

ECSE 446/546

IMAGE SYNTHESIS

Direct Illumination II



image credit: [feelgrafix](#)

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derek@cim.mcgill.ca

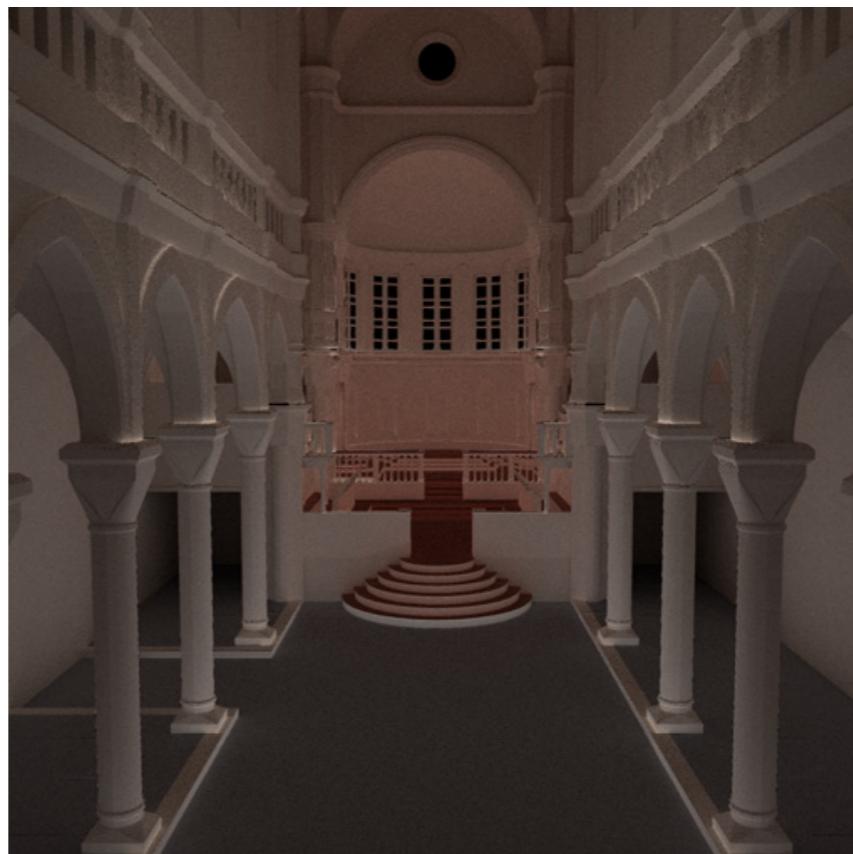
(with some slides adapted from Drs. Novák, Jarosz, and Debevec)

Direct vs. Indirect Illumination

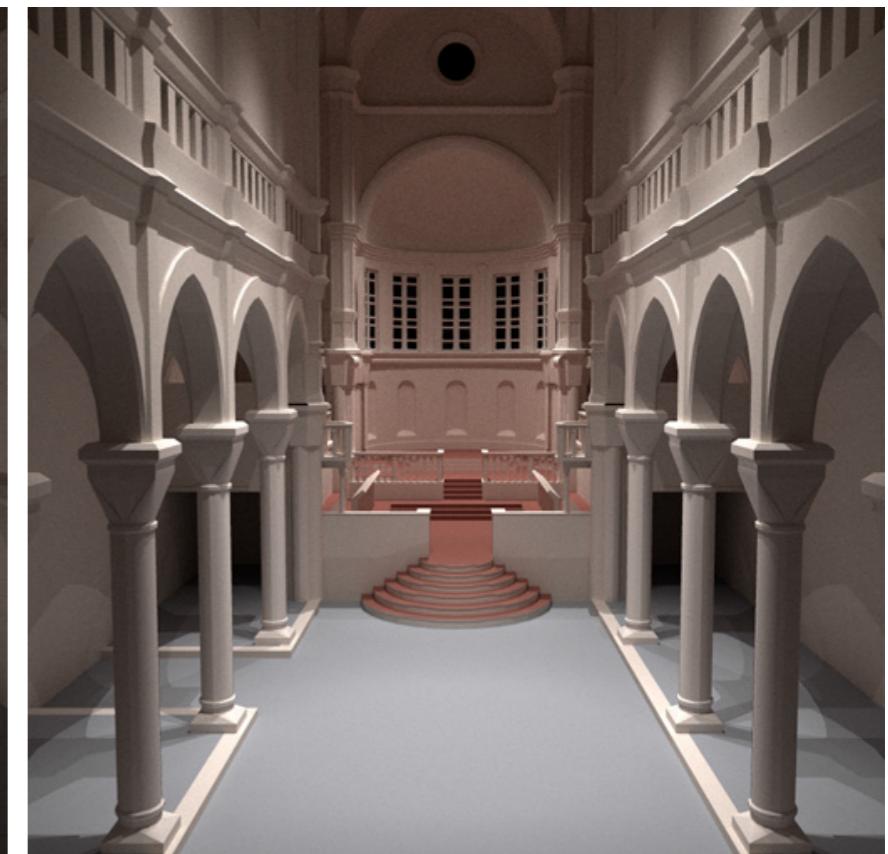
Direct illumination



Indirect illumination

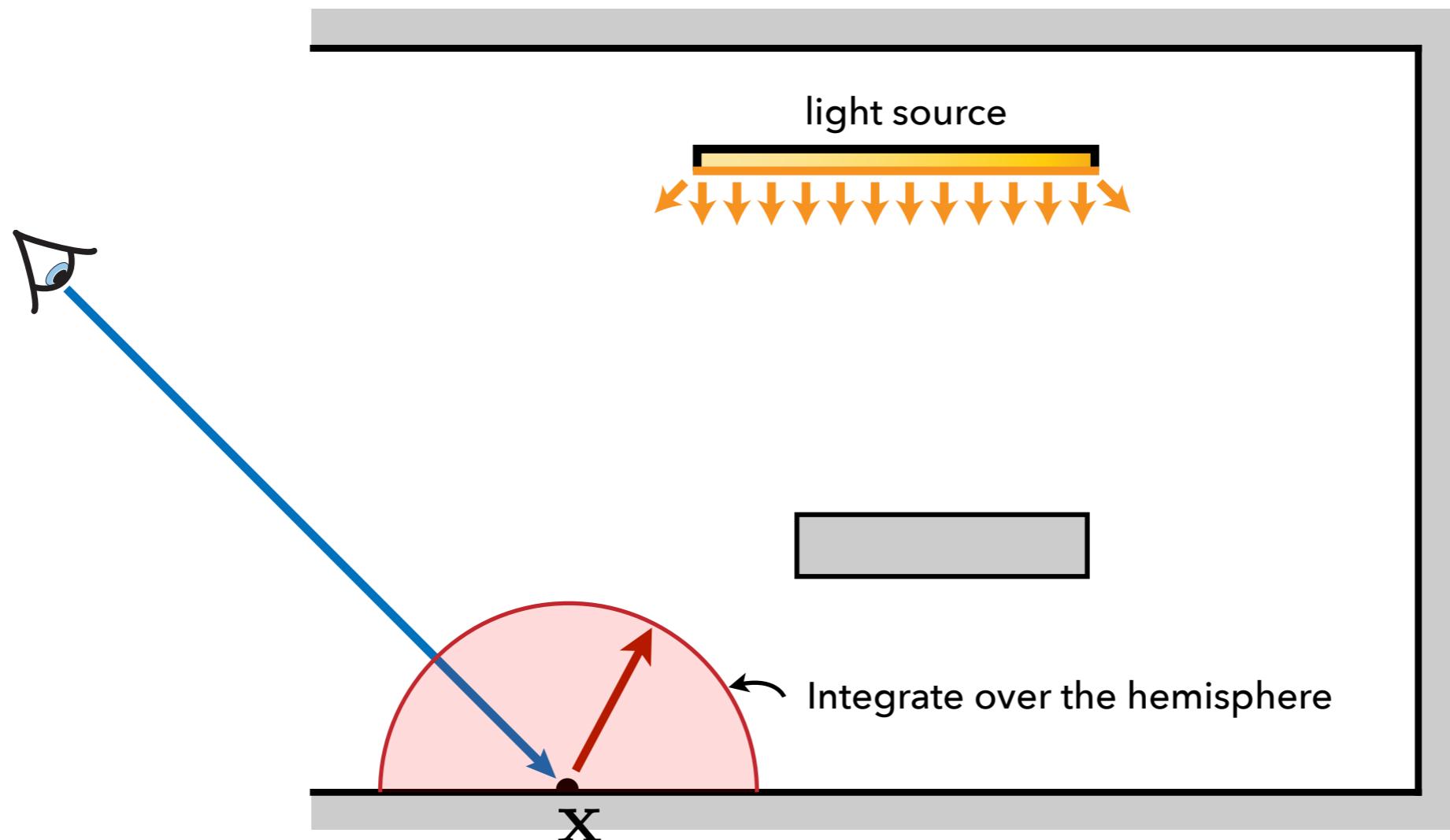


Direct + indirect
illumination



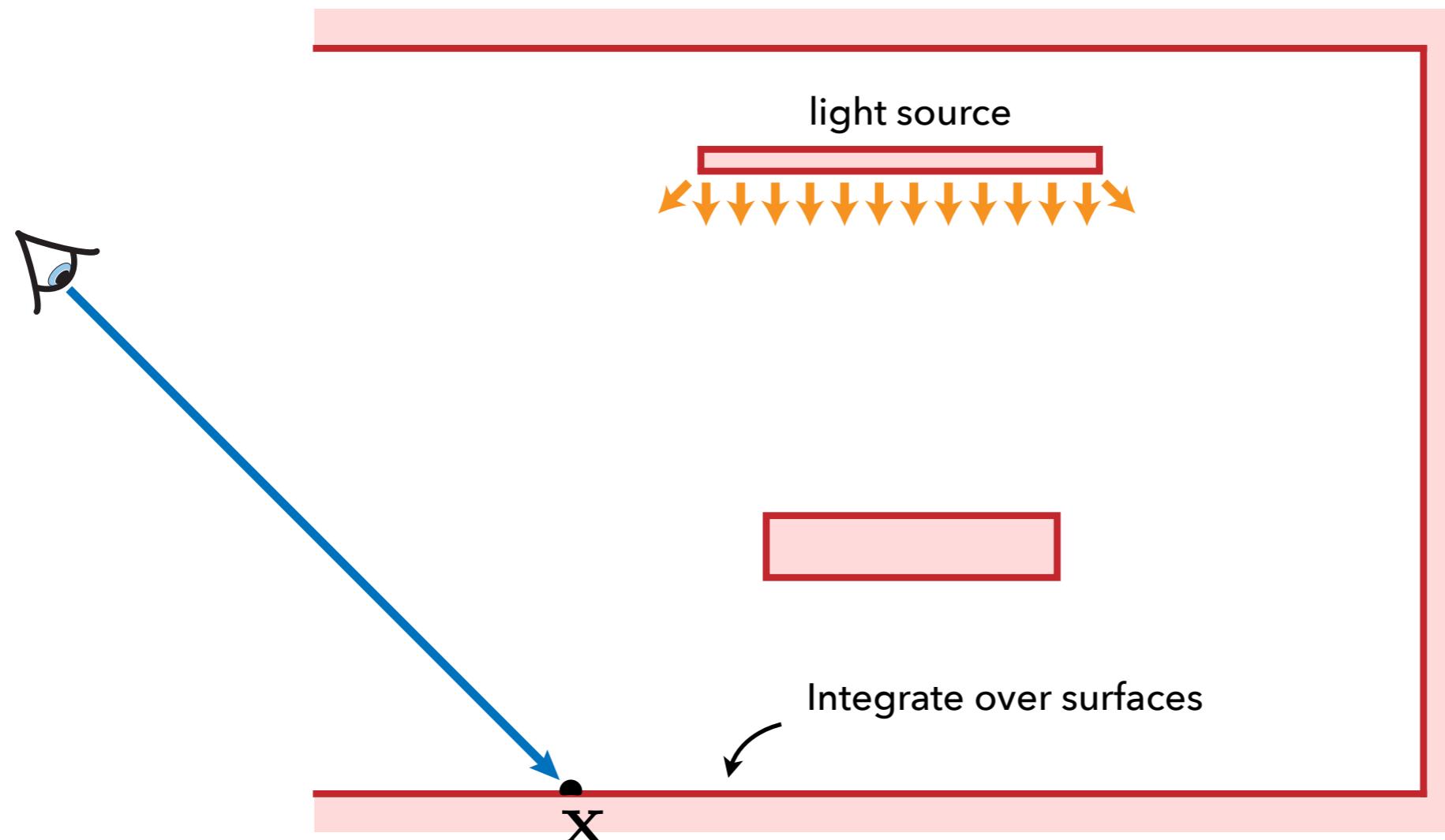
Reflection Eq. – Hemispherical Form

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



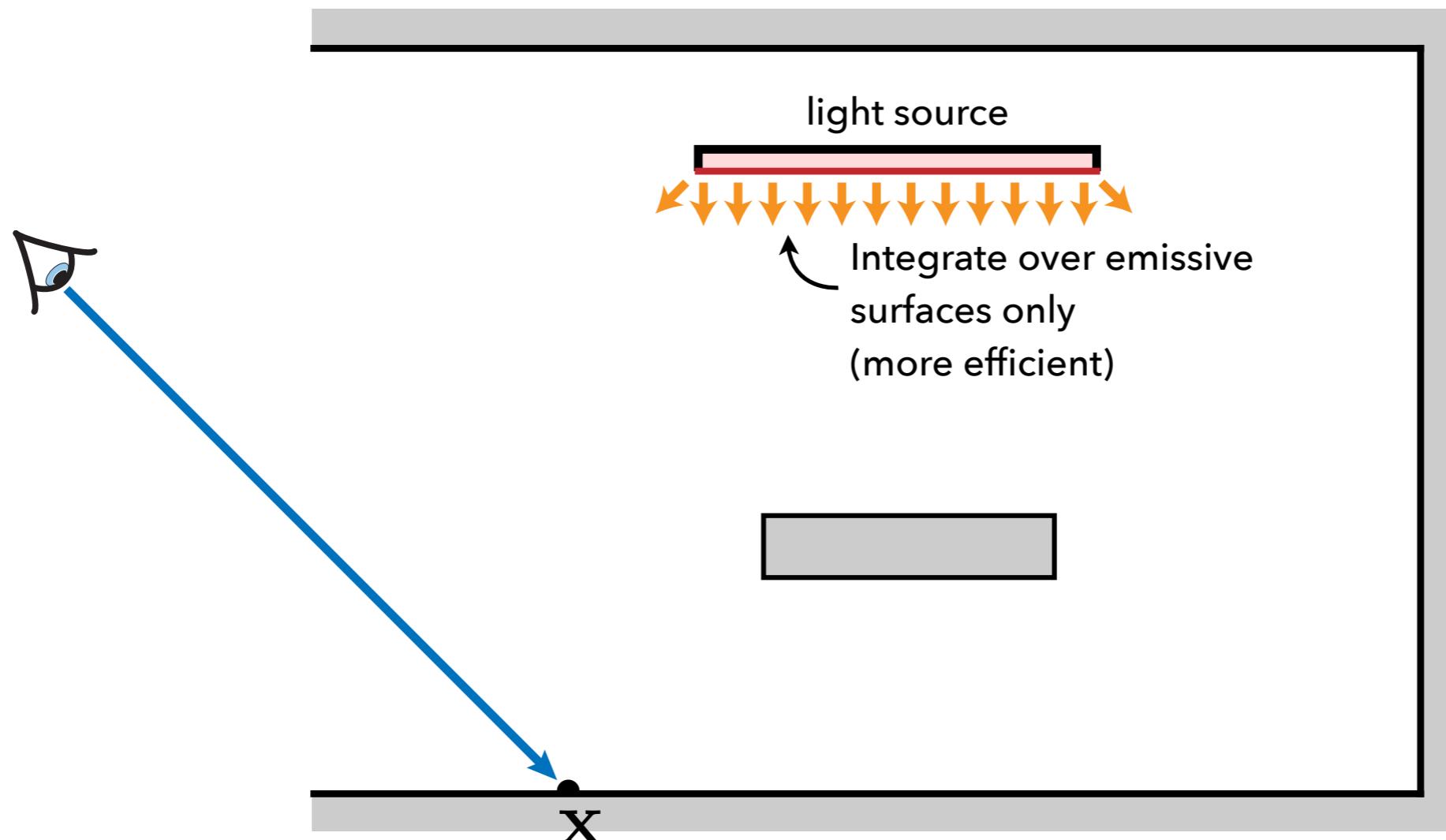
Reflection Eq. – Surface Area Form

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



Reflection Eq. – Surface Area Form

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



Light Sources

Point
light



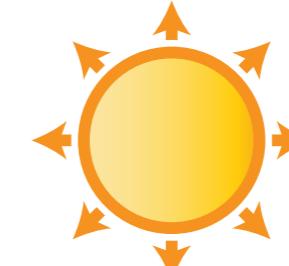
Directional
light



Quad
light



Sphere
light



Mesh
light



Delta lights
(create hard
shadows)

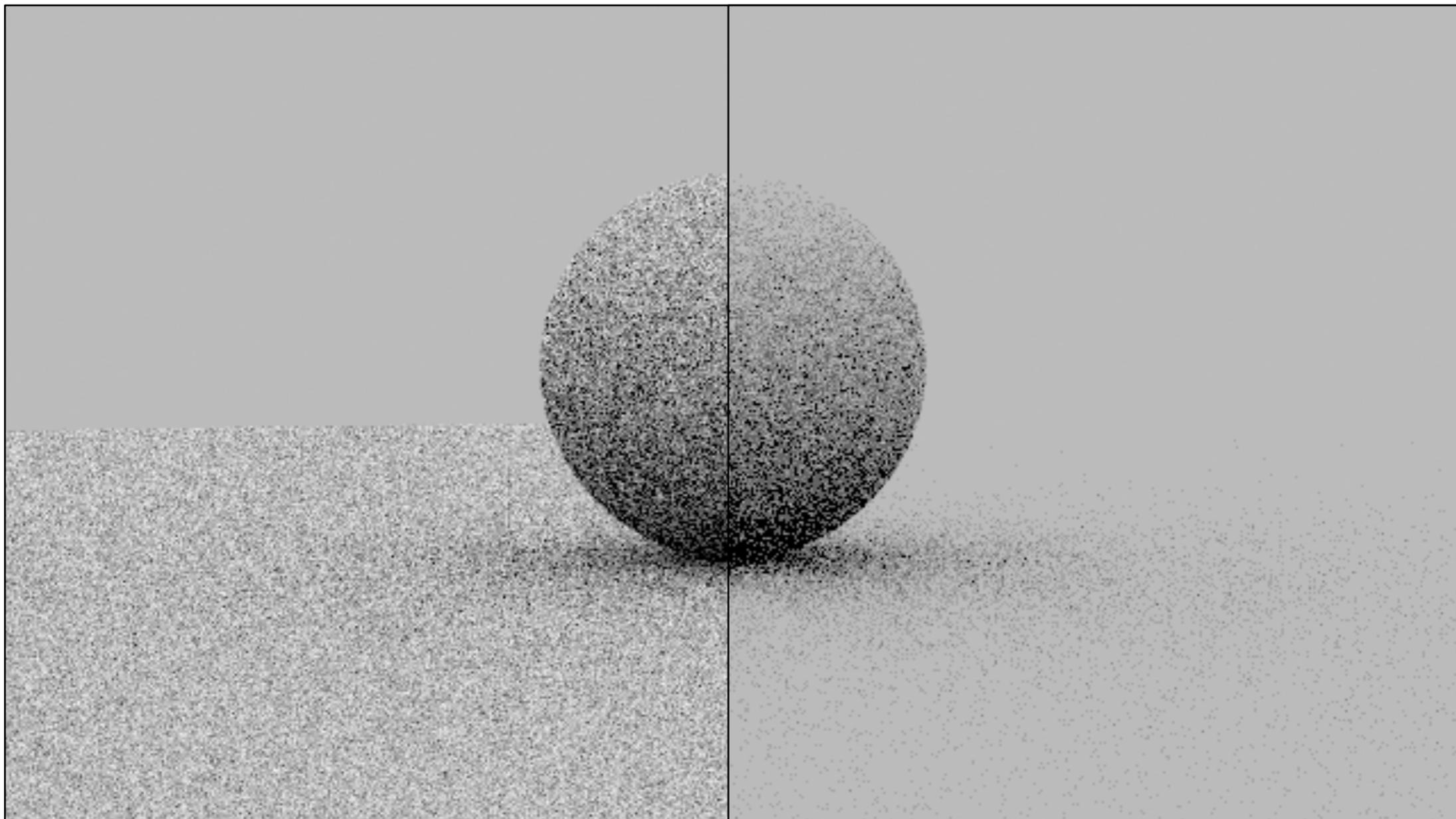
Area/Shape lights
(create soft shadows)

