

Rendering Tutorial 4: Precomputed Radiance Transfer (预计算辐射传输)



徐昆

清华大学

2018-10-11

Rendering under environment lighting

$$L(\mathbf{o}) = \int_{\Omega} L(\mathbf{i})V(\mathbf{i})\rho(\mathbf{i}, \mathbf{o}) \max(0, \mathbf{n} \cdot \mathbf{i}) d\mathbf{i}$$

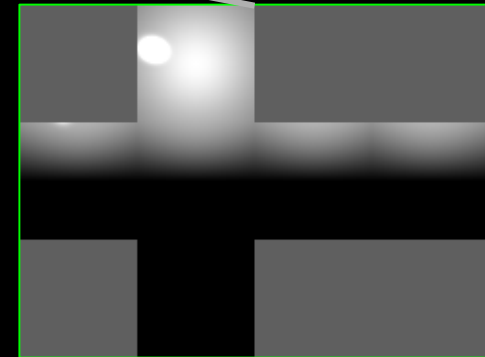
lighting



visibility



BRDF



- ◎ \mathbf{i}/\mathbf{o} : incoming/view directions
- ◎ Brute-force computation
 - Resolution: $6*64*64$
 - Needs $6*64*64$ times for each point!



Precomputed Radiance Transfer (PRT)

- ◎ Introduced by Sloan in SIGGRAPH 2002
 - *Precomputed Radiance Transfer for Real-Time Rendering in Dynamic, Low-Frequency Lighting Environments* [Sloan 02]



Basic idea of PRT [Sloan 02]

$$L(\mathbf{o}) = \int_{\Omega} L(\mathbf{i}) V(\mathbf{i}) \rho(\mathbf{i}, \mathbf{o}) \max(0, \mathbf{n} \cdot \mathbf{i}) d\mathbf{i}$$

lighting light transport

- ⊙ Approximate lighting using basis functions
 - $L(\mathbf{i}) \approx \sum l_i B_i(\mathbf{i})$
- ⊙ Precomputation stage
 - compute light transport, and project to basis function space
- ⊙ Runtime stage
 - dot product (diffuse) or matrix-vector multiplication (glossy)

Diffuse Case

$$L(\mathbf{o}) = \rho \int_{\Omega} \underline{L(\mathbf{i})} V(\mathbf{i}) \max(0, \mathbf{n} \cdot \mathbf{i}) d\mathbf{i}$$

$$L(\mathbf{i}) \approx \sum l_i B_i(\mathbf{i})$$

lighting coefficient basis function



$$L(\mathbf{o}) \approx \rho \sum l_i \int_{\Omega} B_i(\mathbf{i}) V(\mathbf{i}) \max(0, \mathbf{n} \cdot \mathbf{i}) d\mathbf{i}$$



Precompute

$$L(\mathbf{o}) \approx \rho \sum l_i T_i$$

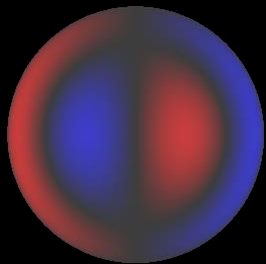
◎ Reduce rendering computation to dot product

Basis functions $B(\mathbf{i})$

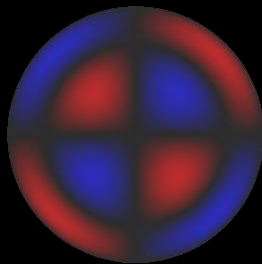
◎ Spherical Harmonics (SH)

◎ SH have nice properties:

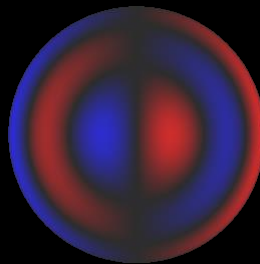
- orthonormal
- simple projection/reconstruction
- rotationally invariant (no aliasing)
- simple rotation
- simple convolution
- few basis functions: low freqs



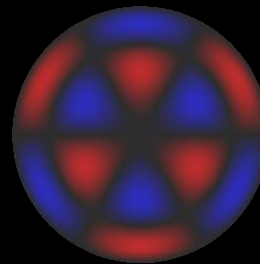
$l=2, m=1$



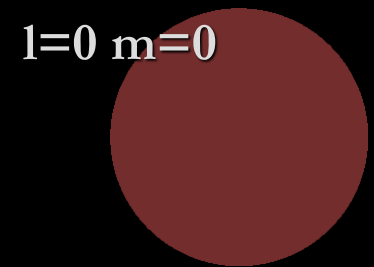
$l=3, m=-1$



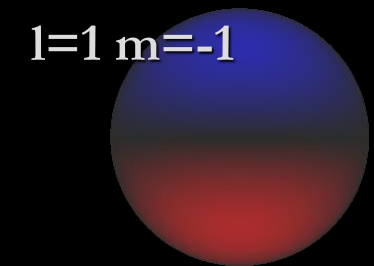
$l=3, m=2$



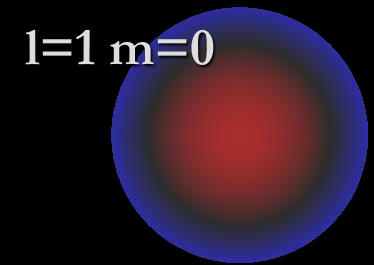
$l=4, m=-2$



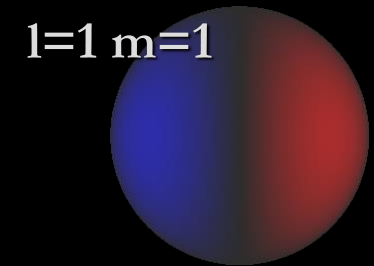
$l=0, m=0$



$l=1, m=-1$



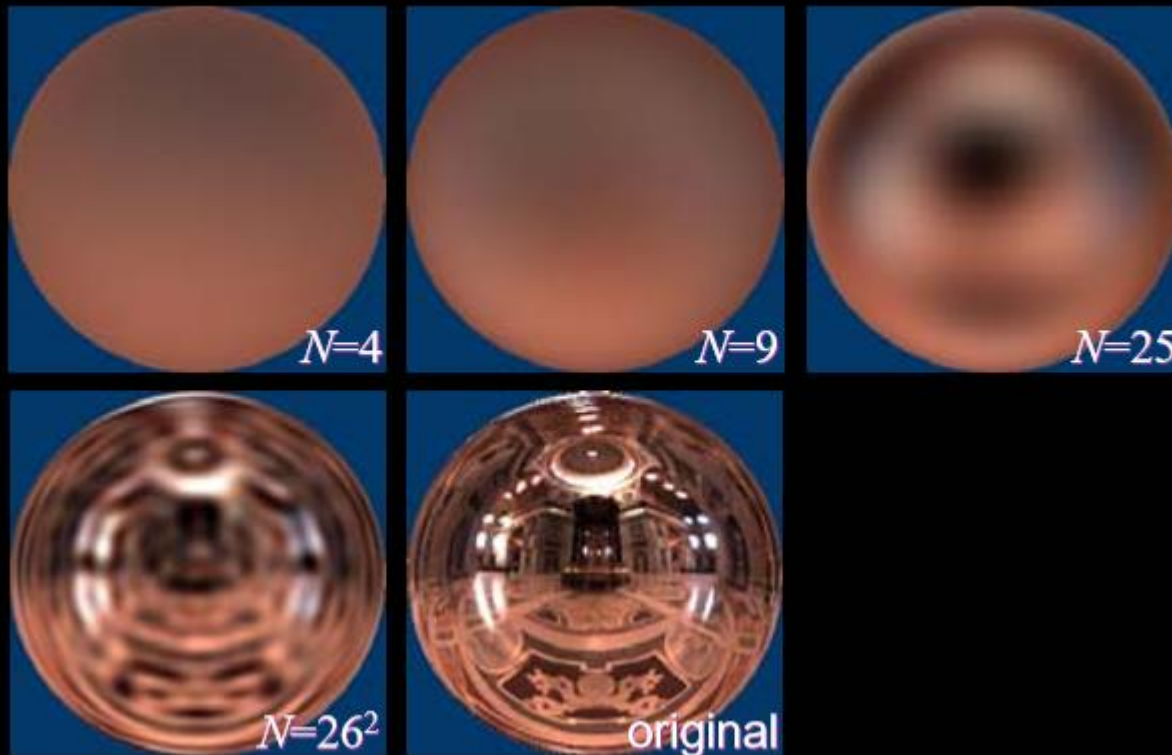
$l=1, m=0$



$l=1, m=1$

Basis functions $B(\mathbf{i})$

- ◎ Spherical Harmonics (SH)
- ◎ Light Approximation Examples



Low frequency

Basis functions $B(\mathbf{i})$

© SH is orthonormal, we have:

$$\int_{\Omega} B_i(\mathbf{i}) \cdot B_j(\mathbf{i}) d\mathbf{i} = \mathbf{1} \quad (\mathbf{i} = j)$$

$$\int_{\Omega} B_i(\mathbf{i}) \cdot B_j(\mathbf{i}) d\mathbf{i} = \mathbf{0} \quad (\mathbf{i} \neq j)$$

Basis functions $B(\mathbf{i})$

Original space



lighting

SH space



lighting coefficients

$$L(\mathbf{i}) \approx \sum l_i B_i(\mathbf{i})$$

⊙ Projection to SH space

$$l_i = \int_{\Omega} L(\mathbf{i}) \cdot B_i(\mathbf{i}) d\mathbf{i}$$

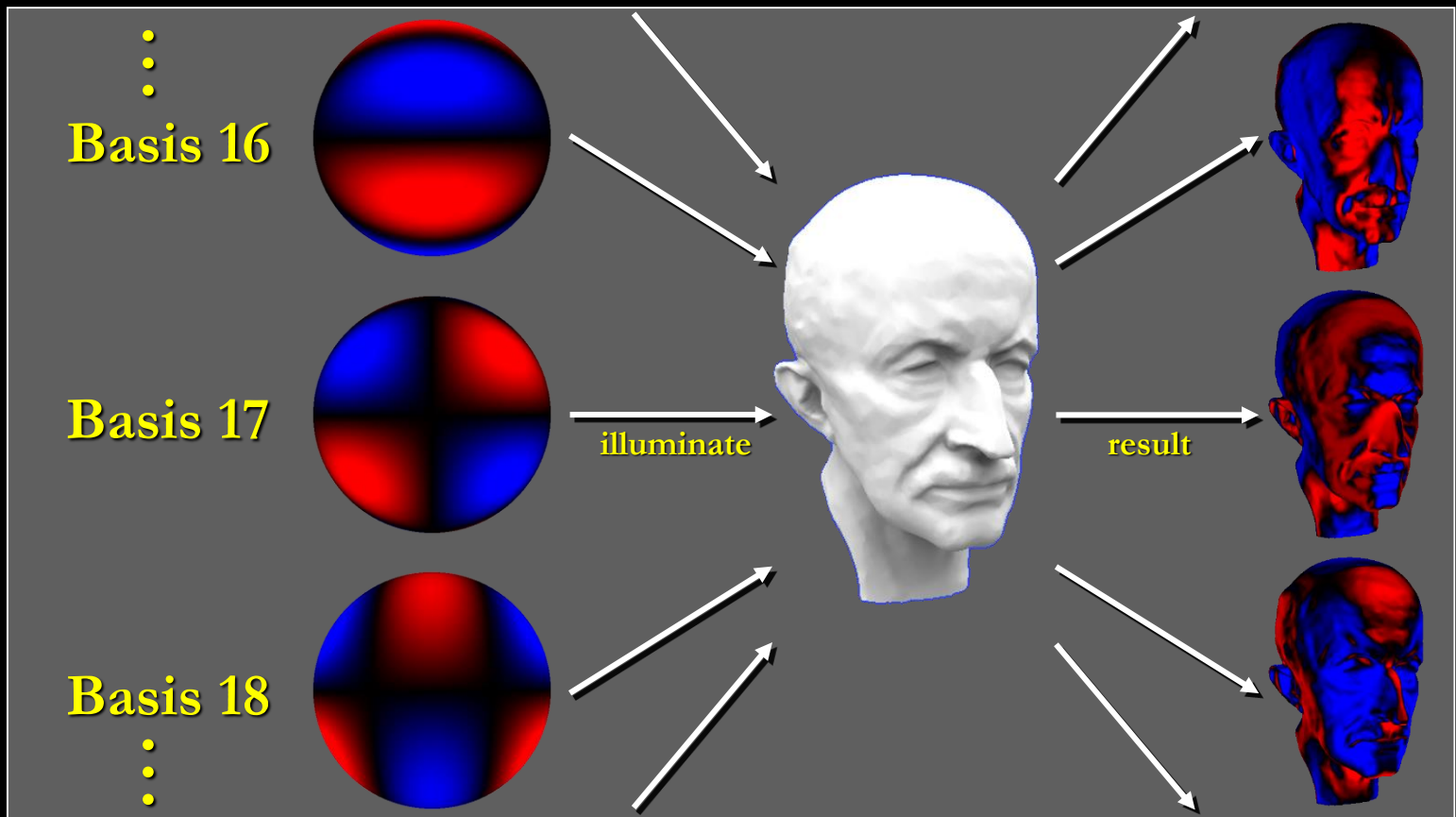
⊙ Reconstruction

$$L(\mathbf{i}) \approx \sum l_i B_i(\mathbf{i})$$

Precomputation

light transport $T_i \approx \int_{\Omega} B_i(\mathbf{i}) V(\mathbf{i}) \max(0, \mathbf{n} \cdot \mathbf{i}) d\mathbf{i}$

⊙ No shadow/ **shadow**/ inter-reflection



Run-time Rendering

$$L(\mathbf{o}) \approx \rho \sum l_i T_i$$

- ◎ Rendering at each point is reduced to a dot product
 - First, project the lighting to the basis to obtain l_i
 - Or, rotate the lighting instead of re-projection
 - Then, compute the dot product
- ◎ Real-time: easily implemented in shader

Diffuse Rendering Results



No Shadows



Shadows



Shadows+Inter

Glossy Case

$$L(\mathbf{o}) = \int_{\Omega} L(\mathbf{i})V(\mathbf{i})\rho(\mathbf{i}, \mathbf{o}) \max(0, \mathbf{n} \cdot \mathbf{i}) d\mathbf{i}$$

$$L(\mathbf{o}) \approx \sum l_i \underline{T_i(\mathbf{o})}$$

$$L(\mathbf{o}) \approx \sum (\sum l_i t_{ij}) B_j(\mathbf{o})$$

$$T_i(\mathbf{o}) \approx \sum t_{ij} B_j(\mathbf{o})$$

transport matrix basis function

reflected radiance

coefficient

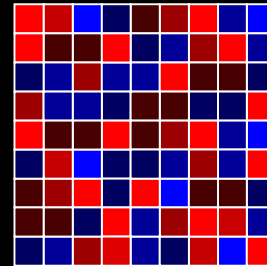


≈

light coefficient



*



transport matrix

© Rendering: vector-matrix multiplication

Time Complexity

◎ #SH Basis : 9/**16**/25

◎ Diffuse Rendering

- At each point: dot-product of size 16

◎ Glossy Rendering

- At each point: $\text{vector}(16) * \text{matrix}(16*16)$

Glossy Rendering Results



No Shadows/Inter



Shadows



Shadows+Inter

- Glossy object, 50K mesh
- Runs at 3.6 fps on 2.2Ghz P4, ATI Radeon 8500

Interreflections and Caustics

interreflections



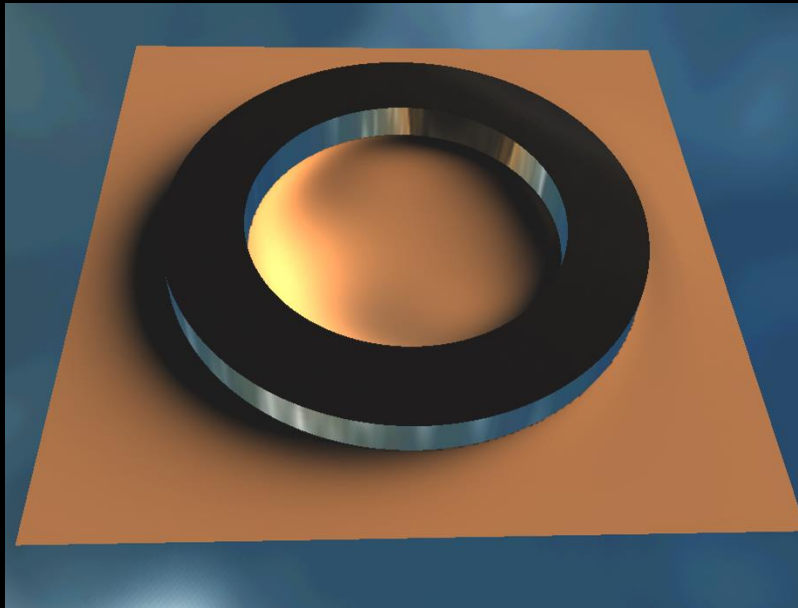
none



1 bounce



2 bounces



caustics

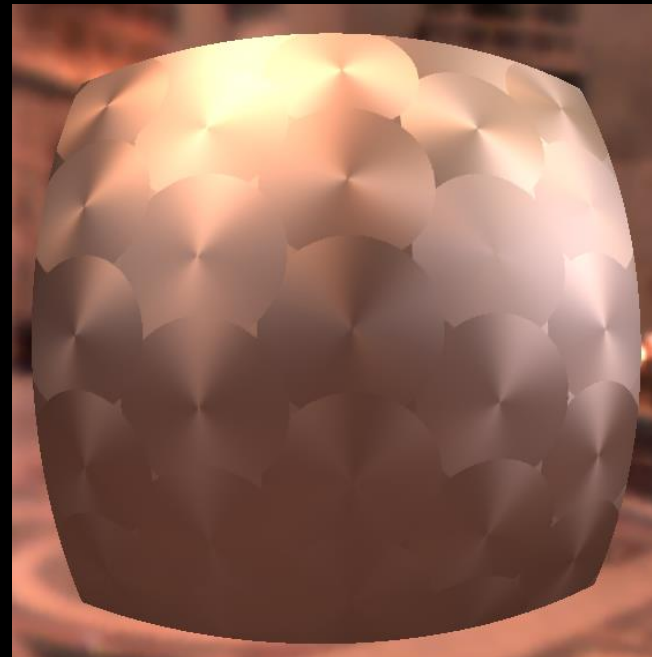
Transport Paths

LP
Runtime is independent
 LGP
of transport complexity

$$L(D|G)^*P$$

$$LS^*(S^*(D|G))^*P$$

Arbitrary BRDF Results



Anisotropic BRDFs

Other BRDFs

Spatially Varying

Results

Acquired Environments

Geometry: 50k vertex mesh

Summary of [Sloan 02]

- ◎ Approximate Lighting and light transport using basis functions (SH)
 - Lighting -> lighting coefficients
 - light transport -> coefficients / matrices
- ◎ Precompute and store light transport
- ◎ Rendering reduced to:
 - Diffuse: dot product
 - Glossy: vector matrix multiplication

Limitations [Sloan 02]

- ◎ Low-frequency
 - Due to the nature of SH
- ◎ Dynamic lighting, but static scene/material
 - Changing scene/material invalidates precomputed light transport
- ◎ Big precomputation data

Follow up works

- ◎ More basis functions
- ◎ dot product => triple products
- ◎ Static scene => dynamic scene
- ◎ Fix material => dynamic material
- ◎ Other effects: translucent, hair, ...
- ◎ Precomputation => analytic computation
- ◎ ...

