

Computational Photography

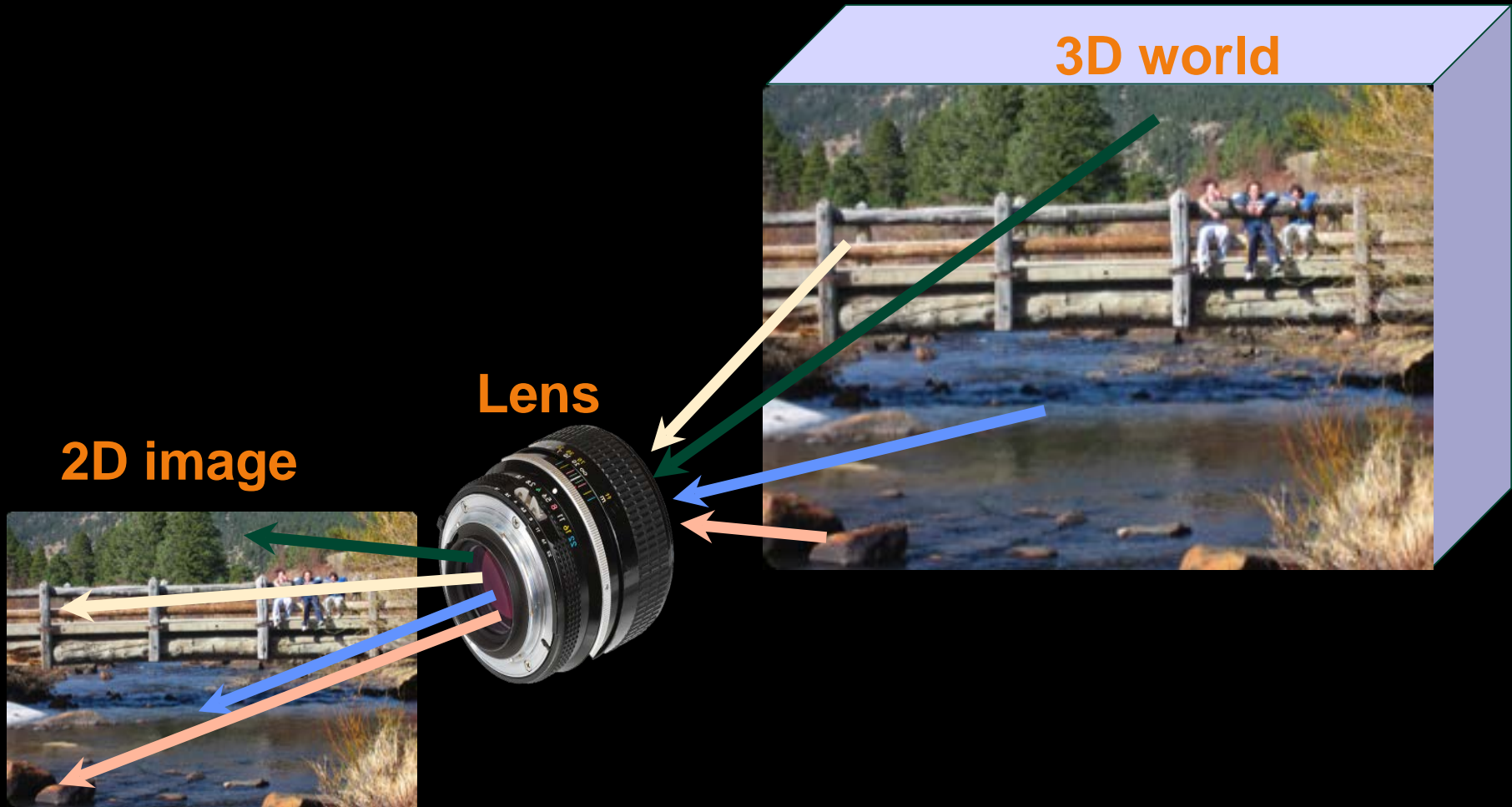
Introduction

Anat Levin

Department of Electrical Engineering
Technion

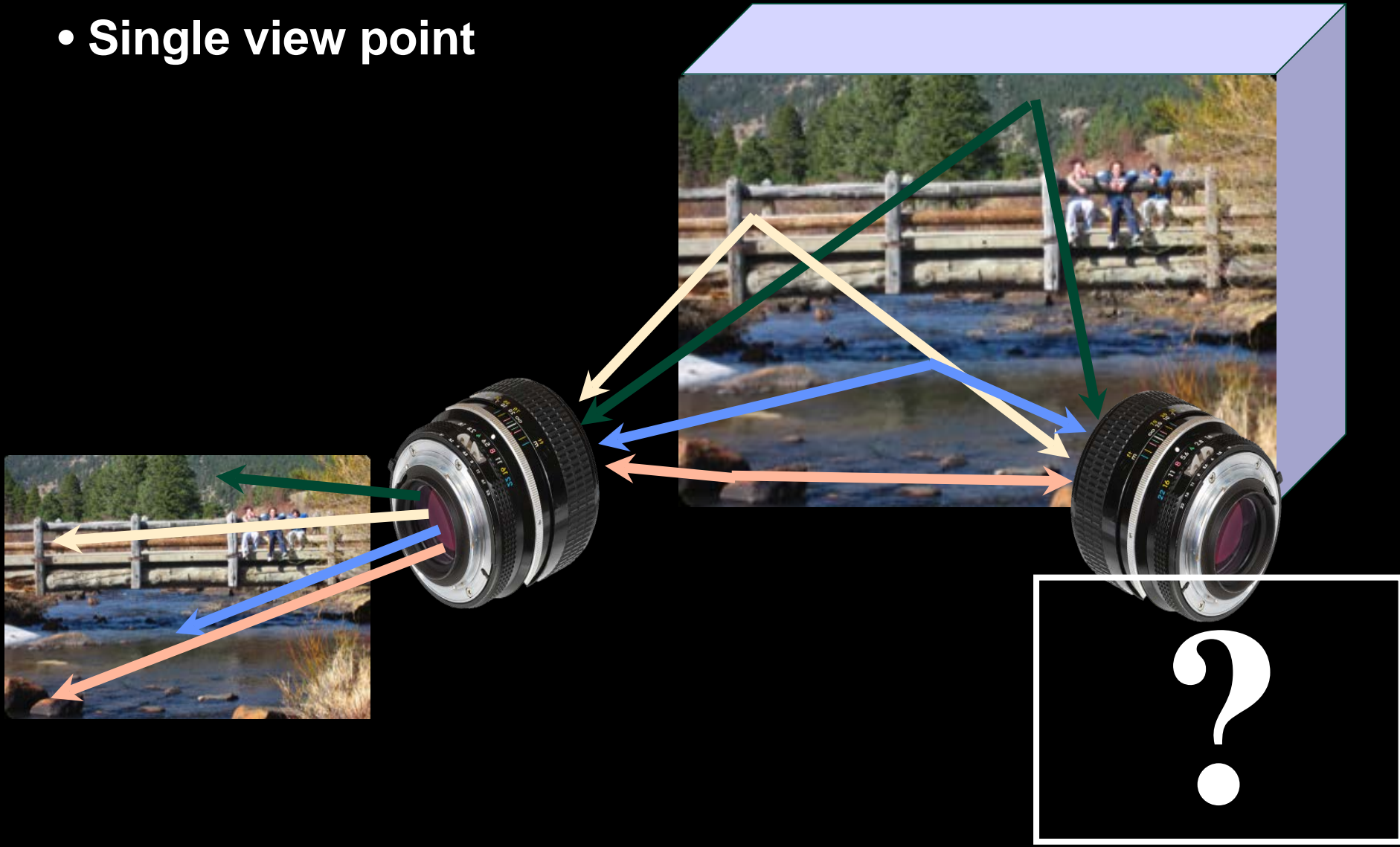
Traditional imaging limitations

- 3D world \rightarrow 2D image. Loss of depth information



Traditional imaging limitations

- 3D world \rightarrow 2D image. Loss of depth information
- Single view point



Traditional imaging limitations

- 3D world -> 2D image. Loss of depth information
- Single view point
- Limited depth of field



Traditional imaging limitations

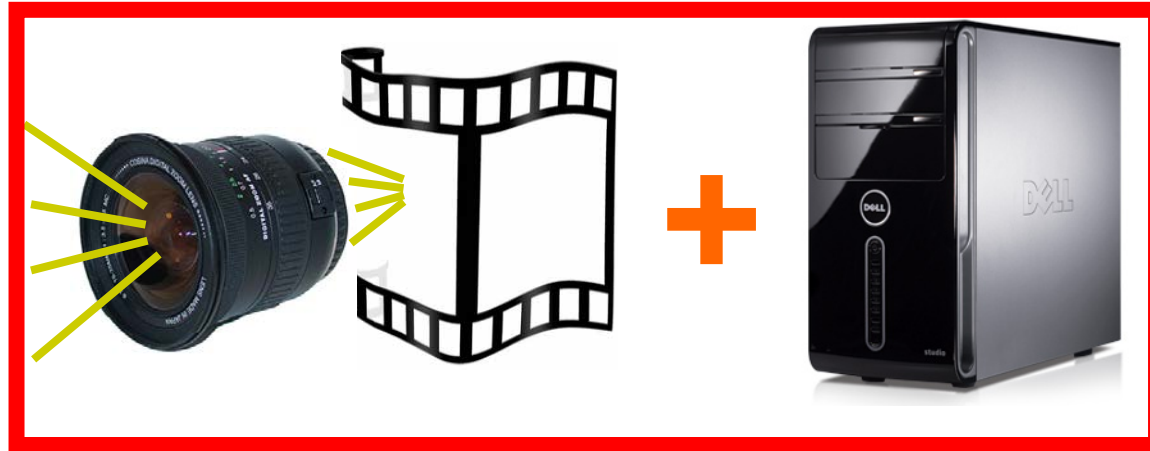
- 3D world -> 2D image. Loss of depth information
- Single view point
- Limited depth of field
- Motion blur
- Limited spatial resolution
- Noise
-
-
-

Traditional Photography



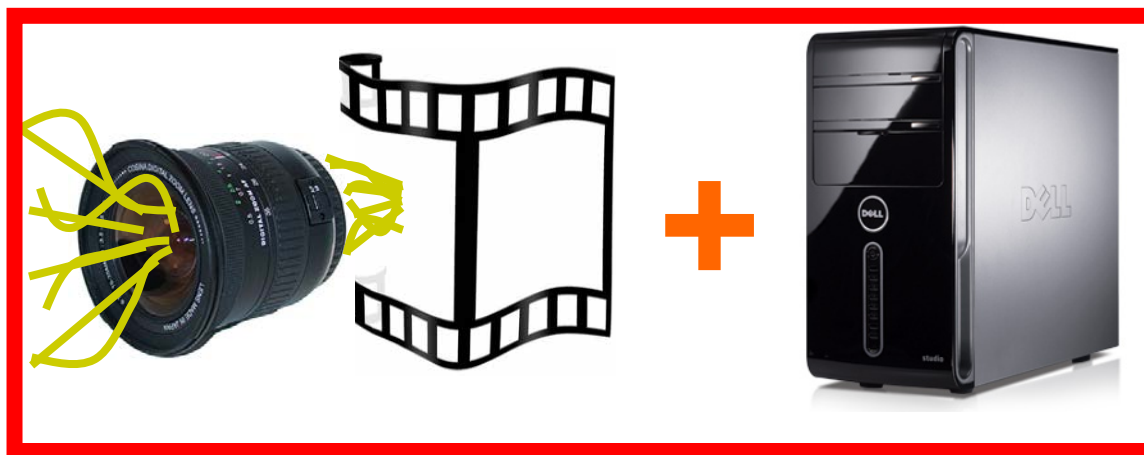
Optical Camera

Computational Photography



Computational Camera

Computational Photography



Computational Camera

Redesign optics to
account for computation

**Compact
camera**



**Video
camera**



Telescope



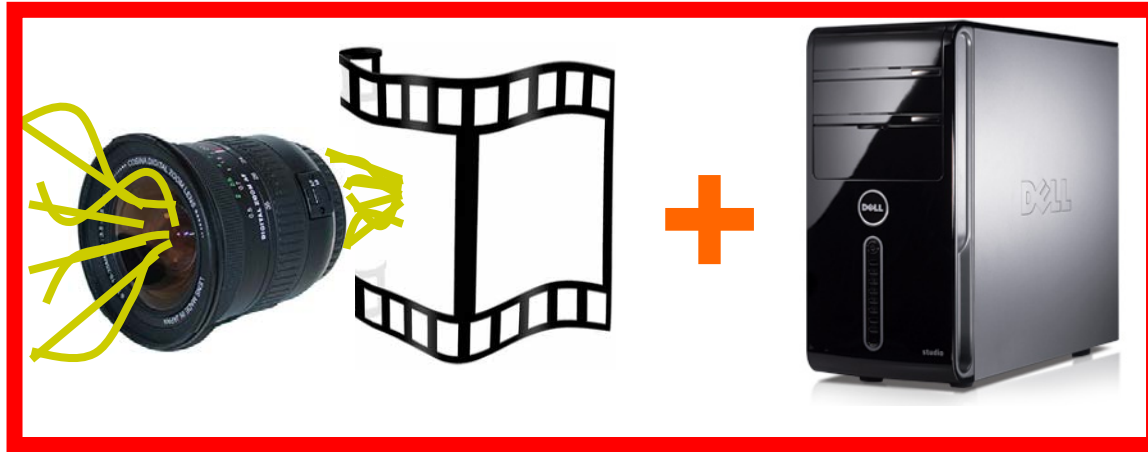
Microscope



**Medical Imaging
CT / MRI**



Computational Photography



Computational Camera

**Redesign optics to
account for computation**

Goal:

- Break bounds on traditional optics using computation
- Develop unified mathematical theory for computational cameras

Examples of computational imaging systems

Overcoming motion blur

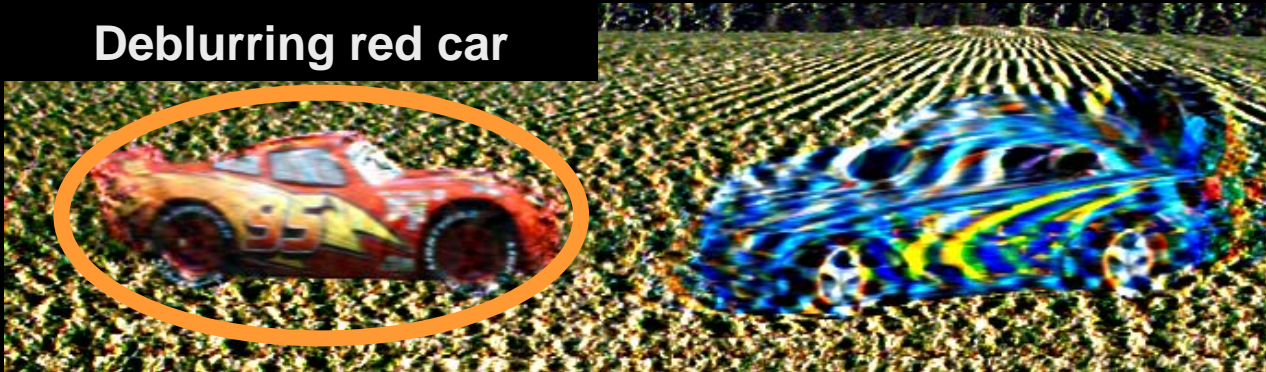
.....
Levin et al. Motion Invariant Photography SIGGRAPH, 2008.



Overcoming motion blur

Levin et al. **Motion Invariant Photography** SIGGRAPH, 2008.

Deblurring red car



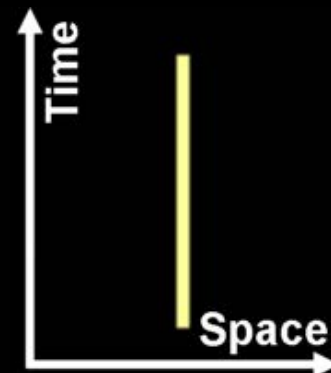
Removing motion blur is hard:

- Need to know exact motion velocity (blur kernel)
- Need to segment image

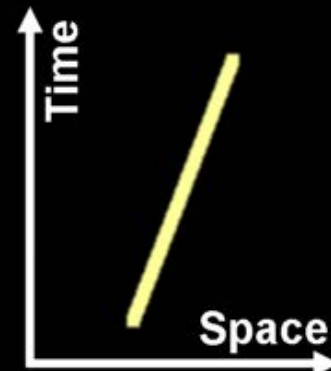
Overcoming motion blur

Levin et al. **Motion Invariant Photography** SIGGRAPH, 2008.

Static- recorded image



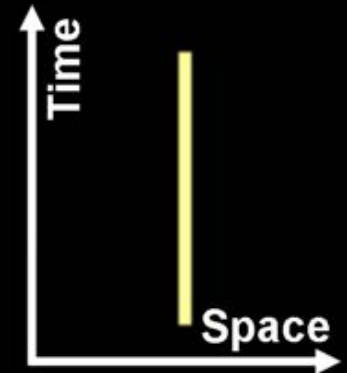
Tracking- recorded image



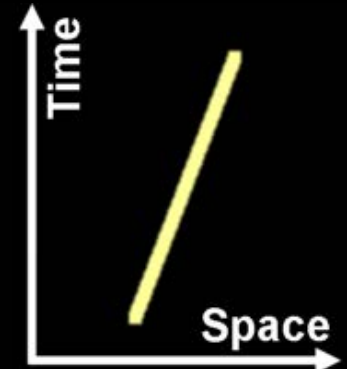
Overcoming motion blur

Levin et al. **Motion Invariant Photography** SIGGRAPH, 2008.

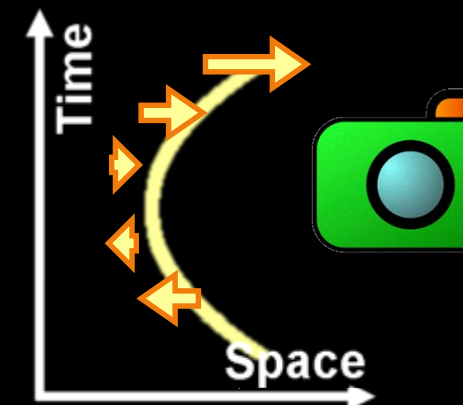
Static- recorded image



Tracking- recorded image



Motion invariant blur



Motion Invariant Photography

Levin et al. Motion Invariant Photography SIGGRAPH, 2008.

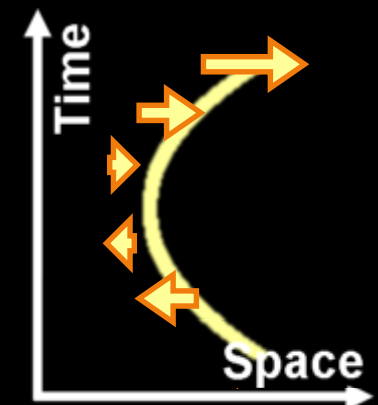
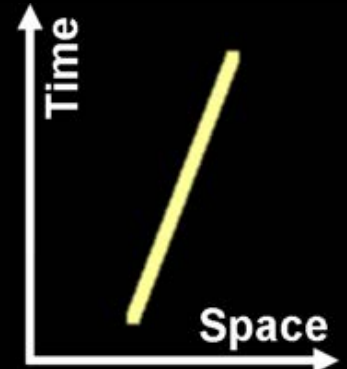
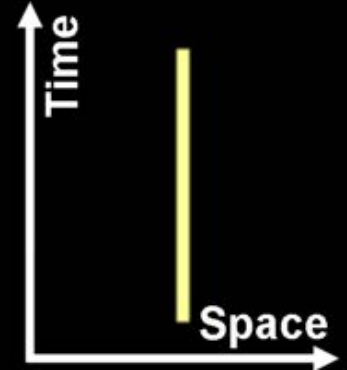
Static- recorded image