Importance Sampling the Cosine

Uniform hemispherical sampling

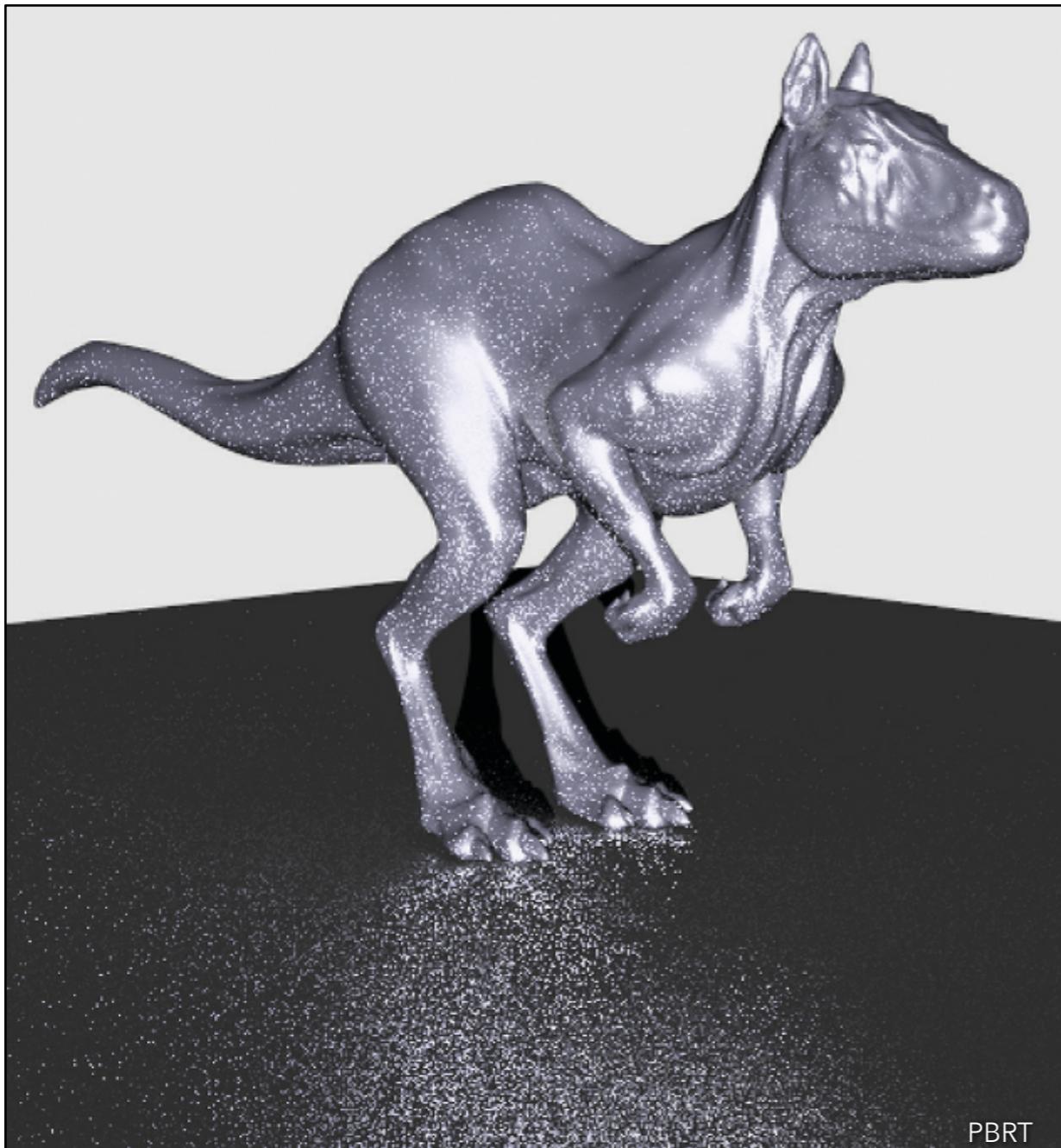
Cosine-weighted importance sampling

4 samples/pixel

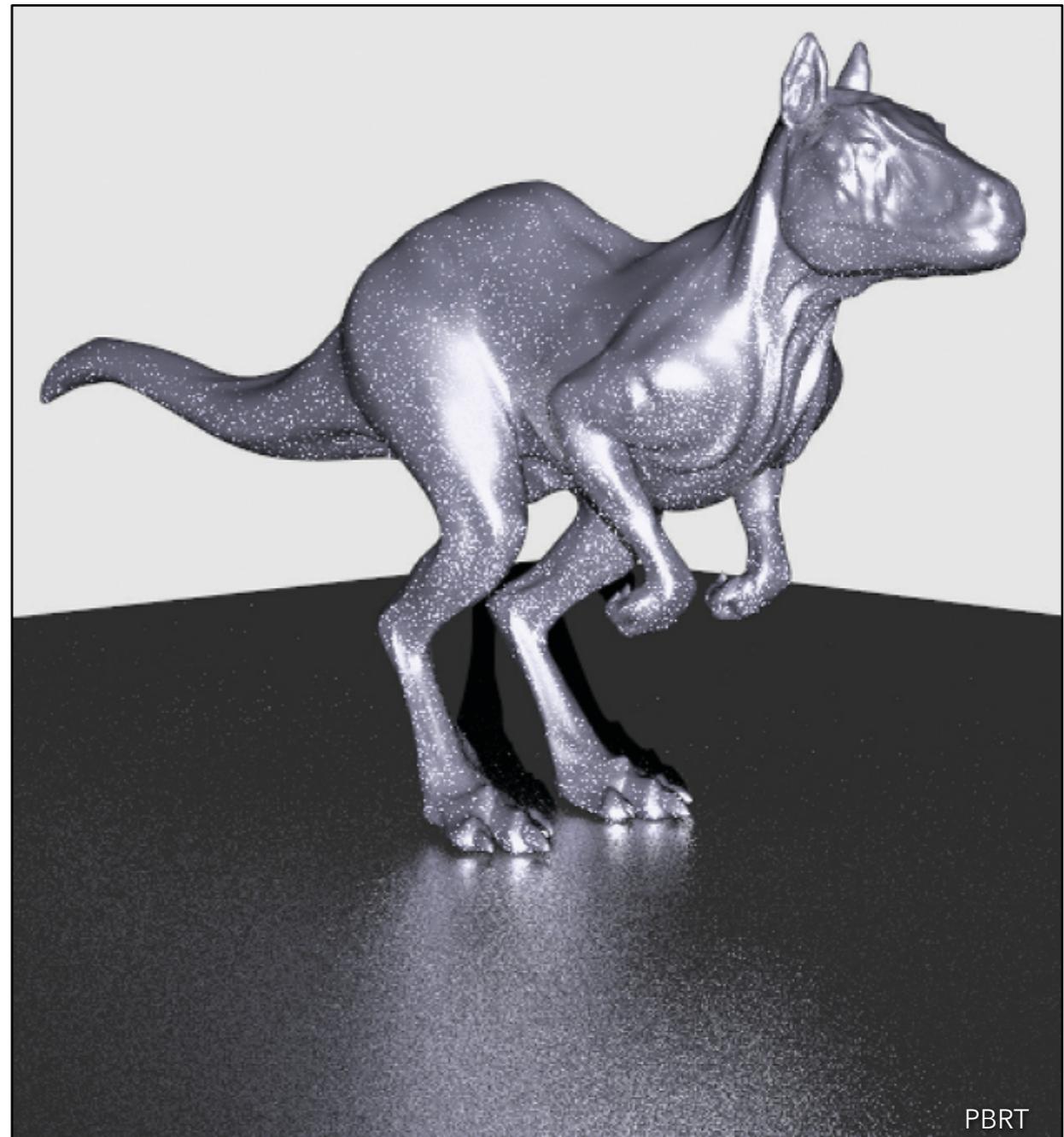


Importance Sampling the BRDF

Uniform hemispherical sampling



BRDF importance sampling

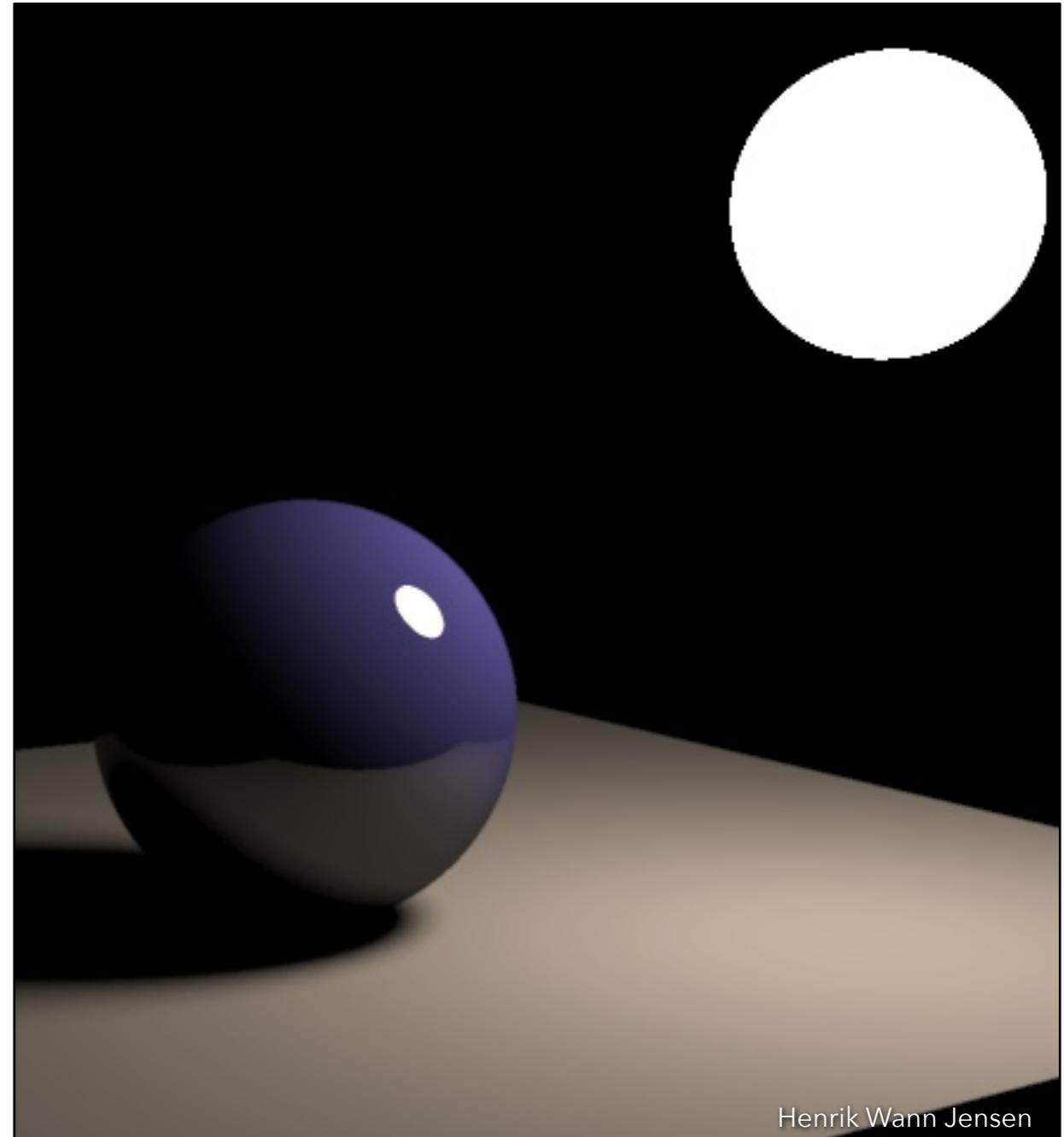


Importance Sampling the Light

Uniform hemispherical sampling



Uniform surface area sampling



Multiple Importance Sampling

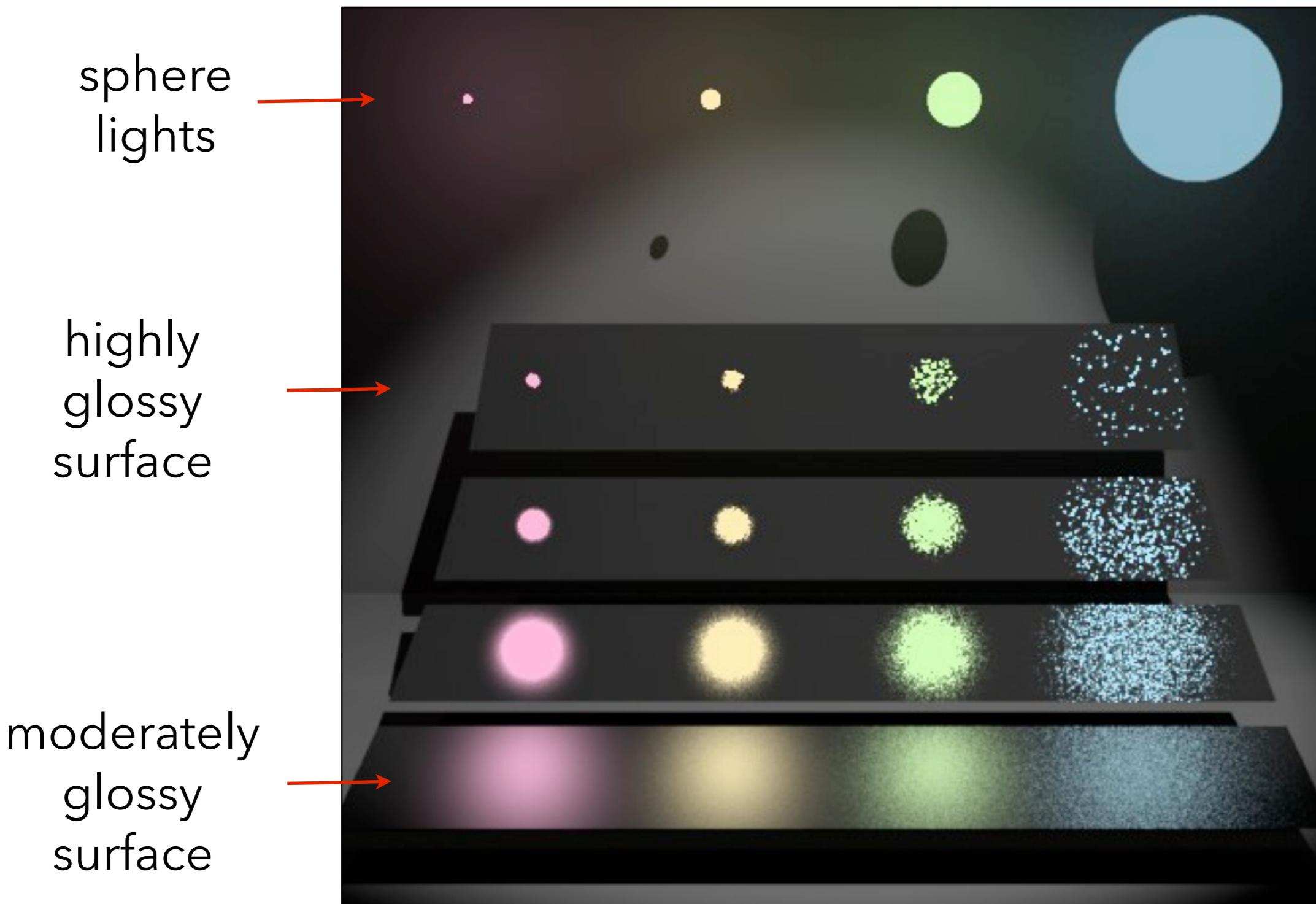
- *Multi-sample* model

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

- *Balance* heuristic:

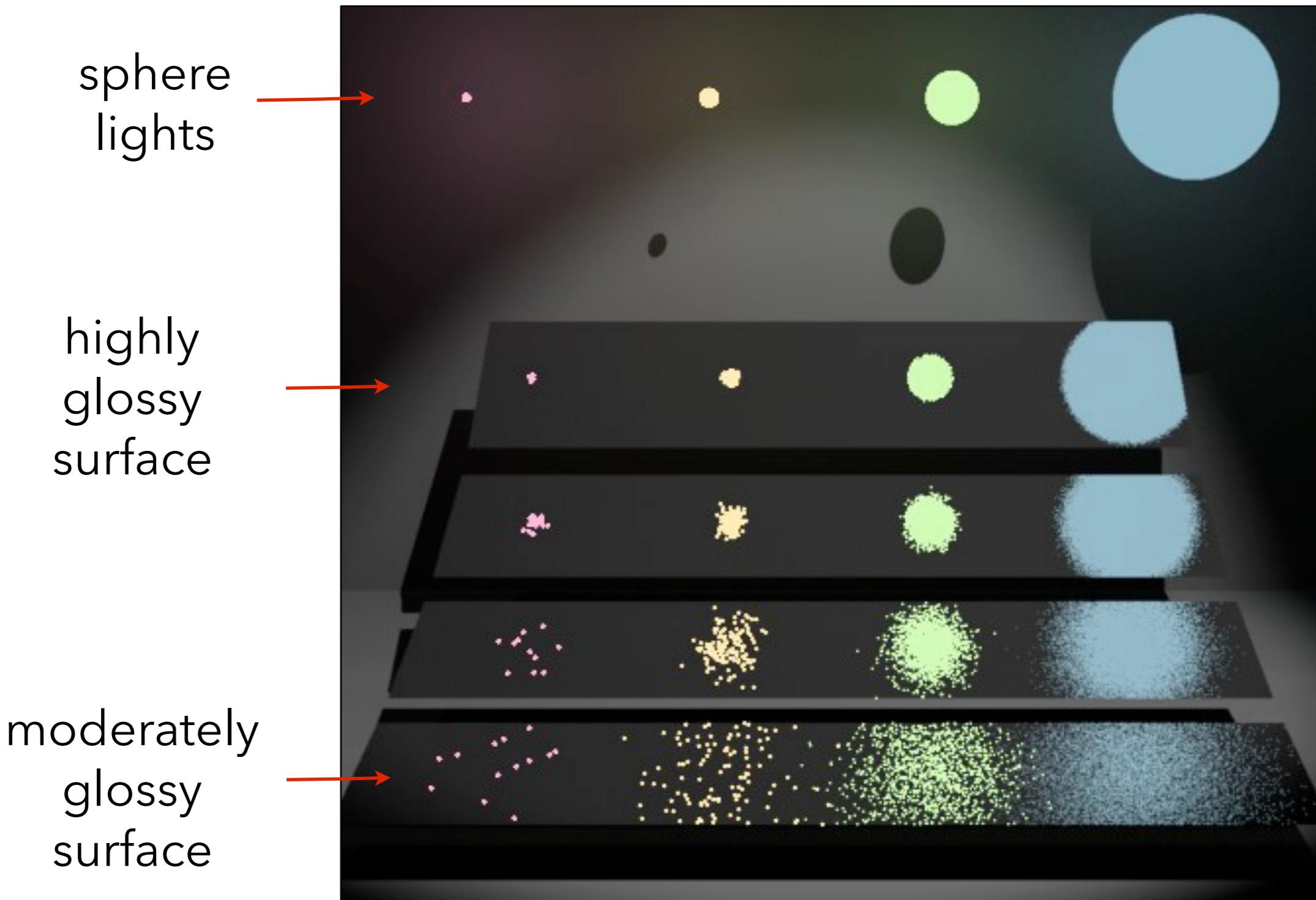
$$w_s(x) = \frac{N_s p_s(x)}{\sum_j N_j p_j(x)}$$

Sampling the Light



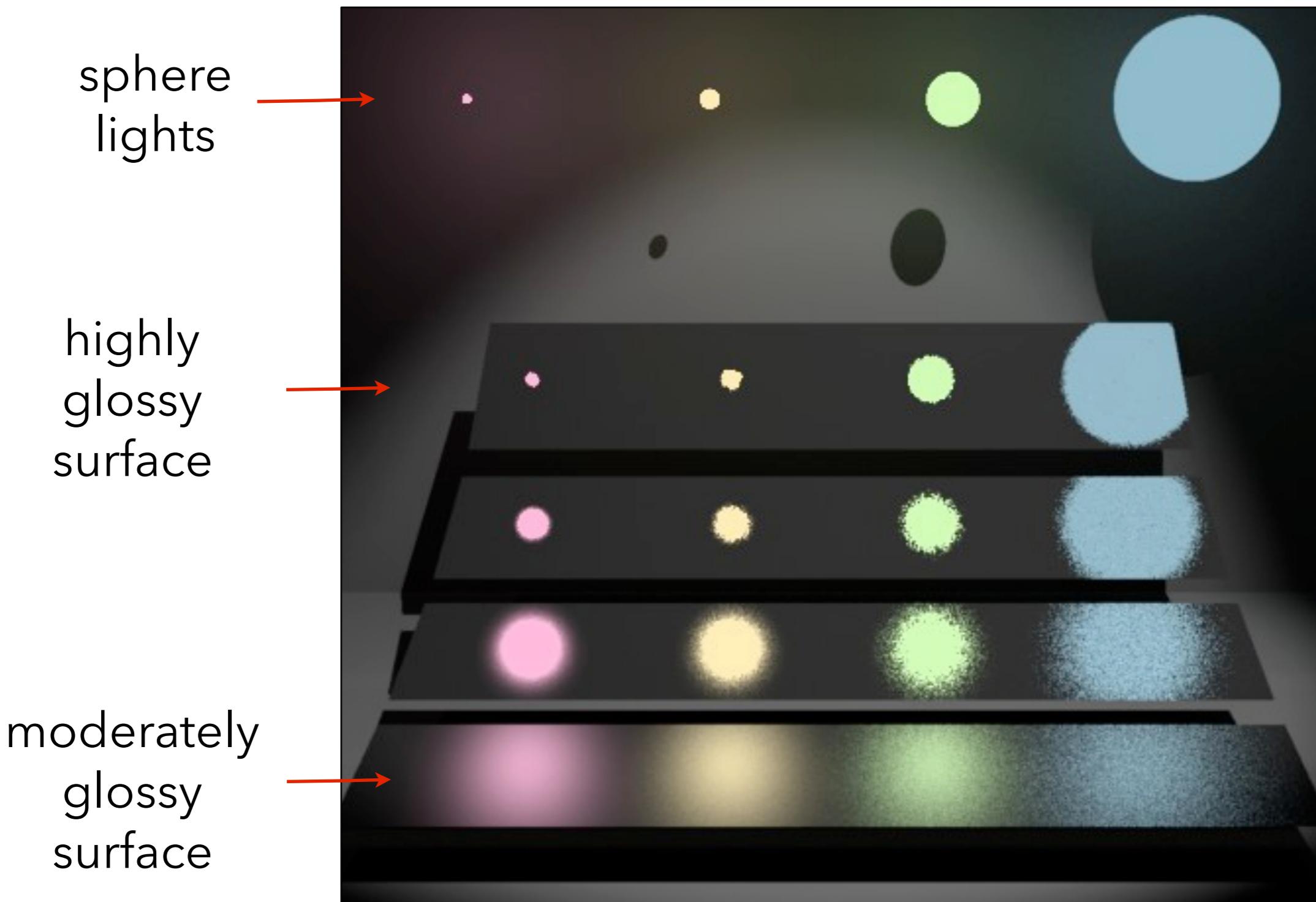
Eric Veach and Leonidas J. Guibas 1995.

Sampling the BRDF



Eric Veach and Leonidas J. Guibas 1995.

Multiple Importance Sampling



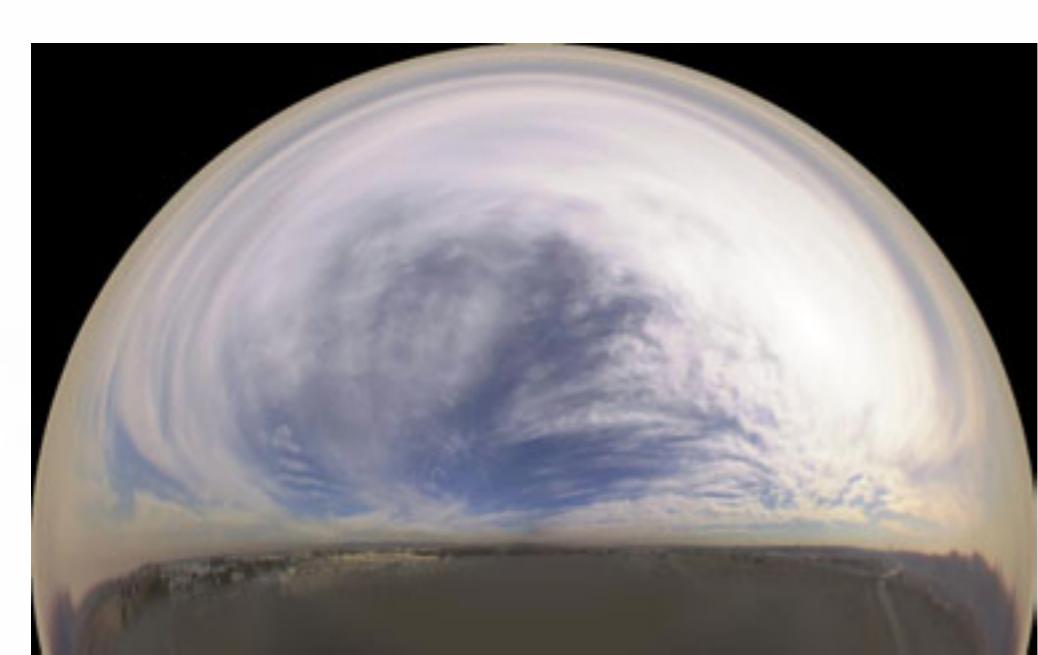
Eric Veach and Leonidas J. Guibas 1995.

