Computational Photography

Independent Study Presentation 2:
Generalized Optics for Computational Photography

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Overview of the Presentation

Introduction

Animal Eyes

Generalised Optics

Light Field

Understanding Light field

Light Field Parameterization

Optical Element Formulation

Digital Camera Formulation

Optical Coding Approaches

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Object Side Coding

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Wavefront Coding

Sensor Side Coding

Basic Concepts of Sensor Side Coding

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Coding In-front of the Sensor Plane

Conclusion

Summary of Research using Generalized Optics

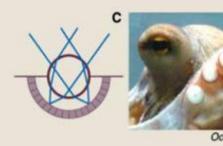
References

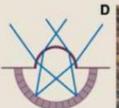
Introduction

Chambered eyes

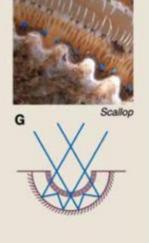










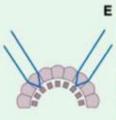


Compound eyes



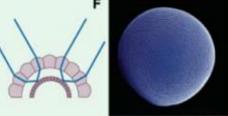


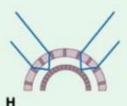
a tan





Dragonfly







Lobster

Animal Eyes

Only 8 types of eyes in whole nature!!!

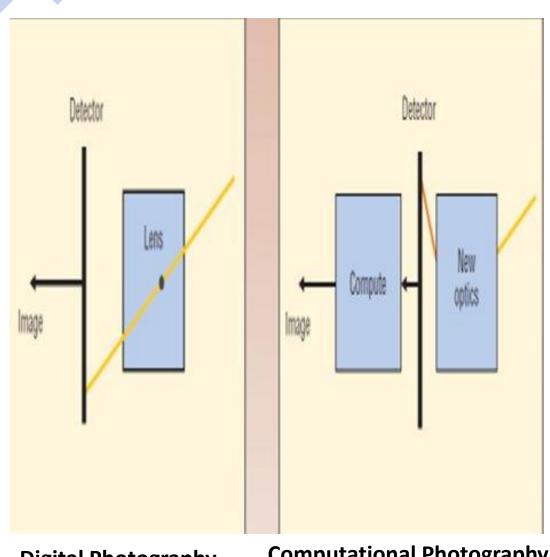
In 33 Animal Phyla-

- 1/3rd have no organs to detect eyes
- 1/3rd have light sensitive organs
- Only 1/3rd have EYES!!!

3 categories of eyes:

- Light detecting spots
- Chambered eyes
- Compound eyes

Generalized Optics



Digital Photography

Computational Photography

Rays are the fundamental Element: Traces all possible ways energy can be transferred by rays from scene to sensor to eyes

Generalized Optics: Manipulates incoming 4D Light Field into new 4D Light Field going to the sensors

Computational **Imaging**

Generalized Sensors: **Un-Conventional** sensors to capture new scene properties not possible to be captured by traditional sensors

Computational Illumination: Replacing the simple flash with customized light to discover the unseen scene elements

Generalized Processing and reconstruction: Application of ML& CV methods to get the desired results and express image semantics

Light Field

Understanding Light Field

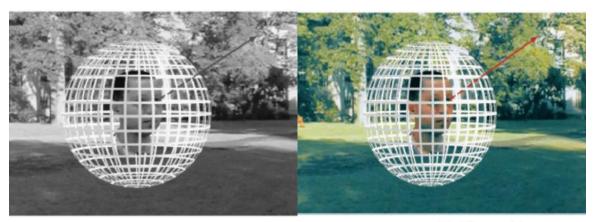
Complete Light Field is represented by:

$$L = P(x, y, z, \theta, \phi, t, \lambda)$$

- a position in 3D space (3D)
- □ in a certain direction (2D)
- at a particular point in time (1D)
- in a particular wavelength (1D)

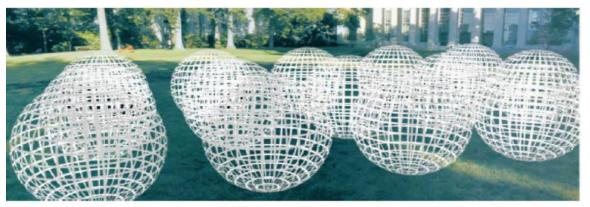
Space with obstacles Light Field: 5D Space without obstacle Light Field: 4D

