

*Steerable Importance Sampling
for
Efficient Direct Distant Illumination*

Kartic Subr

James Arvo

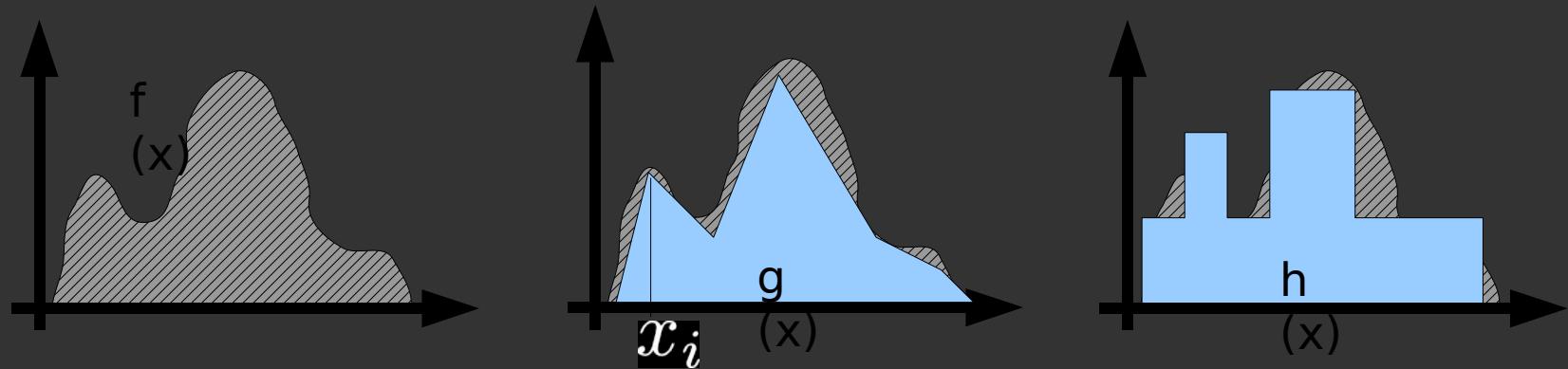
Review: Importance sampling

$$\int_{\mathcal{D}} f(x) \, dx \approx \frac{1}{N} \int_{\mathcal{D}} g(x) \, dx \sum_{i=1}^N \frac{f(x_i)}{g(x_i)}$$

where $x_i \sim g(x)$

Review: Importance sampling

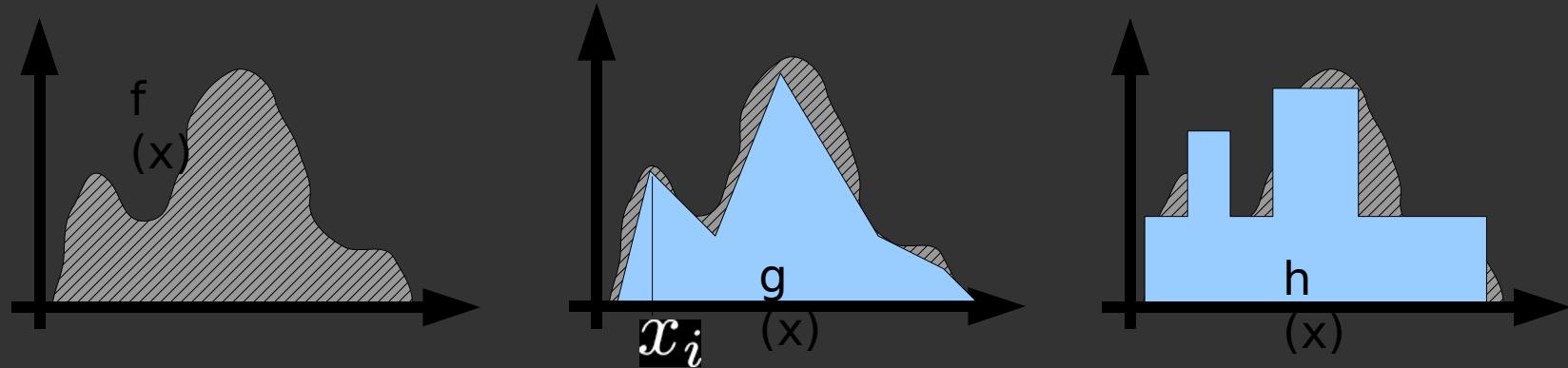
$$\int_{\mathcal{D}} f(x) \, dx \approx \frac{1}{N} \int_{\mathcal{D}} g(x) \, dx \sum_{i=1}^N \frac{f(x_i)}{g(x_i)}$$



Which is a better importance function, g or h ?