More basis functions

- ◎ Spherical Harmonics (SH)
- ◎ Wavelet
- ◎ Zonal Harmonics
- ◎ Spherical Gaussian (SG)
- ◎ Piecewise Constant

Wavelet [Ng 03]

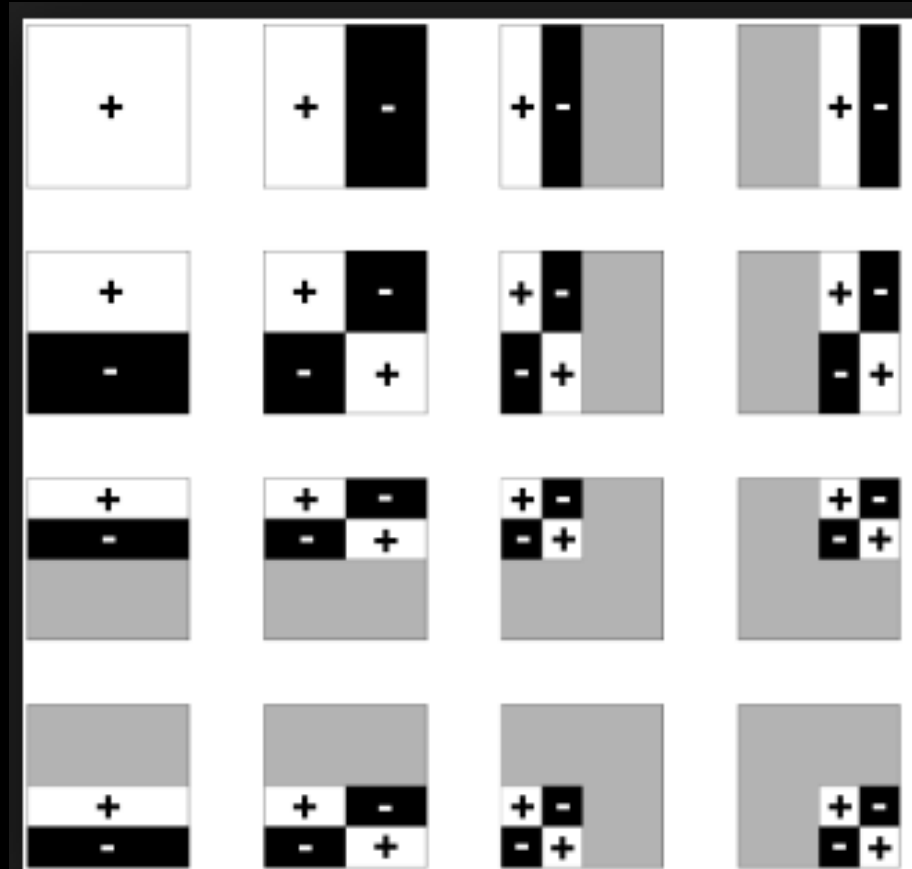
◎ 2D Haar wavelet

◎ Projection:

- Wavelet Transformation
- Retain a small number of non-zero coefficients

◎ A non-linear approximation

◎ All-frequency representation



low frequency vs all frequency

Teapot in Grace Cathedral



Low frequency (SH)



All frequency (Wavelet)

Relighting as Matrix-Vector Multiply

$$\begin{bmatrix} P_1 \\ P_2 \\ P_3 \\ M \\ P_N \end{bmatrix}$$

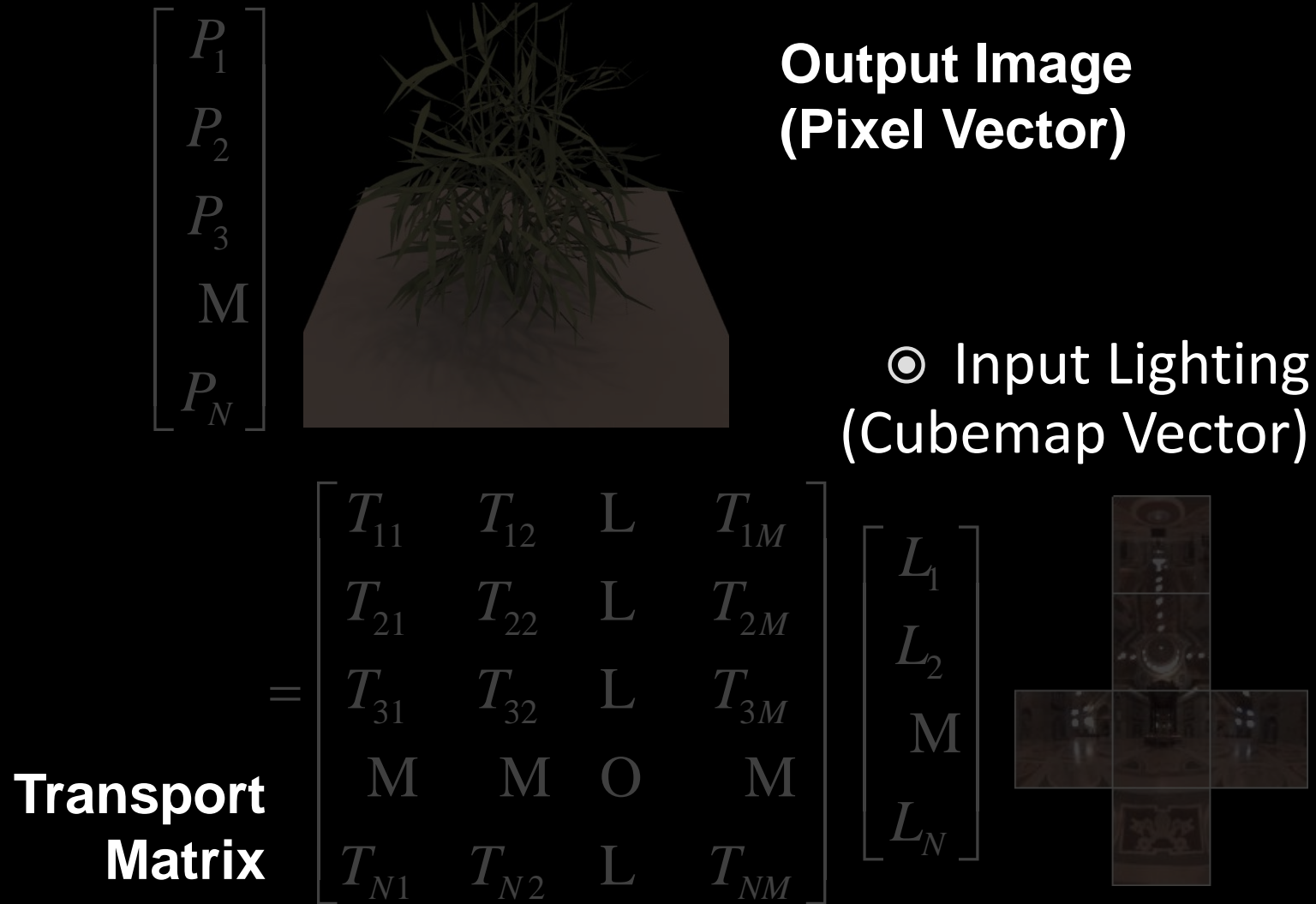


$$= \begin{bmatrix} T_{11} & T_{12} & \mathbf{L} & T_{1M} \\ T_{21} & T_{22} & \mathbf{L} & T_{2M} \\ T_{31} & T_{32} & \mathbf{L} & T_{3M} \\ \mathbf{M} & \mathbf{M} & \mathbf{O} & \mathbf{M} \\ T_{N1} & T_{N2} & \mathbf{L} & T_{NM} \end{bmatrix}$$

$$\begin{bmatrix} L_1 \\ L_2 \\ M \\ L_N \end{bmatrix}$$



Relighting as Matrix-Vector Multiply



Non-linear Wavelet Light Approximation

Wavelet Transform



Non-linear Wavelet Light Approximation

$$\begin{bmatrix} 0 \\ L_2 \\ 0 \\ 0 \\ 0 \\ L_6 \\ M \\ 0 \end{bmatrix}$$



**Non-linear
Approximation**

Retain 0.1% – 1% terms

Matrix Row Wavelet Encoding

$$\begin{bmatrix} T_{11} & T_{12} & T_{13} & T_{14} & L & T_{1M} \\ T_{21} & T_{22} & T_{23} & T_{24} & L & T_{2M} \\ T_{31} & T_{32} & T_{24} & T_{34} & L & T_{3M} \\ T_{41} & T_{42} & T_{43} & T_{44} & L & T_{4M} \\ T_{51} & T_{52} & T_{53} & T_{54} & L & T_{5M} \\ T_{61} & T_{62} & T_{63} & T_{64} & L & T_{6M} \\ M & M & M & M & O & T_{7M} \\ T_{N1} & T_{N2} & T_{N3} & T_{N4} & L & T_{NM} \end{bmatrix}$$

Matrix Row Wavelet Encoding

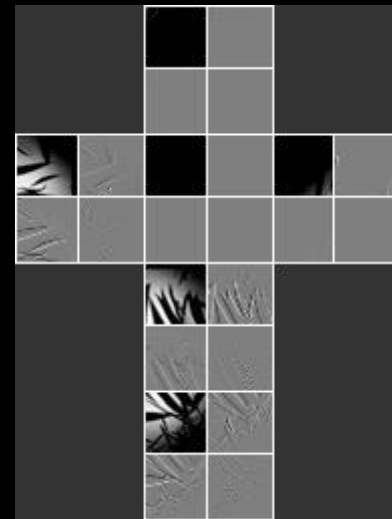
$$\begin{bmatrix} T_{11} & T_{12} & T_{13} & T_{14} & L & T_{1M} \\ T_{21} & T_{22} & T_{23} & T_{24} & L & T_{2M} \\ T_{31} & T_{32} & T_{24} & T_{34} & L & T_{3M} \\ T_{41} & T_{42} & T_{43} & T_{44} & L & T_{4M} \\ T_{51} & T_{52} & T_{53} & T_{54} & L & T_{5M} \\ T_{61} & T_{62} & T_{63} & T_{64} & L & T_{6M} \\ M & M & M & M & O & T_{7M} \\ T_{N1} & T_{N2} & T_{N3} & T_{N4} & L & T_{NM} \end{bmatrix}$$



Extract Row

Matrix Row Wavelet Encoding

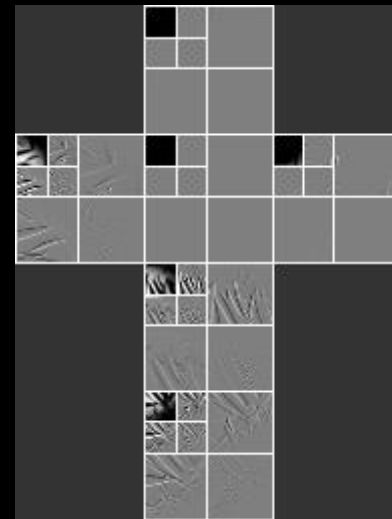
$$\begin{bmatrix} T_{11} & T_{12} & T_{13} & T_{14} & L & T_{1M} \\ T_{21} & T_{22} & T_{23} & T_{24} & L & T_{2M} \\ T_{31} & T_{32} & T_{24} & T_{34} & L & T_{3M} \\ T_{41} & T_{42} & T_{43} & T_{44} & L & T_{4M} \\ T_{51} & T_{52} & T_{53} & T_{54} & L & T_{5M} \\ T_{61} & T_{62} & T_{63} & T_{64} & L & T_{6M} \\ M & M & M & M & O & T_{7M} \\ T_{N1} & T_{N2} & T_{N3} & T_{N4} & L & T_{NM} \end{bmatrix}$$



Wavelet Transform

Matrix Row Wavelet Encoding

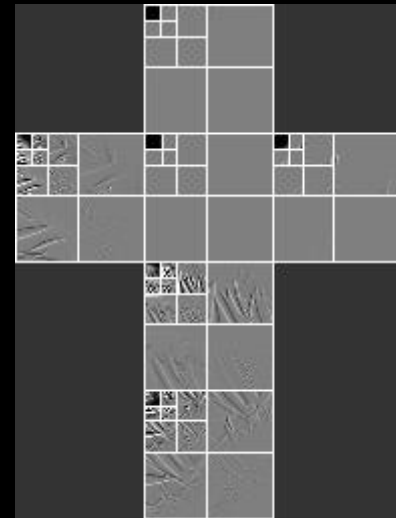
$$\begin{bmatrix} T_{11} & T_{12} & T_{13} & T_{14} & L & T_{1M} \\ T_{21} & T_{22} & T_{23} & T_{24} & L & T_{2M} \\ T_{31} & T_{32} & T_{24} & T_{34} & L & T_{3M} \\ T_{41} & T_{42} & T_{43} & T_{44} & L & T_{4M} \\ T_{51} & T_{52} & T_{53} & T_{54} & L & T_{5M} \\ T_{61} & T_{62} & T_{63} & T_{64} & L & T_{6M} \\ M & M & M & M & O & T_{7M} \\ T_{N1} & T_{N2} & T_{N3} & T_{N4} & L & T_{NM} \end{bmatrix}$$



Wavelet Transform

Matrix Row Wavelet Encoding

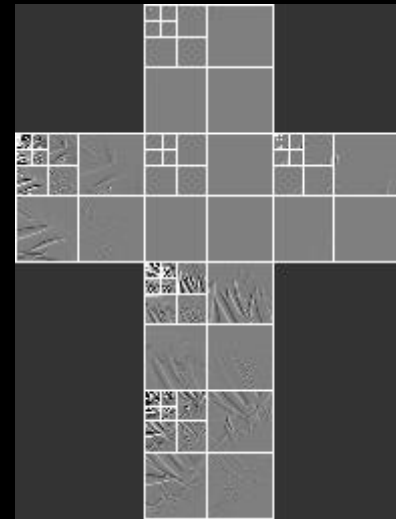
$$\begin{bmatrix} T_{11} & T_{12} & T_{13} & T_{14} & L & T_{1M} \\ T_{21} & T_{22} & T_{23} & T_{24} & L & T_{2M} \\ T_{31} & T_{32} & T_{24} & T_{34} & L & T_{3M} \\ T_{41} & T_{42} & T_{43} & T_{44} & L & T_{4M} \\ T_{51} & T_{52} & T_{53} & T_{54} & L & T_{5M} \\ T_{61} & T_{62} & T_{63} & T_{64} & L & T_{6M} \\ M & M & M & M & O & T_{7M} \\ T_{N1} & T_{N2} & T_{N3} & T_{N4} & L & T_{NM} \end{bmatrix}$$



Wavelet Transform

Matrix Row Wavelet Encoding

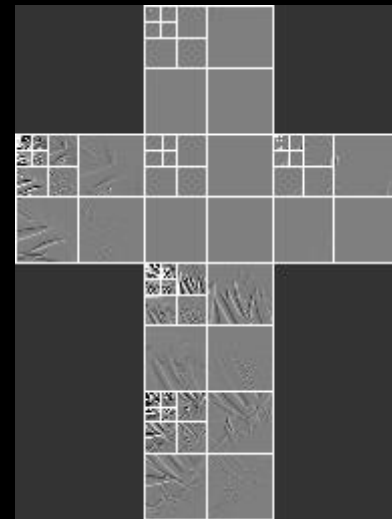
$$\begin{bmatrix} T_{11} & T_{12} & T_{13} & T_{14} & L & T_{1M} \\ T_{21} & T_{22} & T_{23} & T_{24} & L & T_{2M} \\ T_{31} & T_{32} & T_{24} & T_{34} & L & T_{3M} \\ T_{41} & T_{42} & T_{43} & T_{44} & L & T_{4M} \\ T_{51} & T_{52} & T_{53} & T_{54} & L & T_{5M} \\ T_{61} & T_{62} & T_{63} & T_{64} & L & T_{6M} \\ M & M & M & M & O & T_{7M} \\ T_{N1} & T_{N2} & T_{N3} & T_{N4} & L & T_{NM} \end{bmatrix}$$



Wavelet Transform

Matrix Row Wavelet Encoding

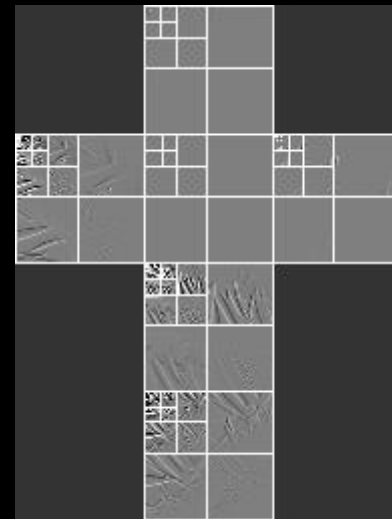
$$\begin{bmatrix} T'_{11} & 0 & 0 & T'_{14} & L & 0 \\ T_{21} & T_{22} & T_{23} & T_{24} & L & T_{2M} \\ T_{31} & T_{32} & T_{24} & T_{34} & L & T_{3M} \\ T_{41} & T_{42} & T_{43} & T_{44} & L & T_{4M} \\ T_{51} & T_{52} & T_{53} & T_{54} & L & T_{5M} \\ T_{61} & T_{62} & T_{63} & T_{64} & L & T_{6M} \\ M & M & M & M & O & T_{7M} \\ T_{N1} & T_{N2} & T_{N3} & T_{N4} & L & T_{NM} \end{bmatrix}$$



Store Back in Matrix

Matrix Row Wavelet Encoding

$$\begin{bmatrix} T_{11}' & 0 & 0 & T_{14}' & L & 0 \\ T_{21} & T_{22} & T_{23} & T_{24} & L & T_{2M} \\ T_{31} & T_{32} & T_{24} & T_{34} & L & T_{3M} \\ T_{41} & T_{42} & T_{43} & T_{44} & L & T_{4M} \\ T_{51} & T_{52} & T_{53} & T_{54} & L & T_{5M} \\ T_{61} & T_{62} & T_{63} & T_{64} & L & T_{6M} \\ M & M & M & M & O & T_{7M} \\ T_{N1} & T_{N2} & T_{N3} & T_{N4} & L & T_{NM} \end{bmatrix}$$



Only 3% – 30% are non-zero

Why Non-linear Approximation?

