

Sampling and Reconstruction of Visual Appearance: From Denoising to View Synthesis

CSE 274 [Fall 2022], Lecture 9

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Applications

- Monte Carlo Rendering
- *Light Transport Acquisition / Many Light Rendering*
- Light Fields and Computational Photography
- View Synthesis
- Animation/Simulation (not covered in course)
- Introduce concepts of sparsity, coherence, compressive sensing for reconstruction

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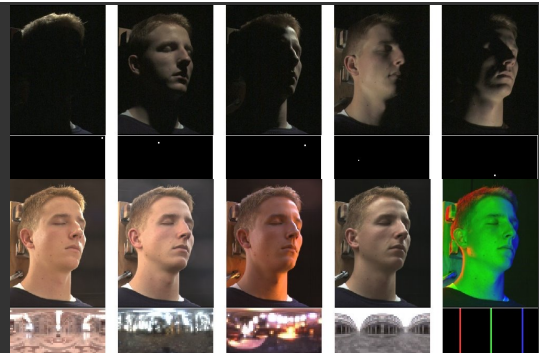
Acquiring Reflectance Field of Human Face [Debevec et al. SIGGRAPH 00]

Illuminate subject from many incident directions



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Example Images



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Motivation: Image-based Relighting



Sample Lighting Directions

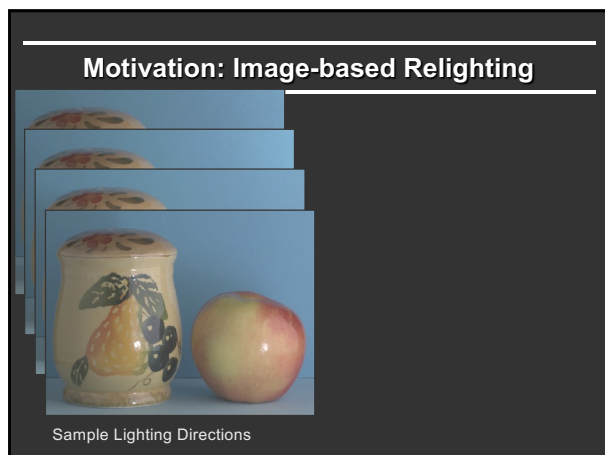
5

Motivation: Image-based Relighting

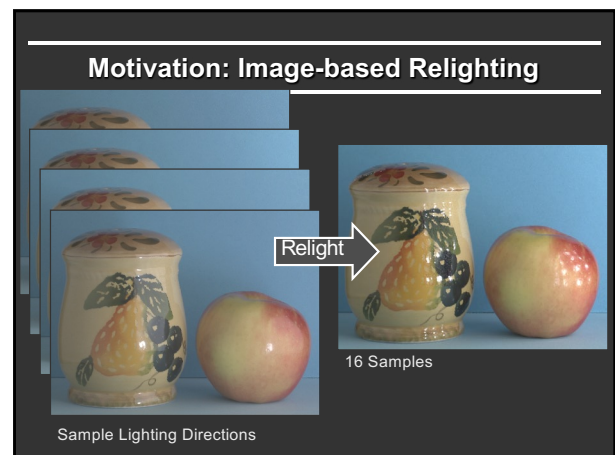


Sample Lighting Directions

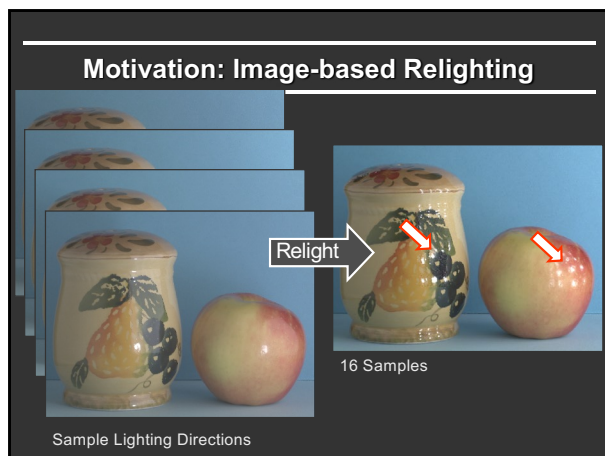
