# Pre-computed Radiance Transfer 

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## Goal

- Real-time rendering with complex lighting, shadows, and possibly GI
- Infeasible - too much computation for too small a time budget
- Approaches
- Lift some requirements, do specific-purpose tricks
- Environment mapping, irradiance environment maps
- SH-based lighting
- Split the effort
- Offline pre-computation + real-time image synthesis
- "Pre-computed radiance transfer"


## SH-based Irradiance Env. Maps



## SH-based Irradiance Env. Maps




Standard Eucalyptus Grove


Images courtesy Ravi Ramamoorthi \& Pat Hanrahan

## SH-based Arbitrary BRDF Shading 1

- [Kautzet al. 2003]
- Arbitrary, dynamic env. map
- Arbitrary BRDF
- No shadows
- SH representation

(a) point light

(b) glossy

(c) anisotropic
- Environment map (one set of coefficients)
- Scene BRDFs (one coefficient vector for each discretized view direction)



## SH-based Arbitrary BRDF Shading 3

- Rendering: for each vertex / pixel, do

$$
L_{o}\left(\omega_{o}\right)=\int_{\Omega} L_{i}\left(\omega_{i}\right) \cdot \operatorname{BRDF}\left(\omega_{i}, \omega_{o}\right) \cdot \cos \theta_{i} \cdot d \omega_{i}
$$

(


Environment map


BRDF
$=$ coeff. dot product

$$
L_{o}\left(\omega_{o}\right)=\Lambda_{\text {intp }}(\mathbf{p}) \bullet F\left(\mathbf{p}, \omega_{o}\right)
$$

## Pre-computed Radiance Transfer

## Pre-computed Radiance Transfer

- Goal
- Real-time + complex lighting, shadows, and GI
- Infeasible - too much computation for too small a time budget
- Approach
- Precompute (offline) some information (images) of interest
- Must assume something about scene is constant to do so
- Thereafter real-time rendering. Often hardware accelerated


## Assumptions

- Precomputation
- Static geometry
- Static viewpoint (some techniques)

- Real-Time Rendering (relighting)
- Exploit linearity of light transport


## Relighting as a Matrix-Vector Multiply



$$
=\left[\begin{array}{cccc}
T_{11} & T_{12} & \cdots & T_{1 M} \\
T_{21} & T_{22} & \cdots & T_{2 M} \\
T_{31} & T_{32} & \cdots & T_{3 M} \\
\vdots & \vdots & \ddots & \vdots \\
T_{N 1} & T_{N 2} & \cdots & T_{N M}
\end{array}\right]\left[\begin{array}{c}
L_{1} \\
L_{2} \\
\vdots \\
L_{M}
\end{array}\right] \text { in }
$$

## Relighting as a Matrix-Vector Multiply



## Output Image (Pixel Vector)

Input Lighting

Precomputed
Transport
Matrix

## Matrix Columns (Images)



## Problem Definition

Matrix is Enormous

- $512 \times 512$ pixel images
- $6 \times 64 \times 64$ cubemap environments

Full matrix-vector multiplication is intractable

- On the order of $10^{10}$ operations per frame

How to relight quickly?

## Outline

- Compression methods
- Spherical harmonics-based PRT [Sloan et al. 02]
- (Local) factorization and PCA
- Non-linear wavelet approximation
- Changing view as well as lighting
- Clustered PCA
- Triple Product Integrals
- Handling Local Lighting
- Direct-to-Indirect Transfer


## SH-based PRT

- Better light integration and transport
- dynamic, env. lights
- self-shadowing
- interreflections

point light


Env. light


Env. lighting, shadows

## SH-based PRT: Idea



## PRT Terminology



## Relation to a Matrix-Vector Multiply

a) SH coefficients of transferred radiance
b) Irradiance (per vertex)

$$
=\left[\begin{array}{cccc}
T_{11} & T_{12} & \cdots & T_{1 M} \\
T_{21} & T_{22} & \cdots & T_{2 M} \\
T_{31} & T_{32} & \cdots & T_{3 M} \\
\vdots & \vdots & \ddots & \vdots \\
T_{N 1} & T_{N 2} & \cdots & T_{N M}
\end{array}\right]\left[\begin{array}{c}
L_{1} \\
L_{2} \\
\vdots \\
L_{M}
\end{array}\right]\left[\begin{array}{c}
\text { SH coefficients } \\
\text { of EM (source } \\
\text { radiance) }
\end{array}\right.
$$