Today's Menu

- Environment lighting (image-based lighting)
- Marginal/conditional inversion sampling
- Hierarchical sample warping
- Product sampling

Environment Lighting









Environment Lighting

- Sidesteps tedious modelling of the environment by representing it using one or more *images*
- The image “wraps” around the virtual scene, serving as a *distant source of illumination*

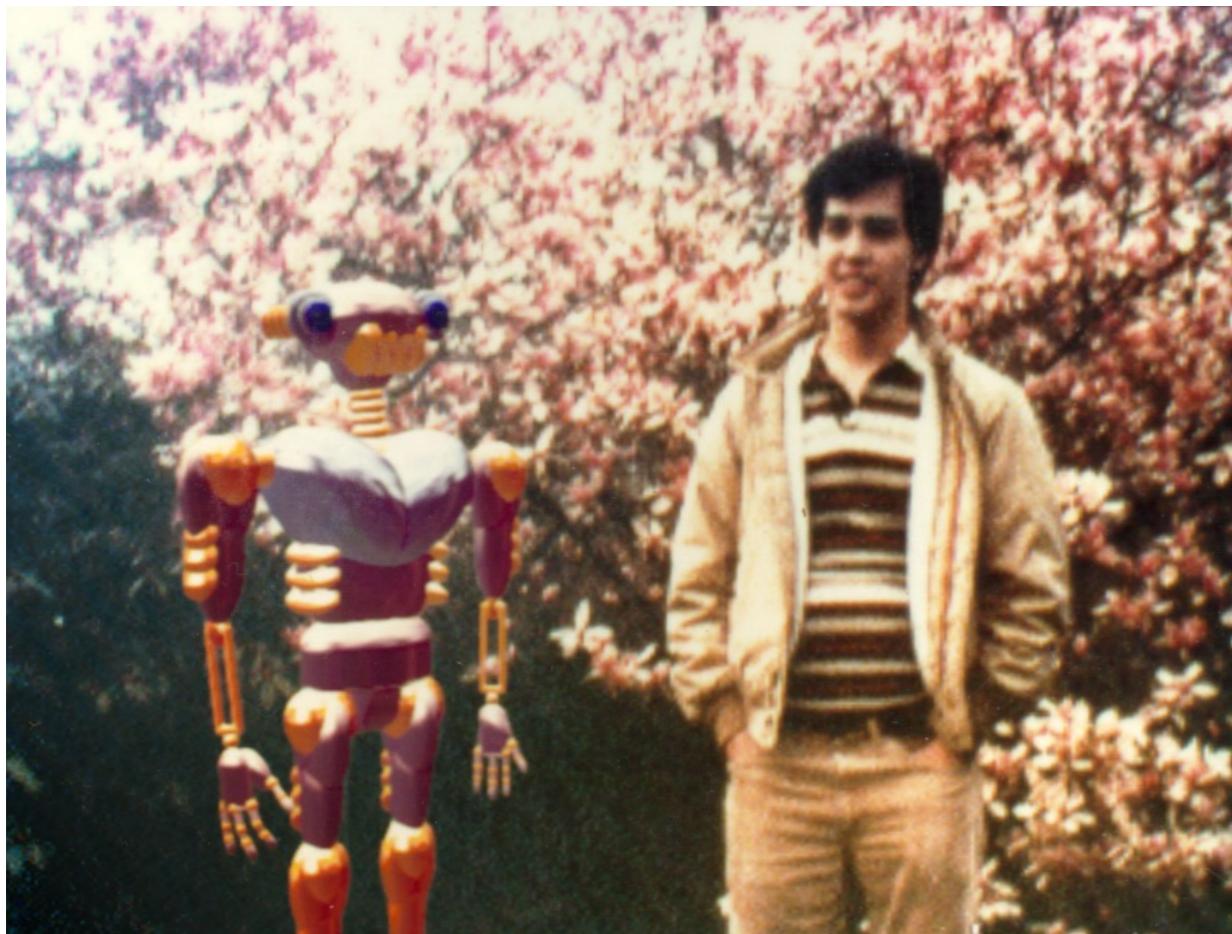


source: [3delight](#)

Related Techniques

- Reflection mapping
- Environment mapping/lighting
- Image-based lighting (IBL)

Reflection Mapping (1982)

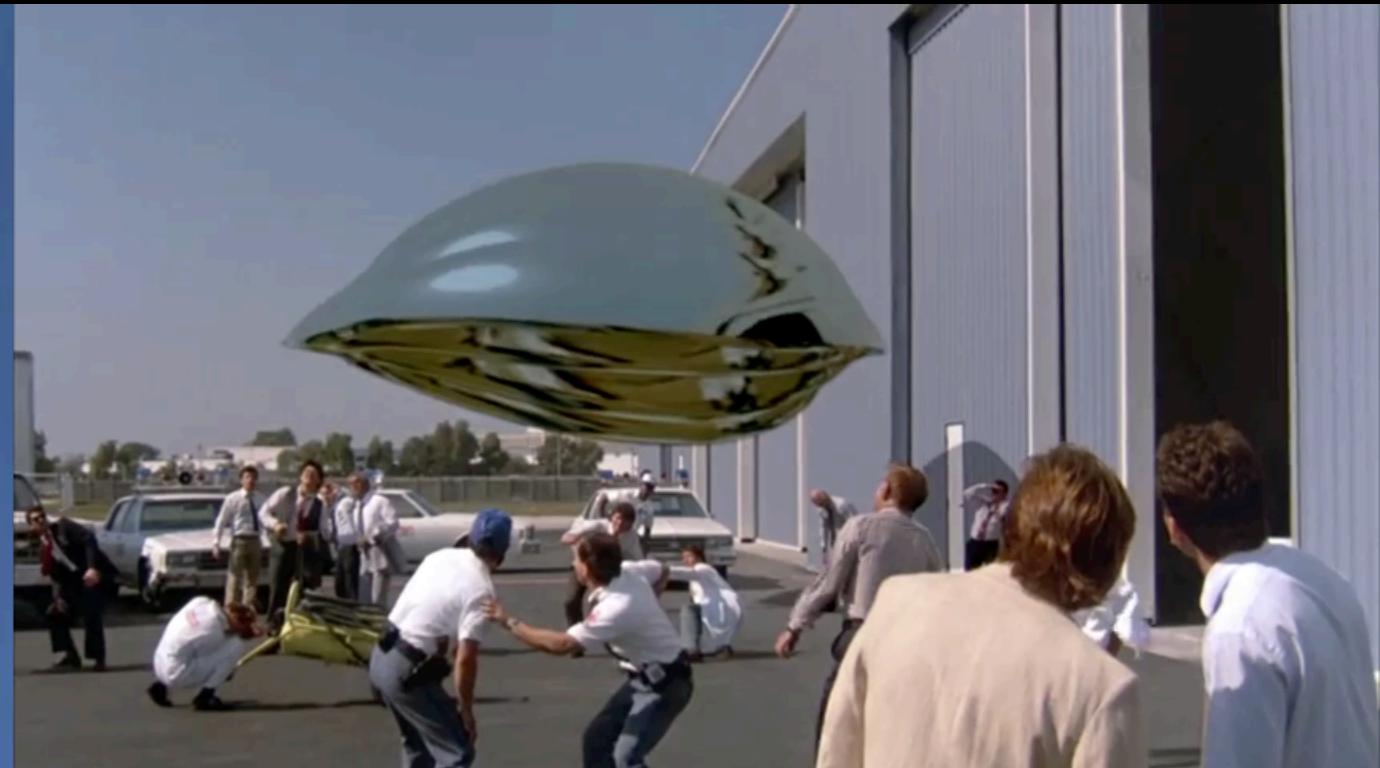


Mike Chou and Lance Williams



Gene Miller and Ken Perlin

Flight of the Navigator (1986)



The Abyss (1989)



Terminator 2 (1991)



Acquiring Environment Lighting

- Mirrored ball + camera
- Fisheye lens images
- Stitching images together
- Panoramic camera

Acquisition – Low Tech.



Maurits C. Escher



Mirror sphere



omnidirectional,
360° panoramic,
HDR image

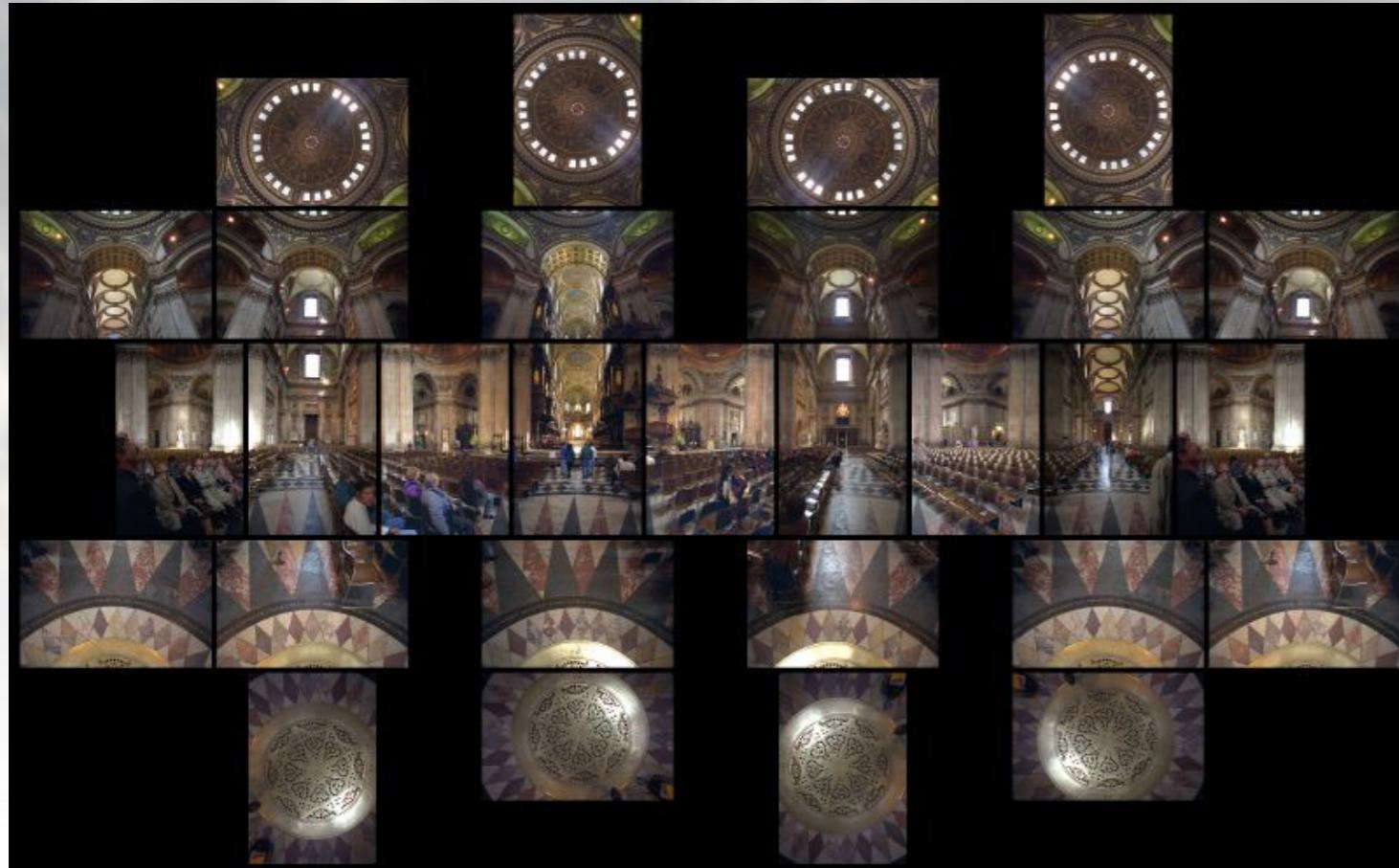
Lightprobe

Acquisition – Even Lower Tech.

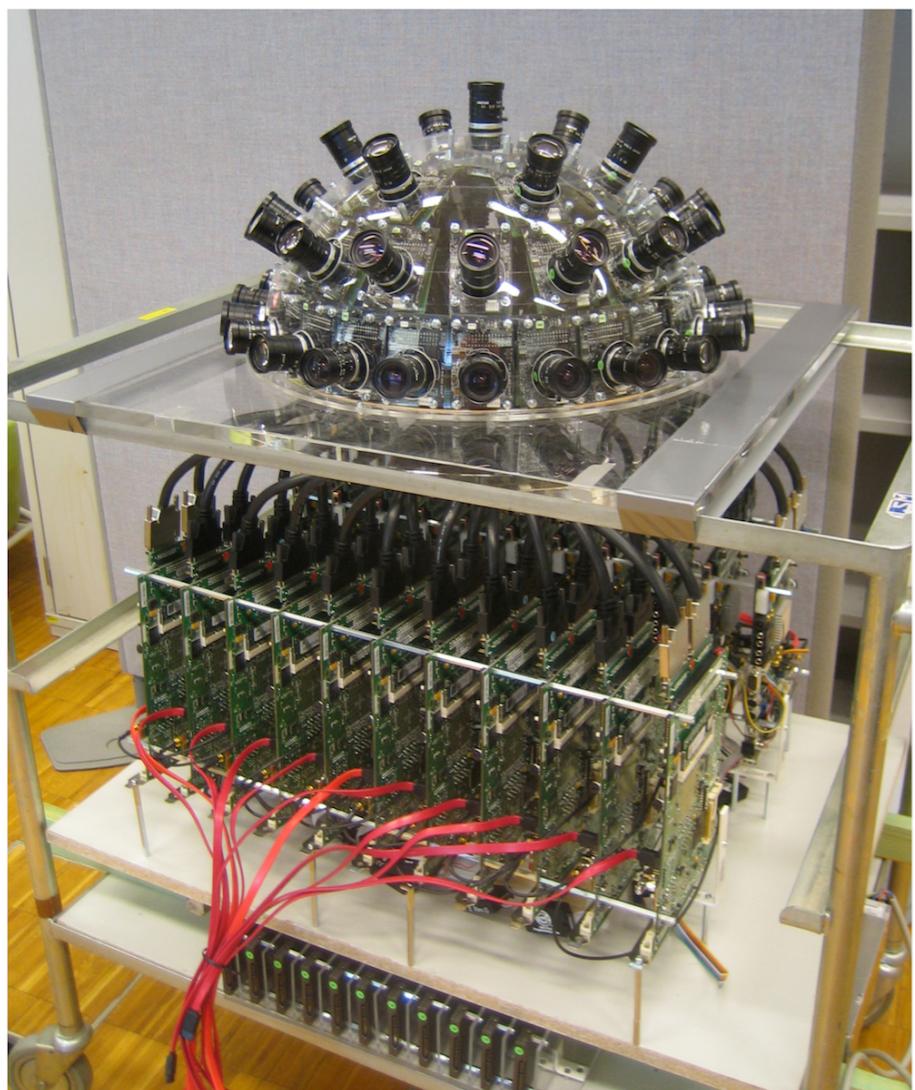


Gene Miller and Ken Perlin

Acquisition – Stitched Panorama



Acquisition – High Tech.



lsm.epfl.ch



immersivemedia.com

Storing Environment Maps

- Various ways to represent/parametrize environment maps
- Related to cartography
 - Projecting the earth (sphere) onto a plane

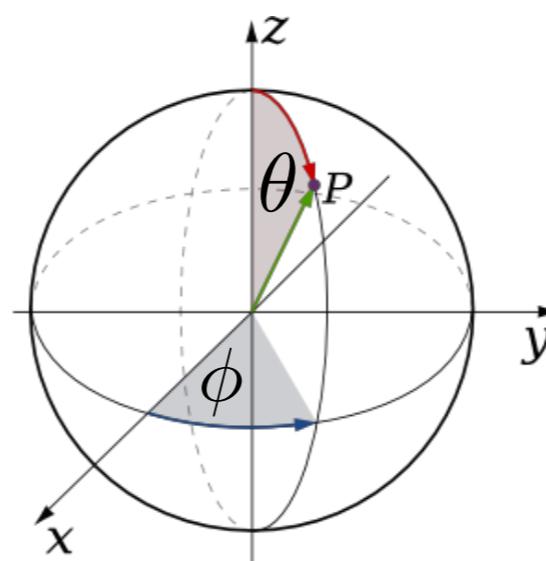
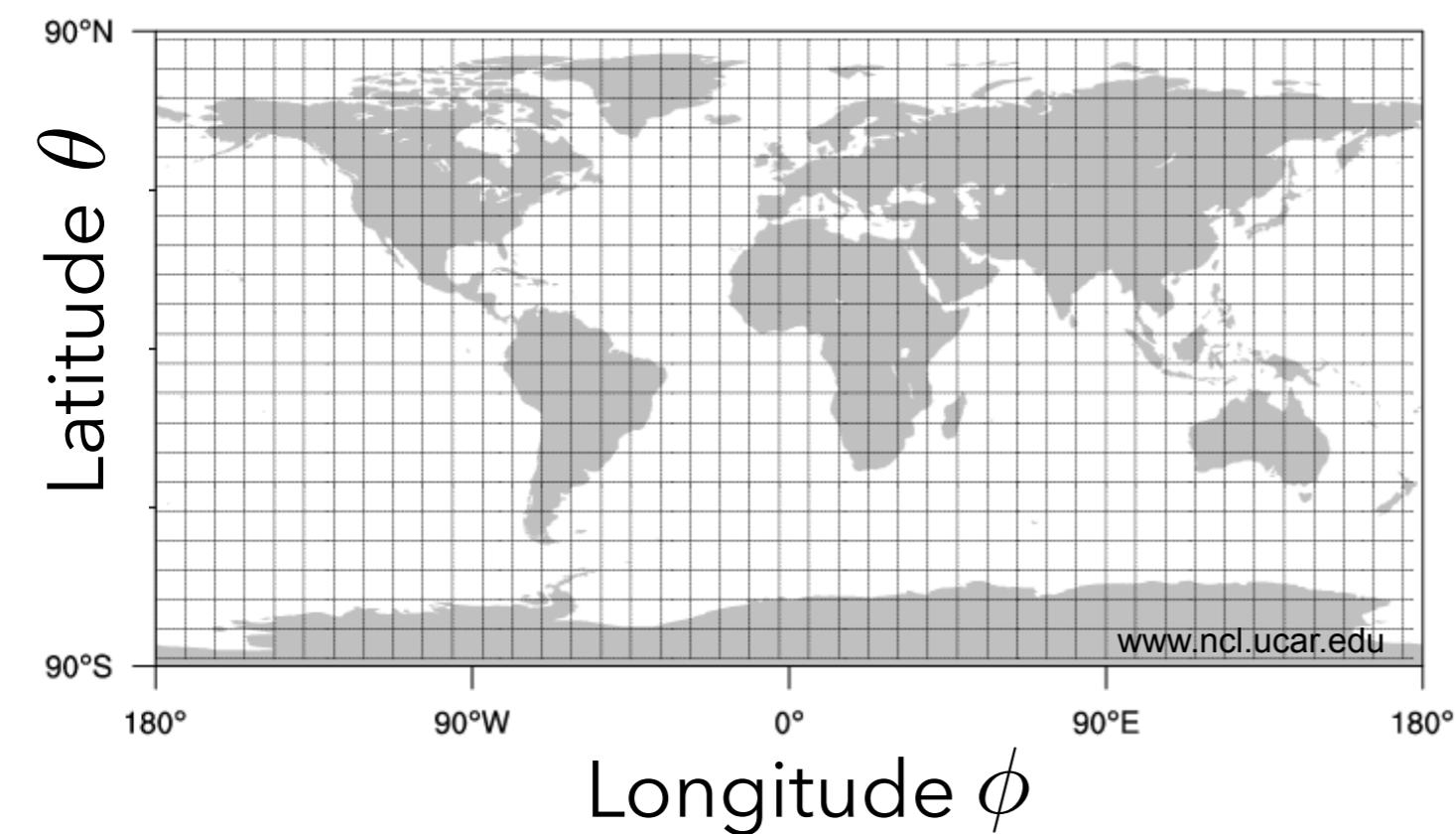
Mirror Ball



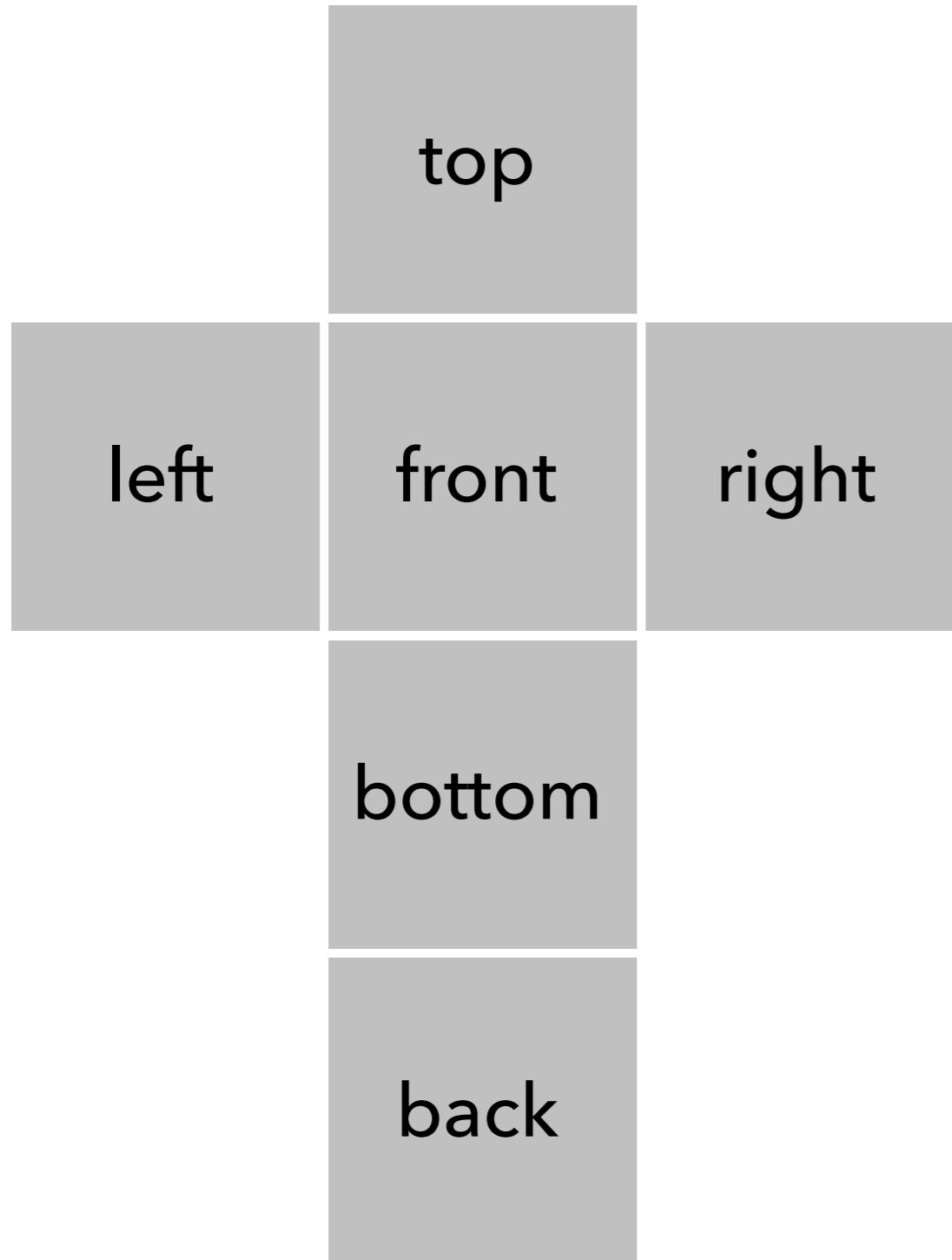
Angular Map



Latitude/Longitude Map

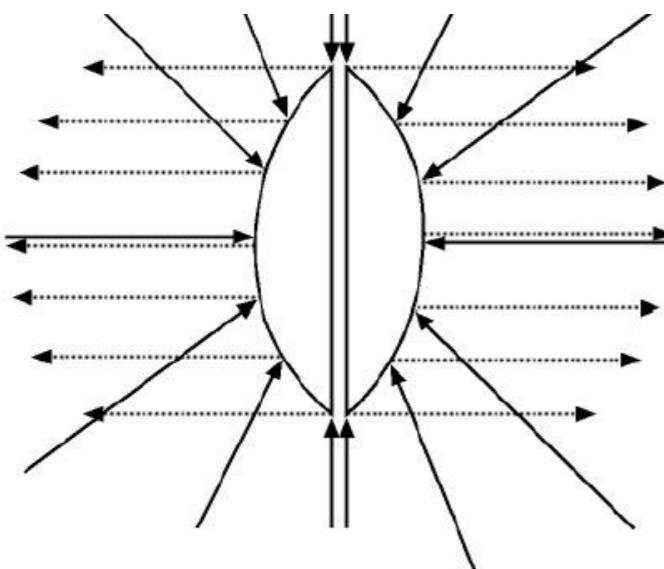
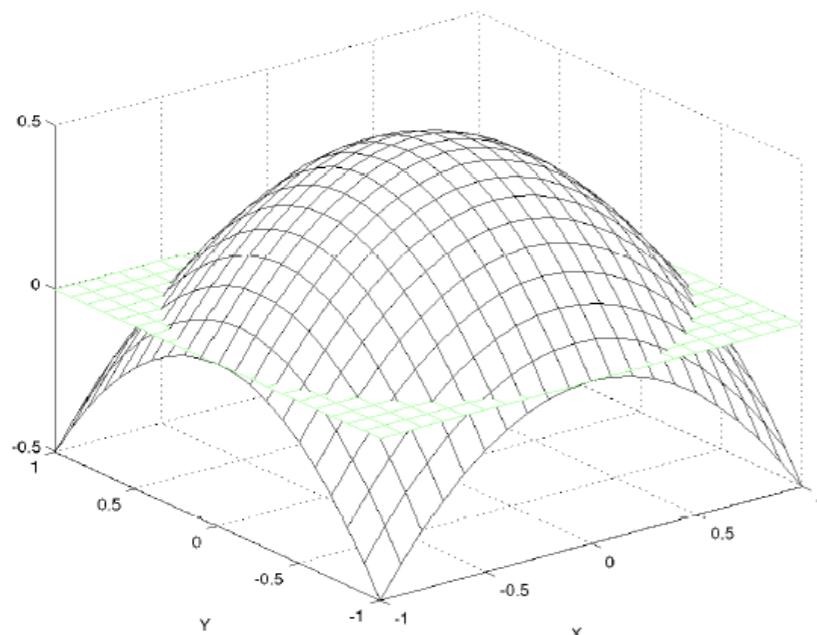


