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

CSE 274 [Fall 2021], Lecture 5

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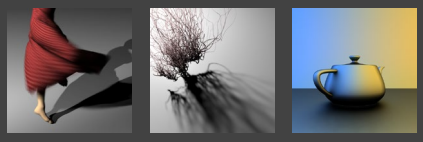
## Applications: Sampling/Reconstruction

- *Monte Carlo Rendering*
- Light Transport Acquisition
- Light Fields and Computational Photography
- View synthesis
- Animation/Simulation (not covered in course)

- Brief overview of these applications today, and opportunities/history for sampling/reconstruction

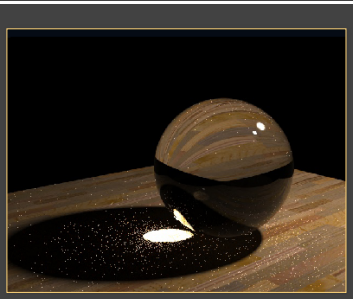
## Motivation

- Distribution effects (depth of field, motion blur, global illumination, soft shadows) are slow. Many dimensions sample



- Ray Tracing physically accurate but slow, not real-time
- Can we adaptively sample and filter for fast, real-time?

## Monte Carlo Path Tracing



1000 paths/pixel

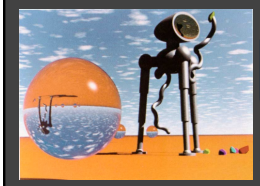
Jensen

## Sampling and Reconstruction

- Monte Carlo is noisy at low sample counts
- Can we reduce time/samples by smart adaptive sampling and smart filtering/reconstruction?
- General area of Monte Carlo denoising
- Long history [Mitchell 91, Guo 98]

## History

- Adaptive sampling old technique Mitchell et al. 87, 91,…



- But not very widely used… artifacts, can miss features
- After seminal papers in 87-91, not much follow on

## Directional Coherence Maps

- Allocate samples to edges (Guo 98) Most of variance at those edges in the image

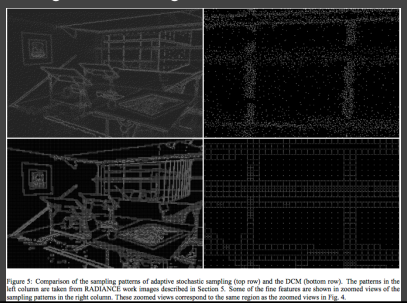


Figure 5: Comparison of the sampling patterns of adaptive stochastic sampling (top row) and the DCM (bottom row). The patterns in the left column are taken from RADIANCE work images described in Section 5. Some of the fine features are shown in zoomed views of the sampling patterns in the right column. These zoomed views correspond to the same region as the zoomed views in Fig. 4.

## Directional Coherence Maps (Guo 98)

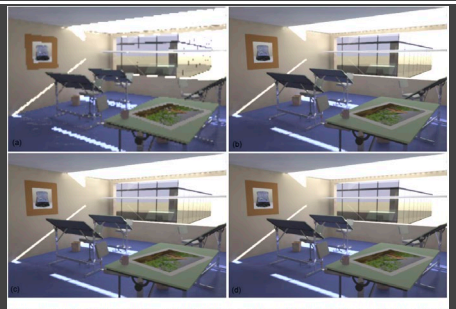
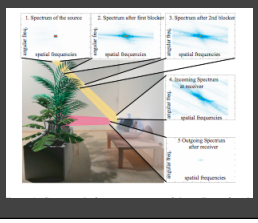


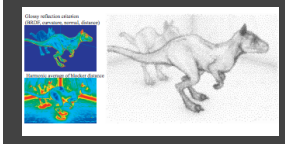
Figure 2: Progressive renderings of an office scene lit by sunlight transferred through a light shelf. (a) The approximate image at the end of the regular subdivision, with 1.6% evaluated pixels located at the corners of the 8 × 8 blocks in the work image. (b) The approximate image after boundary evaluations for all 8 × 8 edge blocks in the work image, with 5% of pixels evaluated. (c) The approximate image after evaluating about 6% of the pixels, whose locations are shown in Fig. 5 (bottom left). (d) The final image as rendered by the baseline RADIANCE system. The scene model was supplied courtesy of Greg W. Larson.

