

Computational Photography

- * Study the basics of computation and its impact on the entire workflow of photography, from capturing, manipulating and collaborating on, and sharing photographs.



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Stereo Photography / Imagery

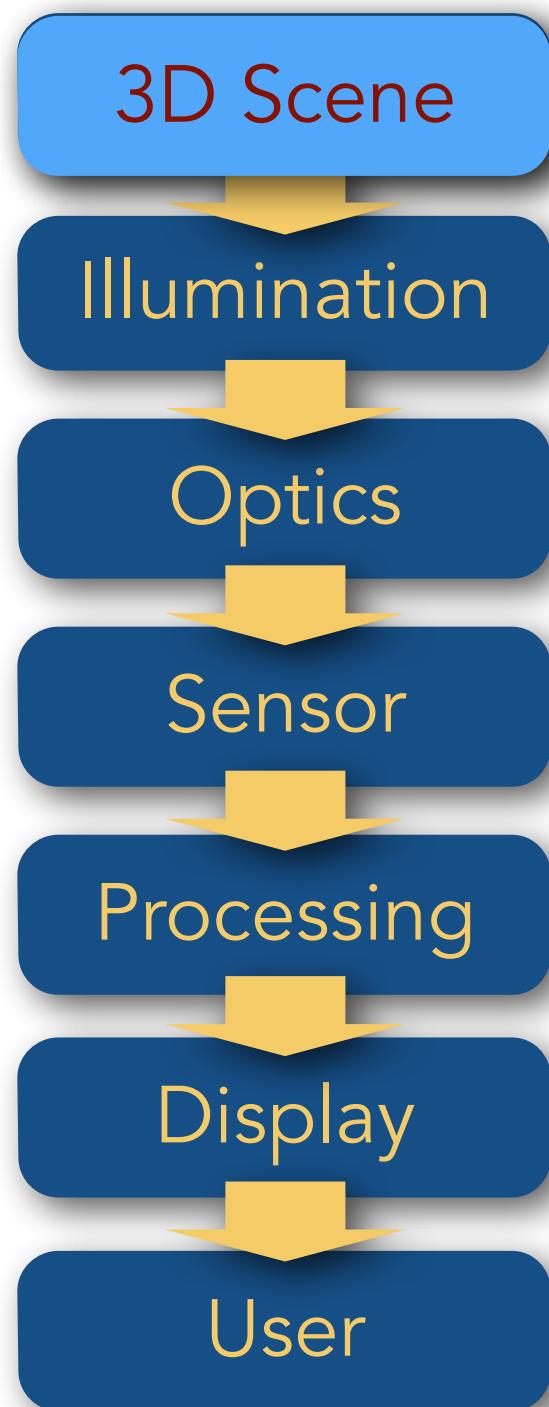
- * Depth / Range Imaging
- * Structure / 3D Scenes
- * Stereo . . Depth from 2 views



Lesson Objectives

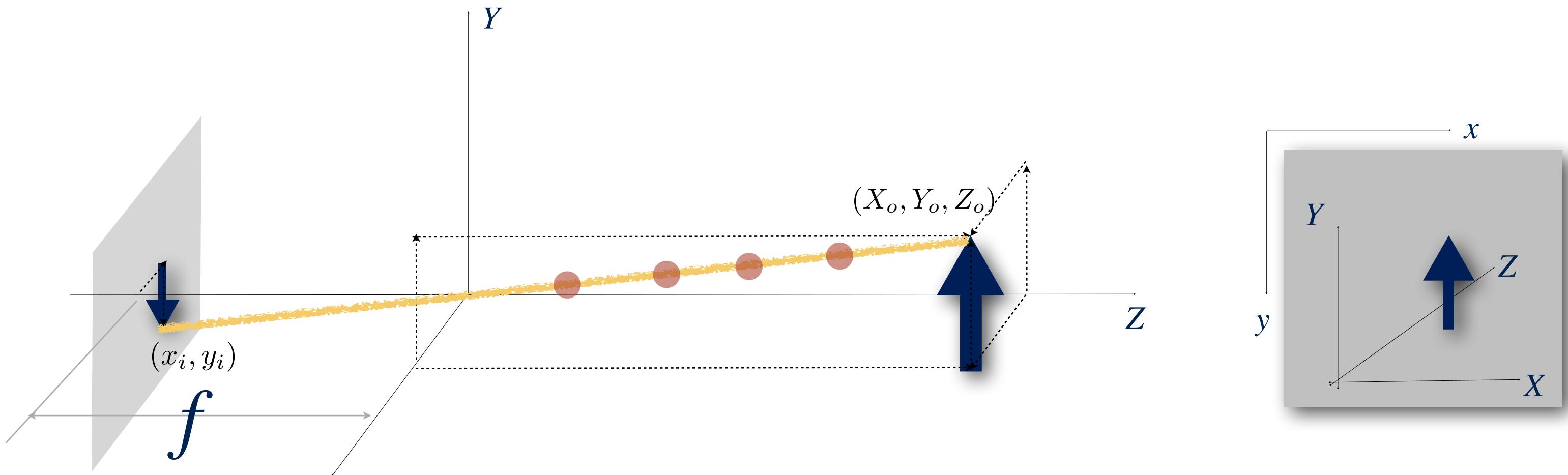
1. Geometry (Depth/Structure) in a scene
2. Stereo
3. Parallax
4. Compute Depth from a stereo image pair