Tracking- recorded image

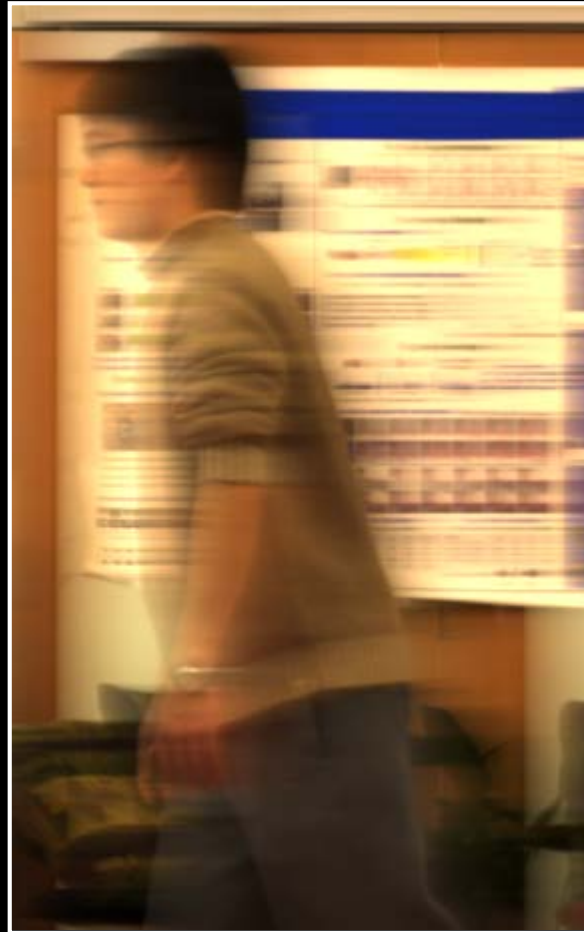


Motion invariant deblurring



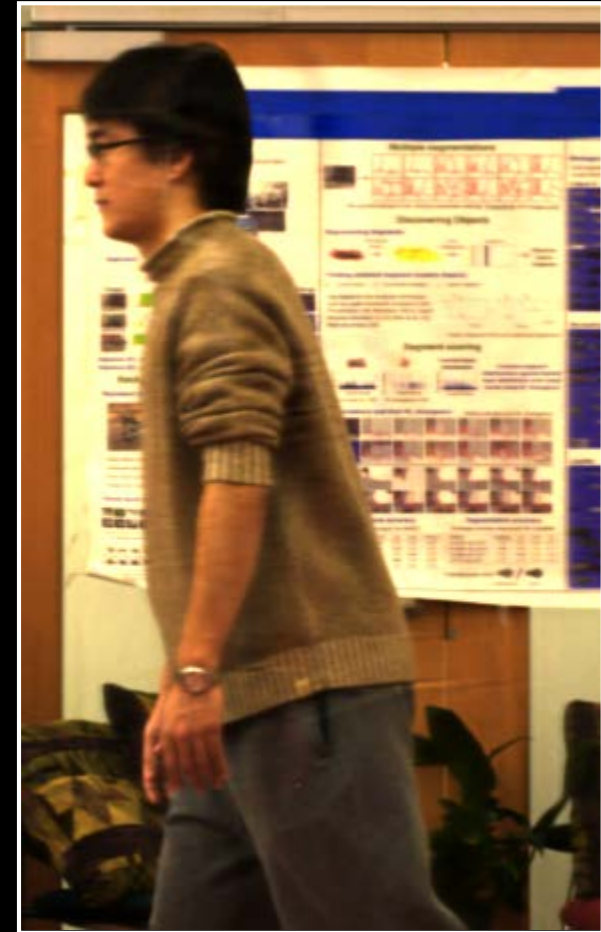


Static camera
**Unknown and
variable blur**

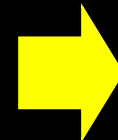


Our parabolic input

**Blur invariant to
velocity**



**Our output after
deblurring**



***NON-BLIND
deblurring***

Defocus blur

Lens and defocus

Lens' aperture

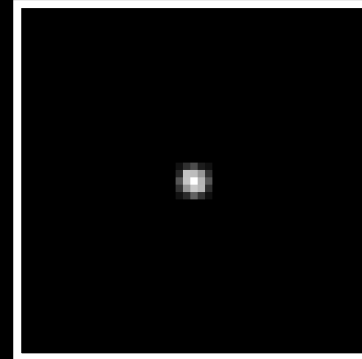
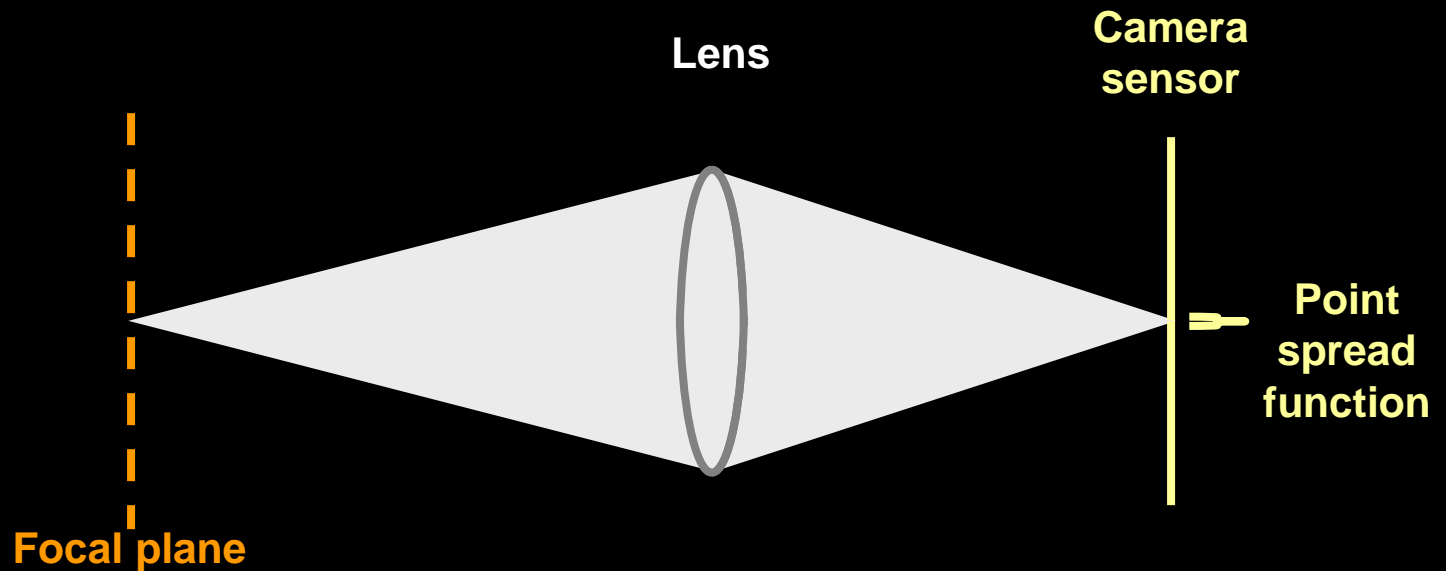


Image of a point light source



Lens and defocus

Lens' aperture

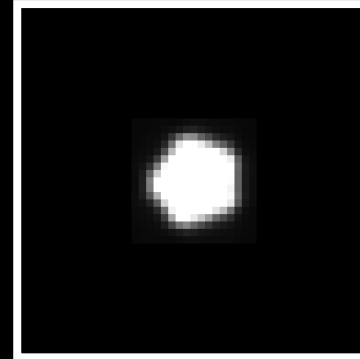
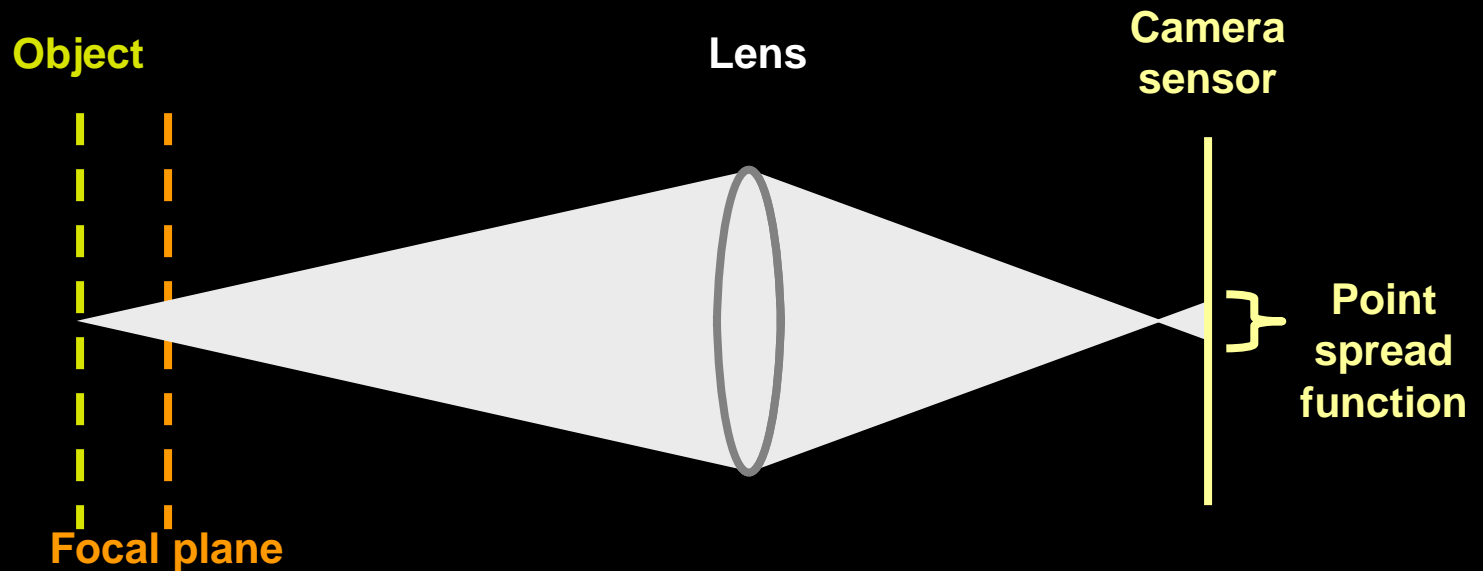


Image of a
defocused point
light source



Lens and defocus

Lens' aperture

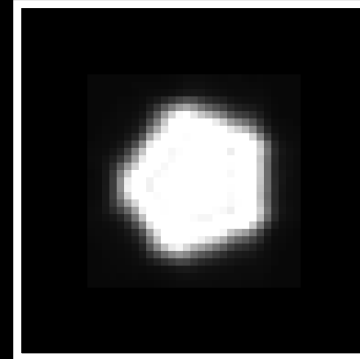
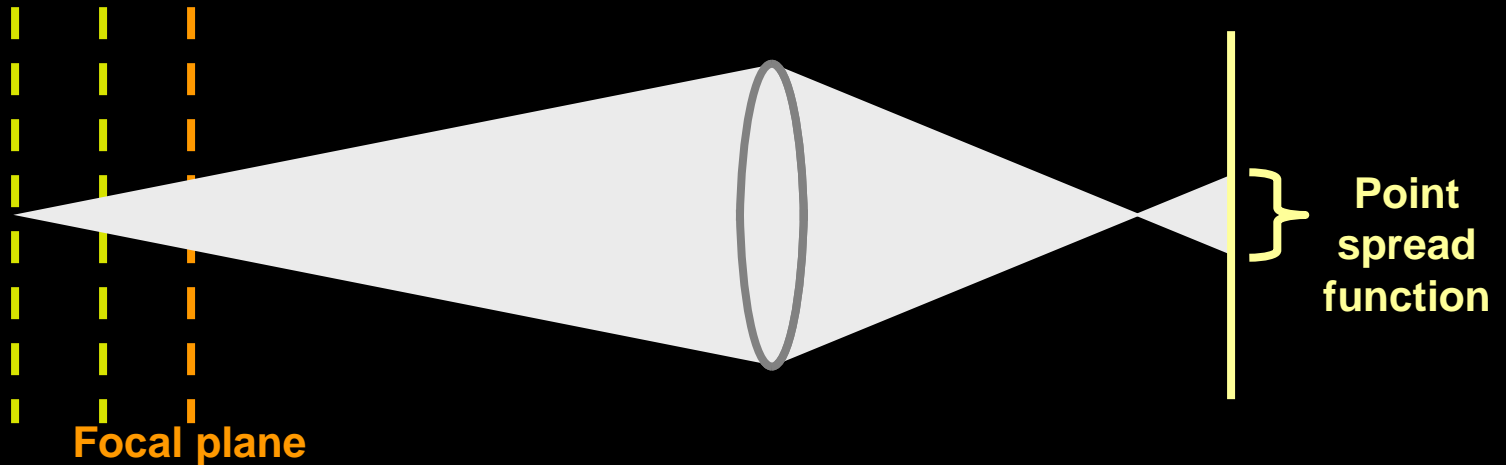


Image of a
defocused point
light source

Object

Lens

Camera
sensor



Lens and defocus

Lens' aperture

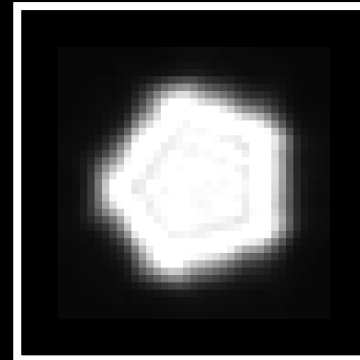
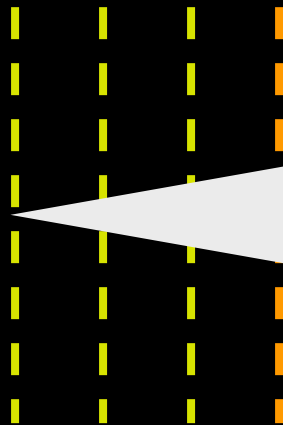


Image of a
defocused point
light source

Object

Lens

Camera
sensor



Focal plane

Point
spread
function

Lens and defocus

Lens' aperture

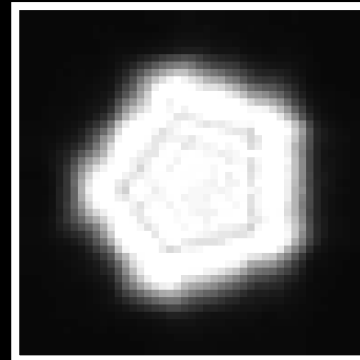
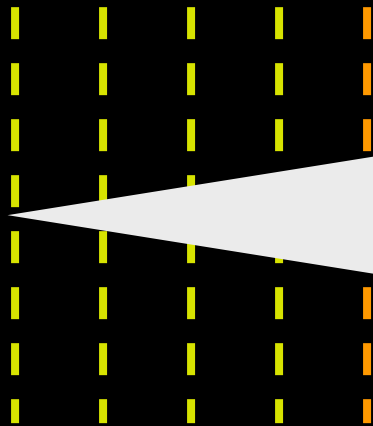


Image of a
defocused point
light source

Object

Lens

Camera
sensor

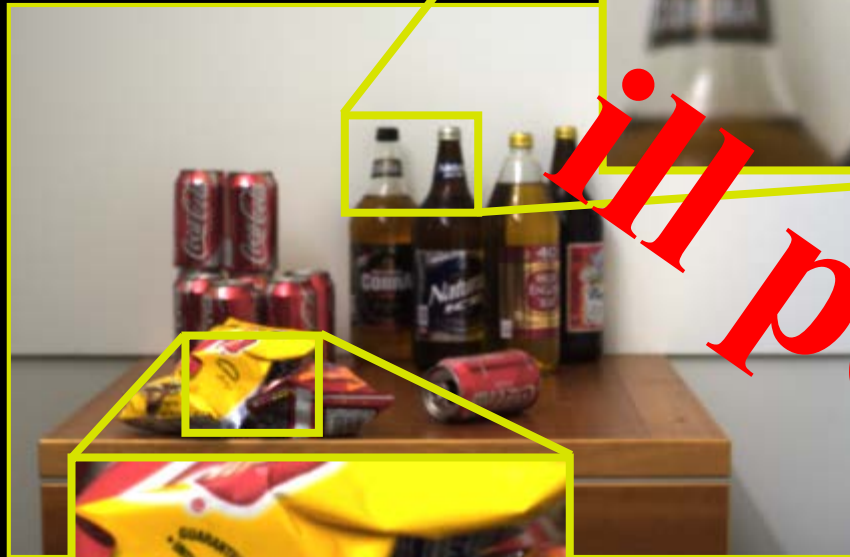


Focal plane

Point
spread
function

Depth and defocus

Out of focus



In focus



Depth from defocus:

Infer depth by analyzing
local scale of defocus blur

Challenges

- Hard to discriminate a smooth scene from defocus blur

?