Guo 98

## Resurgence (2008 - )

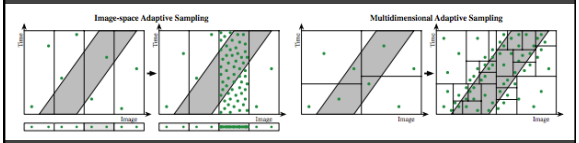
- Eurographics 2015 STAR report by Zwicker et al. [former UCSD faculty, now at Maryland]
- [Durand et al. 2005] Frequency analysis light transport. Proposed use for adaptive sampling. Not very practical



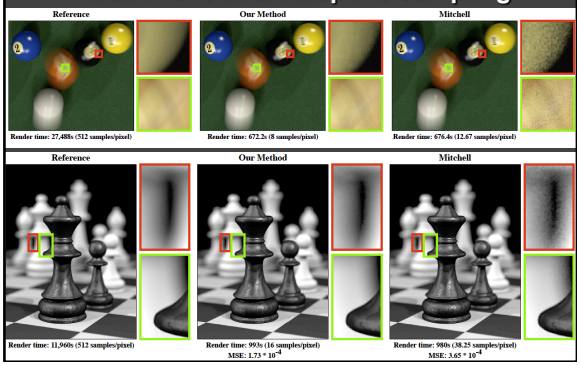


## Multi-Dimensional Adaptive Sampling

- Hachisuka, Jarosz, … Zwicker, Jensen [MDAS 2008]
- Scenes with motion blur, depth of field, soft shadows
- Involves high-dimensional integral, converges slowly
- Exploit high-dimensional info to sample adaptively
- Sampling in 2D image plane or other dims inadequate
  - Need to consider full joint high-dimensional space




## Multidimensional Adaptive Sampling




## Multidimensional Adaptive Sampling

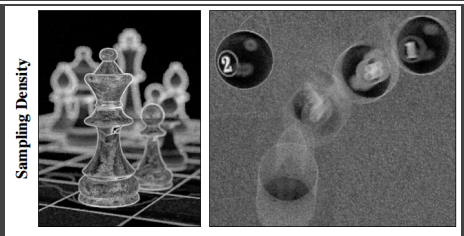


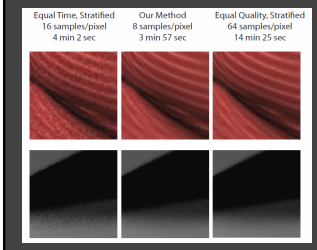

**Figure 10:** *Visualizations of projected sample distributions using our method for the chess scene from Figure 8 and the pool scene from Figure 7. Our adaptive sampler places samples both around high frequency image discontinuities (in focus chess piece and stationary pool ball) as well as in regions which exhibit significant motion blur or depth of field effects.*

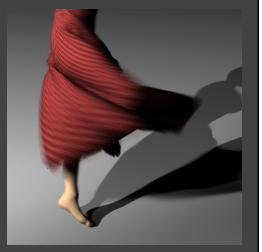
## Multi-Dimensional Adaptive Sampling

Motion Blur and Depth of Field 32 samples per pixel

## A-Priori Methods

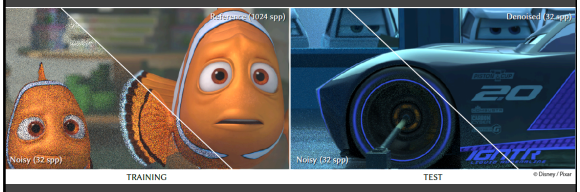
- Egan et al. 2009: Frequency Analysis and Sheared Filtering for Motion Blur; first deep use frequency anal.



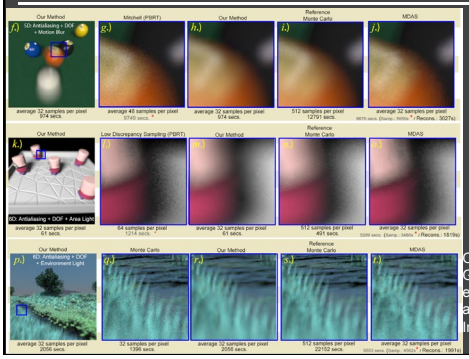
## A-Posteriori Methods

- Adaptive Wavelet Rendering (Overbeck et al. 2009)
- Handle general effects. Sample and denoise
- Many more sophisticated methods available now; used in almost every major production rendering software




Bako et al. 17

## Adaptive Wavelet Rendering



FF VIDEO

Overbeck et al 09
General high-D effects. Simple and fast (renders into wavelet dom)

## Real-Time

- Axis-Aligned Filtering (Mehta et al. 12,13,14)
- Optix plus image-space filtering
- Newer extensions to sheared filtering
- Most recent work (NVIDIA) is fully general, 1 sample per pixel, using modern machine learning methods (similar ideas relevant in offline rendering as well)
- Huge impact in real-time, video games, essential in modern real-time rendering based on deep learning

## Real-Time MAAF Video

**Multiple Axis-Aligned Filters for Rendering of Combined Distribution Effects**

Lifan Wu[1] Ling-Qi Yan[2] Alexandr Kuznetsov[1] Ravi Ramamoorthi[1]
[1] University of California, San Diego [2] University of California, Berkeley

**NO AUDIO**

Recurrent Autoencoder Video (Chaitanya et al. 17)

