Computational Imaging from Nanoscopic to Astronomical Scales





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A Centuries Old Design

Camera Obscura



http://www.camerapixela.net

Digital Camera



Jean François WITZ

Both cameras produce perspective images



Traditional vs. Computational Imaging





A Generalized Camera Model





A Generalized Camera Model



Incoherent Imaging (e.g. sun, incandescent, LEDs)

 $y \in \mathbb{R}^M_+$

 $x \in \mathbb{R}^N_+$

 $A \in \mathbb{R}^{M \times N}_+$

Coded Coding Matrix Noise $y = A \cdot x + n$

Optical Coding Equation



A Generalized Camera Model



Coherent Imaging (e.g. LASER illumination)

 $\sum_{\text{Image}}^{\text{Coded}} \sum_{\text{Matrix}}^{\text{Coding}} \text{Noise}$ $y = |A \cdot x|^2 + n$

Optical Coding Equation

 $y \in \mathbb{R}^{M}_{+}$

 $x \in \mathbb{C}^N$

 $A \in \mathbb{C}^{M \times N}$



Decoding and Image Priors

Assume we have a PDF for images, e.g.

$$P(x) = \exp\left(\left\|B \cdot x\right\|^{\alpha}\right)$$



Other priors •Total Variation (TV) •Wavelet/sparsity prior •Learned priors (K-SVD,DNN)

Compute the Maximum A Posteriori (MAP) estimate

$$x^{*} = \underset{\mathcal{X}}{\operatorname{argmax}} \left(\left\| \underbrace{y - H \cdot x}_{\text{Data term}} \right\|^{2} + \left\| \underbrace{B \cdot x}_{\text{Prior term}} \right\|^{\alpha} \right)$$



Previous Work: CI Performance



Long Exposure





Previous Work: CI Performance



When does computational imaging improve performance?



What are the limits of CI Performance?

Stopped Down Camera: F/11, Focal Sweep: F/1 q=.5, R=.5, t=6ms, p=1um, $\sigma_r=4e^-$



[Mitra et al., PAMI '13]



In this Talk: CI Across Scale





Research Timeline

Talk Outline





Talk Outline





Goal: Long Distance Imaging





Diffraction Limits Resolution

- Light diffracts at the edge of the aperture instead of focusing to a point
- Diffraction is represented by an Airy disk, radius (m)

: wavelength of light (m): distance to object (mm): aperture diameter (mm)





Diffraction Blur For Different Cameras





Diffraction Blur For Different Cameras





Cost Of Using Larger Lenses





Remote Fourier Ptychography Imaging





Image Formation Model



¹Goodman, Joseph W. Introduction to Fourier optics. 2005.



Image Formation Model





Image Formation Model

- Lens forms the Fourier transform at the sensor plane
 \$\mathcal{F}\${A(\hat{\psi}(x',y'))}\$
- Sensor captures the squared magnitude

 $I(x', y') \propto \left| \mathcal{F} \{ A(\hat{\psi}(x', y')) \} \right|^2$



Improving Resolution With FP

- Individual images are low resolution, combine multiple bandpass regions to improve resolution
- Move aperture in the Fourier domain to synthetically increase aperture size







Sampling the Fourier domain









Gerchberg-Saxton Phase Retrieval

- high-resolution Fourier field at the aperture plane
- bandpass operator
- complex field at the camera sensor
- measured image
- index of aperture positions

Goal: Recover



Gerchberg-Saxton Phase Retrieval





Effect Of Overlap On Reconstruction



¹Bodmann, Bernhard G., and Nathaniel Hammen. "Stable phase retrieval with low-redundancy frames." *Advances in computational mathematics* 41.2 (2015): 317-331.



Effect Of Overlap On Reconstruction



Experimental Setup For Macroscopic FP

- Camera: Blackfly Camera from Point Grey, 2.2 μ m pixels
- Fujinon Lens: 75 mm focal length, 2.3 mm aperture (f/32)
- Laser: Helium Neon laser from Thorlabs, $\lambda = 633$ nm
- Diffraction: Spot size on sensor = $49 \,\mu\text{m} \sim 20$ pixels





Experimental Setup For Macroscopic FP



Not to scale



USAF Target





Reflection Mode Imaging Geometry





Rough Surfaces and Random Phase











Rough Surfaces and Random Phase













Captured Images for a Diffuse Object

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Experimental Setup





Experimental Setup





Diffuse USAF Resolution Target

Example captured image





Diffuse USAF Resolution Target

SAVI reconstruction





Resolution Improvement with SAVI





















Future Work

- Capturing long distance FP data in a snapshot
- Camera array to simultaneously acquire images
- Multiplexed lasers to fill in gaps in Fourier domain







Talk Outline









Different Methods for 3D capture

- Passive stereo, DFF/DFD
- Active Scanning, Structured Light (SL), ToF





Tradeoffs in SL





Optimal SL System





Motion Contrast Principle

- Traditional photo sensor continually outputs values
- Motion contrast sensor measures temporal changes





Motion Contrast Capture

- Same bandwidth for Video and motion contrast
- Video frames are dense, temporal resolution is low
- Motion contrast stream is sparse in space and time





MC3D Principle





MC3D Advantage: Bandwidth



Requires only one measurement per pixel



Results: Ambient Illumination



Second Generation MC3D works with 50,000lux



Live Outdoor 3D Scanning





Live Outdoor 3D Scanning

Kinect2

MC3D (Zoom Lens)

IR Image



MC3D works with 80,000 lux at 4m stand-off distance



Talk Outline





X-ray Ptychography at ANL





X-ray Ptychography at ANL



nology. um. (b) scaling g 22nm





X-Ray Ptychography



RAVEN / PRISMA Goals:

- Non-invasive IC chip imaging.
- 10nm resolution over 1cm x 1cm FoV.
- 3D reconstruction of 10-15 layers.
- Capture + Reconstruction within 25 days.



GDSII Design File





Big Picture





Connection 1: Coding





Connections I: Coding





Connections 2: Super-resolution





Connections 2: Super-resolution





Connections 3: Self-calibration





Connections 3: Self-calibration

Uncalibrated X-Ray Tomography Unknown Projection Angles:



Calibrated Projection Angles



Cultural Heritage Imaging

Uncalibrated Photometric Stereo:



Gauguin Surface Measurement:

3D Surface Reconstruction



Drawing Reconstruction





Connections 4: Scattering





Connections 4: Scattering



All Photon Imaging ToF Imaging through Fog/Rain:



Model 3D STPSF Response:





Summary

Advantages of Computational Imaging

- Reduce hardware complexity
- Photon efficient imaging
- Introduce new functionality

Many common problems across large range of physical scales





SAVI

MC3D

X-ray CD

Aknowledgements











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