Review: Importance sampling

$$\int_{\mathcal{D}} f(x) \, dx \approx \frac{1}{N} \int_{\mathcal{D}} g(x) \, dx \sum_{i=1}^N \frac{f(x_i)}{g(x_i)}$$



Which is a better importance function, g or h ?

What if $f(x)$ changes ?

Review: Steerable functions

- Transformed functions = linear combination of basis

$$g_T(x) = \langle s(T), b(x) \rangle$$

Diagram illustrating the components of a transformed function:

- Transformed Function: $g_T(x)$
- Inner Product: $\langle \cdot, \cdot \rangle$
- bases: $s(T)$ and $b(x)$
- Transformation-dependent coefficients: implied by the inner product

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graph TD; TF[Transformed Function] --> gT["g_T(x)"]; IP[Inner Product] --> inner["⟨ s(T), b(x) ⟩"]; bases[bases] --> b["b(x)"]; TDC[Transformation-dependent coefficients] --> coeff["coefficients"];
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Direct, distant illumination

Reflected radiance along direction ω_o

$$\int_{\mathcal{S}^2} \frac{V(x, \omega_i) \rho(\omega_o, \omega_i) L(\omega_i) \max(\omega_i \cdot \mathbf{n}, 0)}{} d\omega_i$$

Visibility

Reflectance
Function

Incident
Radiance

Clamped
Cosine

Direct, distant illumination

Reflected radiance along direction ω_o

$$\int_{\mathcal{S}^2} \frac{V(x, \omega_i) \rho(\omega_o, \omega_i)}{S^2} L(\omega_i) \max(\omega_i \cdot \mathbf{n}, 0) d\omega_i$$

Visibility

Reflectance
Function

Incident
Radiance \times Clamped
Cosine

Importance
Function

Domain Partitioning

$$\int \gamma(x, \omega_i, \omega_r, \mathbf{n}) d\omega_i$$


Partition into Spherical Triangles

$$\left(\int + \int + \int + \dots \right) \gamma(x, \omega_i, \omega_r, \mathbf{n}) d\omega_i$$


Change of variables – 1

$$\int \gamma(x, \omega_i, \omega_r, \mathbf{n}) d\omega_i$$

$$\downarrow$$

Spherical to planar triangle

$$\int \gamma(x, \omega_i, \omega_r, \mathbf{n}) \varphi(\omega_i, \mathbf{n}_\Delta) d\omega_i$$


Planar triangle normal

Change of variables – 2

$$\int \gamma(x, \omega_i, \omega_r, \mathbf{n}) d\omega_i$$

$$\downarrow$$

Spherical to planar triangle

$$\int \gamma(x, \omega_i, \omega_r, \mathbf{n}) \varphi(\omega_i, \mathbf{n}_\Delta) d\omega_i$$

$$\downarrow$$

Unit square to triangle parameterization

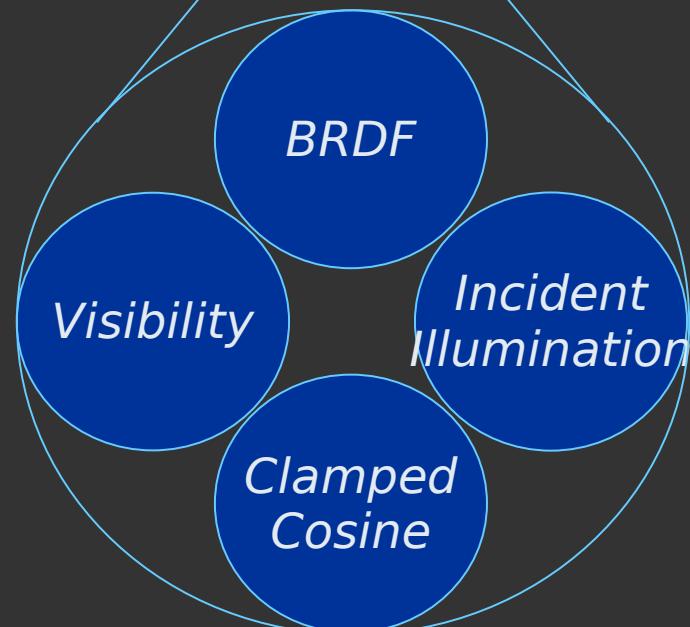
$$\int_0^1 \int_0^1 \gamma(x, \omega_i, \omega_r, \mathbf{n}) \varphi(p_0, p_1, \mathbf{n}_\Delta) |J(p_0, p_1)| dp_0 dp_1$$