◎ Linear

- Use a **fixed** set of approximating functions
- Precomputed radiance transfer uses 25 - 100 of the lowest frequency spherical harmonics

◎ Non-linear

- Use a **dynamic** set of approximating functions (*depends on each frame's lighting*)
- In our case: choose 10's - 100's from a basis of 24,576 wavelets

Overall Rendering Algorithm

◎ Pre-compute (per scene)

- Compute matrix in pixel basis
- Wavelet transform rows
- Quantize, store

◎ Interactive Relighting (each frame)

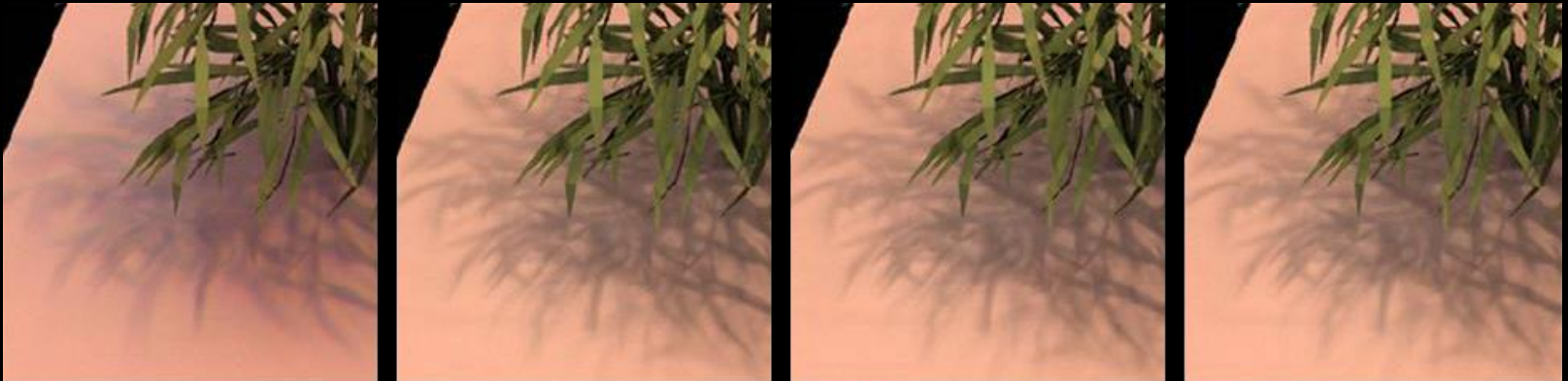
- Wavelet transform lighting
- Prioritize and retain N wavelet coefficients
- Perform sparse-matrix vector multiplication

Output Image Comparison

SH



wavelet



25

200

2,000

20,000

Limitation

- ◎ Wavelet: not rotational invariant
 - Re-projection at each frame
 - Results in flicking

- ◎ Only support dot-product operator
 - Limited to diffuse or fix-view glossy

Results



Zonal Harmonics [Sloan 05]

- ◎ circularly symmetric functions
- ◎ Subset of SH basis ($m=0$)
- ◎ Low-frequency
- ◎ Rotational invariant
- ◎ Much more faster in rotation than SH



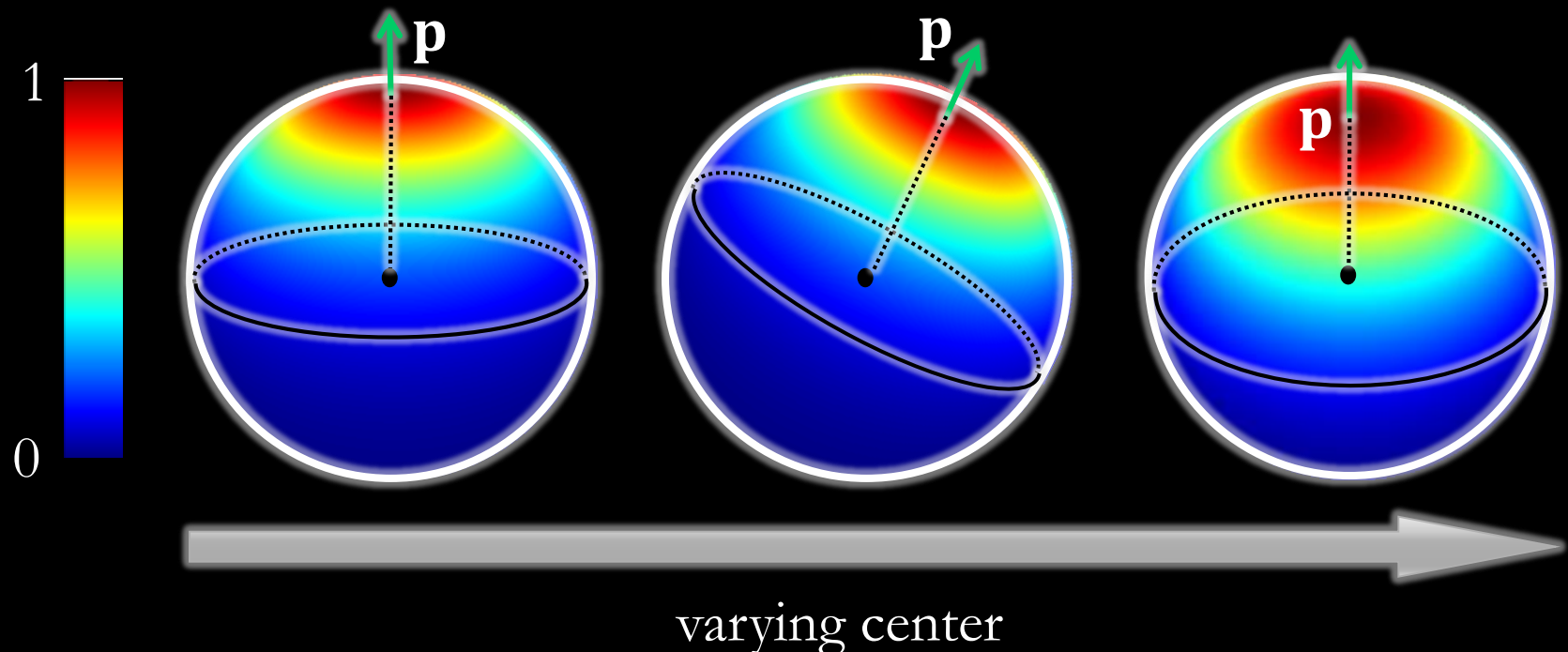
Spherical Gaussian (SG) [Tsai 06]

- SGs (or Spherical Radial Basis Functions, SRBFs)

$$G(\mathbf{v}; \mathbf{p}, \lambda) = e^{\lambda(\mathbf{v} \cdot \mathbf{p} - 1)}$$

center

bandwidth



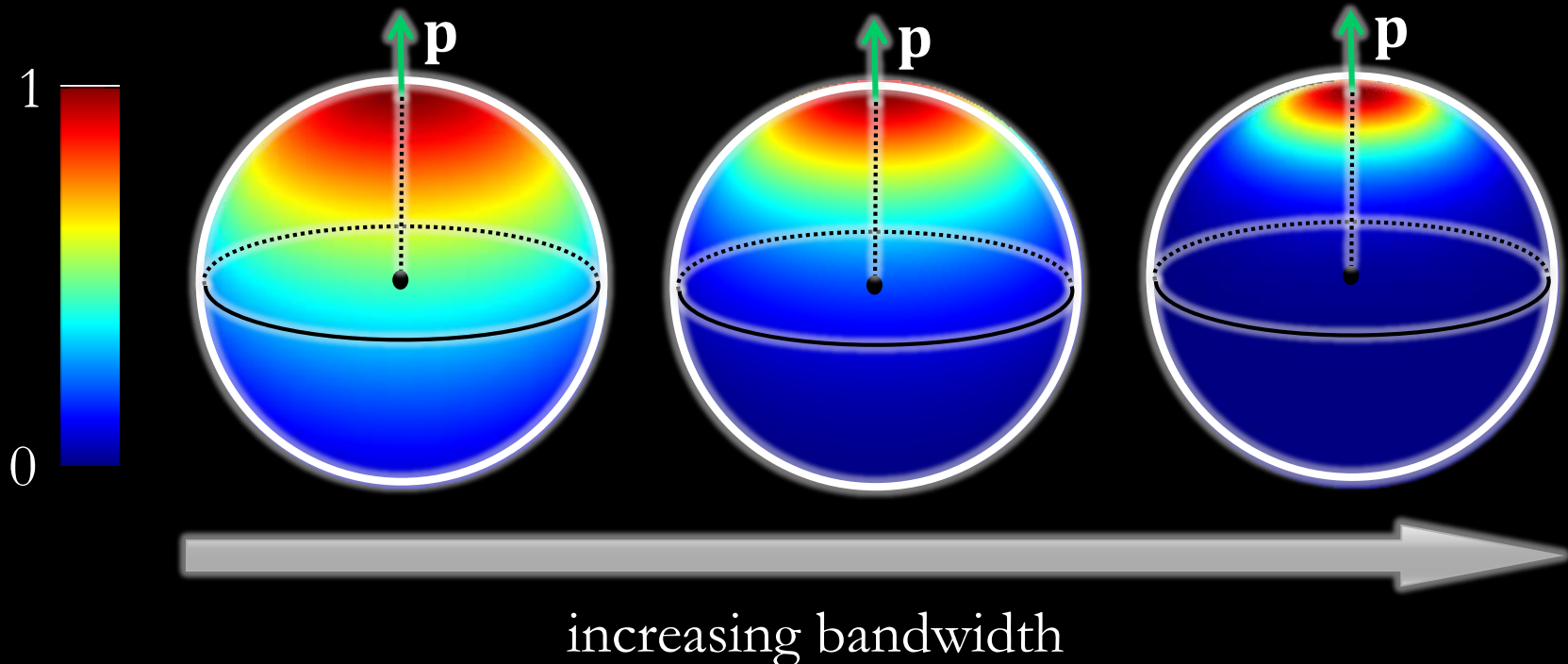
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$$G(\mathbf{v}; \mathbf{p}, \lambda) = e^{\lambda(\mathbf{v} \cdot \mathbf{p} - 1)}$$

center

bandwidth



Mathematical Properties of SGs

◎ Closed-form integral

- The integral of an SG is closed-form

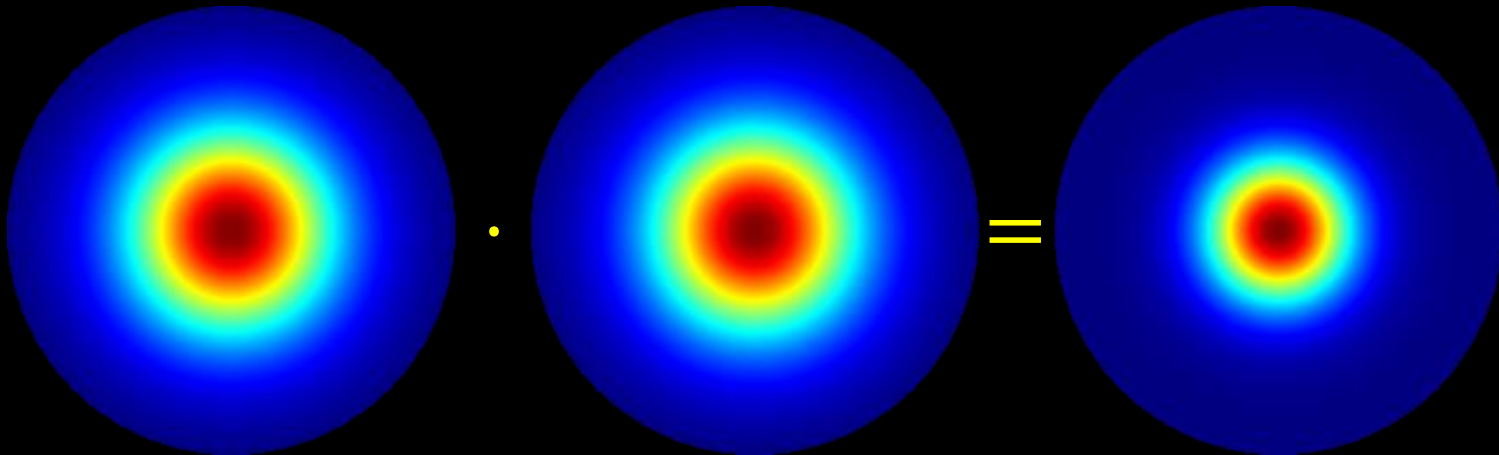
$$\int_{\Omega} G(\mathbf{v}; \mathbf{p}, \lambda) d\mathbf{v} = \frac{2\pi}{\lambda} (1 - e^{-2\lambda})$$

Mathematical Properties of SGs

◎ Closed under multiplication

- The product of two SGs is also an SG

$$G(\mathbf{v}; \mathbf{p}_1, \lambda_1) \cdot G(\mathbf{v}; \mathbf{p}_2, \lambda_2) = cG\left(\mathbf{v}; \frac{\lambda_1 \mathbf{p}_1 + \lambda_2 \mathbf{p}_2}{|\lambda_1 \mathbf{p}_1 + \lambda_2 \mathbf{p}_2|}, |\lambda_1 \mathbf{p}_1 + \lambda_2 \mathbf{p}_2|\right)$$