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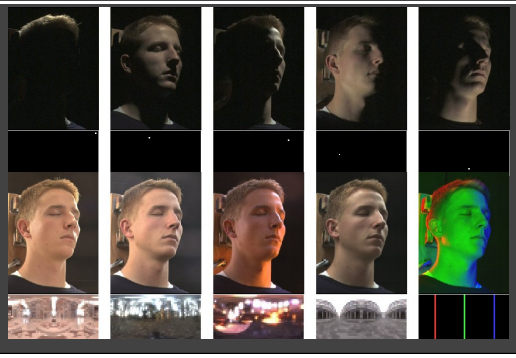
## Acquiring Reflectance Field of Human Face [Debevec et al. SIGGRAPH 00]

Illuminate subject from many incident directions

## Example Images



## Motivation: Image-based Relighting

Sample Lighting Directions

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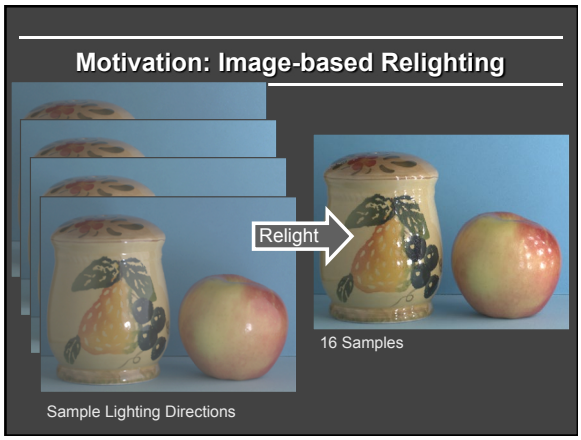
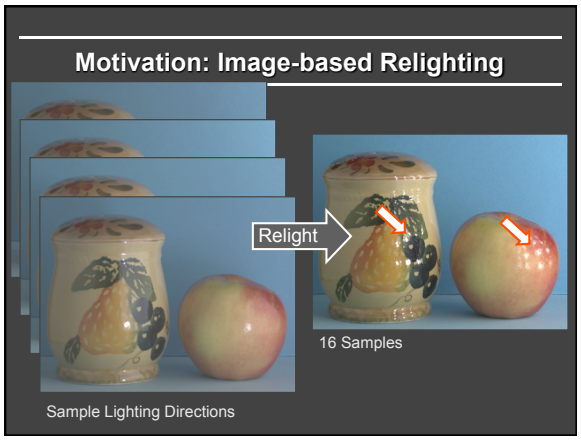
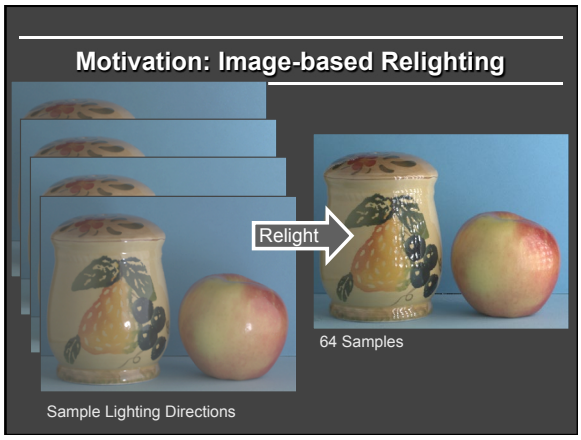
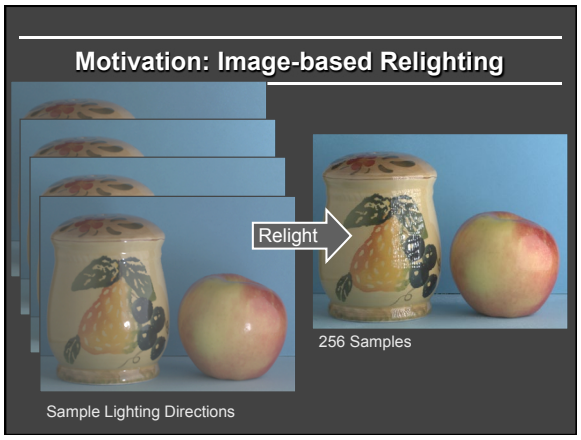
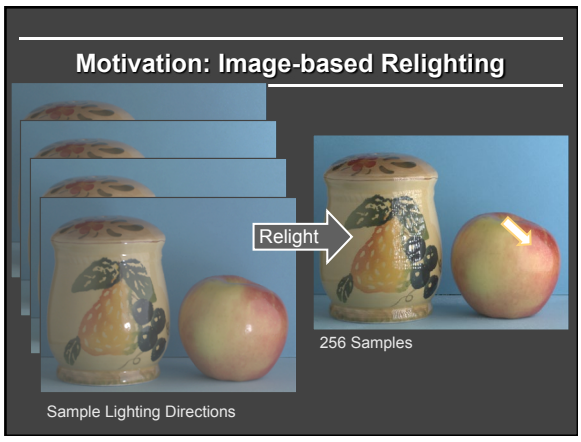
Sample Lighting Directions