$$L = P(x, y, \frac{2}{2}, \theta, \phi, \frac{1}{2}, \lambda)$$



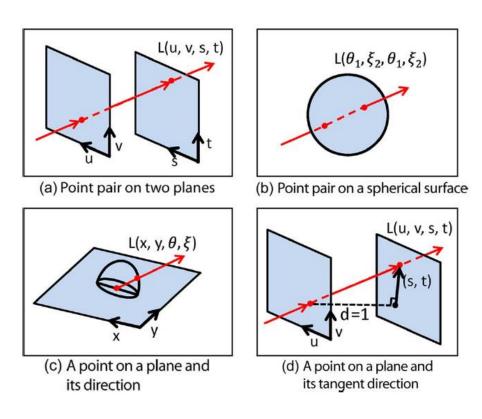
 $P(\theta, \varphi)$

 $P(\theta, \varphi, \lambda)$

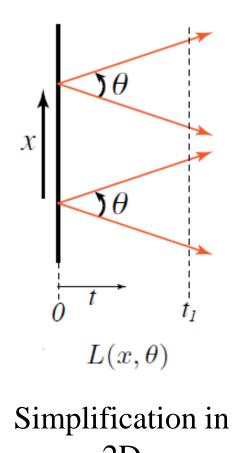


 $P(\theta, \varphi, \lambda, t, x, y, z)$

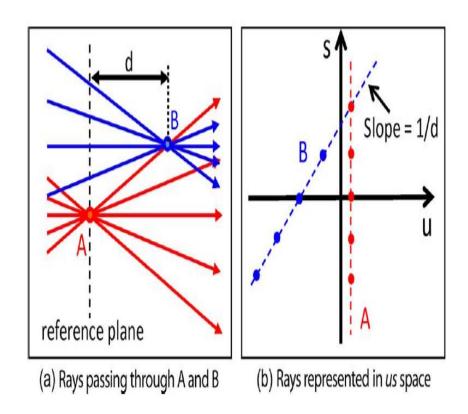
Light Field Parameterization



4 ways of Light Field Parameterization

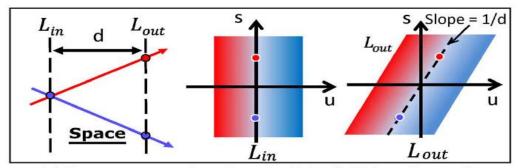


2D

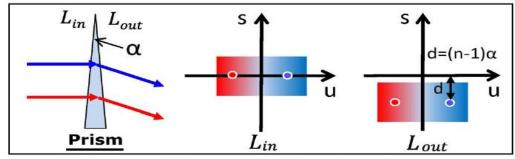


Ray Representation in u-s space

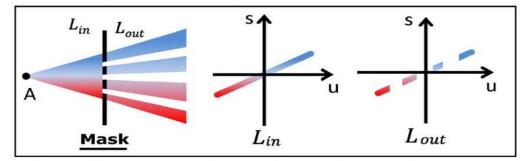
Optical Element Formulation



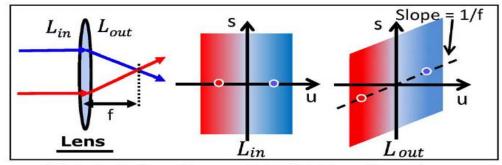
(a) Space propagation shears a light field in u dimension



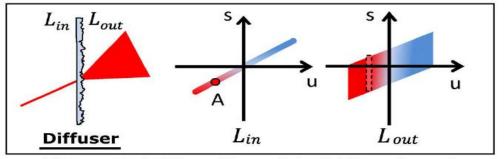
(c) A prism shifts a light field in s dimension



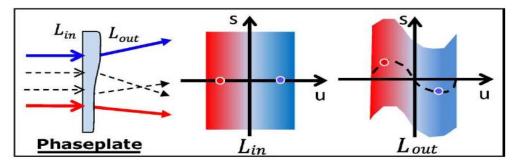
(e) A photomask performs a dot-product in u dimension



