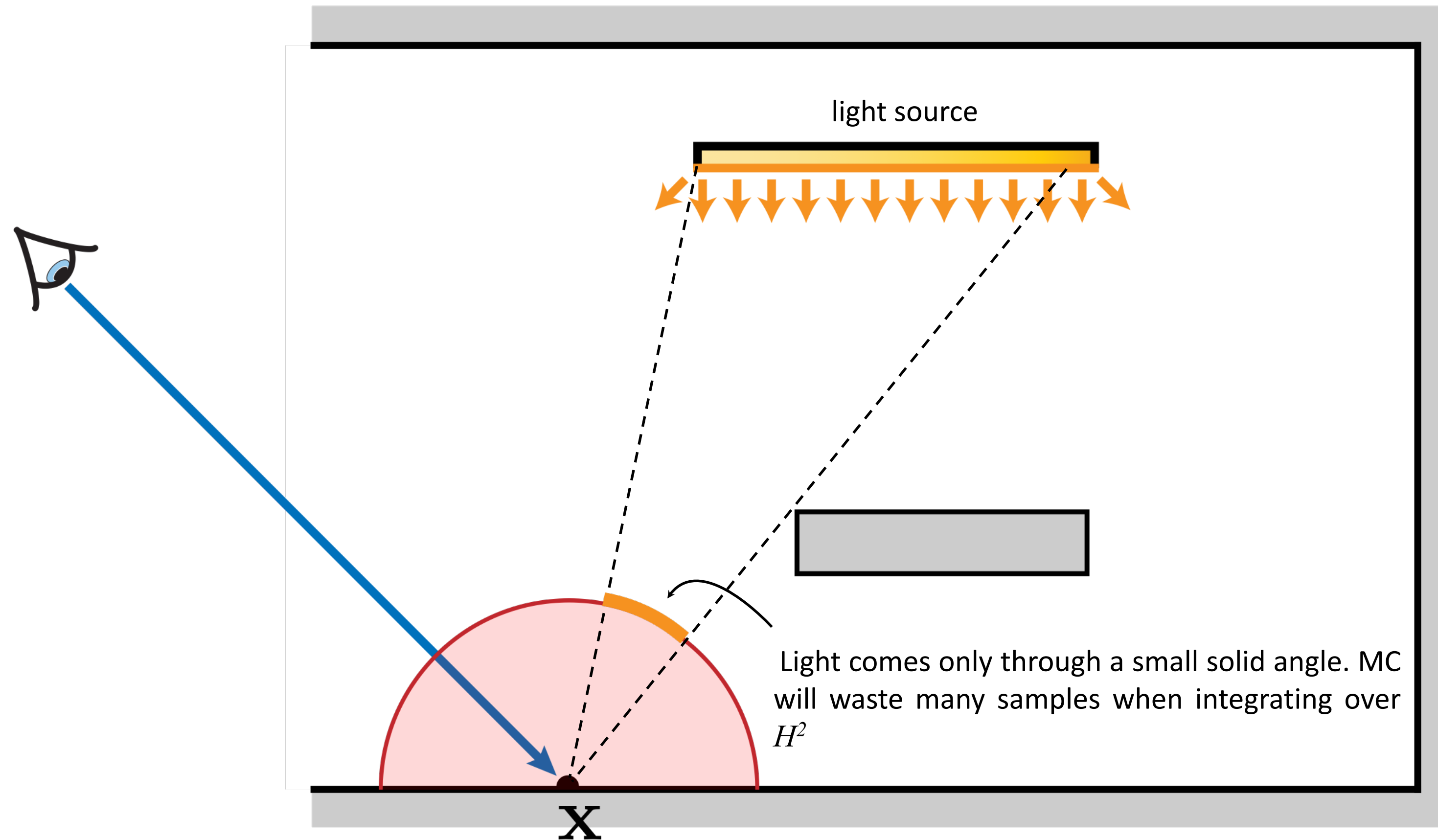
$$L_r(\mathbf{x}, \vec{\omega}_r) = \int_{H^2} f_r(\mathbf{x}, \vec{\omega}_i, \vec{\omega}_r) L_e(r(\mathbf{x}, \vec{\omega}_i), -\vec{\omega}_i) \cos \theta_i d\vec{\omega}_i$$

Any problems?



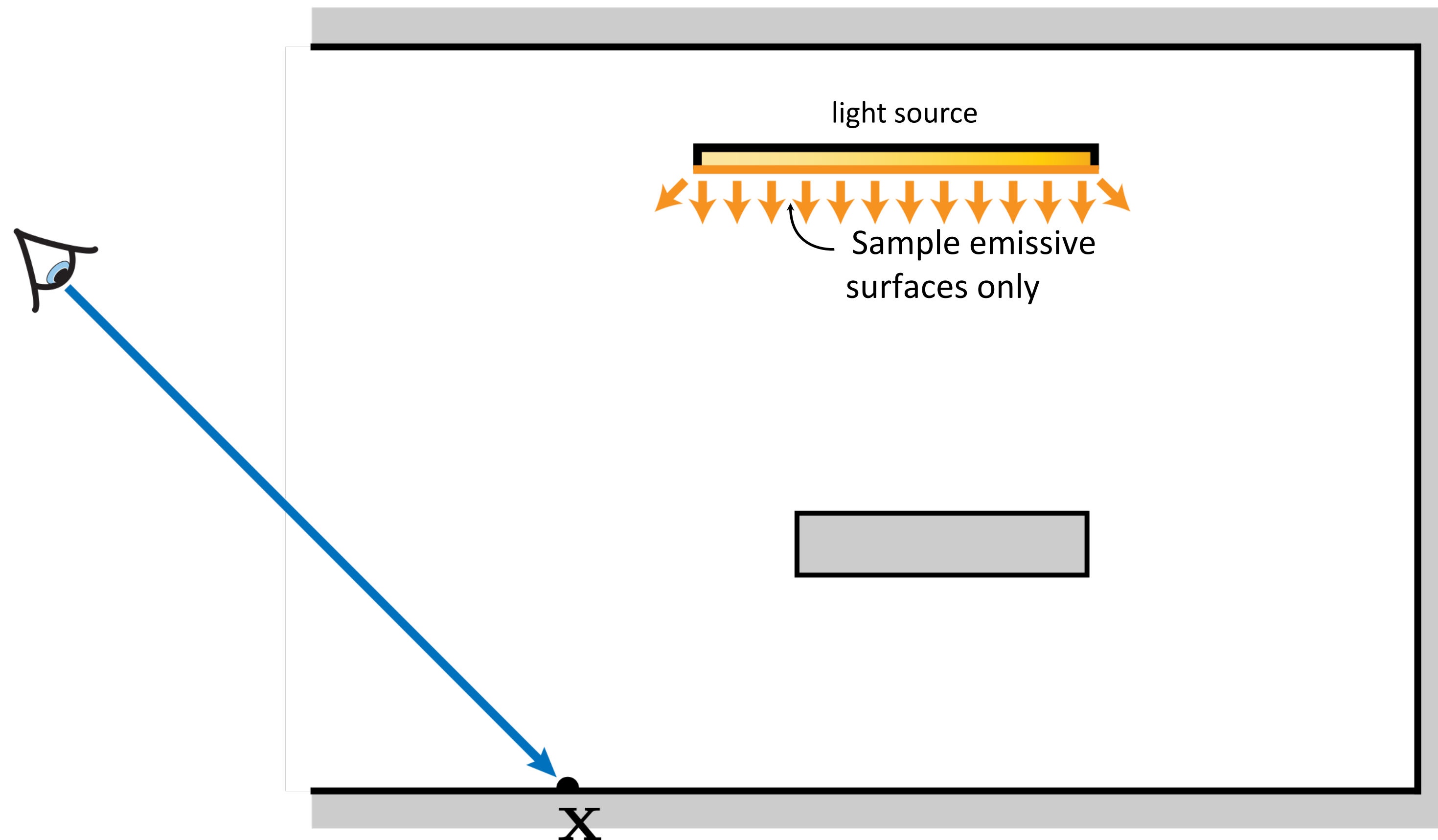
# Direct Illumination

$$L_r(\mathbf{x}, \vec{\omega}_r) = \int_{H^2} f_r(\mathbf{x}, \vec{\omega}_i, \vec{\omega}_r) L_e(r(\mathbf{x}, \vec{\omega}_i), -\vec{\omega}_i) \cos \theta_i d\vec{\omega}_i$$



# Direct Illumination

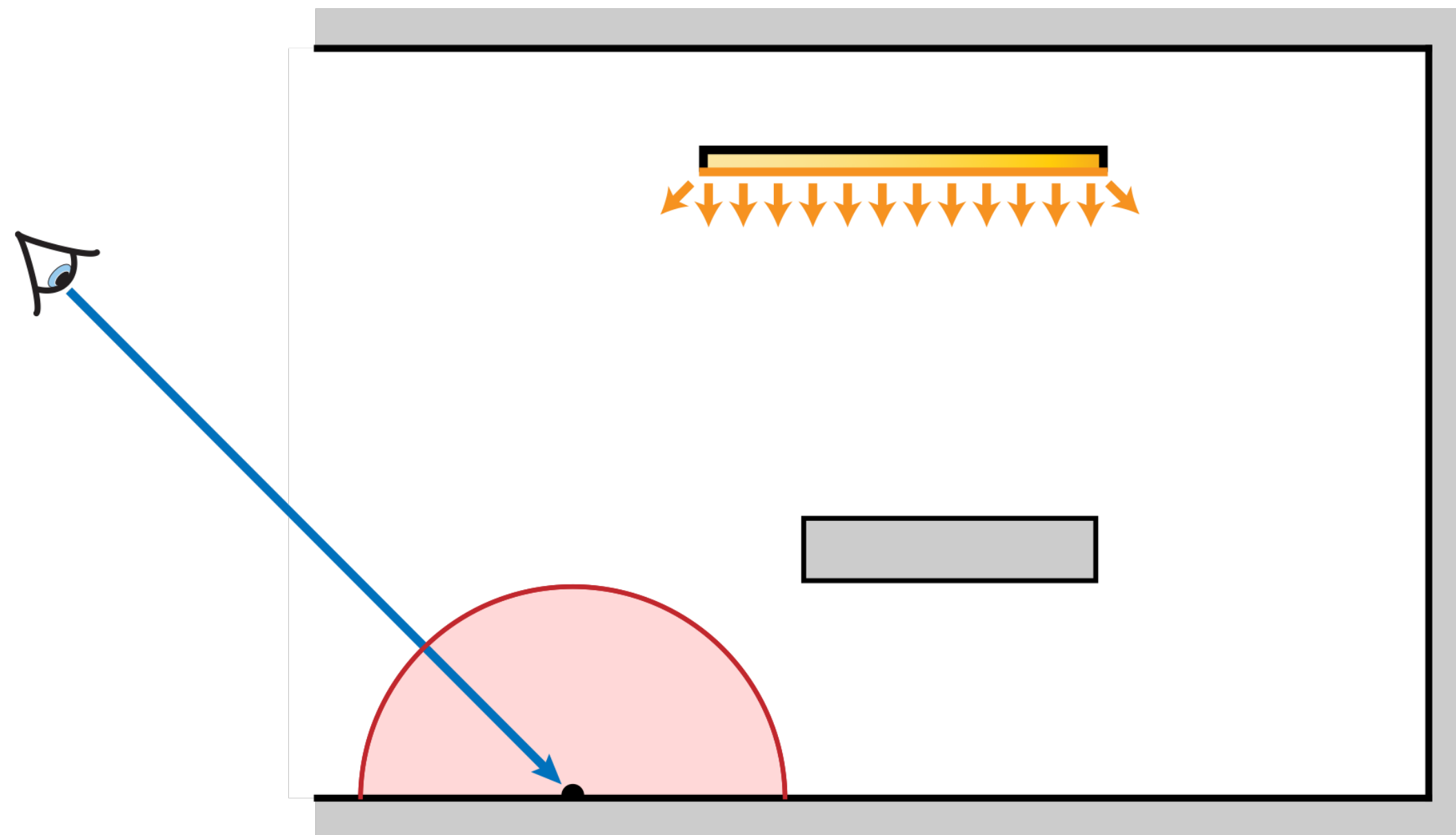
$$L_r(\mathbf{x}, \vec{\omega}_r) = \int_{H^2} f_r(\mathbf{x}, \vec{\omega}_i, \vec{\omega}_r) L_e(r(\mathbf{x}, \vec{\omega}_i), -\vec{\omega}_i) \cos \theta_i d\vec{\omega}_i$$



For direct illumination, it would be better to explicitly sample emissive surfaces

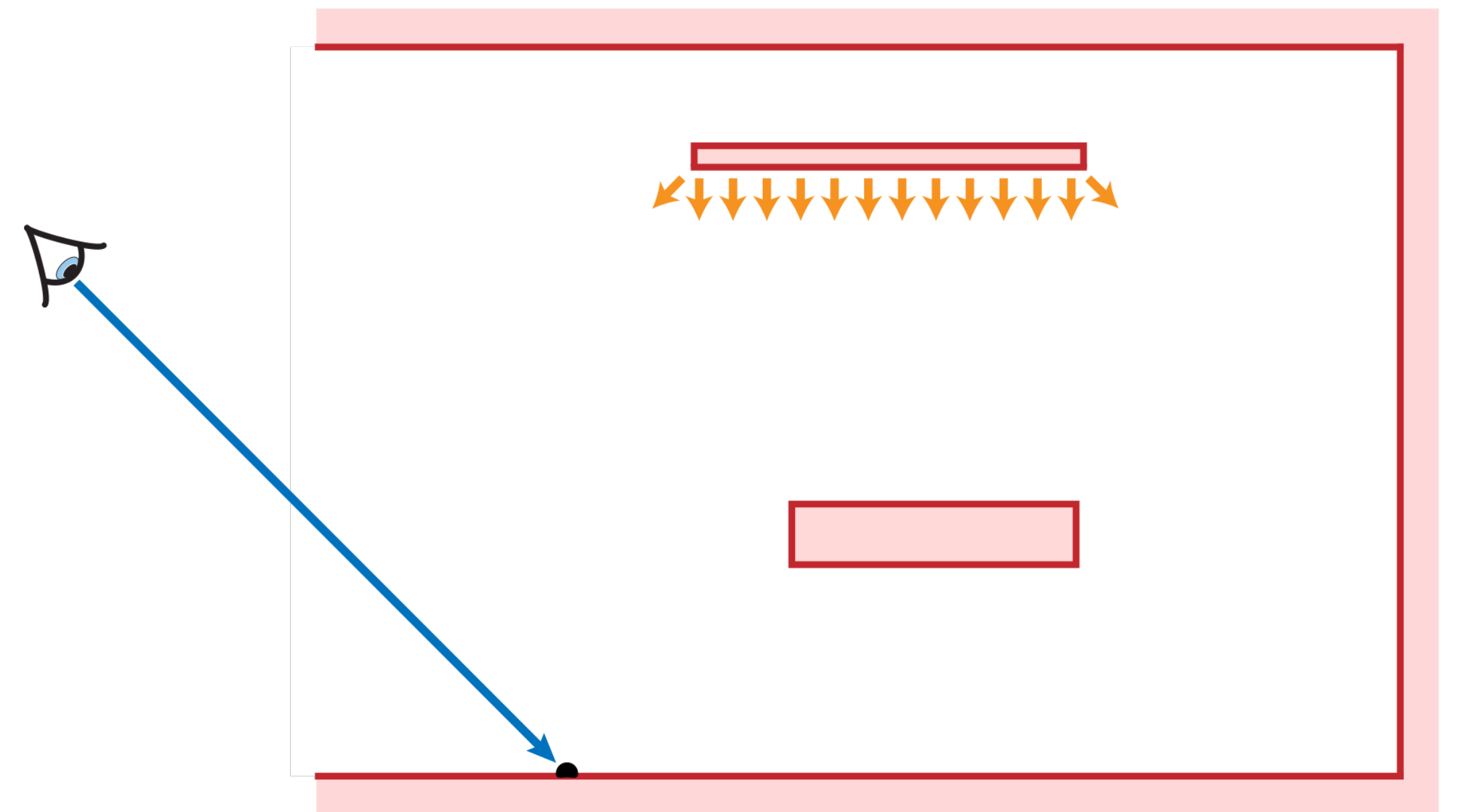
# Forms of Reflection Equation

Hemispherical  
integration



$$L_r(\mathbf{x}, \vec{\omega}_r) = \int_{H^2} f_r(\mathbf{x}, \vec{\omega}_i, \vec{\omega}_r) L_i(\mathbf{x}, \vec{\omega}_i) \cos \theta_i d\vec{\omega}_i$$

Surface Area  
integration



$$L_r(\mathbf{x}, \mathbf{z}) = \int_A f_r(\mathbf{x}, \mathbf{y}, \mathbf{z}) L_i(\mathbf{x}, \mathbf{y}) G(\mathbf{x}, \mathbf{y}) dA(\mathbf{y})$$



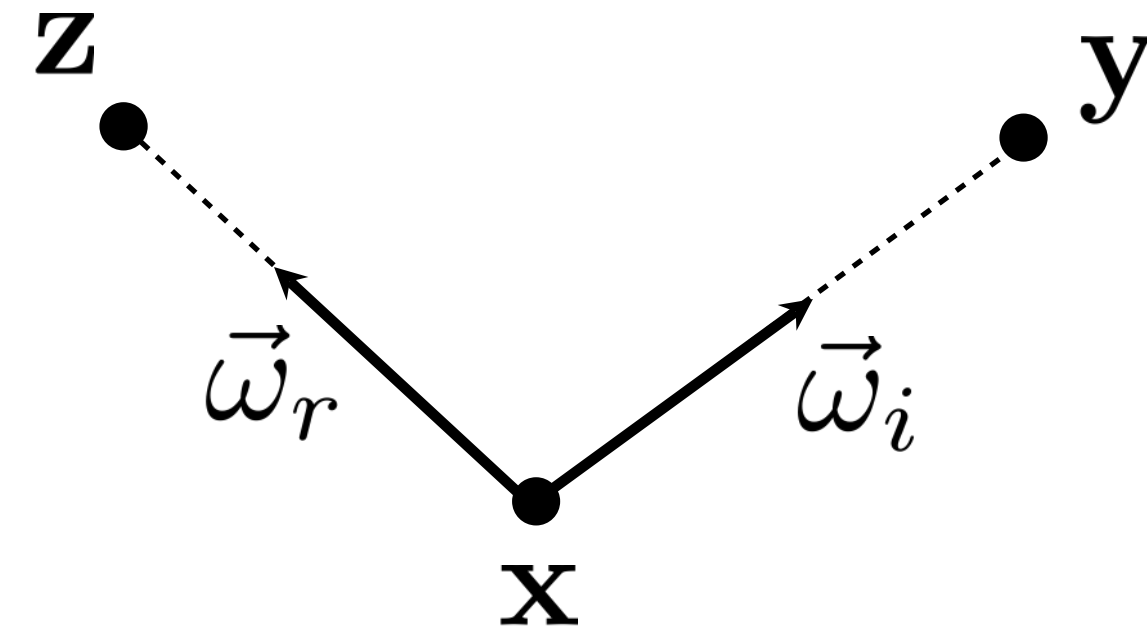
# Forms of Reflection Equation

Change in notation:

$$L_i(\mathbf{x}, \vec{\omega}_i) = L_i(\mathbf{x}, \mathbf{y})$$

$$L_r(\mathbf{x}, \vec{\omega}_r) = L_r(\mathbf{x}, \mathbf{z})$$

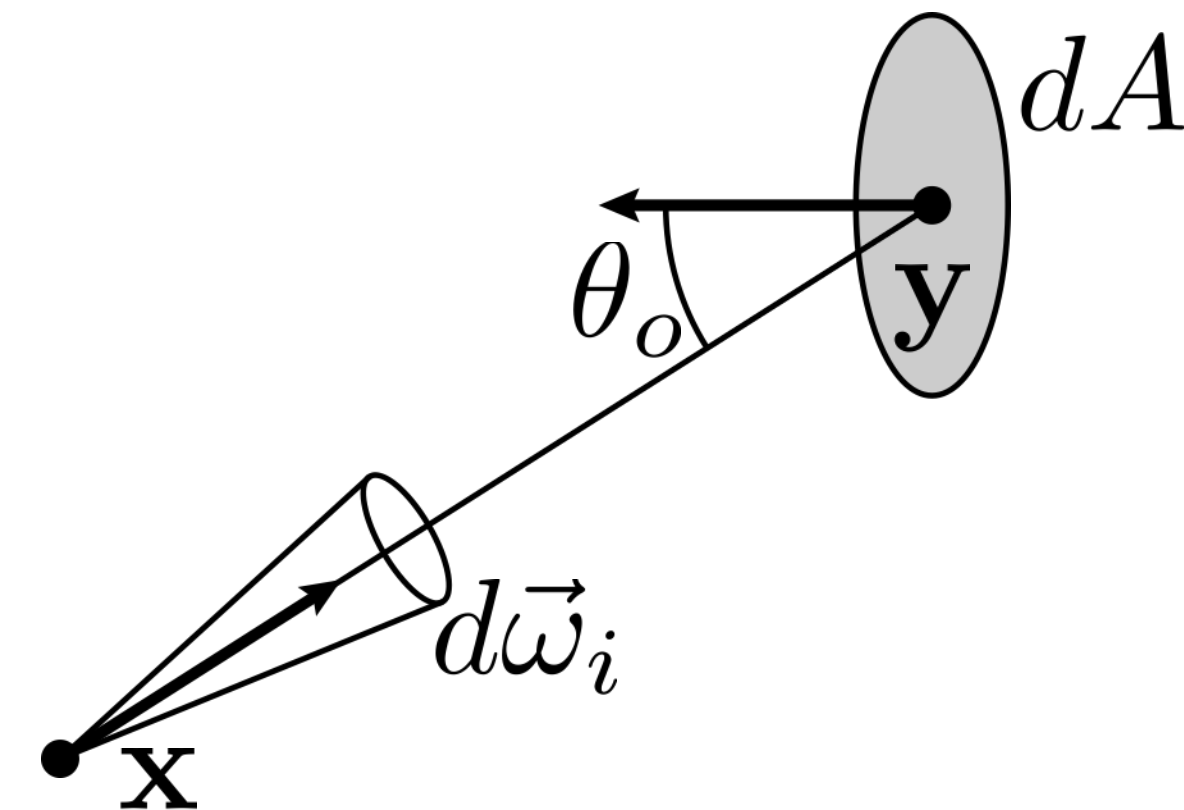
$$f_r(\mathbf{x}, \vec{\omega}_i, \vec{\omega}_r) = f_r(\mathbf{x}, \mathbf{y}, \mathbf{z})$$



Transform integral over directions into integral over surface area.

Jacobian determinant of the trans.:

$$d\vec{\omega}_i = \frac{|\cos \theta_o|}{\|\mathbf{x} - \mathbf{y}\|^2} dA$$



# Forms of Reflection Equation

---

$$L_i(\mathbf{x}, \vec{\omega}_i) = L_i(\mathbf{x}, \mathbf{y})$$

$$L_r(\mathbf{x}, \vec{\omega}_r) = L_r(\mathbf{x}, \mathbf{z})$$

$$f_r(\mathbf{x}, \vec{\omega}_i, \vec{\omega}_r) = f_r(\mathbf{x}, \mathbf{y}, \mathbf{z})$$

$$d\vec{\omega}_i = \frac{|\cos \theta_o|}{\|\mathbf{x} - \mathbf{y}\|^2} dA$$

Hemispherical form:

$$L_r(\mathbf{x}, \vec{\omega}_r) = \int_{H^2} f_r(\mathbf{x}, \vec{\omega}_i, \vec{\omega}_r) L_i(\mathbf{x}, \vec{\omega}_i) \cos \theta_i d\vec{\omega}_i$$

Surface area form:

$$L_r(\mathbf{x}, \mathbf{z}) = \int_A f_r(\mathbf{x}, \mathbf{y}, \mathbf{z}) L_i(\mathbf{x}, \mathbf{y}) G(\mathbf{x}, \mathbf{y}) dA(\mathbf{y})$$

# Area Form of the Reflection Eq.

---

$$L_r(\mathbf{x}, \mathbf{z}) = \int_A f_r(\mathbf{x}, \mathbf{y}, \mathbf{z}) L_i(\mathbf{x}, \mathbf{y}) G(\mathbf{x}, \mathbf{y}) dA(\mathbf{y})$$

Geometry term:

$$G(\mathbf{x}, \mathbf{y}) = V(\mathbf{x}, \mathbf{y}) \frac{|\cos \theta_i| |\cos \theta_o|}{\|\mathbf{x} - \mathbf{y}\|^2}$$

Visibility term:

$$V(\mathbf{x}, \mathbf{y}) = \begin{cases} 1 & : \text{ visible} \\ 0 & : \text{ not visible} \end{cases}$$

# Area Form of the Reflection Eq.

---

$$L_r(\mathbf{x}, \mathbf{z}) = \int_A f_r(\mathbf{x}, \mathbf{y}, \mathbf{z}) L_i(\mathbf{x}, \mathbf{y}) G(\mathbf{x}, \mathbf{y}) dA(\mathbf{y})$$

Original foreshortening term

Geometry term:

$$G(\mathbf{x}, \mathbf{y}) = V(\mathbf{x}, \mathbf{y}) \frac{|\cos \theta_i| |\cos \theta_o|}{\|\mathbf{x} - \mathbf{y}\|^2}$$

Visibility term:

$$V(\mathbf{x}, \mathbf{y}) = \begin{cases} 1 & : \text{ visible} \\ 0 & : \text{ not visible} \end{cases}$$

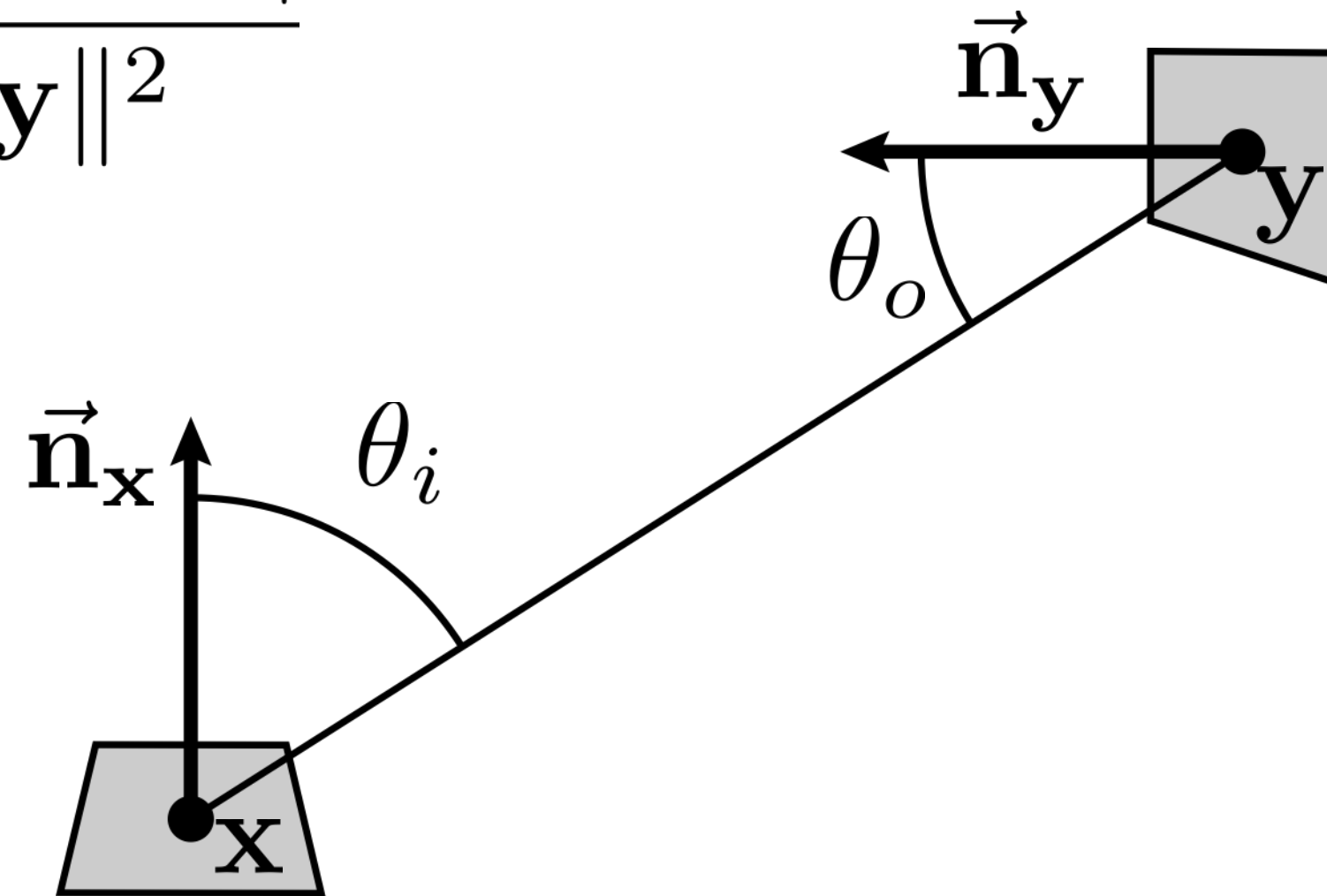
Jacobian determinant  
of the transform

$$d\vec{\omega}_i = \frac{|\cos \theta_o|}{\|\mathbf{x} - \mathbf{y}\|^2} dA$$

# Area Form of the Reflection Eq.

Interpreting

$$\frac{|\cos \theta_i| |\cos \theta_o|}{\|\mathbf{x} - \mathbf{y}\|^2}$$



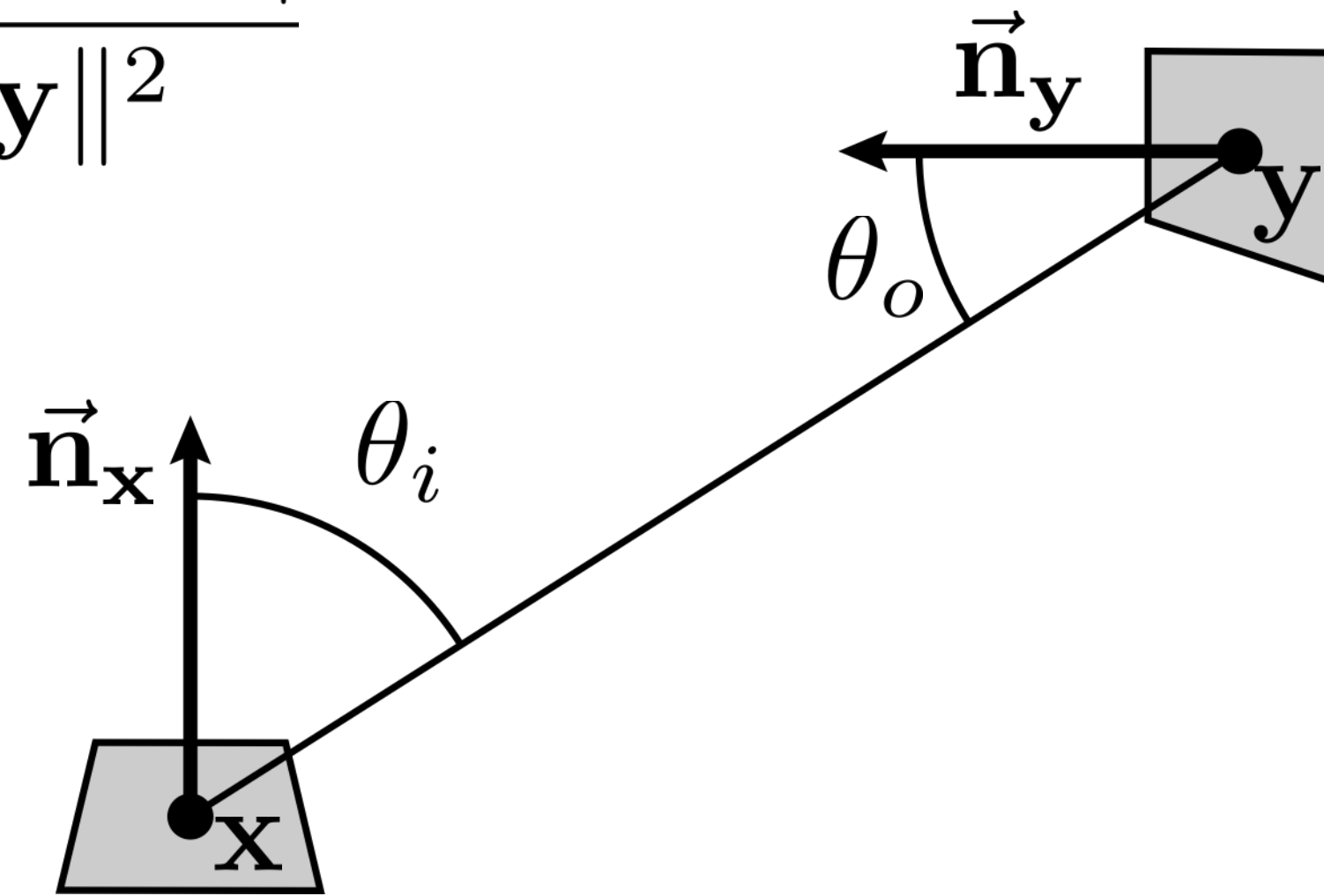
The chance that a photon emitted from a differential patch will hit another diff. patch decreases as:

- the patches face away from each other (numerator)
- the patches move away from each other (denominator)

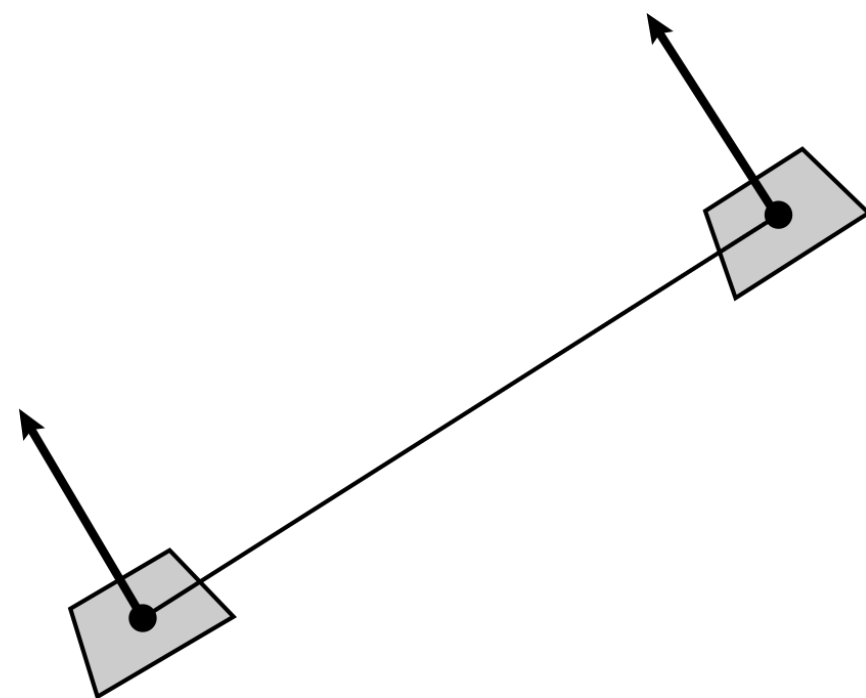
# Area Form of the Reflection Eq.