$G_{iso}(\mathbf{v}; \mathbf{p}_1, \lambda_1)$

$G_{iso}(\mathbf{v}; \mathbf{p}_2, \lambda_2)$

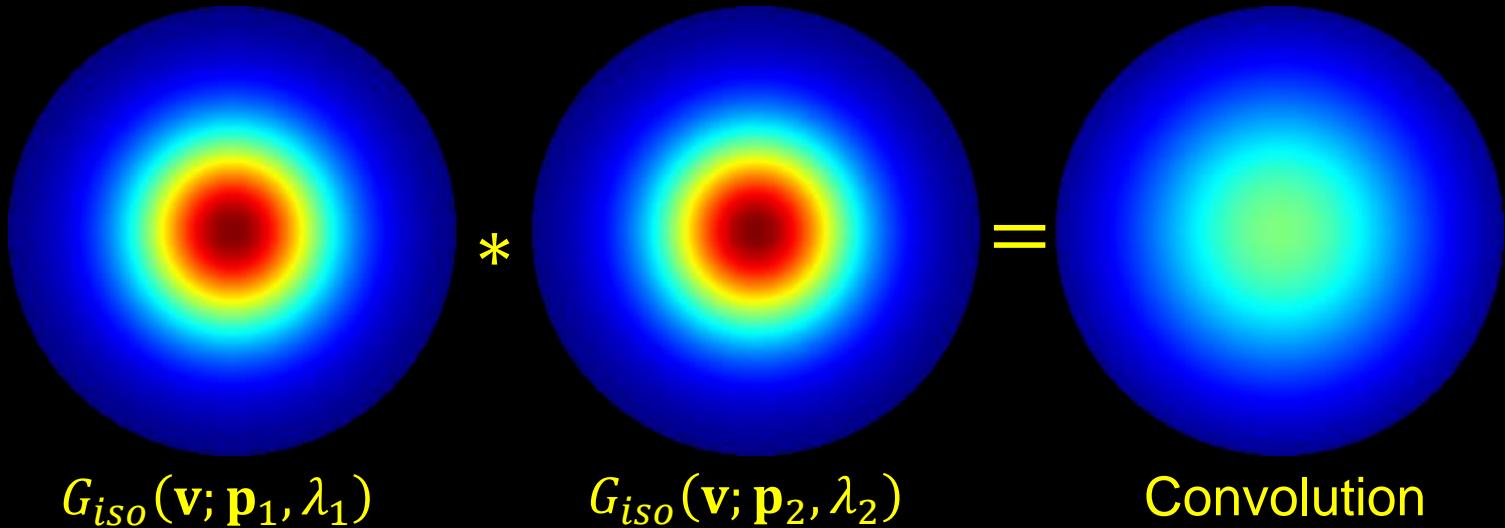
Product

Mathematical Properties of SGs

◎ Closed under convolution approximately

- The convolution of two SGs is still an SG

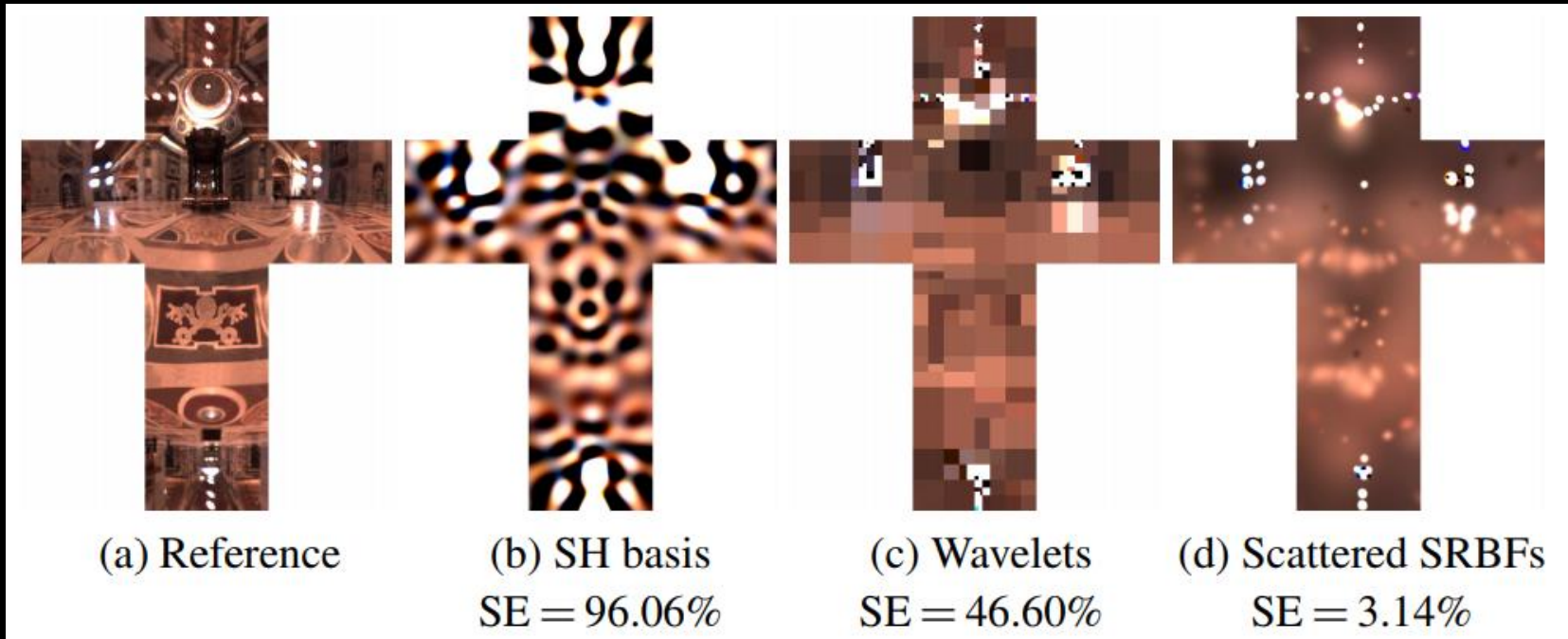
$$\int_{\Omega} G(\mathbf{v}; \mathbf{p}_1, \lambda_1) \cdot G(\mathbf{v}; \mathbf{p}_2, \lambda_2) d\mathbf{v} \approx c_3 G\left(\mathbf{p}_1; \mathbf{p}_2, \frac{\lambda_1 \lambda_2}{\lambda_1 + \lambda_2}\right)$$



Summary of SGs

- ⊙ Rotationally invariant
 - Lighting, BRDFs demand rotation
- ⊙ Capable of representing all-frequency signals
 - All-frequency lighting/BRDFs
- ⊙ Closed-form integral
 - rendering is essentially integration [Kajiya 1986]
- ⊙ Closed under multiplication
 - multiplication of lighting, visibility and BRDFs
- ⊙ Closed under convolution
 - support for various applications
- ⊙ **SGs are non-orthogonal!**

Lighting Approximation



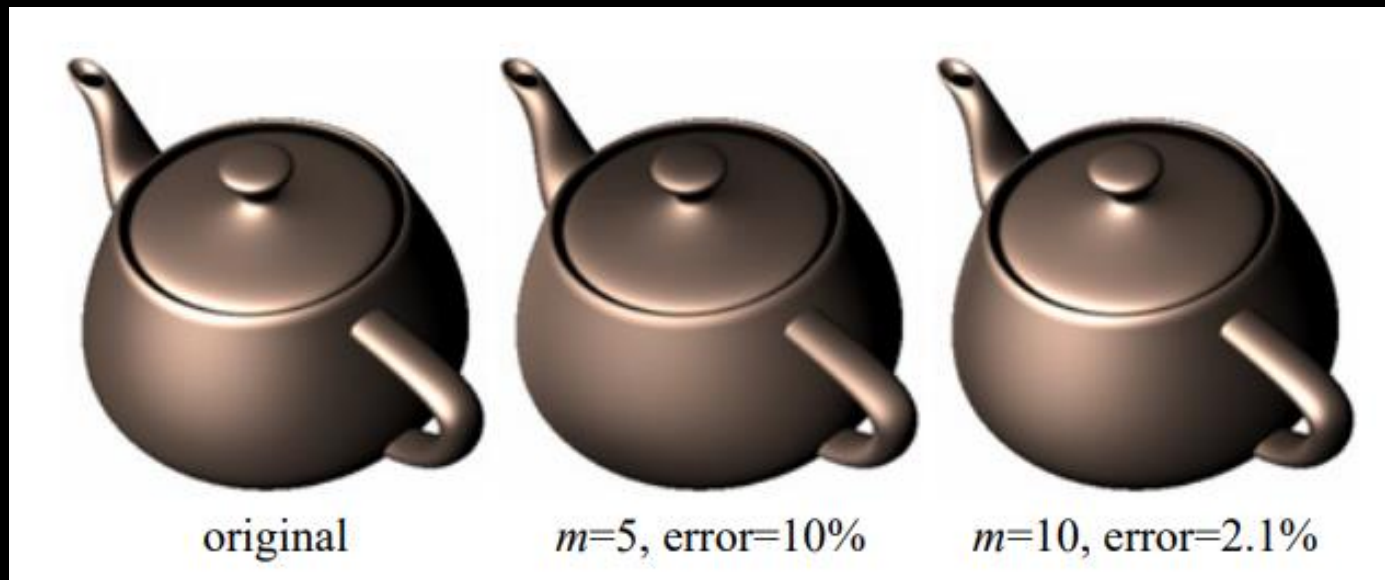
© non-linear process: iterative L-BFGS-B solver (slow)

$$L(\mathbf{i}) \approx \sum l_i G(\mathbf{i}; \mathbf{p}_i, \lambda_i)$$

BRDF Factorization [Wang 03, Liu 03]

- ◎ Precompute the factorization

$$\rho(\mathbf{i}, \mathbf{o}) \approx \sum_m f_m(\mathbf{i}) \cdot g_m(\mathbf{o})$$



Overall Rendering Algorithm

◎ Derivation: Factorizing BRDF

$$L(\mathbf{o}) = \int_{\Omega} L(\mathbf{i})V(\mathbf{i})\rho(\mathbf{i}, \mathbf{o}) \max(0, \mathbf{n} \cdot \mathbf{i}) d\mathbf{i}$$



$$L(\mathbf{o}) = \int_{\Omega} L(\mathbf{i})V(\mathbf{i}) \left(\sum_m f_m(\mathbf{i}) \cdot g_m(\mathbf{o}) \right) \max(0, \mathbf{n} \cdot \mathbf{i}) d\mathbf{i}$$



$$L(\mathbf{o}) = \sum_m g_m(\mathbf{o}) \int_{\Omega} \underbrace{L(\mathbf{i})V(\mathbf{i})f_m(\mathbf{i}) \max(0, \mathbf{n} \cdot \mathbf{i})}_{T(\mathbf{i})} d\mathbf{i}$$

Both represented using SGs

Overall Rendering Algorithm

◎ Derivation: projection to SGs

$$L(\mathbf{o}) = \sum_m g_m(\mathbf{o}) \int_{\Omega} \underline{L(\mathbf{i})T(\mathbf{i})} \, d\mathbf{i}$$

$$L(\mathbf{i}) \approx \sum l_i G_i(\mathbf{i})$$

non-linear approx.



$$T(\mathbf{i}) \approx \sum t_j G_j(\mathbf{i})$$

pre. scattered approx.

$$L(\mathbf{o}) = \sum_m g_m(\mathbf{o}) \sum_{i,j} l_i t_j \int_{\Omega} \underline{G_i(\mathbf{i})G_j(\mathbf{i})} \, d\mathbf{i}$$

analytic solution

- Timing: $O(N*N*M)$, non-orthogonal

Results

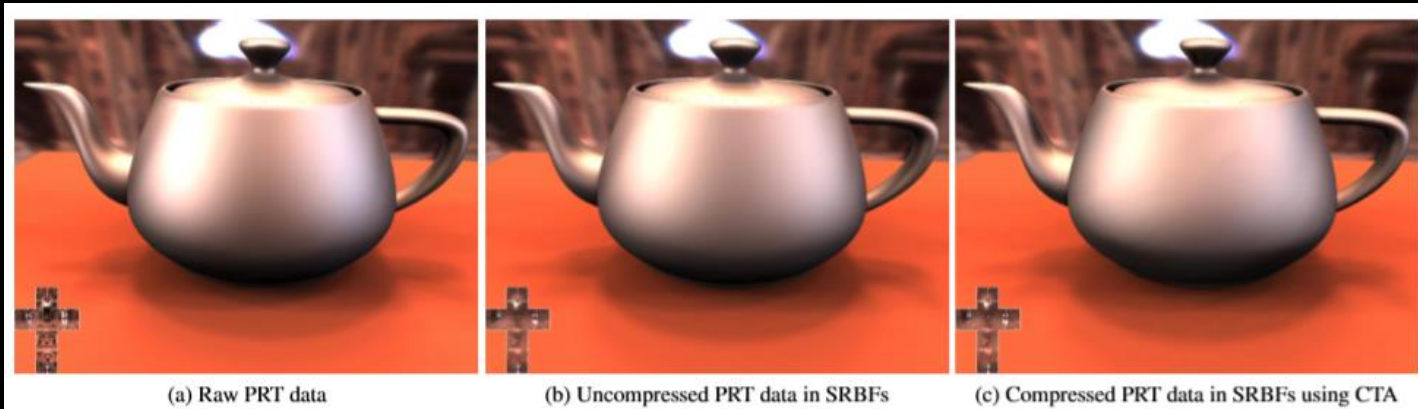
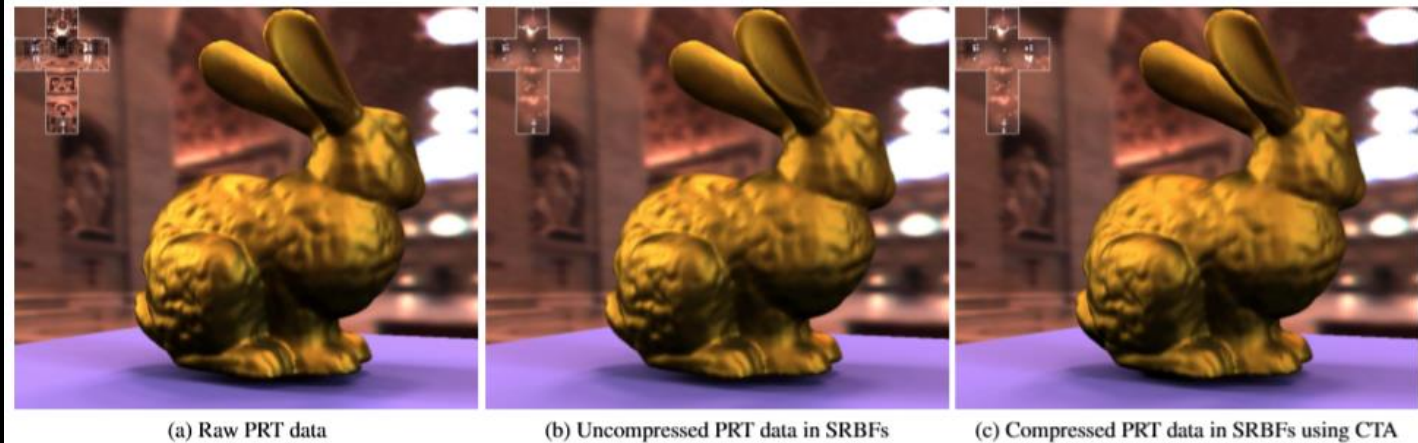


Figure 6: Rendered results of the teapot model.



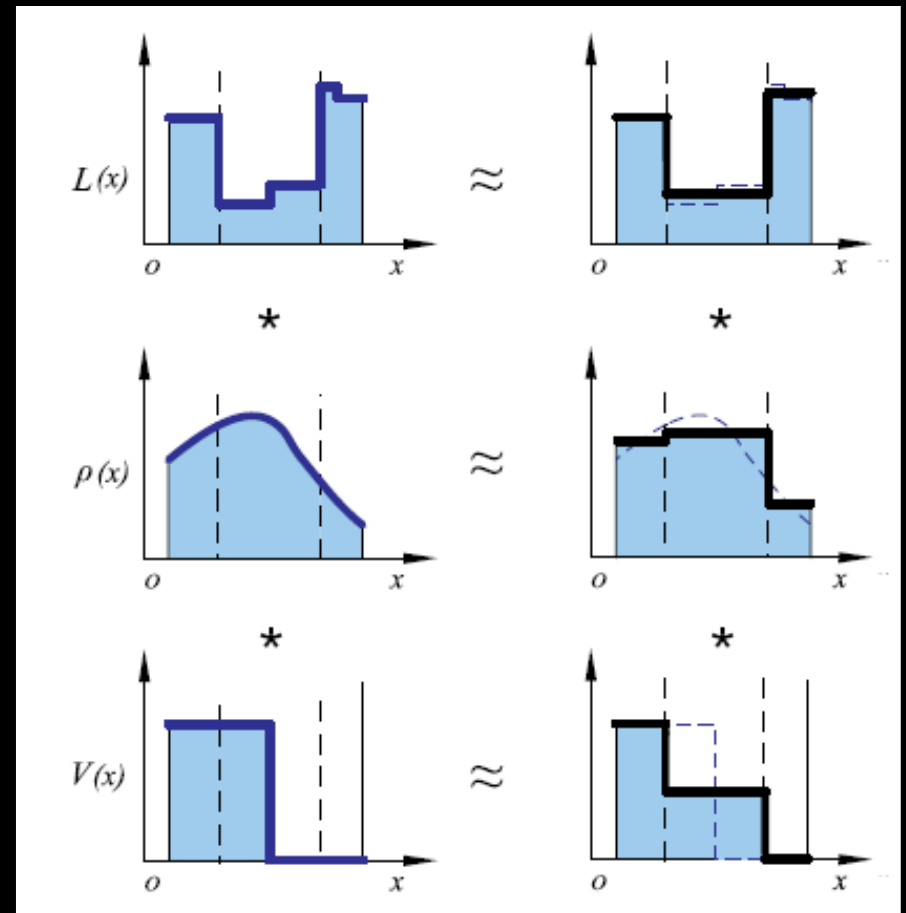
Piecewise Constant [Xu 08]

◎ Spherical Piecewise Constant Basis Function (SPCBF)

- Split sphere into regions
- Each region is represented by a constant

◎ Property

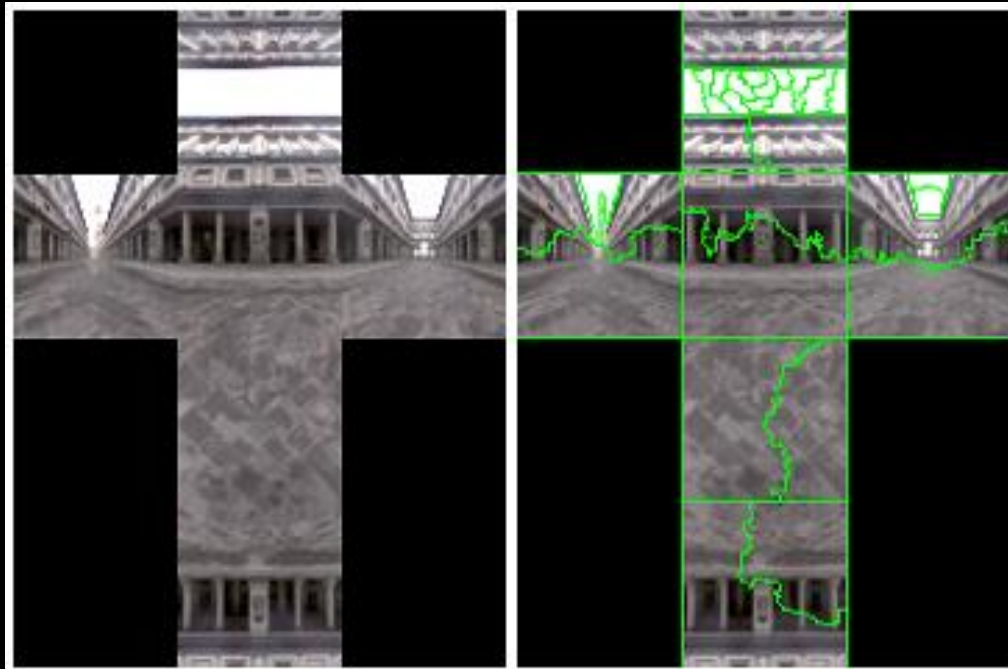
- All-frequency
- Rotation-Invariant
- Multi-product
- Fast projection