Cube Map (Skybox)



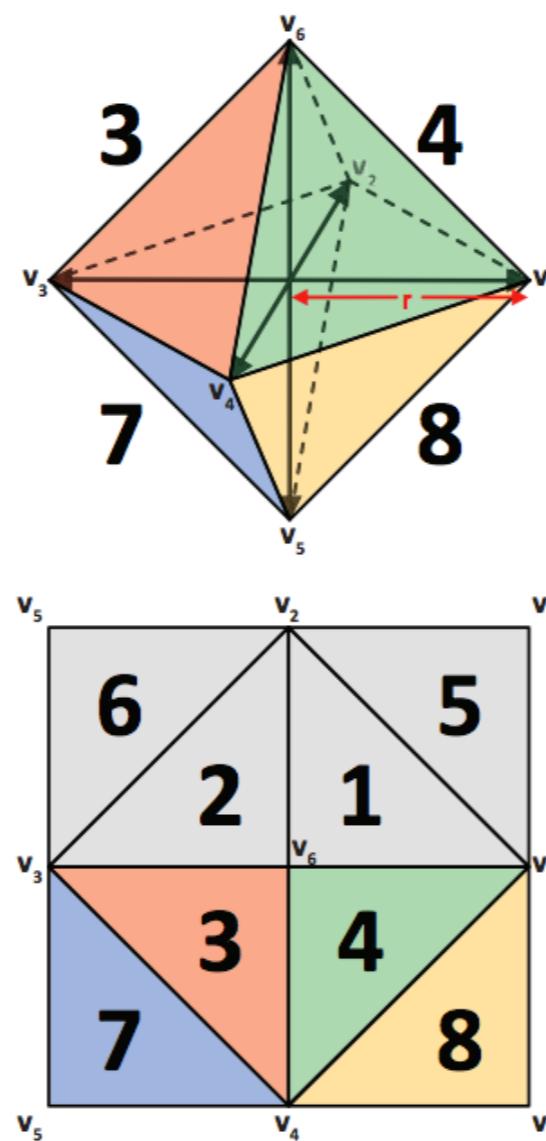
Other Parameterizations

Dual-paraboloid



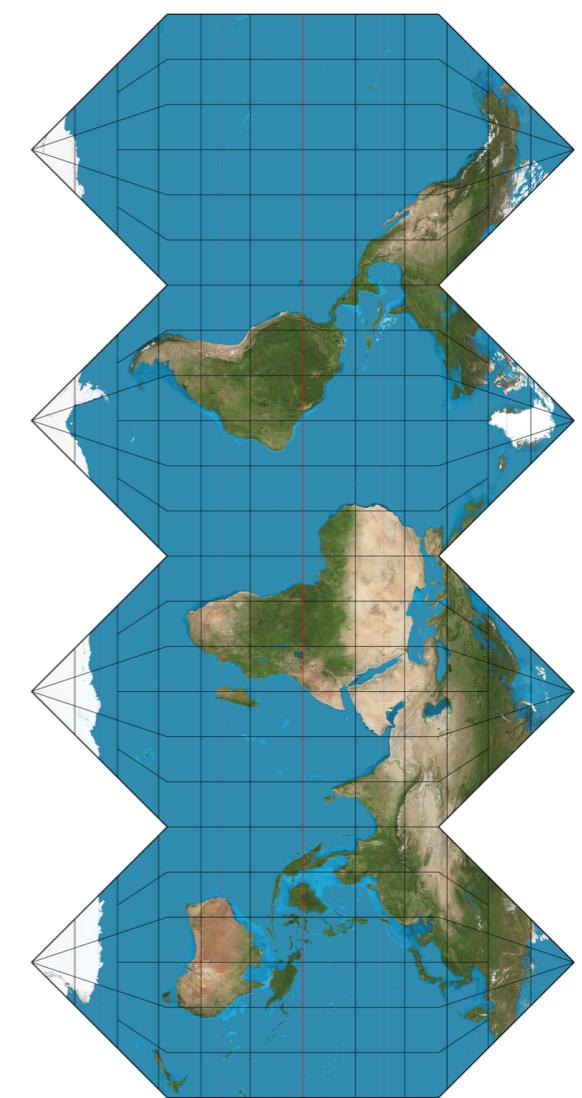
Courtesy of Brabec

Octahedron map



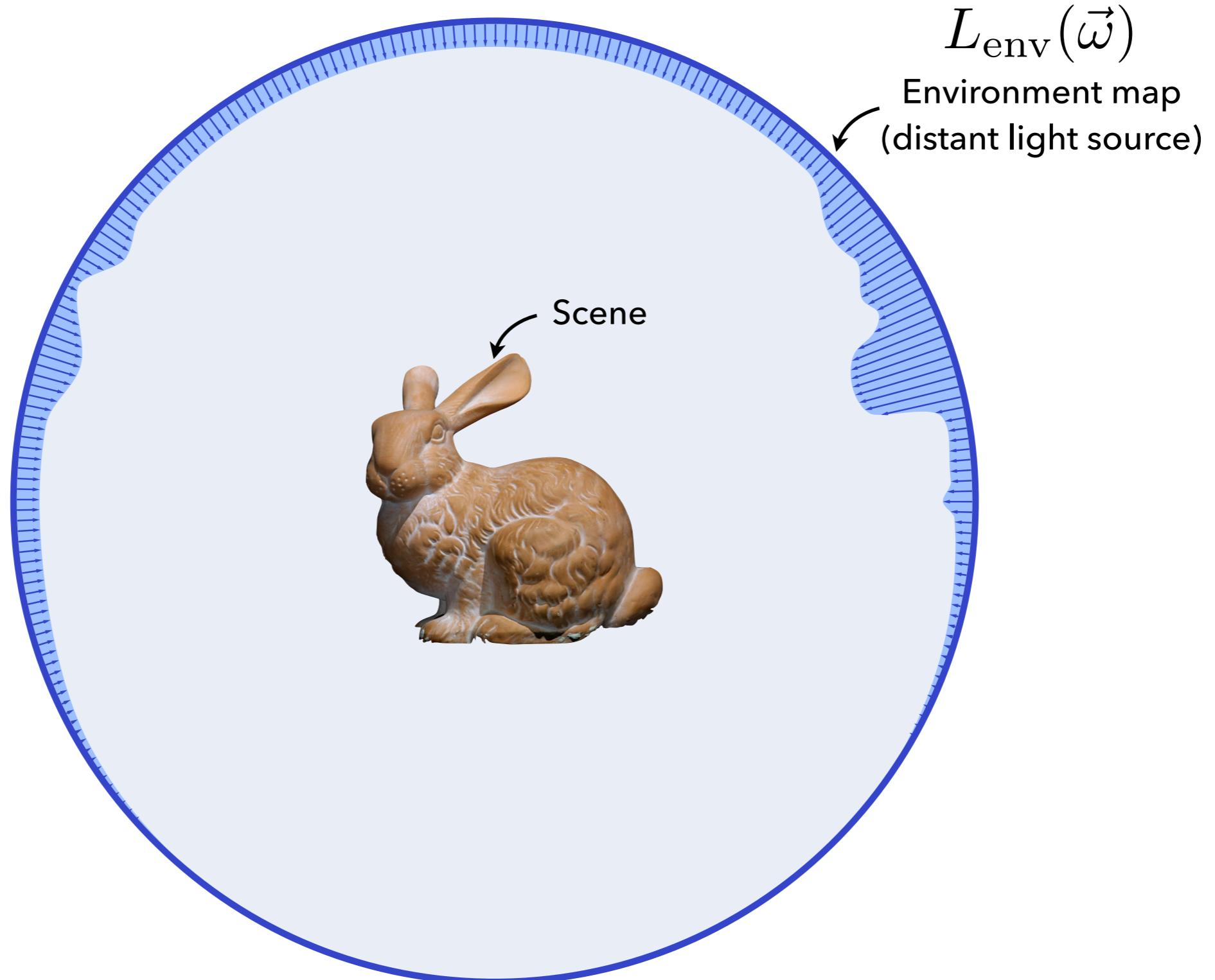
Courtesy of Engelhardt and Dachsbacher

HEALPix



Courtesy of Ryazanov

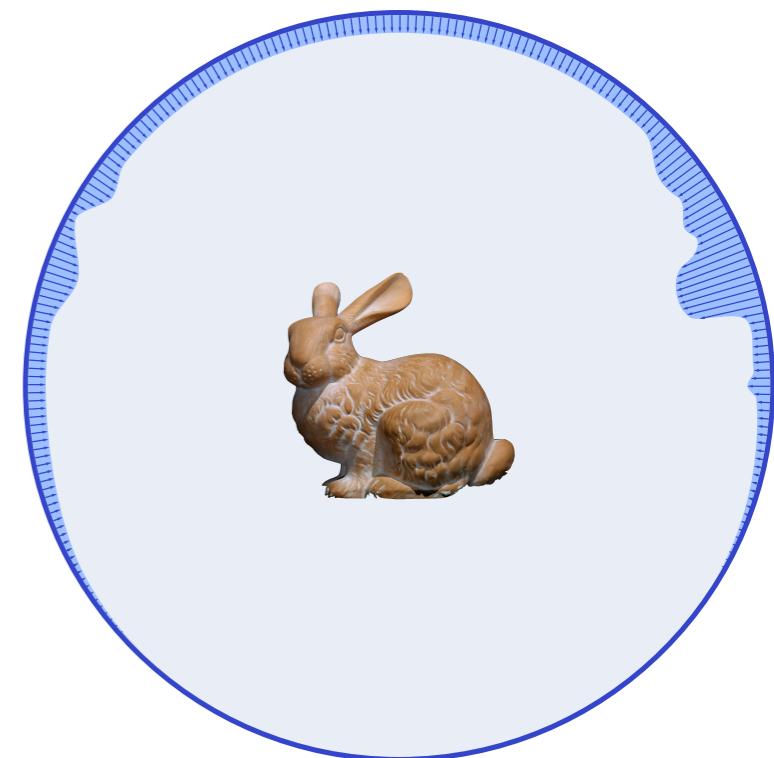
Environment Lighting



Environment Lighting

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

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



Environment Lighting



HENRIK WANN DENSEN - 2002

Environment Lighting



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

Importance Sampling L_{env}



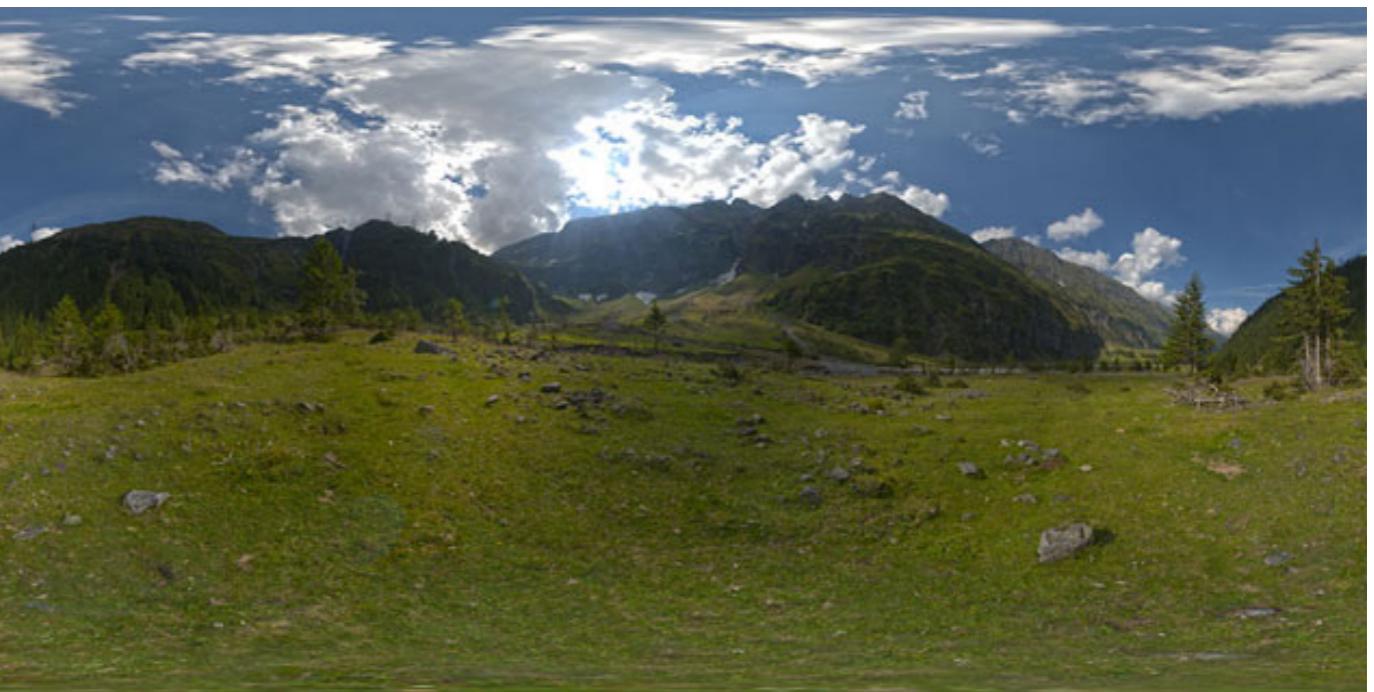
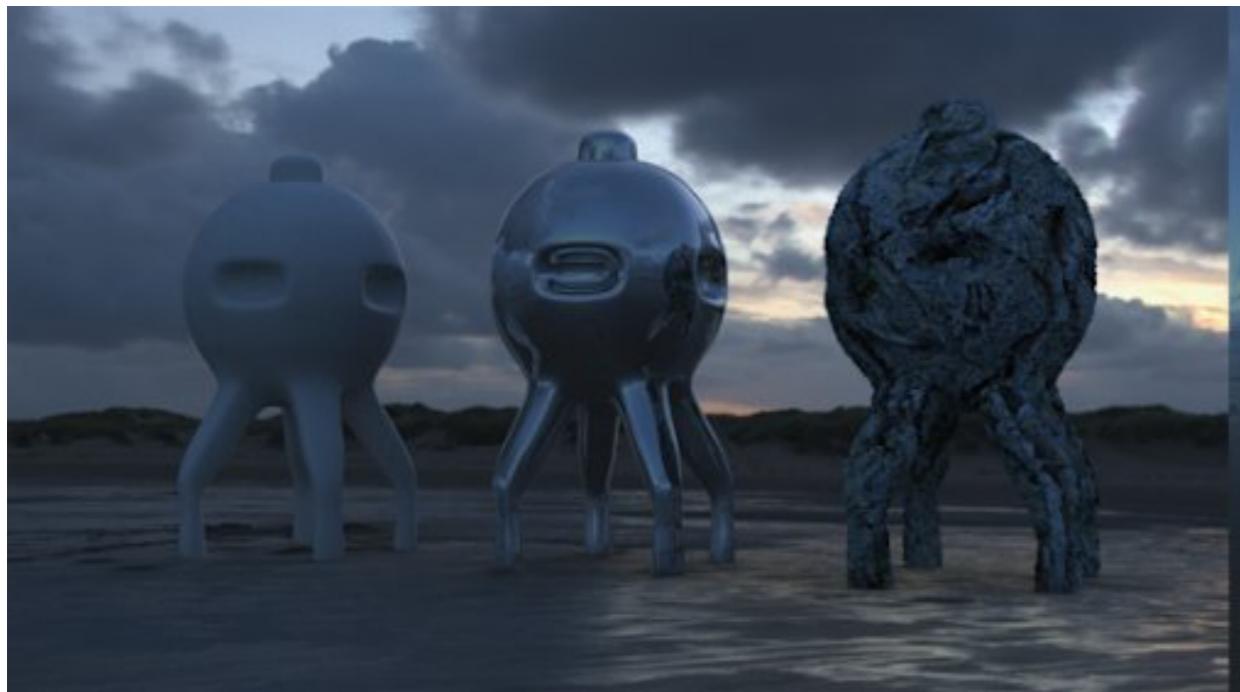
$$p(\vec{\omega}_i) \propto L_{\text{env}}(\vec{\omega}_i)$$

Importance Sampling L_{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

Visual Break



openfootage.net

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

$$d\vec{\omega} = \sin \theta d\theta d\phi$$

↗ Comes from the Jacobian

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

- If we set $p(\theta, \phi) \propto f(\theta, \phi) \sin \theta$ then the sines will effectively cancel out

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

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

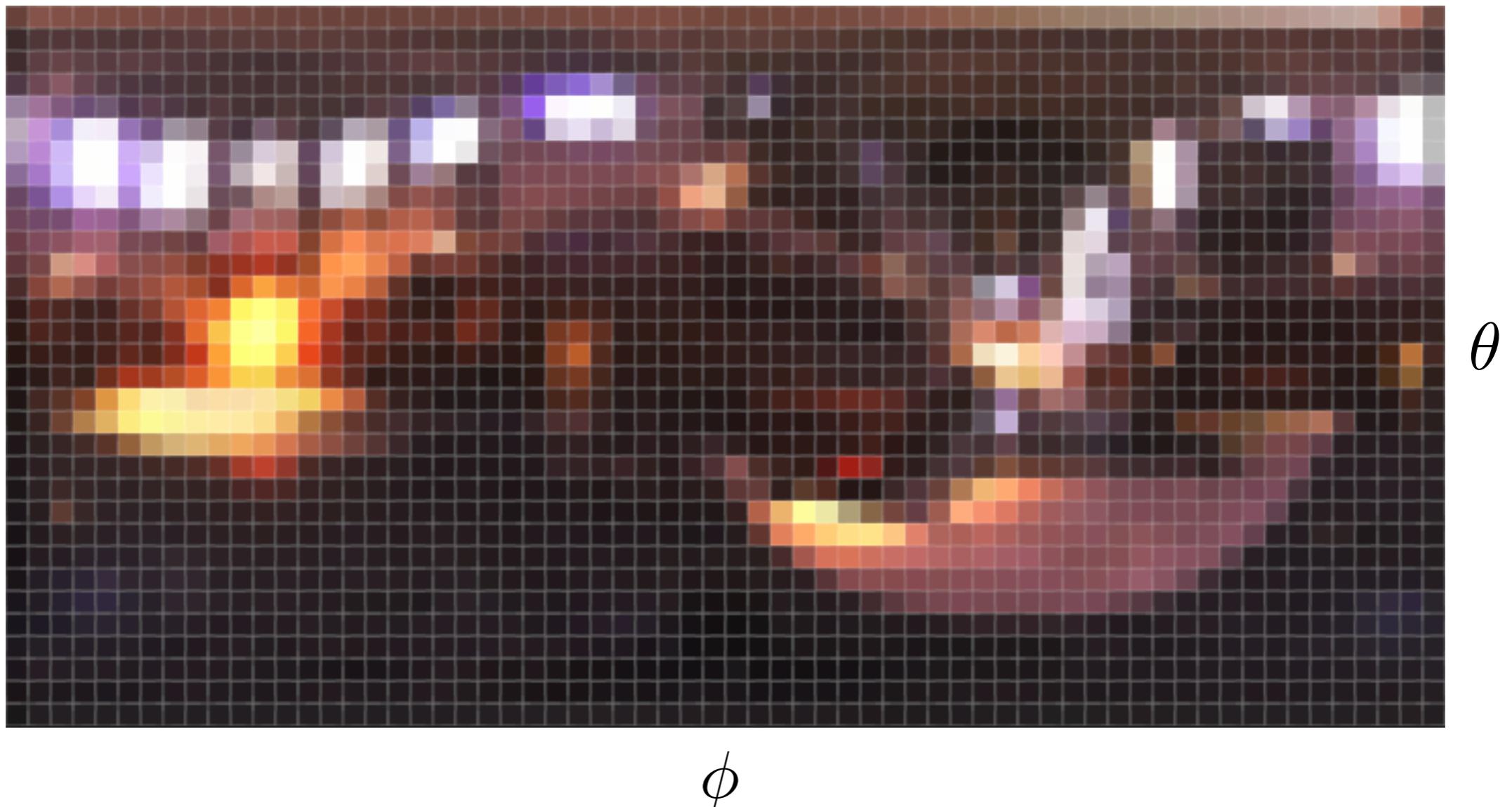
- Step 3: compute conditional PDF

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

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

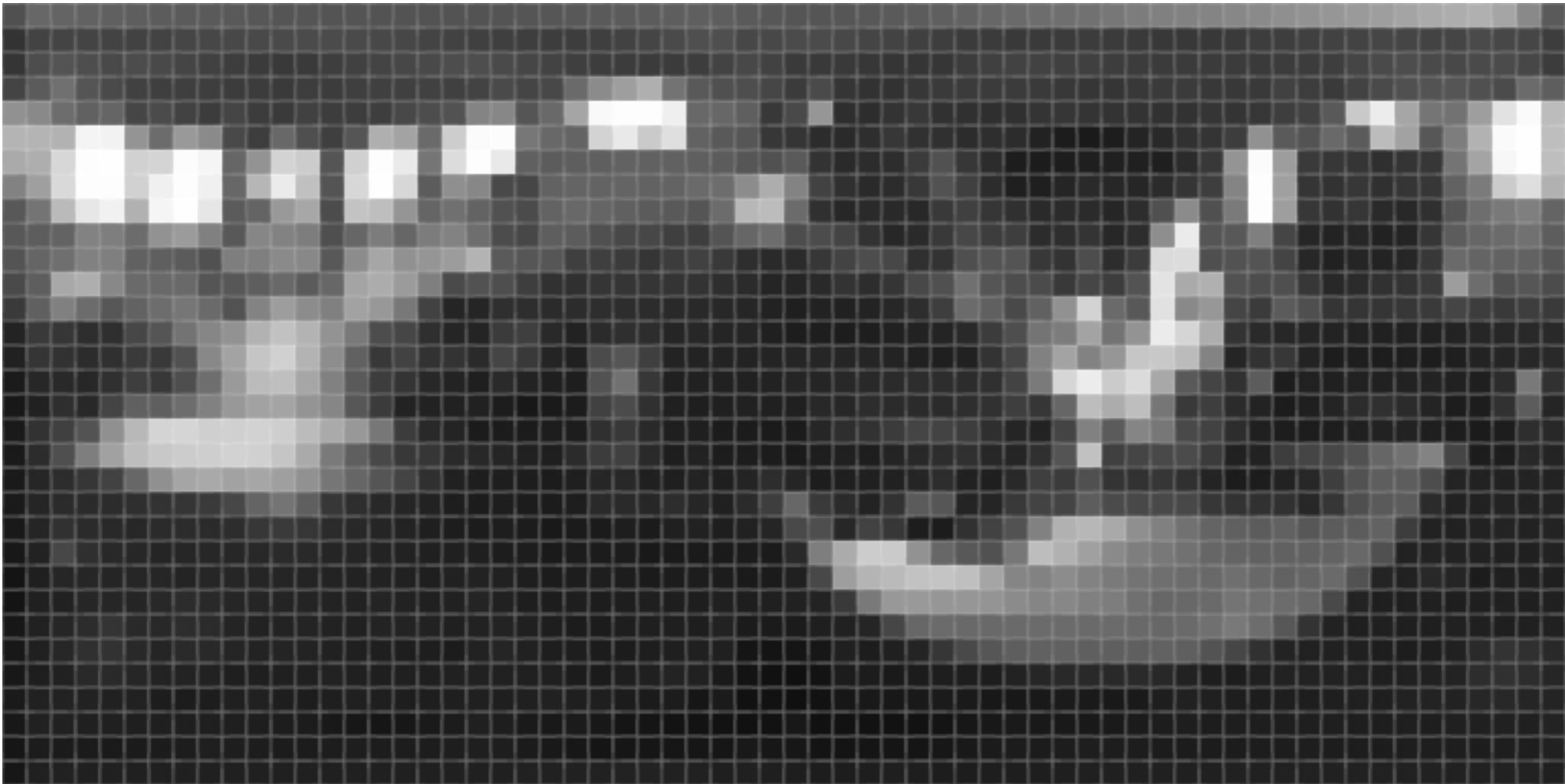
Step 1: Scalar Importance Func.