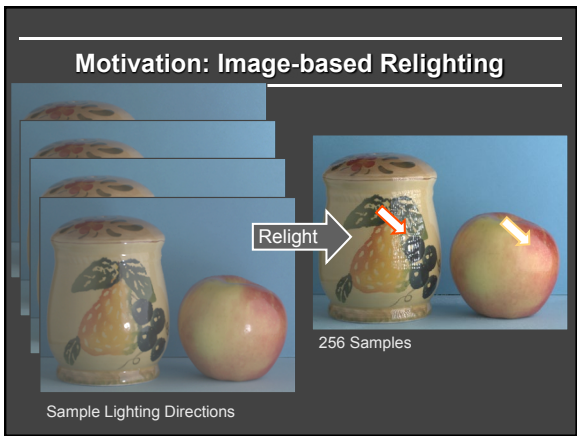

Motivation: Image-based Relighting
Relight
16 Samples
Sample Lighting Directions

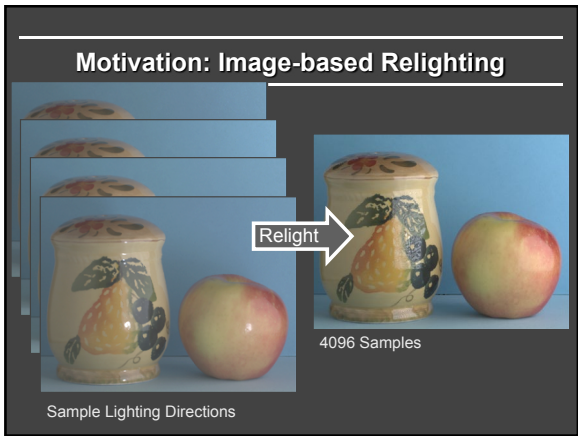

Motivation: Image-based Relighting
Relight
16 Samples
Sample Lighting Directions

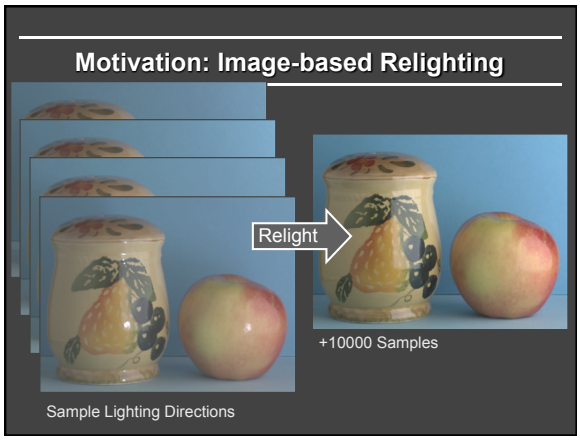

Motivation: Image-based Relighting
Relight
64 Samples
Sample Lighting Directions

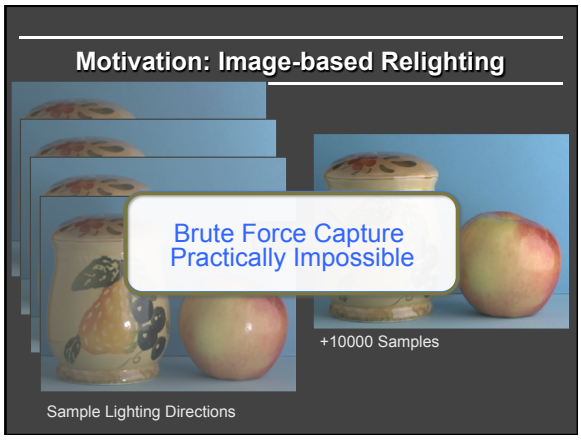

Motivation: Image-based Relighting
Relight
256 Samples
Sample Lighting Directions

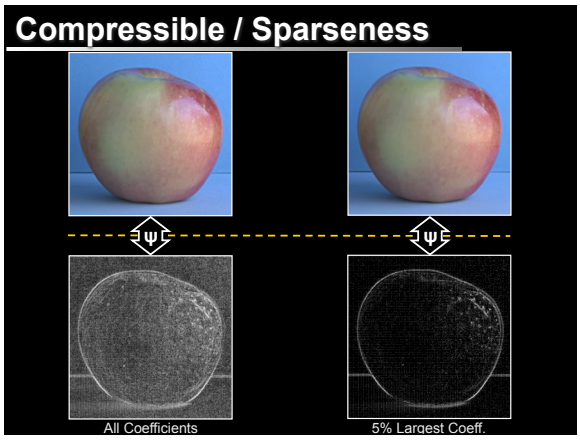

Motivation: Image-based Relighting
Relight
256 Samples
Sample Lighting Directions