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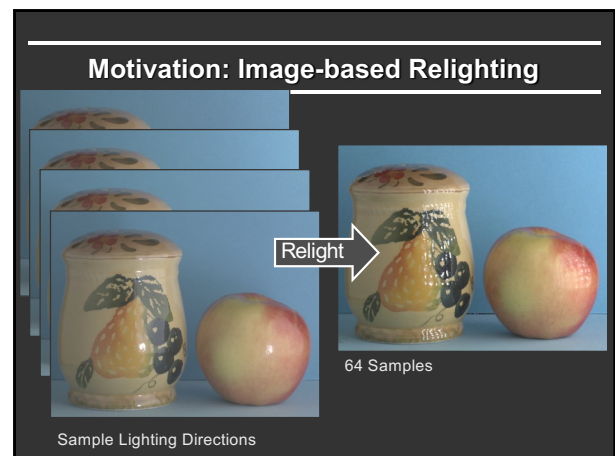
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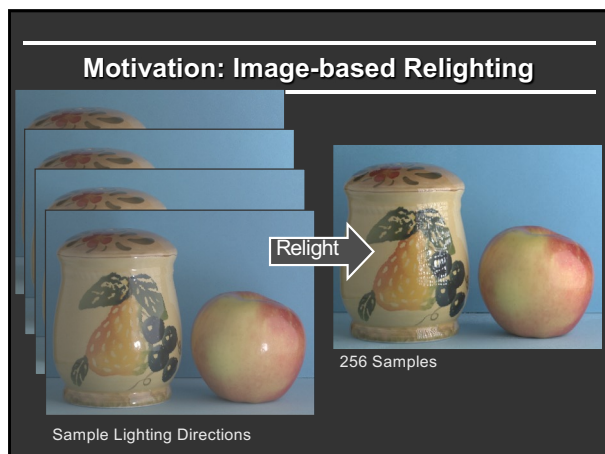
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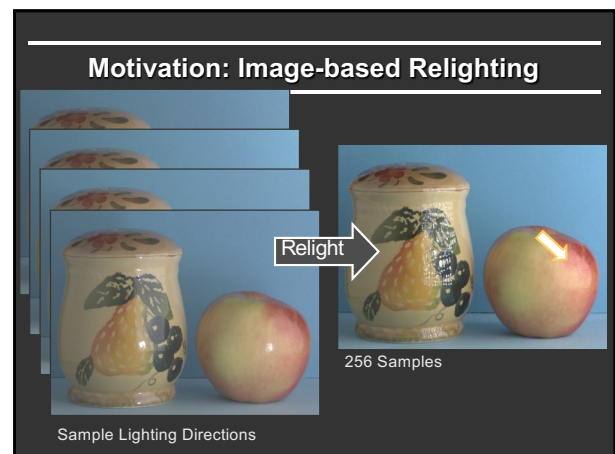
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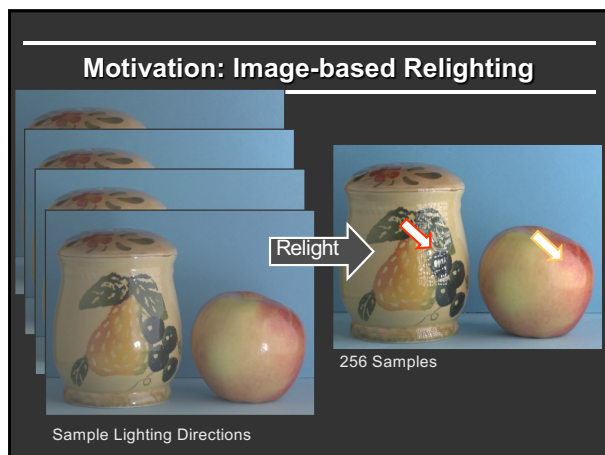
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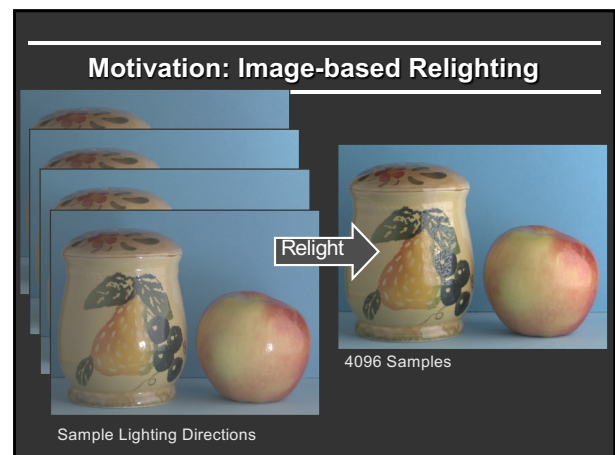
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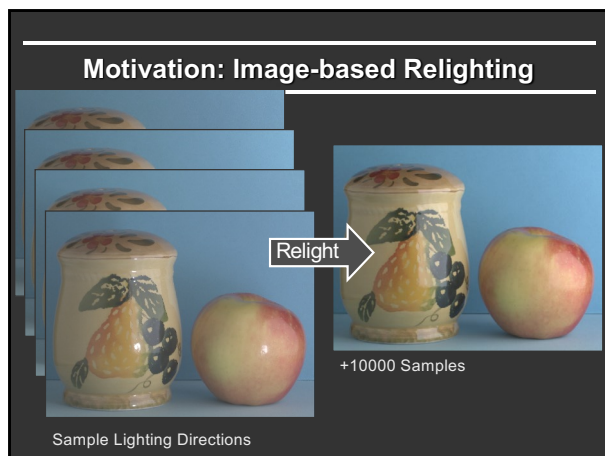
12



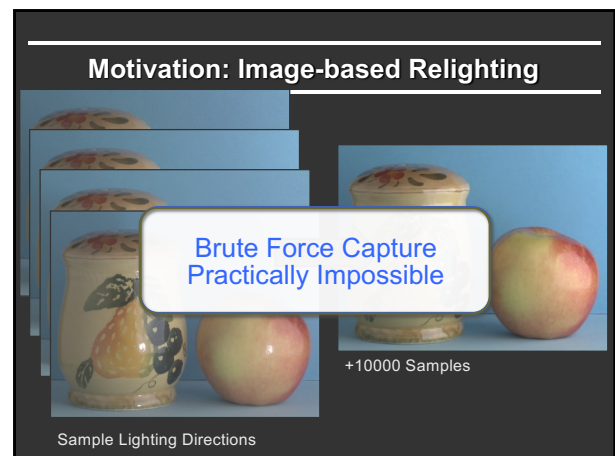
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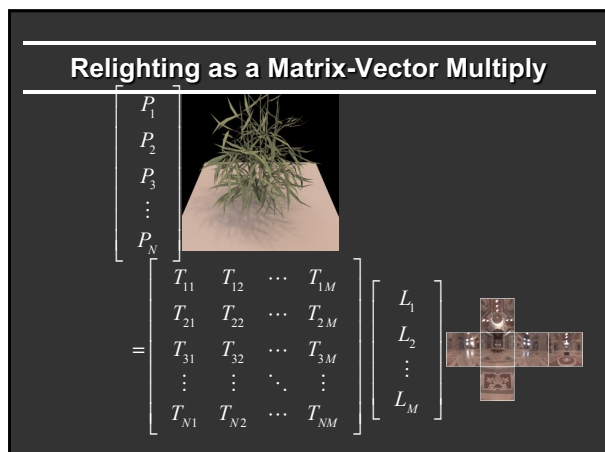
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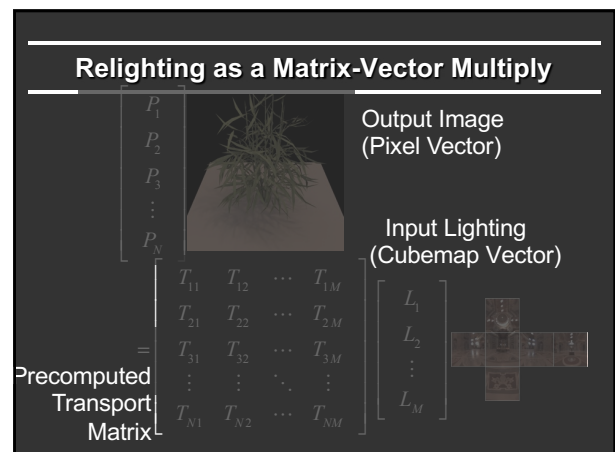
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Matrix Columns (Images)

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



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(Pre)compute: Ray-Trace Image Cols