Out of focus



- Hard to undo defocus blur



Input

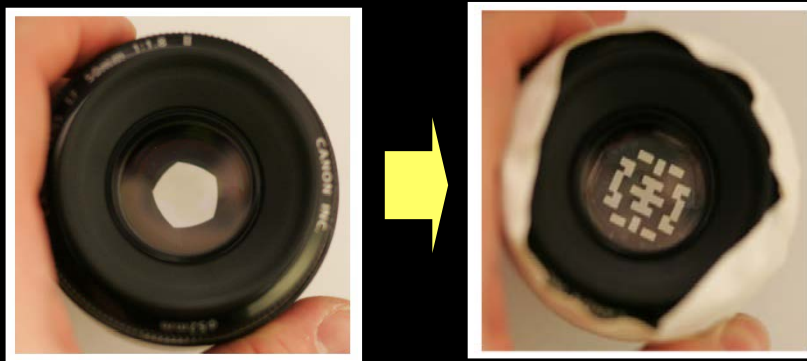


Ringings with conventional deblurring algorithm

Coded aperture

- Coded aperture (mask inside lens)

- make defocus patterns different from natural images and easier to discriminate
- defocus kernel preserves more high frequencies



Coded aperture: lens with occluder

Aperture pattern

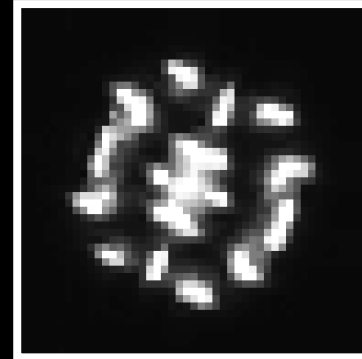


Image of a
defocused point
light source

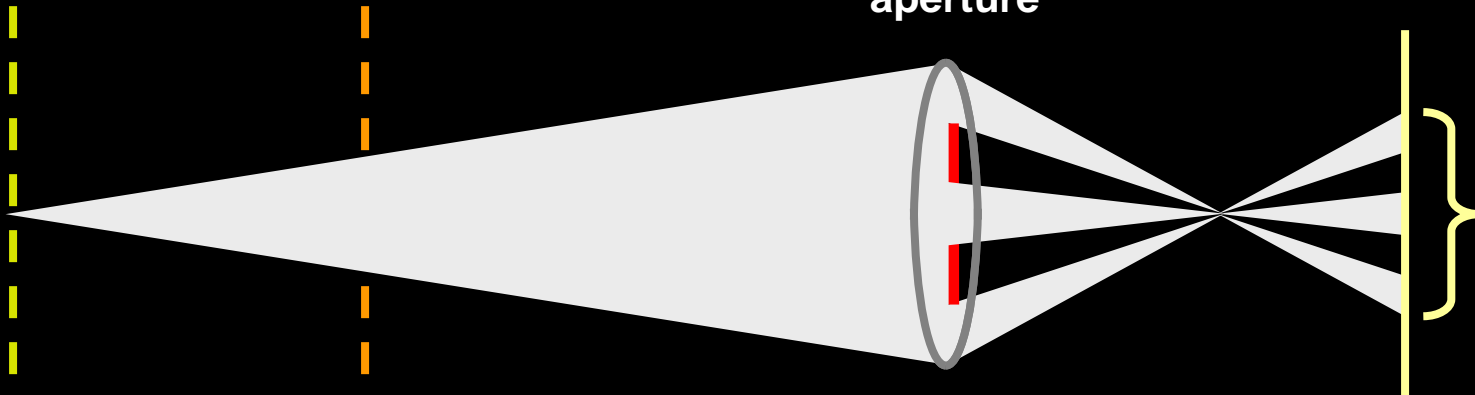
Object

Lens with coded
aperture

Camera
sensor

Focal plane

Point
spread
function



Coded aperture: lens with occluder

Aperture pattern

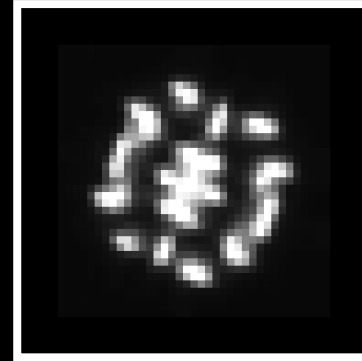
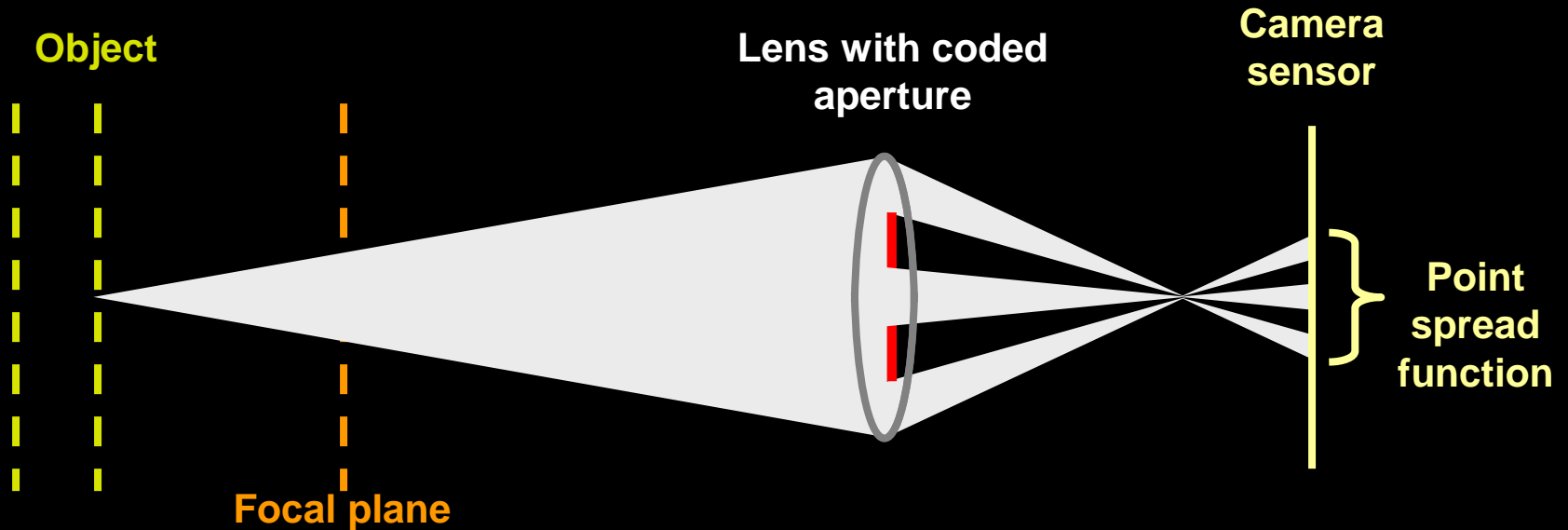


Image of a
defocused point
light source



Coded aperture: lens with occluder

Aperture pattern

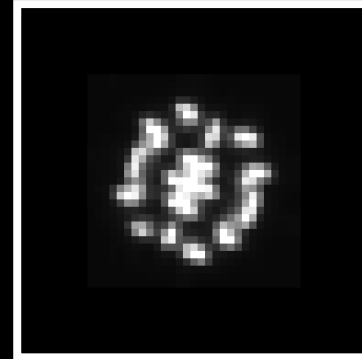
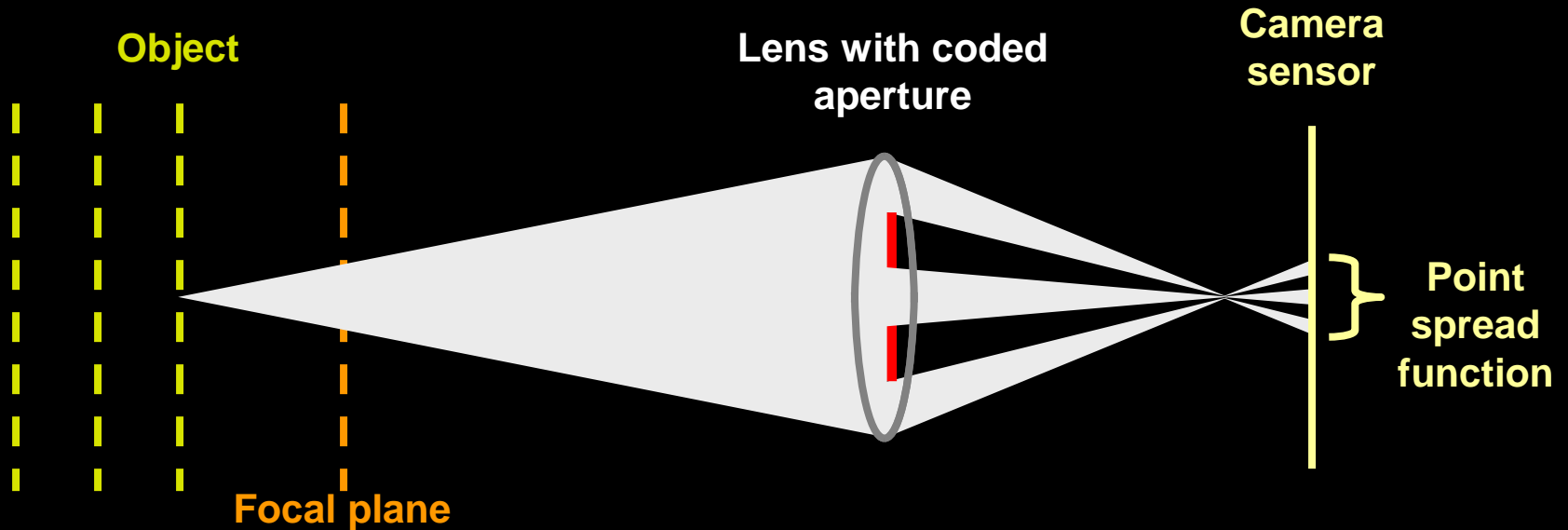


Image of a
defocused point
light source



Coded aperture: lens with occluder

Aperture pattern

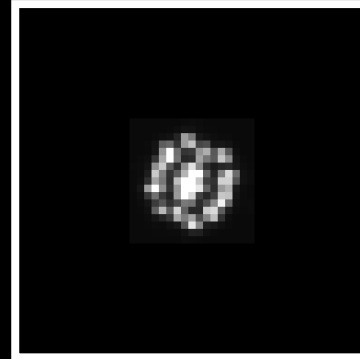
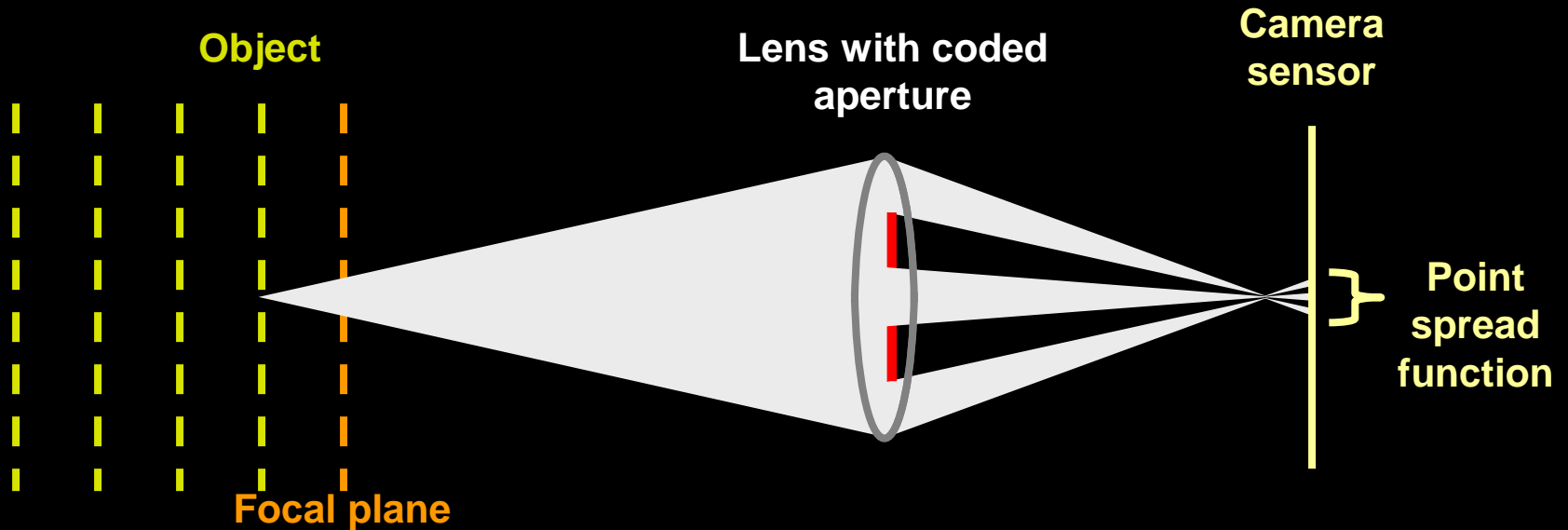