(b) A single lens shears a light field in s dimension

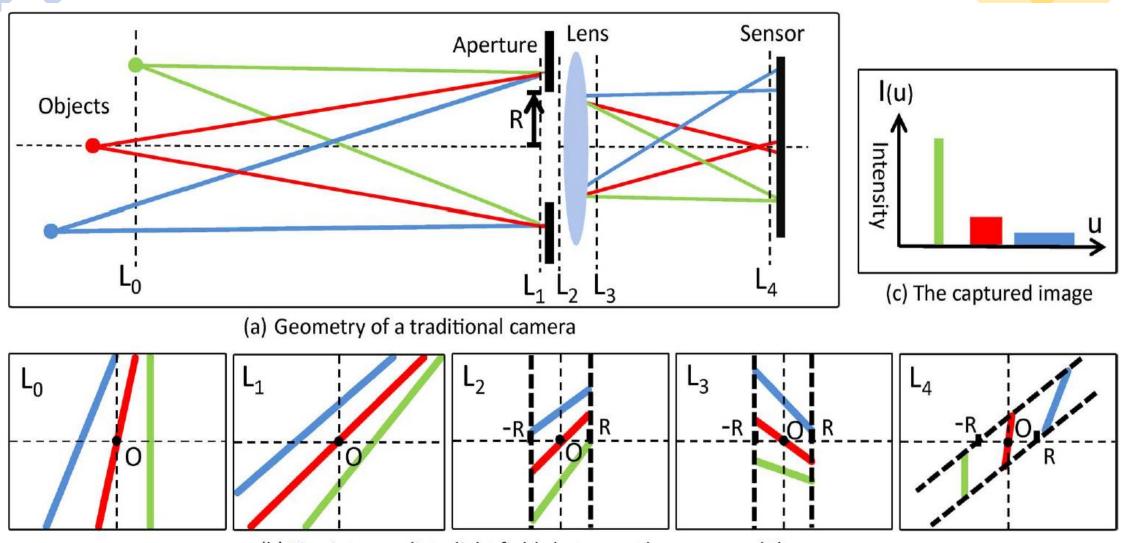


(d) An optical diffuser blurs a light field in s dimension



(f) A phase plate distorts a light field in s dimension

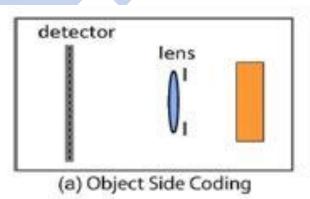
Digital Camera Formulation

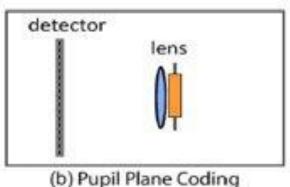


(b) Five intermediate light fields between the scene and the sensor

Optical Coding Approaches

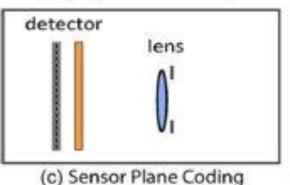
Optical Coding Approaches

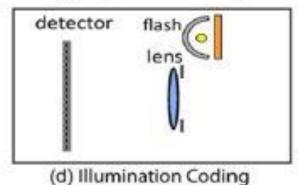




Object side coding: An optical element is attached externally to the conventional Lens.

Pupil plane coding: An optical element is placed at, or close to, the aperture of the lens.



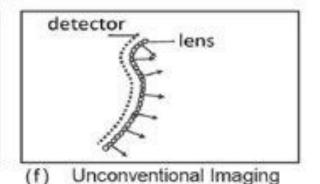


Sensor side coding: An optical element is placed behind the lens.

Illumination Coding: Imaging systems that program the illumination.

camera

(e) Camera Clusters

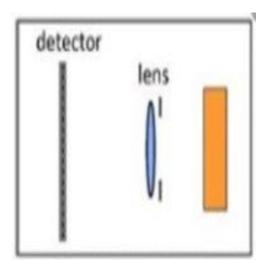


Camera

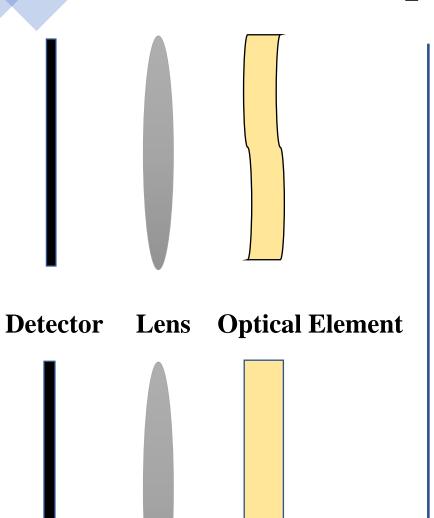
Clusters: Imaging systems that are made up of a cluster or array of traditional camera modules.

Unconventional
Imaging: Imaging
systems using
unconventional camera
architectures
or nonoptical devices.

Object Side Coding



Basic Concepts of Object Side Coding



When Optical Element is Non-Homogenous

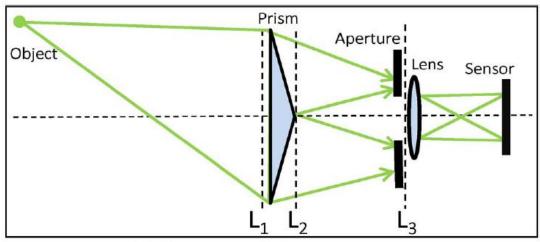
• The cones of light from the same point of object to the optical element will fall on different points on the optical element which are sheared to different locations on the lens. Hence, the modifications become spatially variant with viewing angle in a single shot.