$$\searrow$$

Jacobian of Parameterization

Novel parameterization

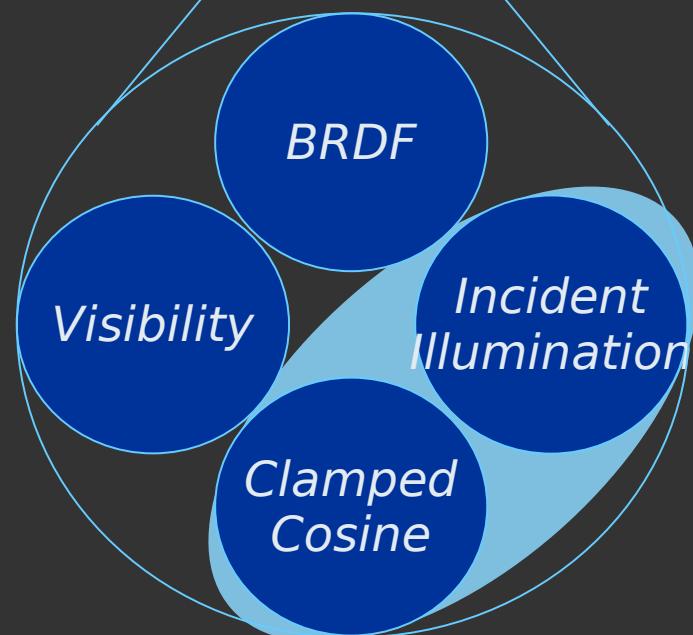
$$\int_0^1 \int_0^1 \boxed{\gamma(x, \omega_i, \omega_r, \mathbf{n})} \varphi(p_0, p_1, \mathbf{n}_\Delta) |J(p_0, p_1)| dp_0 dp_1$$