Piecewise Constant [Xu 08]

◎ Light Projection

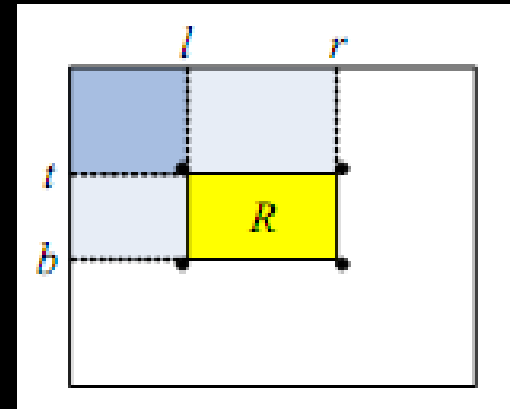
- Bottom-up optimization



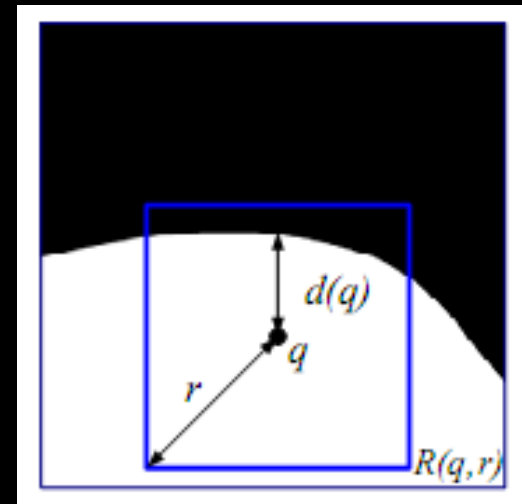
Piecewise Constant [Xu 08]

◎ Projection of visibility and BRDFs

- BRDF
 - using summed area table



- Visibility
 - Using visibility distance table



Results



Comparison of Basis Functions

	SH	Wavelet	SG	SPCBF
Orthogonal	✓	✓	✗	✓
All-frequency	✗	✓	✓	✓
Rotation invariant	✓	✗	✓	✓
Multiple product	✓	✓	✓?	✓
Compact Representation	✓	✓	✓	✗

Triple Product

- Original PRT: light * light transport

$$L(\mathbf{o}) = \int_{\Omega} L(\mathbf{i}) V(\mathbf{i}) \rho(\mathbf{i}, \mathbf{o}) \max(0, \mathbf{n} \cdot \mathbf{i}) d\mathbf{i}$$

lighting

light transport

Not Flexible

- Triple Product

lighting

visibility

BRDF \times cosine

- ... Multiple Product

Wavelet Triple Product [Ng 04]

$$\int_{S^2} \text{img}_1 * \text{img}_2 * \text{img}_3 d\omega$$

The image shows three small images arranged horizontally, separated by asterisks. The first image is a cross-shaped collage of a room. The second image is a silhouette of a person sitting at a desk. The third image is a grayscale silhouette of a person sitting at a desk.

$$B = \int_{S^2} L(\omega) V(\omega) \tilde{\rho}(\omega) d\omega$$

$$= \int_{S^2} \left(\sum_i L_i \Psi_i(\omega) \right) \left(\sum_j V_j \Psi_j(\omega) \right) \left(\sum_k \tilde{\rho}_k \Psi_k(\omega) \right) d\omega$$

$$= \sum_i \sum_j \sum_k L_i V_j \tilde{\rho}_k \int_{S^2} \Psi_i(\omega) \Psi_j(\omega) \Psi_k(\omega) d\omega$$

$$= \sum_i \sum_j \sum_k L_i V_j \tilde{\rho}_k C_{ijk}$$

Wavelet Triple Product [Ng 04]

$$C_{ijk} = \int_{S^2} \Psi_i(\omega) \Psi_j(\omega) \Psi_k(\omega) d\omega$$

Basis Choice	Number Non-Zero C_{ijk}
General (e.g. PCA)	$O(N^3)$
Pixels	$O(N)$
Fourier Series	$O(N^2)$
SH	$O(N^{5/2})$
Haar Wavelets	$O(N \log N)$

Wavelet Triple Product [Ng 04]

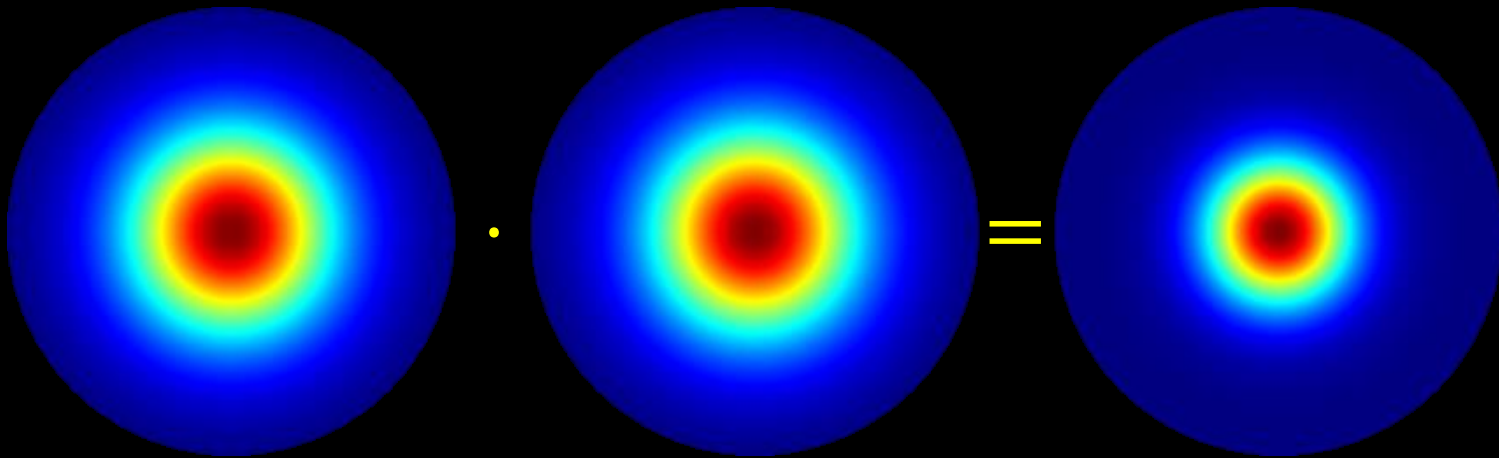


SG Triple Product

⊙ Analytic Computation

- The product of two SGs is also an SG

$$G(\mathbf{v}; \mathbf{p}_1, \lambda_1) \cdot G(\mathbf{v}; \mathbf{p}_2, \lambda_2) = cG\left(\mathbf{v}; \frac{\lambda_1 \mathbf{p}_1 + \lambda_2 \mathbf{p}_2}{|\lambda_1 \mathbf{p}_1 + \lambda_2 \mathbf{p}_2|}, |\lambda_1 \mathbf{p}_1 + \lambda_2 \mathbf{p}_2|\right)$$



$G(\mathbf{v}; \mathbf{p}_1, \lambda_1)$

$G(\mathbf{v}; \mathbf{p}_2, \lambda_2)$

Product

Could be easily extended to multiple product

SH Triple Product

- Precompute all triple products of SH basis

$$C_{ijk} = \int_{\Omega} B_i(\mathbf{i})B_j(\mathbf{i})B_k(\mathbf{i})d\mathbf{i}$$

- Compute the product of two functions directly in SH space:

Original space

SH space

$$L(\mathbf{i}) \approx \sum l_i B_i(\mathbf{i})$$

$$T(\mathbf{i}) \approx \sum t_j B_j(\mathbf{i})$$

$$L(\mathbf{i}) \cdot T(\mathbf{i}) \approx \sum lt_k B_k(\mathbf{i})$$

$$lt_k = \sum_{i,j} l_i t_j C_{ijk}$$

- Could be easily extended to multiple product

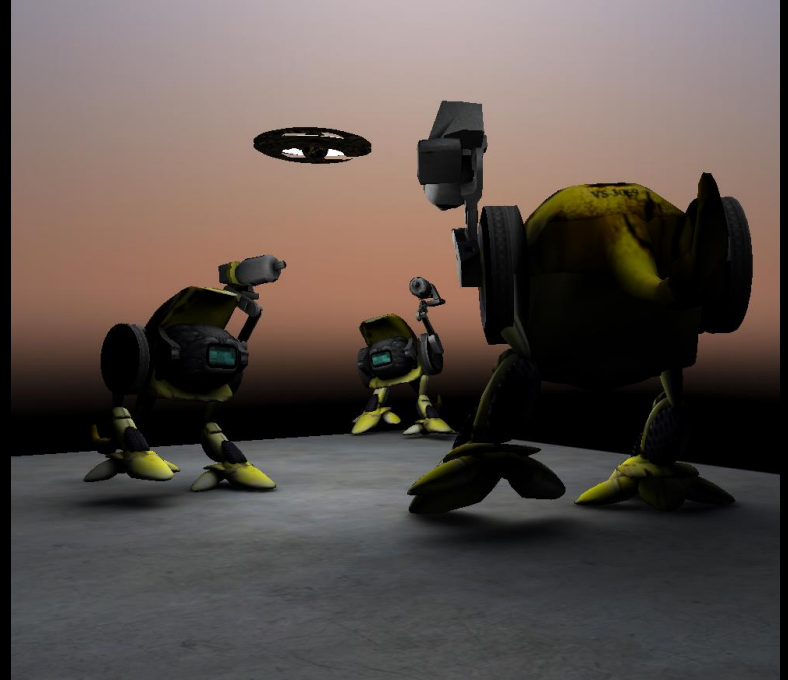
Shadow Field [Zhou05]

◎ PRT

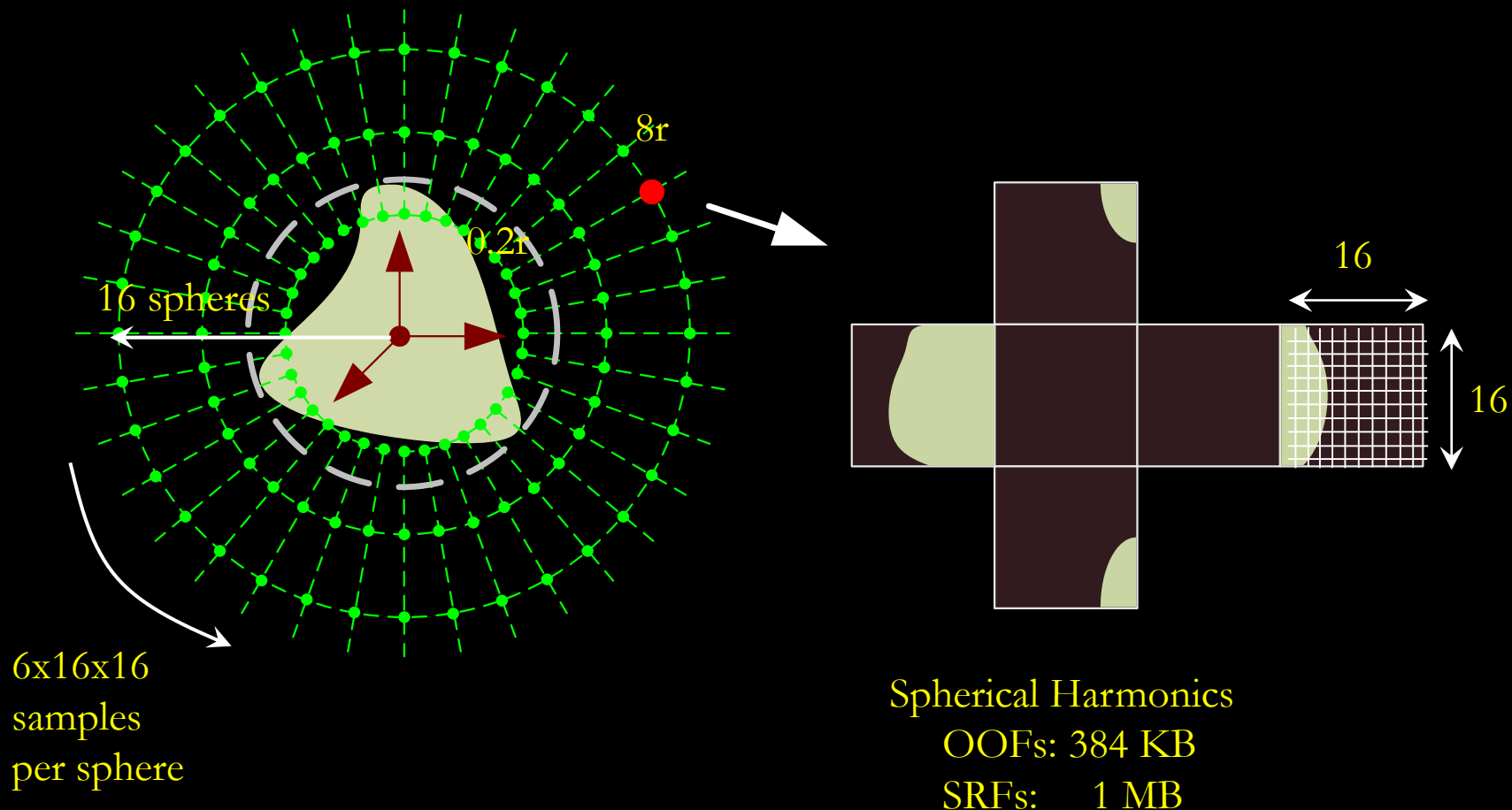
- Handle only static scenes

◎ Shadow Field

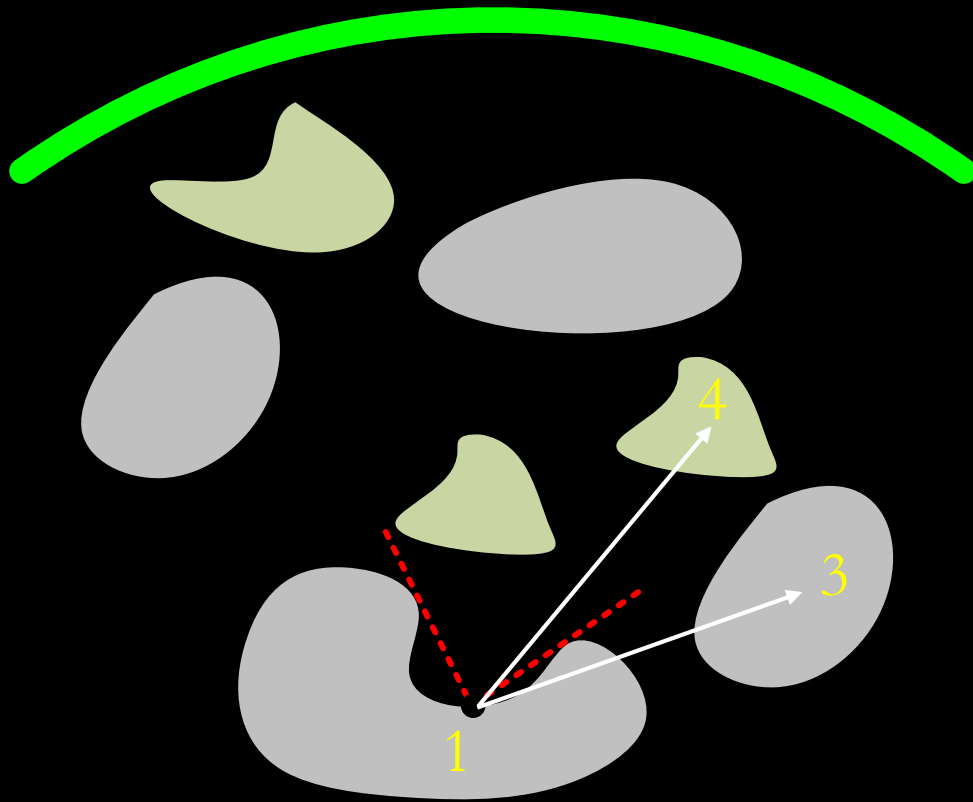
- Handle moving light sources & objects
- Rigid objects + dynamic scene configuration
- Capture SRF/OOF around lights/objects



Sampling & Compression (Low Frequency)



Rendering: Products



BRDF

OOF-1

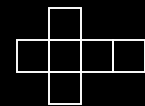
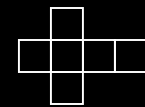
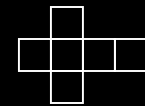
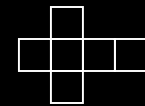
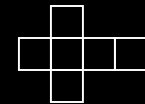
OOF-3

**

SRF-4

=

Reflected
radiance



Results



SH Exponential [Ren 06]