When Optical Element is Homogenous

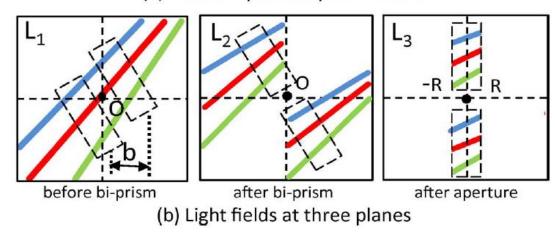
• The Modifications to the Incident Lightfield can be done temporally even though it provides spatially uniform manipulations to all the Lighfields

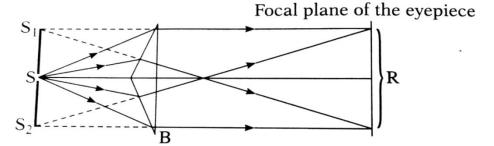
Object Side Coding Examples with Non-Homogenous Optical Element

Depth Estimation



(a) Geometry of a bi-prism camera



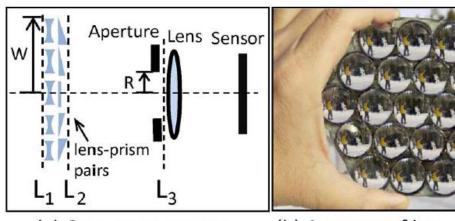


S : Source (a slit illuminated by monochromatic light), B : Biprism, S_1 , S_2 : Virtual sources, R : Region of interference

Bi -Prism

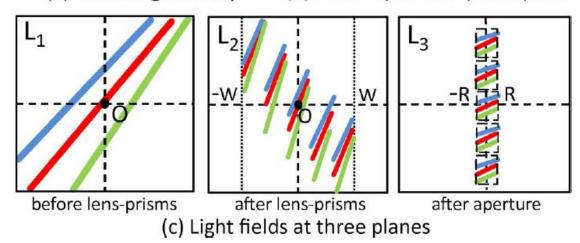
Baseline = $2*tan(\alpha)*d$ $\alpha-->$ Prism Angle D--> distance b/w by-prism and lens

Object Side Coding Examples with Non-Homogenous Optical Element 4D-Light Field Acquisition



(a) Camera geometry

(b) An array of lens-prism pairs



An array of by-prism is kept in-front of the Lens, with each prism having different deviation angle

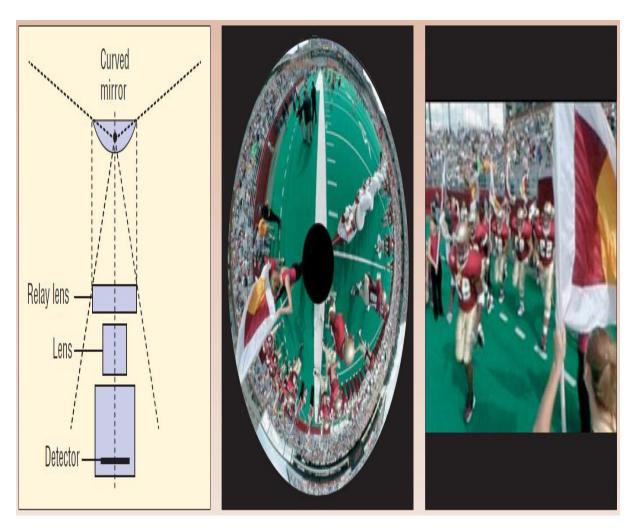
A Concave Lens is kept in-front of prism array to enhance FOV

Multiple copies of the scene are obtained as if viewed from different angle.

Hence, Light field can be aquired from a single image

However, spatial resolution is compromised for the angular resolution

Object Side Coding Examples with Non-Homogenous Optical Element Extending FOV



It's a Catadioptric camera made by combining a lens and a mirror

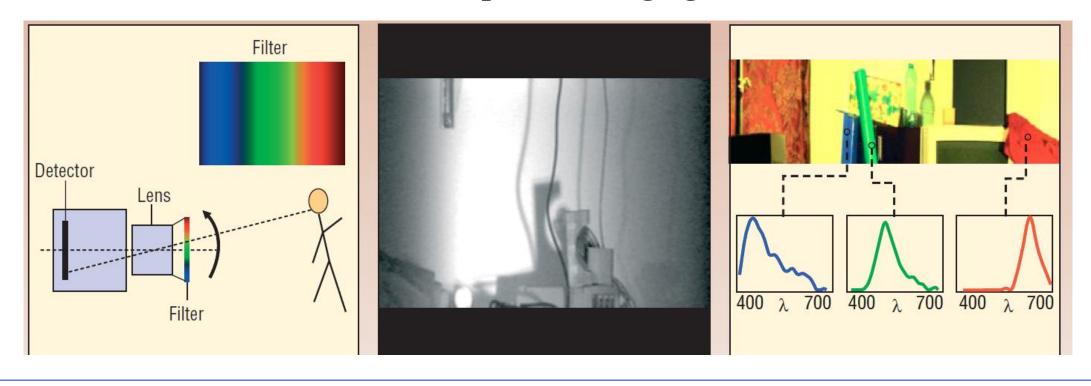
It is designed to ensure that camera captures all the principle rays of the scene pass through the center of projection