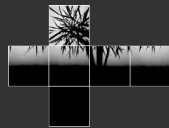
$$\begin{bmatrix} T_{11} & T_{12} & \cdots & T_{1M} \\ T_{21} & T_{22} & \cdots & T_{2M} \\ T_{31} & T_{32} & \cdots & T_{3M} \\ \vdots & \vdots & \ddots & \vdots \\ T_{N1} & T_{N2} & \cdots & T_{NM} \end{bmatrix}$$



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(Pre)compute 2: Rasterize Matrix Rows

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



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Outline

- Matrix Row-Column Sampling (*Many Lights*)
(clustering for matrix completion of light transport)
- Compressive Sensing for Light Transport
- Matrix Completion

Hasan, Pellacini, Bala SIGGRAPH 07

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Complex Illumination: A Challenge



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Conversion to Many Lights

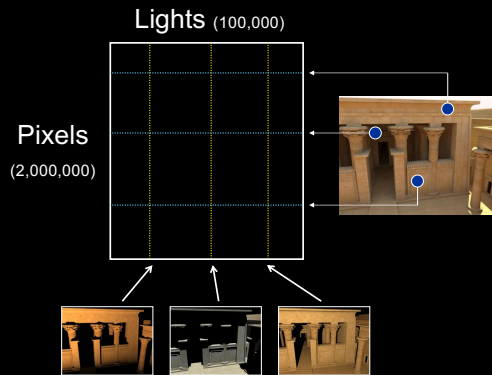
- Area, indirect, sun/sky



Courtesy Walter et al., Lightcuts, SIGGRAPH 05/06

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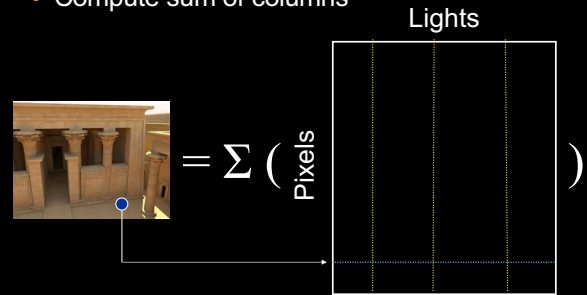
A Matrix Interpretation



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Problem Statement

- Compute sum of columns

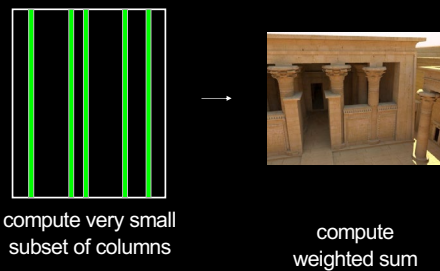


- **Note:** We don't have the matrix data

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Image as a Weighted Column Sum

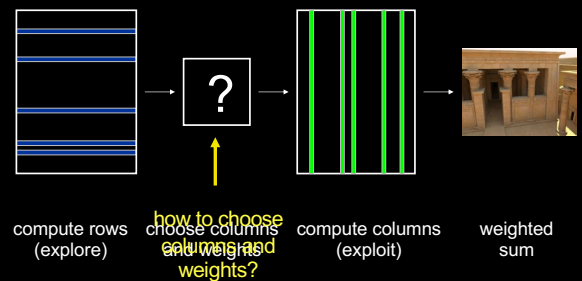
- The following is possible:



- Use rows to choose a good set of columns!

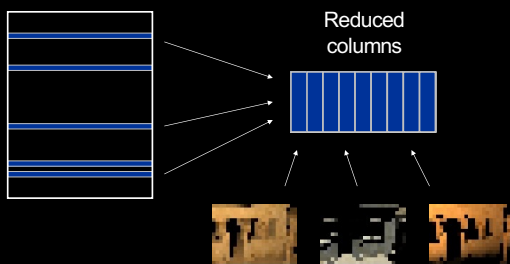
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Exploration and Exploitation



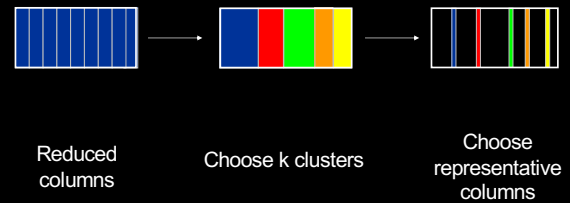
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Reduced Matrix

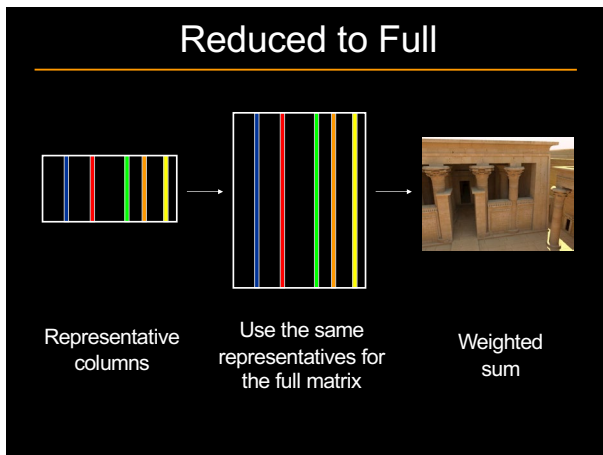


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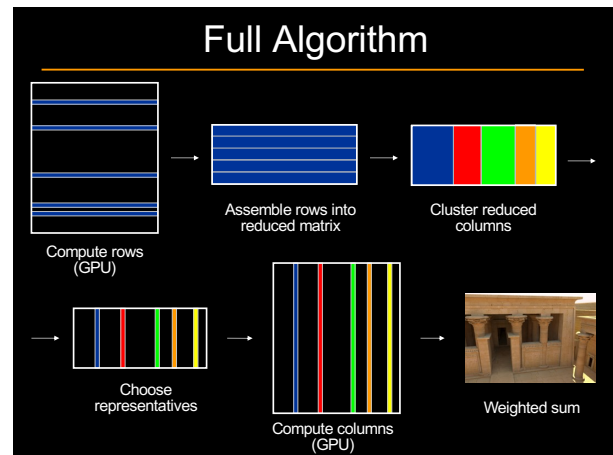
Clustering Approach



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Results

- We show 5 scenes:

Kitchen Temple Trees Bunny Grand Central

- Show reference and 5x difference image
- All scenes have 100,000+ lights
- Timings
 - NVidia GeForce 8800 GTX
 - Light / surface sample creation not included

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Results: Kitchen

- 388k polygons
- Mostly indirect illumination
- Glossy surfaces
- Indirect shadows

Our result: 13.5 sec
(432 rows + 864 columns)

Reference: 13 min
(using all 100k lights)

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Results: Temple

- 2.1m polygons
- Mostly indirect & sky illumination
- Indirect shadows

Our result: 16.9 sec
(300 rows + 900 columns)

Reference: 20 min
(using all 100k lights)

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Results: Trees

- 328k polygons
- Complex incoherent geometry

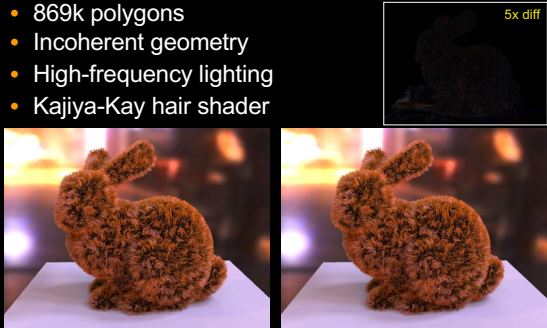
Our result: 2.9 sec
(100 rows + 200 columns)

Reference: 14 min
(using all 100k lights)

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Results: Bunny

- 869k polygons
- Incoherent geometry
- High-frequency lighting
- Kajiya-Kay hair shader




Our result: 3.8 sec
(100 rows + 200 columns)

Reference: 10 min
(using all 100k lights)

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Results: Grand Central

- 1.5m polygons
- Point lights between stone blocks



Our result: 24.2 sec
(588 rows + 1176 columns)

Reference: 44 min
(using all 100k lights)

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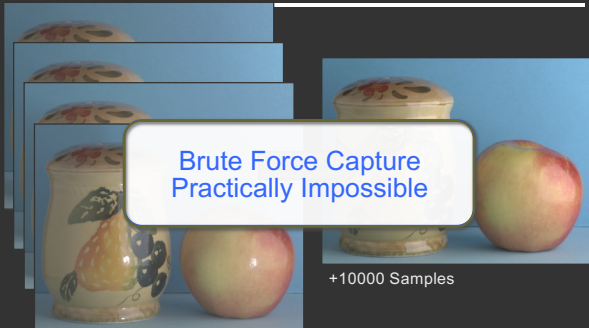
Outline

- Matrix Row-Column Sampling (Many Lights)
(clustering for matrix completion of light transport)
- Compressive Sensing for Light Transport*
- Matrix Completion

Gu et al. ECCV 08
Peers et al. SIGGRAPH 09
Sen and Darabi EG 09 (reading)

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Motivation: Image-based Relighting



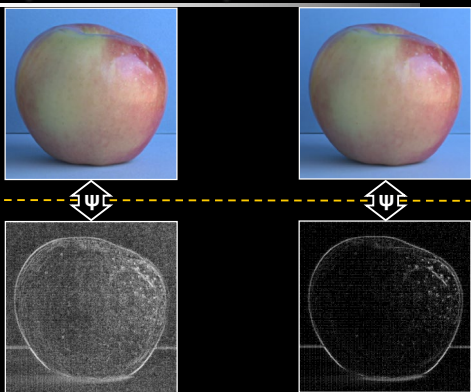
Brute Force Capture
Practically Impossible

+10000 Samples

Sample Lighting Directions

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Compressible / Sparseness

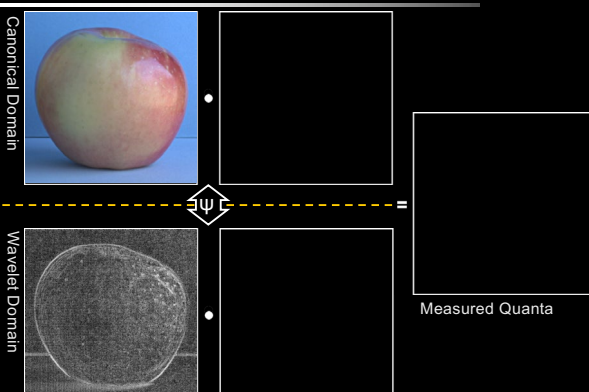


All Coefficients

5% Largest Coeff.