## Idea of SH-based PRT

- The $L$ vector is projected onto low-frequency components (say 25). Size greatly reduced.
- Hence, only 25 matrix columns
- But each pixel/vertex still treated separately
- One RGB value per pixel/vertex:
- Diffuse shading / arbitrary BRDF shading w/ fixed view direction
- SH coefficients of transferred radiance (25 RGB values per pixel/vertex for order 4 SH)
- Arbitrary BRDF shading w/ variable view direction
- Good technique (becoming common in games) but useful only for broad low-frequency lighting


## Diffuse Transfer Results



No Shadows/Inter


Shadows


Shadows+Inter

## SH-based PRT with Arbitrary BRDFs

- Combine with Kautz et al. 03
- Transfer matrix turns SH env. map into SH transferred radiance
- Kautz et al. 03 is applied to transferred radiance



## Arbitrary BRDF Results



Anisotropic BRDFs


Spatially Varying

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## PCA or SVD factorization

- SVD:

- Applying Rank b:

- Absorbing $\mathbf{S i}^{\mathbf{j}}$ values into $\mathbf{C i T}^{\mathbf{i T}}$



## Idea of Compression

- Represent matrix (rather than light vector) compactly
- Can be (and is) combined with SH light vector
- Useful in broad contexts.
- BRDF factorization for real-time rendering (reduce 4D BRDF to 2D texture maps) McCool et al. 01 etc
- Surface Light field factorization for real-time rendering (4D to 2D maps) Chen et al. 02, Nishino et al. 01
- BTF (Bidirectional Texture Function) compression
- Not too useful for general precomput. relighting
" Transport matrix not low-dimensional!!


## Local or Clustered PCA

- Exploit local coherence (in say 16x16 pixel blocks)
- Idea: light transport is locally low-dimensional.
- Even though globally complex
- See Mahajan et al. 07 for theoretical analysis
- Clustered PCA [Sloan et al. 2003]
- Combines two widely used compression techniques: Vector Quantization or VQ and Principal Component Analysis


## Compression Example

## Surface is curve, signal is normal

Following couple of slides courtesy P.-P. Sloan

## Compression Example

Signal Space

## Cluster normals

Replace samples with cluster mean

$$
\mathbf{M}_{p} \approx \tilde{\mathbf{M}}_{p}=\mathbf{M}_{C_{p}}
$$

## PCA

Replace samples with mean + linear combination

$$
\mathbf{M}_{p} \approx \tilde{\mathbf{M}}_{p}=\mathbf{M}^{0}+\sum^{N} w_{p}^{i} \mathbf{M}^{i}
$$

## CPCA

Compute a linear subspace in each cluster

$$
\mathbf{M}_{p} \approx \tilde{\mathbf{M}}_{p}=\mathbf{M}_{C_{p}}^{0}+\sum_{i=1}^{N} w_{p}^{i} \mathbf{M}_{C_{p}}^{i}
$$



## CPCA

- Clusters with low dimensional affine models
- How should clustering be done?
- $k$-means clustering
- Static PCA
- VQ, followed by one-time per-cluster PCA
- optimizes for piecewise-constant reconstruction
- Iterative PCA
- PCA in the inner loop, slower to compute
- optimizes for piecewise-affine reconstruction

Static vs. Iterative


## Equal Rendering Cost



VQ


PCA


CPCA

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## Sparse Matrix-Vector Multiplication

Choose data representations with mostly zeroes
Vector: Use non-linear wavelet approximation on lighting

Matrix: Wavelet-encode transport rows


## Haar Wavelet Basis



## Non-linear Wavelet Approximation

Wavelets provide dual space / frequency locality

- Large wavelets capture low frequency area lighting
- Small wavelets capture high frequency compact features

Non-linear Approximation
" Use a dynamic set of approximating functions (depends on each frame's lighting)

- By contrast, linear approx. uses fixed set of basis functions (like 25 lowest frequency spherical harmonics)
- We choose 10 's - 100's from a basis of 24,576 wavelets (64×64×6)


## Non-linear Wavelet Light Approximation

Wavelet Transform


## Non-linear Wavelet Light Approximation

$\left[\begin{array}{c}0 \\ L_{2} \\ 0 \\ 0 \\ 0 \\ L_{6} \\ \vdots \\ 0\end{array}\right]$