Image of a
defocused point
light source



Coded aperture: lens with occluder

Aperture pattern

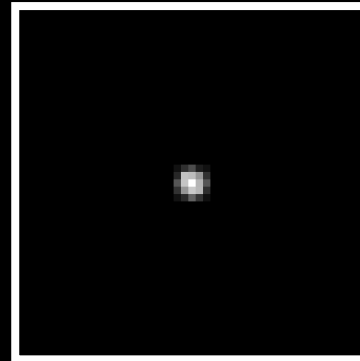
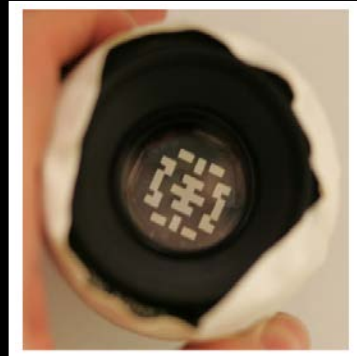
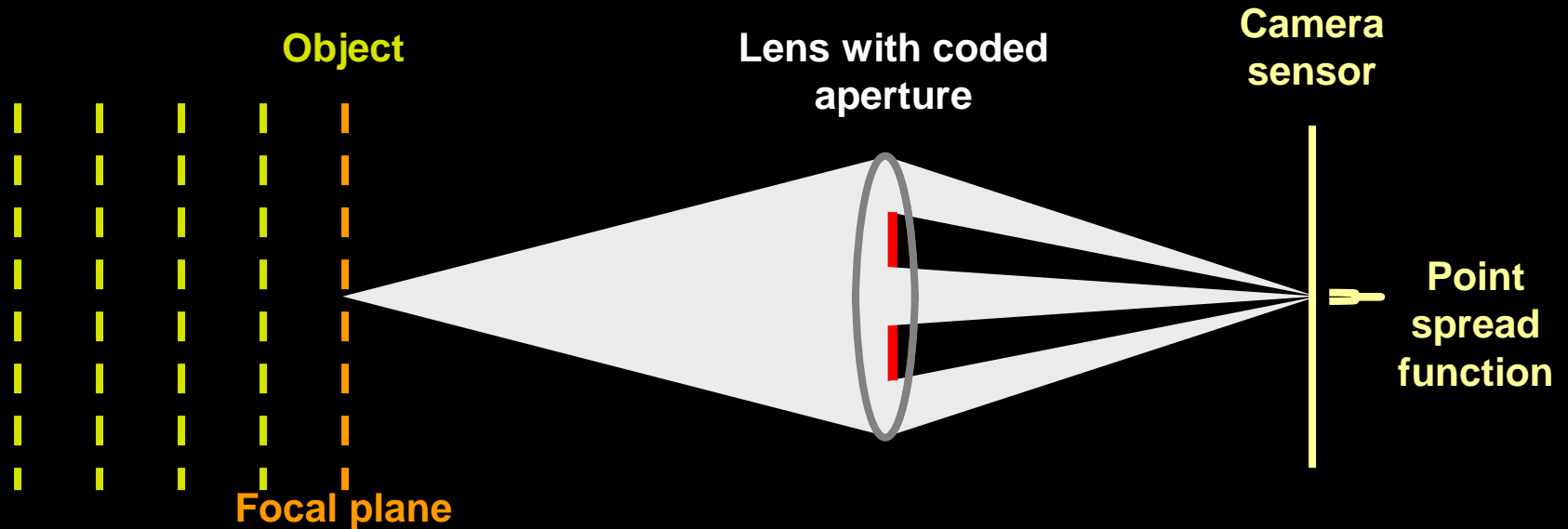
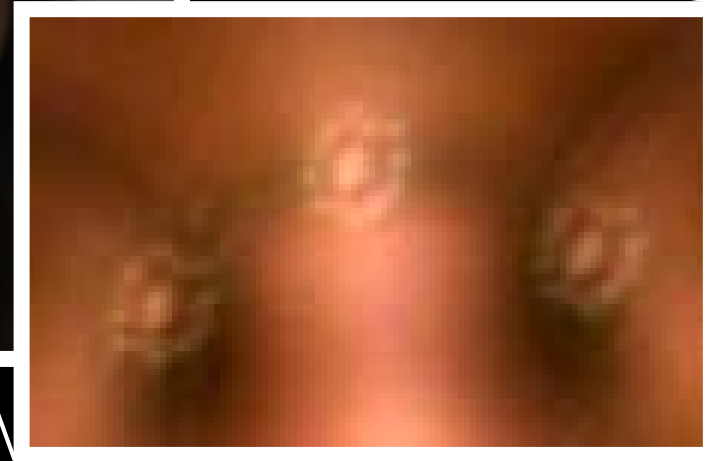
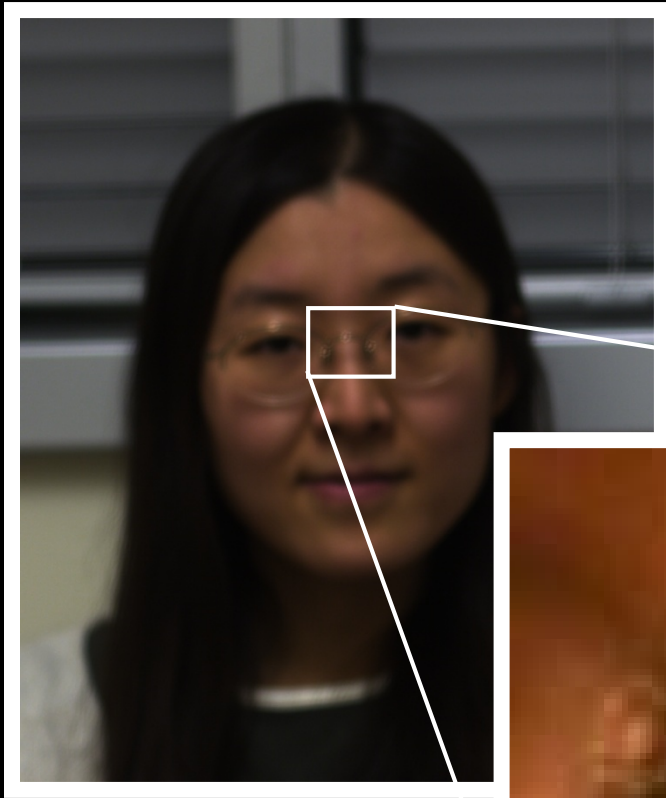


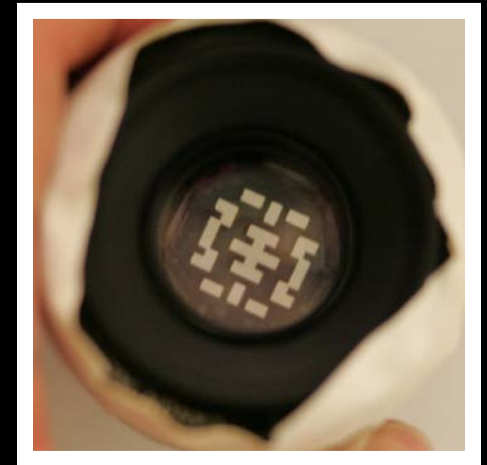
Image of a
defocused point
light source



Defocused images \neq natural images



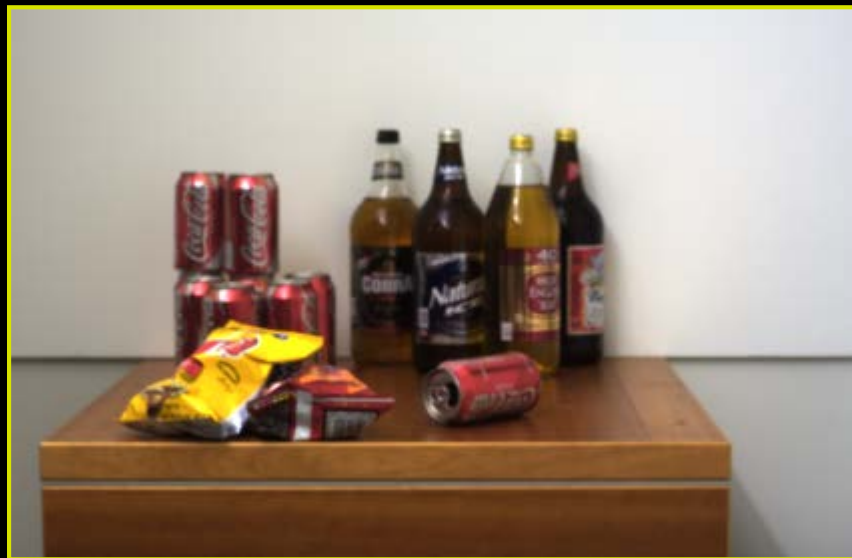
Captured Image



Aperture code

➡ We have strong priors on natural images and can make defocused images look differently

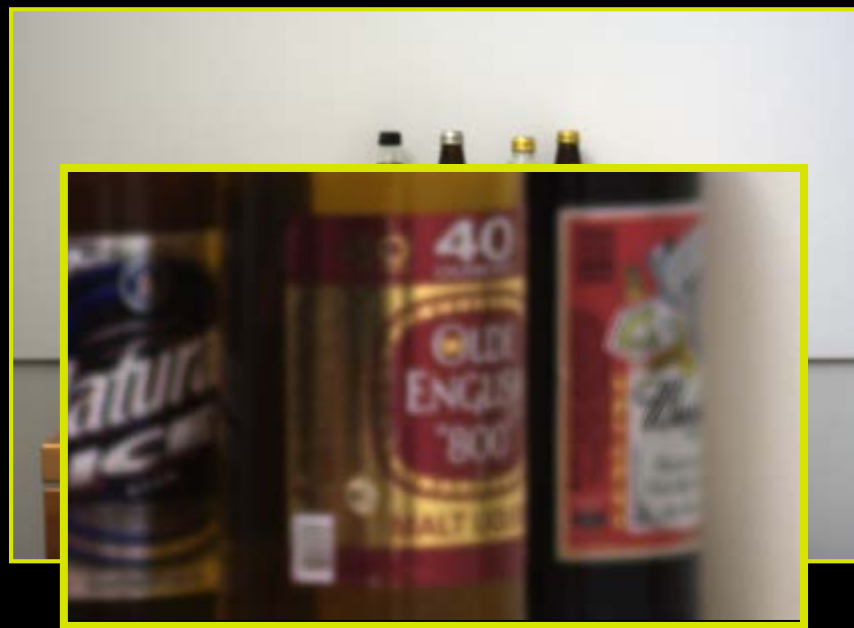
Single input image:



Output #1: Depth map



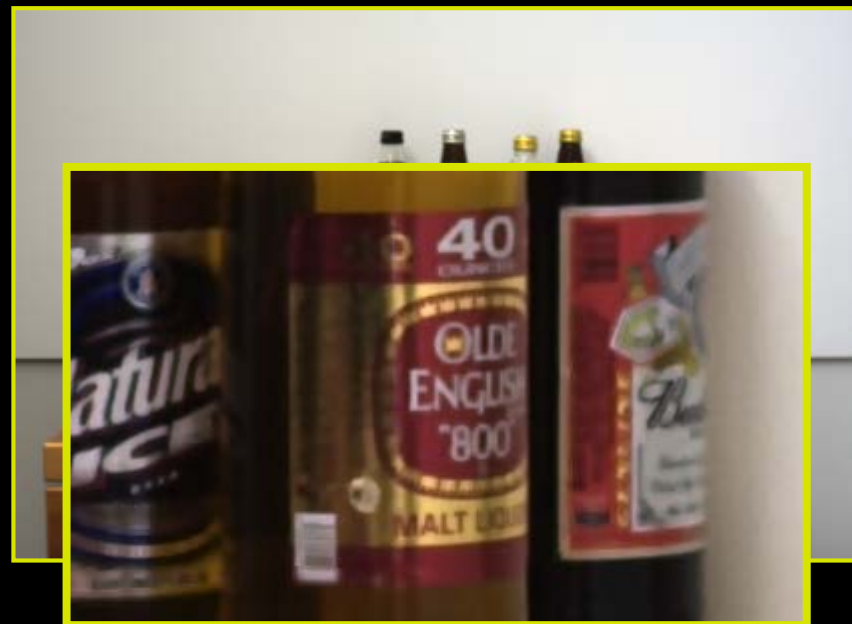
Single input image:



Output #1: Depth map

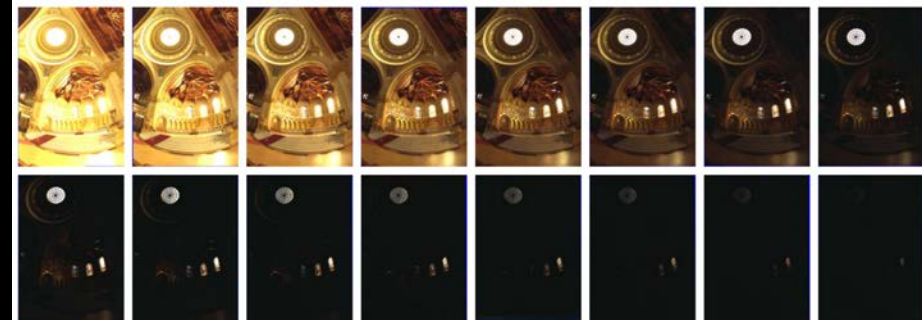
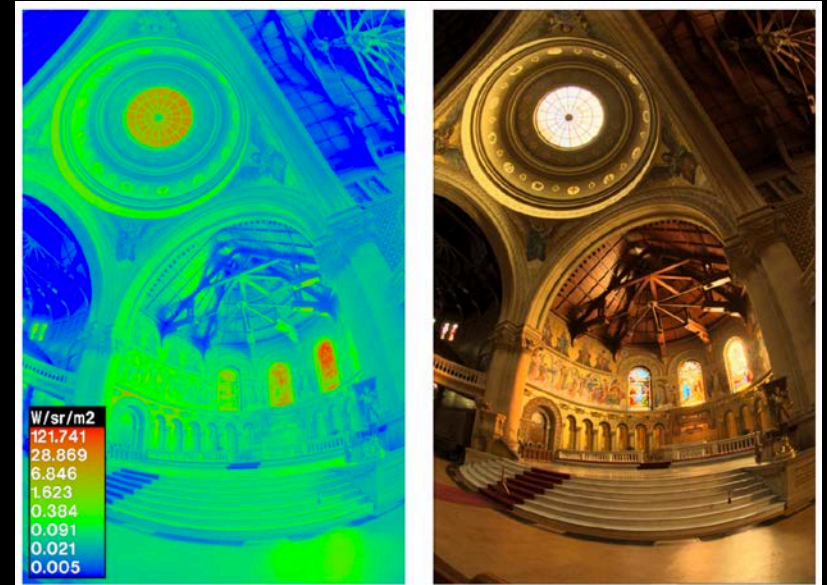


Output #2: All-focused image



Burst Photography

- high dynamic range
- super-resolution
- noisy / blurry
- Focal stack



Light Fields and The Plenoptic Function



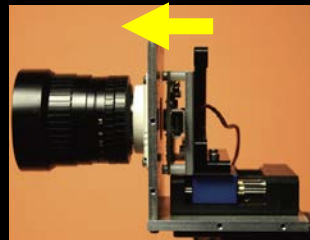
- camera arrays
- integral imaging
- coded masks
- refocus
- fourier slice photography

The Computational Camera Zoo

Goal: a unified mathematical framework

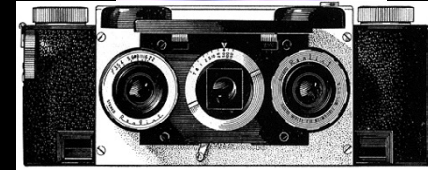
- Evaluate and compare computational cameras
- Systematically design new cameras

Conventional single-lens cameras



Focus sweep

Stereo and trinocular cameras



Coded aperture



Lattice-focal lens



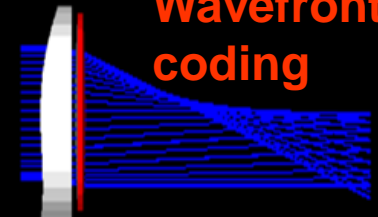
Plenoptic cameras



Multi-aperture



Wavefront coding

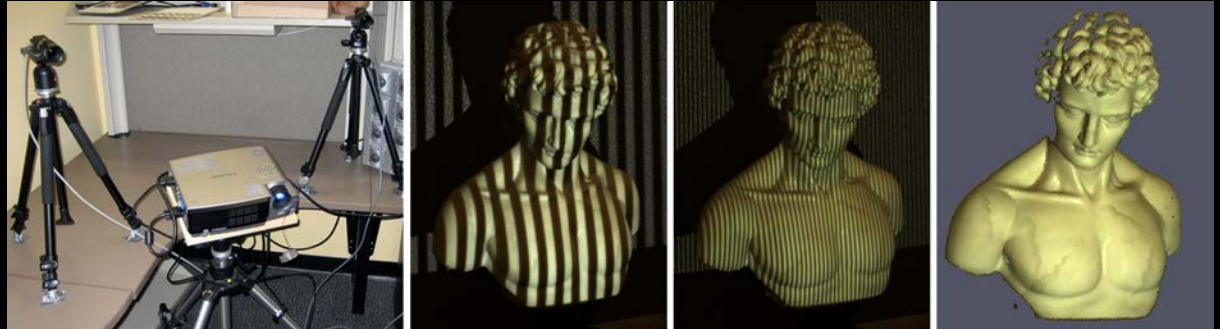


Computational illumination

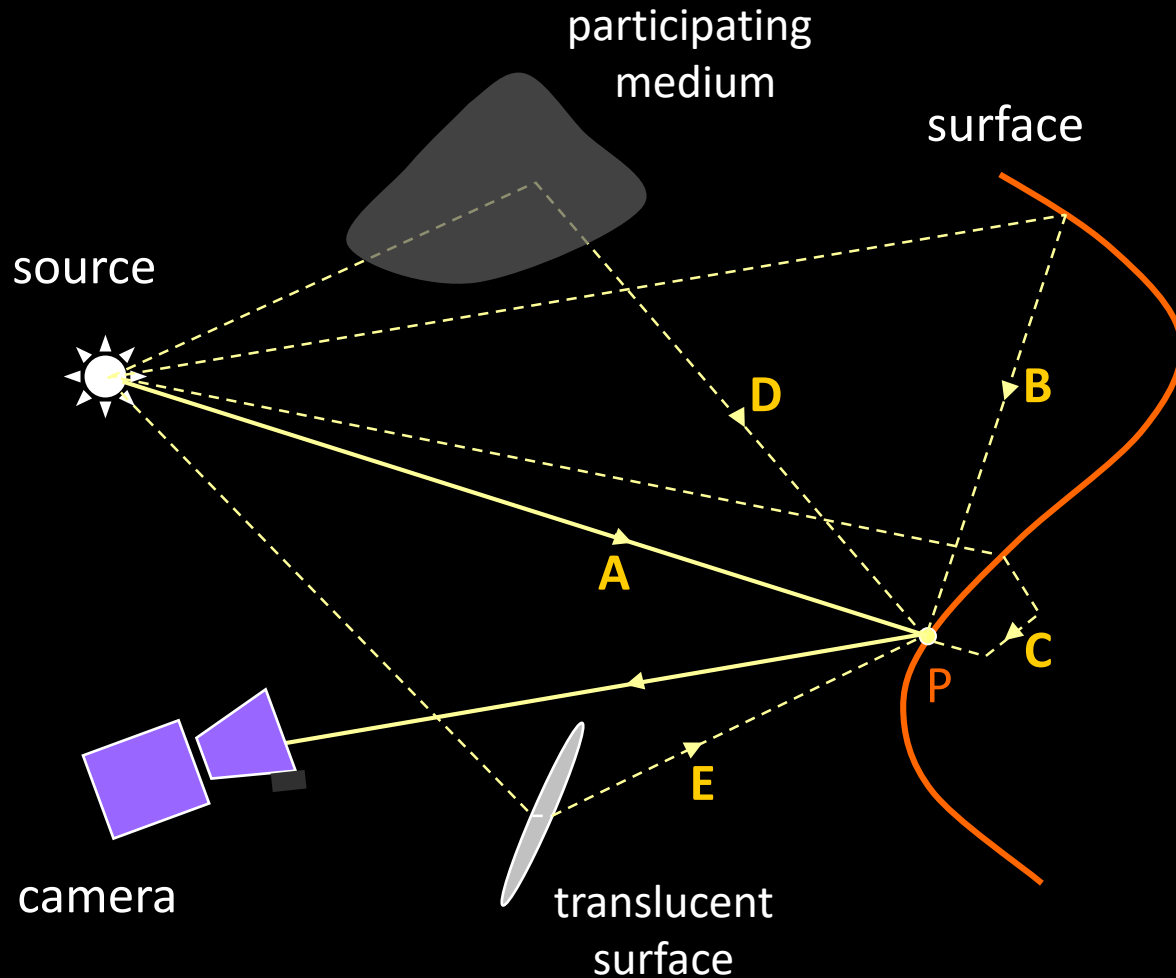
Infer on the interaction between scene and light

Computational illumination

- time of flight
- structured illumination
- photometric stereo
- multi-flash photography
- microsoft kinect
- leap motion

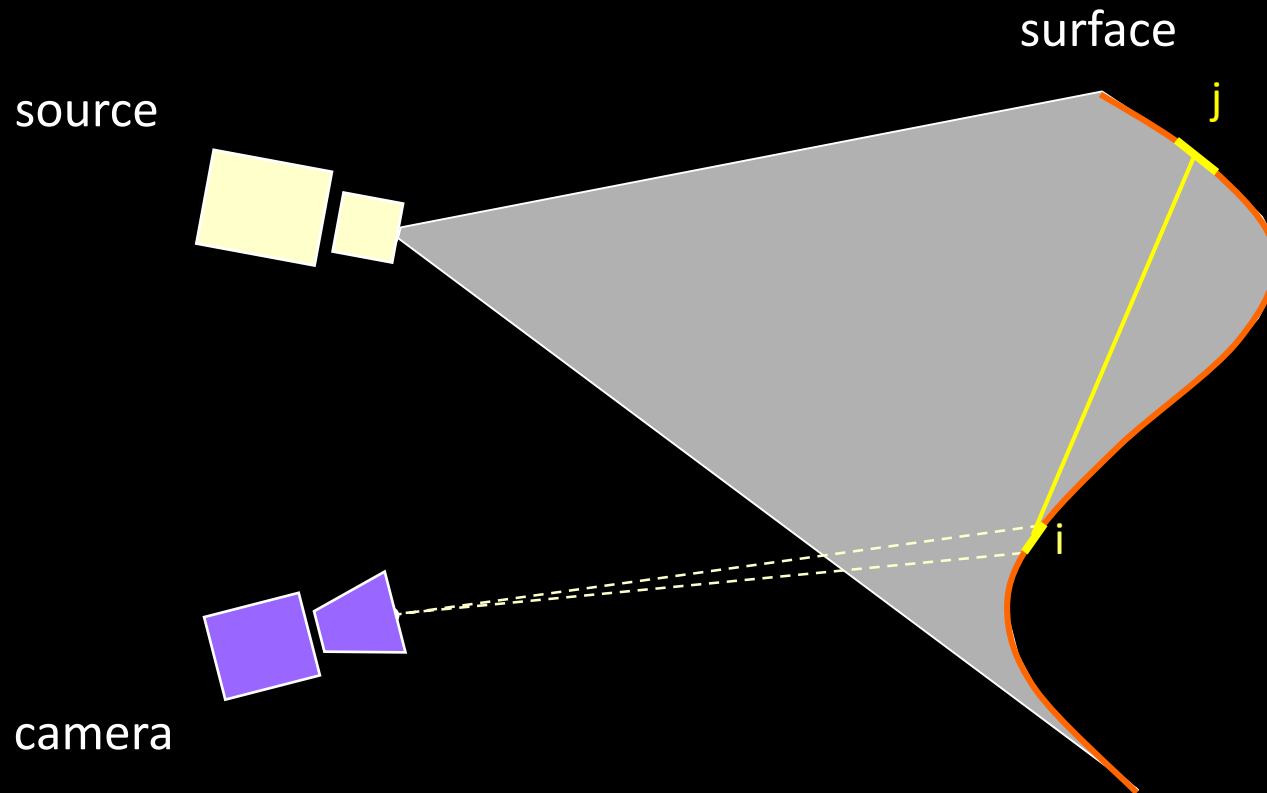


Direct and Global Illumination



- A : Direct
- B : Interreflection
- C : Subsurface
- D : Volumetric
- E : Diffusion