FOV in horizontal plane: 360 degree. FOV in vertical plane: 220 degree

The compressed image can be easily decoded to perspective image by a dedicated computational module at video rate

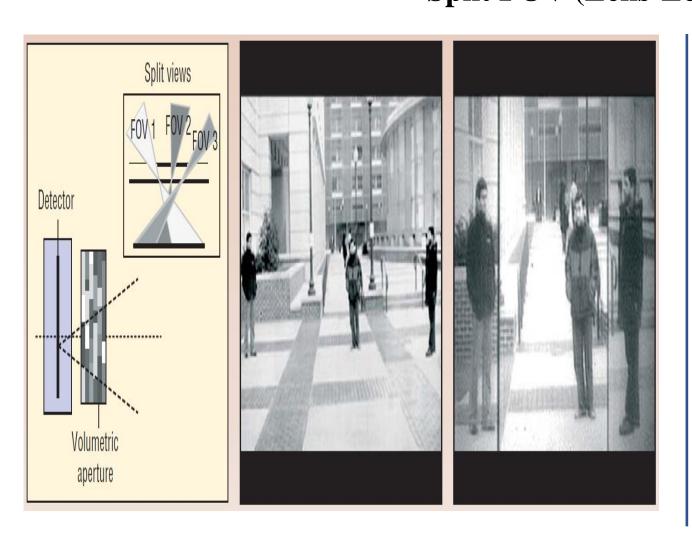
Method doesn't require static scene and only single camera & single shot is required

Object Side Coding Examples with Non-Homogenous Optical Element Multi-Spectral Imaging



Attaching a Linear Interference Filter having spatially varying property of passing different wavelength of visible spectrum, when attached to a black and white video camera it can capture multiple wavelength information of the scene which can be displayed together after interpolation using software.

Object Side Coding Examples with Non-Homogenous Optical Element Split FOV (Lens-Less Imaging)



It is designed by placing 3D(Volumetric Aperture) in front of the detector instead of Lens

Proper arrangement of regions of full transmittance and zero transmittance across the volume can lead disconnected FOV to be mapped on adjacent sensor area

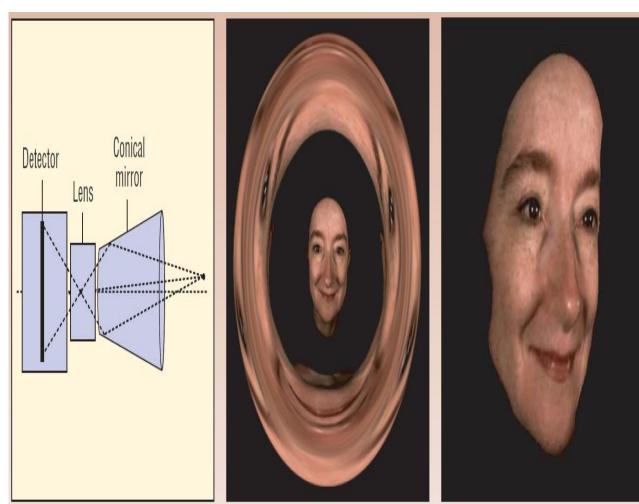
The aperture is implemented as a stack of controllable 2D apertures. Each aperture is a liquid crystal (LC) sheet.

One application is to image the desired FOV in higher resolution than wasting pixels on non-interest scene areas

Further research scope lies in making the transmittance programmable and controllable.

Object Side Coding Examples with Homogenous Optical Element

Multi-View Radial Camera for 3D Imaging



A hollow cone, mirrored inside, is placed in-front of the traditional camera with the axes aligned.

Each scene point is imaged 3 times, once directy and 2 times after getting reflected from 2 point on the conical mirror.

Hence, 3 viewpoints are obtained. One is at the center of the sensor and the other 2 virtual viewpoints are equal and opposite wrt Optical axis on the sensor

A Stereo Matching algo can find coreespondence between the 3 views and can give the depth information.

Object Side Coding Examples with Homogenous Optical Element

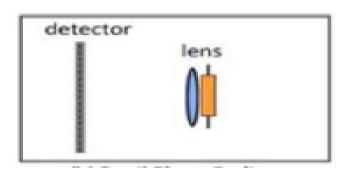
Other Applications

Capturing image with different polarization direction in order to remove specular reflection

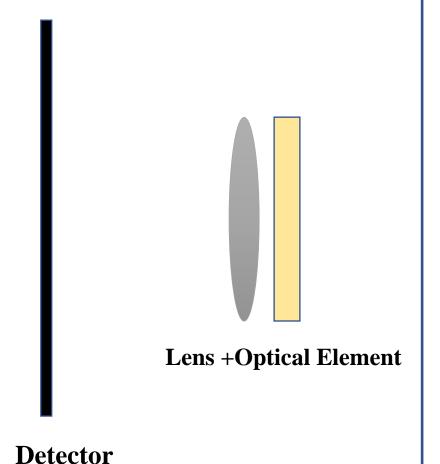
Coding external shutter for coded exposure to remove motion blurring

Placing Optical Diffuser on the object side for depth estimation as it gives larger angle triangulation etc.