## Non-linear <br> Approximation

Retain 0.1\% - 1\% terms

## Error in Lighting: St Peter's Basilica



## Output Image Comparison

## Top: <br> Linear Spherical Harmonic Approximation

 Bottom: Non-linear Wavelet Approximation

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## SH + Clustered PCA

- Described earlier (combine Sloan 03 with Kautz 03)
- Use low-frequency source light and transferred light variation (Order 5 spherical harmonic $=25$ for both; total $=$ 25*25=625)
- 625 element vector for each vertex
- Apply CPCA directly (Sloan et al. 2003)
- Does not easily scale to high-frequency lighting
- Really cubic complexity (number of vertices, illumination directions or harmonics, and view directions or harmonics)
- Practical real-time method on GPU


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## Problem Characterization

6D Precomputation Space

- Distant Lighting
(2D)
View
(2D)
Rigid Geometry
(2D)

With ~ 100 samples per dimension
$\sim 10^{12}$ samples total!! : Intractable computation, rendering

## Factorization Approach



## Triple Product Integral Relighting



## Relit Images (3-5 sec/frame)



## Triple Product Integrals

$$
\begin{aligned}
B & =\int_{S^{2}} L(\omega) V(\omega) \tilde{\rho}(\omega) d \omega \\
& =\int_{S^{2}}\left(\sum_{i} L_{i} \Psi_{i}(\omega)\right)\left(\sum_{j} V_{j} \Psi_{j}(\omega)\right)\left(\sum_{k} \tilde{\rho}_{k} \Psi_{k}(\omega)\right) d \omega \\
& =\sum_{i} \sum_{j} \sum_{k} L_{i} V_{j} \tilde{\rho}_{k} \int_{S^{2}} \Psi_{i}(\omega) \Psi_{j}(\omega) \Psi_{k}(\omega) d \omega \\
& =\sum_{i} \sum_{j} \sum_{k} L_{i} V_{j} \tilde{\rho}_{k} C_{i j k}
\end{aligned}
$$

## Basis Requirements



1. Need few non-zero "tripling" coefficients

$$
C_{i j k}=\int_{S^{2}} \Psi_{i}(\omega) \Psi_{j}(\omega) \Psi_{k}(\omega) d \omega
$$

2. Need sparse basis coefficients


## 1. Number Non-Zero Tripling Coeffs

$$
C_{i j k}=\int_{S^{2}} \Psi_{i}(\omega) \Psi_{j}(\omega) \Psi_{k}(\omega) d \omega
$$

| Basis Choice | Number Non-Zero $C_{i j k}$ |
| :--- | :---: |
| General (e.g. PCA) | $O\left(N^{3}\right)$ |
| Sph. Harmonics | $O\left(N^{5 / 2}\right)$ |
| Haar Wavelets | $O(N \log N)$ |

## 2. Sparsity in Light Approx.




Approximation Terms

## Summary of Wavelet Results

- Derive direct $O(N \log N)$ triple product algorithm
- Dynamic programming can eliminate log $N$ term
- Final complexity linear in number of retained basis coefficients


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## Direct-to-Indirect Transfer

- Lighting non-linear w.r.t. light source parameters (position, orientation etc.)
- Indirect is a linear function of direct illumination
- Direct can be computed in real-time on GPU
- Transfer of direct to indirect is pre-computed
- Hašan et al. 06
- Fixed view - cinematic relighting with GI


## DTIT: Matrix-Vector Multiply



$$
=\left[\begin{array}{cccc}
T_{11} & T_{12} & \cdots & T_{1 M} \\
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T_{31} & T_{32} & \cdots & T_{3 M} \\
\vdots & \vdots & \ddots & \vdots \\
T_{N 1} & T_{N 2} & \cdots & T_{N M}
\end{array}\right]\left[\begin{array}{c}
L_{1} \\
L_{2} \\
\vdots \\
L_{M}
\end{array}\right]\left[\begin{array}{c}
\text { Direct } \\
\text { illumination on a } \\
\text { set of samples } \\
\text { distributed on } \\
\text { scene surfaces }
\end{array}\right]
$$

Compression: Matrix rows in Wavelet basis

## DTIT: Demo

## Summary

- Really a big data compression and signalprocessing problem
- Apply many standard methods
- PCA, wavelet, spherical harmonic, factor compression
- And invent new ones
- VQPCA, wavelet triple products
- Guided by and gives insights into properties of illumination, reflectance, visibility
" How many terms enough? How much sparsity?