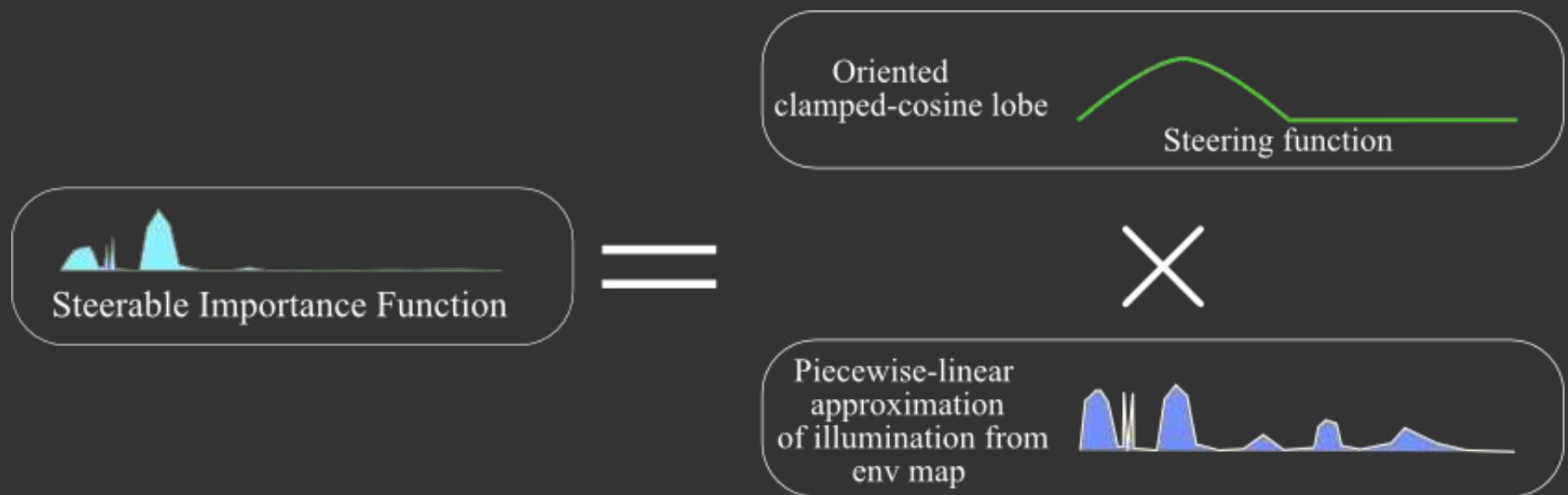
Novel parameterization

$$\int_0^1 \int_0^1 \boxed{\gamma(x, \omega_i, \omega_r, \mathbf{n})} \varphi(p_0, p_1, \mathbf{n}_\Delta) \boxed{|J(p_0, p_1)|} dp_0 dp_1$$

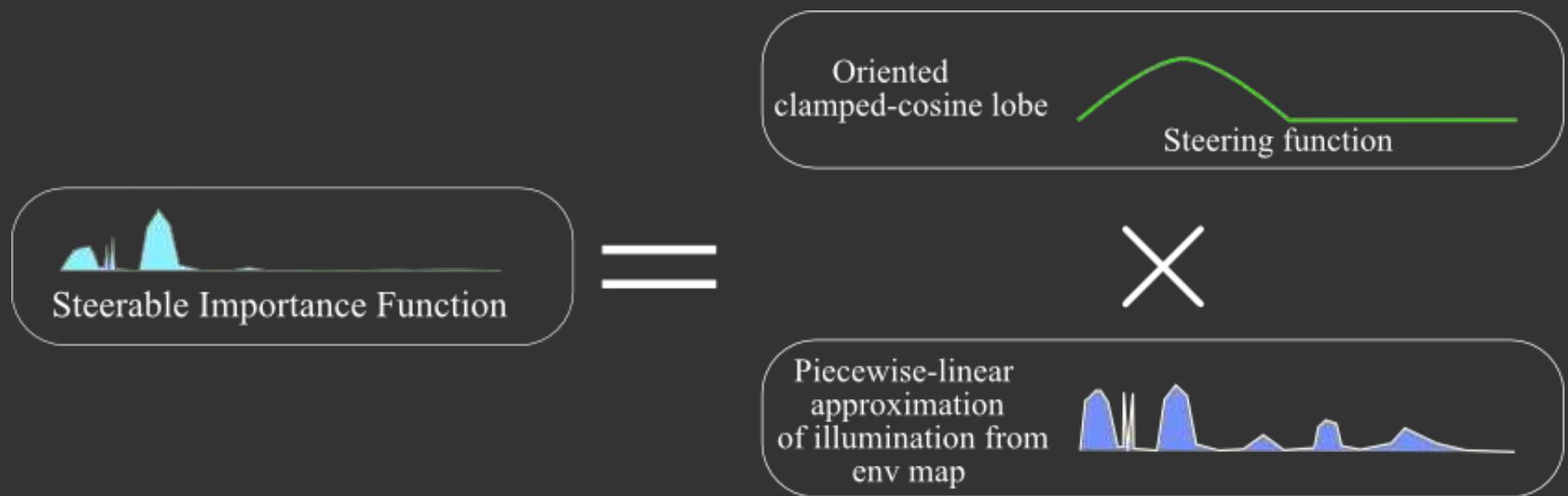
Derive parameterization so that
Jacobian \approx Illumination * Clamped Cosine



Steerable importance function



Steerable importance function



Is this steerable?

Steerable importance function

$$g_n(\mathbf{u}) = L(\mathbf{u}) \max(\mathbf{n} \cdot \mathbf{u}, 0)$$

Steerable importance function

$$g_n(\mathbf{u}) = L(\mathbf{u}) \max(\mathbf{n} \cdot \mathbf{u}, 0)$$

Represent using SH bases
8 coefficients – good approximation
[Ramamoorthi & Hanrahan]