◎ Shadow Field

- Rigid objects
- Computation of SH multiple product is still costly

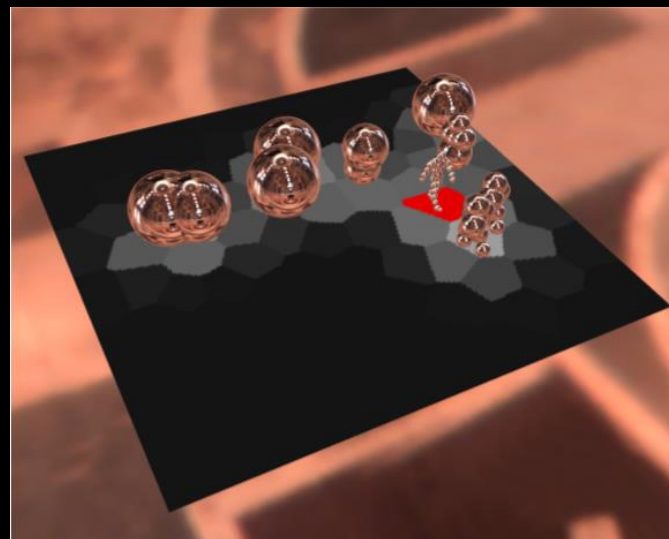
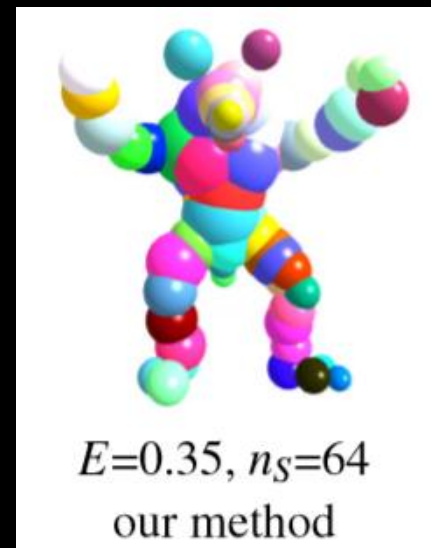
◎ SH Exponential

- Dynamic, deformation scene (objects)
- Derive Exp/Log operators in SH space
- Convert costly multiple product to summation in log space



Blocker Geometry Approximation

- ⦿ Using sphere sets
- ⦿ Dynamically update at each frame



Rendering Computation

◎ Multiple Product (a very big number)

$\text{Light} * \text{Self_Vis} * \text{Occlusion1} * \text{Occlusion2} * \dots * \text{OcclusionN} * \text{BRDF} ***$

◎ Approach

$$f_1 * f_2 * \dots * f_n$$



$$\exp(\log f_1 + \log f_2 + \dots + \log f_n)$$

- Implement exp/log directly in SH space
- Much faster

Results

Dinosaur Demo

120k vertices (75k static, 45k dynamic)

500 spheres in blocker approximation

250 receiver clusters

12.6 Hz average frame rate

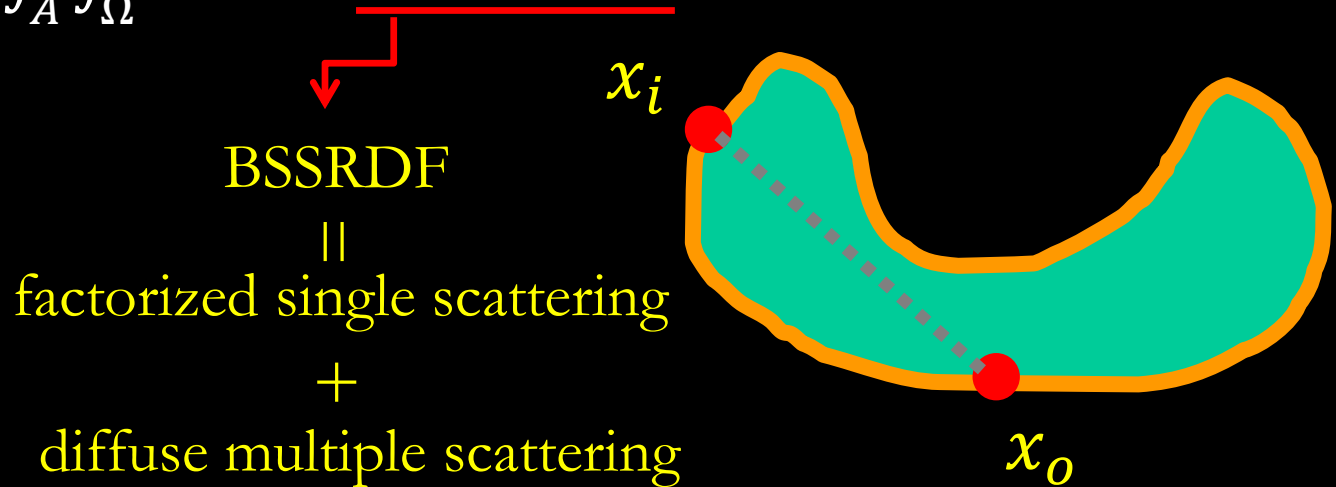
More Rendering Applications

- ◎ Translucent Rendering
- ◎ Hair Rendering
- ◎ BRDF Editing
- ◎ Translucent Editing
- ◎ Hair Editing

Translucent Rendering [Wang05]

- Extend PRT to handle translucent materials

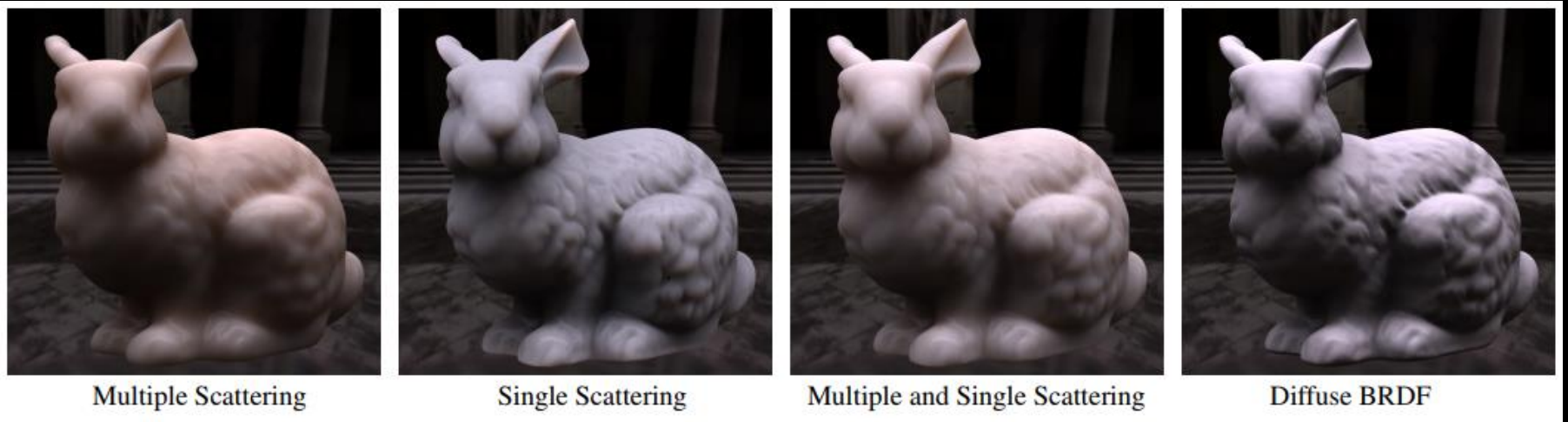
$$L(x_o, \mathbf{o}) = \int_A \int_{\Omega} L(x_i, \mathbf{i}) S(x_i, \mathbf{i}; x_o, \mathbf{o}) \max(0, \mathbf{n} \cdot \mathbf{i}) d\mathbf{i}$$



- Precompute the transport for multiple scattering and single scattering separately

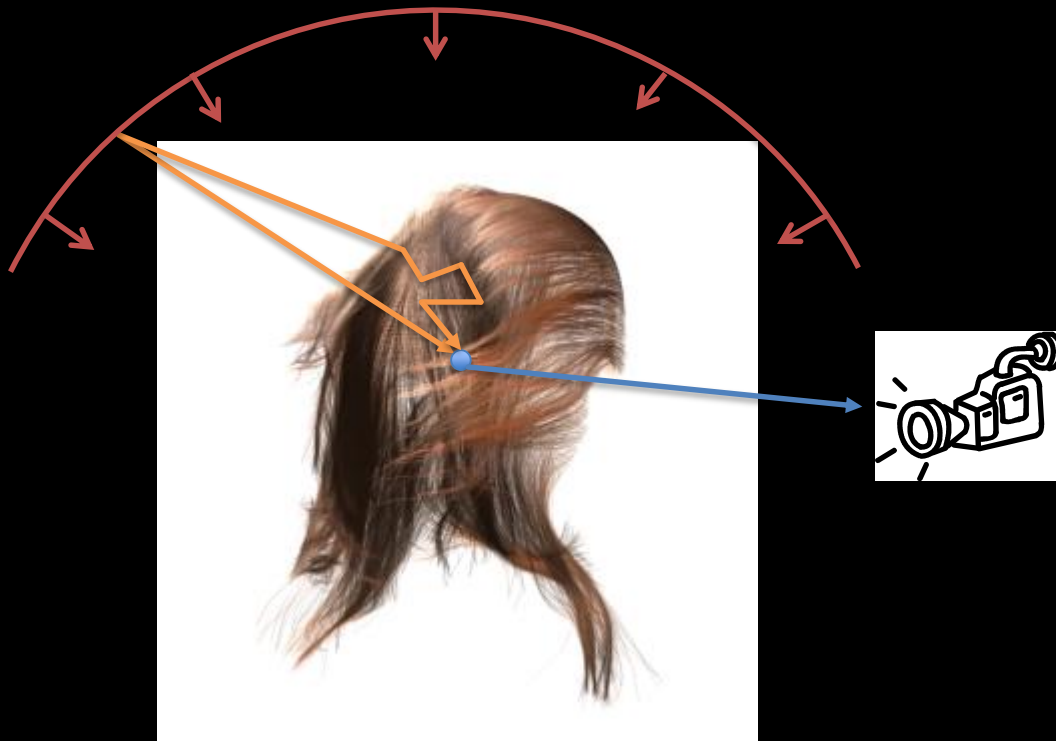
Results

- ◎ Dynamic environment lighting
- ◎ Fix: geometry + materials
- ◎ Real-time



Hair Rendering [Ren 10]

- ◎ Extend PRT to handle hair rendering
 - Support environment lighting



- Single Scattering
 - Self-shadowing
 - Fiber scattering
 - Transparency
- Multiple Scattering
- Natural Illumination

Single Scattering Computation

$$L(\mathbf{o}) = D \int_{\Omega} L(\mathbf{i}) T(\mathbf{i}) S(\mathbf{i}, \mathbf{o}) \max(0, \mathbf{n} \cdot \mathbf{i}) d\mathbf{i}$$

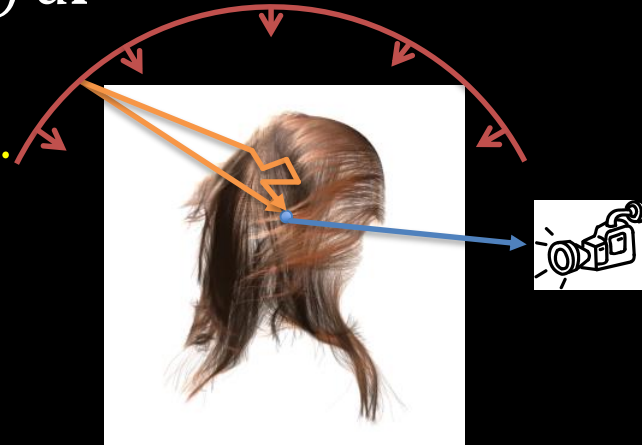
self shadow

hair scattering func.

- Approximate $L(\mathbf{o})$ by N SGs
- Move T out from the integral
 - small variation of T

$$L(\mathbf{o}) = D \sum_{j=1}^N L_j \tilde{T} \int_{\Omega} G_j(\mathbf{i}) S(\mathbf{i}, \mathbf{o}) \max(0, \mathbf{n} \cdot \mathbf{i}) d\mathbf{i}$$

Precompute as 4D table



Results

- ◎ Dynamic lighting, geometry
- ◎ Fix hair scattering parameters
- ◎ Interactive framerates



BRDF Editing [Ben-Artzi 06]

◎ PRT

- dynamic lighting + precompute light transport
- Fix: material + geometry

$$L(\mathbf{o}) = \int_{\Omega} \underbrace{L(\mathbf{i})}_{\text{lighting}} \underbrace{V(\mathbf{i}) \max(0, \mathbf{n} \cdot \mathbf{i}) \rho(\mathbf{i}, \mathbf{o})}_{\text{light transport}} d\mathbf{i}$$

$\underbrace{\hspace{10em}}_{\text{material transport}} \quad \underbrace{\hspace{2em}}_{\text{material}}$

◎ PRT based BRDF editing

- dynamic material + precompute material transport
- Fix: lighting + geometry + viewpoint

BRDF Editing [Ben-Artzi 06]

- Approach: parameterize BRDF as 1D curve

$$\rho(\mathbf{i}, \mathbf{o}) = \underbrace{\rho_q(\mathbf{i}, \mathbf{o})}_{\text{quotient term}} \underbrace{f(\gamma(\mathbf{i}, \mathbf{o}))}_{\text{1D curve}}$$

quotient term 1D curve

$$f(\gamma) \approx \sum c_j \underbrace{b_j(\gamma)}_{\text{wavelet basis}}$$

wavelet basis

- Rendering Algorithm

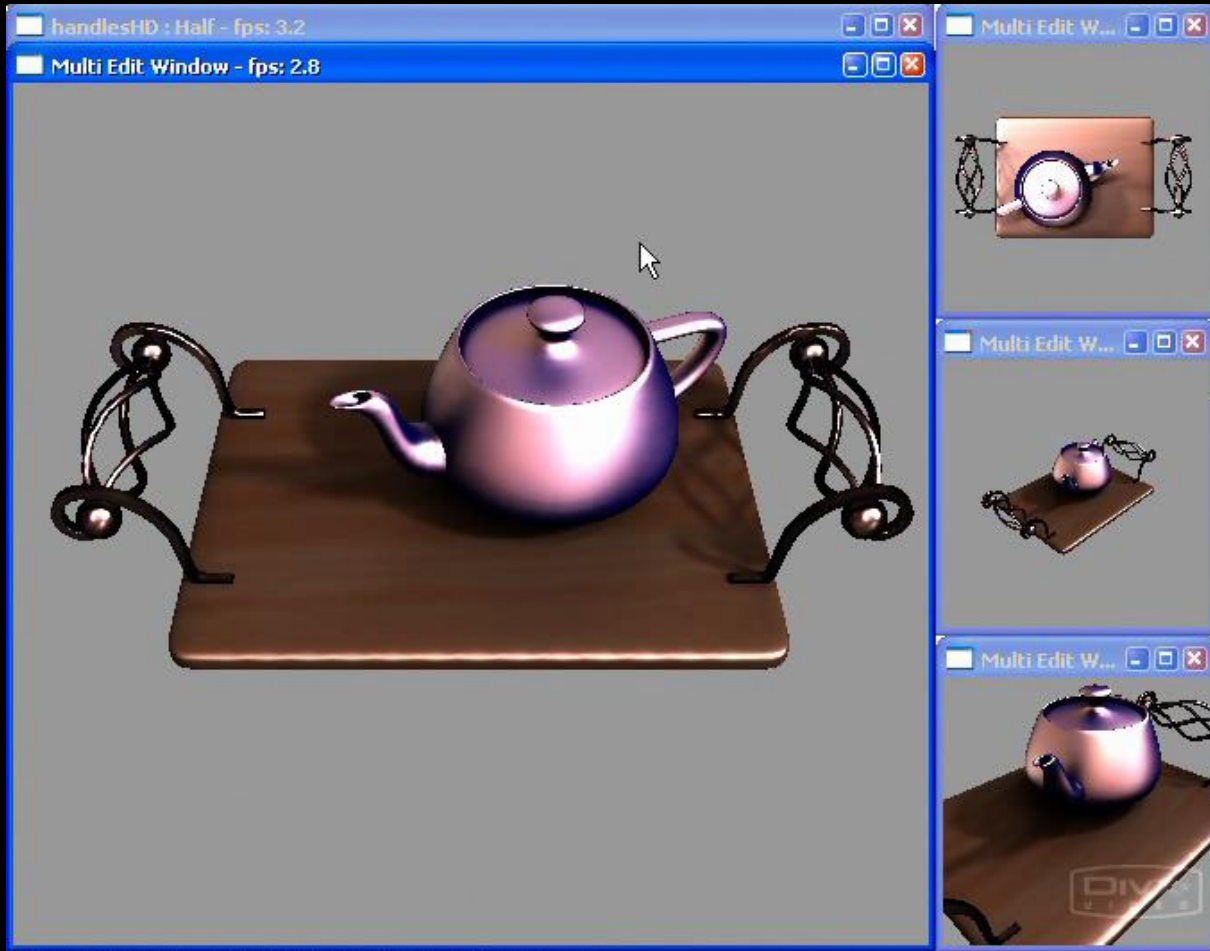
- Precompute:

$$T_j = \int_{\Omega} L(\mathbf{i}) V(\mathbf{i}) \max(0, \mathbf{n} \cdot \mathbf{i}) \rho_q(\mathbf{i}, \mathbf{o}) b_j(\gamma(\mathbf{i}, \mathbf{o})) d\mathbf{i}$$

- Runtime: (viewing direction \mathbf{o} is fixed)