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Measurements

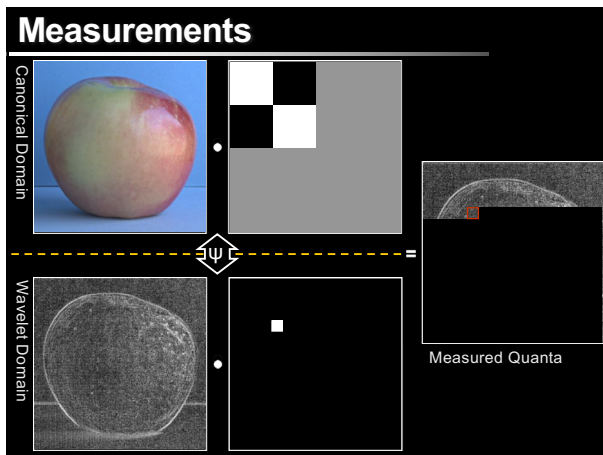


Canonical Domain

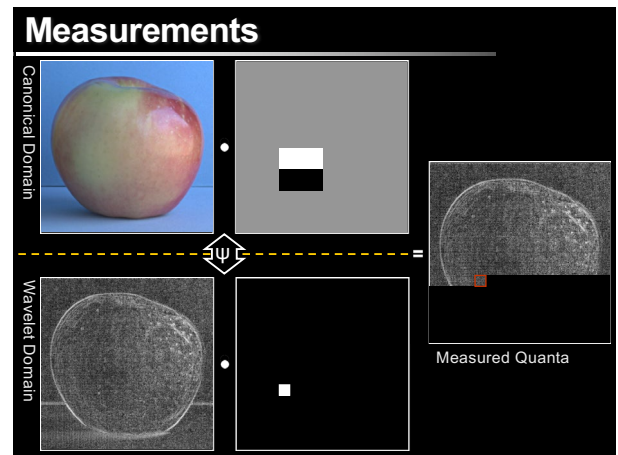
Wavelet Domain

Measured Quanta

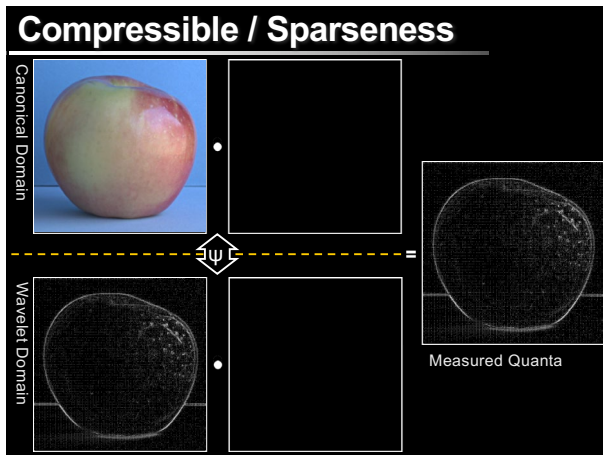
42



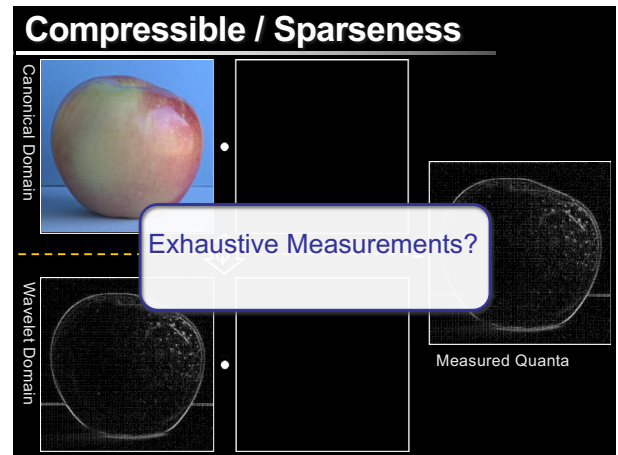
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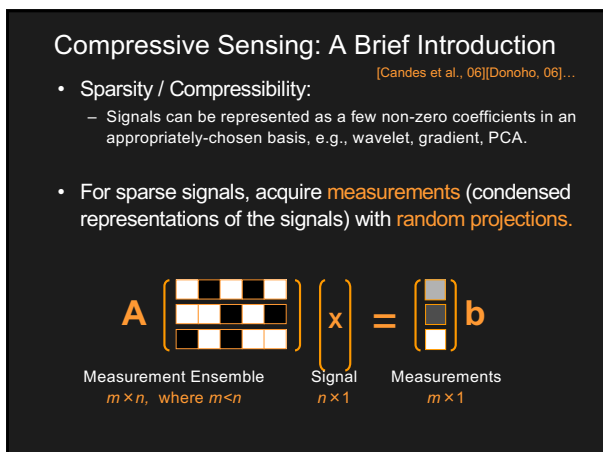
44



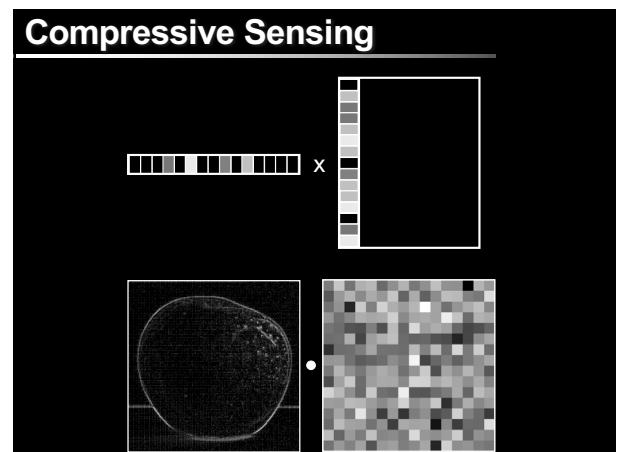
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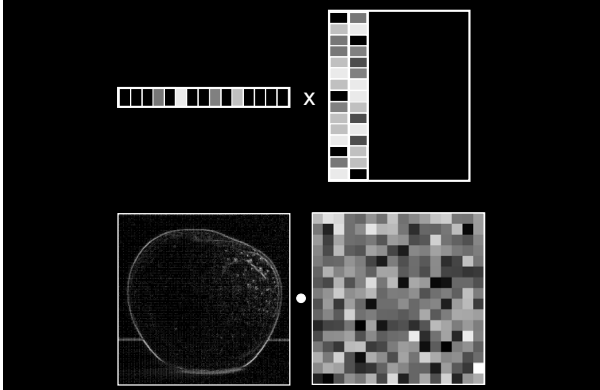


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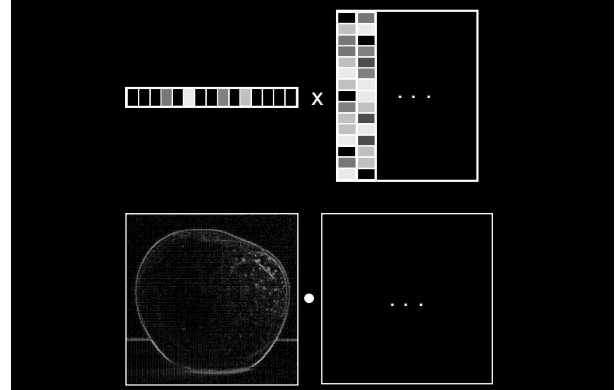
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Compressive Sensing