$\langle \mathbf{a}(n), Y(u) \rangle$

Rotated coefficients Spherical Harmonic bases

Steerable importance function

rewrite

$$\begin{aligned} g_n(\mathbf{u}) &= L(\mathbf{u}) \max(\mathbf{n} \cdot \mathbf{u}, 0) \\ &\quad \downarrow \\ &< \mathbf{a}(n), Y(\mathbf{u}) > \\ g_n(\mathbf{u}) &= < \mathbf{a}(n), L(\mathbf{u}) Y(\mathbf{u}) > \end{aligned}$$

Steerable importance function

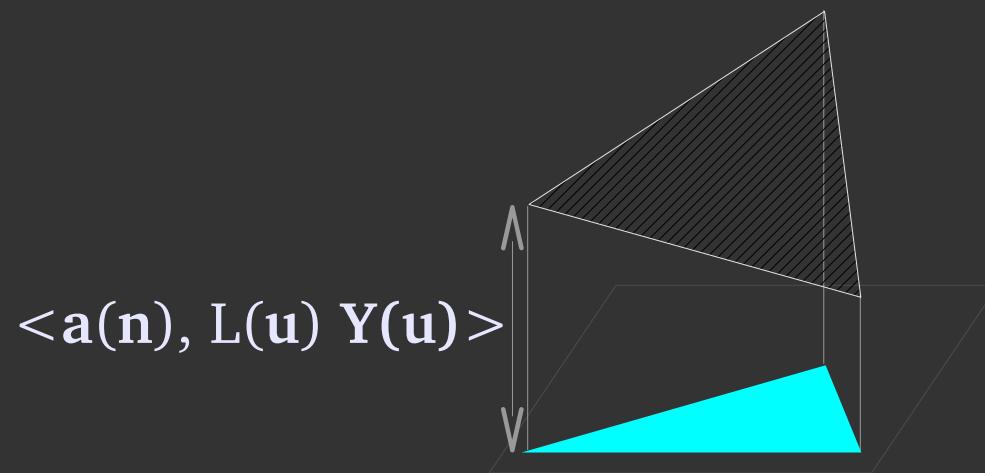
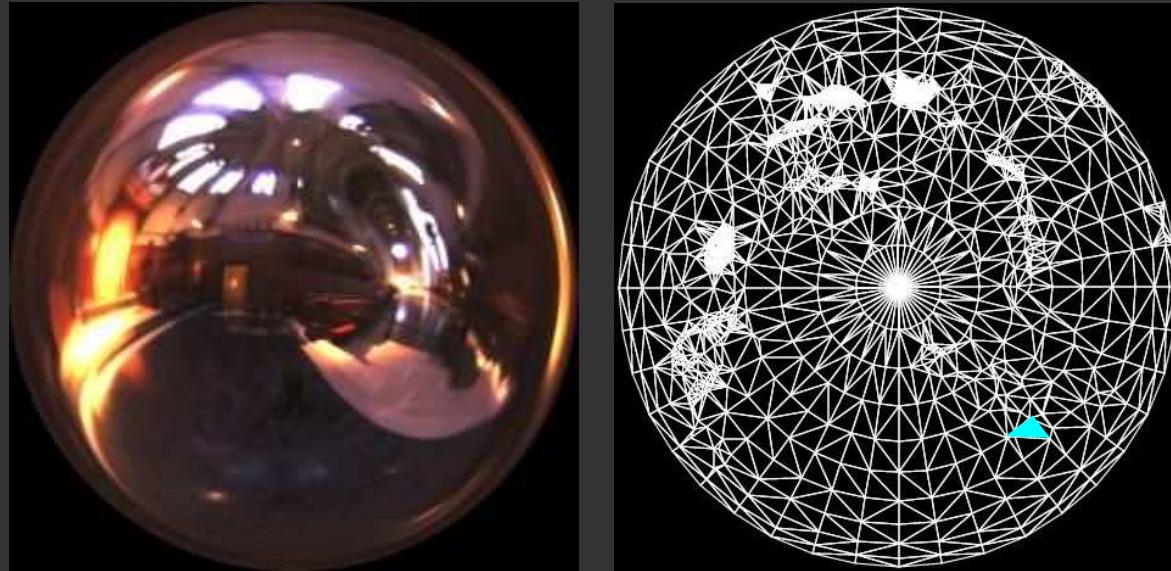
$$g_n(\mathbf{u}) = L(\mathbf{u}) \max(\mathbf{n} \cdot \mathbf{u}, 0)$$