Interpreting

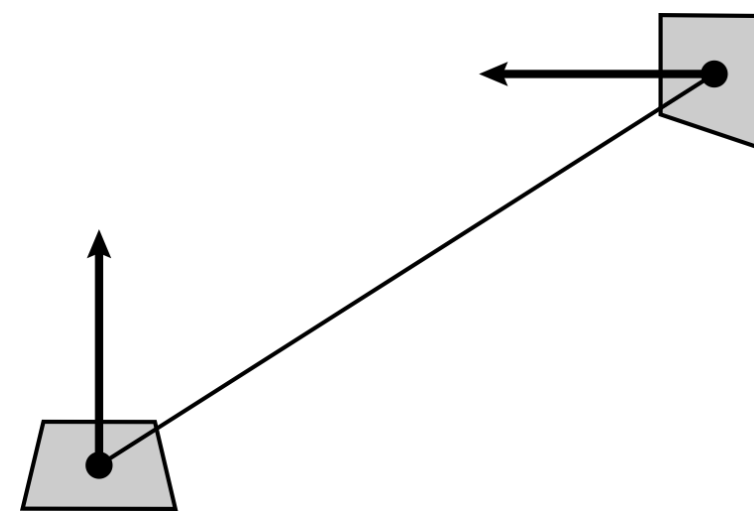
$$\frac{|\cos \theta_i| |\cos \theta_o|}{\|\mathbf{x} - \mathbf{y}\|^2}$$



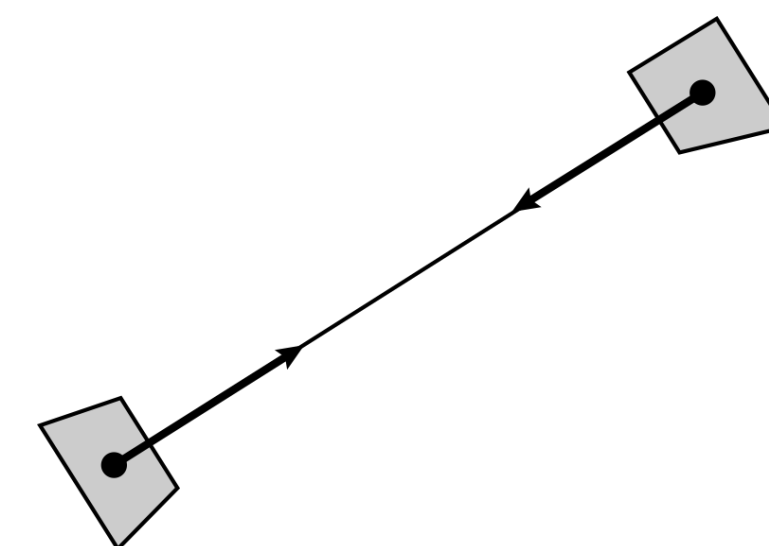
numerator = 0



$0 < \text{numerator} < 1$

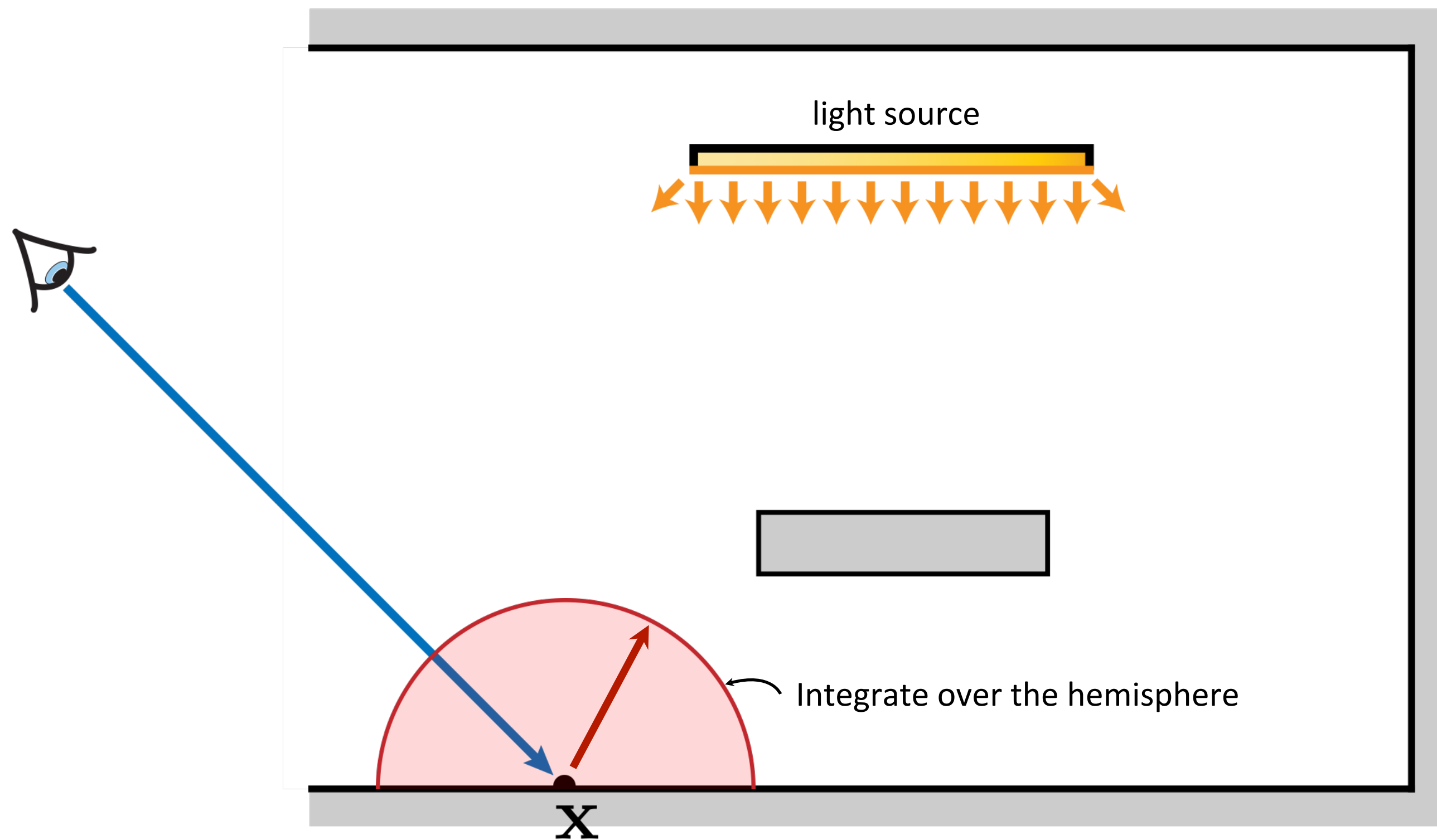


numerator = 1



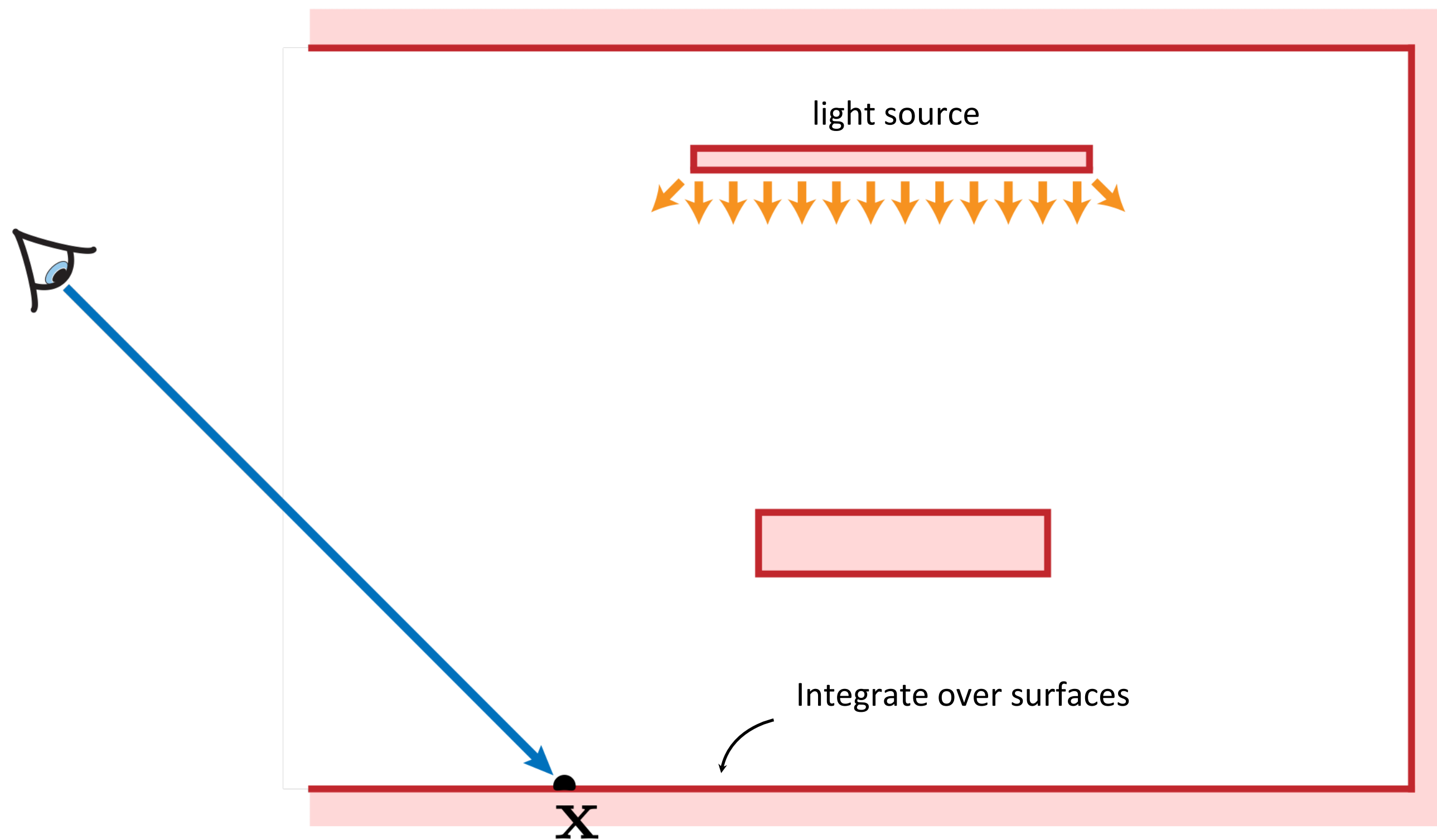
# Direct Illumination

$$L_r(\mathbf{x}, \vec{\omega}_r) = \int_{H^2} f_r(\mathbf{x}, \vec{\omega}_i, \vec{\omega}_r) L_e(r(\mathbf{x}, \vec{\omega}_i), -\vec{\omega}_i) \cos \theta_i d\vec{\omega}_i$$



# Direct Illumination

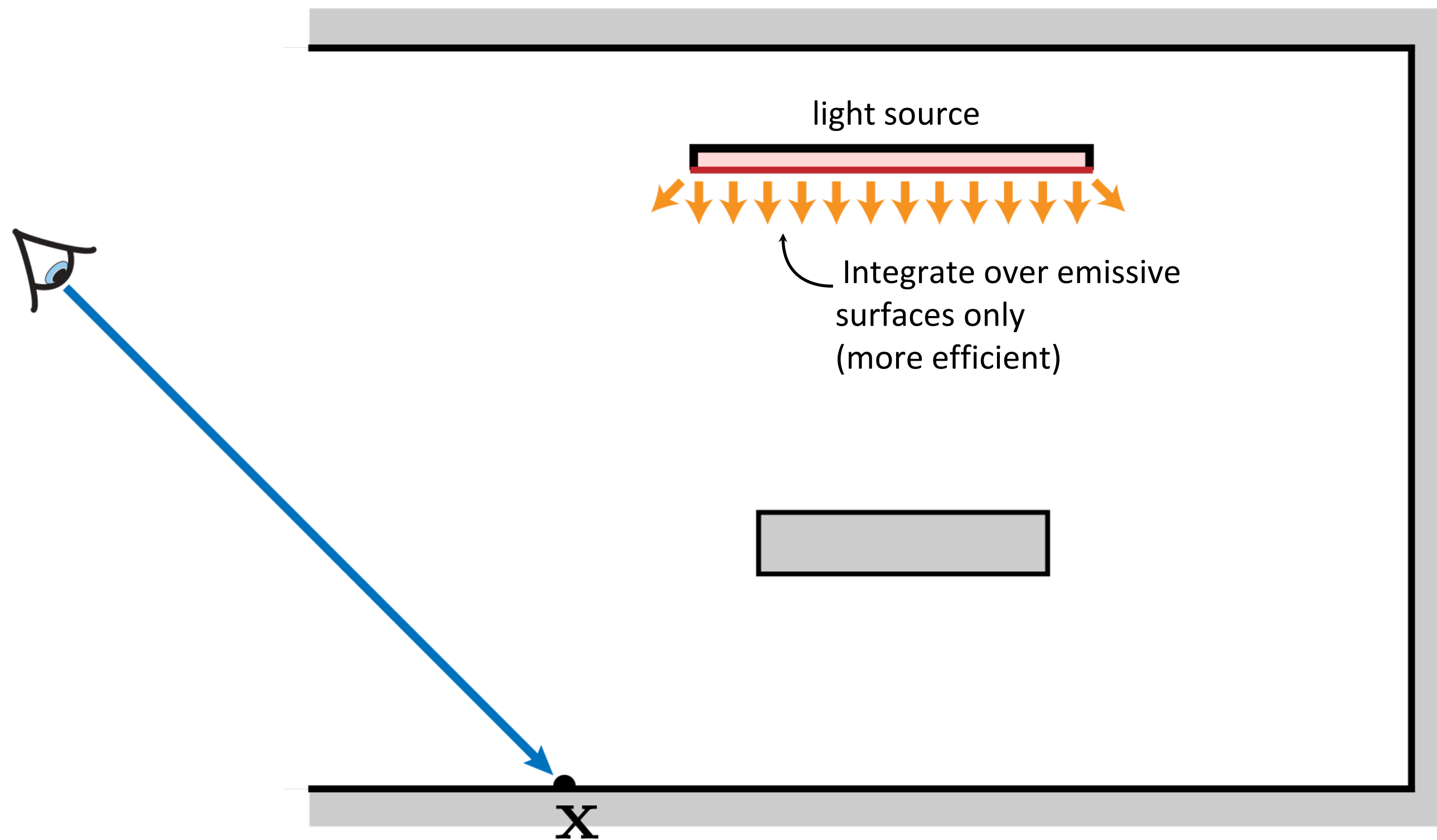
$$L_r(\mathbf{x}, \mathbf{z}) = \int_A f_r(\mathbf{x}, \mathbf{y}, \mathbf{z}) L_e(\mathbf{y}, \mathbf{x}) V(\mathbf{x}, \mathbf{y}) \frac{|\cos \theta_i| |\cos \theta_o|}{\|\mathbf{x} - \mathbf{y}\|^2} dA(\mathbf{y})$$





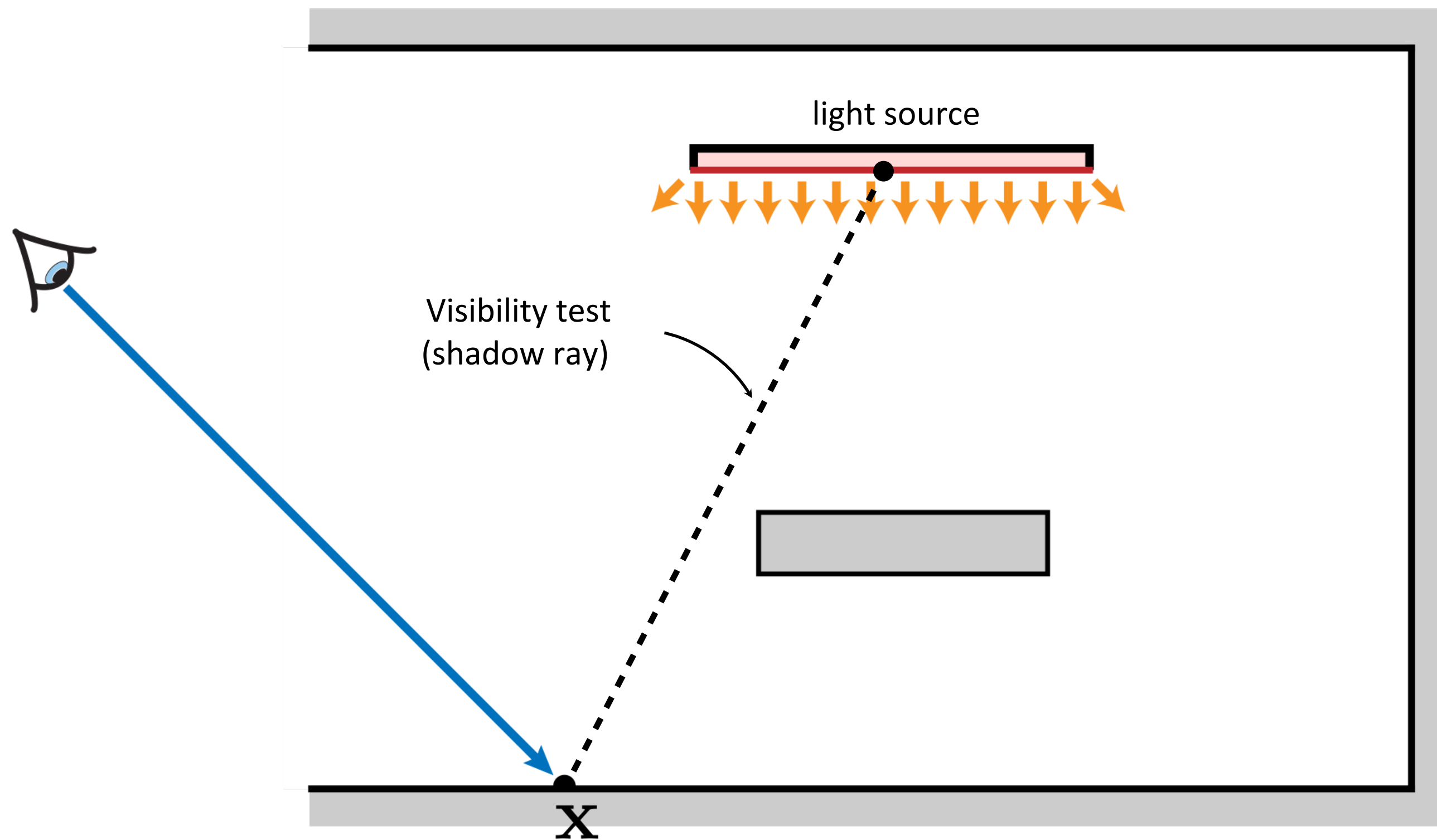
# Direct Illumination

$$L_r(\mathbf{x}, \mathbf{z}) = \int_{A_e} f_r(\mathbf{x}, \mathbf{y}, \mathbf{z}) L_e(\mathbf{y}, \mathbf{x}) V(\mathbf{x}, \mathbf{y}) \frac{|\cos \theta_i| |\cos \theta_o|}{\|\mathbf{x} - \mathbf{y}\|^2} dA(\mathbf{y})$$



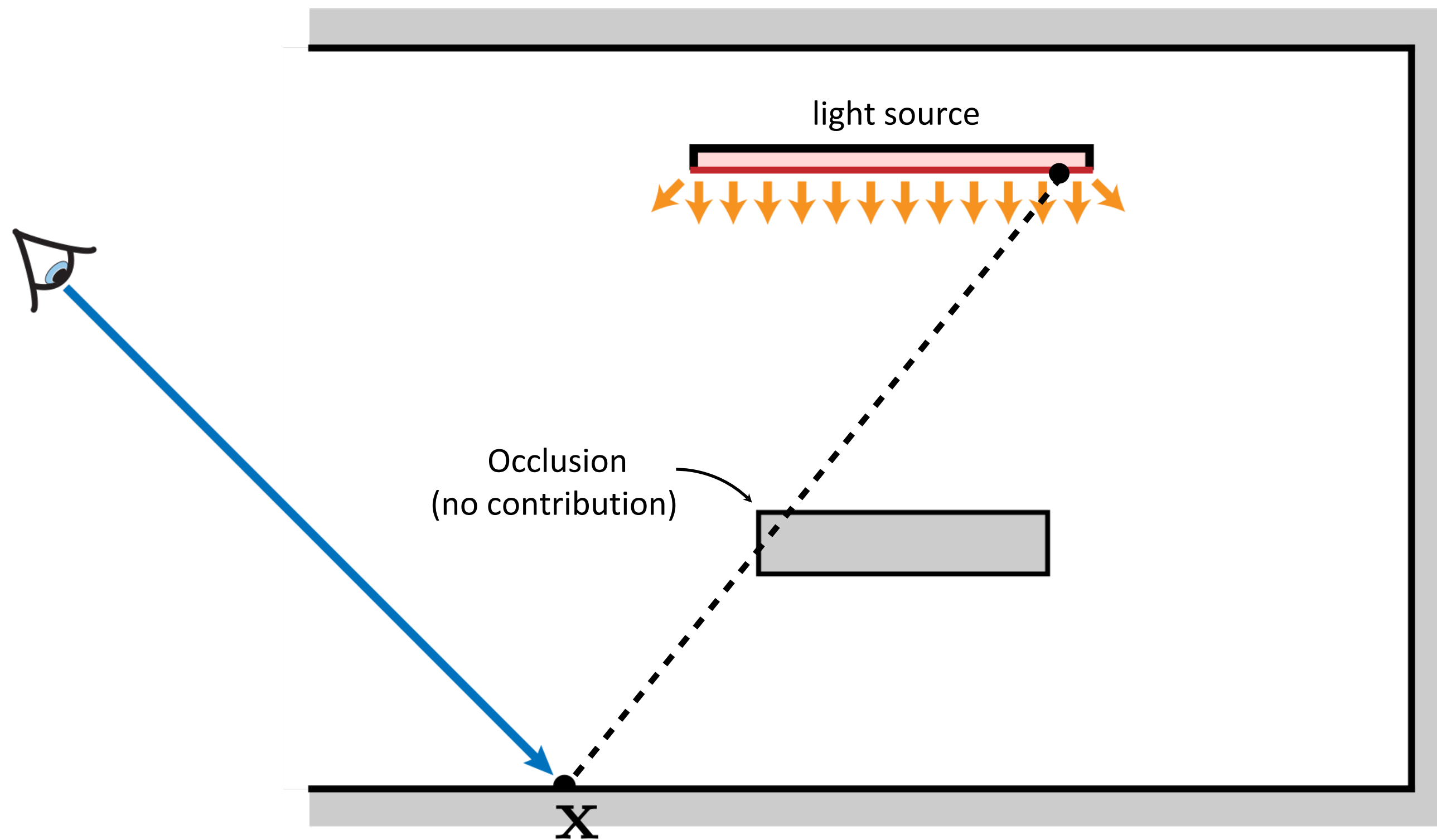
# Direct Illumination

$$L_r(\mathbf{x}, \mathbf{z}) = \int_{A_e} f_r(\mathbf{x}, \mathbf{y}, \mathbf{z}) L_e(\mathbf{y}, \mathbf{x}) V(\mathbf{x}, \mathbf{y}) \frac{|\cos \theta_i| |\cos \theta_o|}{\|\mathbf{x} - \mathbf{y}\|^2} dA(\mathbf{y})$$



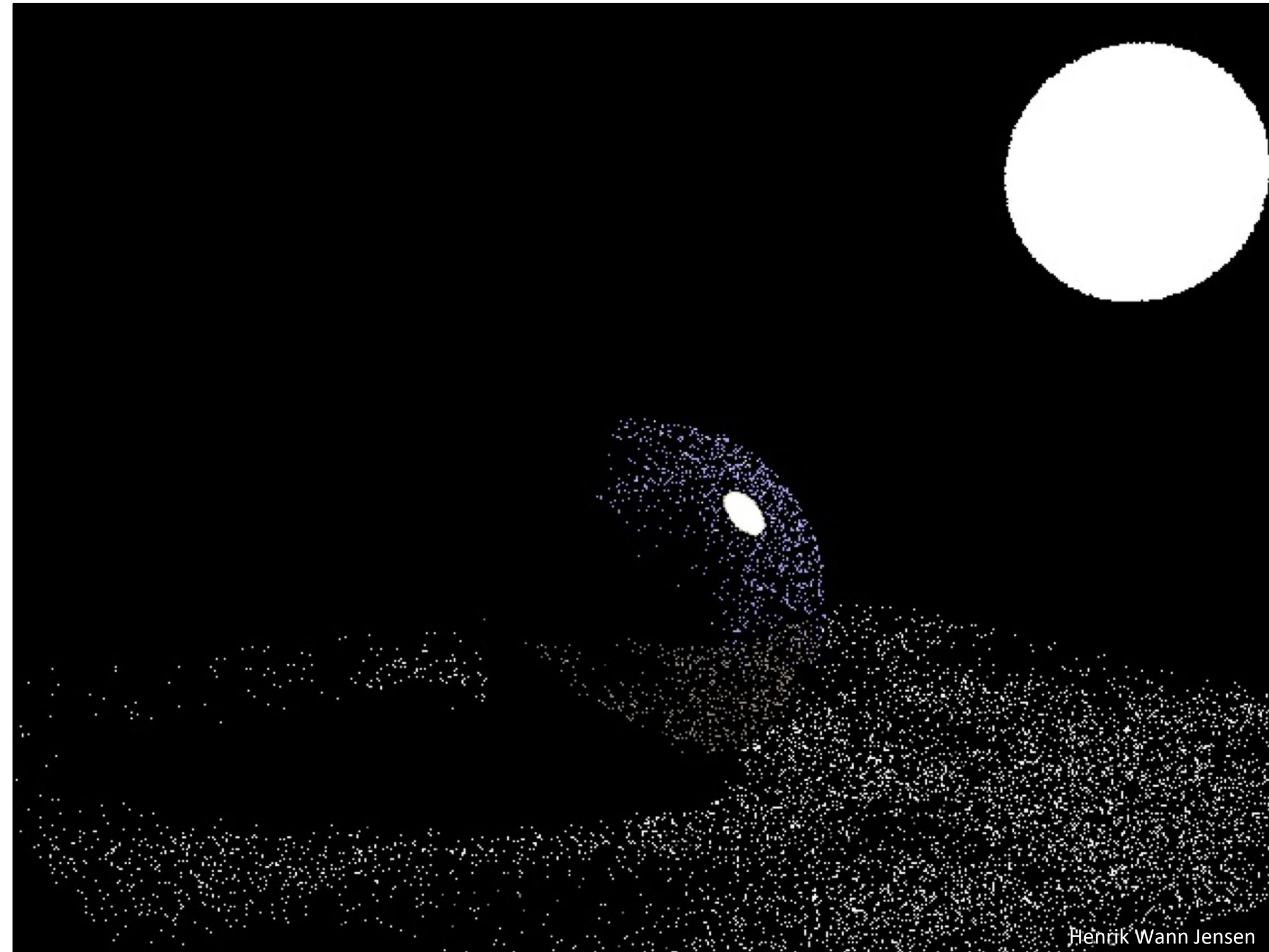
# Direct Illumination

$$L_r(\mathbf{x}, \mathbf{z}) = \int_{A_e} f_r(\mathbf{x}, \mathbf{y}, \mathbf{z}) L_e(\mathbf{y}, \mathbf{x}) V(\mathbf{x}, \mathbf{y}) \frac{|\cos \theta_i| |\cos \theta_o|}{\|\mathbf{x} - \mathbf{y}\|^2} dA(\mathbf{y})$$



# Direct Illumination

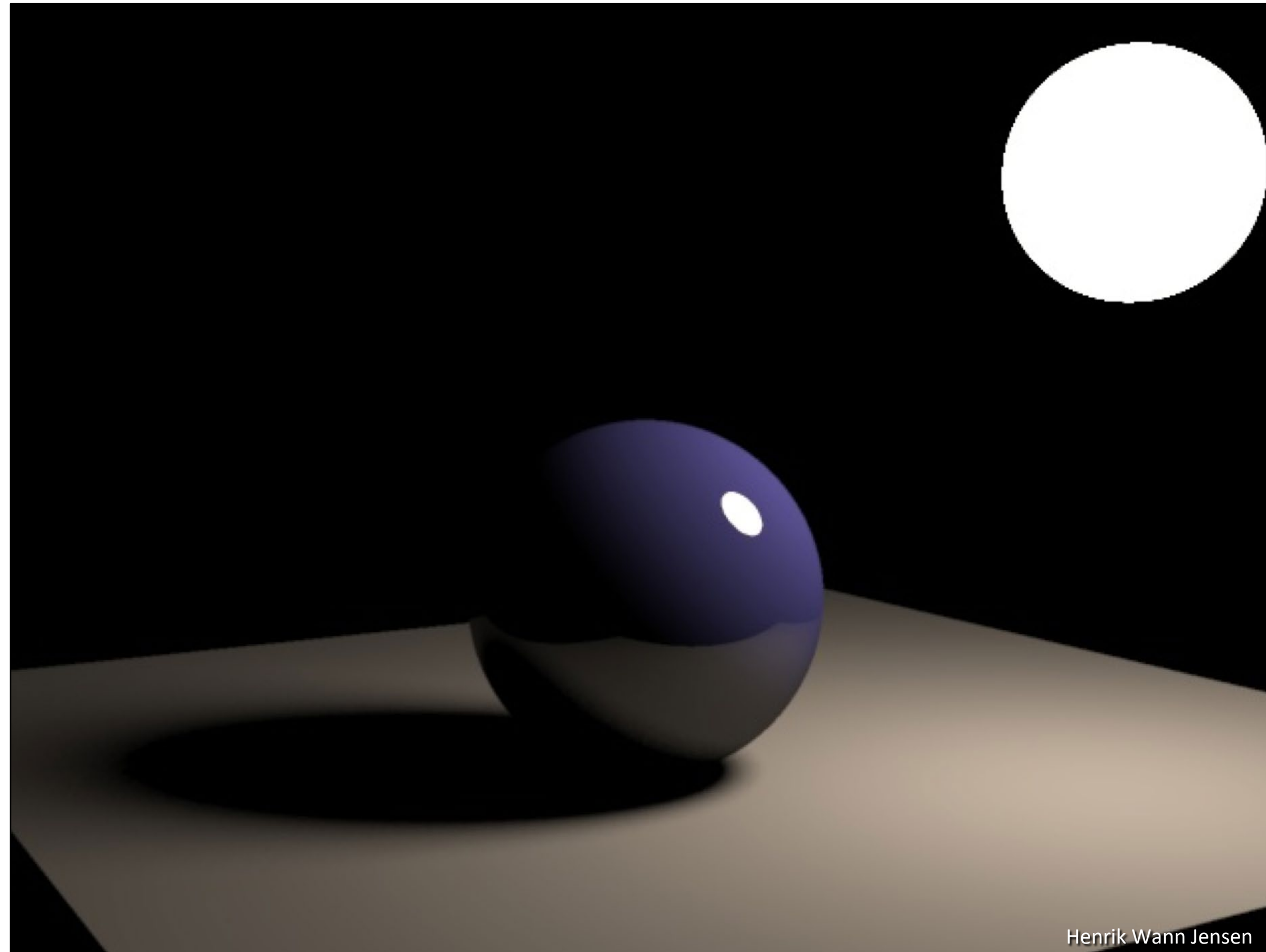
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Sampling the hemisphere

# Direct Illumination

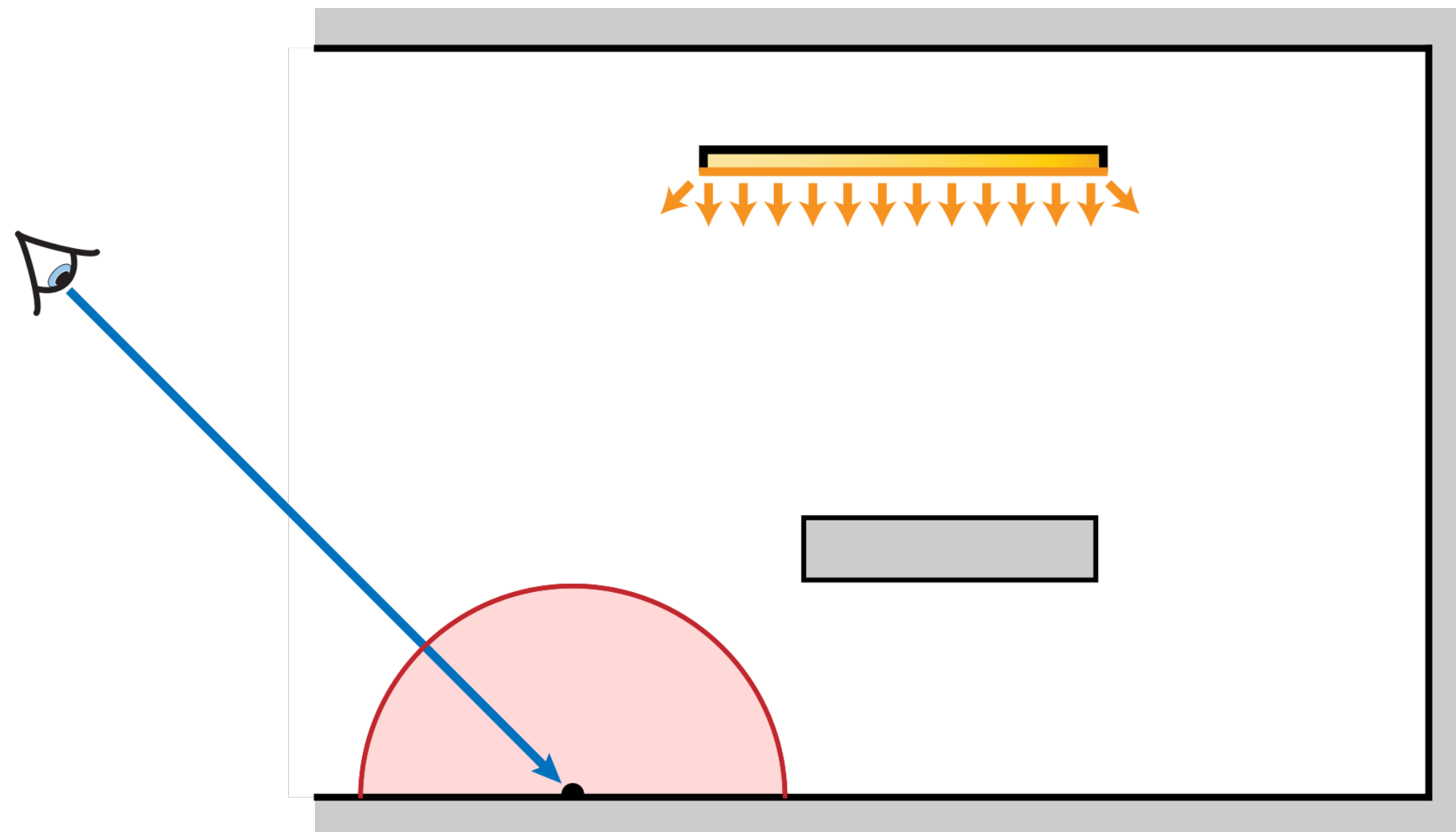
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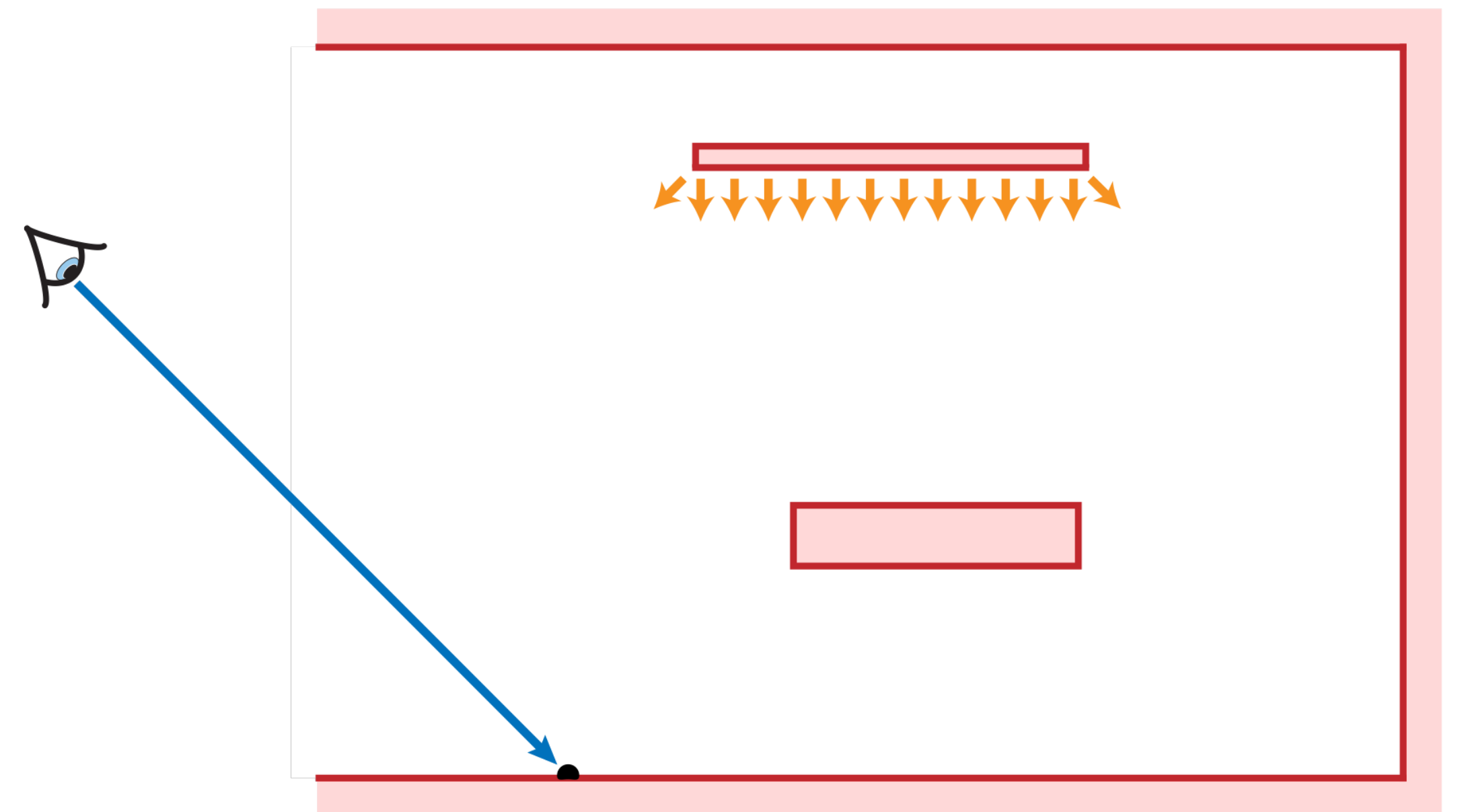
Sampling the area of the light

# Forms of Reflection Equation

Hemispherical  
integration



Surface Area  
integration



$$L_r(\mathbf{x}, \vec{\omega}_r) = \int_{H^2} f_r(\mathbf{x}, \vec{\omega}_i, \vec{\omega}_r) L_i(\mathbf{x}, \vec{\omega}_i) \cos \theta_i d\vec{\omega}_i$$

$$L_r(\mathbf{x}, \mathbf{z}) = \int_A f_r(\mathbf{x}, \mathbf{y}, \mathbf{z}) L_i(\mathbf{x}, \mathbf{y}) G(\mathbf{x}, \mathbf{y}) dA(\mathbf{y})$$

How do we decide which one to use for sampling direct illumination?

- The answer depends on the types of light sources in the scene.