Original environment map



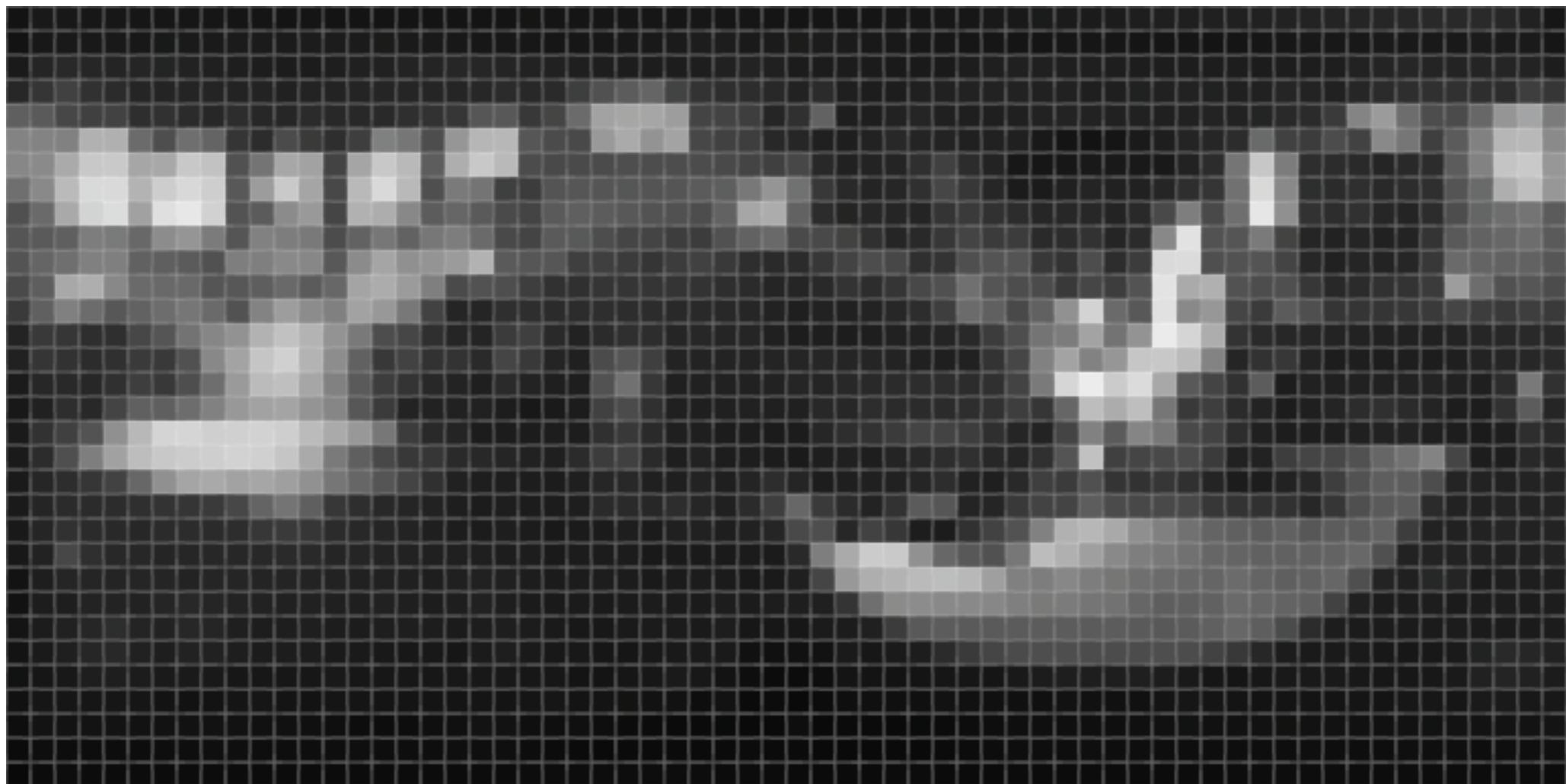
Step 1: Scalar Importance Func.

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



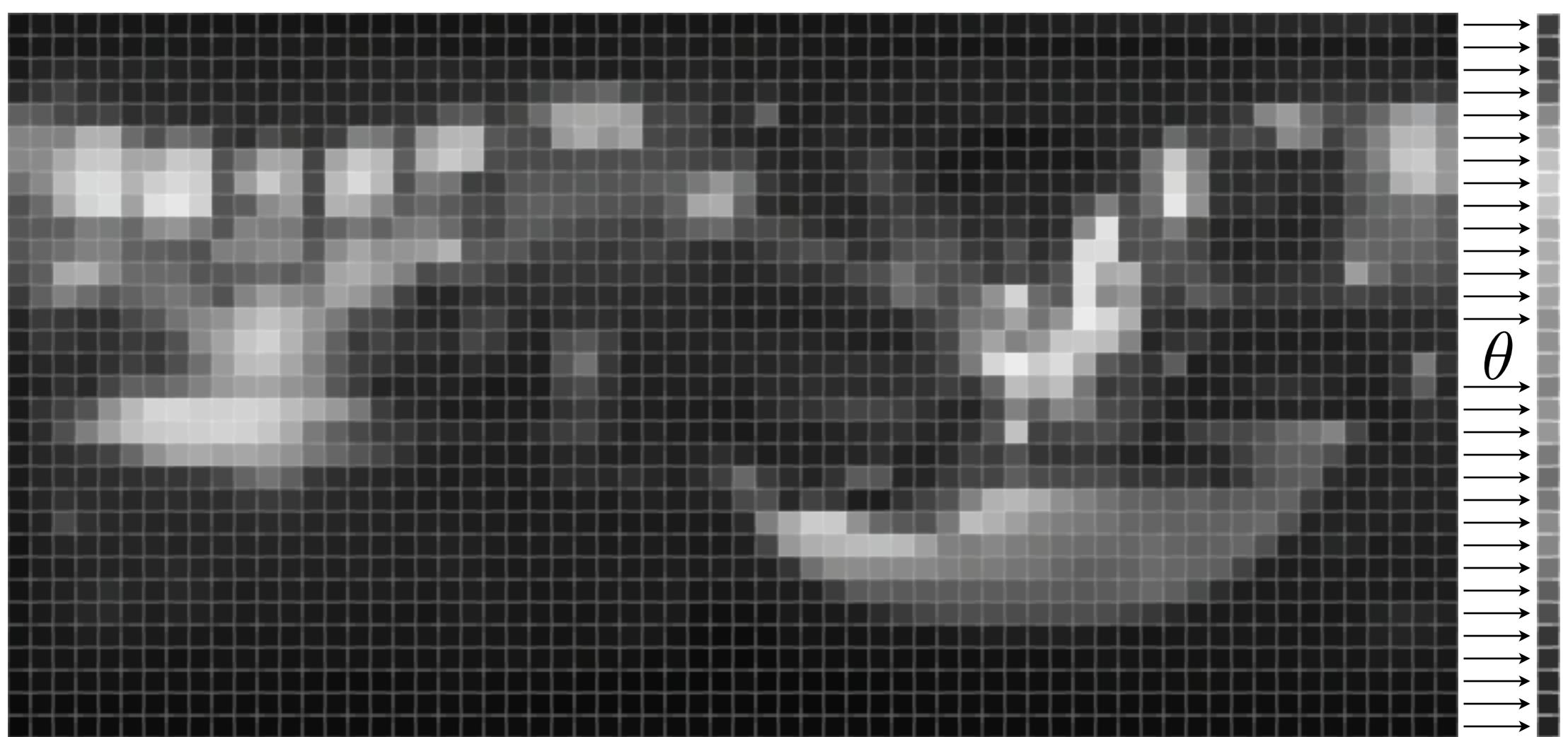
Step 1: Scalar Importance Func.

Multiplied by $\sin \theta$



Step 2: Marginalization

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

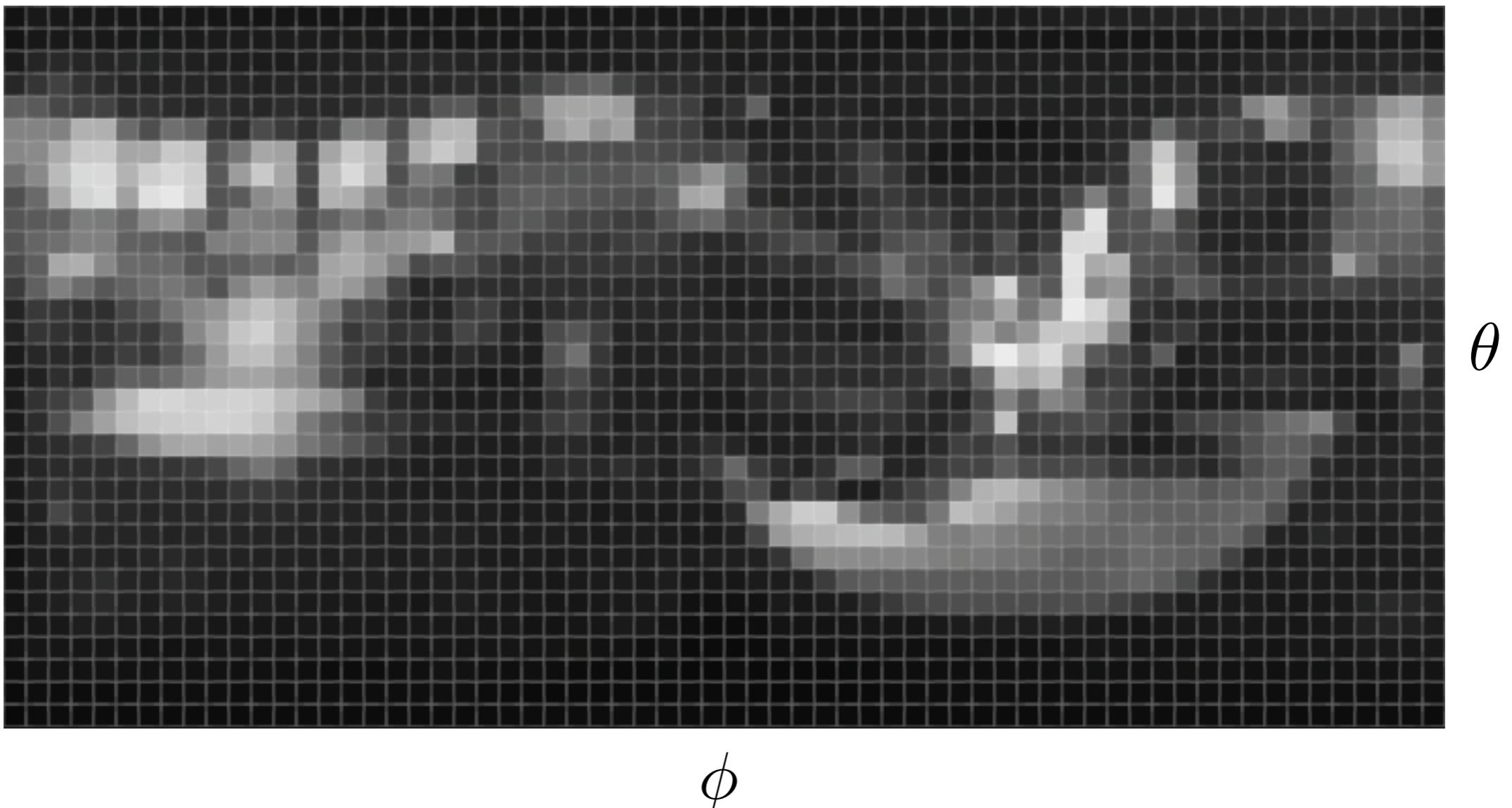


ϕ

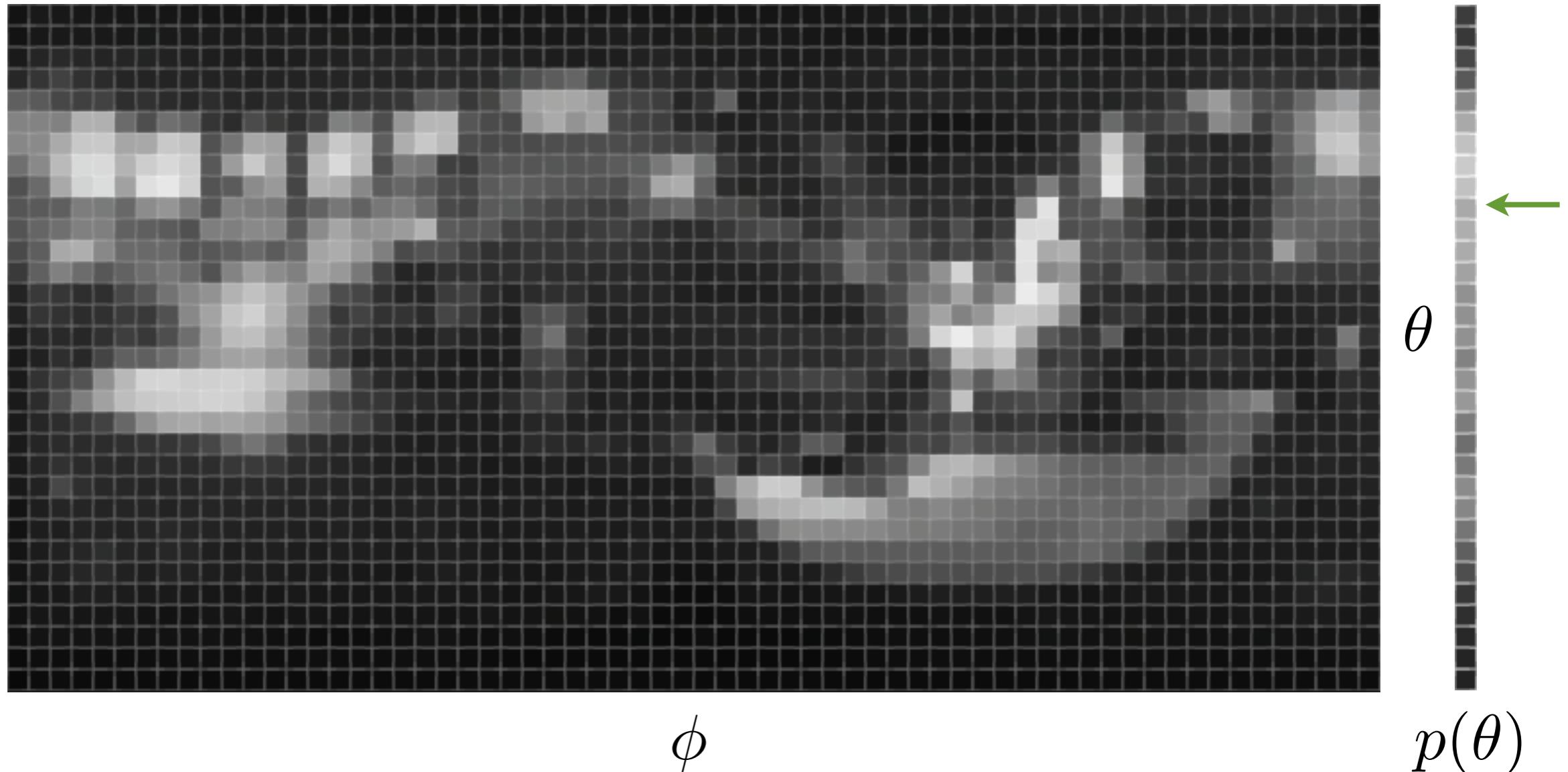
Marginalize to get $p(\theta)$

Step 3: Conditional PDFs

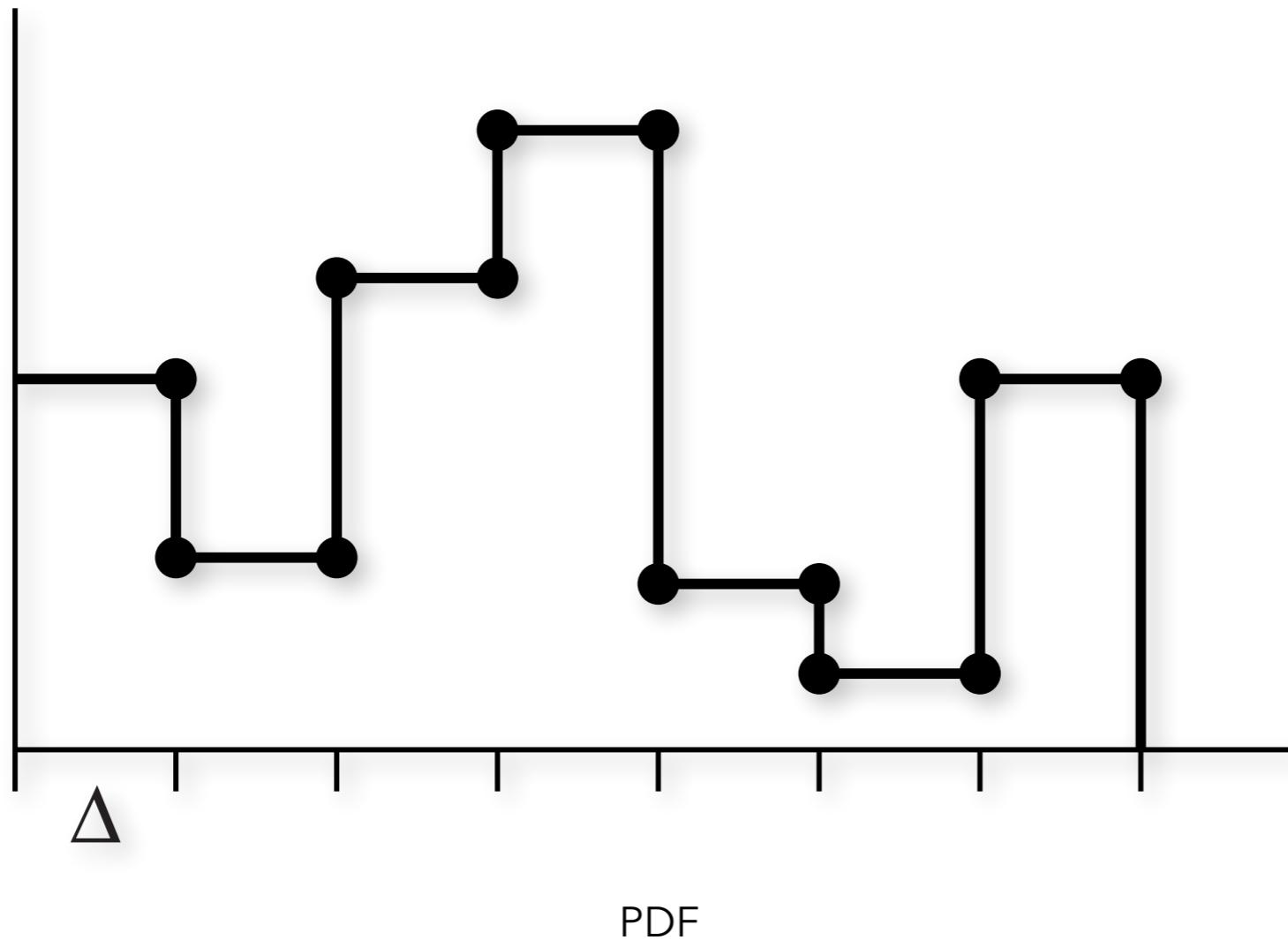
Once normalized, each row can serve as the conditional PDF