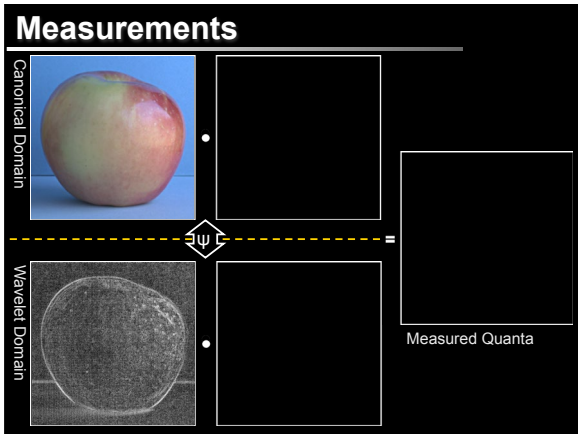

Motivation: Image-based Relighting
Relight
256 Samples
Sample Lighting Directions

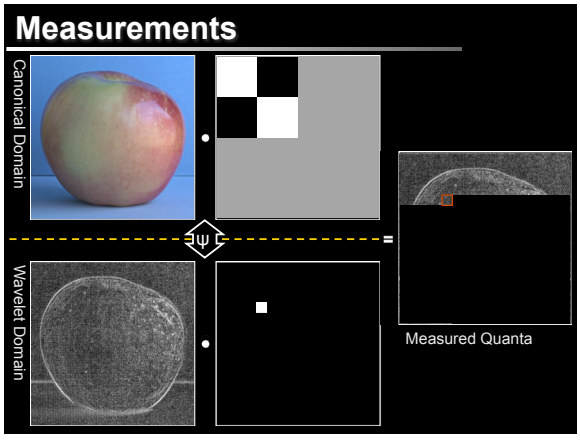

Motivation: Image-based Relighting
Relight
4096 Samples
Sample Lighting Directions


Motivation: Image-based Relighting
Relight
+10000 Samples
Sample Lighting Directions


Motivation: Image-based Relighting
Brute Force Capture
Practically Impossible
+10000 Samples
Sample Lighting Directions


Compressible / Sparseness
All Coefficients
5% Largest Coeff.


Measurements
Canonical Domain
Wavelet Domain
Measured Quanta


Measurements
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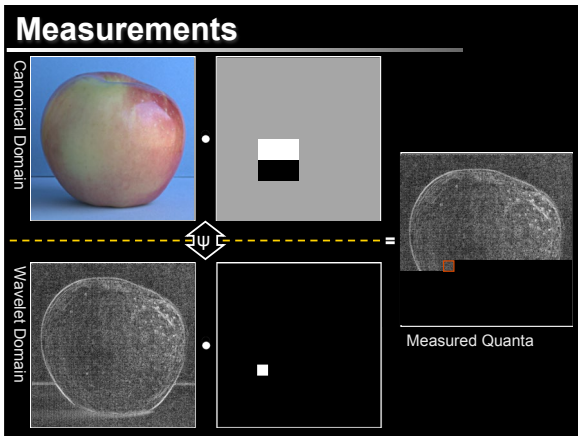
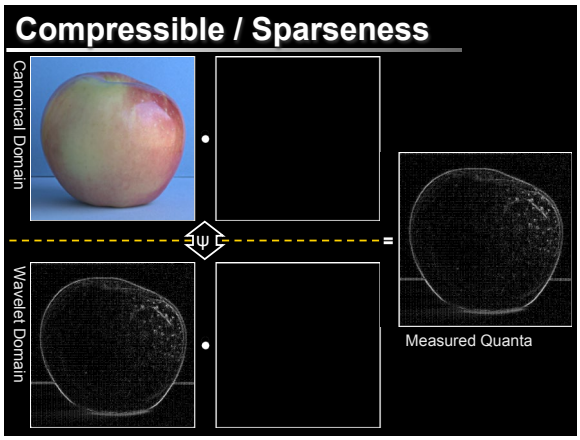
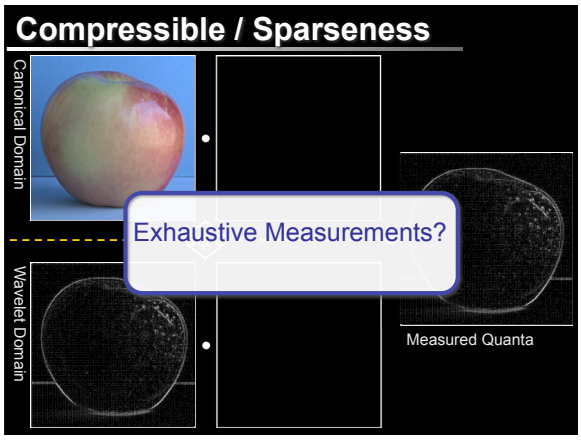