# Light Sources

---

Point  
light



Spot  
light



Directional  
light



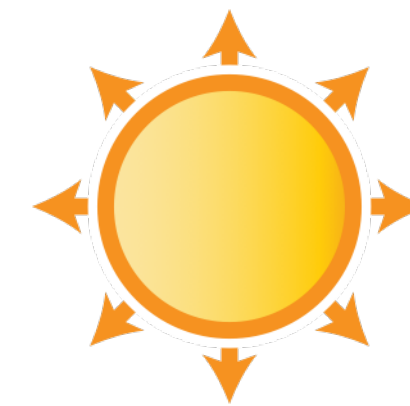
Environment  
light



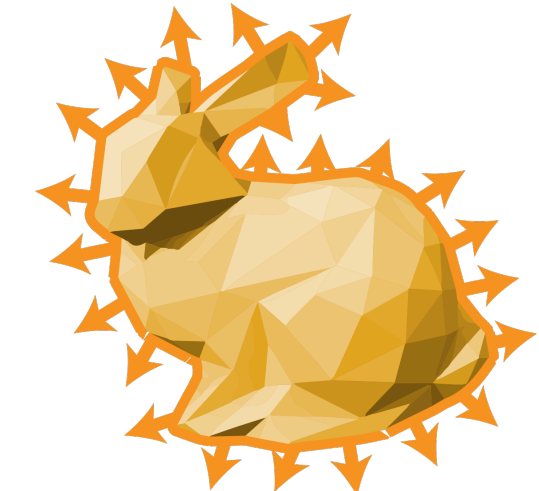
Quad  
light



Sphere  
light



Mesh  
light



Delta lights

(create hard shadows)

Finite lights

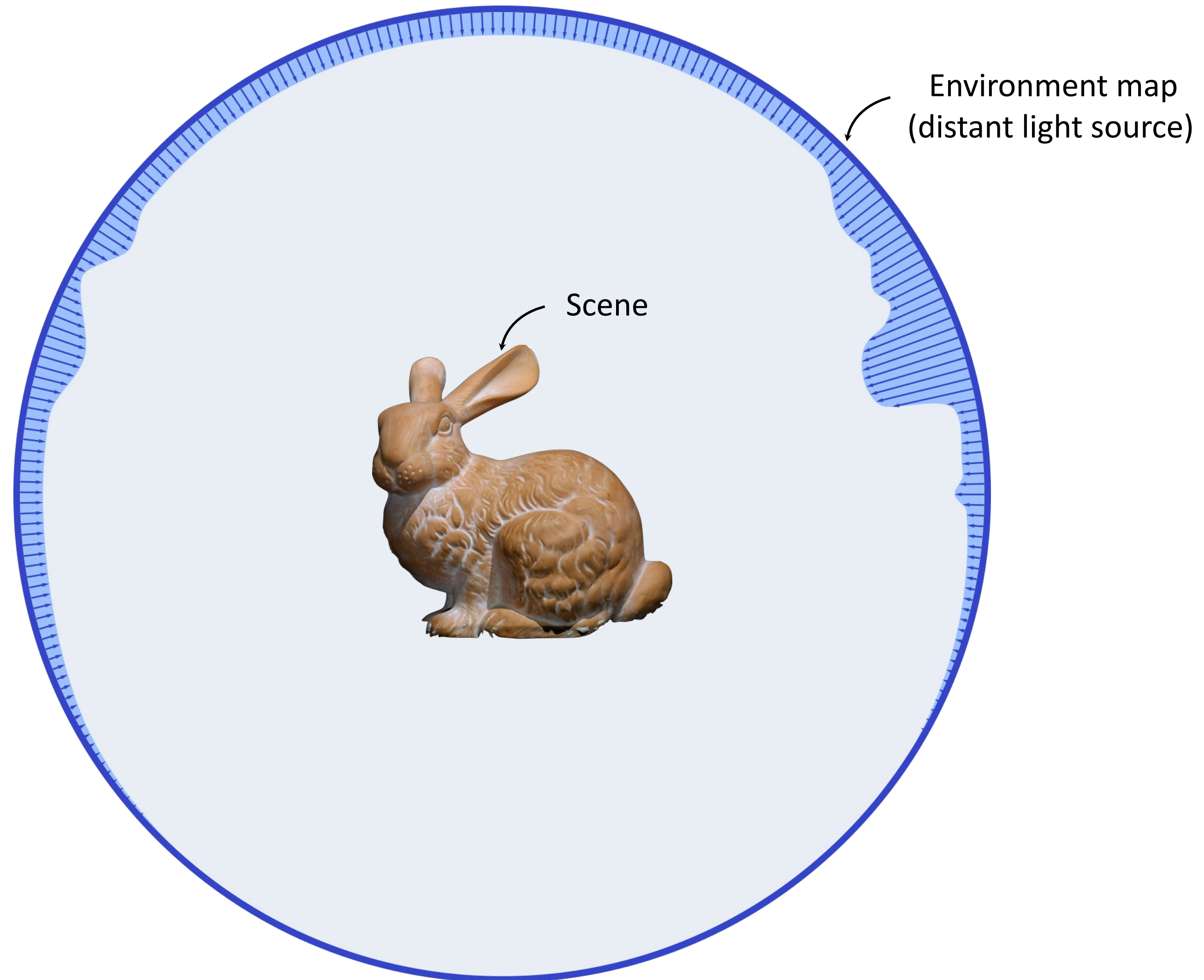
(create soft shadows)

# Environment Lighting



# Environment Lighting

---



# Environment Lighting

---

The image “wraps” around the virtual scene, serving as a *distant* source of illumination

Convenient to express using the *hemispherical* form of the reflectance equation

$$\begin{aligned} L_r(\mathbf{x}, \vec{\omega}_r) &= \int_{\Omega} f_r(\vec{\omega}_i, \vec{\omega}_r) L_i(\mathbf{x}, \vec{\omega}_i) \cos \theta_i d\vec{\omega}_i \\ &= \int_{\Omega} f_r(\vec{\omega}_i, \vec{\omega}_r) L_{\text{env}}(\vec{\omega}_i) V(\mathbf{x}, \vec{\omega}_i) \cos \theta_i d\vec{\omega}_i \end{aligned}$$



# Environment Lighting

---



# Environment Lighting

---



$$L_r(\mathbf{x}, \vec{\omega}_r) = \int_{\Omega} f_r(\vec{\omega}_i, \vec{\omega}_r) L_{\text{env}}(\vec{\omega}_i) V(\mathbf{x}, \vec{\omega}_i) \cos \theta_i d\vec{\omega}_i$$

# Importance Sampling $L_{\text{env}}$

---



Sample using the *hemispherical form* of the reflectance equation and pdf

$$p(\vec{\omega}_i) \propto L_{\text{env}}(\vec{\omega}_i)$$

# Importance Sampling $L_{\text{env}}$

---

$$p(\vec{\omega}_i) \propto L_{\text{env}}(\vec{\omega}_i)$$

Several strategies exist

We'll discuss:

- Marginal/Conditional CDF method
- Hierarchical warping method

# Importance Sampling

---

## Recipe:

1. Express the desired distribution in a convenient coordinate system
  - requires computing the Jacobian
2. Compute marginal and conditional 1D PDFs
3. Sample 1D PDFs using the inversion method

# Marginal/Conditional CDF

---

Assume the lat/long parameterization

Draw samples from joint  $p(\theta, \phi) \propto L_{\text{env}}(\theta, \phi) \sin \theta$



# Why the Sine?

---

General case of integrating some  $f(\vec{\omega})$  over  $S^2$

If we set  $d\vec{\omega} = \sin \theta d\theta d\phi$  we want to cancel out the sine.

↖ Comes from the Jacobian

$$\begin{aligned}\int_{S^2} f(\vec{\omega}) d\vec{\omega} &= \int_0^{2\pi} \int_0^\pi f(\theta, \phi) \sin \theta d\theta d\phi \\ &\approx \frac{1}{N} \sum_{i=1}^N \frac{f(\theta_i, \phi_i) \sin \theta_i}{p(\theta_i, \phi_i)}\end{aligned}$$

$$p(\theta, \phi) \propto f(\theta, \phi) \sin \theta$$

# Marginal/Conditional CDF

---

Assume the lat/long parameterization

Draw samples from joint  $p(\theta, \phi) \propto L_{\text{env}}(\theta, \phi) \sin \theta$

- Step 1: create scalar version  $L'(\theta, \phi)$  of  $L_{\text{env}}(\theta, \phi) \sin \theta$

- Step 2: compute marginal PDF

$$p(\theta) = \int_0^{2\pi} L'(\theta, \phi) d\phi$$

- Step 3: compute conditional PDF

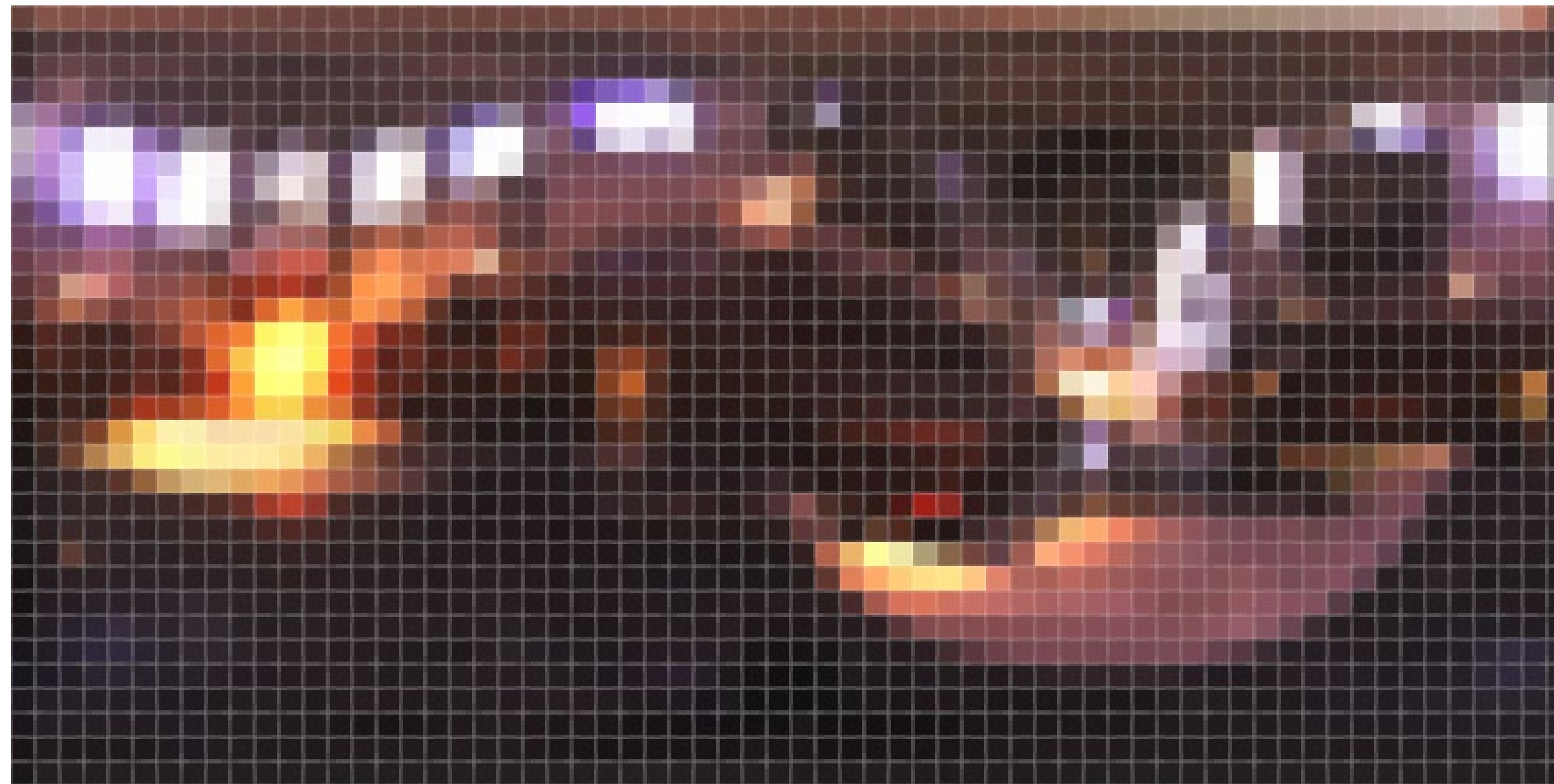
$$p(\phi|\theta) = \frac{p(\theta, \phi)}{p(\theta)}$$

- Step 4: draw samples  $\theta_i \sim p(\theta)$  and  $\phi_i \sim p(\phi|\theta)$

# Step 1: Scalar Importance Func.

---

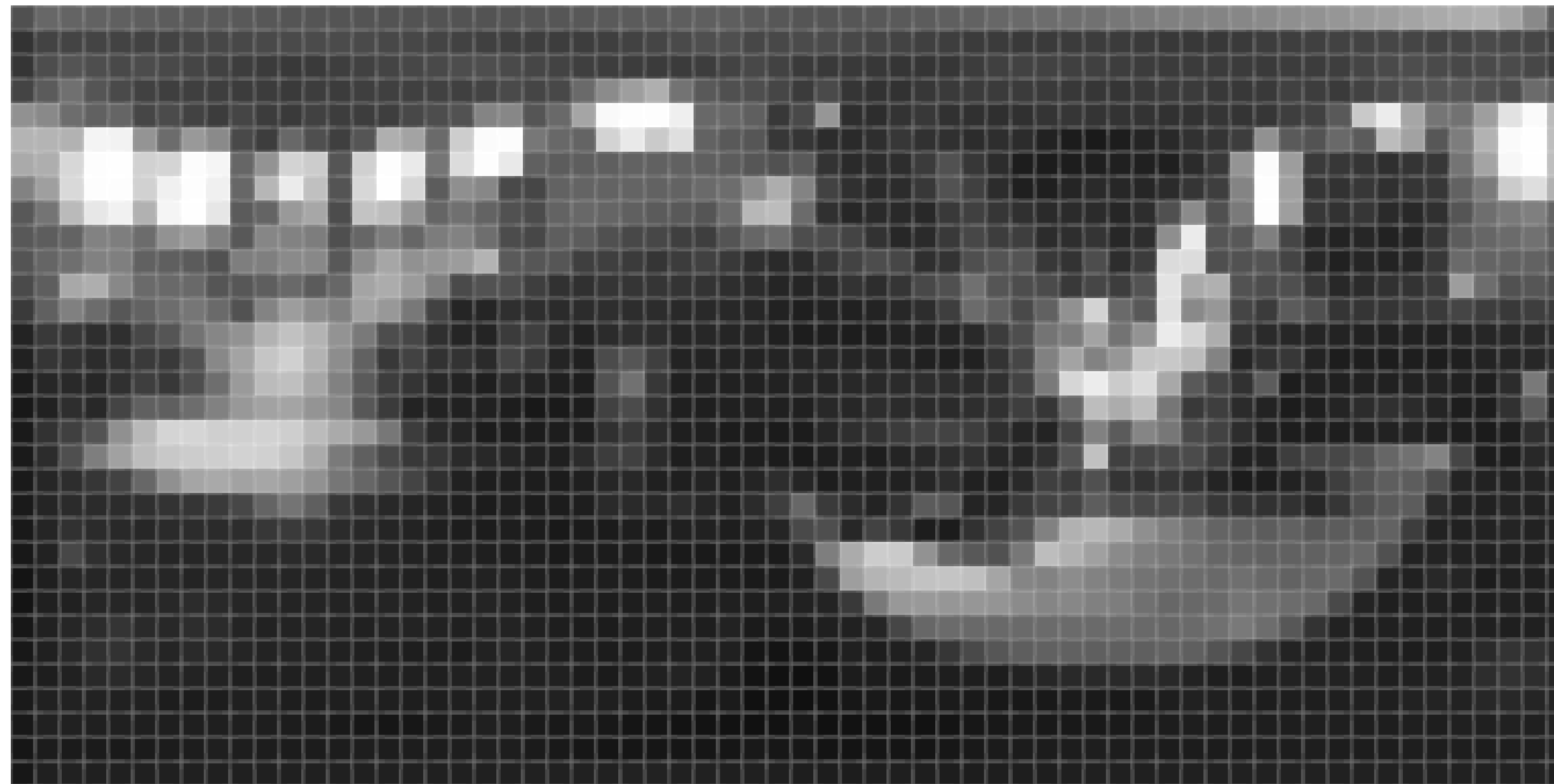
Original environment map



# Step 1: Scalar Importance Func.

---

Scalar version  
(average, max, or luminance of RGB channels)