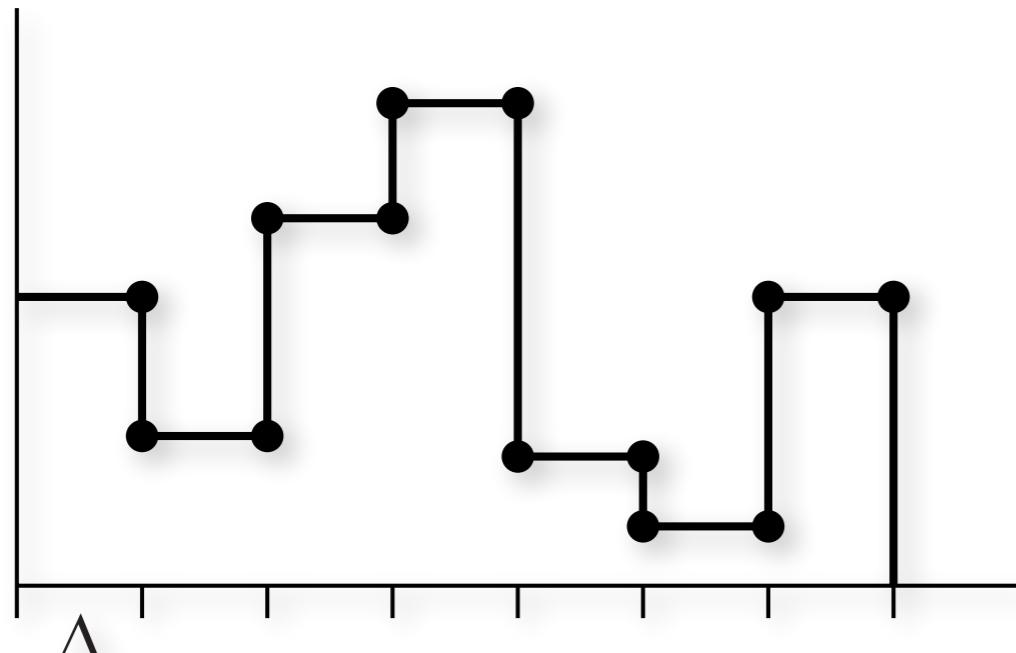
Step 4: Sampling



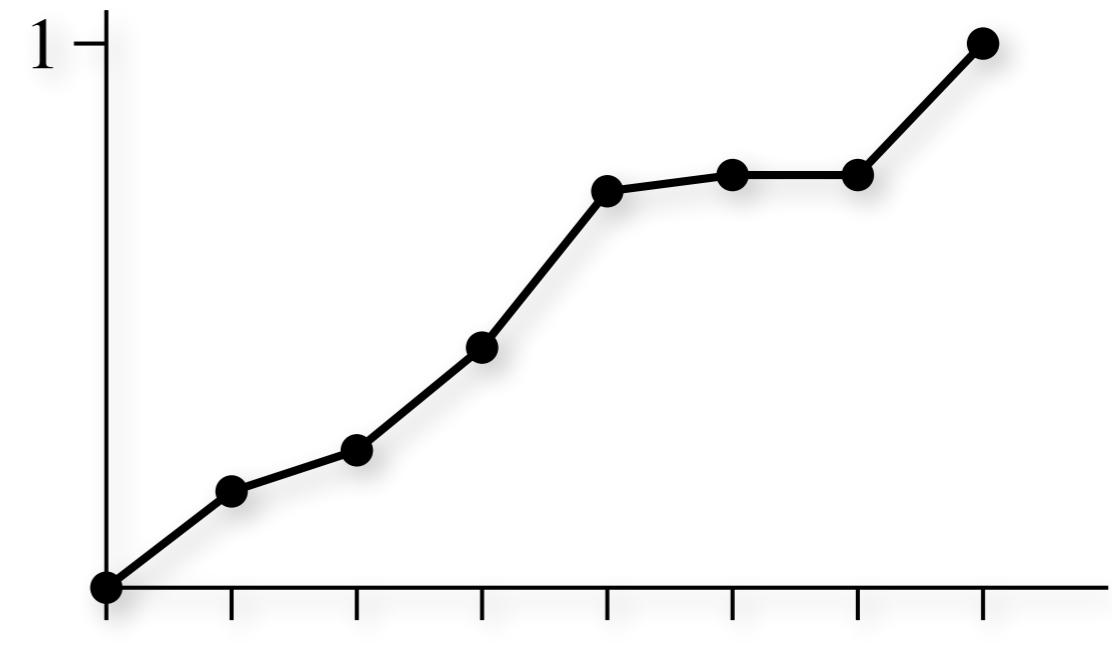
Sampling Discrete 1D PDFs



Sampling Discrete 1D PDFs



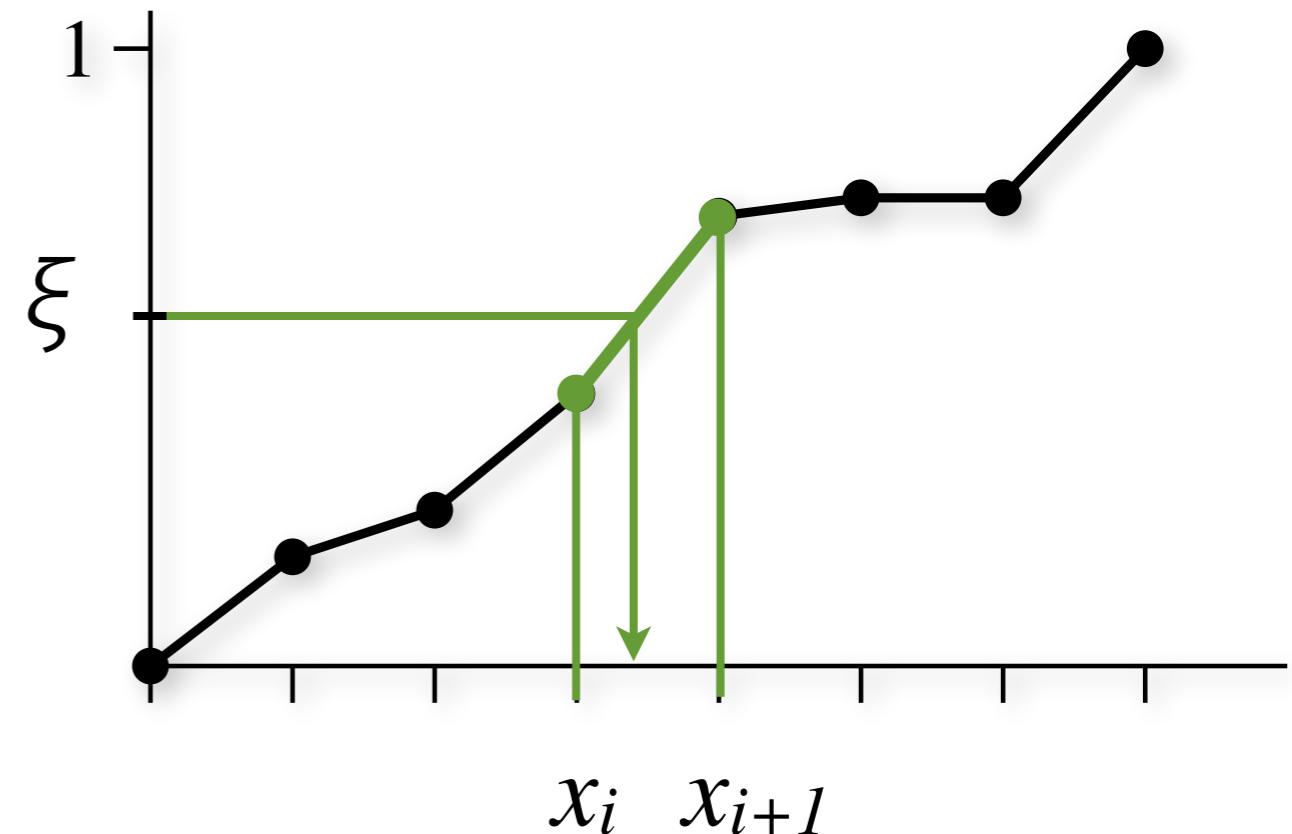
PDF



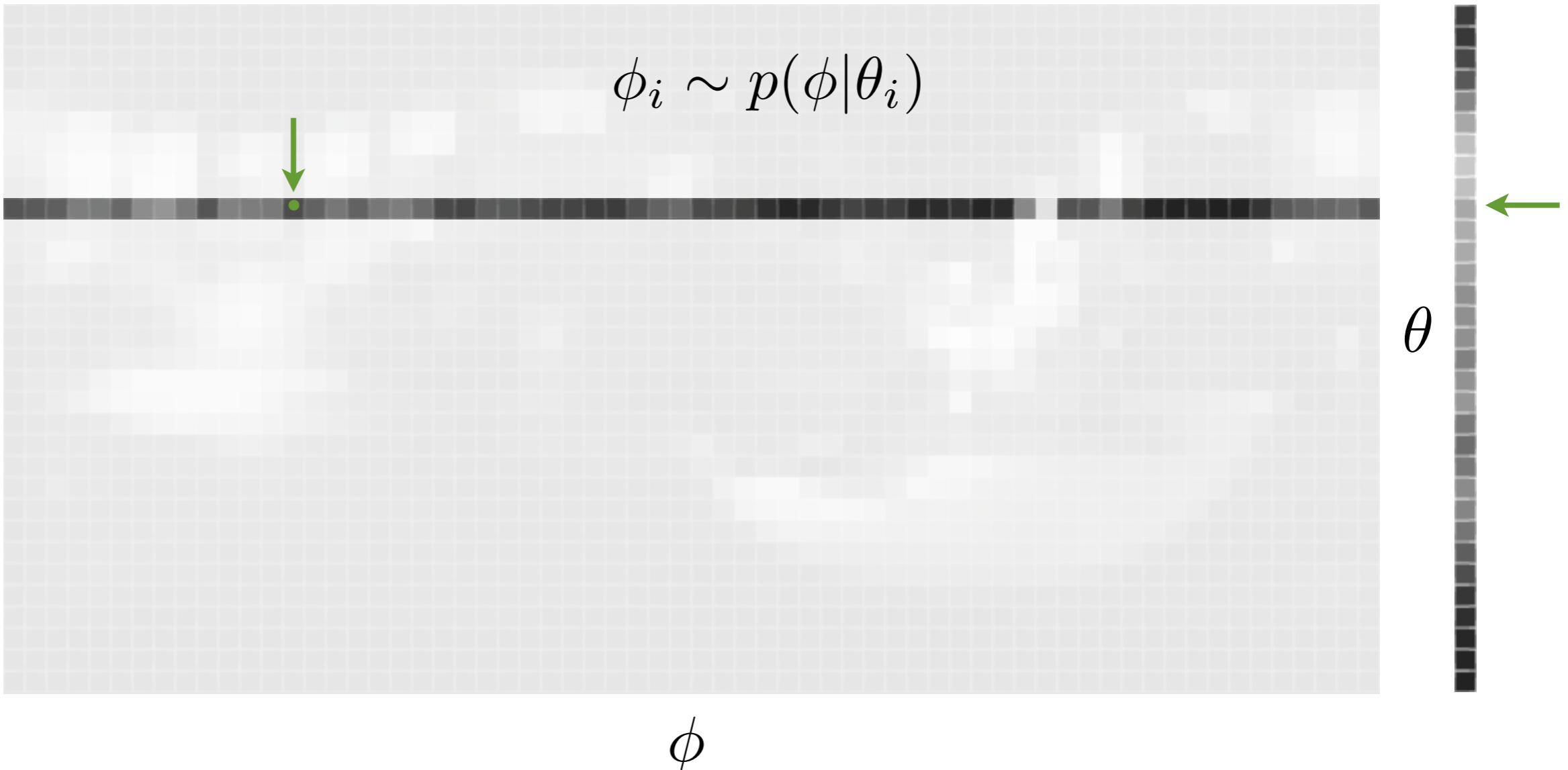
CDF

Sampling Discrete 1D PDFs

- Given a uniform random value ξ
- Using binary search find x_i and x_{i+1}
- Linearly interpolate to find x



Step 4: Sampling



Resulting Sample Distribution



Visual Break



openfootage.net

Hierarchical Sample Warping

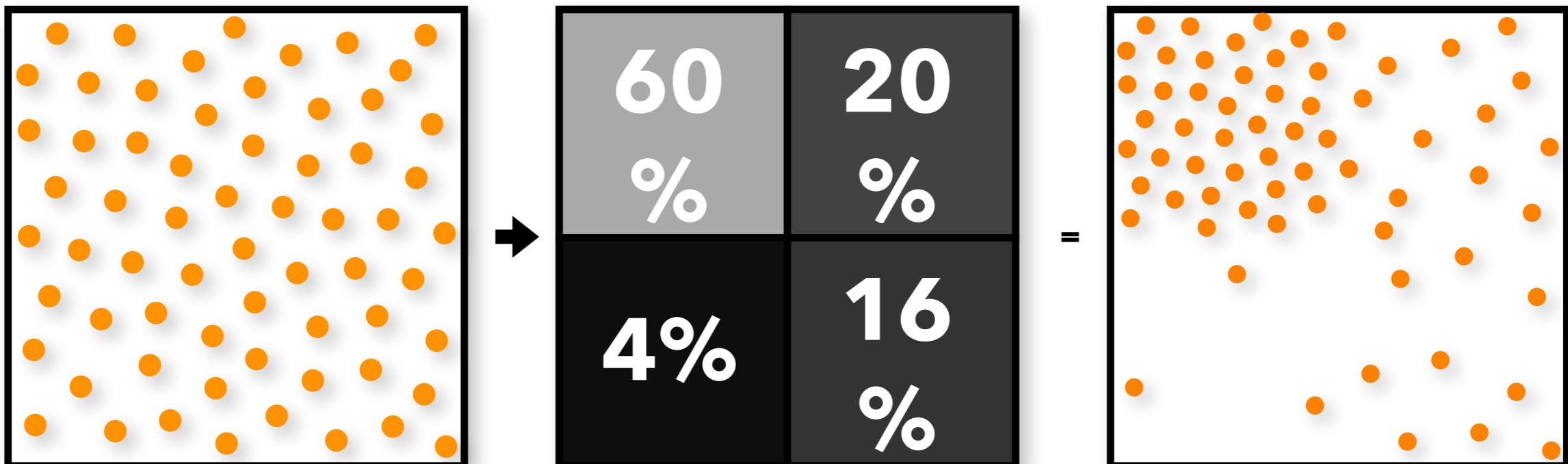
Clarberg, Jarosz, Akenine-Möller, Jensen. "Wavelet Importance Sampling," 2005.

- Given:
 - input point set, and
 - hierarchical representation of density function (mip-map)

Hierarchical Sample Warping

Clarberg, Jarosz, Akenine-Möller, Jensen. "Wavelet Importance Sampling," 2005.

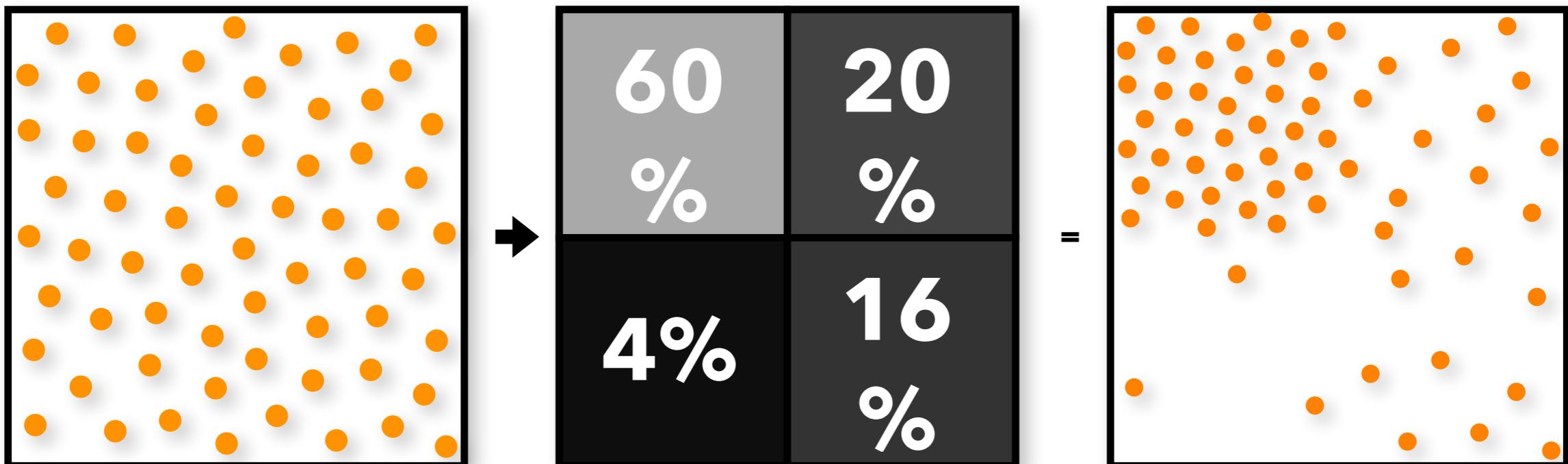
- Given:
 - **input point set**, and
 - hierarchical representation of density function (mip-map)
- Recursively warp point set to match the importance function



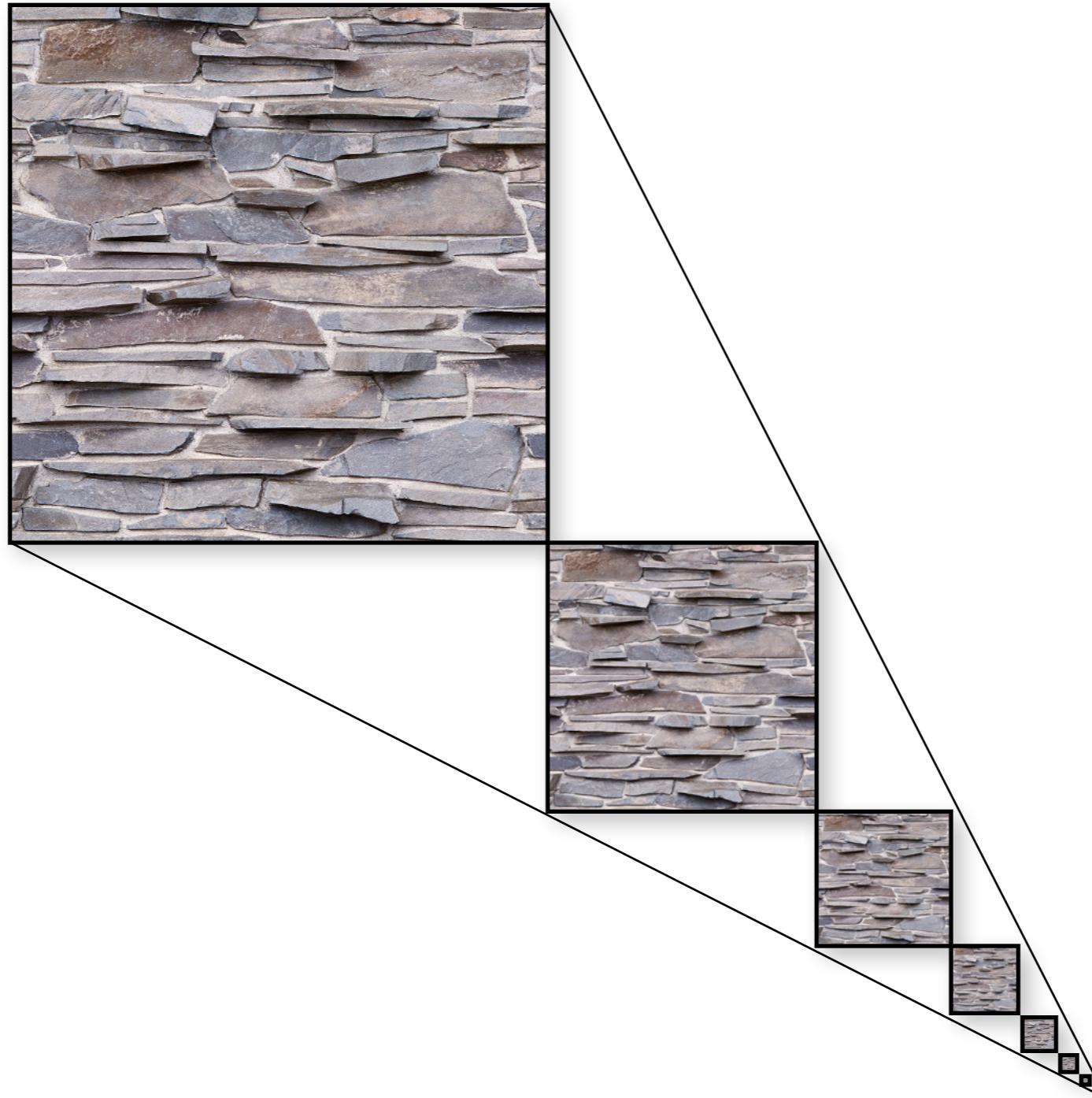
Hierarchical Sample Warping

Clarberg, Jarosz, Akenine-Möller, Jensen. "Wavelet Importance Sampling," 2005.

- Given:
 - input point set, and
 - **hierarchical representation of density function** (mip-map)
- Recursively warp point set to match the importance function



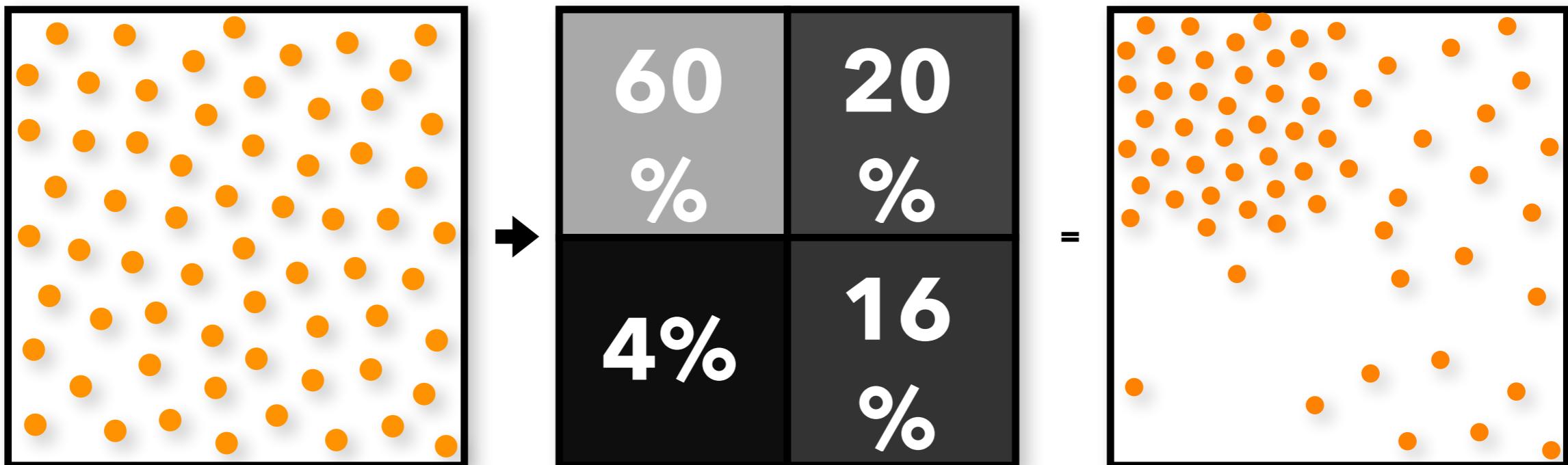
Mip-Map

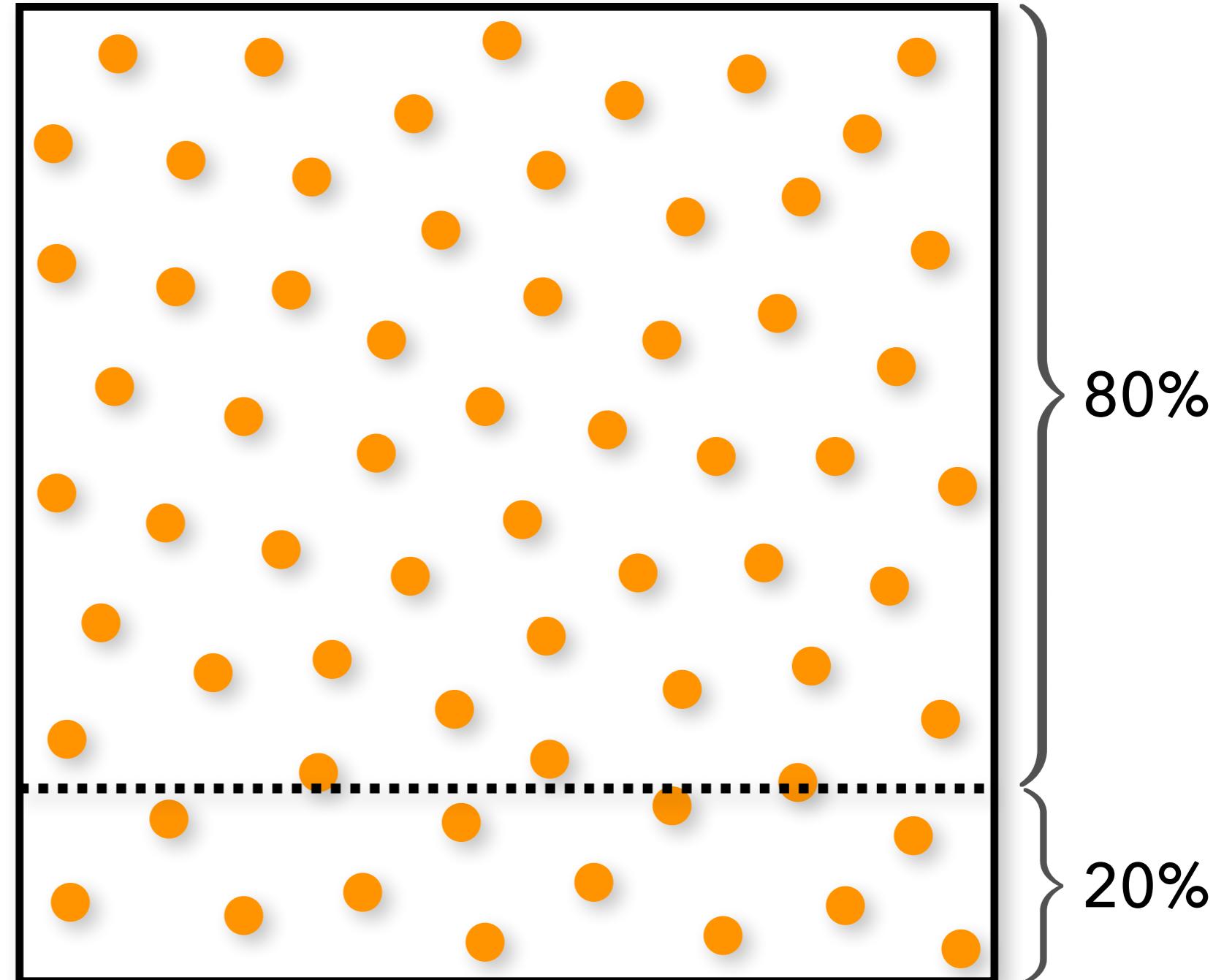
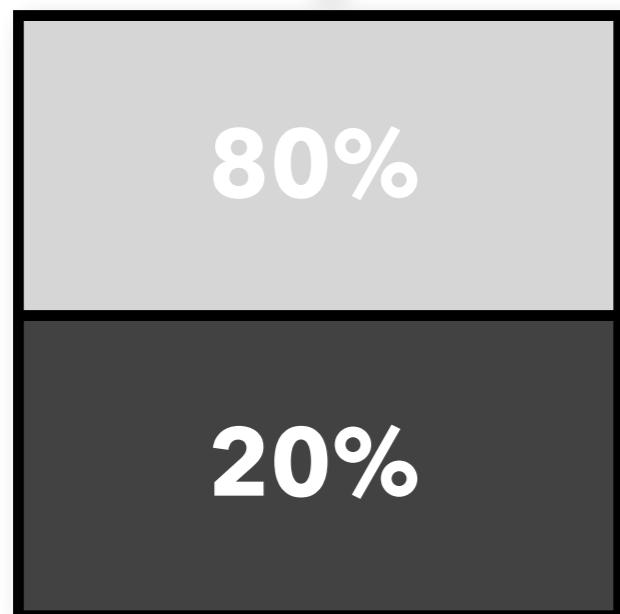
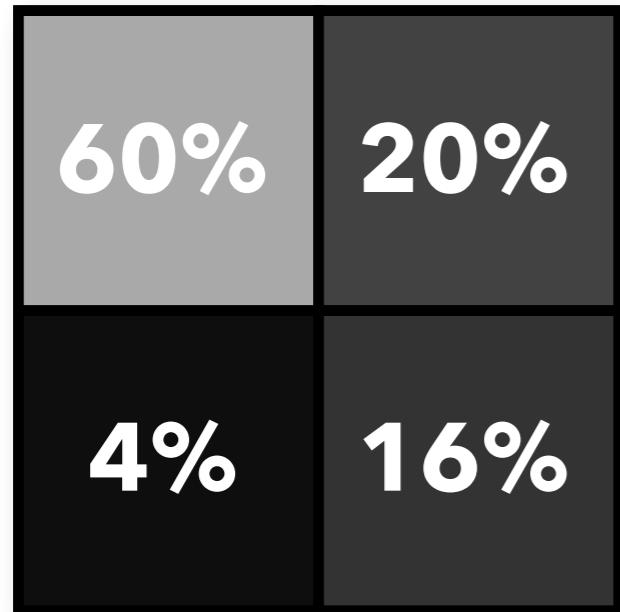