↓

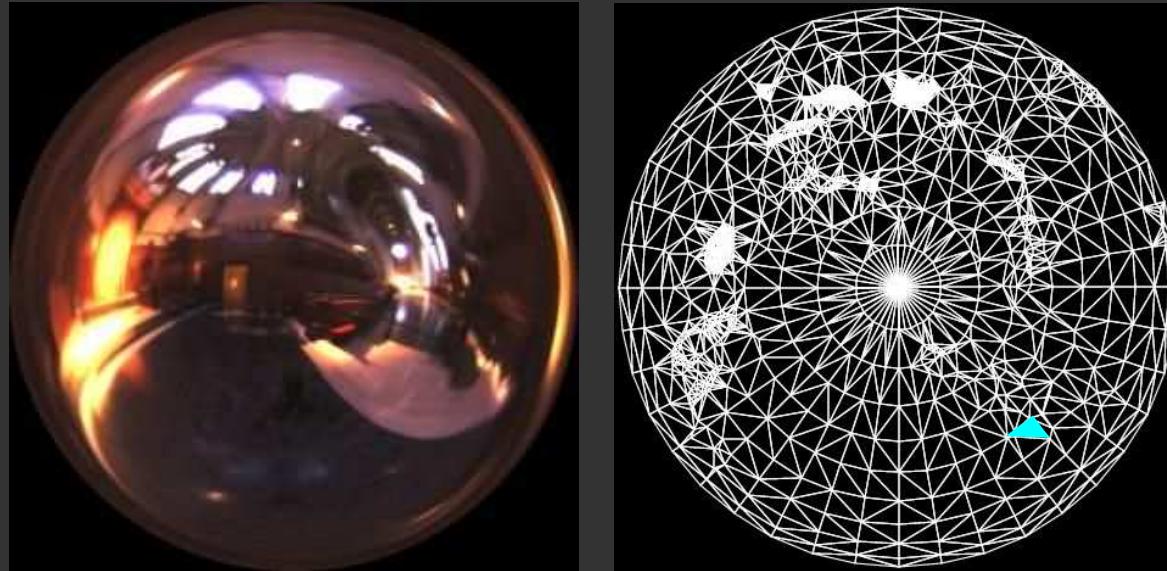
$$\langle \mathbf{a}(n), Y(\mathbf{u}) \rangle$$

→ $g_n(\mathbf{u}) = \langle \mathbf{a}(n), L(\mathbf{u}) Y(\mathbf{u}) \rangle$
Precomputed

Steerable importance function



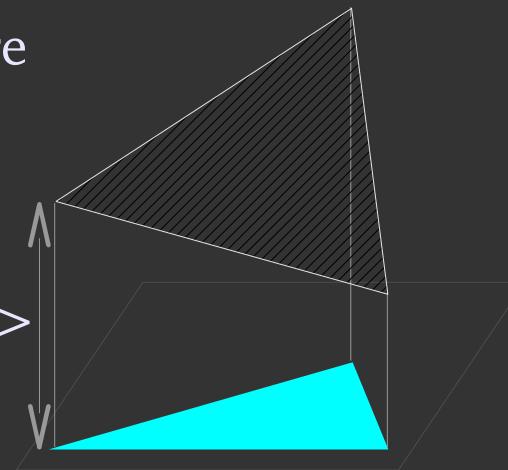
Steerable importance function



Precompute and store
(per vertex)

Function of normal

$$\langle \mathbf{a}(\mathbf{n}), \mathbf{L}(\mathbf{u}) \mathbf{Y}(\mathbf{u}) \rangle$$



Drawing samples

- Triangle selection
- Stratified sampling of selected triangle

Drawing samples

- Triangle selection
 - proportional to function integral within triangle
 - $O(\log N)$ cost (N triangles)
- Stratified sampling of selected triangle
 - according to linear function
 - $O(1)$ cost