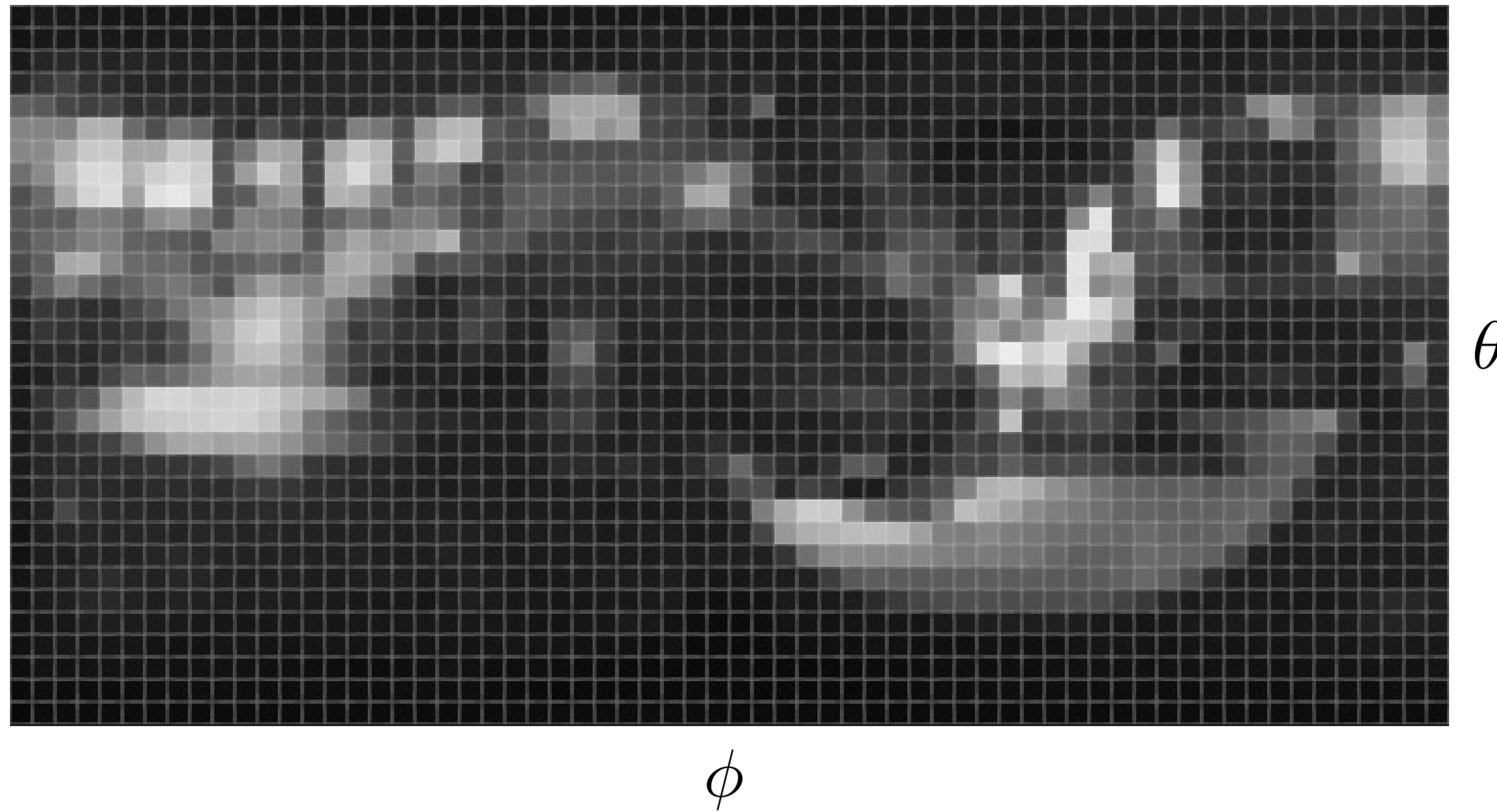
$\theta$

$\phi$

# Step 1: Scalar Importance Func.

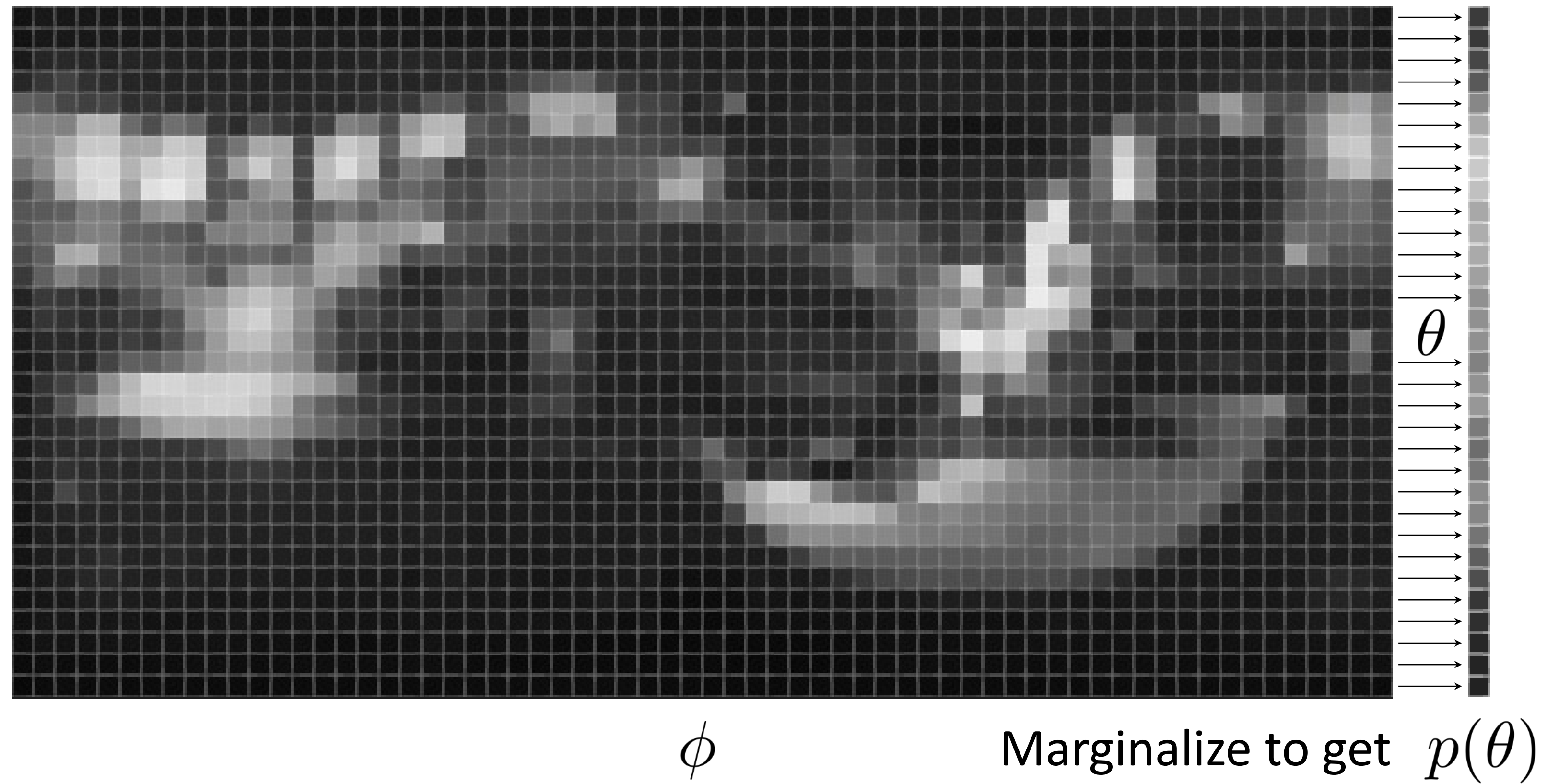
---

Multiplied by  $\sin \theta$



# Step 2: Marginalization

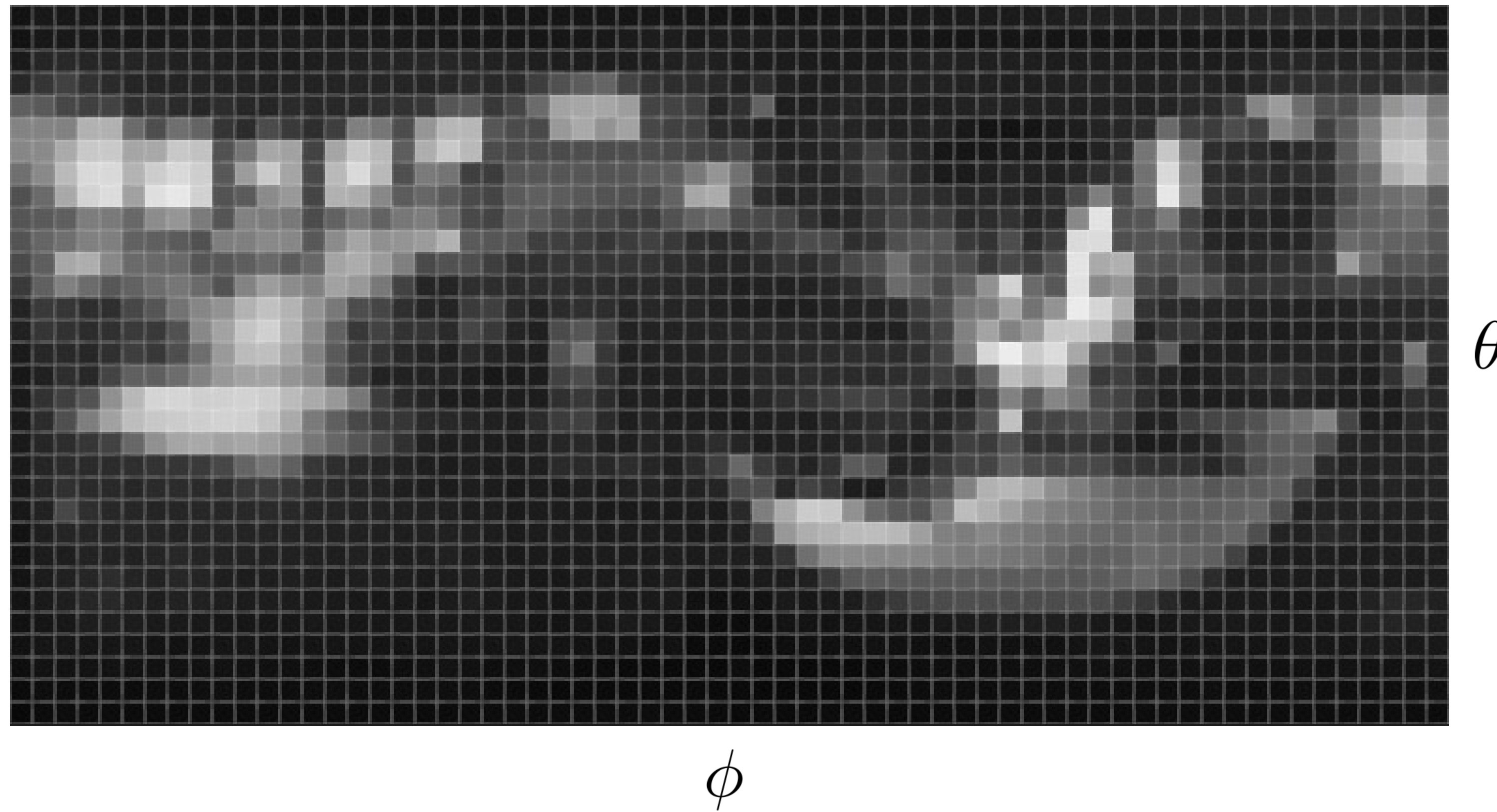
---



# Step 3: Conditional PDFs

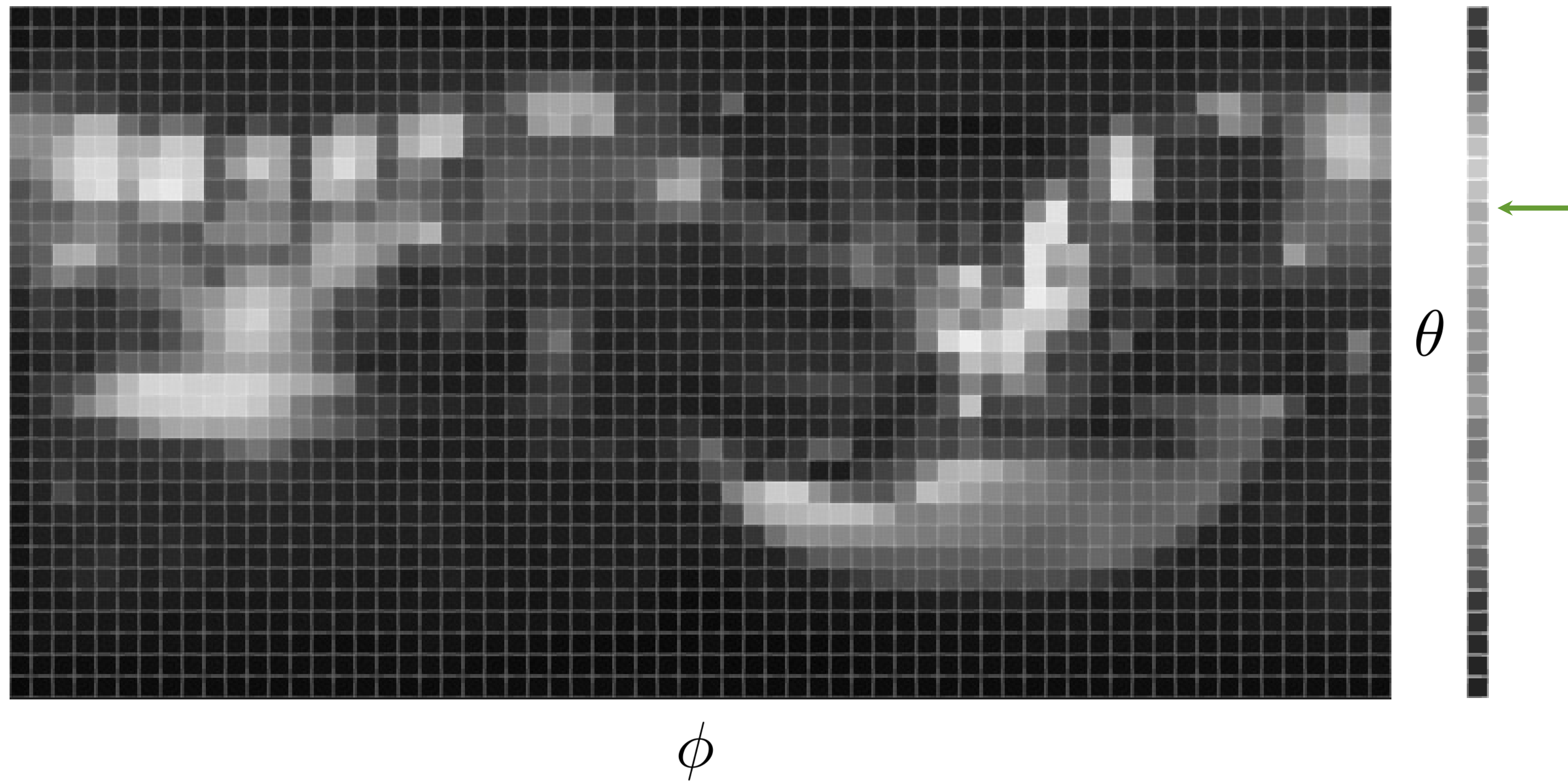
---

Once normalized, each row can serve as  
the conditional PDF



# Step 4: Sampling

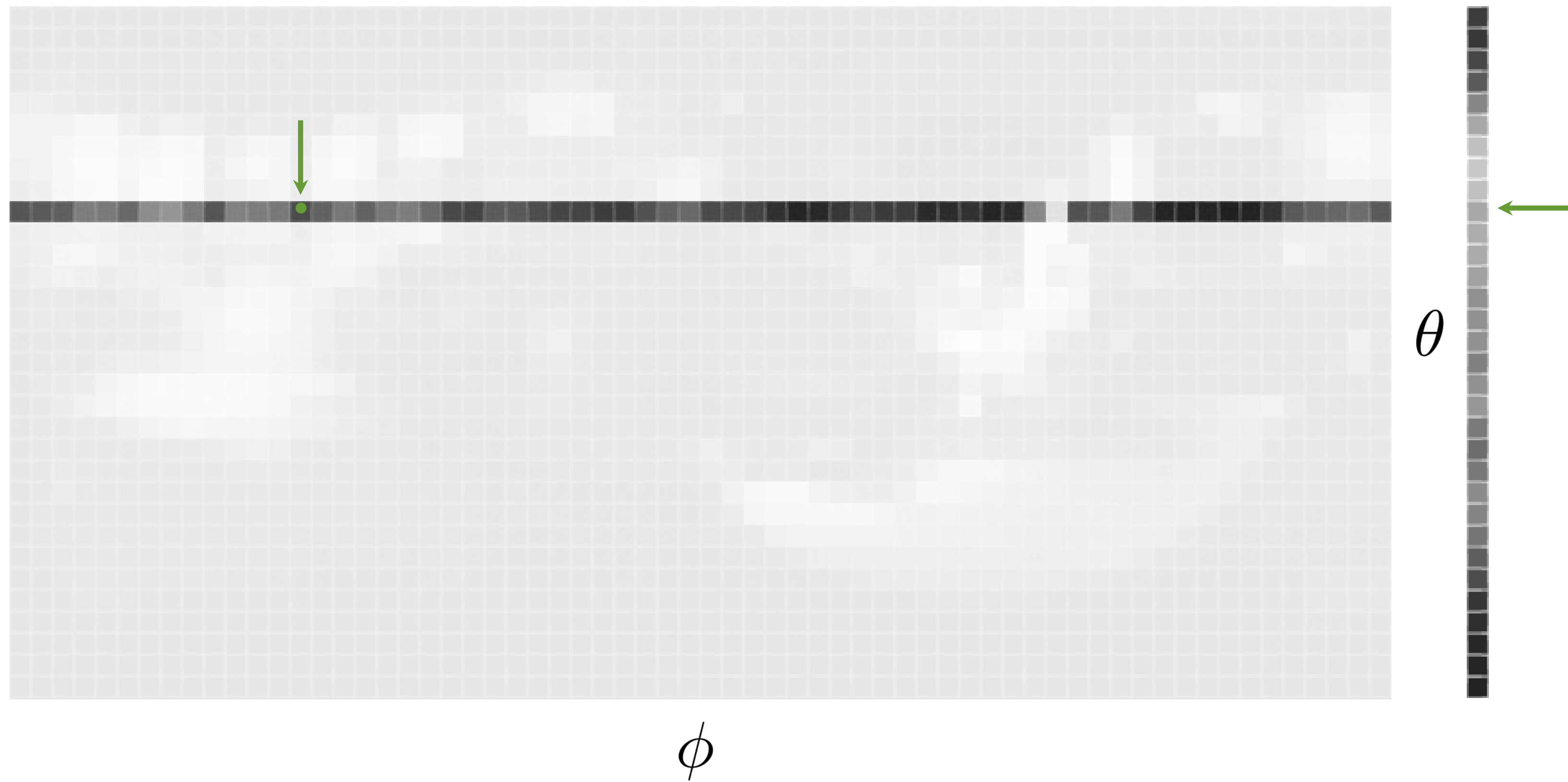
---





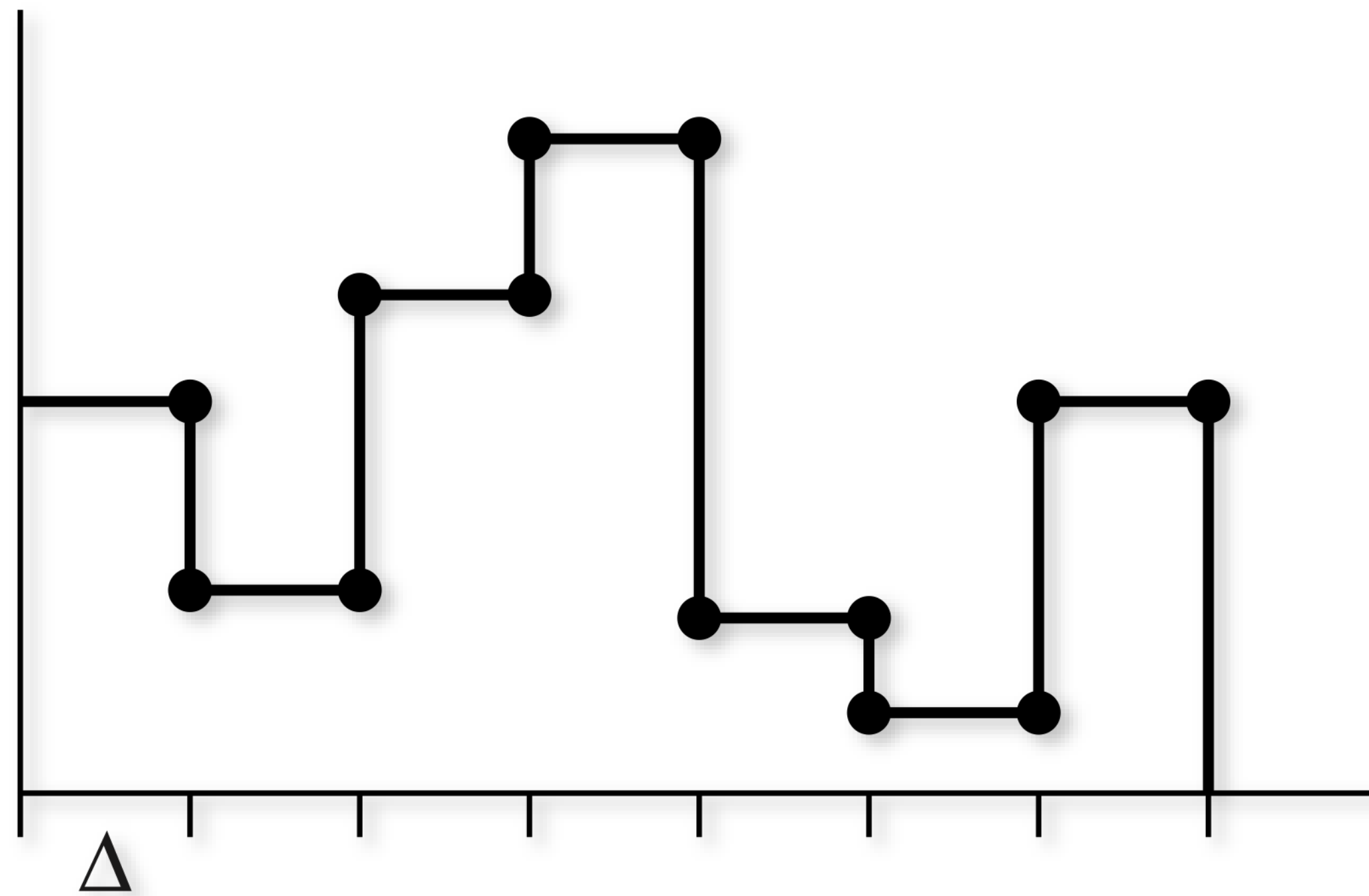
# Step 4: Sampling

---

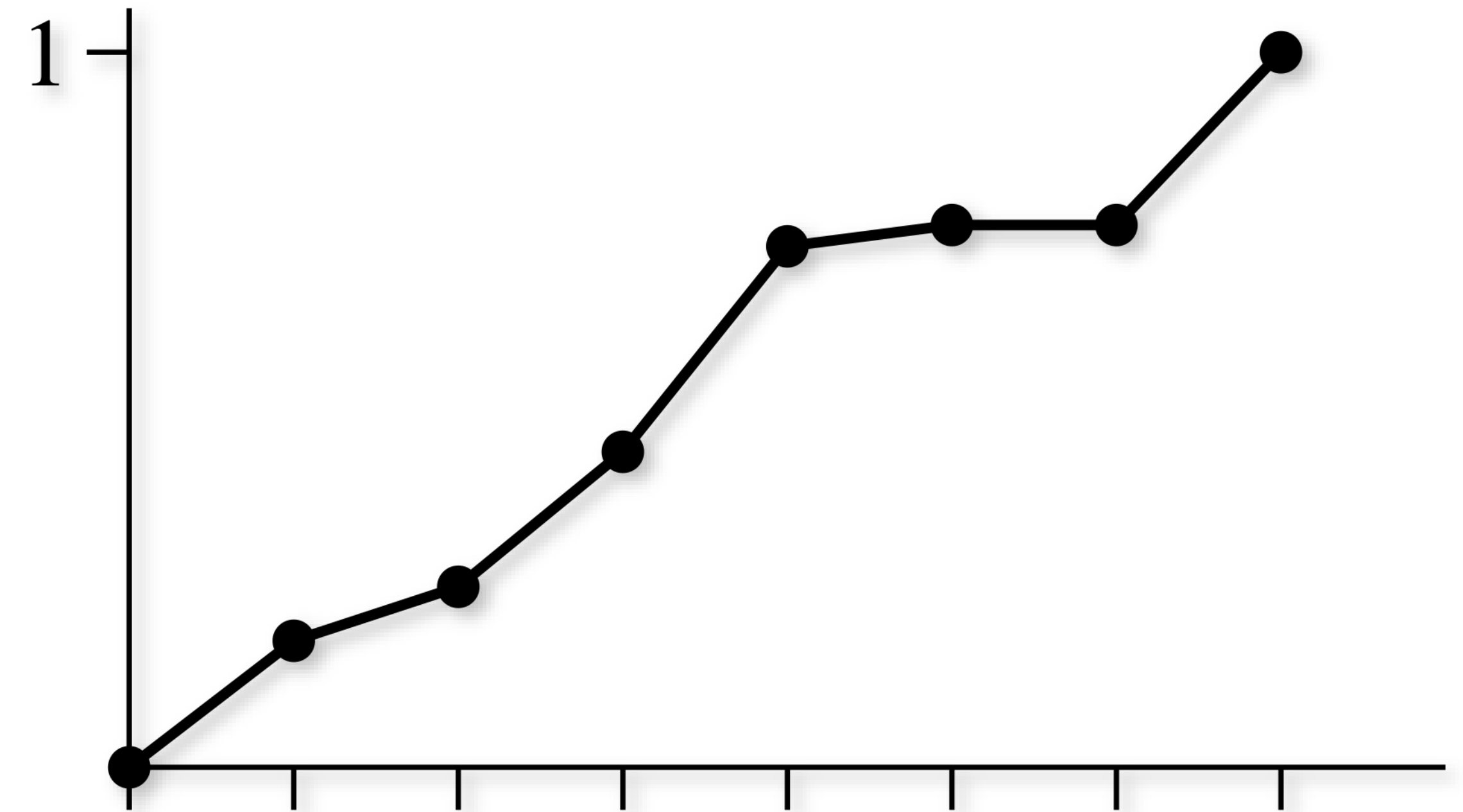


# Sampling Discrete 1D PDFs

---



PDF



CDF

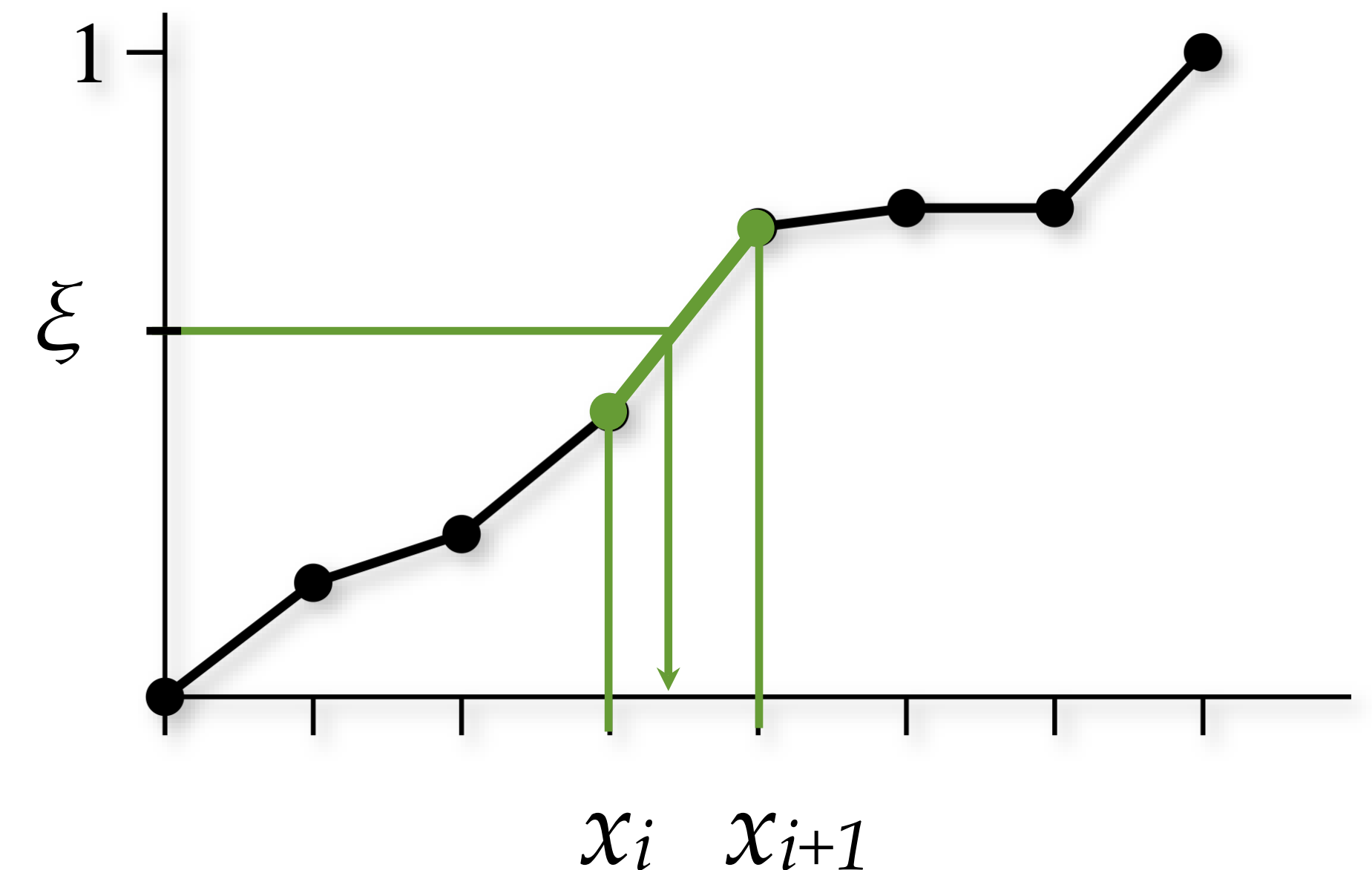
# Sampling Discrete 1D PDFs

---

Given a uniform random value  $\xi$

Find  $x_i$  and  $x_{i+1}$  using binary search

Linearly interpolate to find  $x$



# C++ details

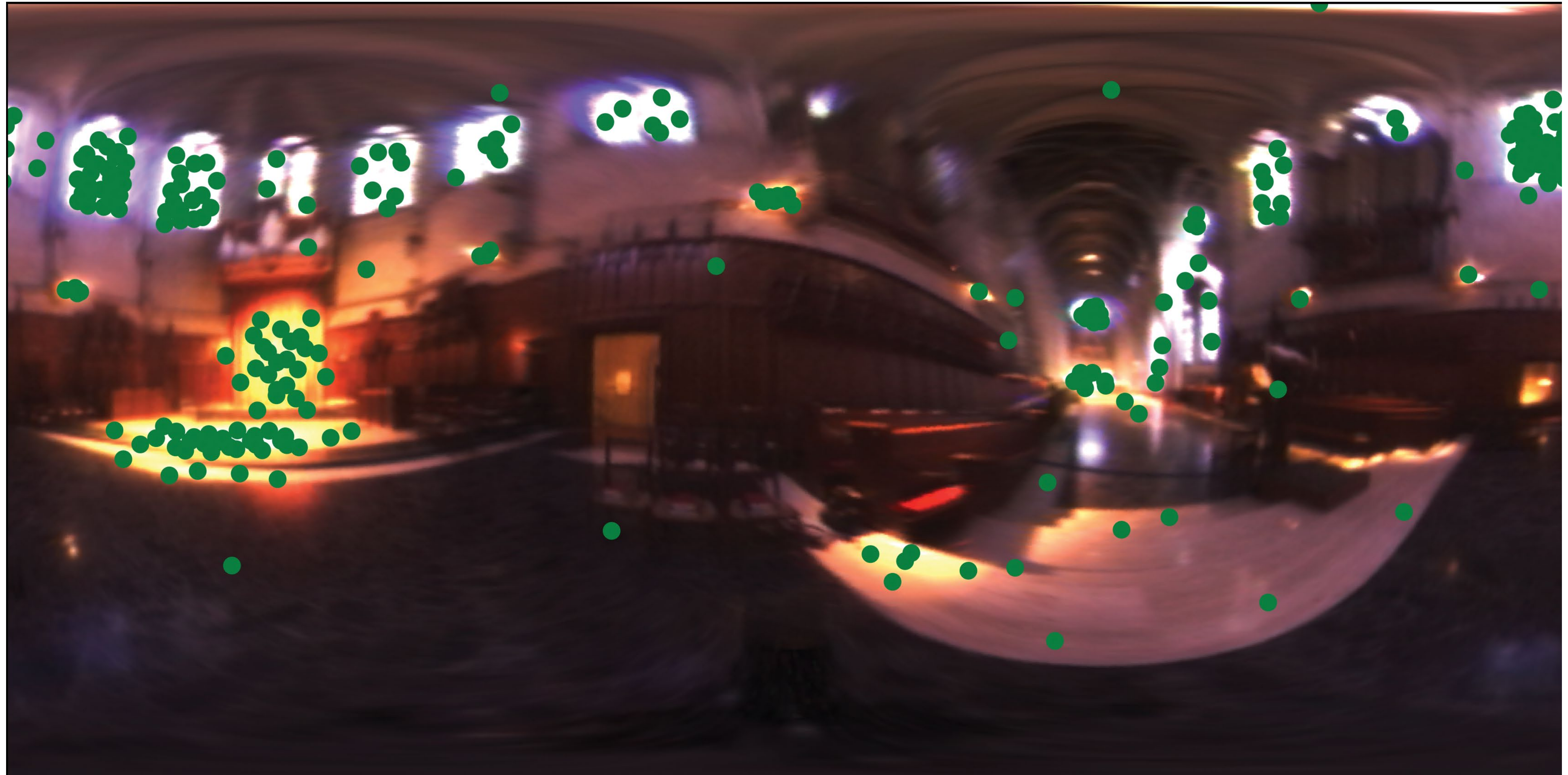
---

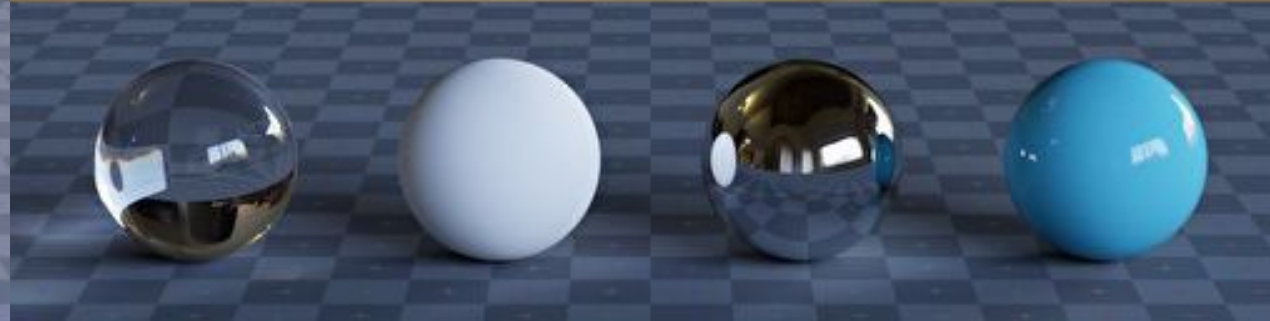
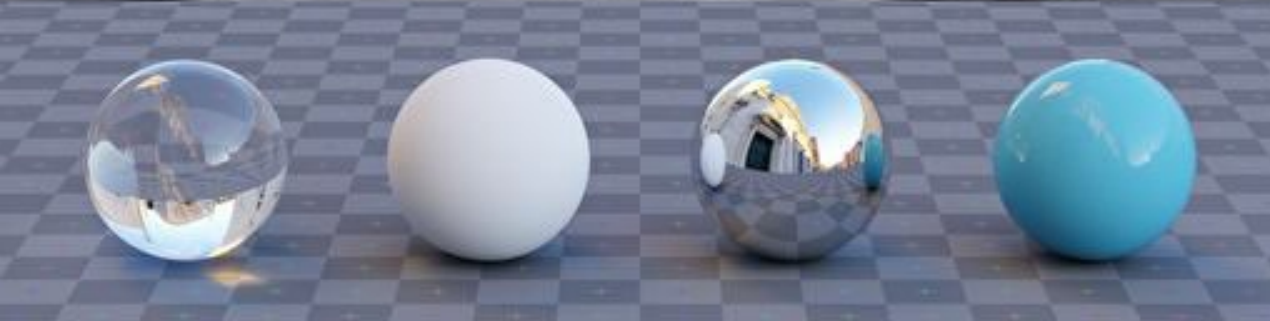
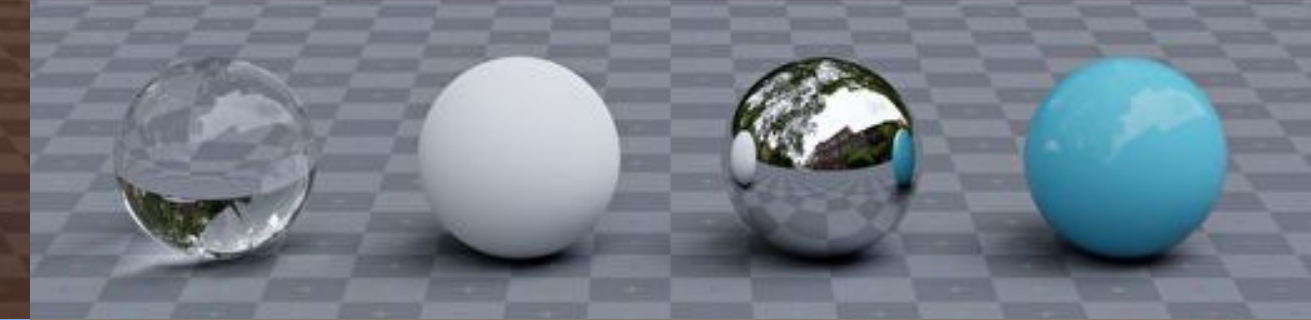
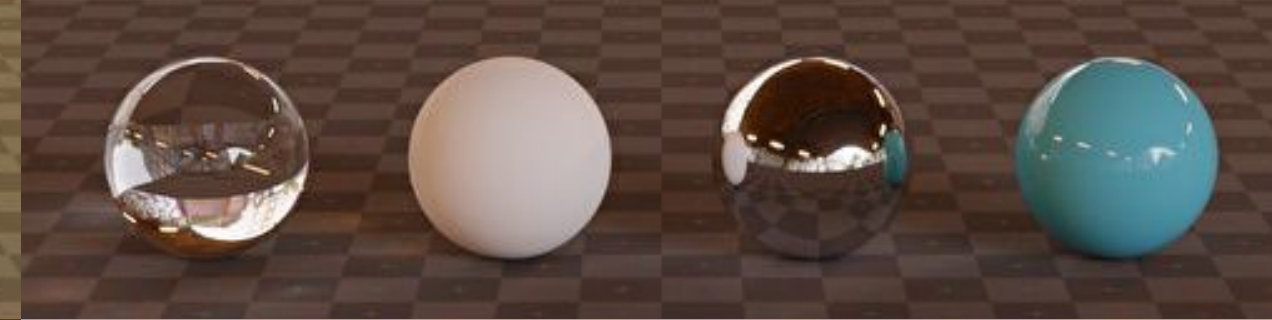
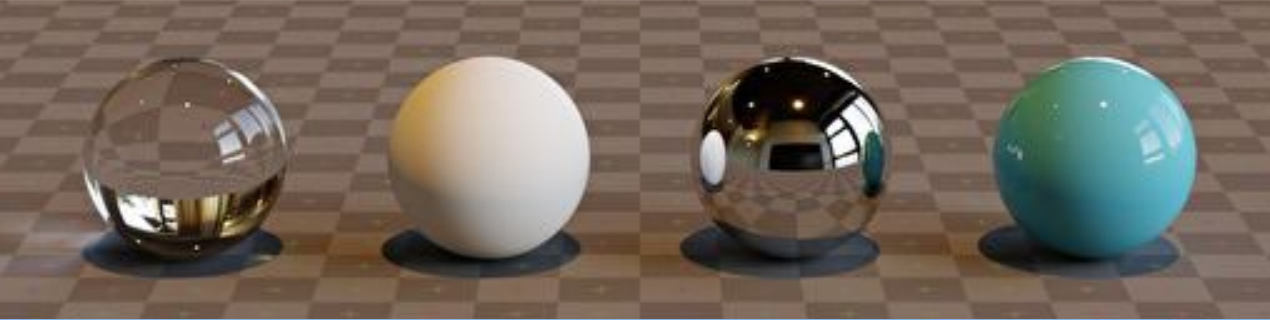
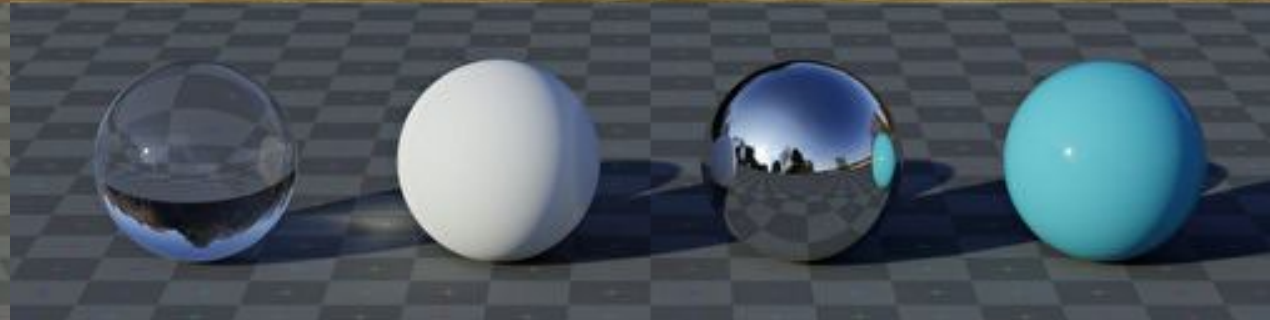
Don't need to implement binary search yourself!

- Given sorted list, use `std::lower_bound(...)`
- See implementation in PBRT

# Resulting Sample Distribution

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# Light Sources

# Light Sources

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Point  
light



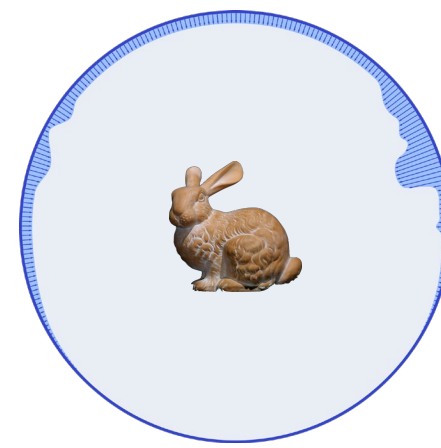
Spot  
light



Directional  
light



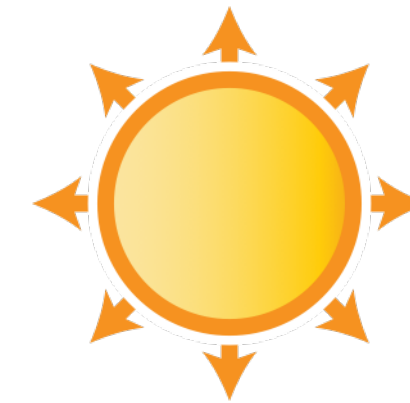
Environment  
light



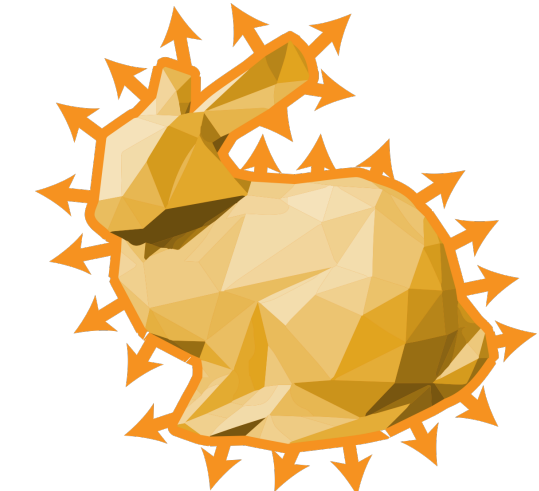
Quad  
light



Sphere  
light



Mesh  
light



Delta lights

(create hard shadows)

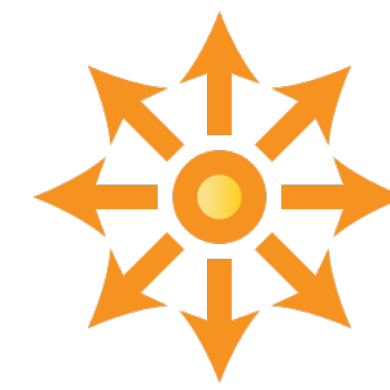
Finite lights

(create soft shadows)



# Point Light

---

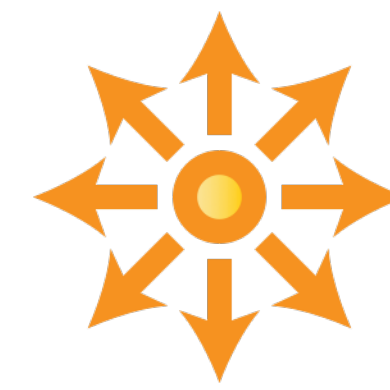


Omnidirectional emission from a single point

Typically defined using a point  $\mathbf{p}$  and emitted power  $\Phi$

- delta function with respect to which form of the reflection equation?

# Point Light



Omnidirectional emission from a single point

Typically defined using a point  $\mathbf{p}$  and emitted power  $\Phi$

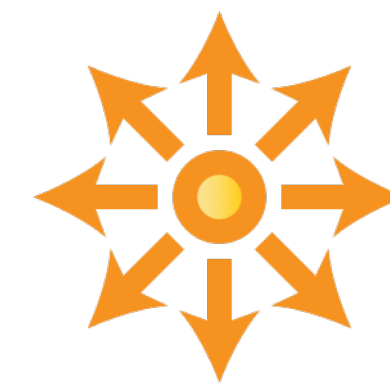
- delta function with respect to surface integral form of the reflection equation

$$L_r(\mathbf{x}, \mathbf{z}) = \int_{A_e} f_r(\mathbf{x}, \mathbf{y}, \mathbf{z}) L_e(\mathbf{y}, \mathbf{x}) V(\mathbf{x}, \mathbf{y}) \frac{|\cos \theta_i| |\cos \theta_o|}{\|\mathbf{x} - \mathbf{y}\|^2} dA(\mathbf{y})$$

$$L_e(\mathbf{y}, \mathbf{x}) = \frac{\Phi}{4\pi} \delta(\mathbf{y} - \mathbf{p})$$

$$L_r(\mathbf{x}, \mathbf{z}) = \frac{\Phi}{4\pi} f_r(\mathbf{x}, \mathbf{p}, \mathbf{z}) V(\mathbf{x}, \mathbf{p}) \frac{|\cos \theta_i|}{\|\mathbf{x} - \mathbf{p}\|^2}$$

# Point Light



Omnidirectional emission from a single point

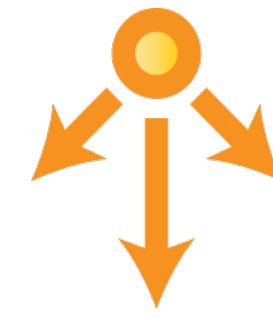
Typically defined using a point  $\mathbf{p}$  and emitted power  $\Phi$

- delta function with respect to surface integral form of the reflection equation

$$L_r(\mathbf{x}, \mathbf{z}) = \frac{\Phi}{4\pi} f_r(\mathbf{x}, \mathbf{p}, \mathbf{z}) V(\mathbf{x}, \mathbf{p}) \frac{|\cos \theta_i|}{\|\mathbf{x} - \mathbf{p}\|^2}$$

# Spot Light?

---



**Directionally dependent** emission from a single point

Typically defined using a point  $\mathbf{p}$  and ...

$$L_r(\mathbf{x}, \mathbf{z}) = \frac{\Phi}{4\pi} f_r(\mathbf{x}, \mathbf{p}, \mathbf{z}) V(\mathbf{x}, \mathbf{p}) \frac{|\cos \theta_i|}{\|\mathbf{x} - \mathbf{p}\|^2}$$

# Spot Light

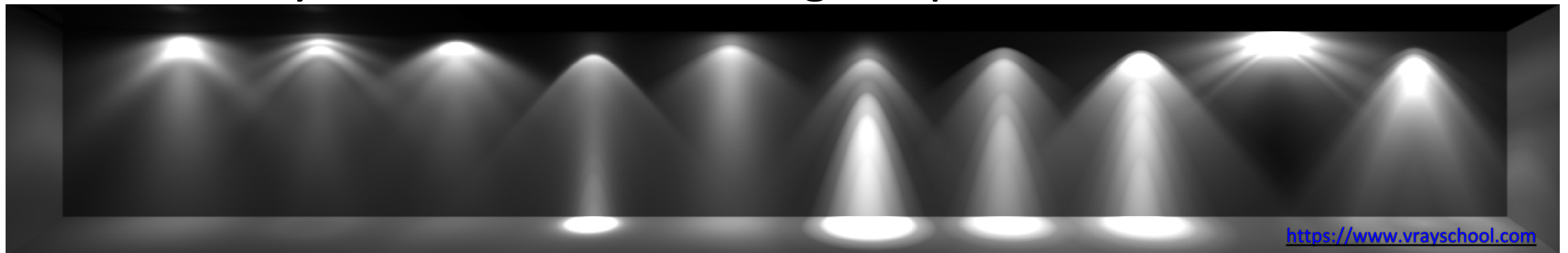


**Directionally dependent** emission from a single point

Typically defined using a point  $\mathbf{p}$  and a directionally dependent radiant intensity function  $I$

$$L_r(\mathbf{x}, \mathbf{z}) = I(\mathbf{p}, \mathbf{x}) f_r(\mathbf{x}, \mathbf{p}, \mathbf{z}) V(\mathbf{x}, \mathbf{p}) \frac{|\cos \theta_i|}{\|\mathbf{x} - \mathbf{p}\|^2}$$

The intensity can be defined using IES profiles:



# Directional Light

---



Far-away emission from single direction (delta environment map)

Typically defined using a direction  $\vec{\omega}_d$  and radiance  $L_d$

- delta function with respect to which form of the reflection equation?

# Directional Light



Far-away emission from single direction (delta environment map)

Typically defined using a direction  $\vec{\omega}_d$  and radiance  $L_d$

- delta function with respect to hemispherical integral form of the reflection equation

$$L_r(\mathbf{x}, \vec{\omega}_r) = \int_e \int_{H^2} f_r(\mathbf{x}, \vec{\omega}_i, \vec{\omega}_r) L_e(r(\mathbf{x}, \vec{\omega}_i), -\vec{\omega}_i) \cos \theta_i d\vec{\omega}_i$$

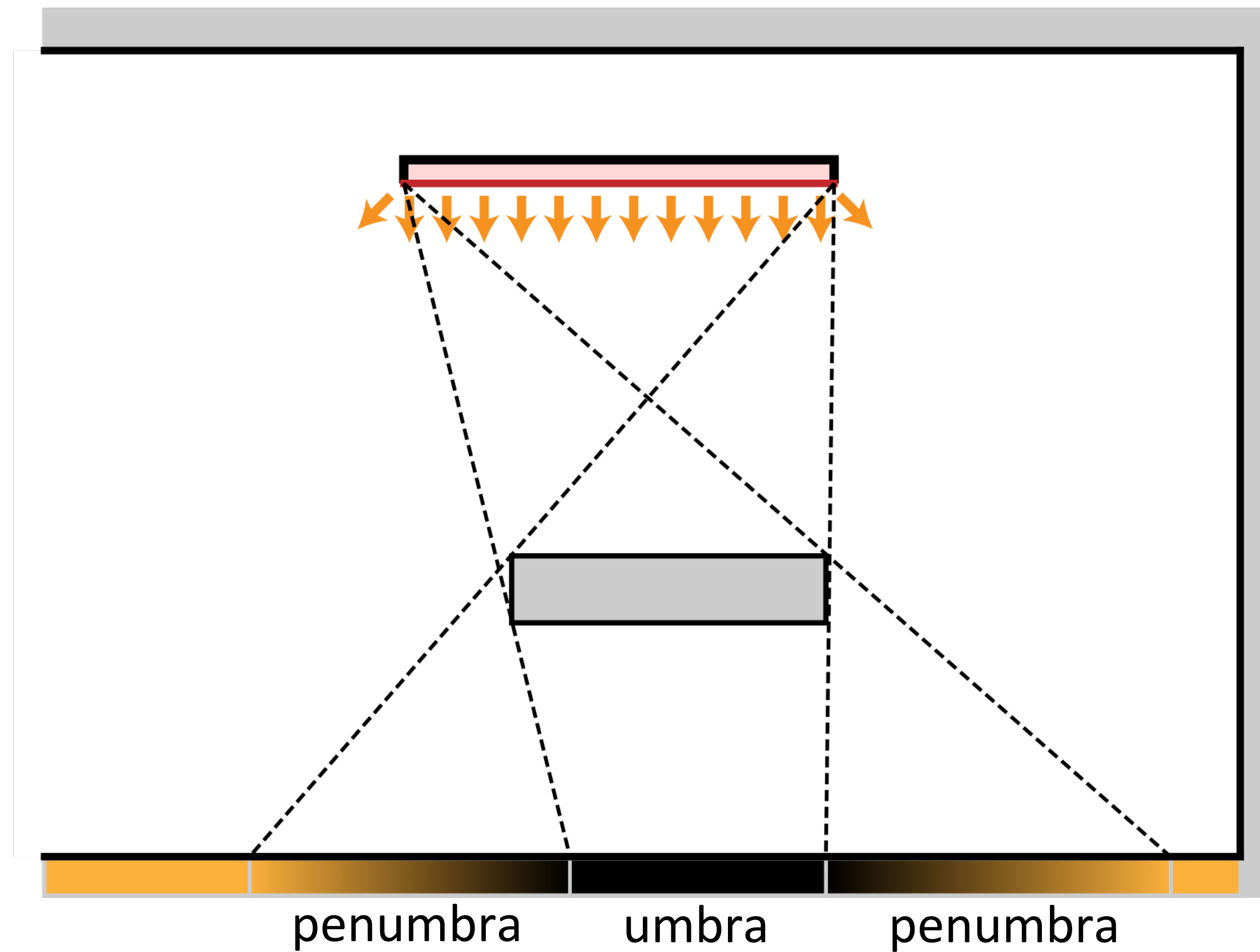
$$L_e(\mathbf{y}, \vec{\omega}) = V(\mathbf{y}, \vec{\omega}_d) L_d \delta(\vec{\omega}_d - \vec{\omega})$$

$$L(\mathbf{x}, \vec{\omega}_r) = f_r(\mathbf{x}, \vec{\omega}_d, \vec{\omega}_r) V(\mathbf{x}, \vec{\omega}_d) L_d \cos \theta_d$$

# Quad Light



Has finite area... creates soft shadows

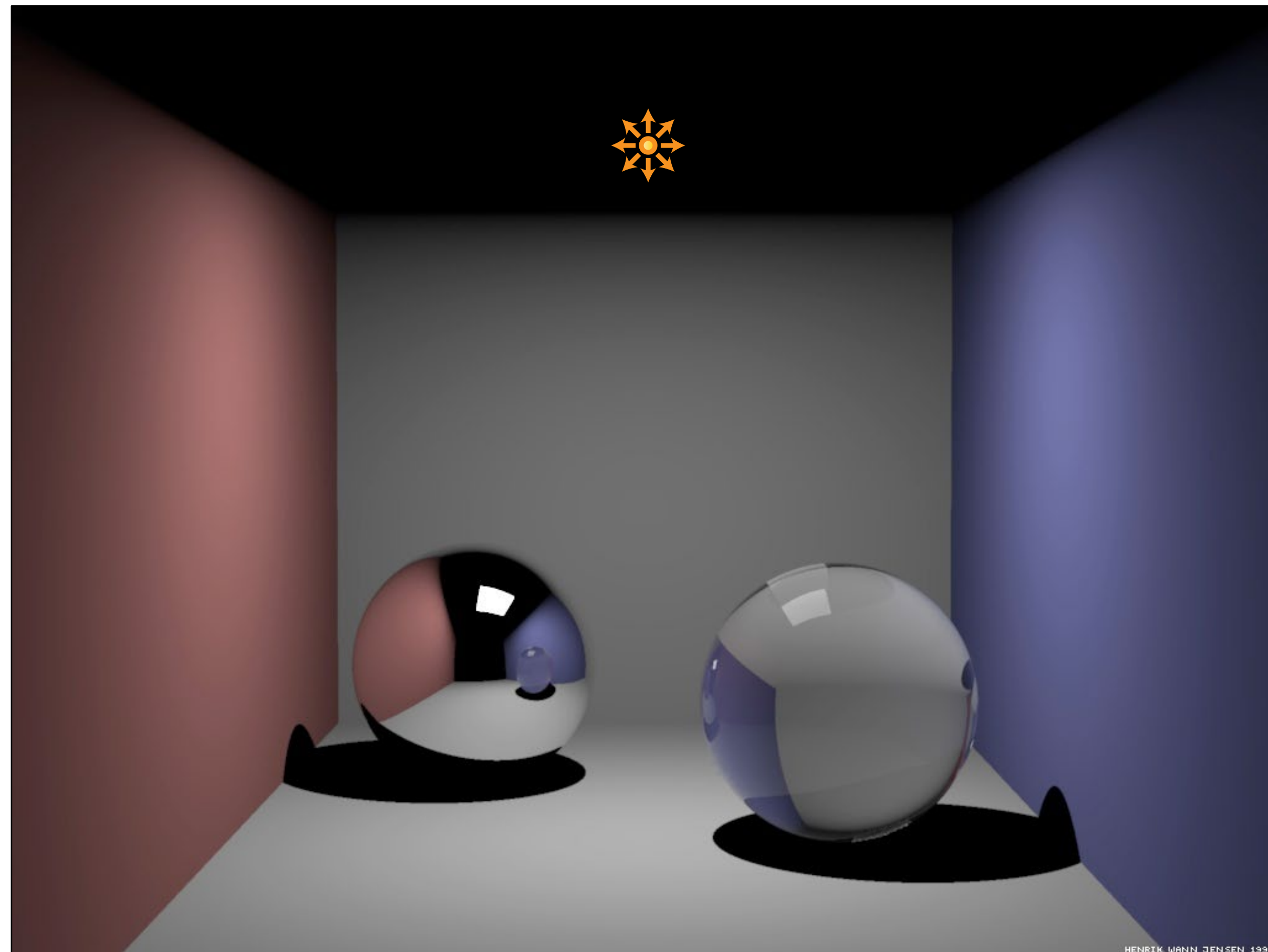




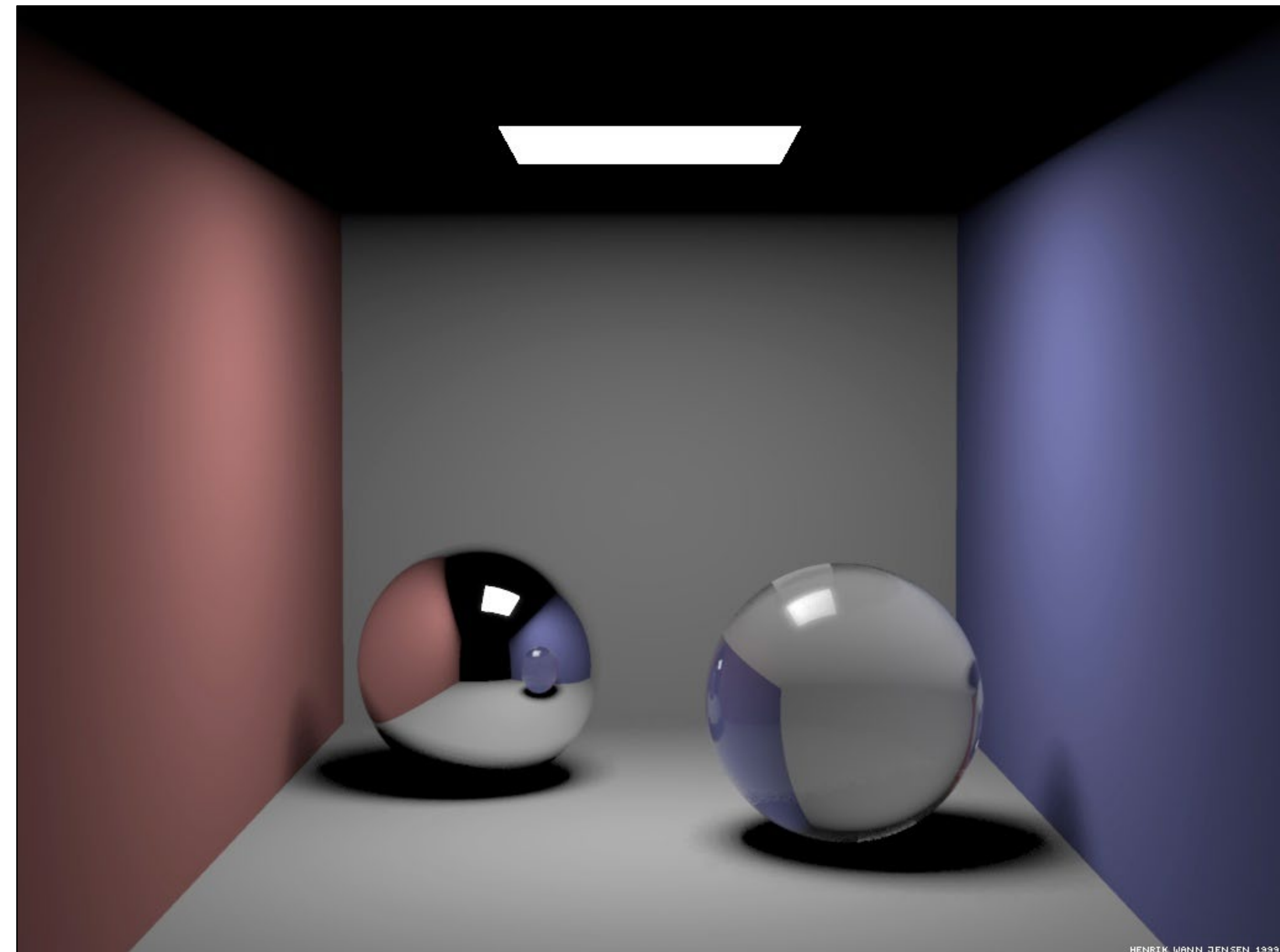
# Quad Light



Point light

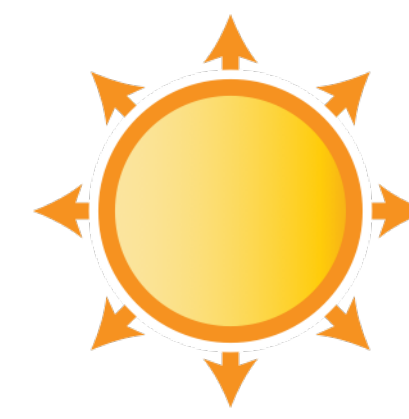


Quad light



# Sphere Light

---

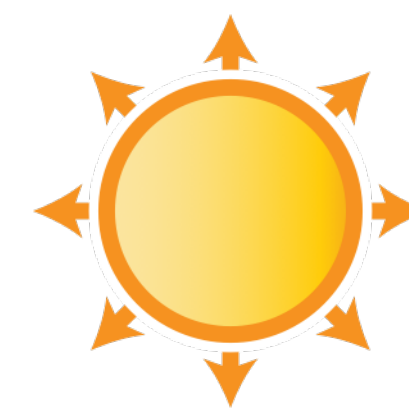


Typically defined using a center  $\mathbf{p}$ , radius  $r$ , and emitted power  $\Phi$  (or emitted radiance  $L_e$ )

Has finite surface area  $4\pi r^2$

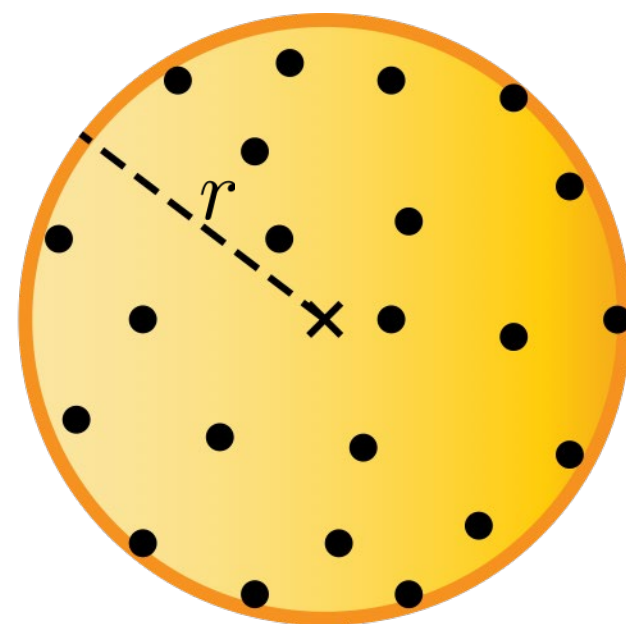
# Sphere Light

---

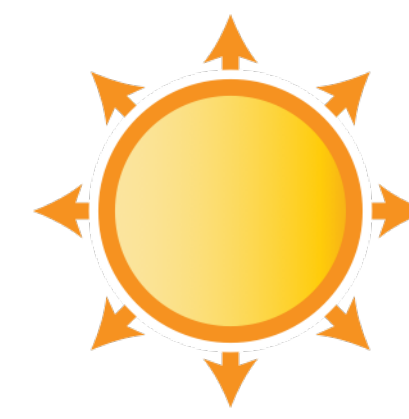


How to sample points on the sphere light?