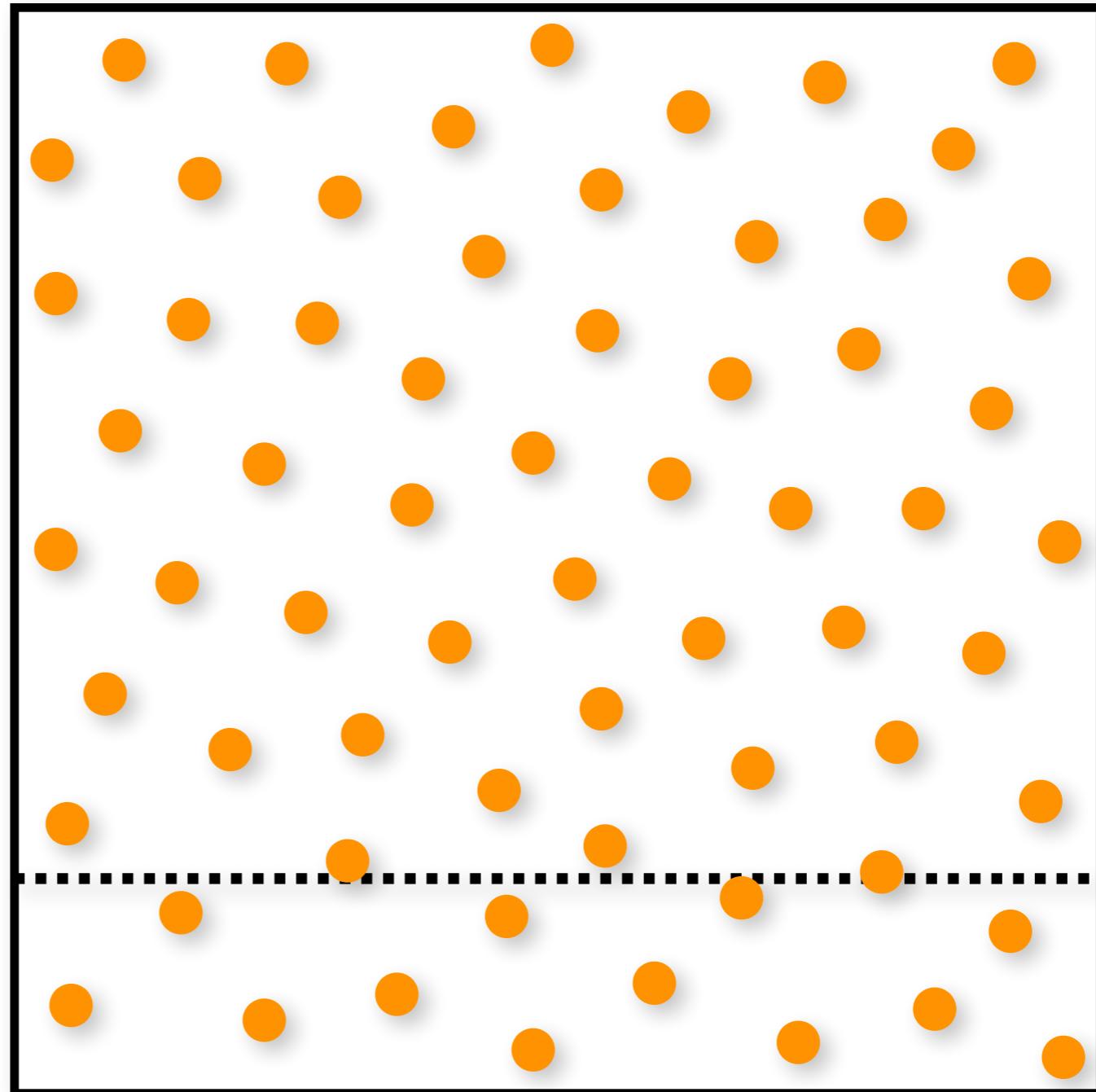
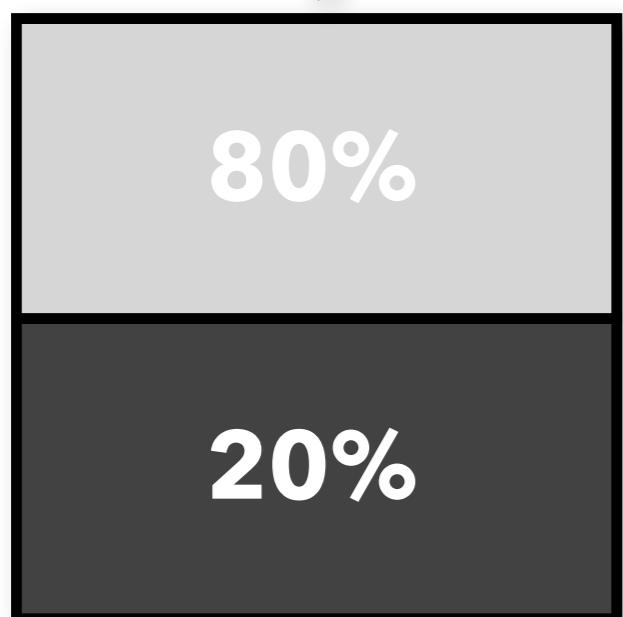
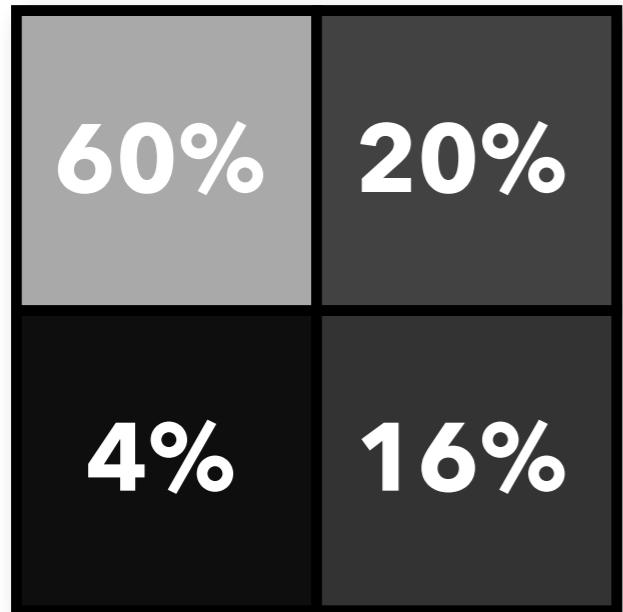
Hierarchical Sample Warping

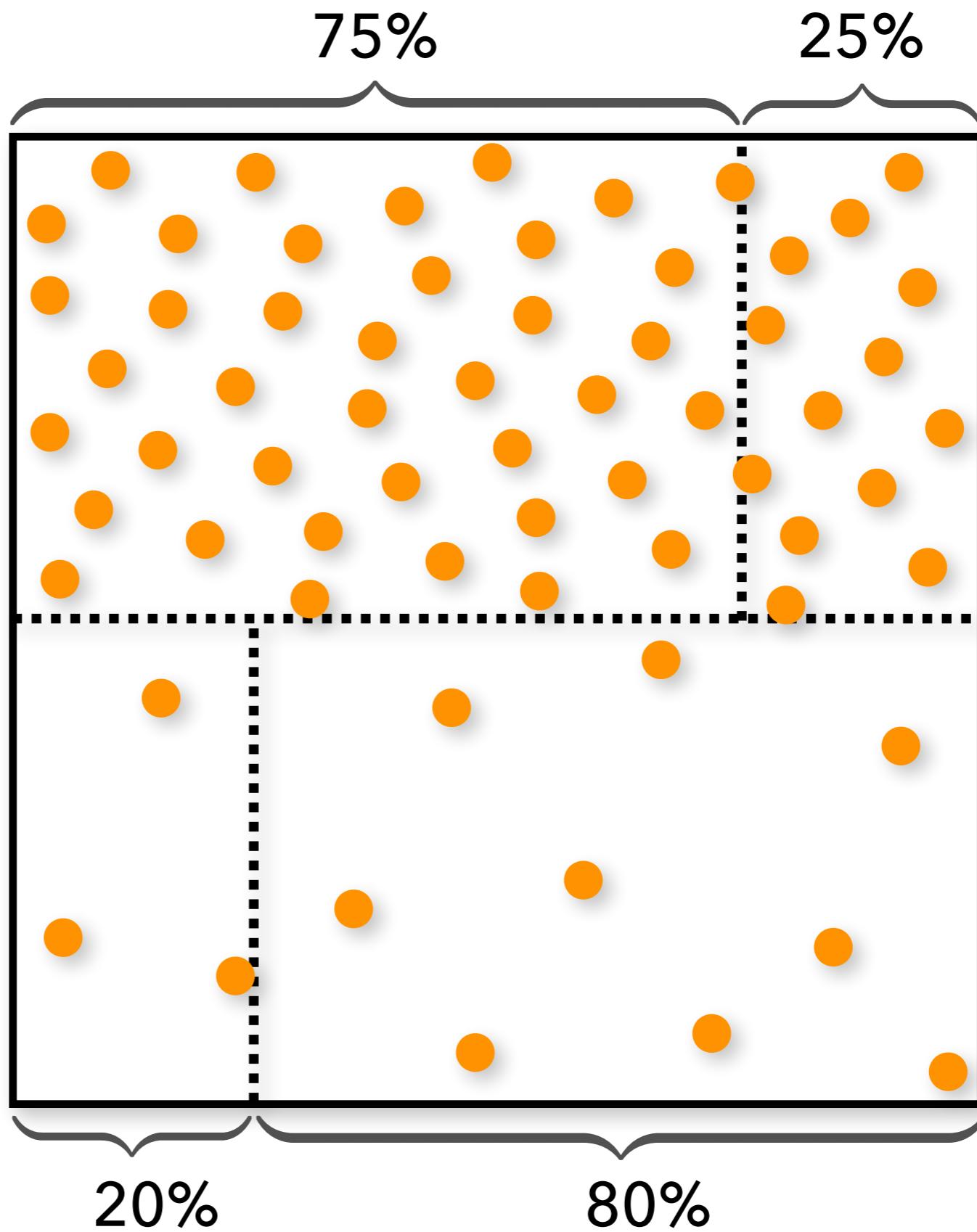
Clarberg, Jarosz, Akenine-Möller, Jensen. "Wavelet Importance Sampling," 2005.

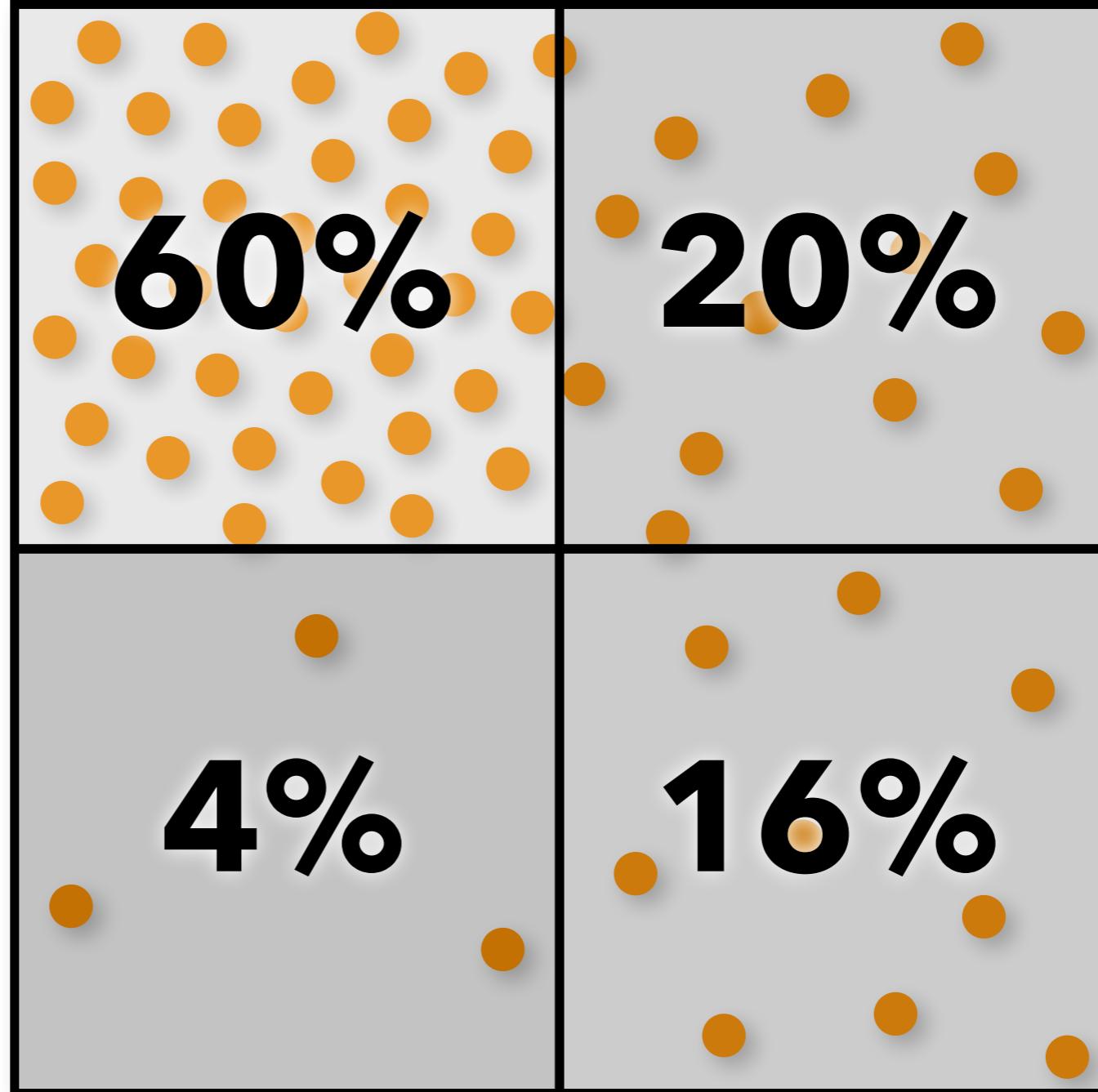
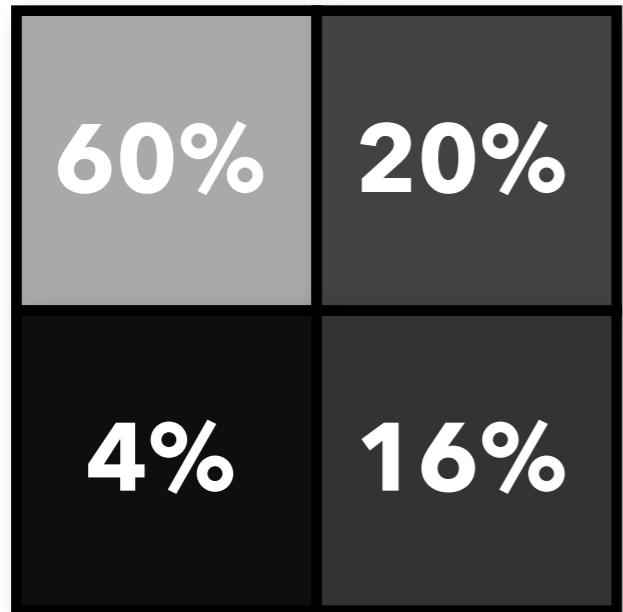
- Given:
 - input point set, and
 - hierarchical representation of density function (mip-map)
- Recursively warp point set to match the importance function











repeat on each quadrant recursively

Environment Lighting

$$\begin{aligned} L_r(\mathbf{x}, \vec{\omega}_r) &= \int_{\Omega} f_r(\mathbf{x}, \vec{\omega}_i, \vec{\omega}_r) L_i(\mathbf{x}, \vec{\omega}_i) \cos \theta_i d\vec{\omega}_i \\ &= \int_{\Omega} f_r(\mathbf{x}, \vec{\omega}_i, \vec{\omega}_r) L_{\text{env}}(\vec{\omega}_i) V(\mathbf{x}, \vec{\omega}_i) \cos \theta_i d\vec{\omega}_i \\ &\approx \frac{1}{N} \sum_{i=1}^N \frac{f_r(\mathbf{x}, \vec{\omega}_i, \vec{\omega}_r) L_{\text{env}}(\vec{\omega}_i) V(\mathbf{x}, \vec{\omega}_i) \cos \theta_i}{\text{pdf}(\vec{\omega}_i)} \end{aligned}$$

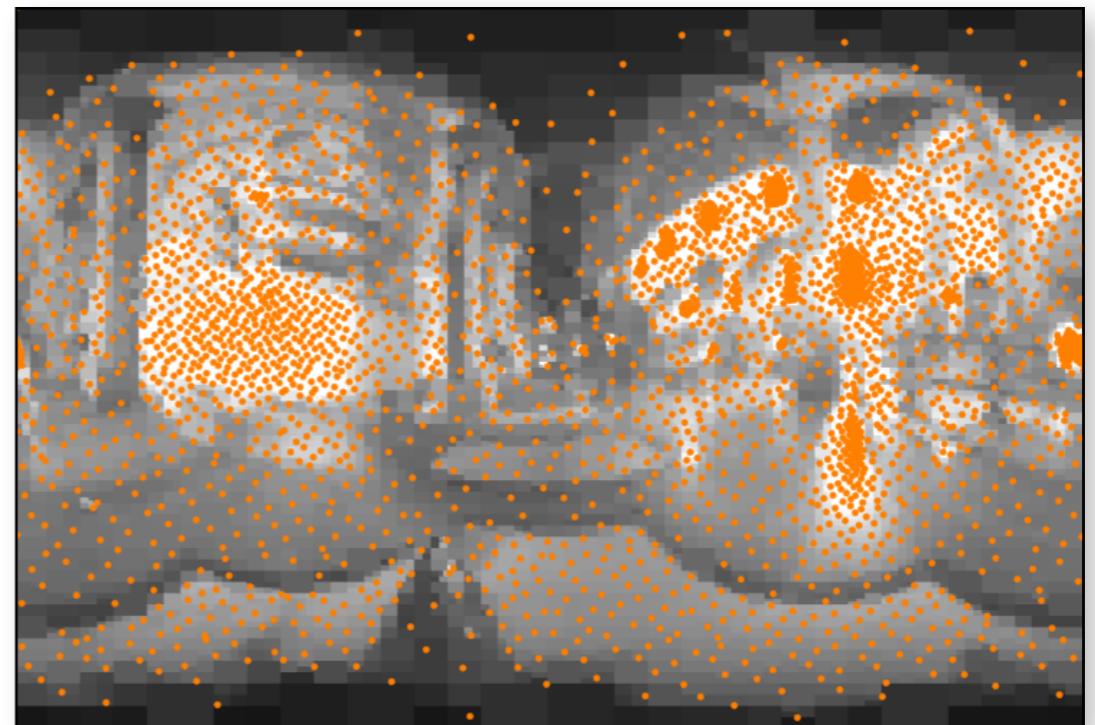
- What pdf should/can we use?
 - prop. to f_r
 - prop. to L_{env}