Results

Environment map

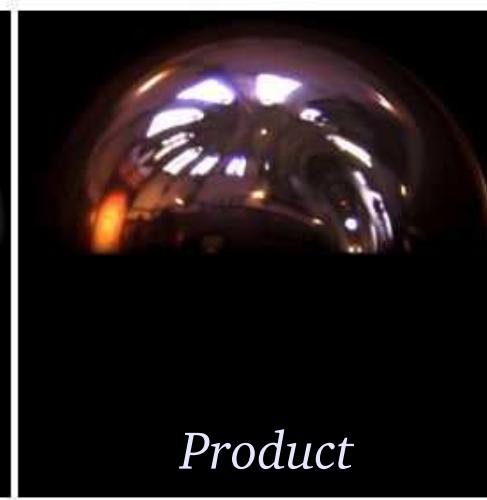


Input

Clamped cosine



Product

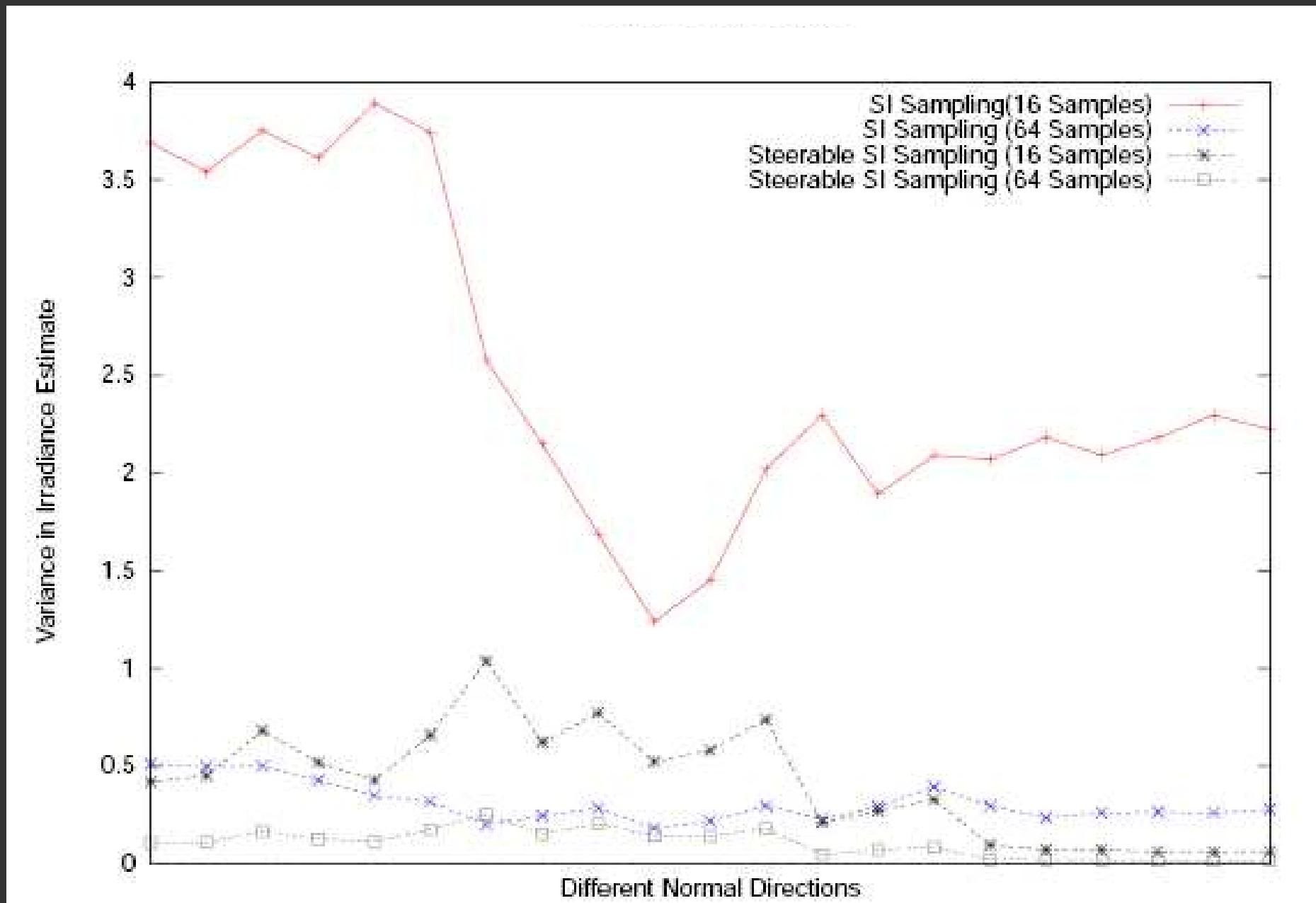


Samples (green)