Pupil Plane Coding



Basic Concepts of Pupil Plane Coding



Unlike Object Side Coding, Light reaching the Pupil Plane of the Lens is spatially invariant

Coding Done on the Pupil Plane has the aim of modifying the Point Spread Function of the Lens

The Fresnel equation describes the relation between the Pupil Plane Coding and Resulting PSF:

$$f(x) = |\mathcal{F}(W(x) \cdot Q_d(x))|^2$$

f(x) PSF function

 $\mathcal{F}(\cdot)$ Fourier Transform

W(x) Aperture coding function

 $Q_d(x)$ Quadratic Phase Term

Point Spread Function

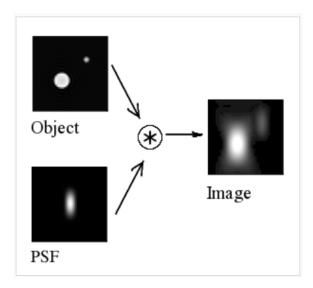
Point Spread Function:

It is a function that describes the response of a focussed imaging system to a point source at a given distance.

Finding Optimum Aperture Size at a given distance:

$$\delta = 2.44\lambda f/\delta$$

$$\delta_{opt} = \sqrt{2.44 \, \lambda f}$$



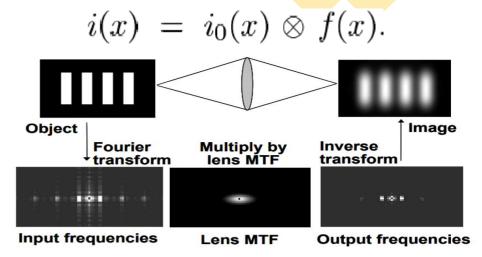
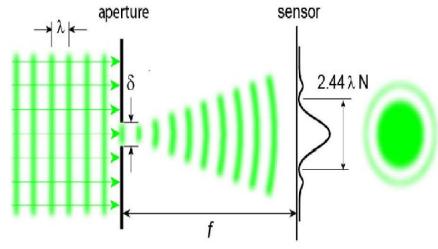
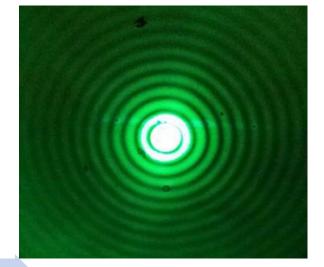


Figure 2.4 Frequency domain filtering in an optical system.





Small Aperture and the diffraction leading to Airy Disk and Diffraction Rings

Basic Concepts of Pupil Plane Coding

Coded Aperture

Involves Pupil Plane
Coding with
Intensity Modulators

Aperture Coding Function is Real

Involves Pupil Plane Coding with Phase Modulators

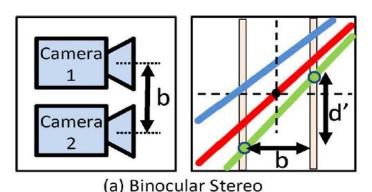
Aperture Coding Function is Complex

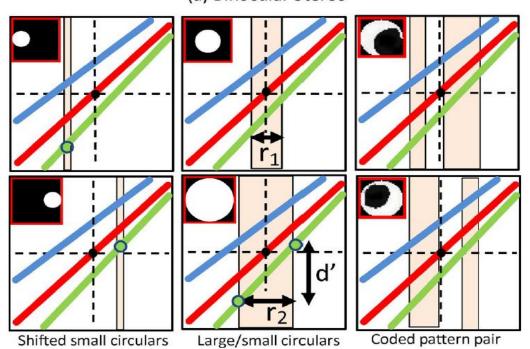
Pupil Plane Coding

Wavefront Coding

Coded Aperture

Depth from Defocus





(b) Depth from Defocus

Here we compare Binocular Stereo with the Coded Aperture Technique for calculating the Scene Depth

In terms of Light Field Representation, Slope of a stripe is inverse of the depth. So both the techniques go for Slope Calculation

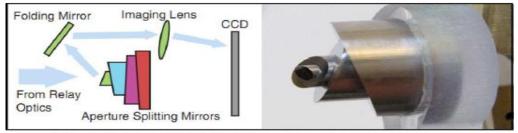
Binocular Stereo establishes the correspondences and calculate (focal length normalized disparity)/(baseline)

DFD using Coded Aperture calculate different integral across each stripe to calculate blur kernel size by DFD algo

Slope for DFD is (focal length normalized blur kernel radius)/(aperture diameter)