Sample Distributions

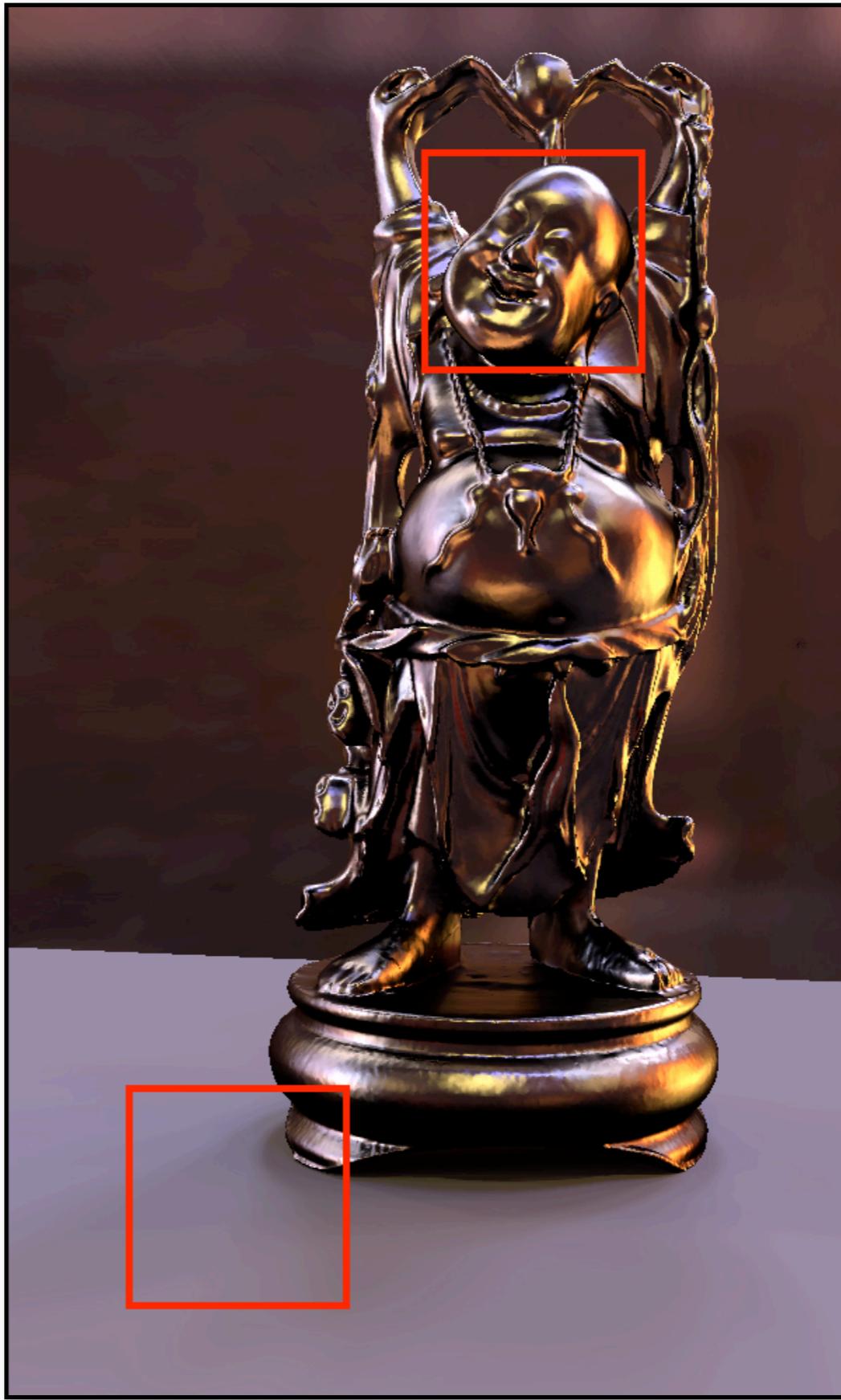


BRDF

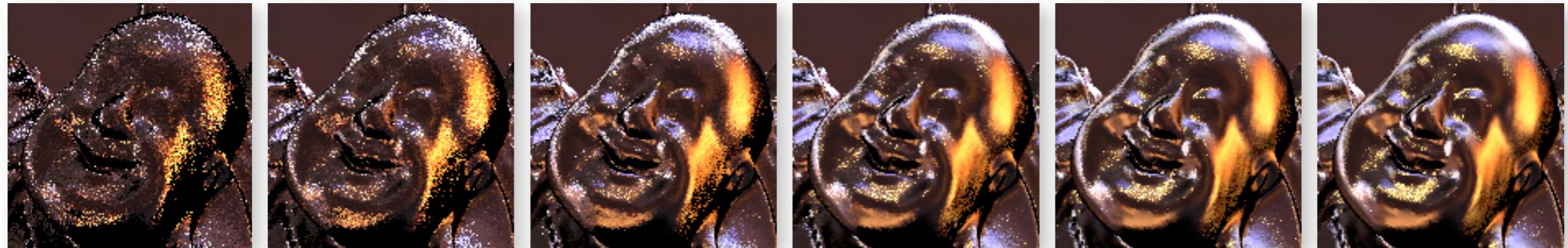
or



Lighting



BRDF



Lighting



1

3

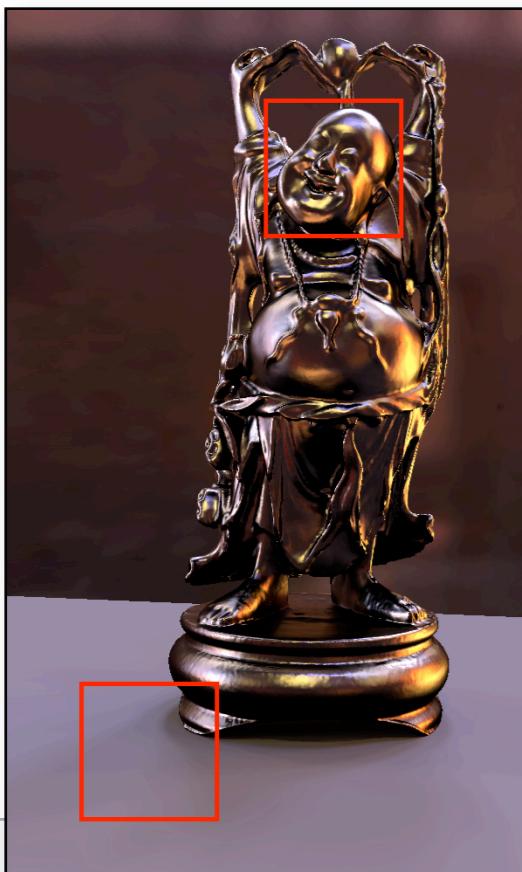
10

30

60

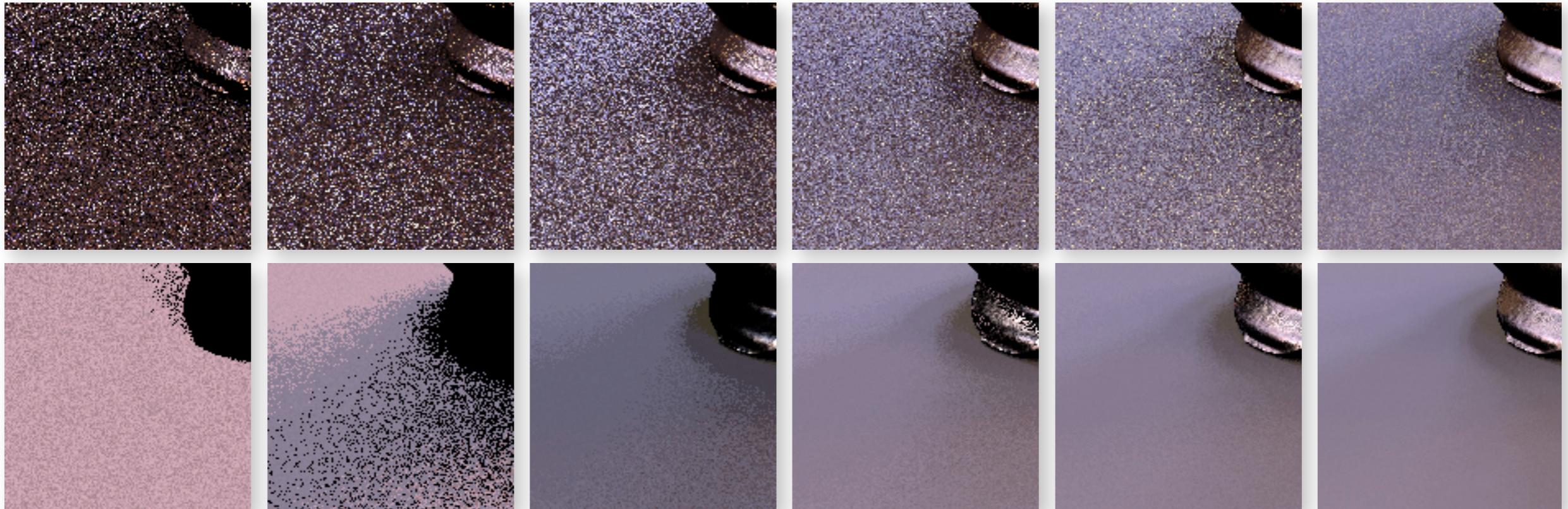
100

Samples / Pixel



BRDF

Lighting



1

3

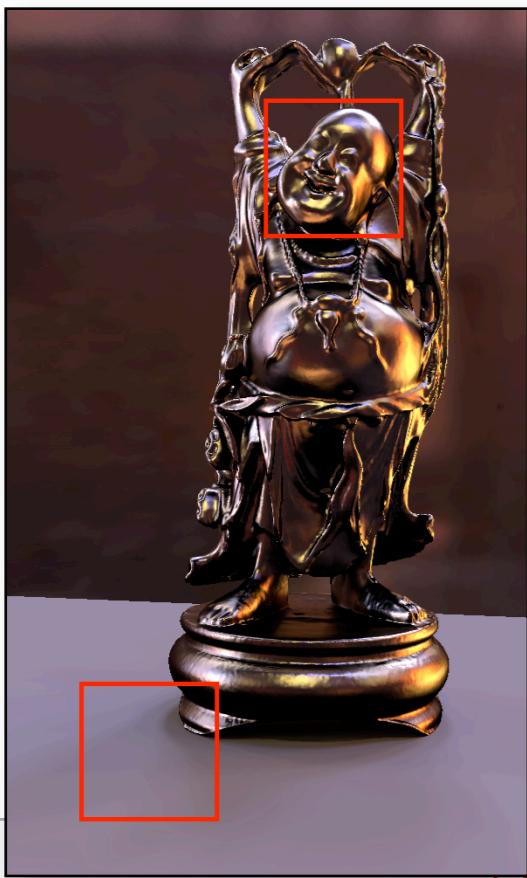
10

30

60

100

Samples / Pixel



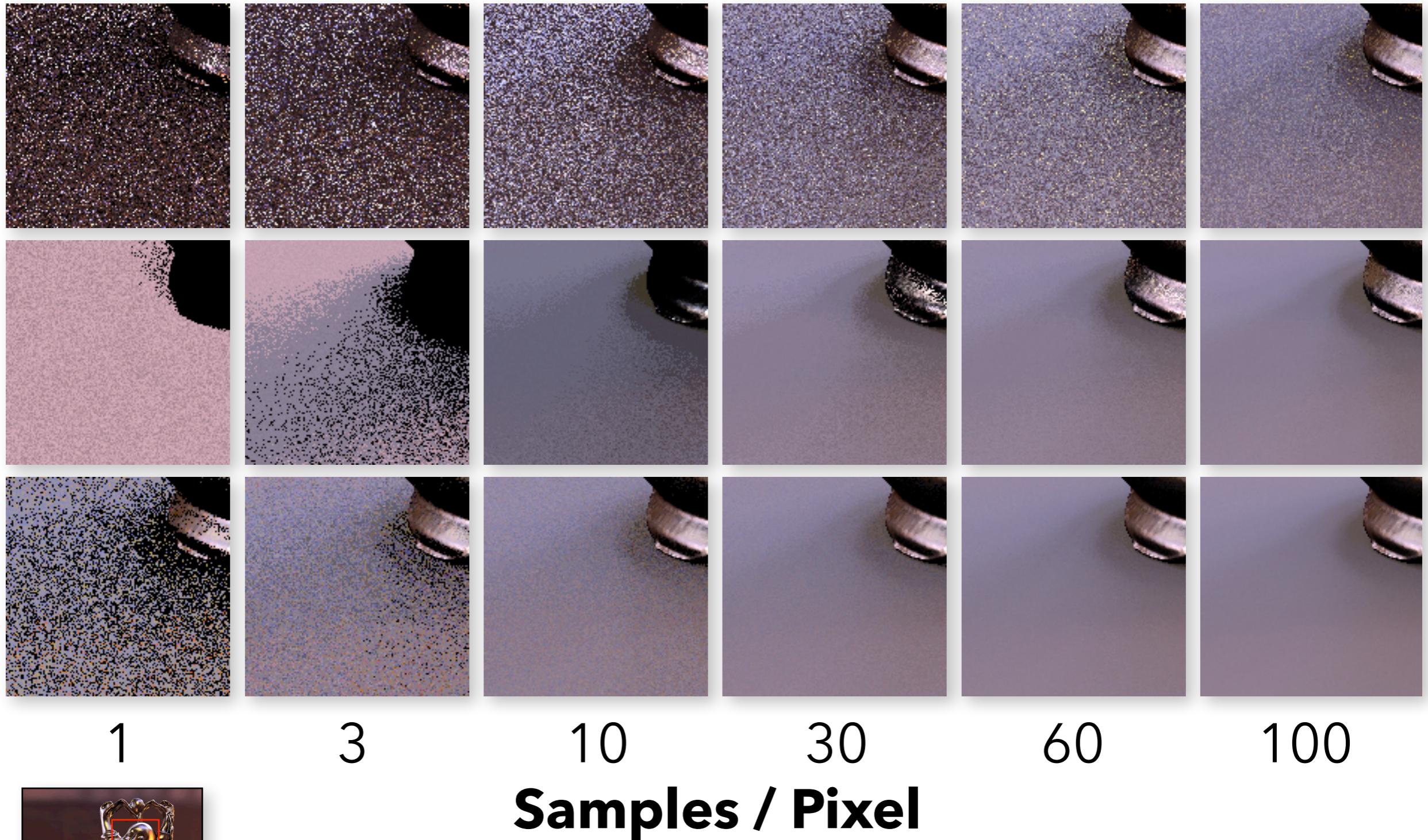
Product Sampling

Environment Lighting

- Could use MIS
 - pdf is a linear combination of f_r and L_{env}
- Can we sample the *product* ($f_r \cdot L_{\text{env}} \cdot \cos \theta_i$) directly?
 - ★ Clarberg, Jarosz, Akenine-Möller, Jensen. "Wavelet Importance Sampling," 2005.
 - ★ Subr and Arvo. "Steerable Importance Sampling," 2007.
 - ★ Jarosz, Carr, Jensen. "Importance Sampling Spherical Harmonics," 2009.
 - ★ many more...

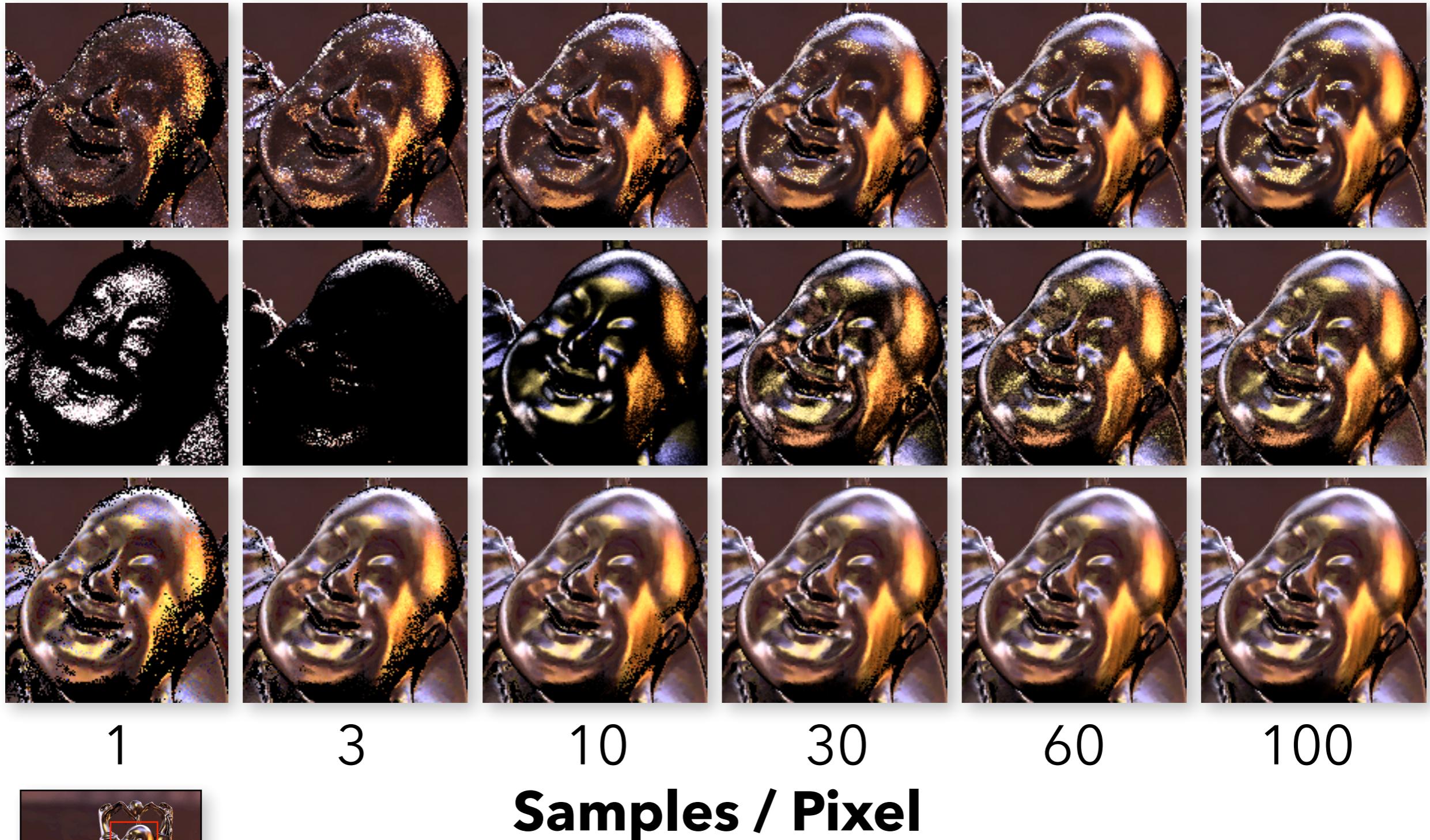
Product Lighting

BRDF

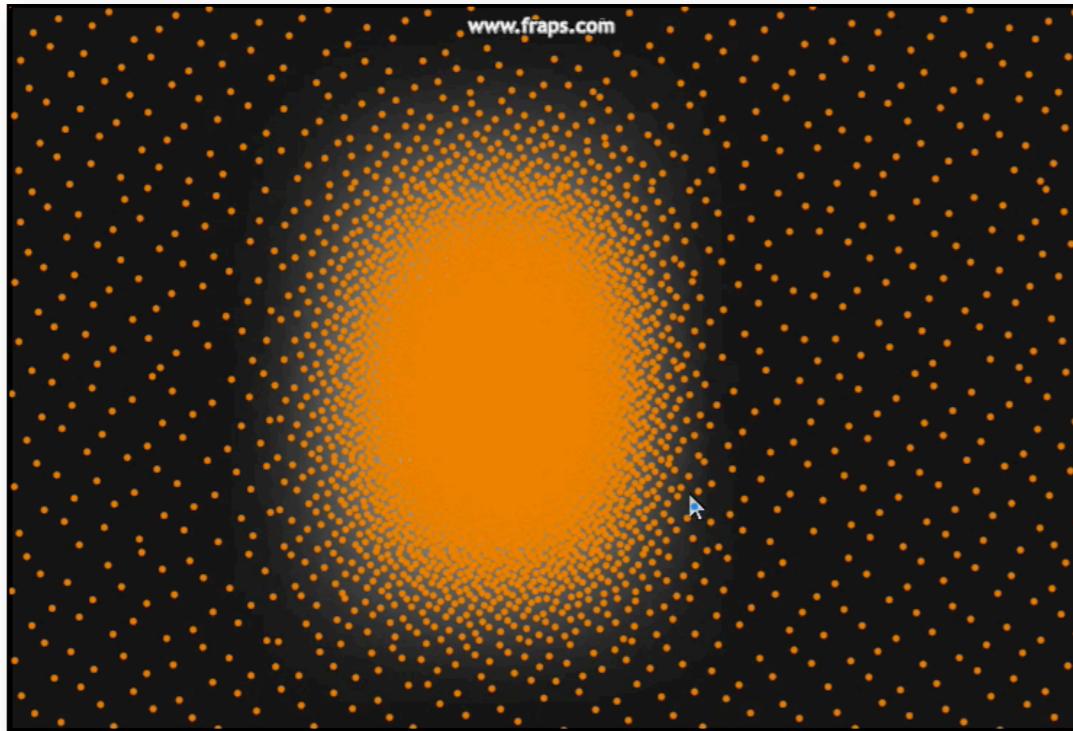


Product Lighting

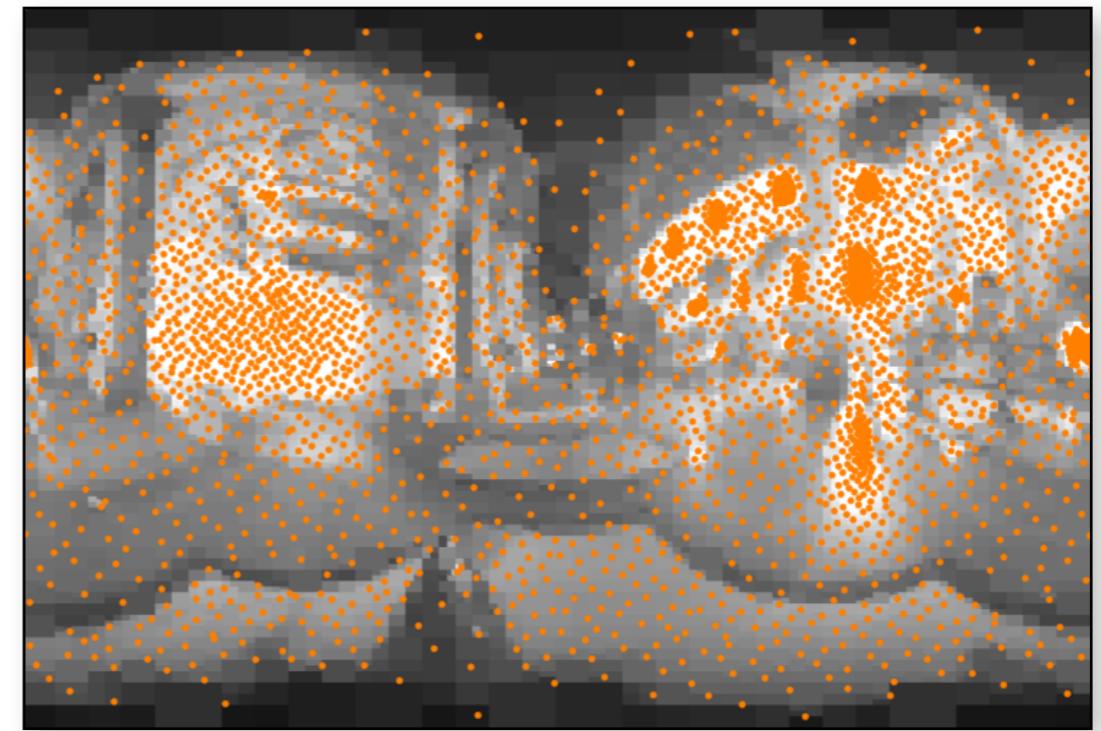
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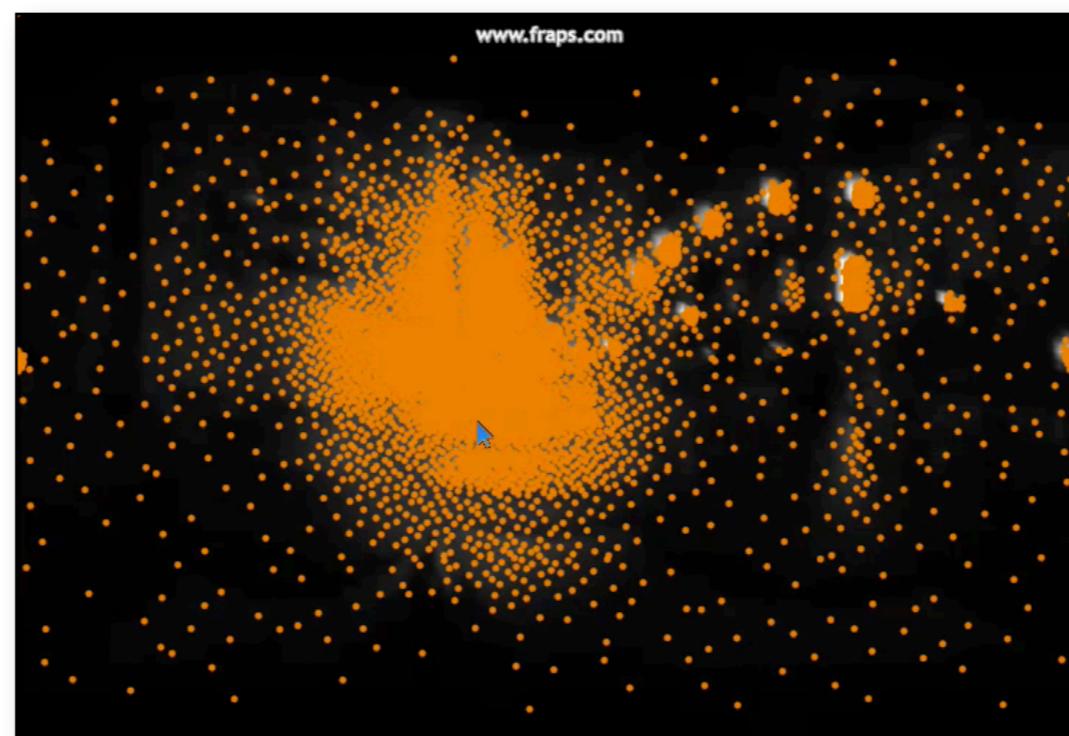
Sample Distributions



BRDF



Lighting



BRDF x Lighting



Blue Fabric

Grace Cathedral

Questions?
