#### **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**



Incident Radiance (Illumination Environment Map)

Irradiance Environment Map

## **SH-based Irradiance Env. Maps**



Images courtesy Ravi Ramamoorthi & Pat Hanrahan

# **SH-based Arbitrary BRDF Shading 1**

- [Kautz et 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



#### **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
  - **D** 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**



$$= \begin{bmatrix} T_{11} & T_{12} & \cdots & T_{1M} \\ T_{21} & T_{22} & \cdots & T_{2M} \\ T_{31} & T_{32} & \cdots & T_{3M} \\ \vdots & \vdots & \ddots & \vdots \\ T_{N1} & T_{N2} & \cdots & T_{NM} \end{bmatrix} \begin{bmatrix} L_1 \\ L_2 \\ \vdots \\ L_M \end{bmatrix}$$

### **Relighting as a Matrix-Vector Multiply**



#### **Output Image** (Pixel Vector)

#### 

Precomputed Transport Matrix

$$\begin{bmatrix} T_{11} & T_{12} & \cdots & T_{1M} \\ T_{21} & T_{22} & \cdots & T_{2M} \\ T_{31} & T_{32} & \cdots & T_{3M} \\ \vdots & \vdots & \ddots & \vdots \\ T_{N1} & T_{N2} & \cdots & T_{NM} \end{bmatrix} \begin{pmatrix} Cubemap Vector \\ L_1 \\ L_2 \\ \vdots \\ L_M \end{bmatrix}$$

# Matrix Columns (Images)

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



# **Problem Definition**

#### Matrix is Enormous

- 512 x 512 pixel images
- 6 x 64 x 64 cubemap environments

# Full matrix-vector multiplication is intractable On the order of 10<sup>10</sup> 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
- For diffuse and glossy surfaces
- At real-time rates
- Sloan et al. 02



point light



Env. light



Env. lighting, no shadows

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)



$$= \begin{bmatrix} T_{11} & T_{12} & \cdots & T_{1M} \\ T_{21} & T_{22} & \cdots & T_{2M} \\ T_{31} & T_{32} & \cdots & T_{3M} \\ \vdots & \vdots & \ddots & \vdots \\ T_{N1} & T_{N2} & \cdots & T_{NM} \end{bmatrix} \begin{bmatrix} L_1 \\ L_2 \\ \vdots \\ L_M \end{bmatrix}$$

SH coefficients of EM (source radiance)

# 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** 

**Other BRDFs** 

**Spatially Varying** 

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# **PCA or SVD factorization**

• SVD:



• Applying Rank **b**:



• Absorbing **S<sup>j</sup>** values into **C<sup>iT</sup>**:



# **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

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





#### Replace samples with mean + linear combination

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



i=1



Compute a linear subspace in each cluster

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





 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



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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 (64x64x6)

## **Non-linear Wavelet Light Approximation**

#### Wavelet Transform



### Non-linear Wavelet Light Approximation



#### Non-linear Approximation

Retain 0.1% – 1% terms

#### **Error in Lighting: St Peter's Basilica**



Ng, Ramamoorthi, Hanrahan 03

# **Output Image Comparison**

Top:Linear Spherical Harmonic ApproximationBottom:Non-linear Wavelet Approximation



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# Outline

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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 ~ 10<sup>12</sup> samples total!! : Intractable computation, rendering

## **Factorization Approach**



# **Triple Product Integral Relighting**







# Relit Images (3-5 sec/frame)









# **Triple Product Integrals**

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

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

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

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

## **Basis Requirements**

$$B = \sum_{i} \sum_{j} \sum_{k} L_{i} V_{j} \tilde{\rho}_{k} C_{ijk}$$

1. Need few non-zero "tripling" coefficients

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

2. Need sparse basis coefficients  $L_i, V_j, \tilde{\rho}_k$ 

## **1. Number Non-Zero Tripling Coeffs**

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

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

# 2. Sparsity in Light Approx.



# **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**



$$= \begin{bmatrix} T_{11} & T_{12} & \cdots & T_{1M} \\ T_{21} & T_{22} & \cdots & T_{2M} \\ T_{31} & T_{32} & \cdots & T_{3M} \\ \vdots & \vdots & \ddots & \vdots \\ T_{N1} & T_{N2} & \cdots & T_{NM} \end{bmatrix} \begin{bmatrix} L_1 \\ L_2 \\ \vdots \\ L_M \end{bmatrix}$$
Direct illumination on a set of samples distributed on scene surfaces

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?