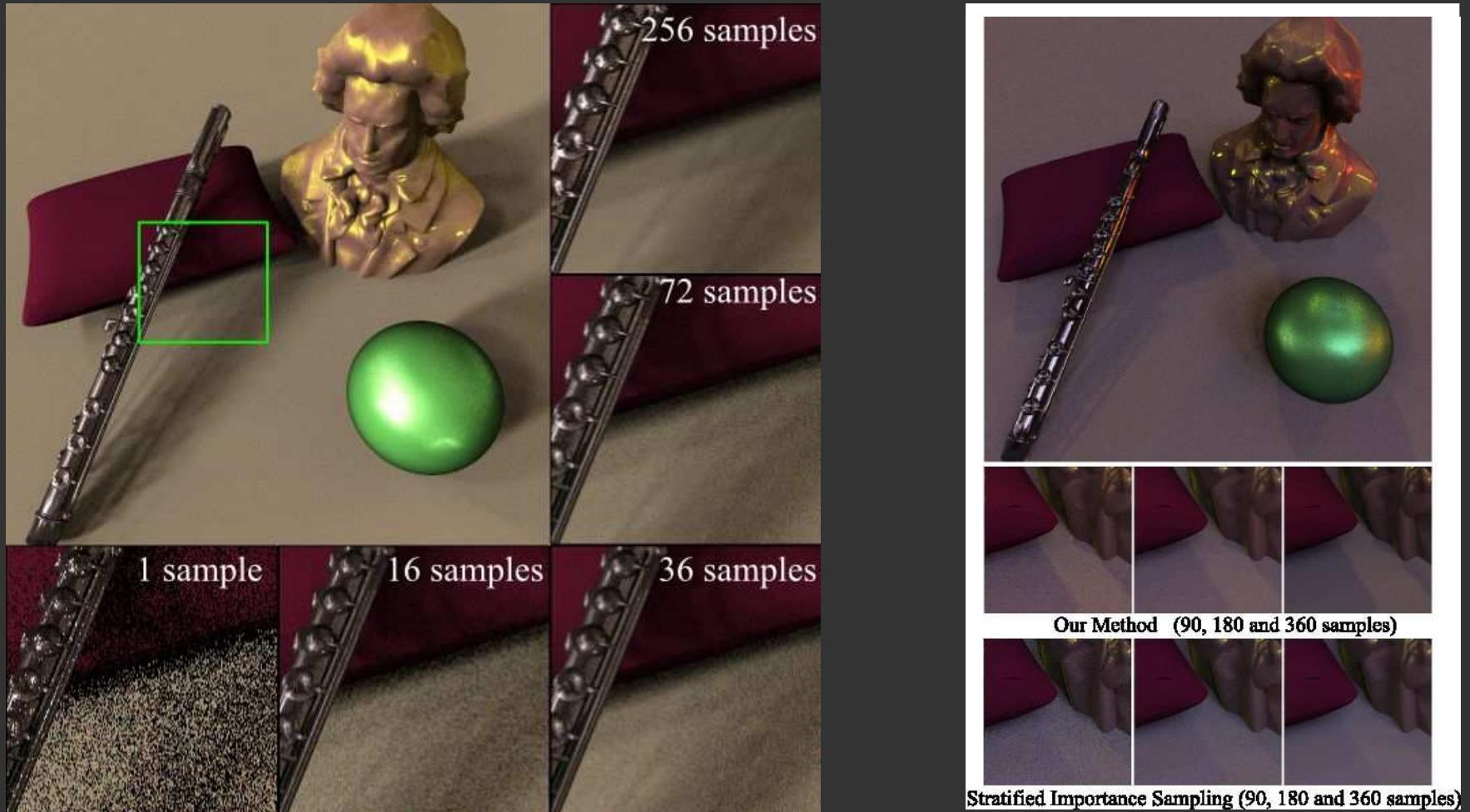


Output

Results: Reduced variance



Results: Images generated



Questions ?

[Ramamoorthi & Hanrahan]

An efficient representation for irradiance environment maps.
SIGGRAPH 2001.

[Teo]

Theory and applications of steerable functions.
PhD thesis, 1998.

[W. Freeman]

Steerable filters and the local analysis of image structure.
PhD thesis, 1992.