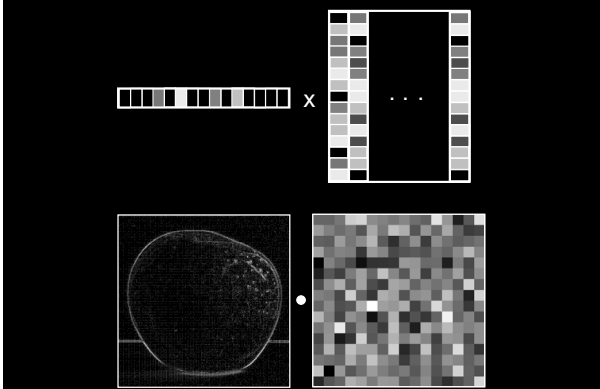
49

Compressive Sensing



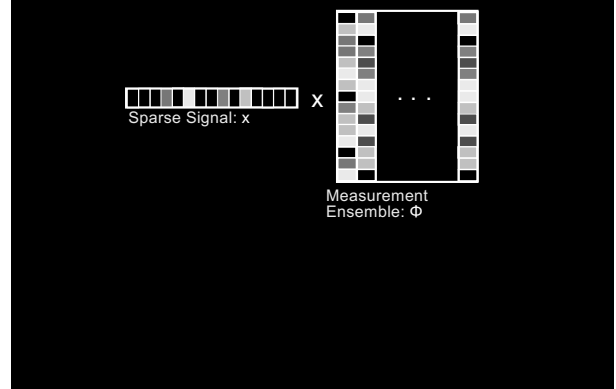
50

Compressive Sensing



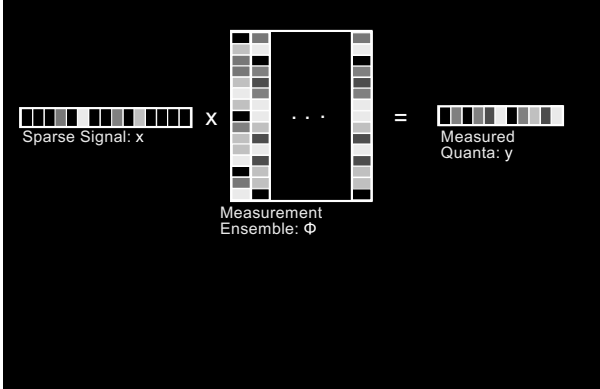
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Compressive Sensing



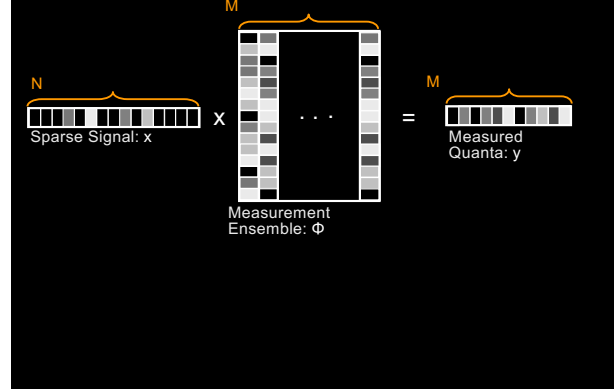
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Compressive Sensing



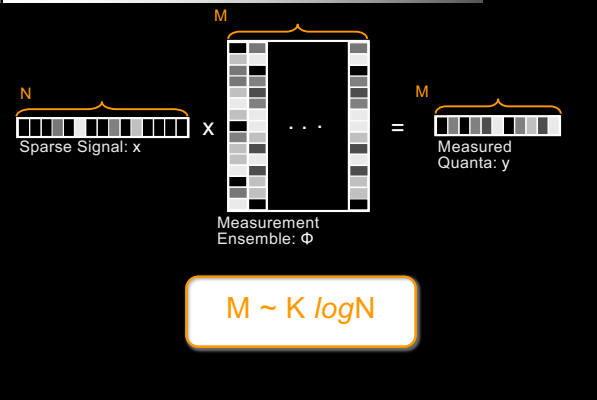
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Compressive Sensing



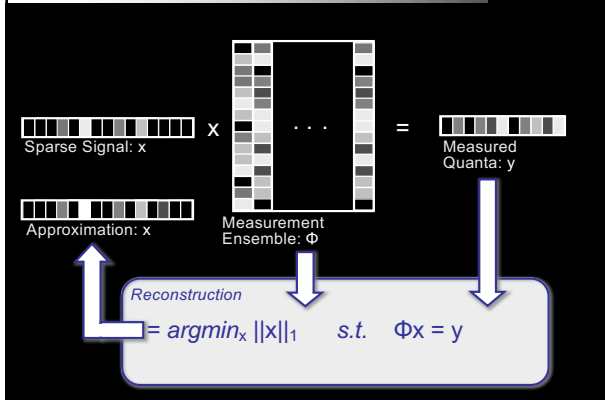
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Compressive Sensing



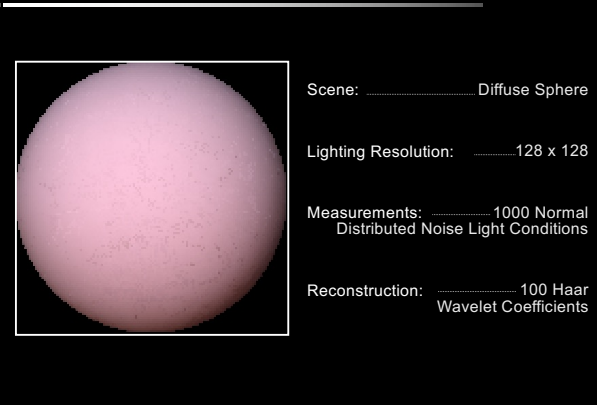
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Compressive Sensing



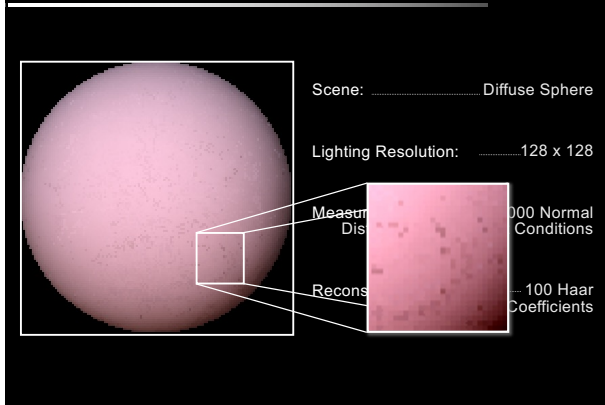
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Brute Force: Result



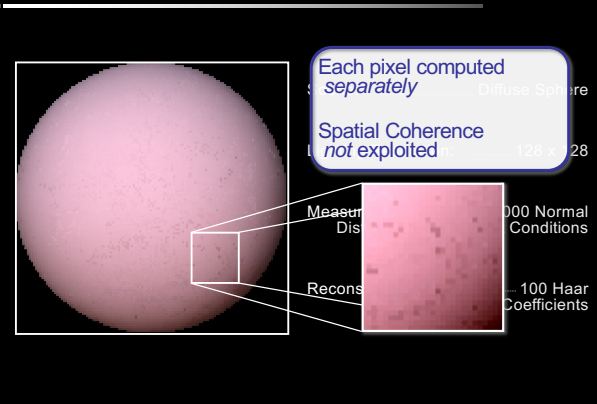
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Brute Force: Result