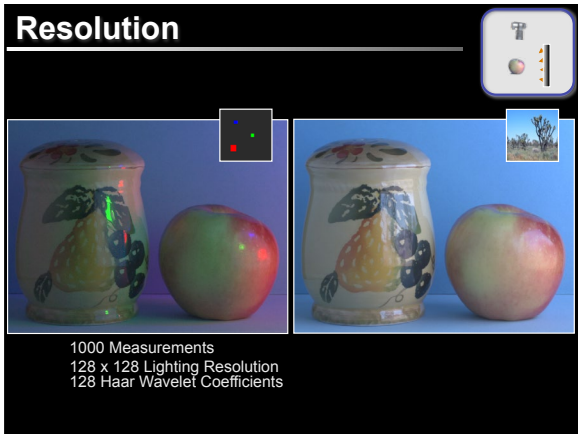

## Compressive Sensing: A Brief Introduction

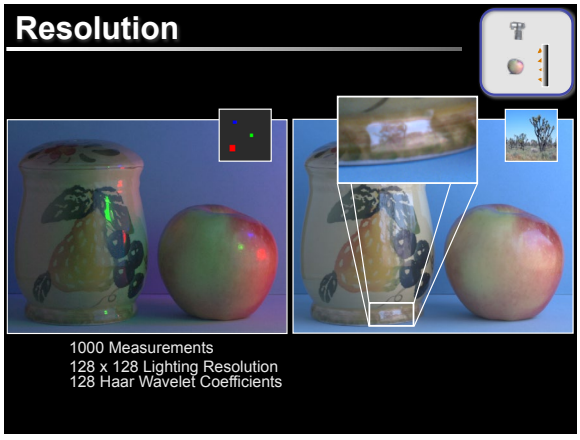
[Candes et al., 06][Donoho, 06]…

- Sparsity / Compressibility:
  - Signals can be represented as a few non-zero coefficients in an appropriately-chosen basis, e.g., wavelet, gradient, PCA.
- For sparse signals, acquire measurements (condensed representations of the signals) with random projections.

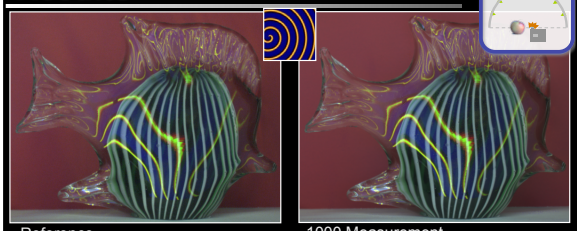
$$A \quad x = b$$

Measurement Ensemble $m \times n$, where $m < n$ | Signal $n \times 1$ | Measurements $m \times 1$



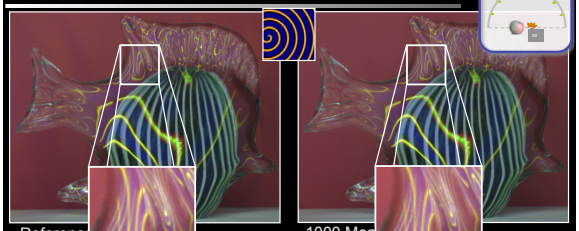
## Results



Reference

1000 Measurement
128 x 128 Lighting Resolution
128 Haar Wavelet Coefficients

## Results



Reference

1000 Mea
128 x 128
128 Haar

ution
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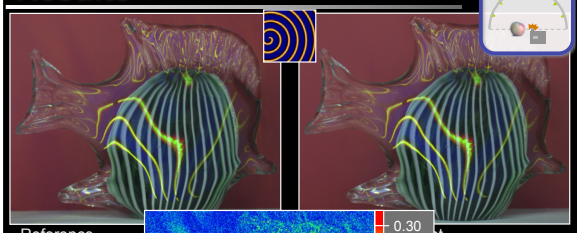
## Results




Reference

Resolution
Coefficients

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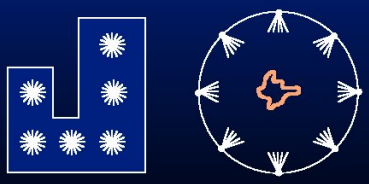
## Light Field Rendering

*Marc Levoy* *Pat Hanrahan*

Computer Science Department
Stanford University

## Apple's QuickTime VR



Outward

Inward

## Generating New Views

Problem: fixed vantage point/center

One Solution: view interpolation

- Interpolating between range images (Chen and Willams, 1993)
- Correspondences and epipolar analysis (McMillan and Bishop, 1995)

-> Requires depths or correspondences:

must be extracted from acquired imagery

relatively expensive and error-prone morph

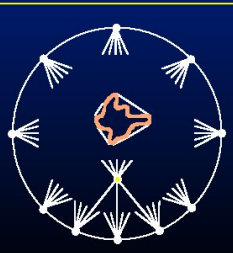
## Light Fields

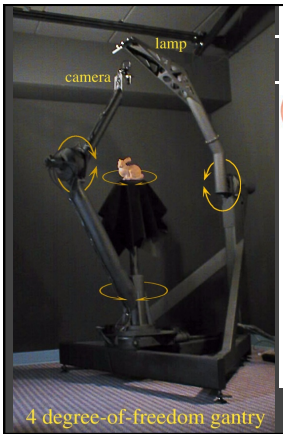
Gershun's and Moon's idea of a light field:

*Radiance as a function of a ray or line: L ( x, y, z, θ, ϕ )*

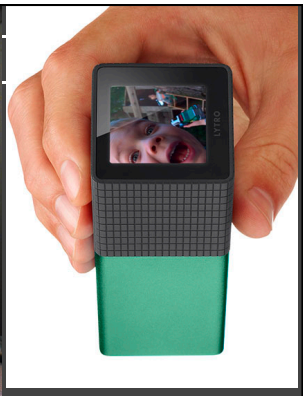
- In "free space" (no occluders) 5D reduces to 4D
  - Exterior of the convex hull of an object
  - Interior of an environment
- Images are 2D slices
  - Insert acquired imagery
  - Extract image from a given viewpoint

## 4D Light Field

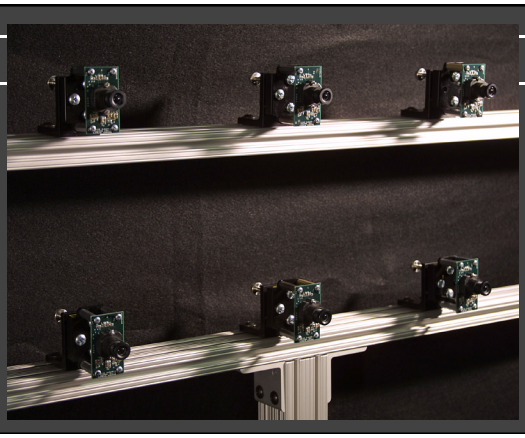





4 degree-of-freedom gantry

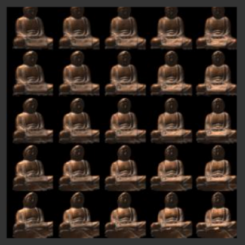


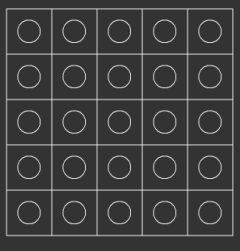
Lytro Camera



## Light Field as a 2D Array of Image

Camera Array



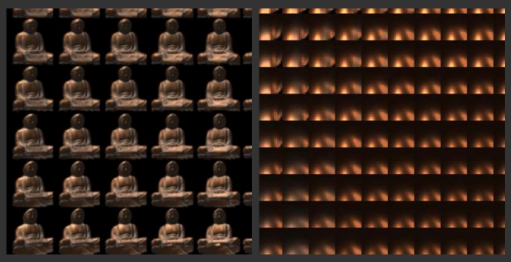


$$L(r) = L(u, v, s, t)$$

## Dual Interpretation of Light Field

Plenoptic Light Field
Field radiance

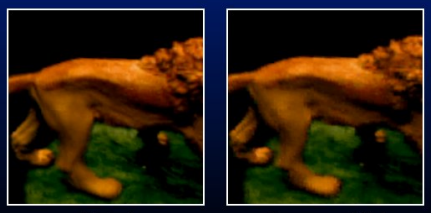
Surface Light Field
Surface radiance



UV Array of ST Images

ST Array of UV Images

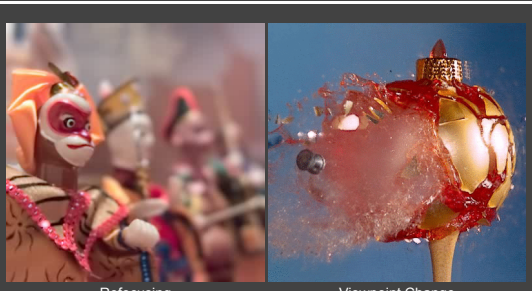
## Compression Example



Original

Compressed 120:1

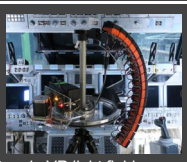
## Light fields



Refocusing

Viewpoint Change

## Recent Light Field Cameras



Google VR light field camera

Pelican

Light

RayTrix

Lytro Illum

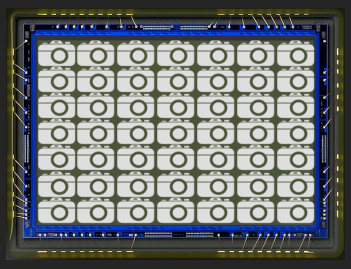
## Recent Light Field Cameras

## Resolution trade-off

Limited resolution

High angular

Low spatial



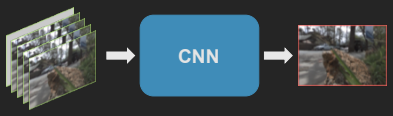
UC San Diego Kalantari et al.

## Solution: angular super-resolution

UC San Diego Kalantari et al.