**Approach 1:** uniformly sample *sphere area*

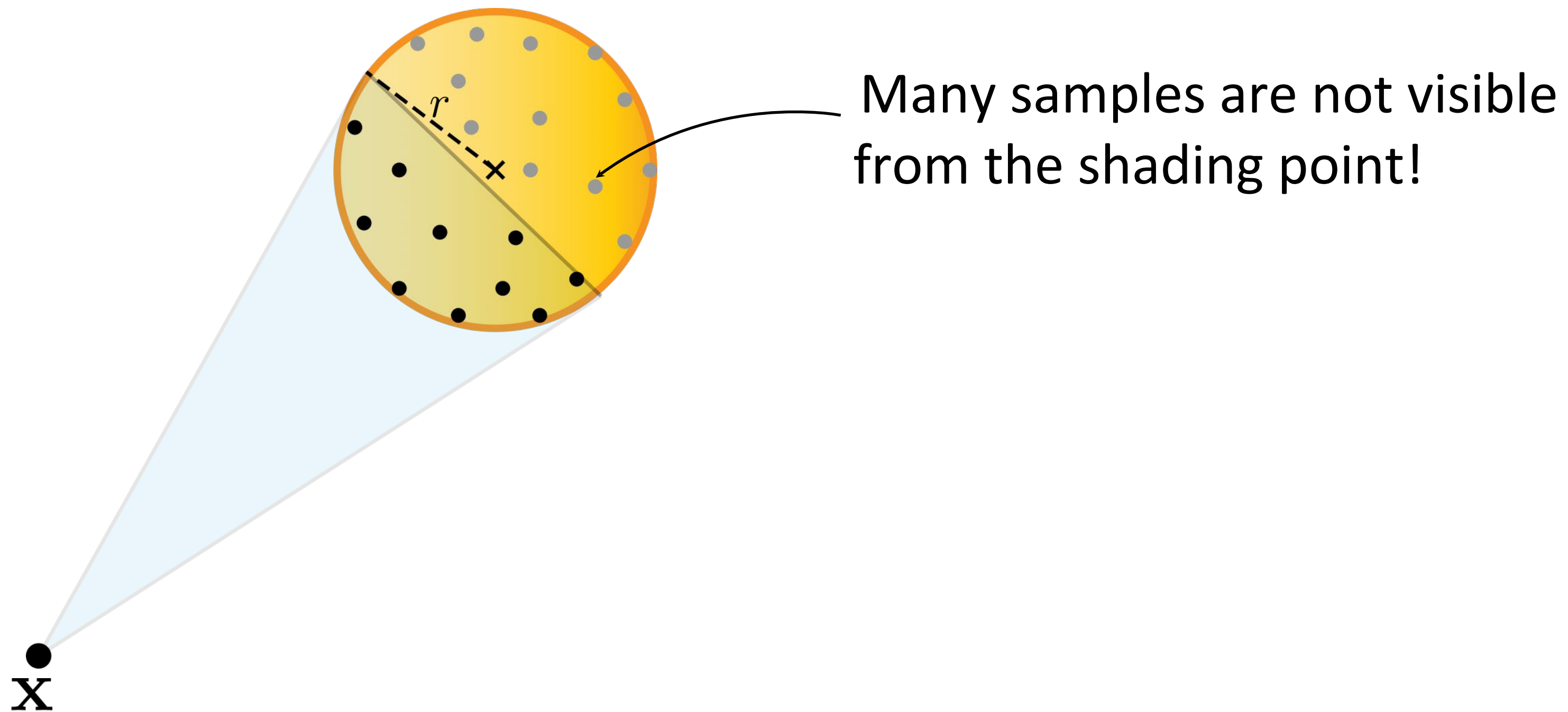


# Sphere Light

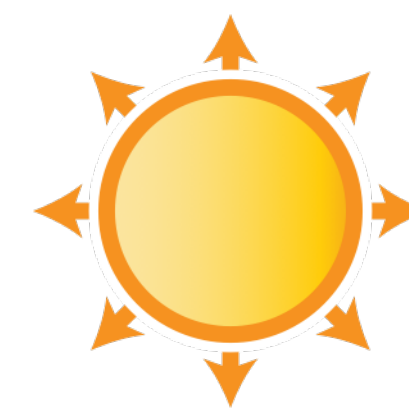


How to sample points on the sphere light?

**Approach 1:** uniformly sample *sphere area*

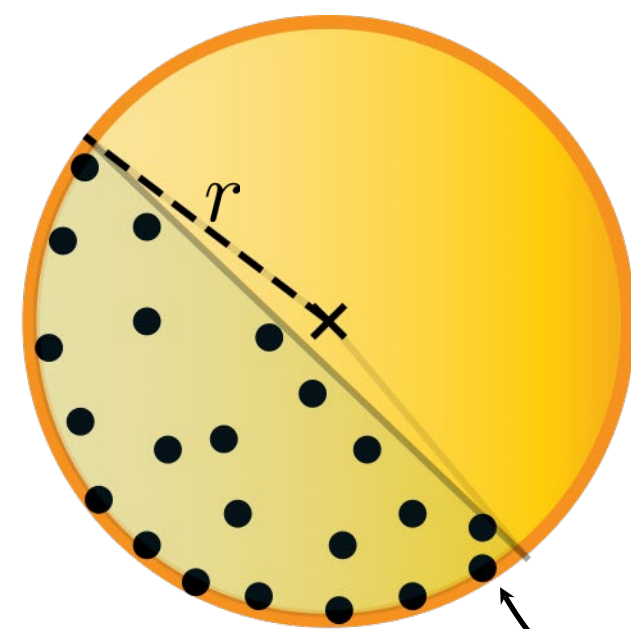


# Sphere Light



How to sample points on the sphere light?

**Approach 2** (better): uniformly sample area of the *visible spherical cap*

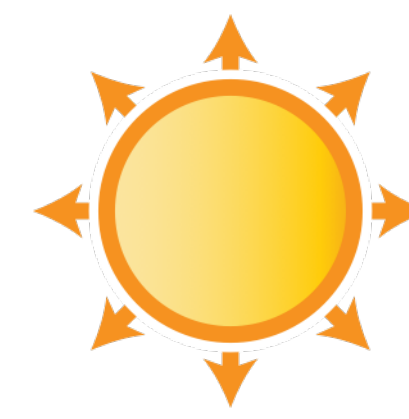


spherical cap on light *area*



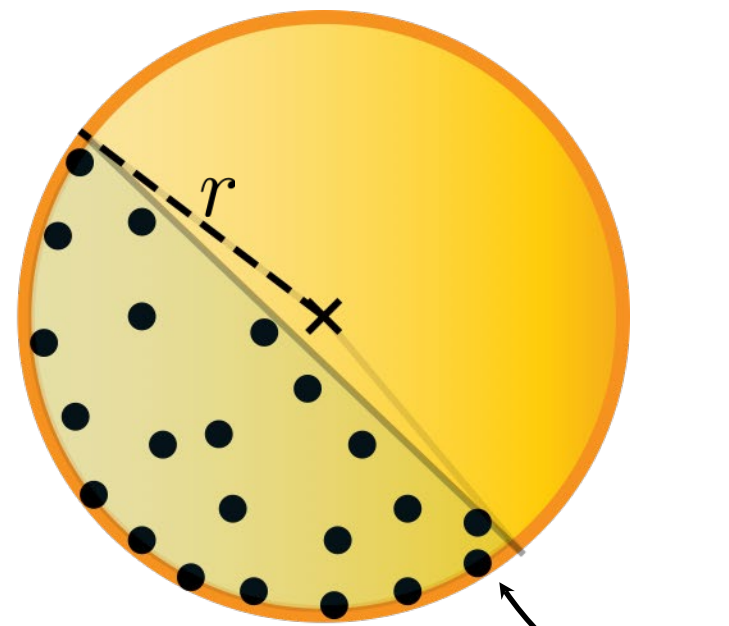
Can sample a spherical cap using Hat-Box theorem!

# Sphere Light



How to sample points on the sphere light?

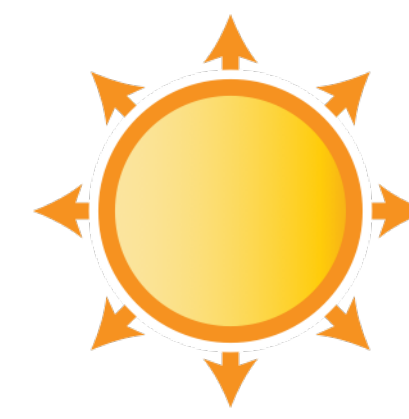
**Approach 2** (better): uniformly sample area of the *visible spherical cap*



Uniform area-density is not ideal as emitted radiance is weighted by the cosine term (recall the form factor in the G term)

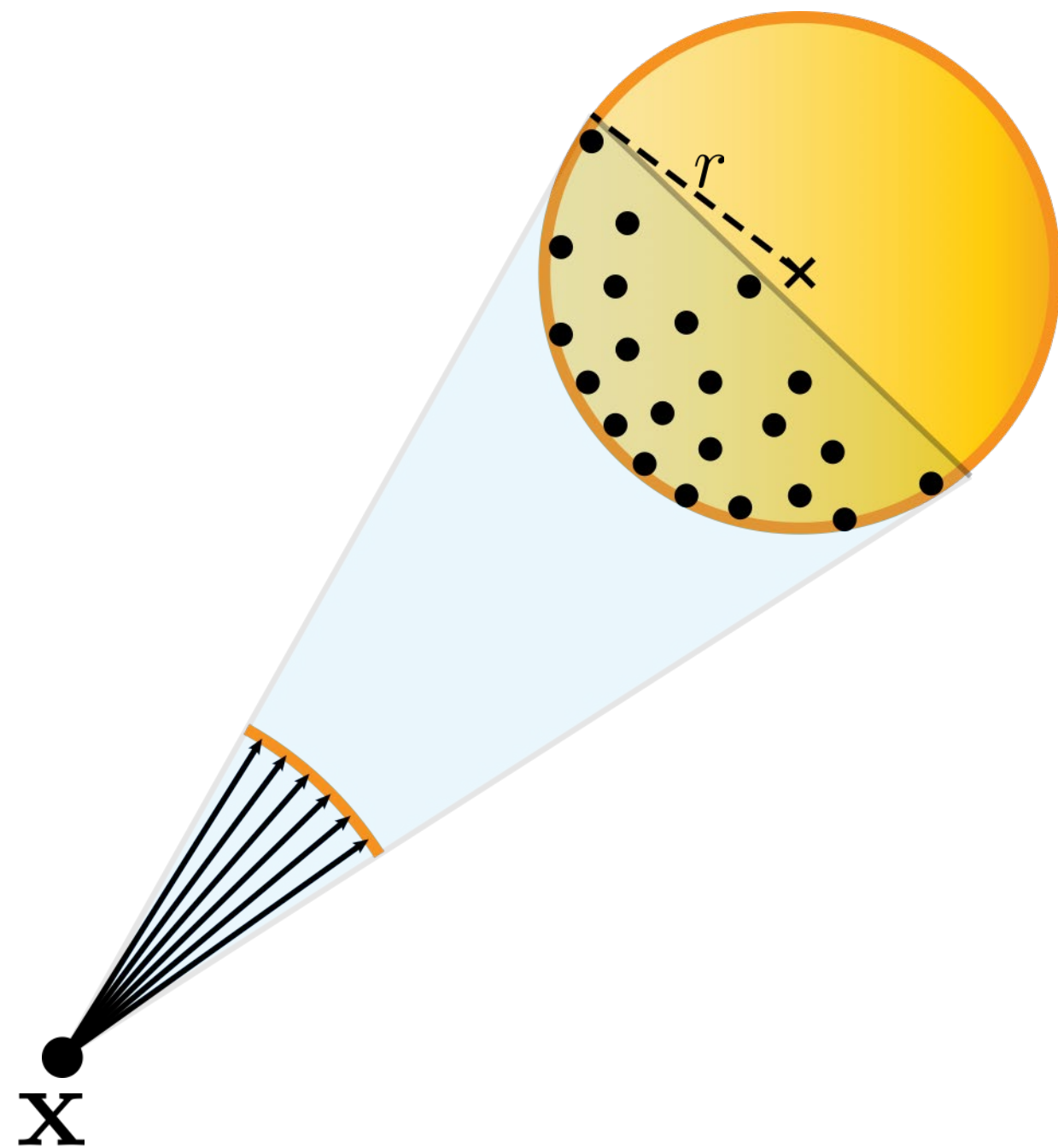
●  
x

# Sphere Light

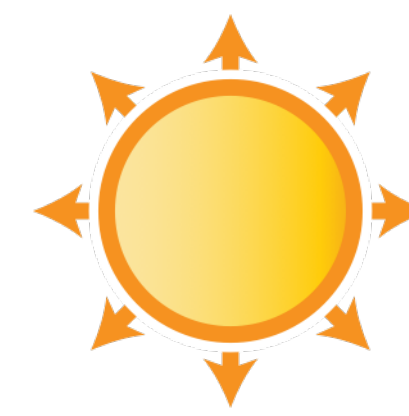


How to sample points on the sphere light?

**Approach 3** (even better): uniformly sample *solid angle* subtended by the sphere

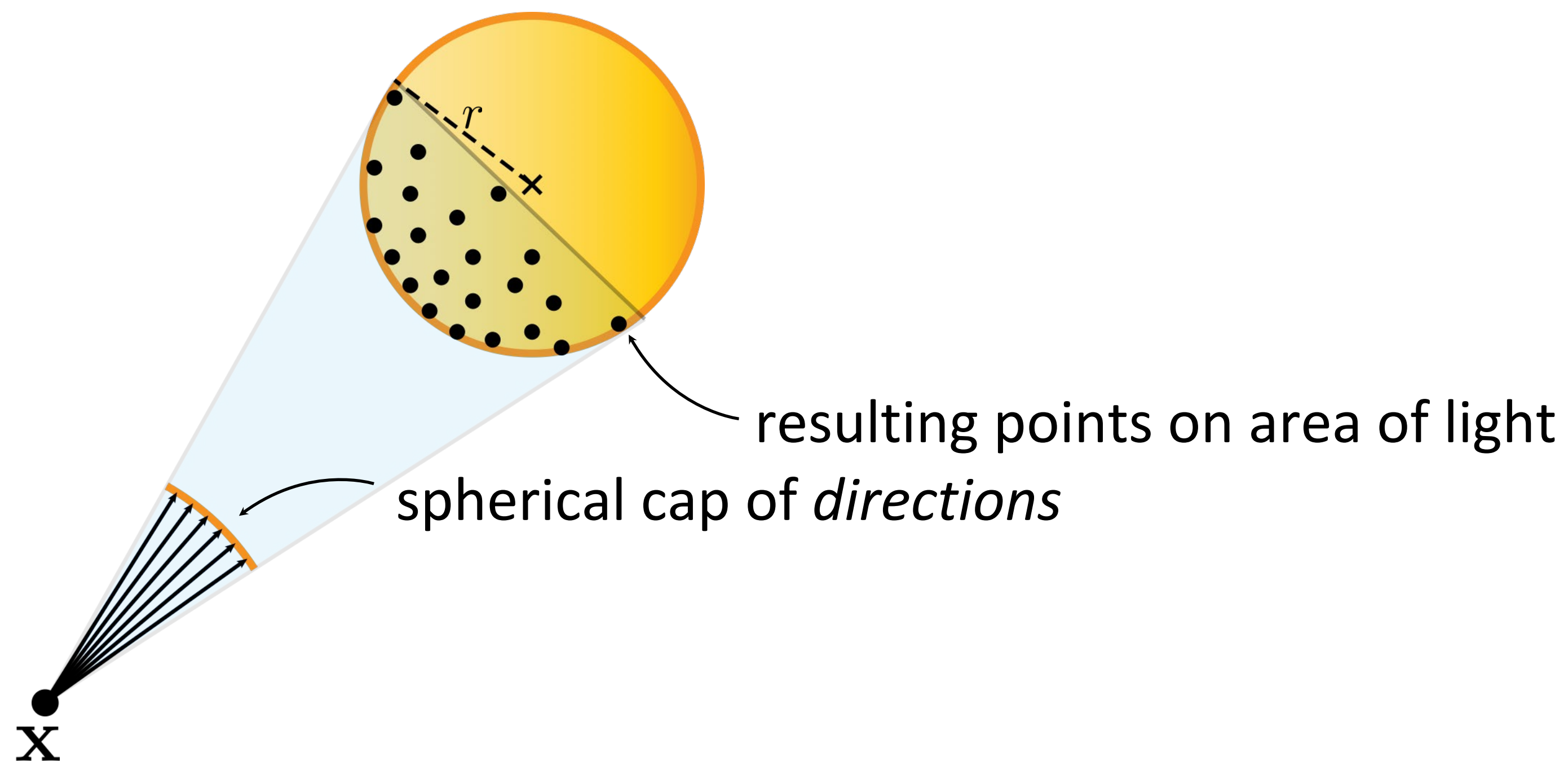


# Sphere Light



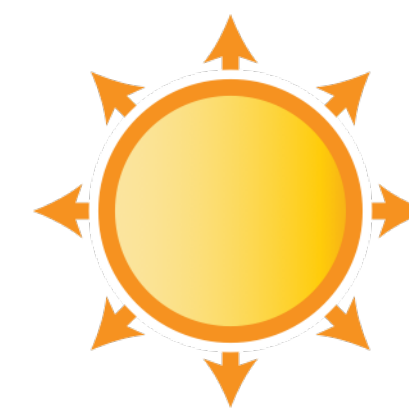
How to sample points on the sphere light?

**Approach 3** (even better): uniformly sample *solid angle* subtended by the sphere





# Sphere Light

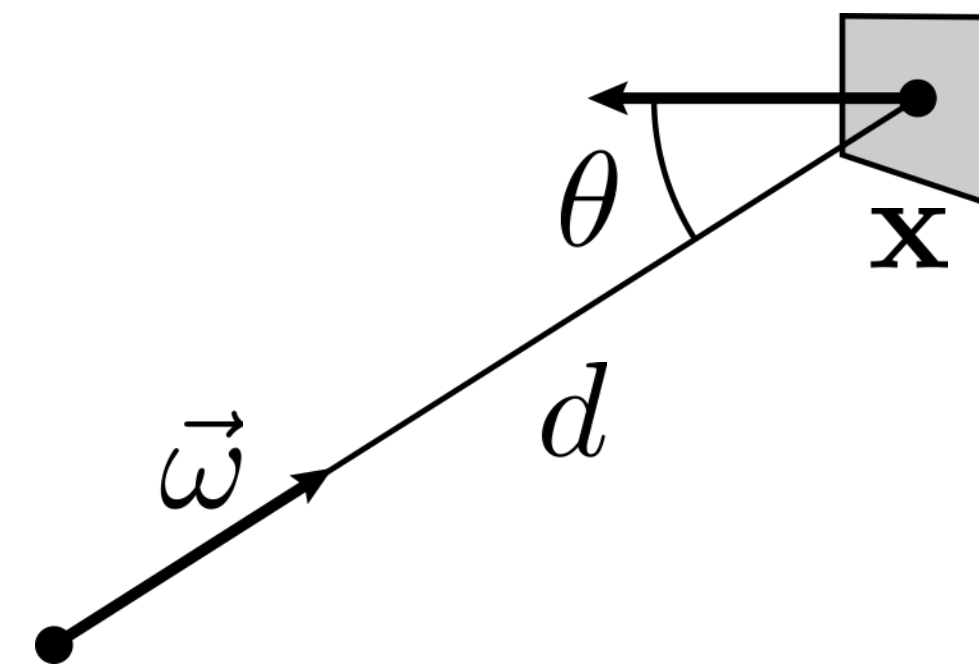


How to sample points on the sphere light?

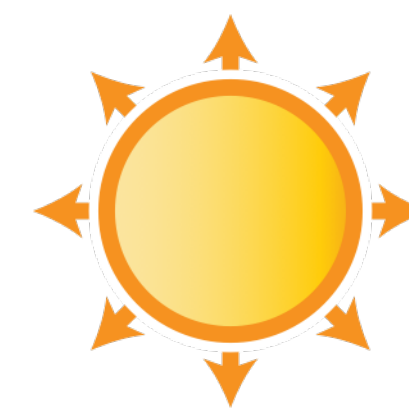
## Caution!

- Approaches use PDFs defined wrt different measures
- Make sure to convert the PDF into the measure of the integral!

$$p_A(\mathbf{x}) = \frac{\cos \theta}{d^2} p_\Omega(\vec{\omega})$$
$$p_\Omega(\vec{\omega}) = \frac{d^2}{\cos \theta} p_A(\mathbf{x})$$



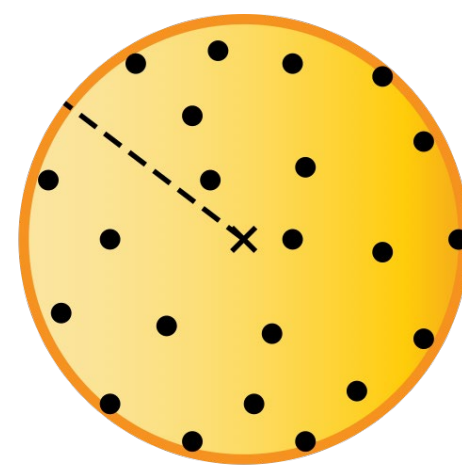
# Sphere Light



How to sample points on the sphere light?

## Caution!

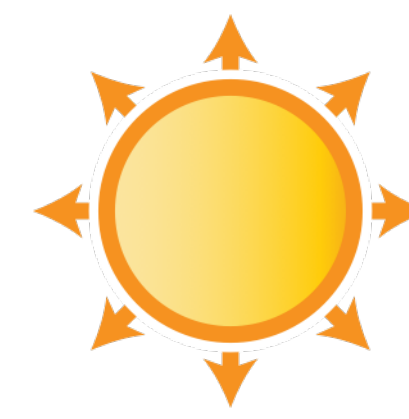
- Approaches use PDFs defined wrt different measures
- Make sure to convert the PDF into the measure of the integral!
- Example: using approach 1 for MC integration of the hemispherical formulation of the reflection eq.



$$\langle L_r(\mathbf{x}, \vec{\omega}_r) \rangle = \frac{1}{N} \sum_{k=1}^N \frac{f_r(\mathbf{x}, \vec{\omega}_{i,k}, \vec{\omega}_r) L_i(\mathbf{x}, \vec{\omega}_{i,k}) \cos \theta_{i,k}}{p_\Omega(\vec{\omega}_{i,k})}$$

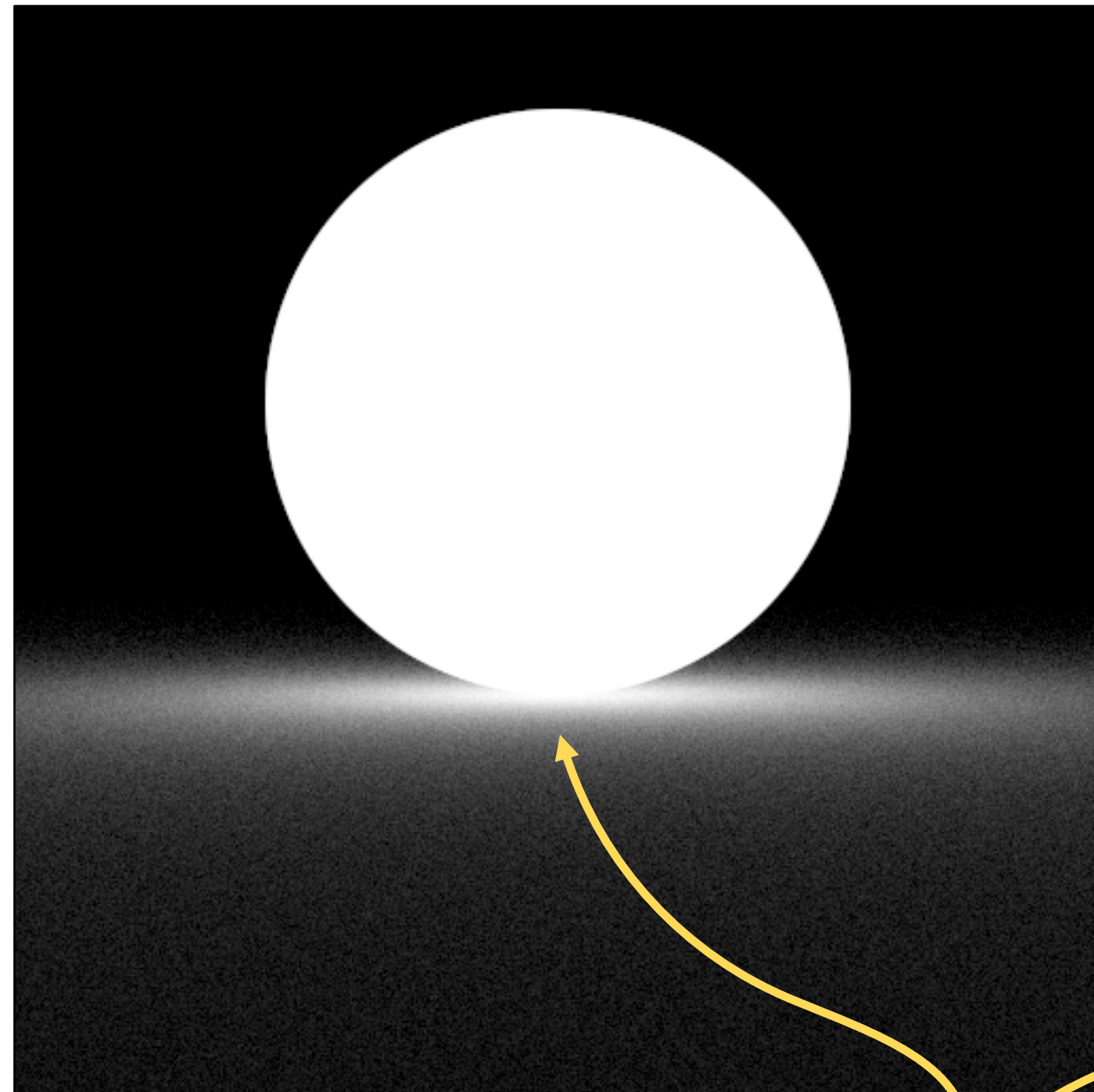
$$p_A(\mathbf{y}) = \frac{1}{4\pi r^2} \quad p_\Omega(\vec{\omega}_i) = \frac{\|\mathbf{x} - \mathbf{y}\|^2}{|-\omega_i \cdot \mathbf{n}_y| 4\pi r^2}$$

# Sphere Light

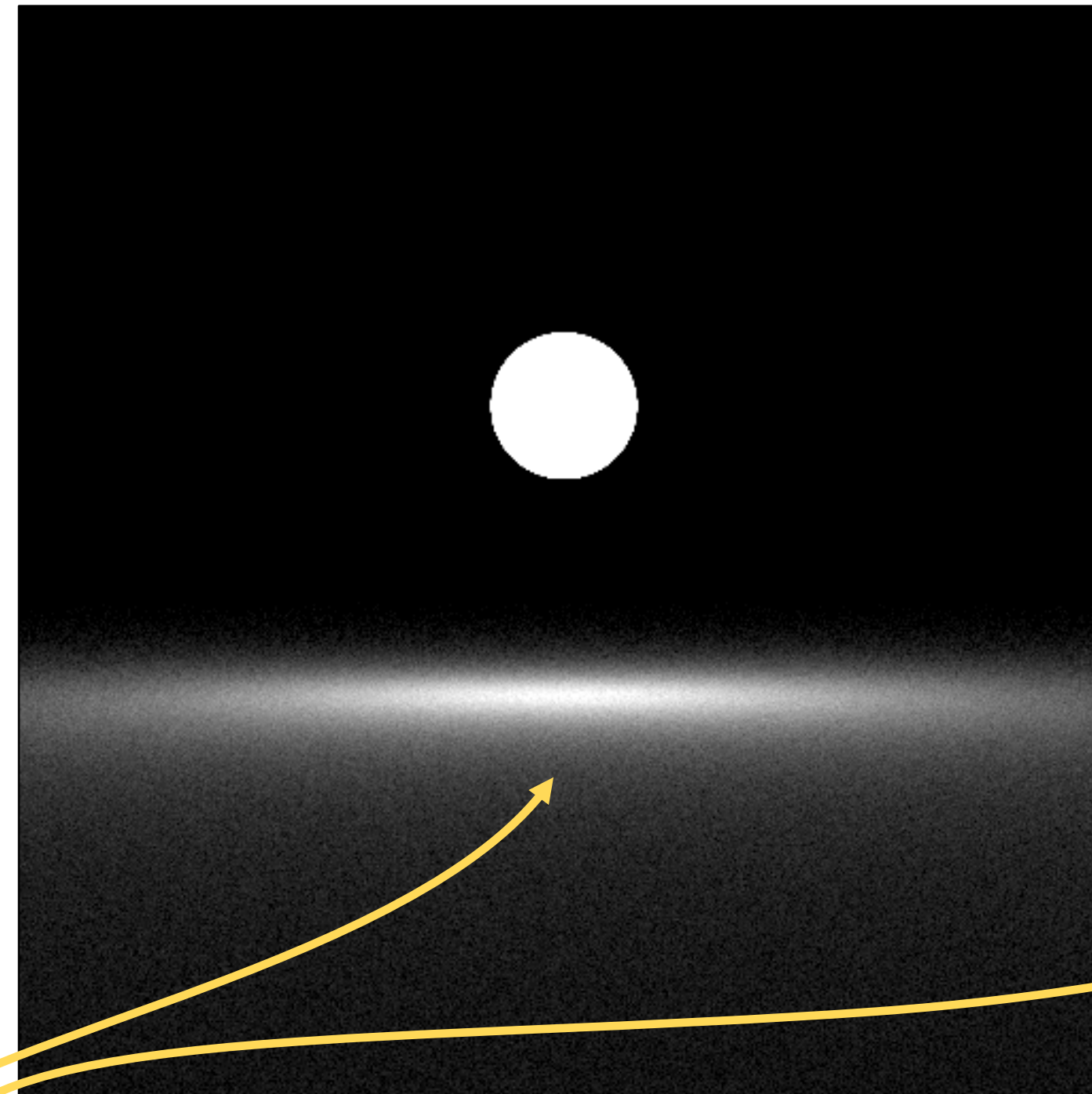


**Validation:** irradiance is independent of radius  
(assuming it emits always the same power & no occluders)

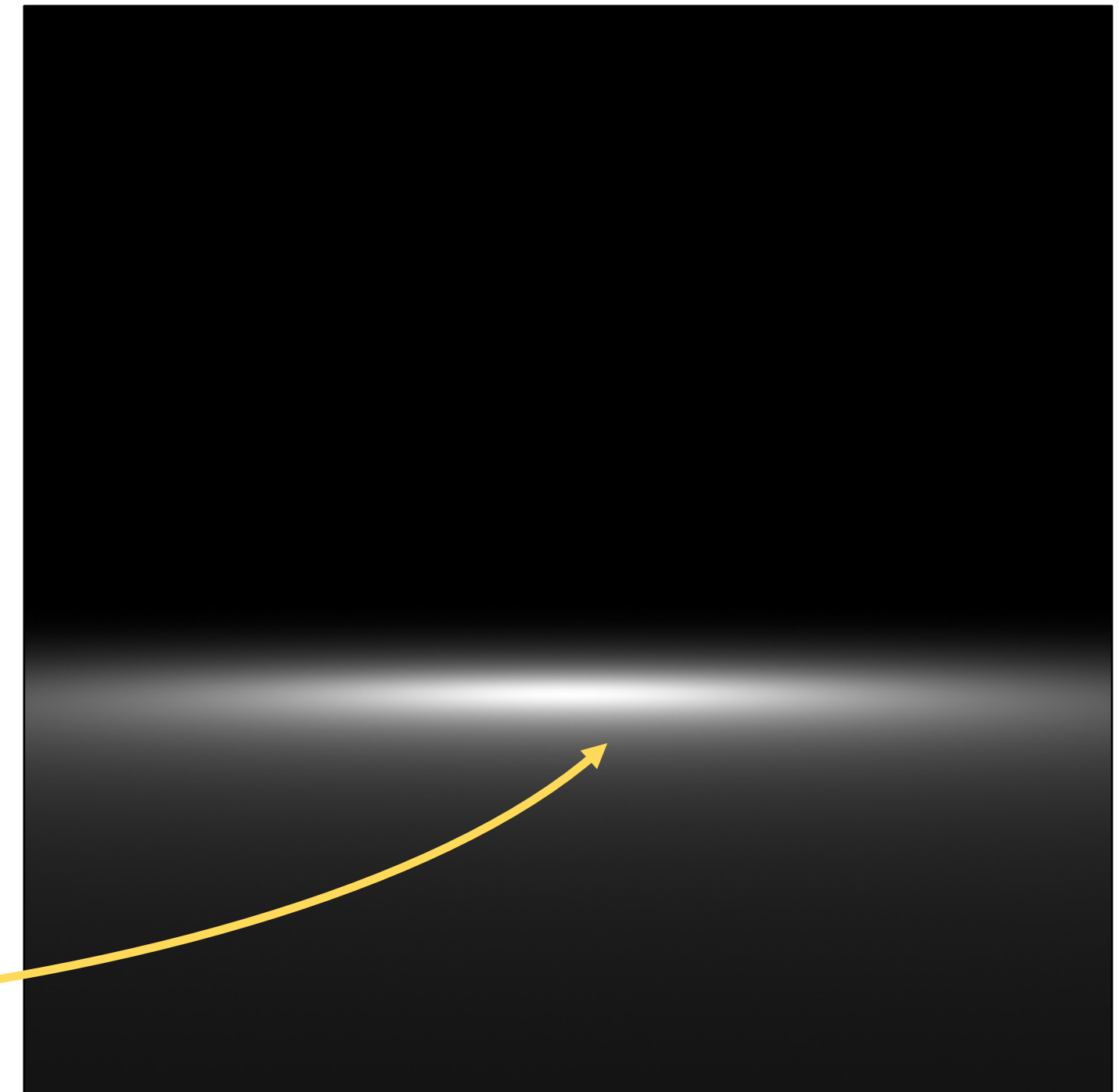
A sphere light



A smaller sphere light



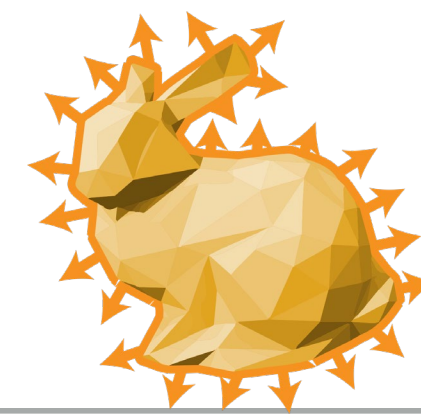
A point light



Identical irradiance profiles

# Mesh Light

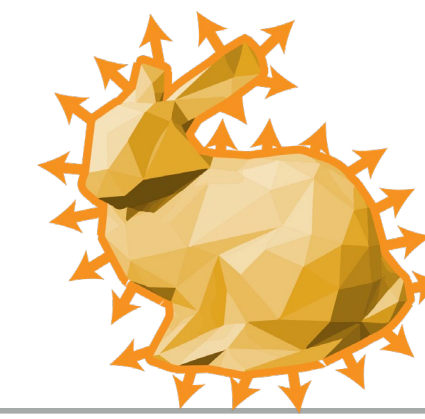
---



An emissive mesh where every surface point emits given radiance  $L_e$

Total area:  $\sum A(k)$

# Mesh Light



How to importance sample?