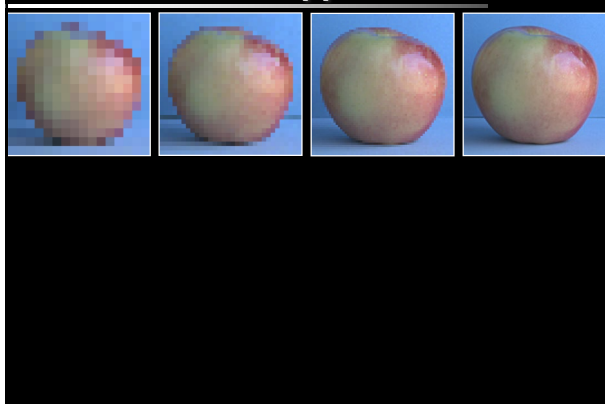
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Brute Force: Result



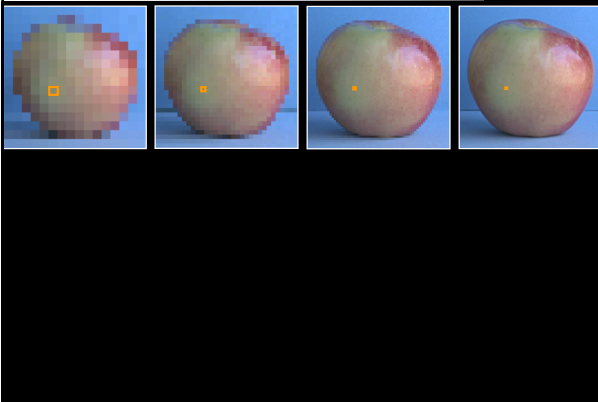
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Multi-resolution Approach



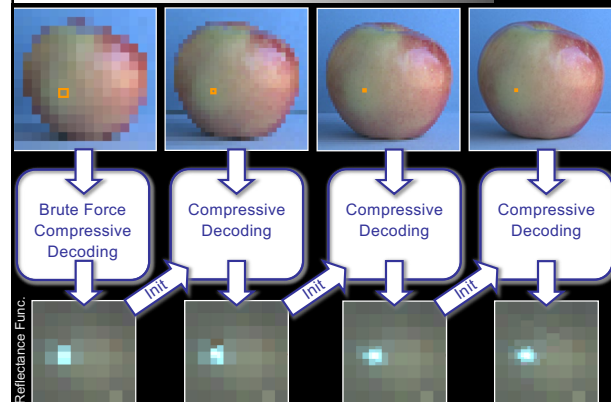
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Multi-resolution Approach



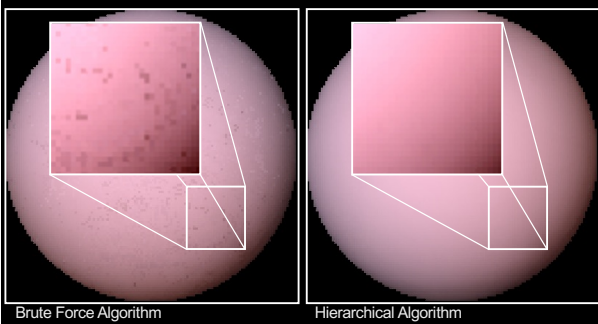
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Multi-resolution Approach



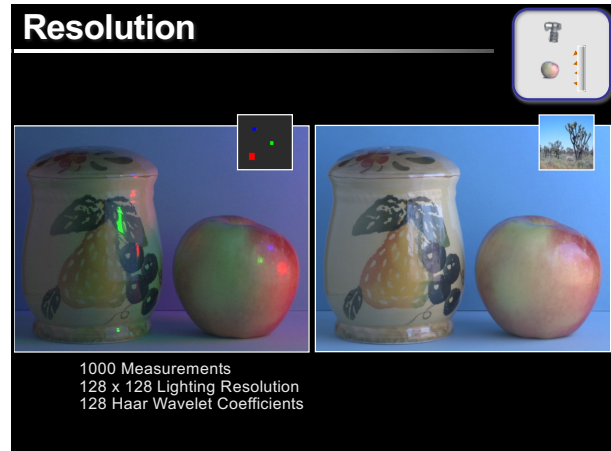
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Results