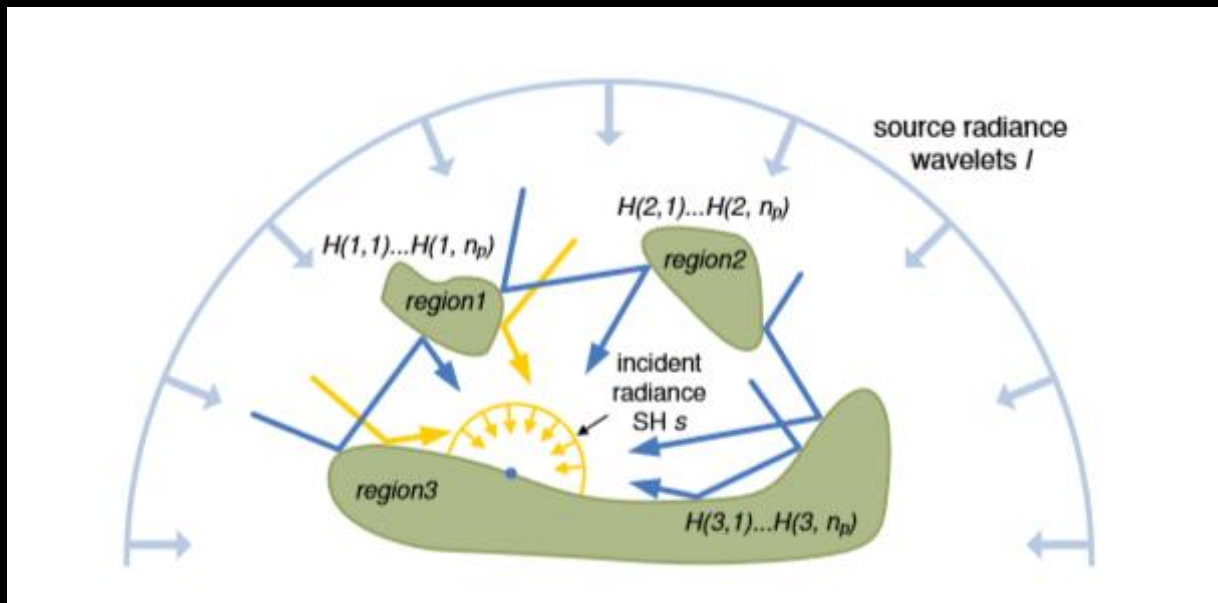
$$L(\mathbf{o}) \approx \sum c_j T_j$$

Results



BRDF Editing with interreflection [Sun06]

- ⦿ dynamic lighting + viewpoint + material
- ⦿ Fix: geometry
- ⦿ all-frequency one bounce interreflection
- ⦿ Introduce PTT: precomputed transfer tensors



Results

◎ Interactive rates



Translucent Editing [Xu 07]

- ◎ Combine the ideas in “BRDF editing” and in “translucent rendering”
 - dynamic dipole parameters + precompute material transport
 - Compute single/multiple scattering separately
 - Basis Function: piecewise linear



Results

- ◎ Real-time, environment lighting
- ◎ Fix: lighting + geometry
- ◎ Changing scattering parameters



PRT vs analytic integration

Rendering Integral $\int_{\Omega} L(\mathbf{i})V(\mathbf{i})\rho(\mathbf{i})d\mathbf{i}$

⊙ PRT (Precomputation)

- Long precomputation time, large storage
- Bake geometry/material/lighting into precomputation, needs to fix them

⊙ Analytic Computation

- No (or small) precomputation
- Everything dynamic, could be run-time changed

SG based analytic Integration

- ◎ SG as a PRT basis [Tsai 2006]
- ◎ rendering with dynamic BR [Wang 2006]



Rendering with dynamic BRDFs [Wang09]

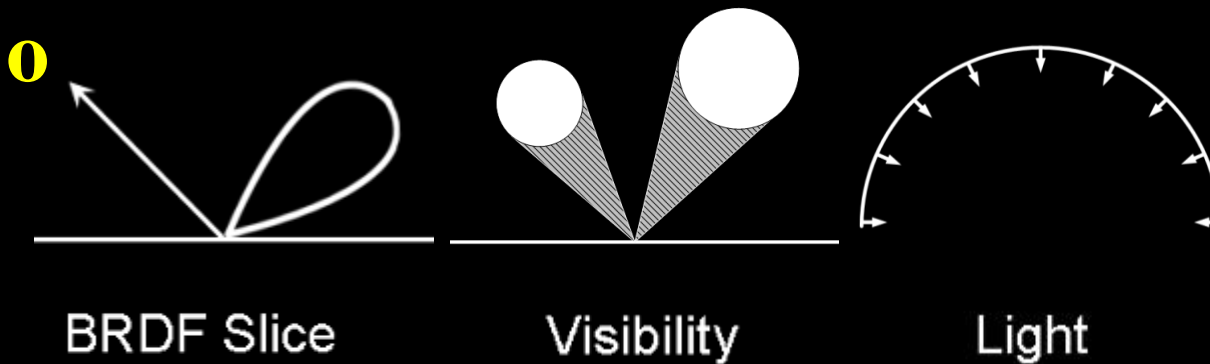
◎ static scene, dynamic lighting, dynamic BRDF

◎ BRDF: microfacet model

- parametric \leftrightarrow measured
- isotropic \leftrightarrow anisotropic
- glossy \leftrightarrow mirror-like

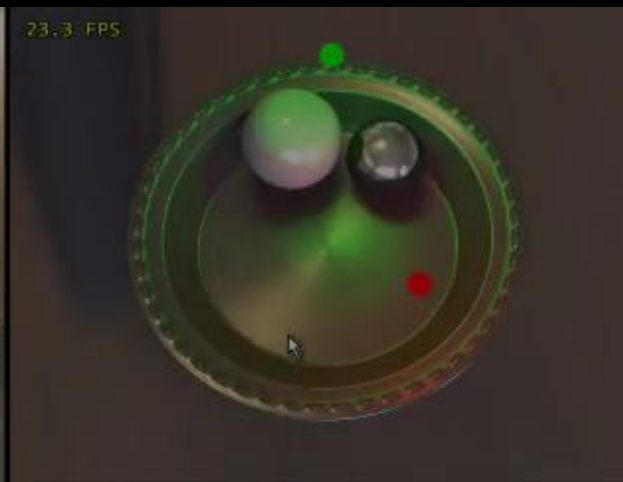


Algorithm Overview



Spherical Gaussians \otimes SSDF \cdot Prefiltered Environment =

Results



Rendering and appearance editing of hairs [Xu 2011]

Single scattering



$$L(\omega_o) = D \int_{\Omega} L(\omega_i) T(\omega_i) S(\omega_i, \omega_o) \cos \theta_i d\omega_i$$

- $L(\omega_i)$: environment lighting
- $T(\omega_i)$: self shadowing
- $S(\omega_i, \omega_o)$: hair scattering function

Rendering and appearance editing of hairs [Xu 2011]

Single scattering



$$L(\omega_o) \approx D \sum_{\Omega} \left(\sum_j L_j(\omega_i) T_j(\omega_i, \omega_o) \right) T(\omega_i, \omega_o) G(\omega_i, \omega_o) d\omega_i$$

- Approximate $L(\omega_i)$ by a set of SGs $G_j(\omega_i)$ [Tsai and Shih 2006]

Rendering and appearance editing of hairs [Xu 2011]

Single scattering



$$L(\omega_o) \approx D \sum_j l_j \tilde{T} \int_{\Omega} \int_{\Omega} G(\omega_i, \omega_o) T(\omega_i, \omega_o, \omega_s) \rho_i \cos \theta_i d\omega_i d\omega_s$$

- Approximate $L(\omega_i)$ by a set of SGs $G(\omega_i)$ [Tsai and Shih 2006]
- Move T out from the integral [Ren]

Problem:
evaluate scattering Integral

Single Scattering Integral

$$\int_{\Omega} G_j(\omega_i) S(\omega_i, \omega_o) \cos \theta_i d\omega_i$$

- ◎ Previous Approach [Ren 2010]

- Precompute the integral into 4D table

- ◎ Our key insight

- Is there an approximated analytic solution?

- **YES**

- Decompose SG $G_j(\omega_i)$ into products of **circular Gaussians**
- Approximate scattering function $S(\omega_i, \omega_o)$ by **circular Gaussians**

Results

- ⦿ No precomputation
- ⦿ all (geometry, lighting, hair scattering param.) dynamic

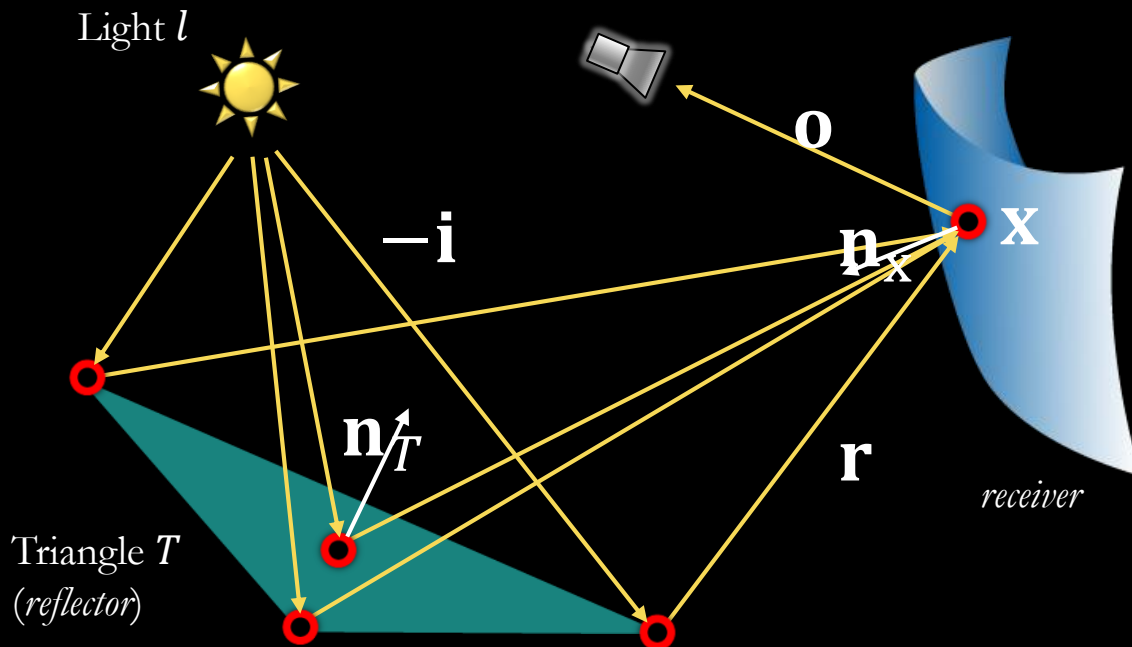


One-bounce interreflection [Xu 14]

- ◎ Aim at accurately and efficiently computing one-bounce interreflections with *all-frequency* BRDFs
- ◎ SG-based representation of BRDFs and lighting
- ◎ A novel *analytic* rendering formula

One-bounce Interreflection Model

$$L_x(\mathbf{o}) = \int_{\Omega_T} \rho_x(-\mathbf{r}, \mathbf{o}) \max(-\mathbf{r} \cdot \mathbf{n}_x, 0) \int_{\Omega} L_l(\mathbf{i}) \rho_T(\mathbf{i}, \mathbf{r}) \max(\mathbf{i} \cdot \mathbf{n}_T, 0) d\mathbf{i} d\mathbf{r}$$



Configuration:

- **Single triangle reflector**
- **Distant lighting**
- **No occlusion between the light, the reflector, and the receiver**
- **Ignore textures on the reflector**
- **Uniform BRDF (reflector)**

One-bounce Interreflection Model

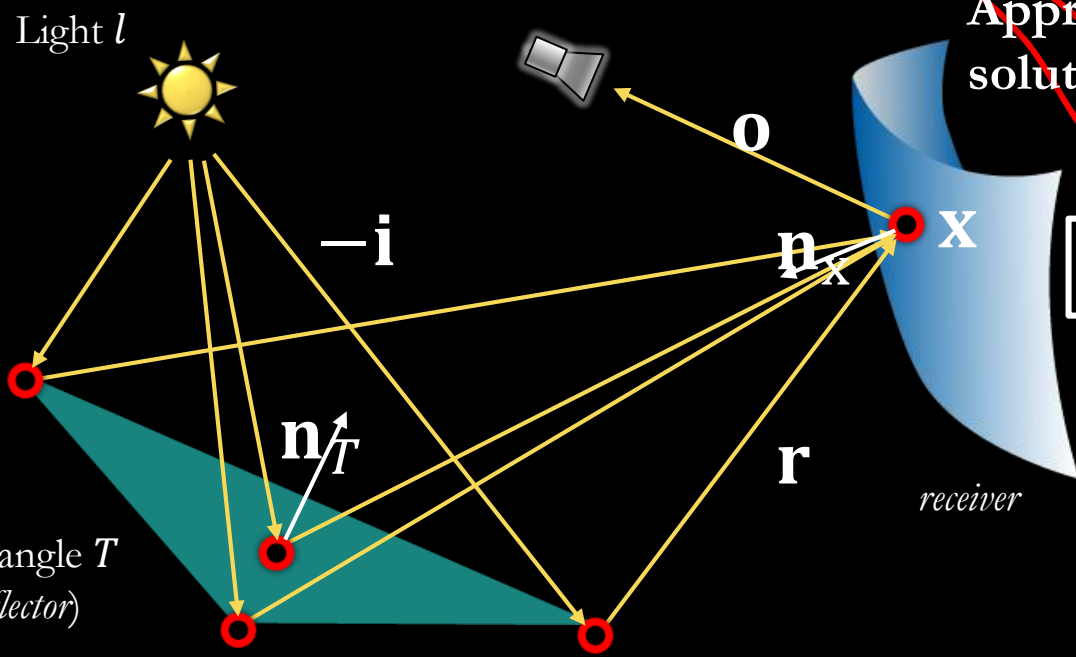
$$L_x(\mathbf{o}) = \int_{\Omega_T} \rho_x(-\mathbf{r}, \mathbf{o}) \max(-\mathbf{r} \cdot \mathbf{n}_x, 0) \int_{\Omega} L_l(\mathbf{i}) \rho_T(\mathbf{i}, \mathbf{r}) \max(\mathbf{i} \cdot \mathbf{n}_T, 0) d\mathbf{i} d\mathbf{r}$$

Piecewise linear approximation

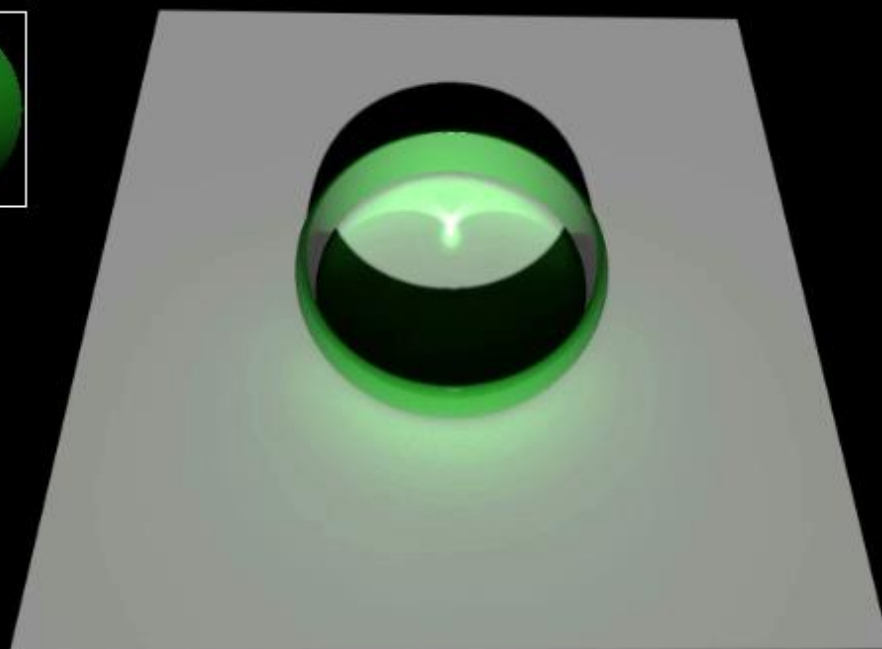
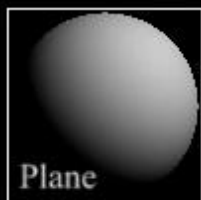
Approximated closed solution

Analytically Evaluated!