Direct and Global Components: Interreflections



$$L[c, i] = L_d[c, i] + L_g[c, i]$$

radiance

direct

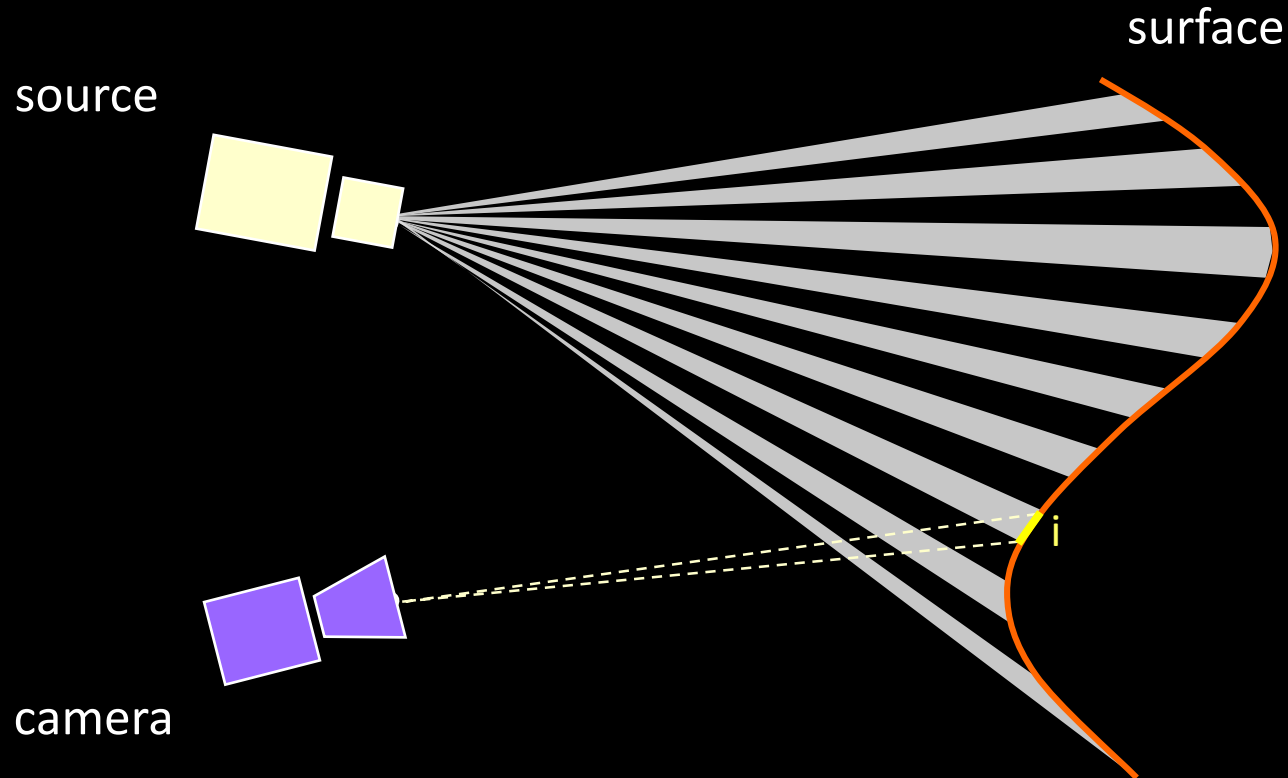
global

$$L_g[c, i] = \sum_P A[i, j] L[i, j]$$

BRDF and geometry

Direct-global separation using high-frequency illumination

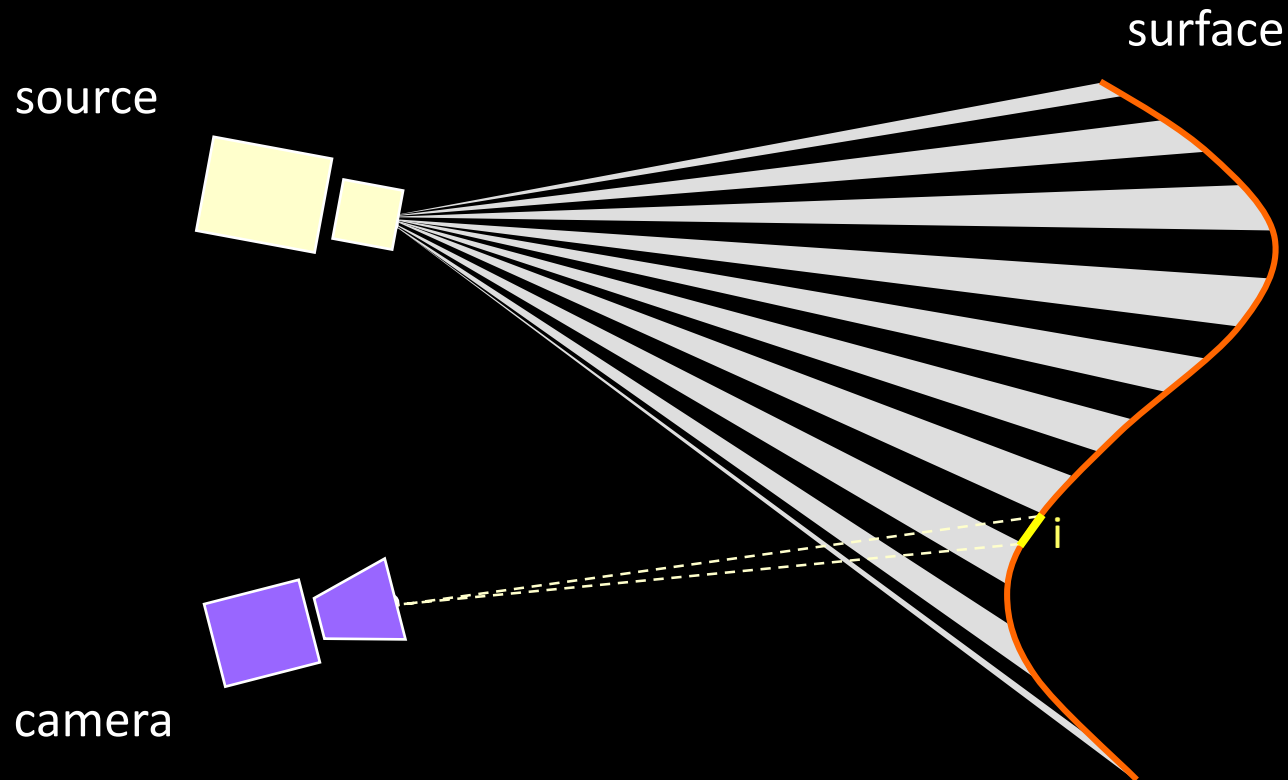
High Frequency Illumination Pattern



$$L^+[c, i] = L_d[c, i] + \alpha L_g[c, i]$$

fraction of activated source elements

High Frequency Illumination Pattern



$$L^+[c, i] = L_d[c, i] + \alpha L_g[c, i]$$

$$L^-[c, i] = (1 - \alpha) L_g[c, i]$$

fraction of activated source elements

Separation from Two Images

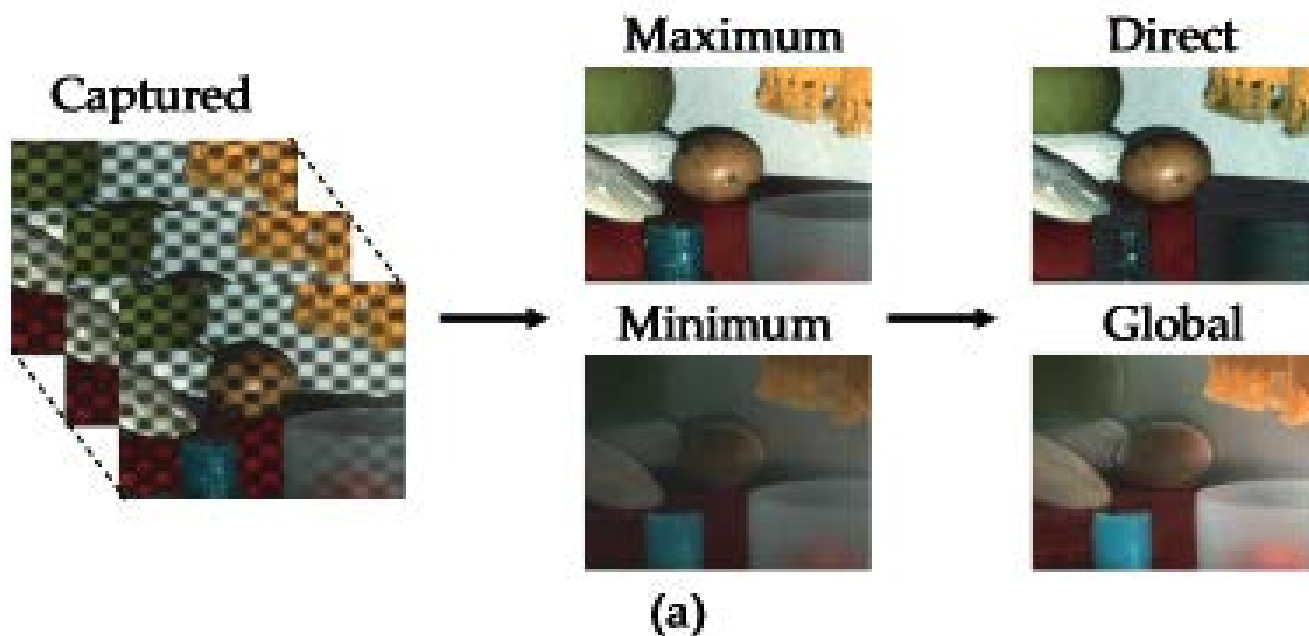
$$\alpha = \frac{1}{2}:$$

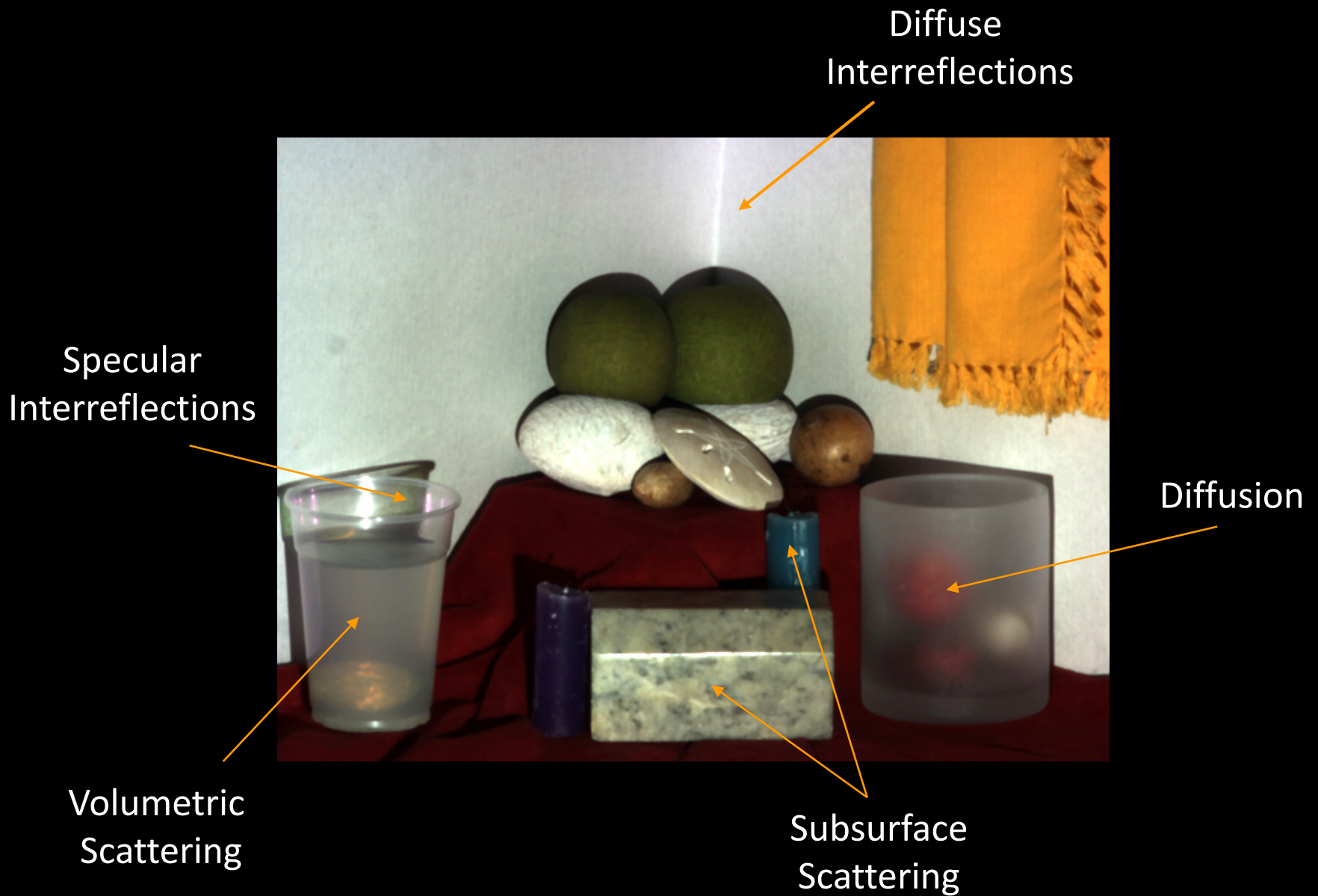
$$L_d = L_{\max} - L_{\min} , L_g = 2L_{\min}$$