## Preprocess:

- build a discrete PDF,  $p_{\Delta}$ , for choosing polygons (triangles) *proportional to their area*:

$$p_{\Delta}(i) = \frac{A(i)}{\sum_k A(k)}$$

## Run-time:

- sample a polygon  $\dot{i}$  and a point  $\mathbf{x}$  on  $\dot{i}$
- compute the PDF of choosing the point:

$$p_A(\mathbf{x}) = p_{\Delta}(i)p_A(\mathbf{x}|i) = \frac{1}{\sum A(k)}$$

# Light Sources

---

Point  
light



Spot  
light



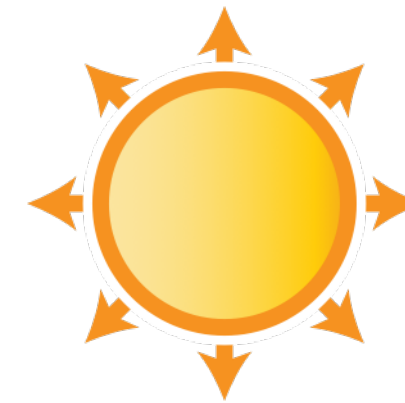
Directional  
light



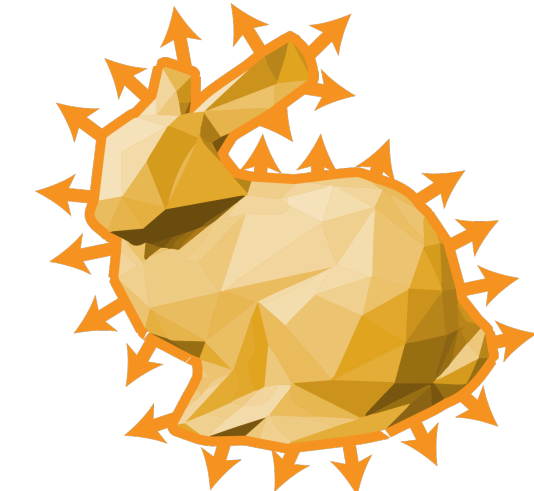
Quad  
light



Sphere  
light



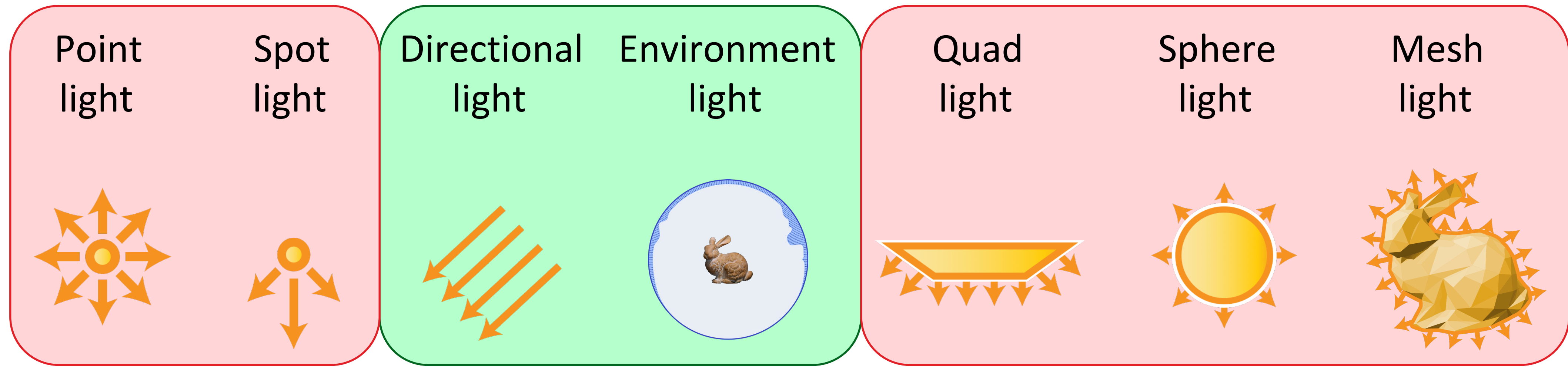
Mesh  
light



Delta lights  
(create hard shadows)

Area/Shape lights  
(create soft shadows)

# Light Sources



Delta lights  
(create hard shadows)

Finite lights  
(create soft shadows)

- sample using surface integral form
- sample using hemispherical integral form

typically, but not always

# Reflection Equation

---

$$L_r(\mathbf{x}, \vec{\omega}_r) = \int_{H^2} f_r(\mathbf{x}, \vec{\omega}_i, \vec{\omega}_r) L_i(\mathbf{x}, \vec{\omega}_i) \cos \theta_i d\vec{\omega}_i$$

What terms can we importance sample?

- BRDF
- incident radiance
- cosine term

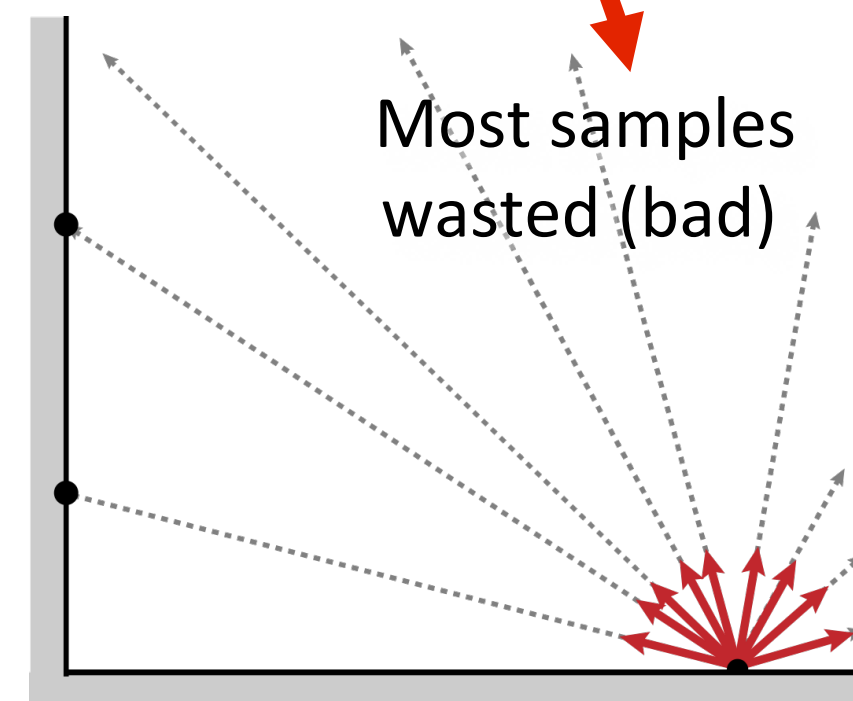
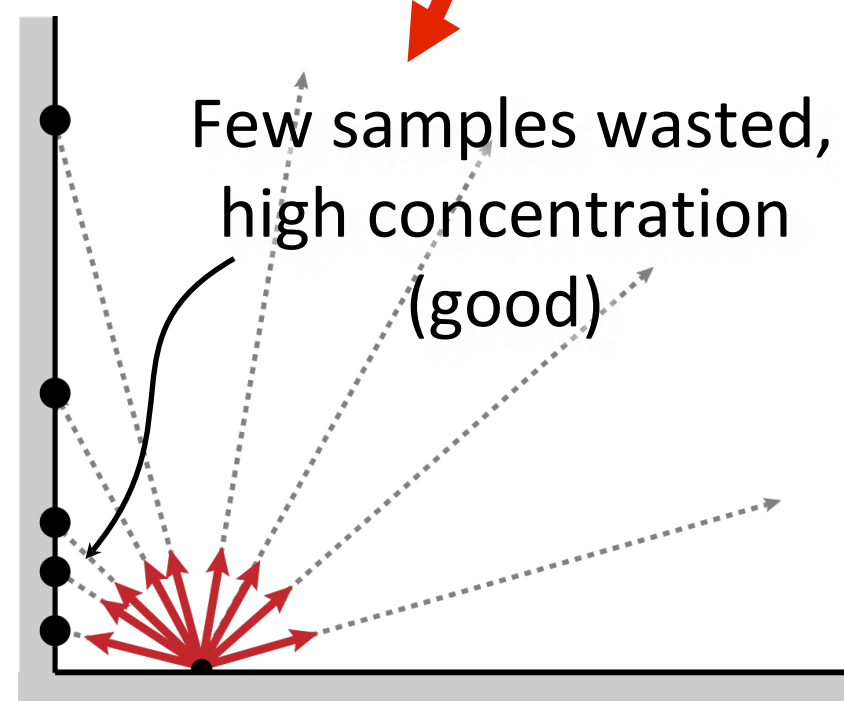
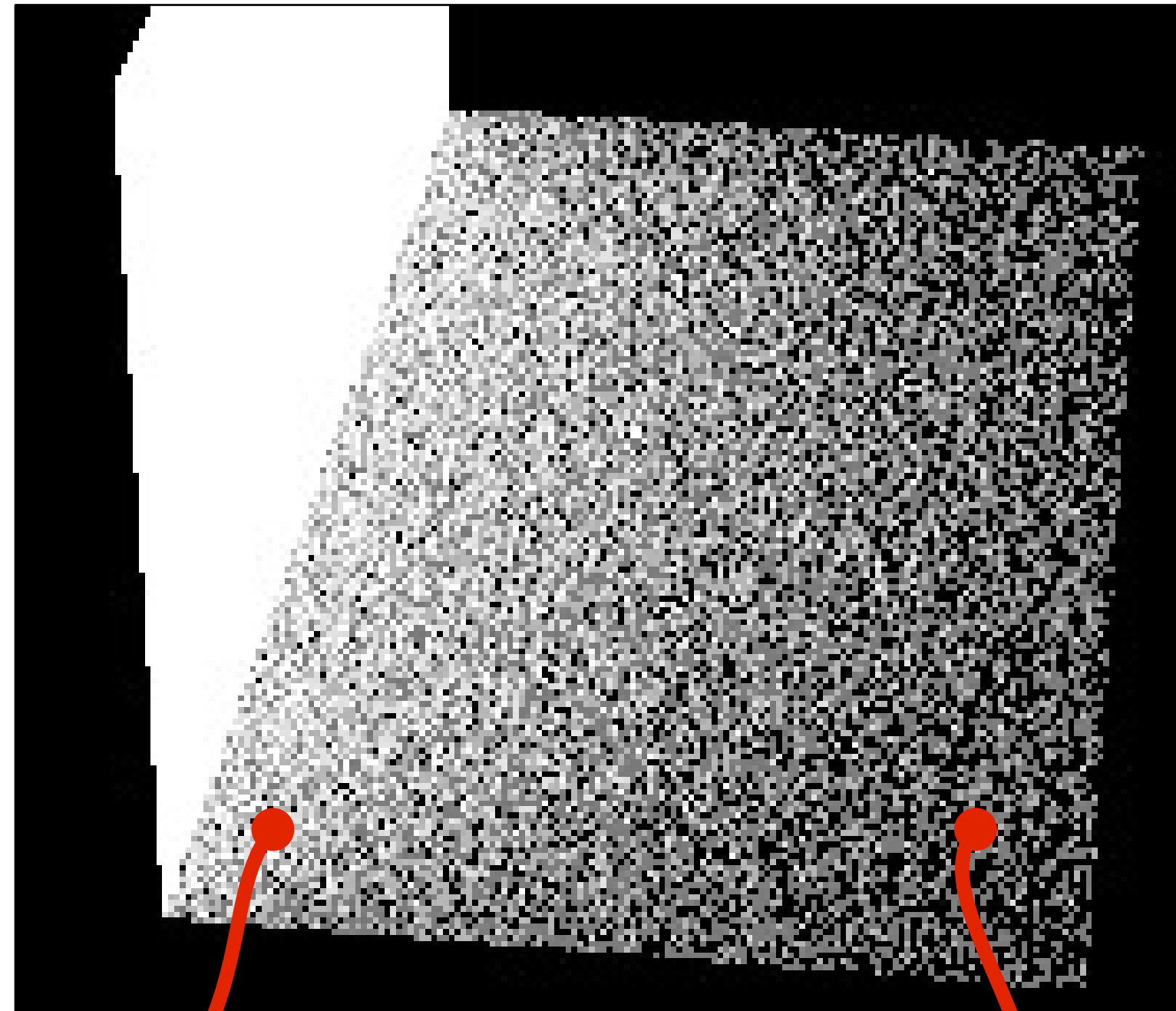
What terms **should** we importance sample?

- depends on the context, hard to make a general statement

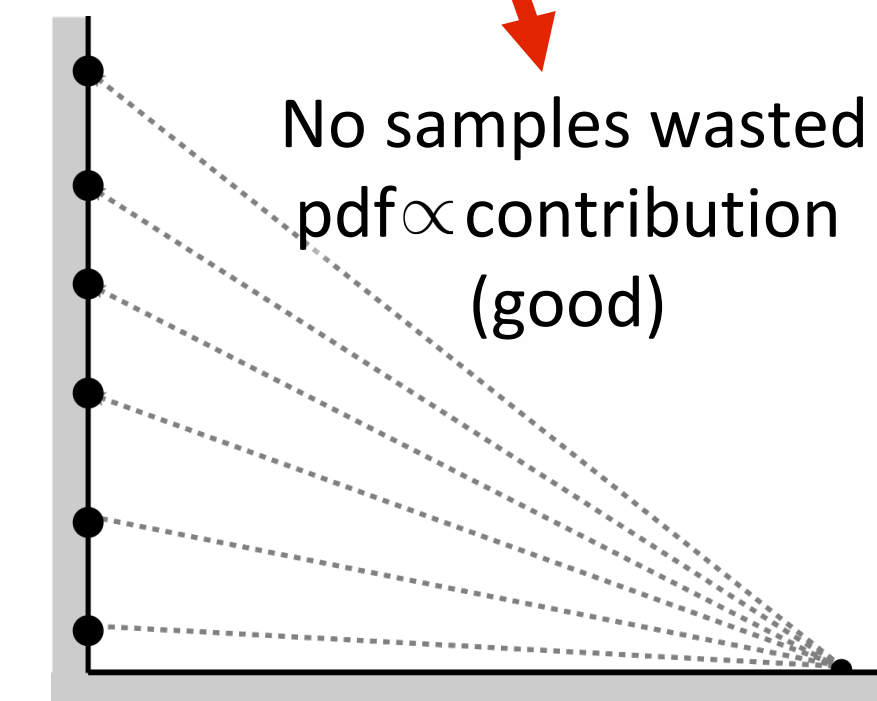
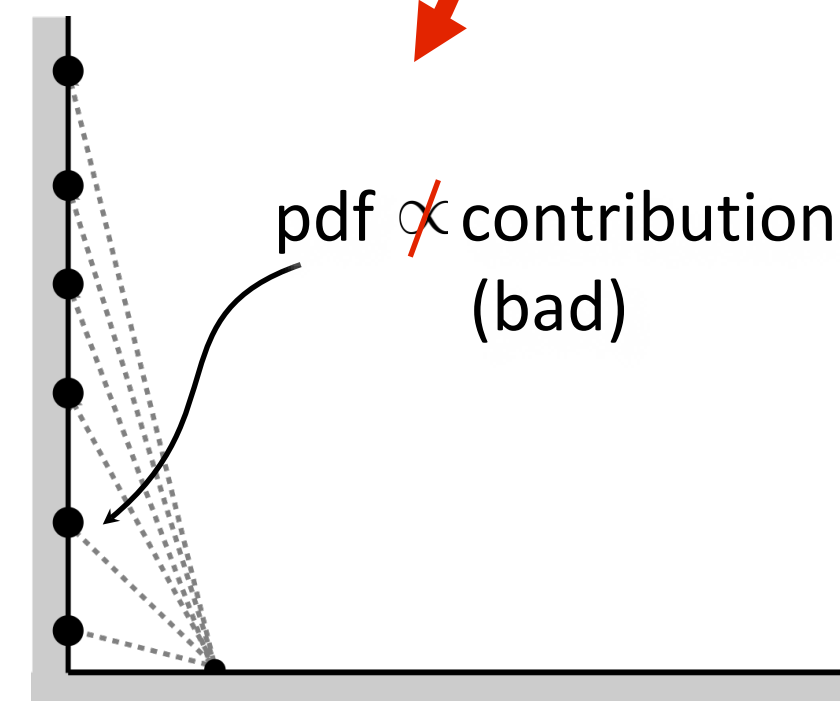
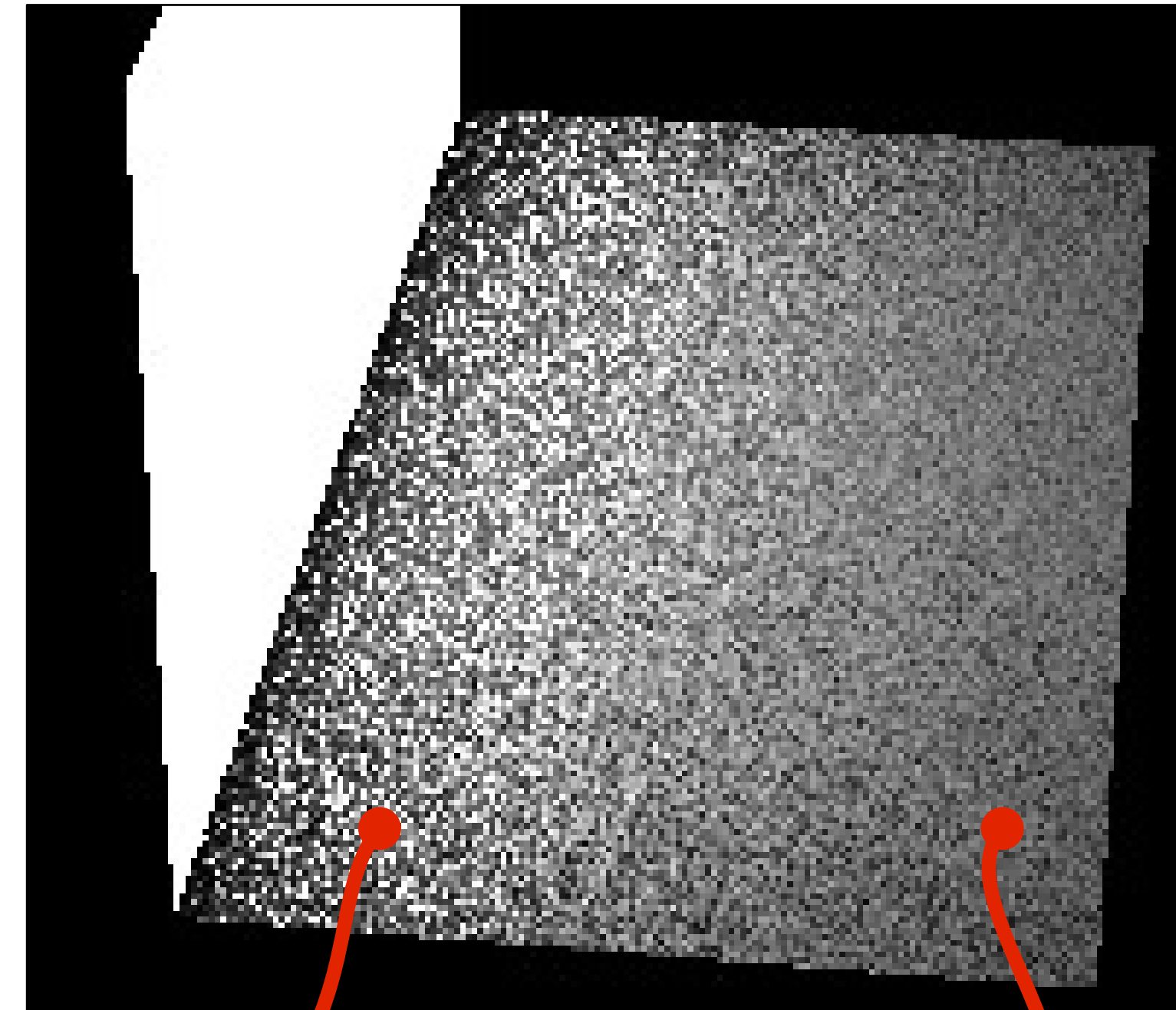


# Multiple Strategies

Cosine-weighted hemisphere



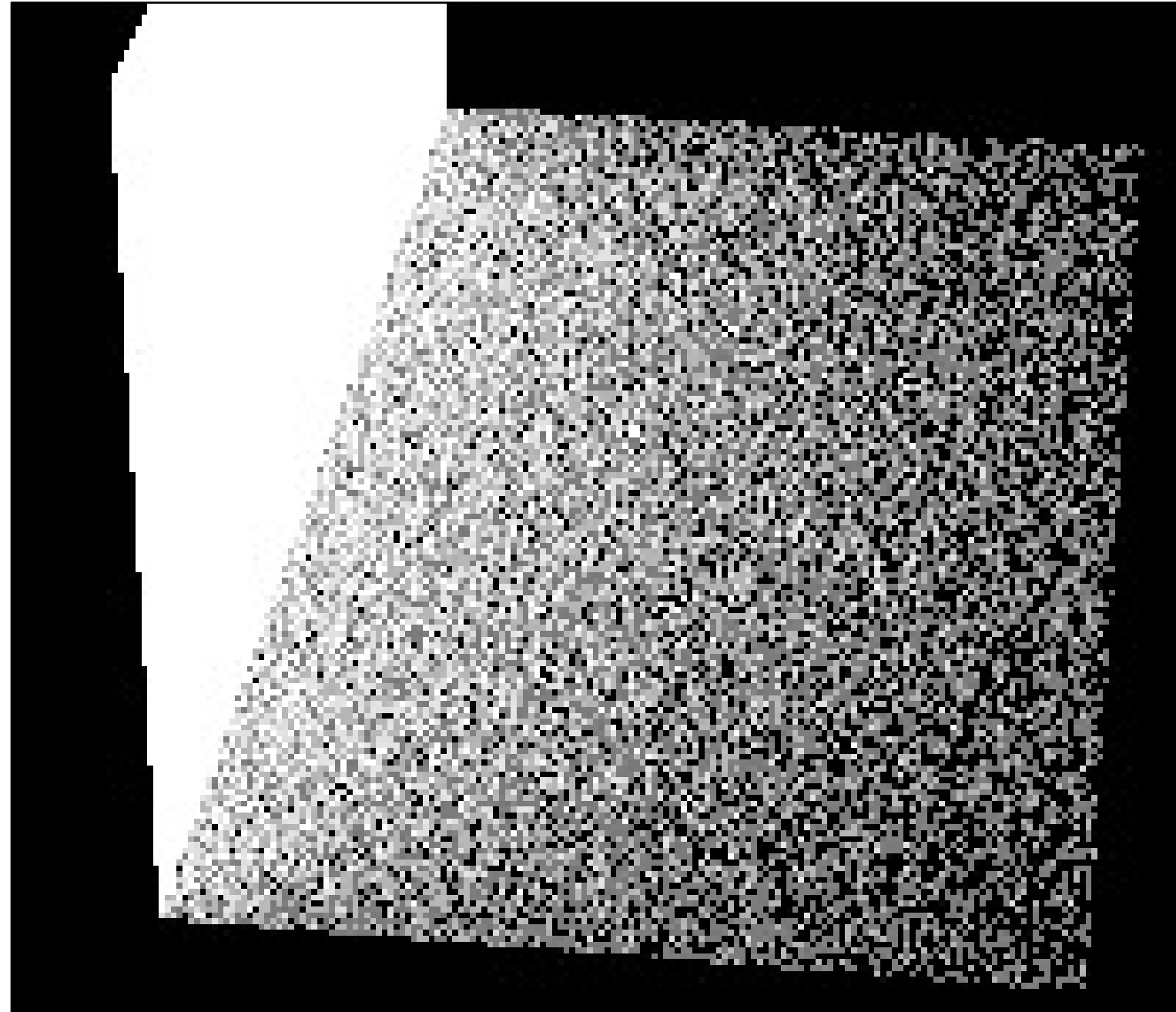
Uniform surface area



# Combining Multiple Strategies

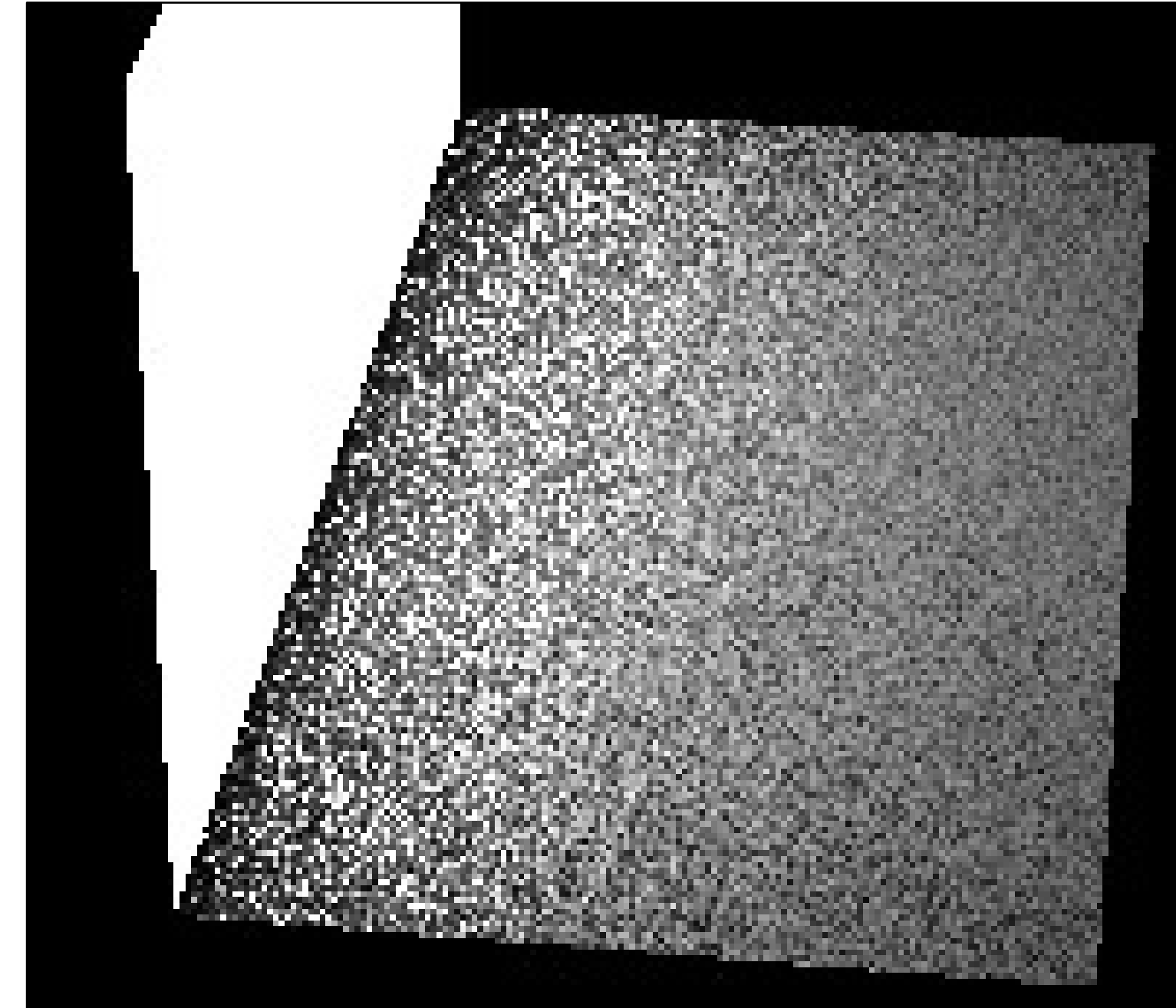
---

Cosine-weighted hemisphere



$$p_1(\vec{\omega}) = \frac{\cos \theta}{\pi}$$

Uniform surface area

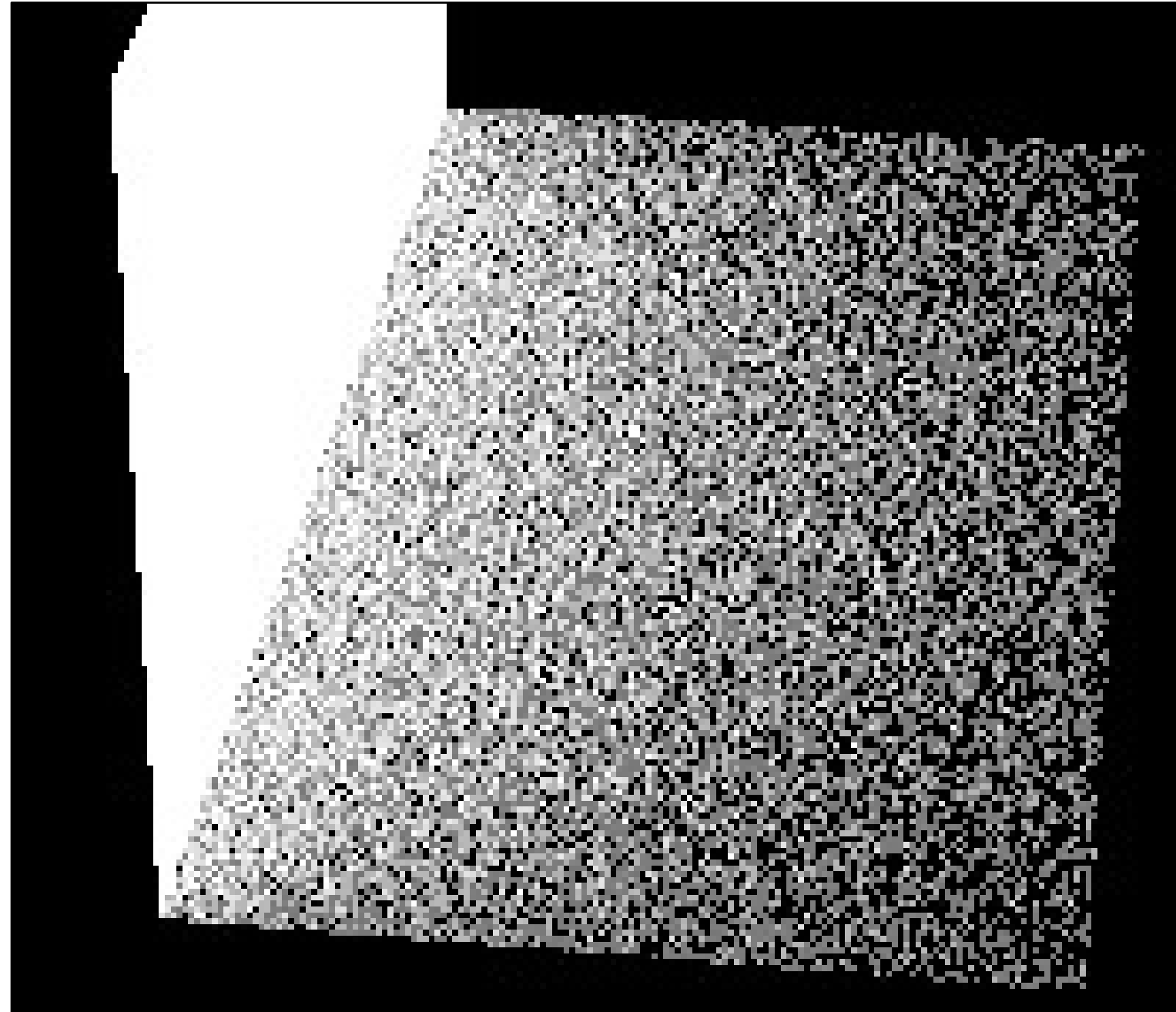


$$p_2(\mathbf{x}) = \frac{1}{A}$$

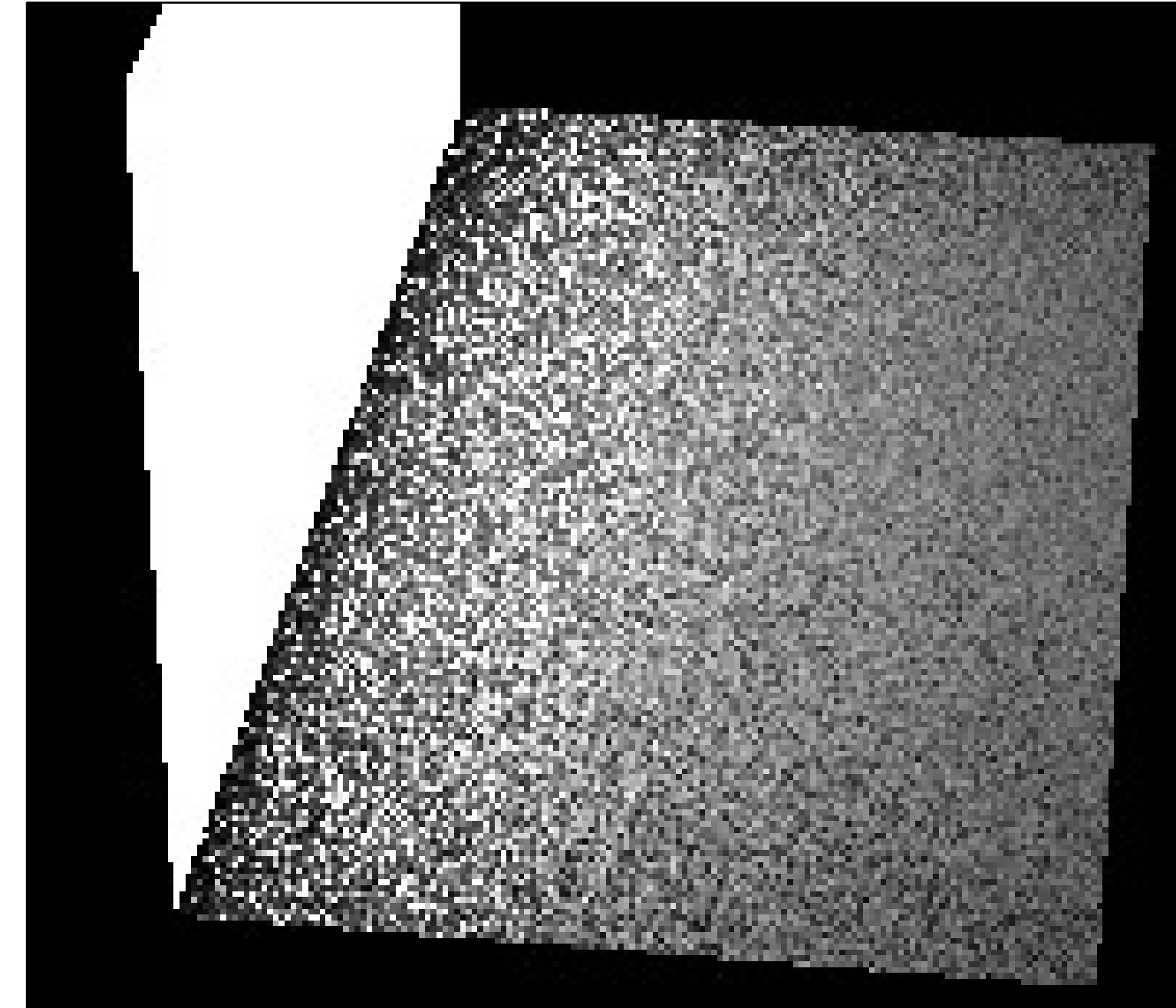
# Combining Multiple Strategies

---

Cosine-weighted hemisphere



Uniform surface area



$$p_1(\vec{\omega}) = \frac{\cos \theta}{\pi}$$

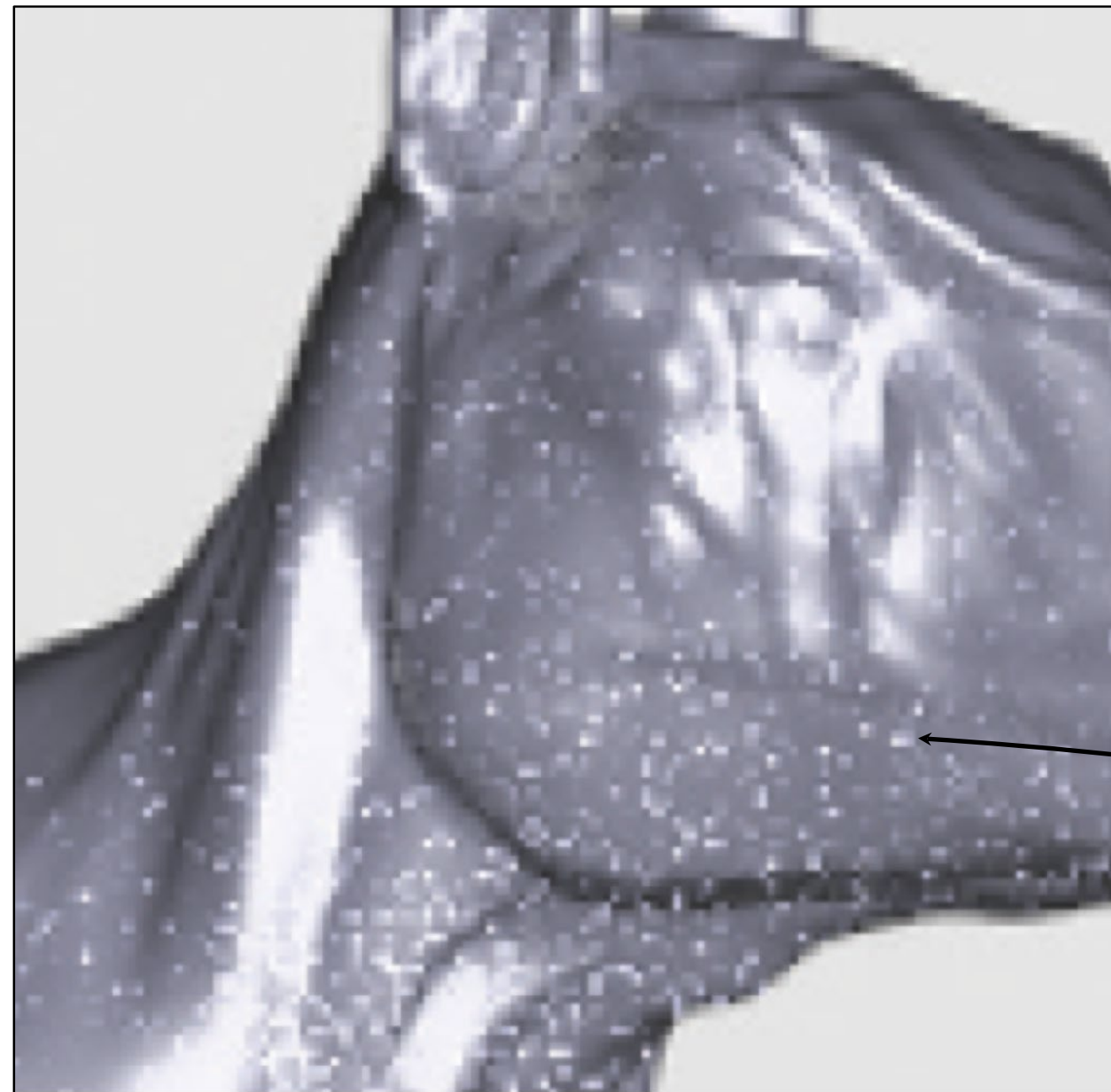
$$p_2(\mathbf{x}) = \frac{1}{A} \quad p_2(\vec{\omega}) = \frac{1}{A} \frac{d^2}{\cos \theta}$$