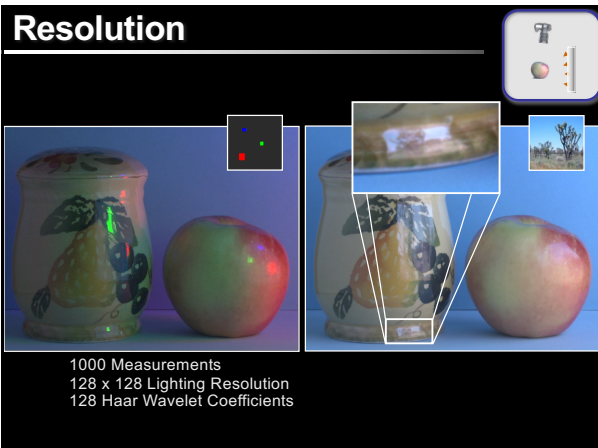
63

Resolution



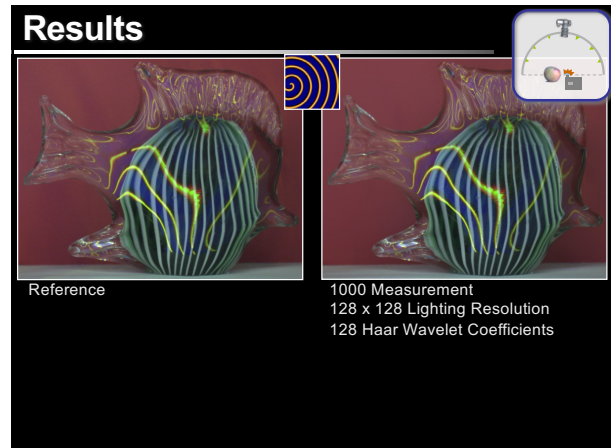
64

Resolution

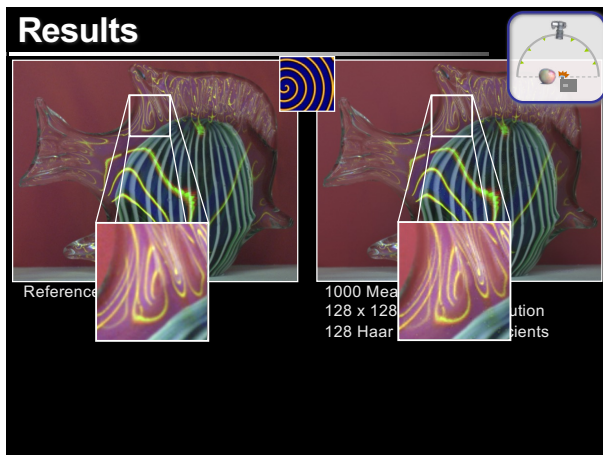


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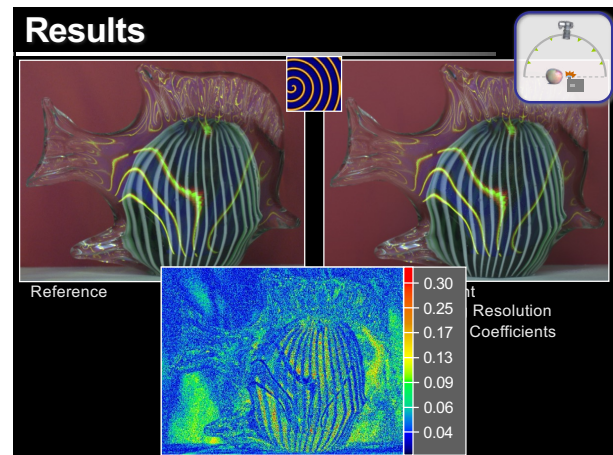
Results



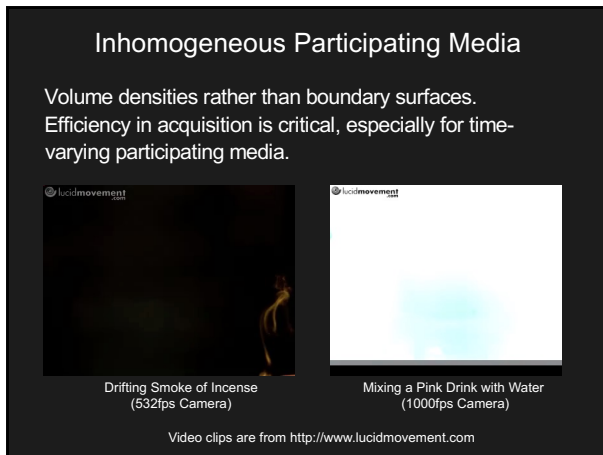
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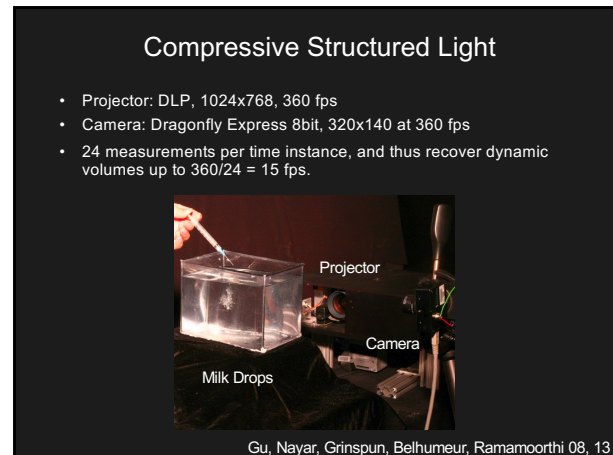
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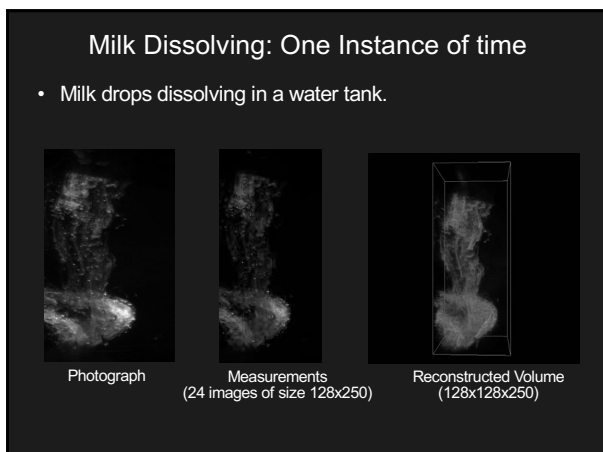
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Outline

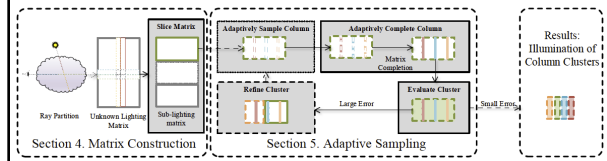
- Matrix Row-Column Sampling (Many Lights)
(clustering for matrix completion of light transport)
- Compressive Sensing for Light Transport
- Matrix Completion
 - Extension to compressive sensing: Low rank matrices
 - Minimize matrix norm (rank), given some entries
 - Combine many ideas seen previously

Huo et al. SIGGRAPH Asia 16

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Outline

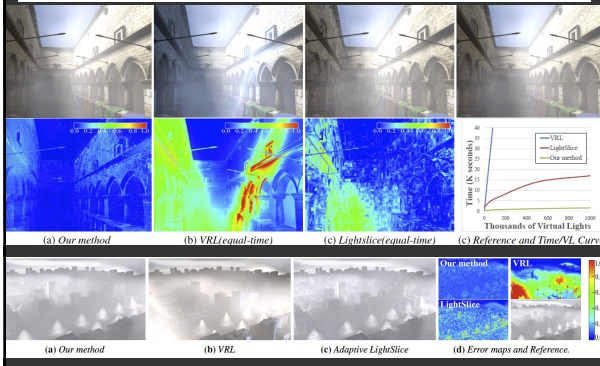
- Matrix Completion
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Huo et al. SIGGRAPH Asia 16

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Results (Participating Media)



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Summary

- Light Transport for Acquisition, Many Light Rendering
- Compressive Sensing for projected patterns
- Matrix Completion for many light rendering
- Leverages popular ideas in applied math
- Consider all forms of coherence
- Think about modern extensions with deep learning

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