## Straightforward solution

- Model the process with a single CNN

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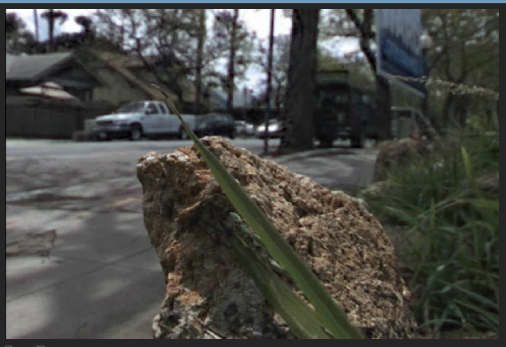
## Single CNN's result



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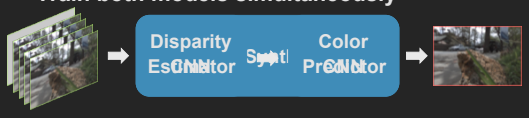
## High-level idea

- Follow the pipeline of existing techniques and break the process into two components
  Goesele et al. [2010]; Chaurasia et al. [2013]
  - Disparity estimator
  - Color predictor
- Model the components using learning
- Train both models simultaneously

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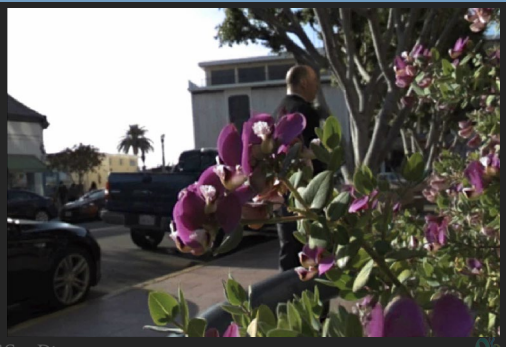
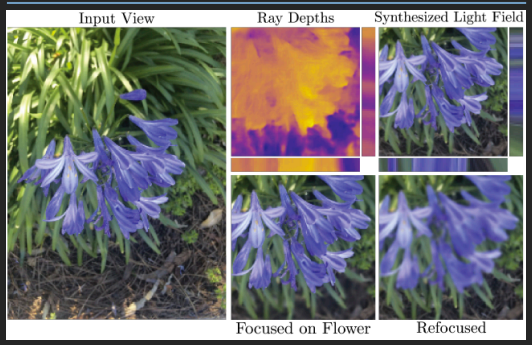
## Our result



UC San Diego Kalantari et al.

## 4D RGBD Light Fields from 2D Image




UC San Diego Srinivasan et al. ICCV 17

## Light field video

- Consumer light field cameras limited bandwidth
- Capture low frame rate videos

Lytro Illum (3 fps video)

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Wang et al. SIGGRAPH 17

## Lytro video

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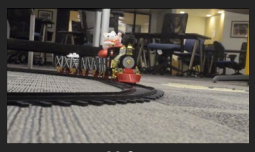
## Hybrid Light Field Video System

DSLR

Lytro



30 fps

3 fps

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## Our result

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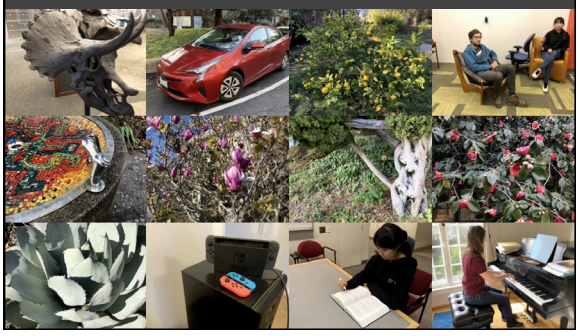
## Applications: Sampling/Reconstruction

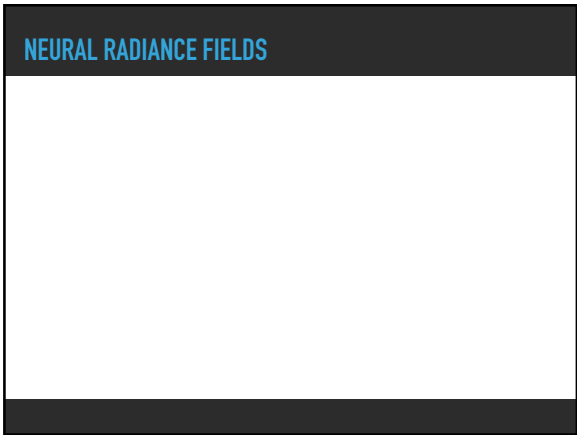
- Monte Carlo Rendering
- Light Transport Acquisition
- Light Fields and Computational Photography
- *View Synthesis*
- Animation/Simulation (not covered in course)

- Brief overview of these applications today, and opportunities/history for sampling/reconstruction

## Light Fields with 4000x fewer views



## NEURAL RADIANCE FIELDS



## Summary

- Brief overview of applications, some algorithms
- Will cover in greater detail in rest of course
- Biggest practical progress in Monte Carlo rendering: order of magnitude speedups
- Widely used in production, also in real-time
- Very relevant in light transport acquisition
- Sampling/Reconstruction key for light fields
- View Synthesis other major focus, huge explosion
- Many other applications: PRT, Animation, etc.