# Fireflies

---

In MC integration, variance is high when the PDF is not proportional to the integrand

Worst case: *rare* samples with *huge* contributions



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

← large value  
← small value

“fireflies”

# Motivation

---

In MC integration, variance is high when the PDF is not proportional to the integrand

Worst case: *rare* samples with *huge* contributions

$$\langle F^N \rangle = \frac{1}{N} \sum_{i=1}^N \frac{f(x_i)}{p(x_i)} \begin{array}{l} \leftarrow \text{large value} \\ \leftarrow \text{small value} \end{array}$$

We often have multiple sampling strategies

If at least one covers each part of the integrand well, then combining them should reduce fireflies

# Mixture sampling and multiple importance sampling (MIS)

# Combining Multiple Strategies

---

Could just average two different estimators:

$$\frac{0.5}{N_1} \sum_{i=1}^{N_1} \frac{f(x_i)}{p_1(x_i)} + \frac{0.5}{N_2} \sum_{i=1}^{N_2} \frac{f(x_i)}{p_2(x_i)}$$

- doesn't really help if weights independent of sample: *variance is additive*

# Mixture sampling

---

Instead of averaging multiple estimators

$$\frac{0.5}{N_1} \sum_{i=1}^{N_1} w_1(x_i) \frac{f(x_i)}{p_1(x_i)} + \frac{0.5}{N_2} \sum_{i=1}^{N_2} w_2(x_i) \frac{f(x_i)}{p_2(x_i)}, \quad N_1 + N_2 = N$$

sample from the average PDF

$$\frac{1}{N} \sum_{i=1}^N \frac{f(x_i)}{0.5(p_1(x_i) + p_2(x_i))}$$



# Sample from Average PDF (mixture sampling)

---

You are given two sampling functions and their corresponding pdfs:

```
float sample1(float rnd); float pdf1(float x);  
float sample2(float rnd); float pdf2(float x);
```

Create a new function:

```
float sampleAvg(float rnd);
```

which has the corresponding pdf:

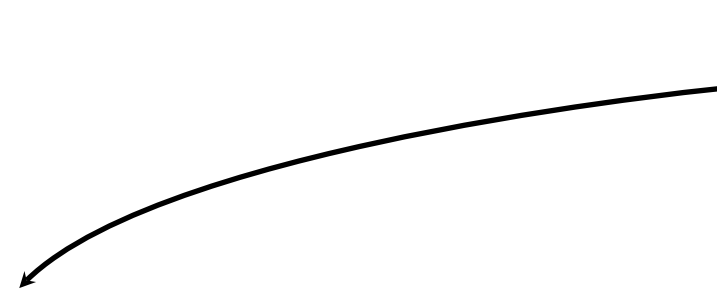
```
float pdfAvg(float x)  
{  
    return 0.5 * (pdf1(x) + pdf2(x));  
}
```

# Sample from Average PDF (mixture sampling)

---

```
float sampleAvg(float rnd)
{
float Prob1 = 0.5;
if (rand.nextFloat() < Prob1)
return sample1(rnd);
else
return sample2(rnd);
}
```

Requires extra random number (can be avoided)

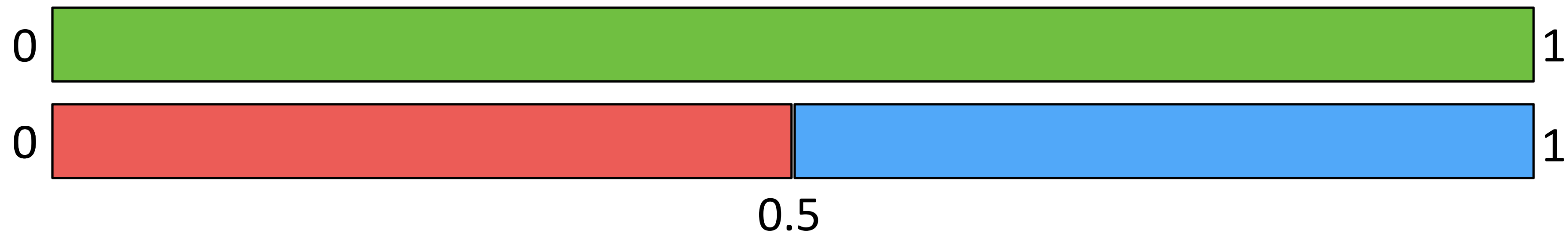


# Sample from Average PDF (mixture sampling)

---

```
float sampleAvg(float rnd)
{
float Prob1 = 0.5;
if (rnd < Prob1)
return sample1(rnd);
else
return sample2(rnd);
}
```

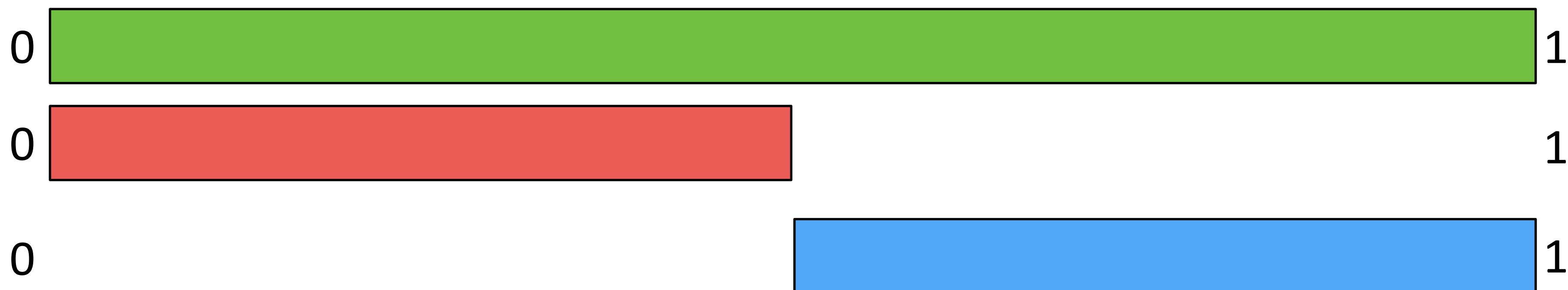
These need to be uniform random numbers in [0..1)



# Sample from Average PDF (mixture sampling)

```
float sampleAvg(float rnd)
{
float Prob1 = 0.5;
if (rnd < Prob1)
return sample1(rnd);
else
return sample2(rnd);
}
```

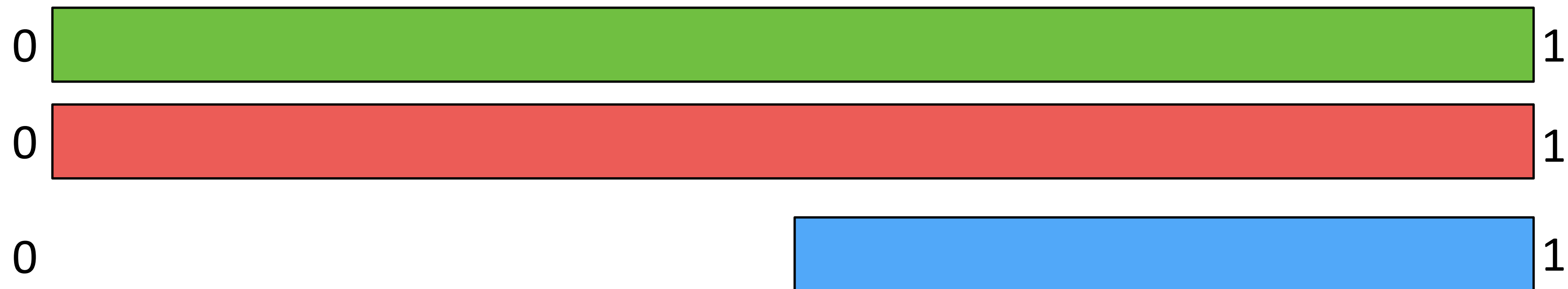
These need to be uniform random numbers in [0..1)



# Sample from Average PDF (mixture sampling)

---

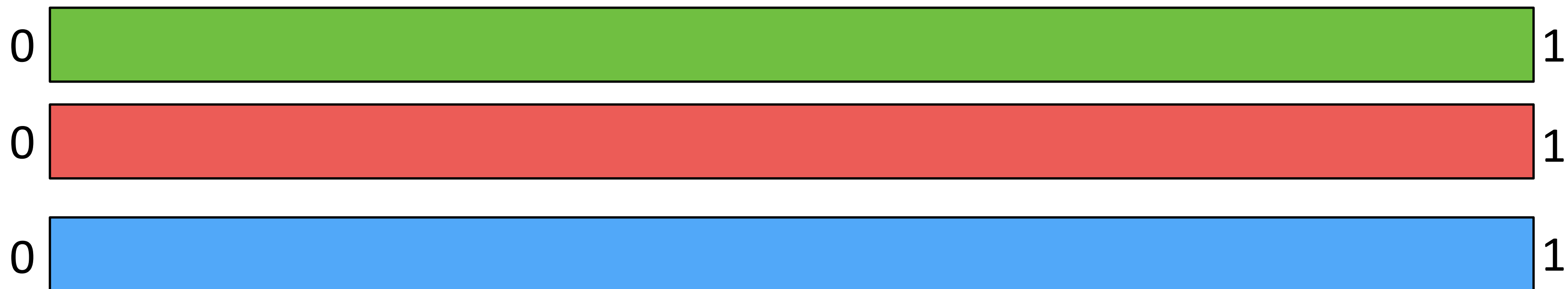
```
float sampleAvg(float rnd)
{
float Prob1 = 0.5;
if (rnd < Prob1)
return sample1(rnd / Prob1);
else
return sample2(rnd);
}
```



# Sample from Average PDF (mixture sampling)

---

```
float sampleAvg(float rnd)
{
float Prob1 = 0.5;
if (rnd < Prob1)
return sample1(rnd / Prob1);
else
return sample2((rnd-Prob1) / (1-Prob1));
}
```



# Sample from Weighted Average

---

```
float sampleWeightedAvg(float rnd)
{
float Prob1 = 0.25;
if (rnd < Prob1)
return sample1(rnd / Prob1);
else
return sample2((rnd-Prob1) / (1-Prob1));
}

float pdfWeightedAvg(float x)
{
return 0.25 * pdf1(x) + 0.75 * pdf2(x);
}
```

Still works, just change Prob1

# Multiple Importance Sampling

---

Combination of 2 strategies using *sample-dependent* weights:

$$\langle F^{\text{MIS}} \rangle = w_1(x_1) \frac{f(x_1)}{p_1(x_1)} + w_2(x_2) \frac{f(x_2)}{p_2(x_2)}$$

– where:

$$w_1(x) + w_2(x) = 1$$



# Multiple Importance Sampling

---

Combination of  $M$  strategies with *sample-dependent* weights:

$$\langle F^{\Sigma N_s} \rangle = \sum_{s=1}^M \frac{1}{N_s} \sum_{i=1}^{N_s} w_s(x_i) \frac{f(x_i)}{p_s(x_i)}$$

– where:

$$\sum_{s=1}^M w_s(x) = 1$$

How to choose the weights?

# Multiple Importance Sampling

---

Balance heuristic (provably good):

$$w_s(x) = \frac{p_s(x)}{\sum_j p_j(x)}$$

Power heuristic (more aggressive, can be better):

$$w_s(x) = \frac{p_s(x)^\beta}{\sum_j p_j(x)^\beta}$$

Other heuristics exist

- e.g. cutoff heuristic, maximum heuristic, ...

# Multiple Importance Sampling

---

**Multi-sample** model: *deterministically* allocate  $N_s$  samples to  $s$ -th strategy

$$\langle F^{\Sigma N_s} \rangle = \sum_{s=1}^M \frac{1}{N_s} \sum_{i=1}^{N_s} w_s(x_i) \frac{f(x_i)}{p_s(x_i)}$$

What if we want to draw just **one** sample?

**One-sample** model: *randomly* select to use  $s$ -th strategy

$$\langle F^1 \rangle = w_s(x) \frac{f(x)}{q_s p_s(x)}$$

where  $q_s$  is the probability of using strategy  $s$ , and  $\sum_{s=1}^N q_s = 1$

# Interpreting the Balance Heuristic

---

Balance heuristic for the one-sample model:

$$w_s(x) = \frac{q_s p_s(x)}{\sum_j q_j p_j(x)}$$

Plugged into the one-sample model:

$$\langle F^1 \rangle = w_s(x) \frac{f(x)}{q_s p_s(x)} = \frac{q_s p_s(x)}{\sum_j q_j p_j(x)} \frac{f(x)}{q_s p_s(x)} = \frac{f(x)}{\sum_j q_j p_j(x)}$$

One-sample model with balance heuristic samples from average PDF (*mixture sampling*)

# Multiple Importance Sampling with Balance Heuristic

---

**Multi-sample** model: Equivalent to mixture sampling with *stratification* (deterministic allocation of samples per strategy).

$$\langle F^{\sum N_s} \rangle = \sum_{s=1}^M \frac{1}{N_s} \sum_{i=1}^{N_s} w_s(x_i) \frac{f(x_i)}{p_s(x_i)}$$

**One-sample** model: Equivalent to mixture sampling.

$$\langle F^1 \rangle = w_s(x) \frac{f(x)}{q_s p_s(x)}$$

where  $q_s$  is the probability of using strategy  $s$ , and  $\sum_{s=1}^N q_s = 1$

# Why Does it Work?

---

Using a single strategy:

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

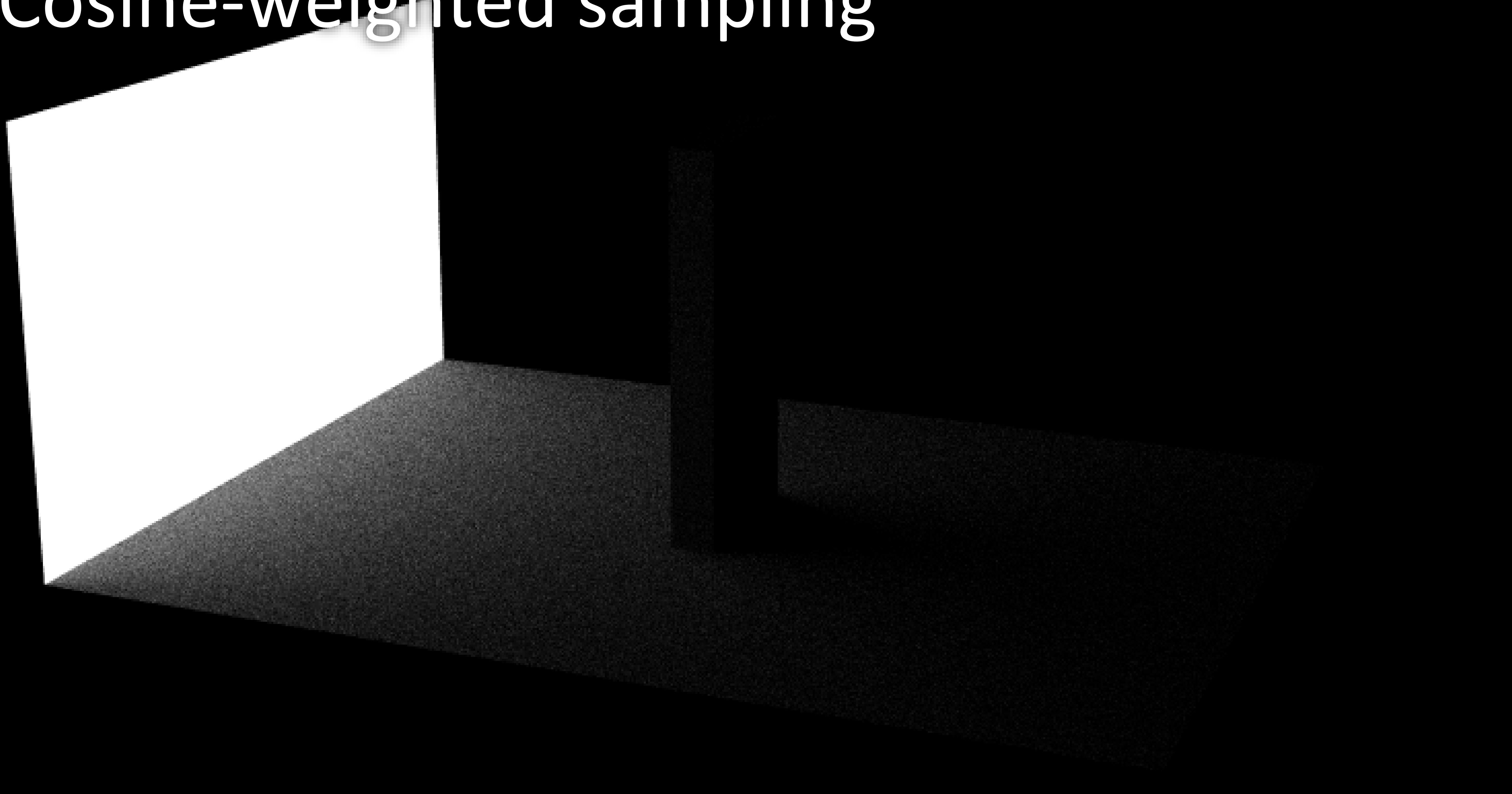
← large value  
← small value

Combining multiple strategies using balance heuristic (MIS or mixture sampling):

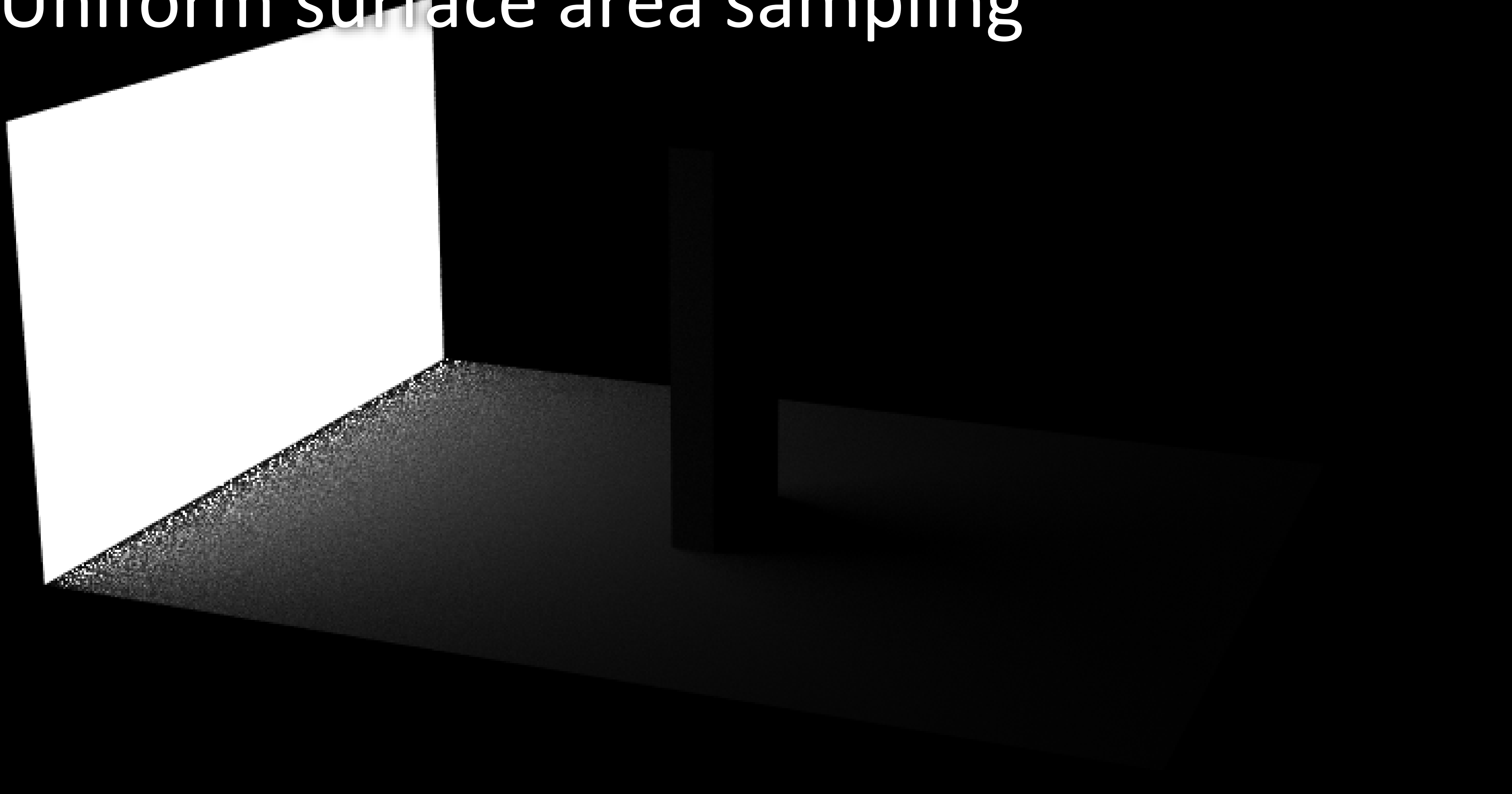
$$\langle F^N \rangle = \frac{1}{N} \sum_{i=1}^N \frac{f(x_i)}{\sum_j q_j p_j(x_i)}$$

← large value  
← relatively large value  
(as long as at least one PDF is large)

# Cosine-weighted sampling

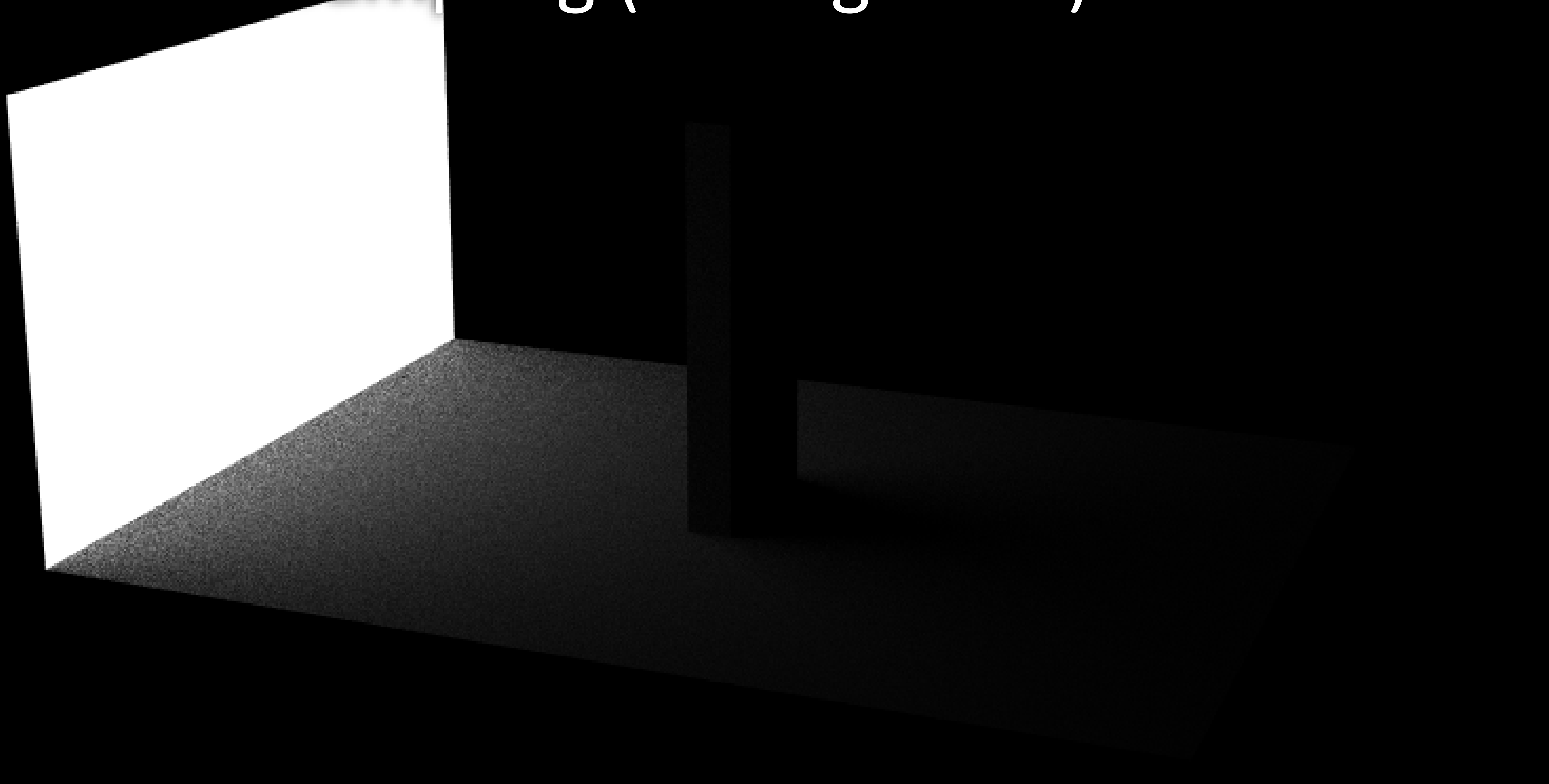


# Uniform surface area sampling

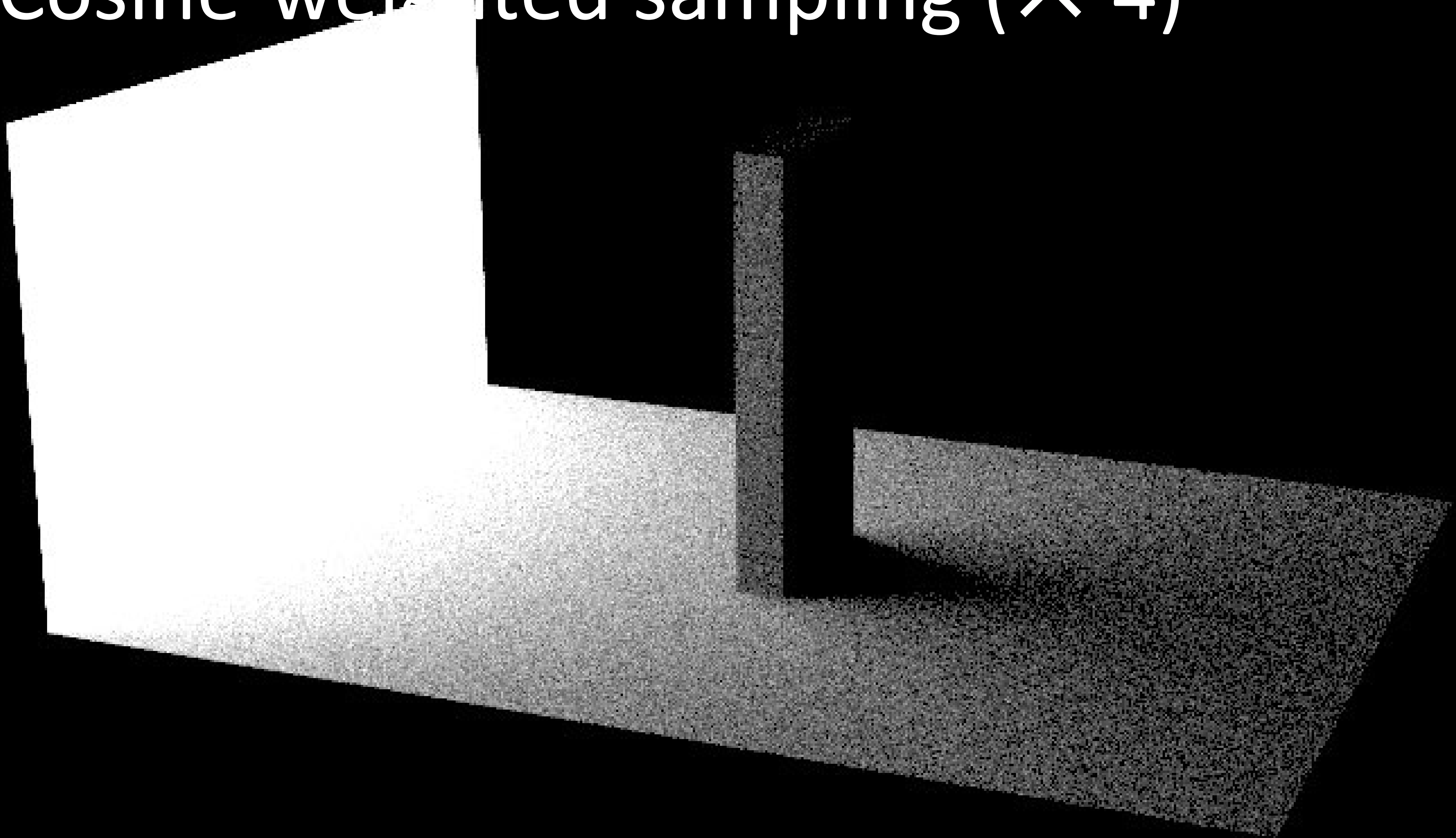




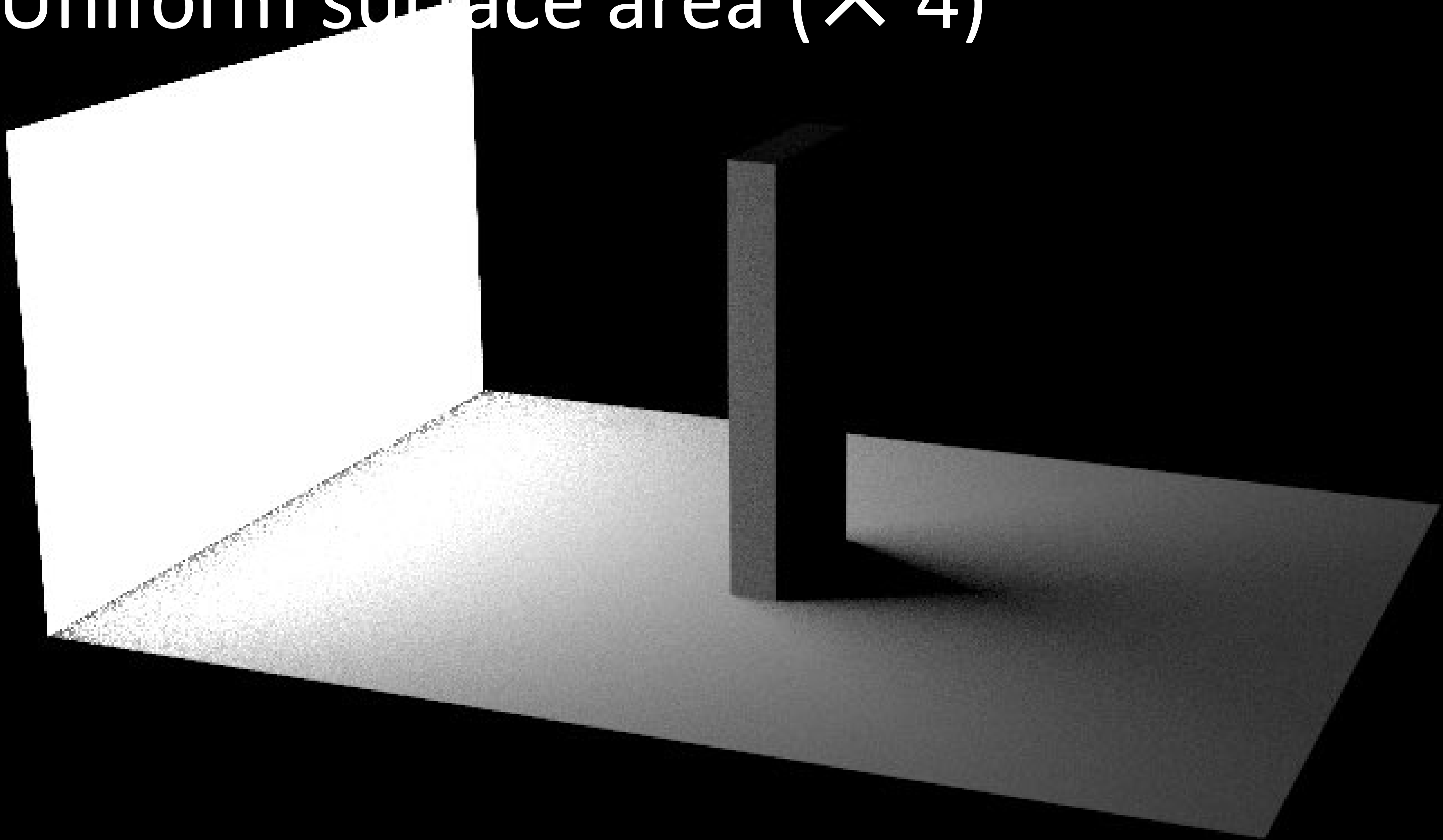
# Mixture sampling (average PDF)



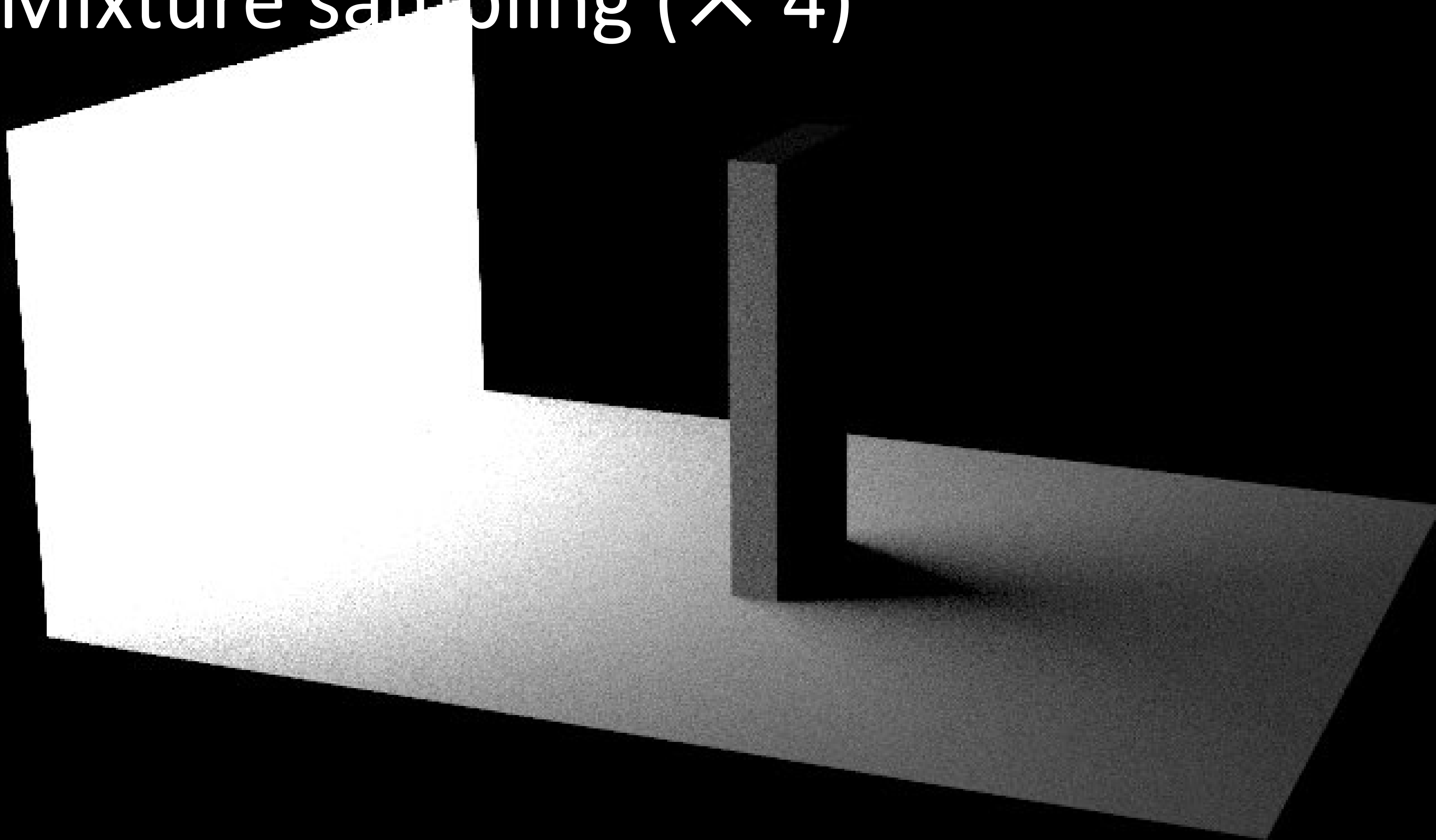
Cosine-weighted sampling ( $\times 4$ )