Coded Aperture

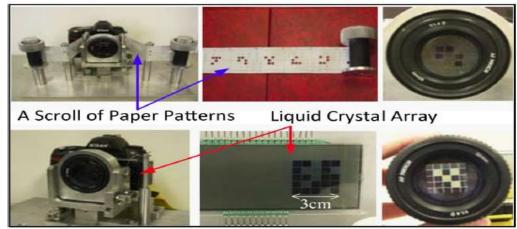
Other Applications



(a) Multi-aperture photography using tilted mirrors



(b) Coded aperture using photomasks



(c) Programmable aperture camera using paper patterns or LCD

Defocus Deblurring

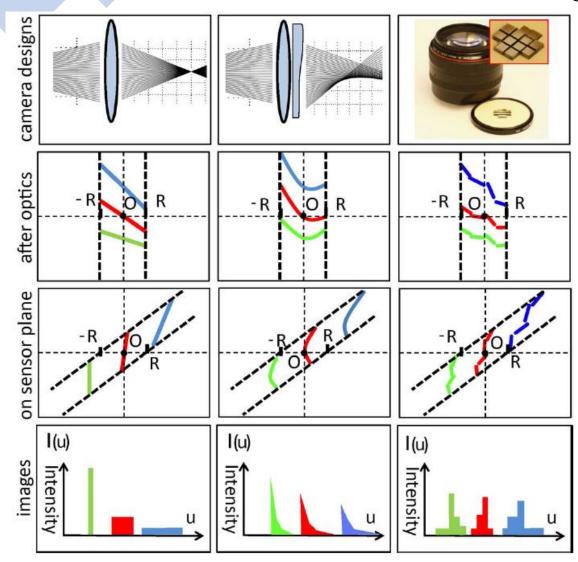
- If Point Spread Function is known, then deconvolving the captured image with PSF gives high quality sharp image.
- Coded Aperture can help in getting broadband MTF with limited and known zero crossing frequencies

Depth Using Coded Aperture for colors

- Color Filter Apertures so that 3 color channels are captured with different apertures
- Depth information can be obtained by processing each channel image seprately

Wavefront Coding

Extending Depth of Field



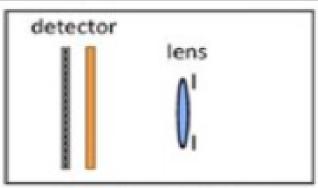
TraditionalCamera Wavefront coding Lattice-focal Lens

A cubic phase plate having Aperture Coding function a third order polynomial is used

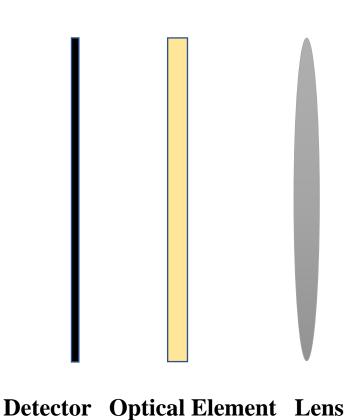
It results in a broadband MTF without zerocrossing frequencies and is depth invariant as cubic polynomial overpowers the quadratic term varying with depth

The Intensity of a ray falling on the sensor plane depends on its alignment with sensor plane. So the phase plate tilts the light field in a way that images are sharp irrespective of the depth

Sensor Side Coding



Basic Concepts of Sensor Side Coding

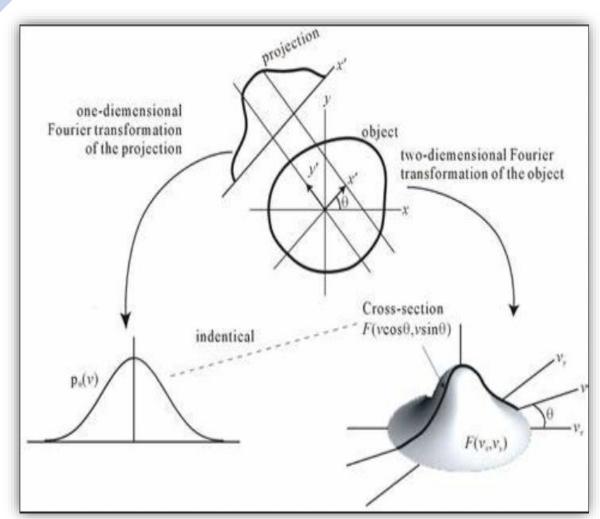


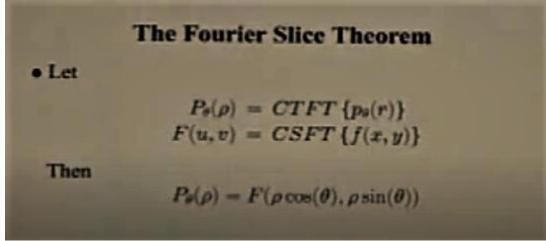
Optical Element kept on the sensor side of the Lens provides non-convetional optics