Represented as



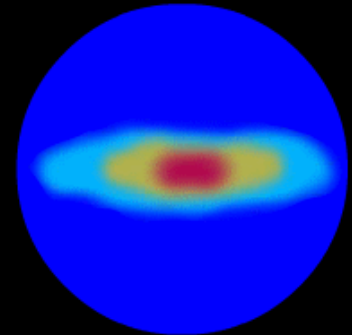
Results



Limitation of SGs

⊙ Representing real functions

- A mixture model of n scattered SGs are required
- Poor scalability
 - More anisotropic functions require more SGs
- Making Trade-off
 - Larger $n \rightarrow$ more accuracy, more cost
 - Smaller $n \rightarrow$ less accuracy, less cost



An example

Anisotropic SG [Xu 14]

$$G(\mathbf{v}; [\mathbf{x}, \mathbf{y}, \mathbf{z}], [\lambda, \mu]) = \max(\mathbf{v} \cdot \mathbf{z}, 0) \cdot e^{-(\lambda(\mathbf{v} \cdot \mathbf{x})^2 + \mu(\mathbf{v} \cdot \mathbf{y})^2)}$$

tangent

bi-tangent

lobe

bandwidth for \mathbf{x} -axis

bandwidth for \mathbf{y} -axis

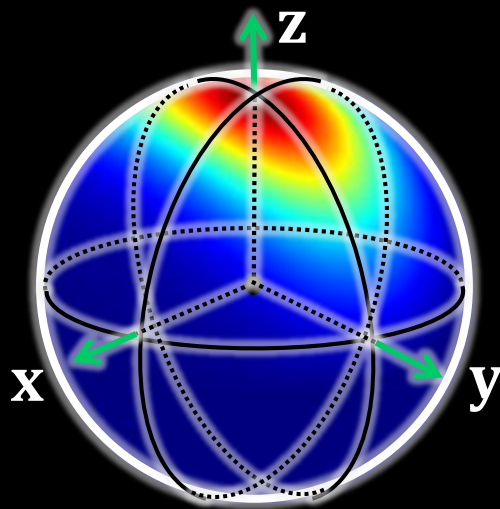
smooth term

exponential term

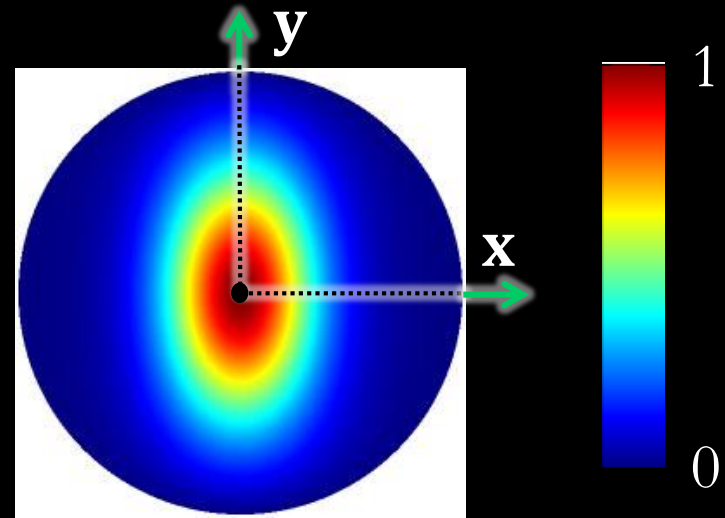
An ASG
example with

$$\lambda = 4$$

$$\mu = 1$$



3D view



Top view (2D)

ASGs

- ◎ Desired operators
 - Closed-form integral
 - Closed-form product
 - Closed-form convolution

Integral of an ASG

◎ Integral

$$\int_{\Omega} G(\mathbf{v}) d\mathbf{v} = \int_{\phi=0}^{2\pi} \left(\int_{\theta=0}^{\pi/2} e^{-\lambda(\sin\theta \cos\phi)^2 - \mu(\sin\theta \sin\phi)^2} \sin\theta \cos\theta d\theta \right) d\phi$$

◎ Our approximation

$$\int_{\Omega} G(\mathbf{v}) d\mathbf{v} \approx \frac{\pi}{\sqrt{\lambda\mu}}$$

- Accurate (error < 0.68%) when $\lambda, \mu > 5$

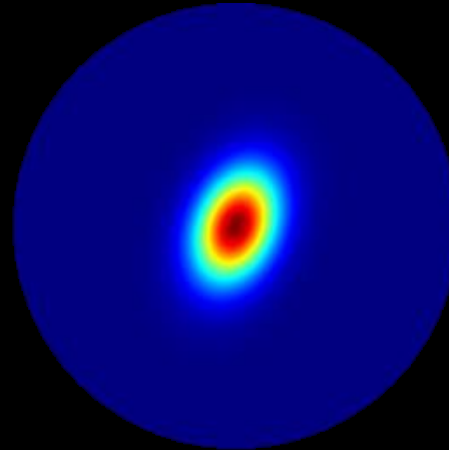
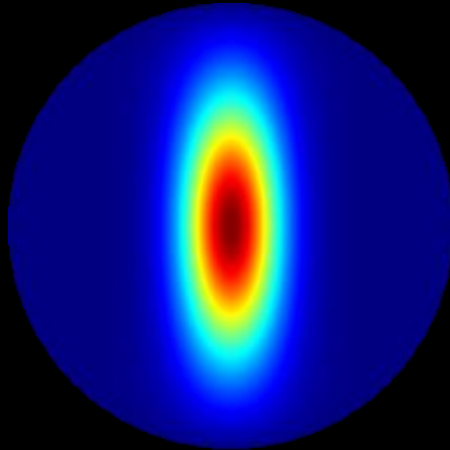
Product of two ASGs

⊙ Our approximation: $G(\mathbf{v}; A_1) \cdot G_2(\mathbf{v}; A_2) \approx S(\mathbf{z}_3; \mathbf{z}_1, \mathbf{z}_2) \cdot G(\mathbf{v}; A_3)$

⊙ Validation

1st ASG

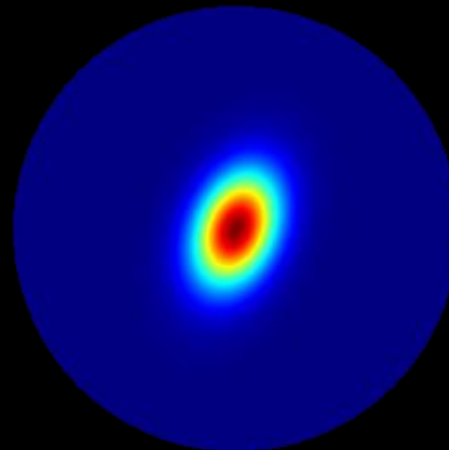
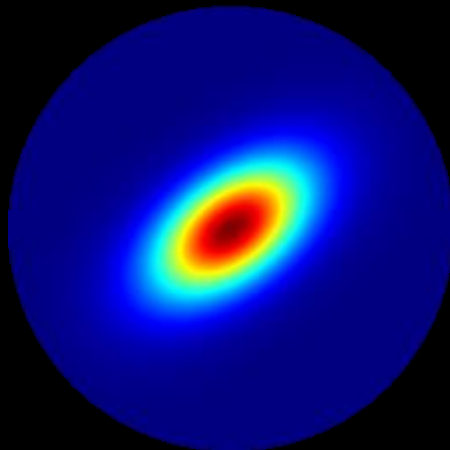
$G(\mathbf{v}; A_1)$



Approximated
product

2nd ASG

$G(\mathbf{v}; A_2)$

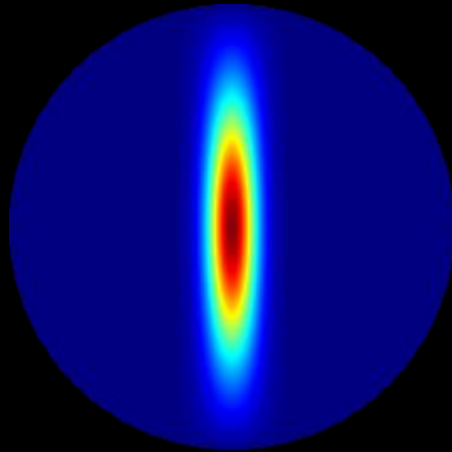


Ground truth
product

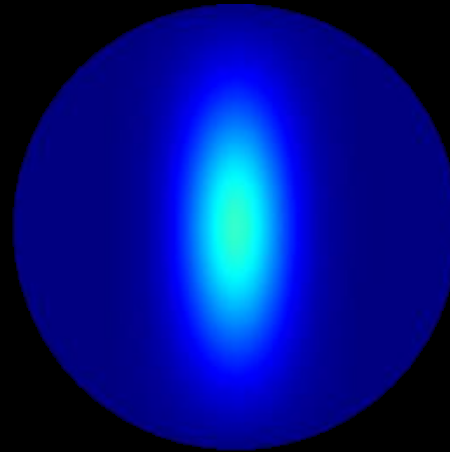
Convolution of an ASG with an SG

- Our approximation: $C(\mathbf{p}) \approx \frac{\pi}{\sqrt{(\lambda+\nu)(\mu+\nu)}} \cdot G\left(\mathbf{p}; [\mathbf{x}, \mathbf{y}, \mathbf{z}], \left[\frac{\nu\lambda}{\nu+\lambda}, \frac{\nu\mu}{\nu+\mu}\right]\right)$

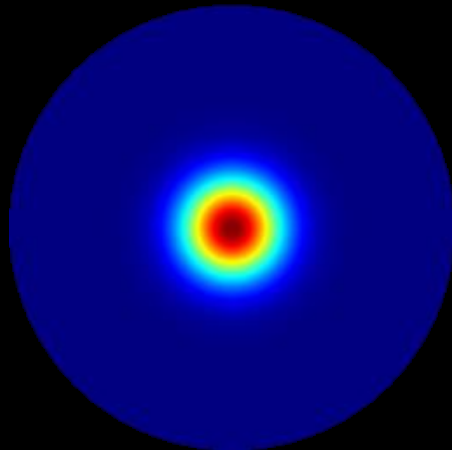
ASG



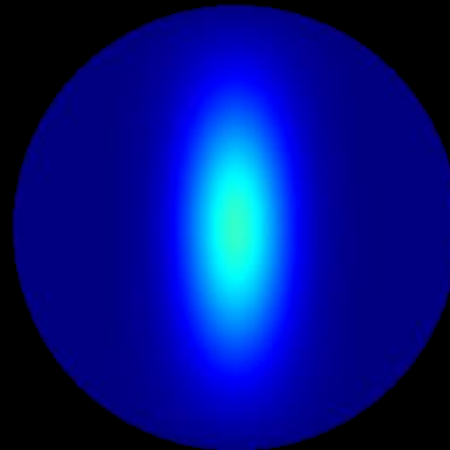
Approximated
convolution



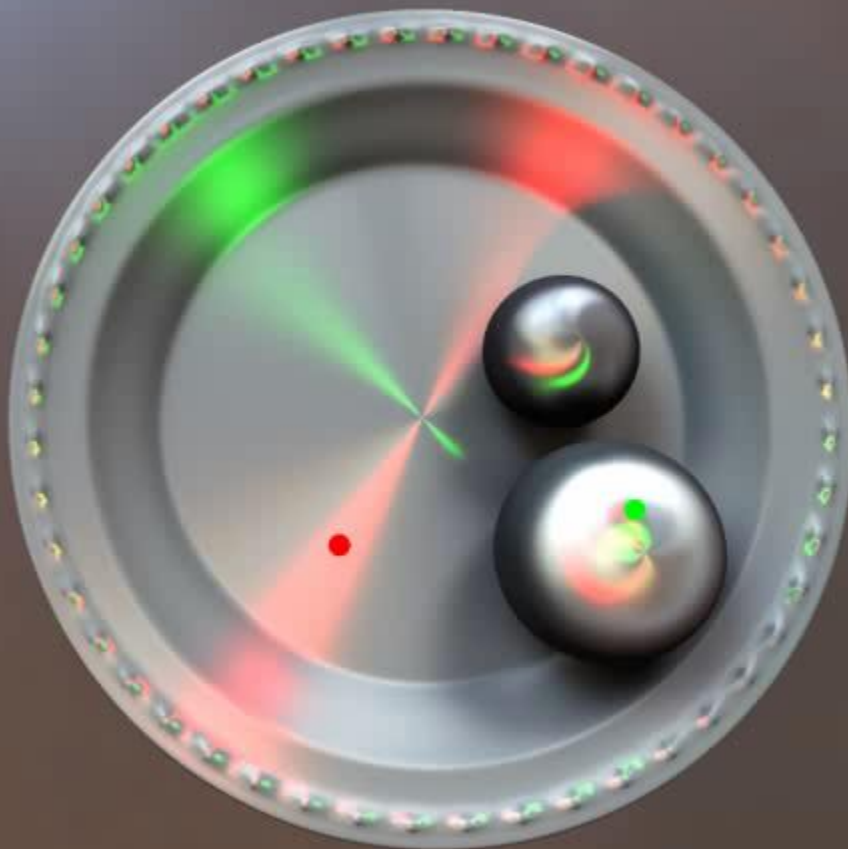
Convolution kernel
(SG)



Ground truth
convolution

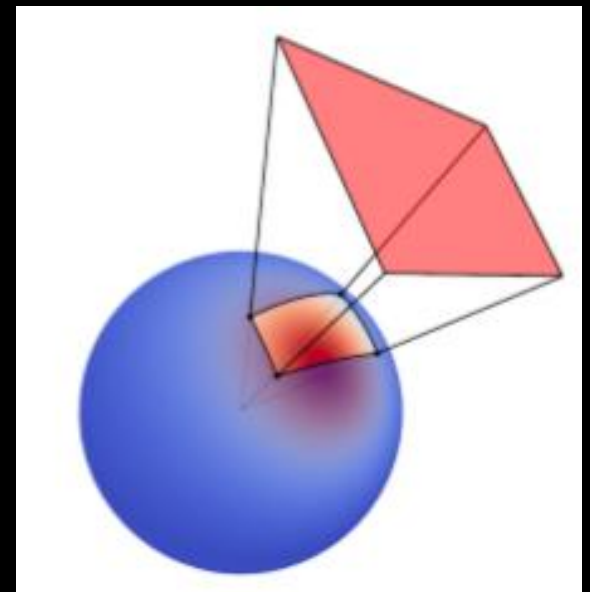
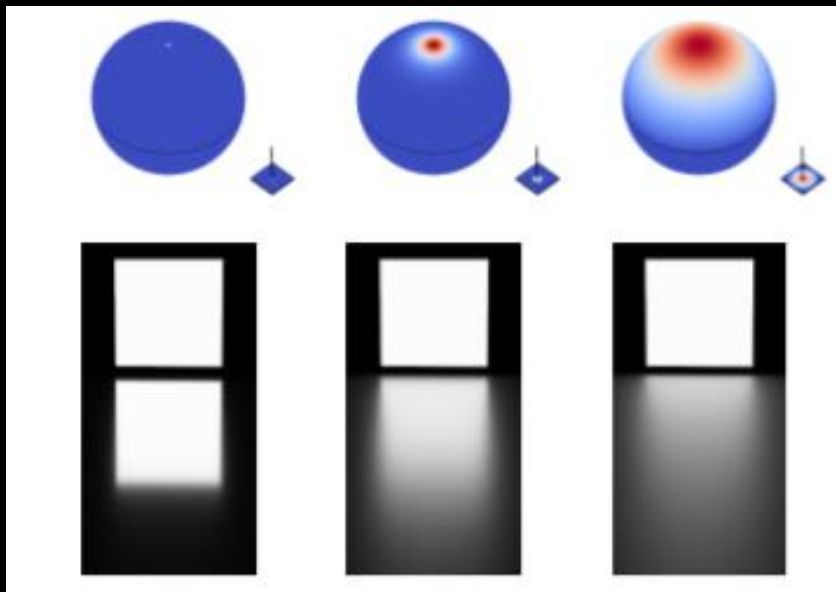


Results



Linearly Transformed Cosines [Heitz 18]

- ◎ Approximate BRDFs using *Linearly Transformed Cosines Functions*
- ◎ analytical integration on spherical polygons



Misc

◎ Compression

- VQ, PCA, Clustered PCA [Sloan 03]

◎ Meshless [Lehtinen 08]

◎ Image space

- Direct-to-indirect Transfer [Hašan 06]



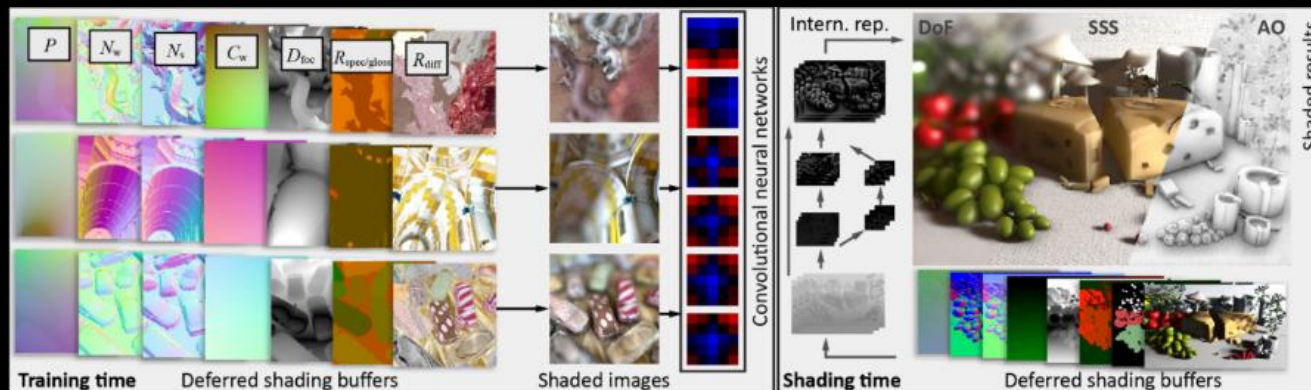
Misc

◎ Neural network as a basis

- Radiance Regression Functions [Ren 2013]



- Deep Shading [Nalbach 2017]



Reading Materials

- ◎ SIGGRAPH 2005 Course , by Jan Kautz et al
 - Precomputed Radiance Transfer: Theory and Practice
www0.cs.ucl.ac.uk/staff/j.kautz/PRTCOURSE/
- ◎ PRT survey, 2007, by Ravi Ramamoorthi
 - Precomputation-Based Rendering
- ◎ EG STAR 2012 Report, by Ritschel et al
 - The State of the Art in Interactive Global Illumination

Conclusion

- ◎ Precomputed Radiance Transfer
 - Project light/transport to basis function space
 - Precompute and save the transport
 - Efficient computing at run-time
- Various rendering applications/features
 - environment lighting, local lighting
 - BRDFs/ translucent
 - Material editing
 - Static/dynamic scenes
 - Interreflection
 - ...

Thanks!
Questions?