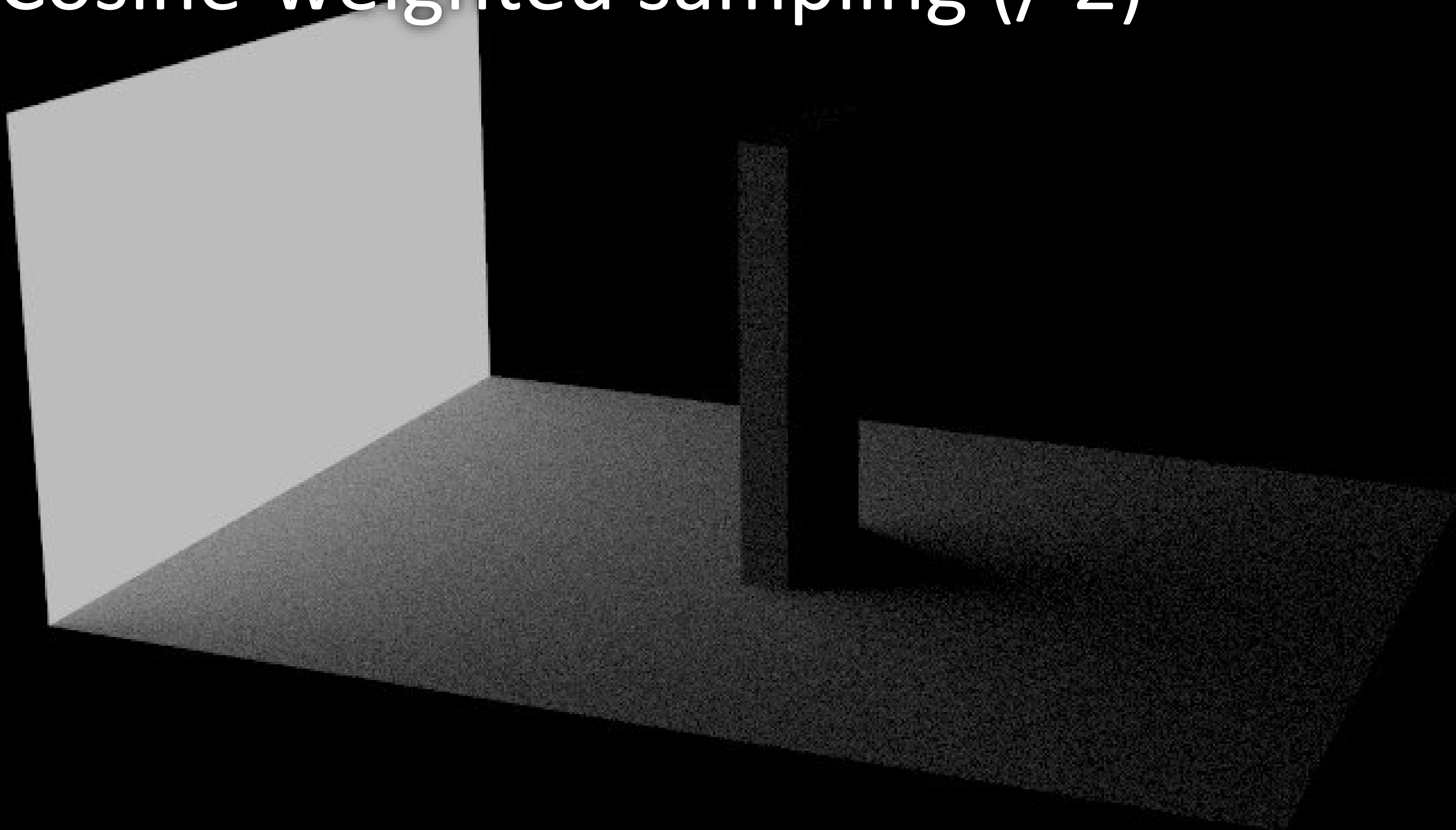
Uniform surface area ( $\times 4$ )



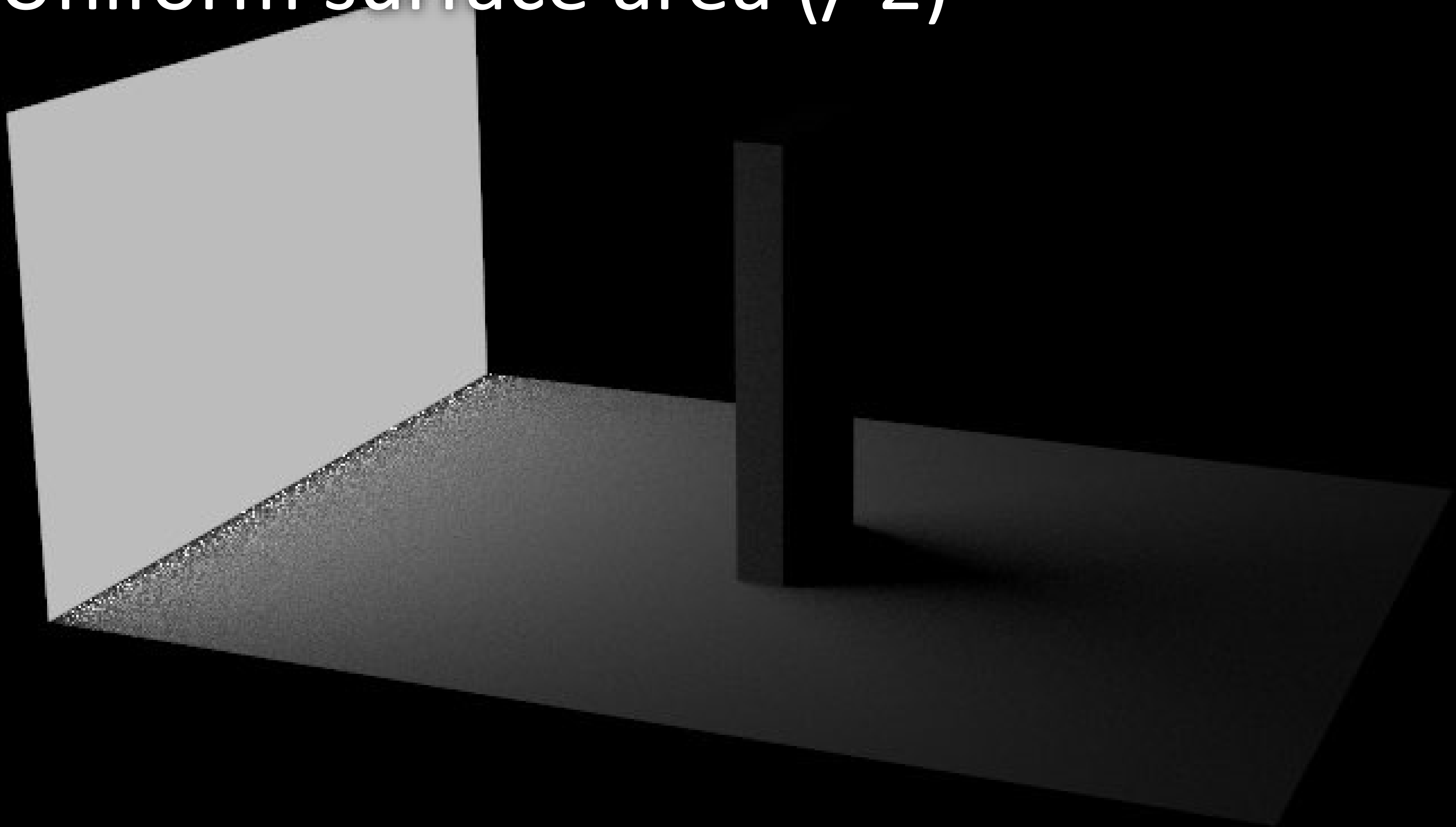
Mixture sampling ( $\times 4$ )



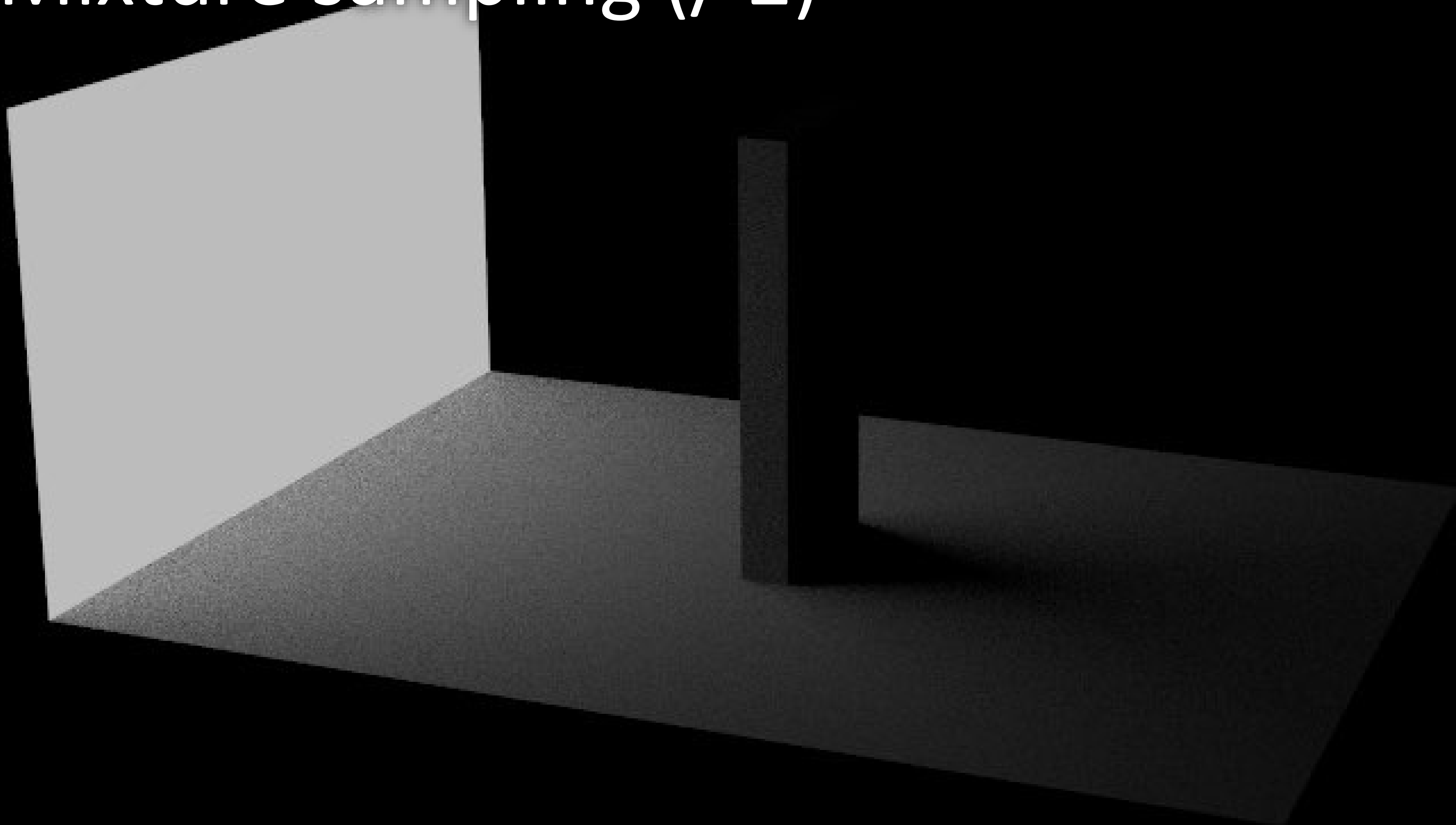
# Cosine-weighted sampling ( $/ 2$ )

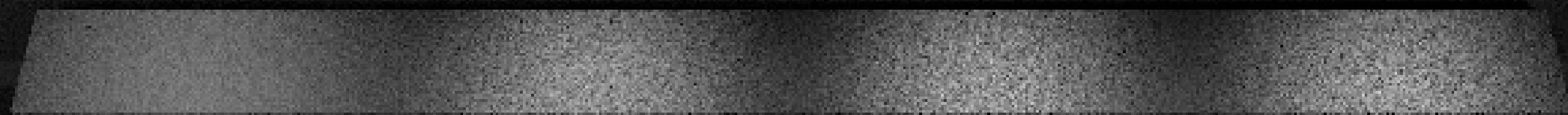
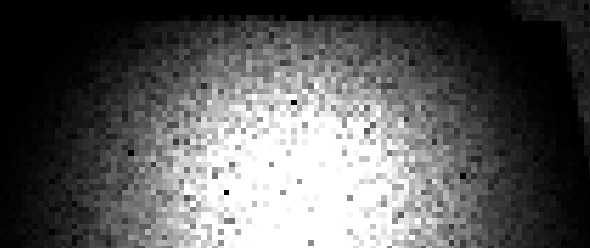
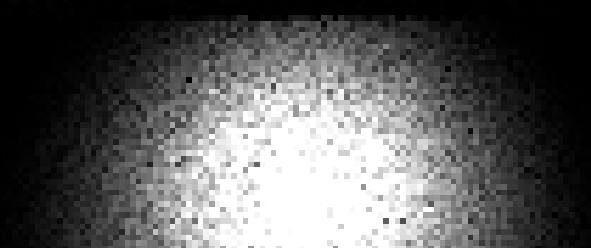
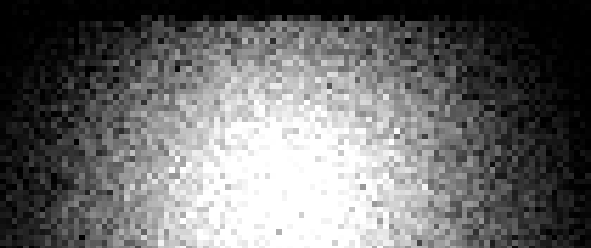
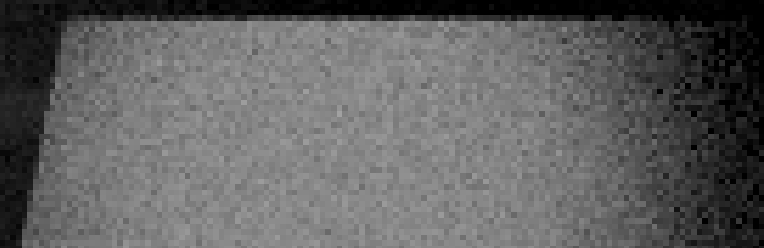
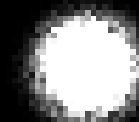
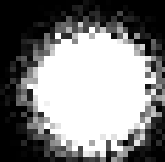
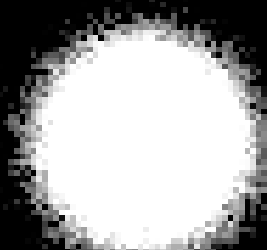
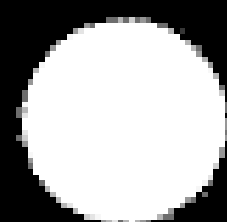
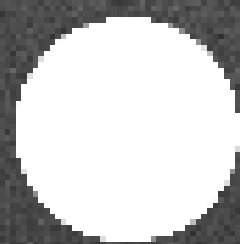
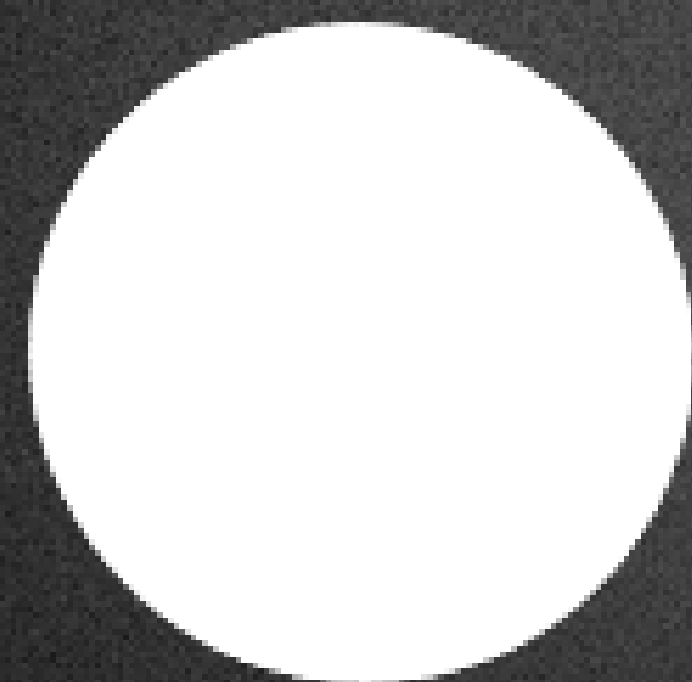


Uniform surface area ( $/ 2$ )



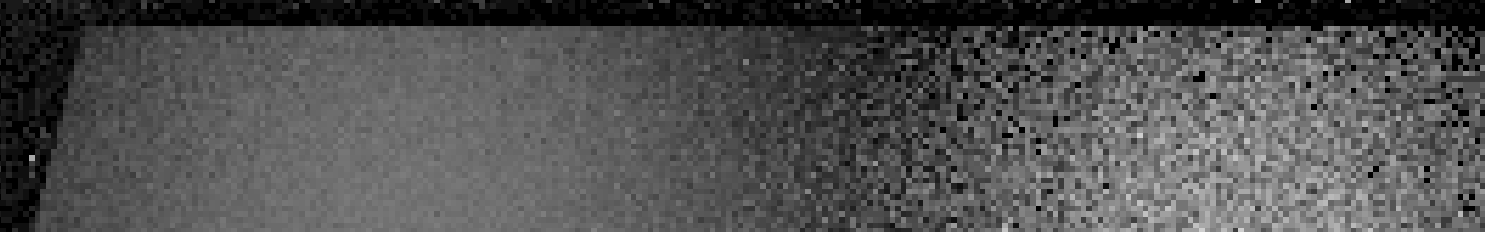
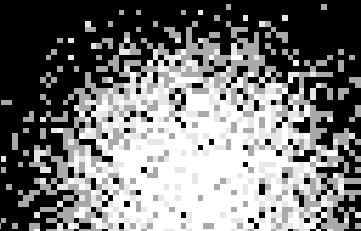
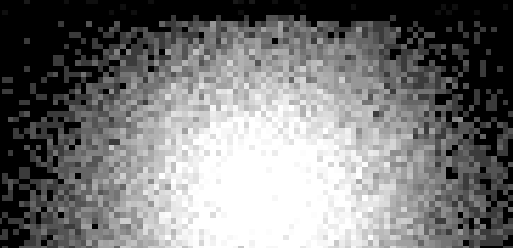
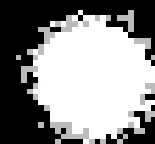
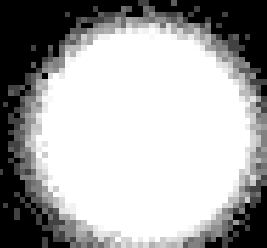
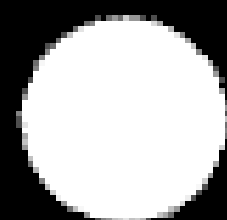
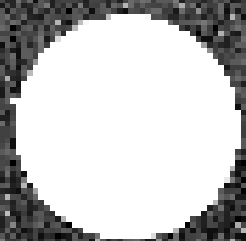
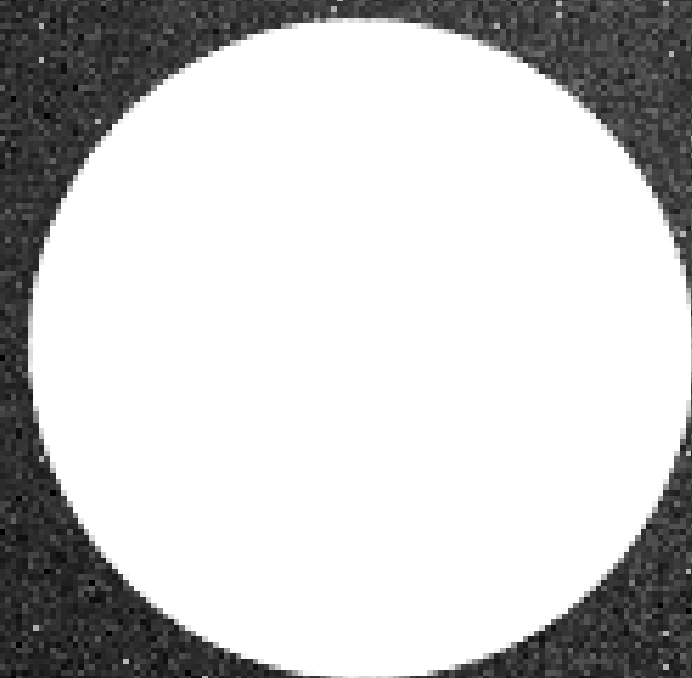
# Mixture sampling (/ 2)



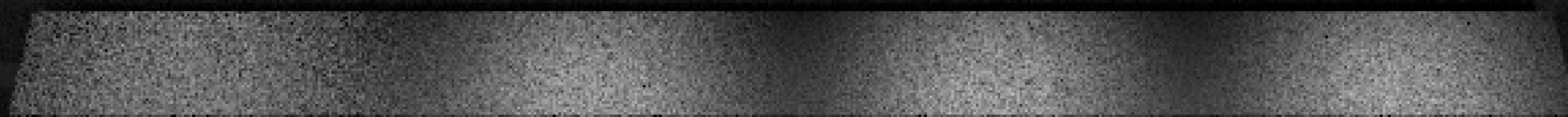
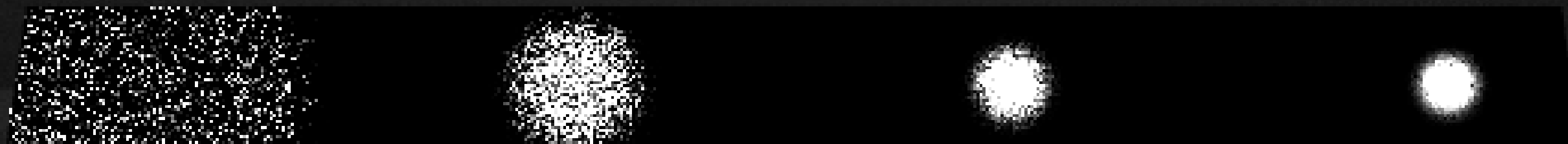
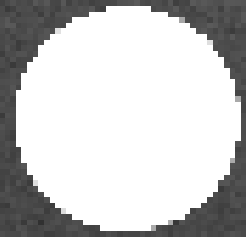
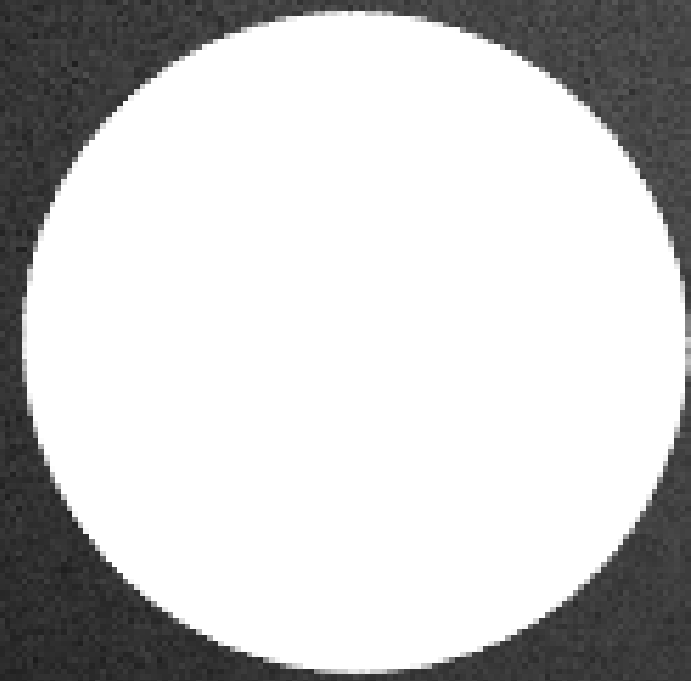




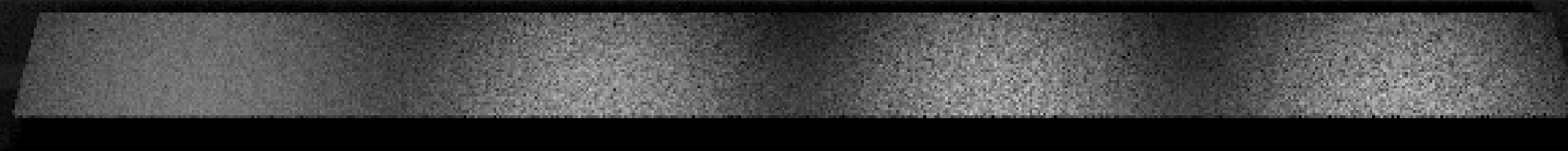
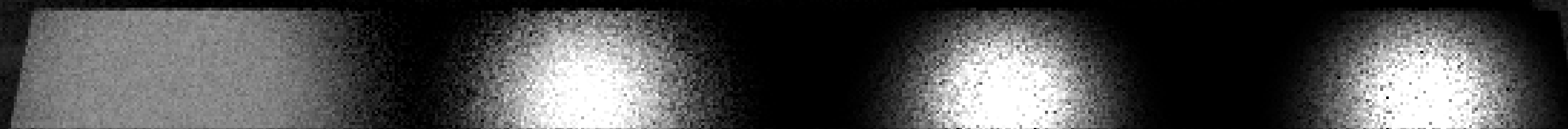
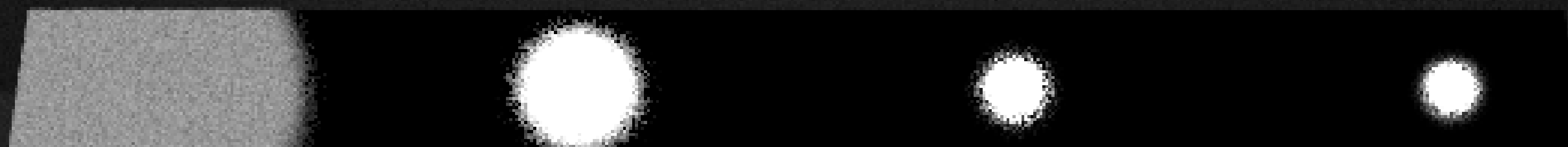
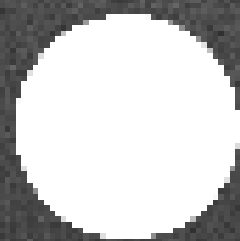
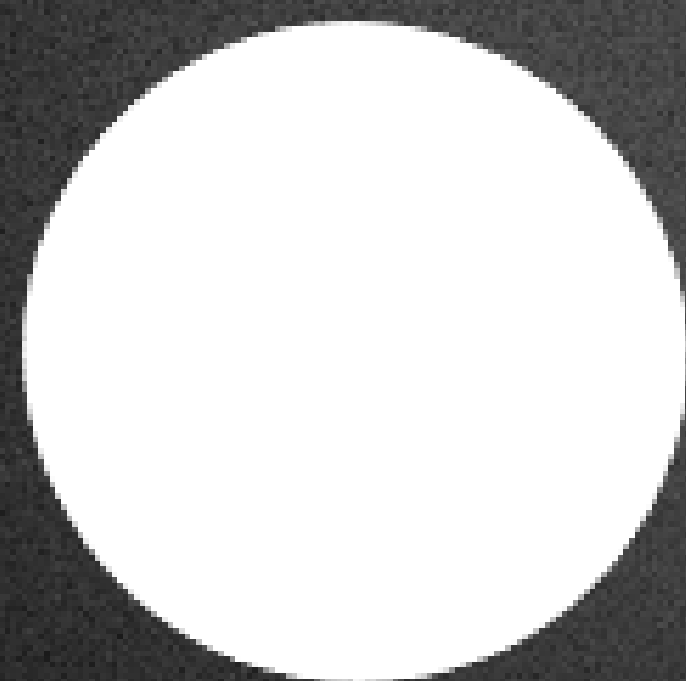
# BSDF sampling



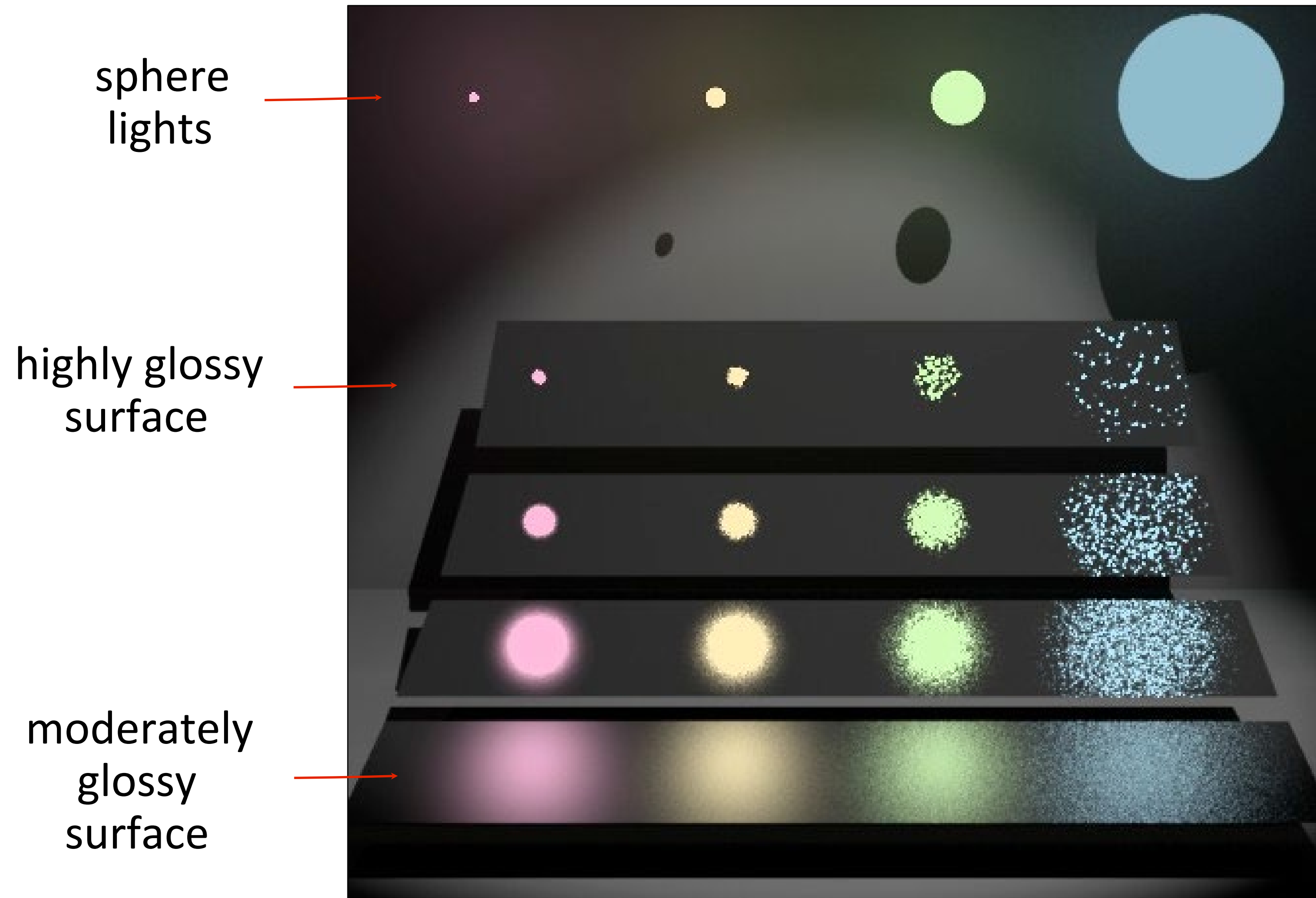
# Light sampling



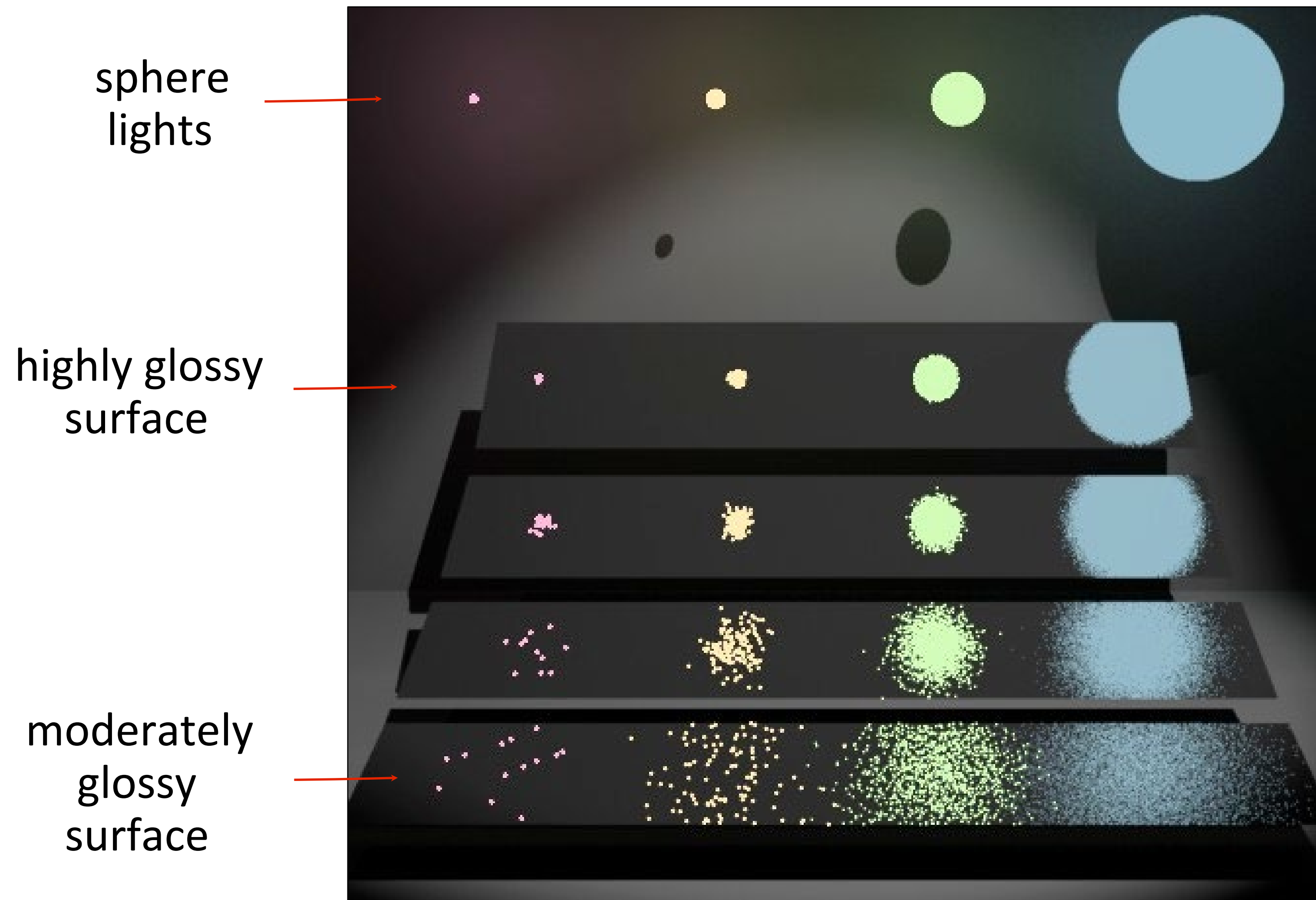
# Mixture sampling



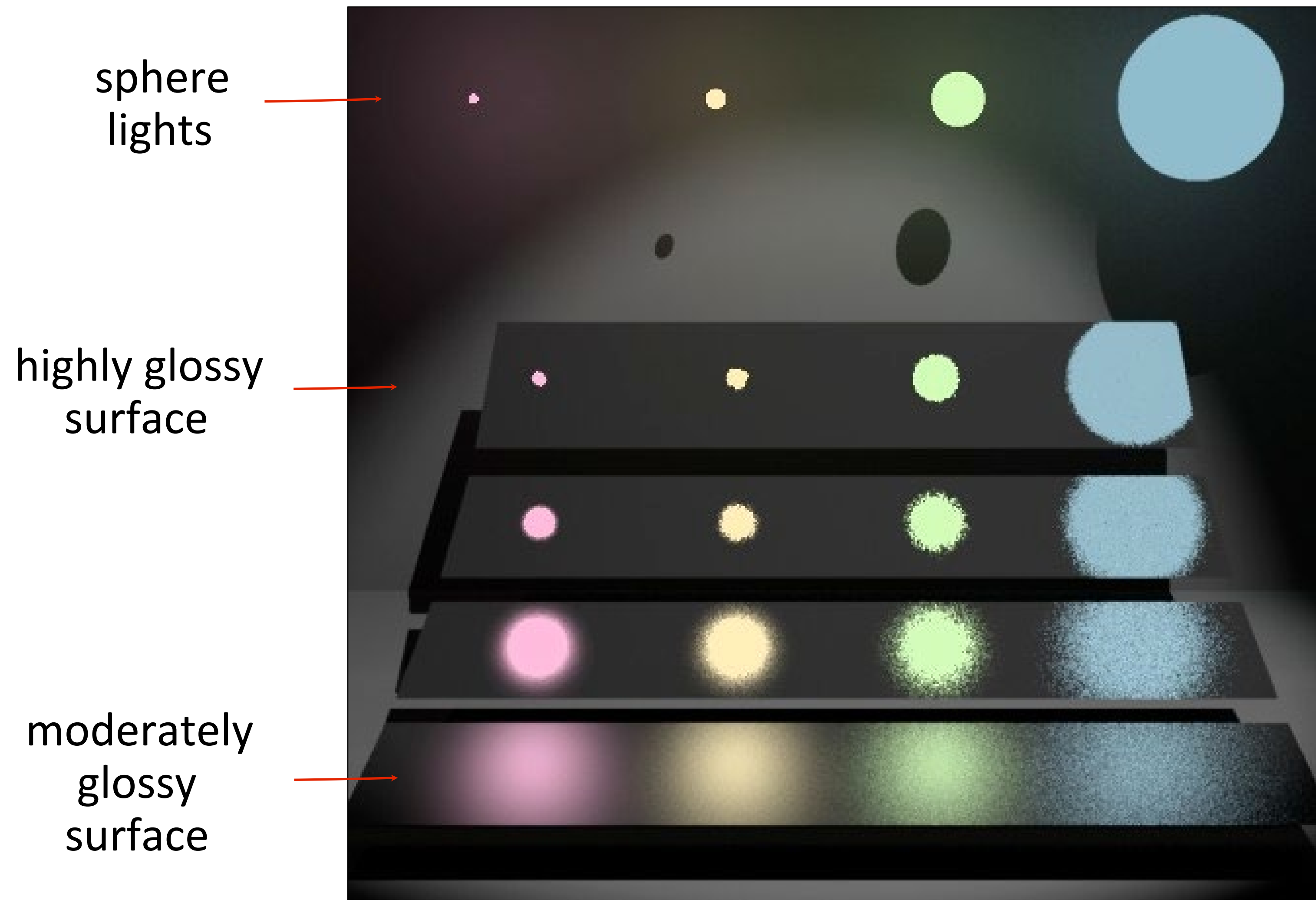
# Sampling the Light



# Sampling the BRDF



# Multiple Importance Sampling



# Multiple Importance Sampling

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See PBRe3 13.10.1 for more details