direct

global

Separation of Direct and Global Images





Scene



Direct



Global

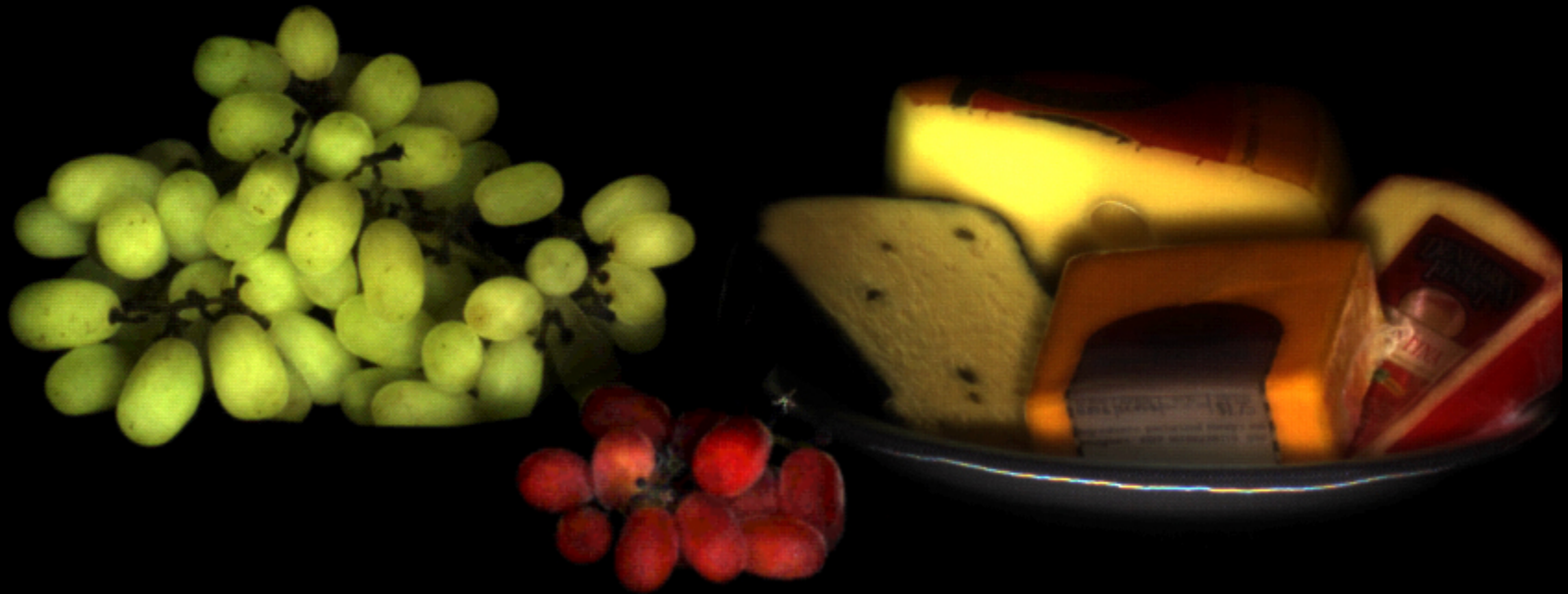
Separation of Direct and Global Images



Separation of Direct and Global Images



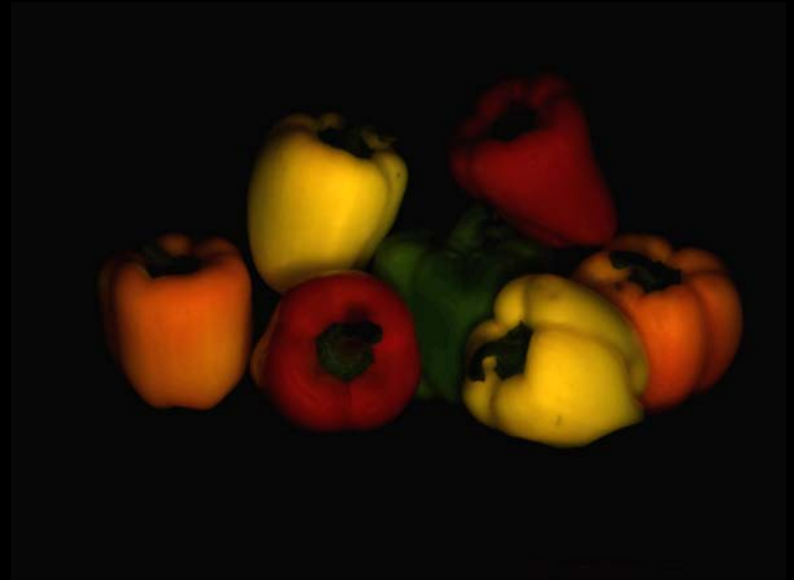
Separation of Direct and Global Images



Peppers: Subsurface Scattering



Direct



Global

Eggs: Diffuse Interreflections

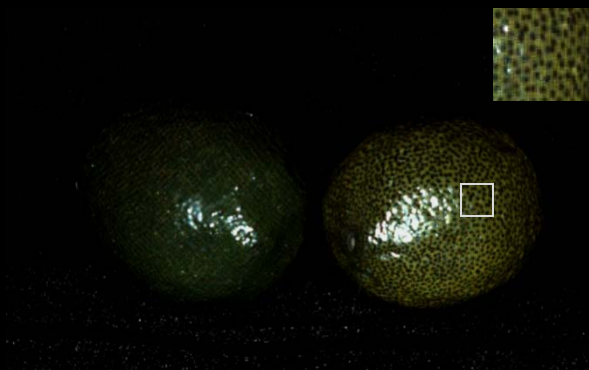
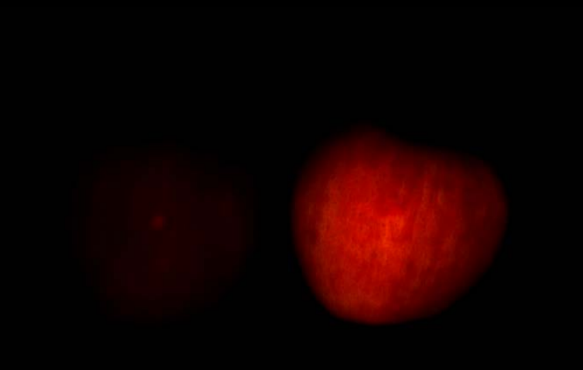
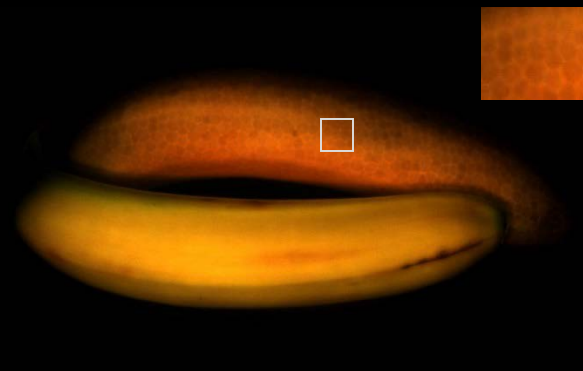
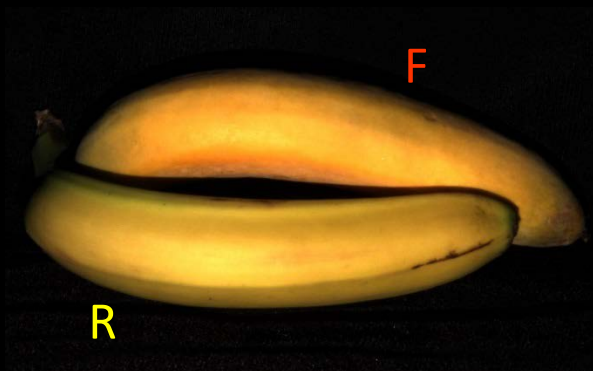


Direct



Global

Real Fake



Direct

Global

Pink Carnation



Spectral Bleeding: Funt et al. 91



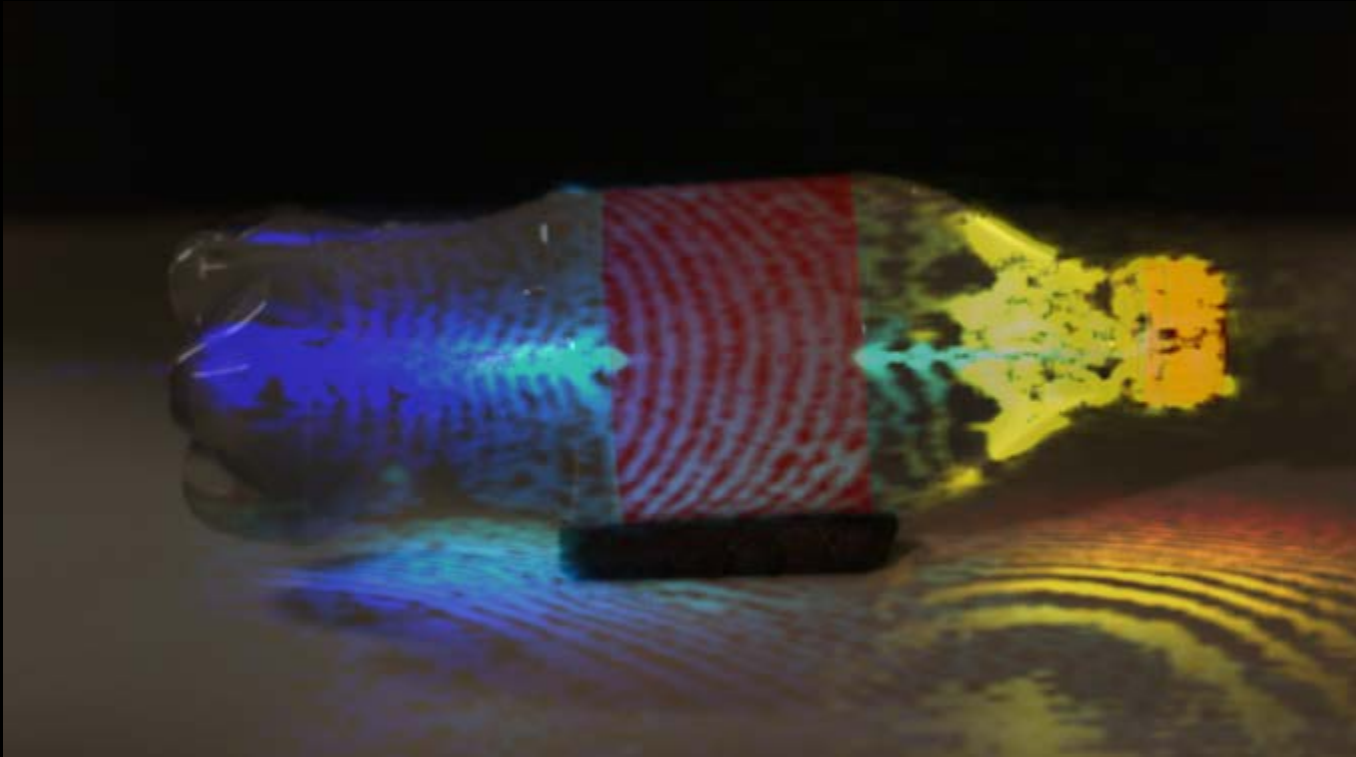
Direct



Global

Transient imaging: visualizing photons in motion

<http://web.media.mit.edu/~raskar//trillionfps/>



Transient imaging: visualizing photons in motion