According to Gauss Law, Optical Devices after the lens are the dual of the optical devices infront of the lens. Hence, sensor side coding can provide same functionalities as object side coding

Advantage of Sensor Side Coding is that it provide compact solutions which are non-intrusive for scene

Fourier Slice Theorem

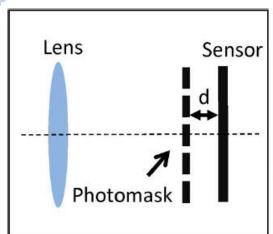




Fourier Slice Theorem Equations

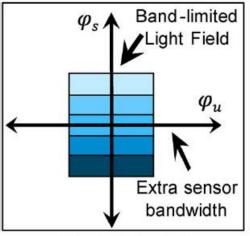


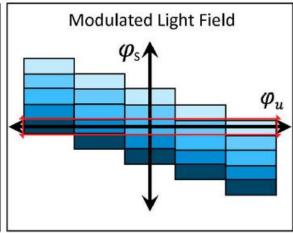
Coding in-front of the Sensor Plane Light Field Acquisition





(a) The geometry and a prototype of heterodyne light field camera





without photomask

with photomask

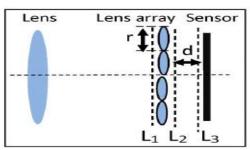
(b) Light field on sensor plane in the Fourier domain

$$I = \mathcal{P}[\mathcal{S}[L \cdot M]] = \mathcal{P}[\mathcal{S}[L] \cdot \mathcal{S}[M]] = \mathcal{P}[L_5 \cdot \mathcal{S}[M]]$$

M is the shifted cosine signal in spatial domain or shifted Dirac Delta function in Fourier Domain which on getting multiplied with L5 gives multiple copies of L5 each tilted with respect to each other in spatial domain. Each tilted copy projects different projection on the sensor which when combined yields 4D Light Field

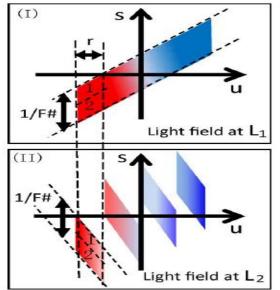
Coding in-front of the Sensor Plane

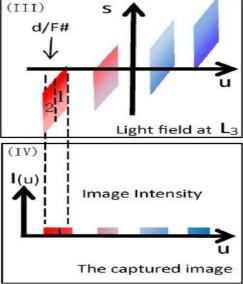
Light Field Acquisition





(a) The geometry and a prototype of plenoptic camera





(b) Light field transform at three planes and the final image

The figure demonstrates how the lens array helps in mapping two field components of different angle on same location to be mapped separately which is not possible with traditional camera.

Conclusion

Summary of Research using Generalized Optics

Devices		Object Side Coding	Pupil Plane Coding	Sensor Side Coding
Phase Modulators	Lens(es)	Lightfield: [52, 11]	Depth: [104]	Lightfield: [113,114,111,115,116,206]
	Prism(s) Plate(s)	Depth: [46,50,208,209] Color: [72]		
	Phaseplate		Depth: [98, 99] EDOF: [100,101,105,106,107,210]	
	Diffuser	Depth: [31] HDR: [77]	EDOF: [32,108]	
Intensity Modulators	Photomask	HDR: [33,70,215] Motion: [76]	Lightfield: [37,38] EDOF: [94,95] Depth: [34,80,81,82,83,84,89,90,217,218] Image: [34,35,36,85,86,87,88,211,212,213]	Lightfield: [35] HDR: [110,121,214]
	Color Filter	Color: [33]	Depth: [45]	Color: [43,116,117,119,120,121]
	Polarizer	Separation: [73,74,75,219,220]		
Others	Motion	EDOF: [132]	Depth: [221]	EDOF: [134] Image: [133,134,135] Motion: [76,112,136]
	Mirror(s)	Depth: [47,48,49,51,71,223,226] FOV: [53,54,55,56,57,58,59,60,64, 65,66,67,68,69,71,222,224,225]		HDR: [41,97,215] FOV: [41]

References

- Computational Photography by Raskar Tumblin
- Computational Cameras: Convergence Of Optics and Processing By Shree
 K. Nayar and Changyin Zhou
- Computational Cameras: Redefining the Image By Shree K. Nayar
- https://ocw.mit.edu/courses/media-arts-and-sciences/mas-531-computational-camera-and-photography-fall-2009/
- https://www.youtube.com/watch?v=IIrPkh0IH14&t=140s
- https://www.youtube.com/watch?v=Koq19e4xVgg&list=PLjRBT0-8H-NHkA7JxZ4b4o5WDgFcbghOd&index=5
- https://www.youtube.com/watch?v=Tkc GOCjx7E
- www.computationalphotography.org

Thank You!!!