Depth (of a Scene)



- * Above all, we are interested in capturing a 3D scene with Geometry
- * Need Depth, Geometry, 3D

Compute depth / structure



$$x_i = \frac{X_o}{Z_o} f$$

$$x_i = \frac{k X_o}{k Z_o} f$$

$$y_i = \frac{Y_o}{Z_o} f$$

$$y_i = \frac{k Y_o}{k Z_o} f$$

Fundamental Ambiguity: Any points along the same ray map to the same point in the image

Depth ambiguity



Images from S. Lazebnik and I. Essa

Depth Cues



Perspective
(Vanishing
Lines/points)

Depth Cues



Objects of
known sizes

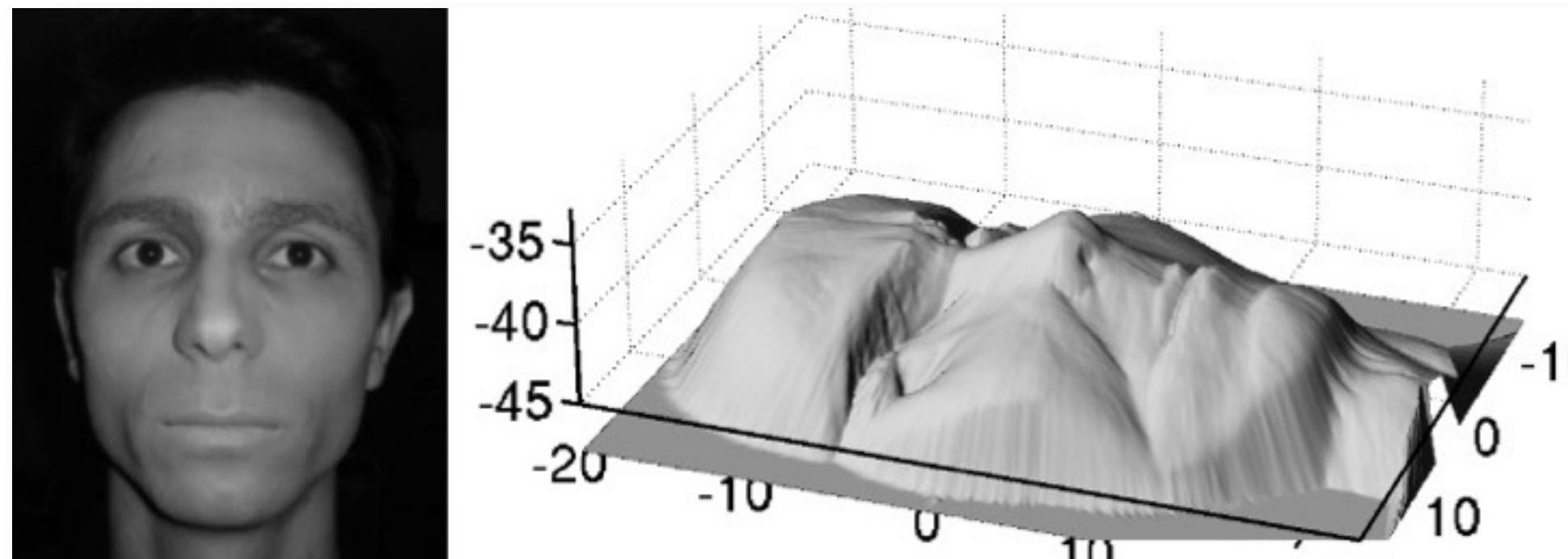
Depth Cues



Occlusions,
etc

Depth Cues

Shape from
Shading

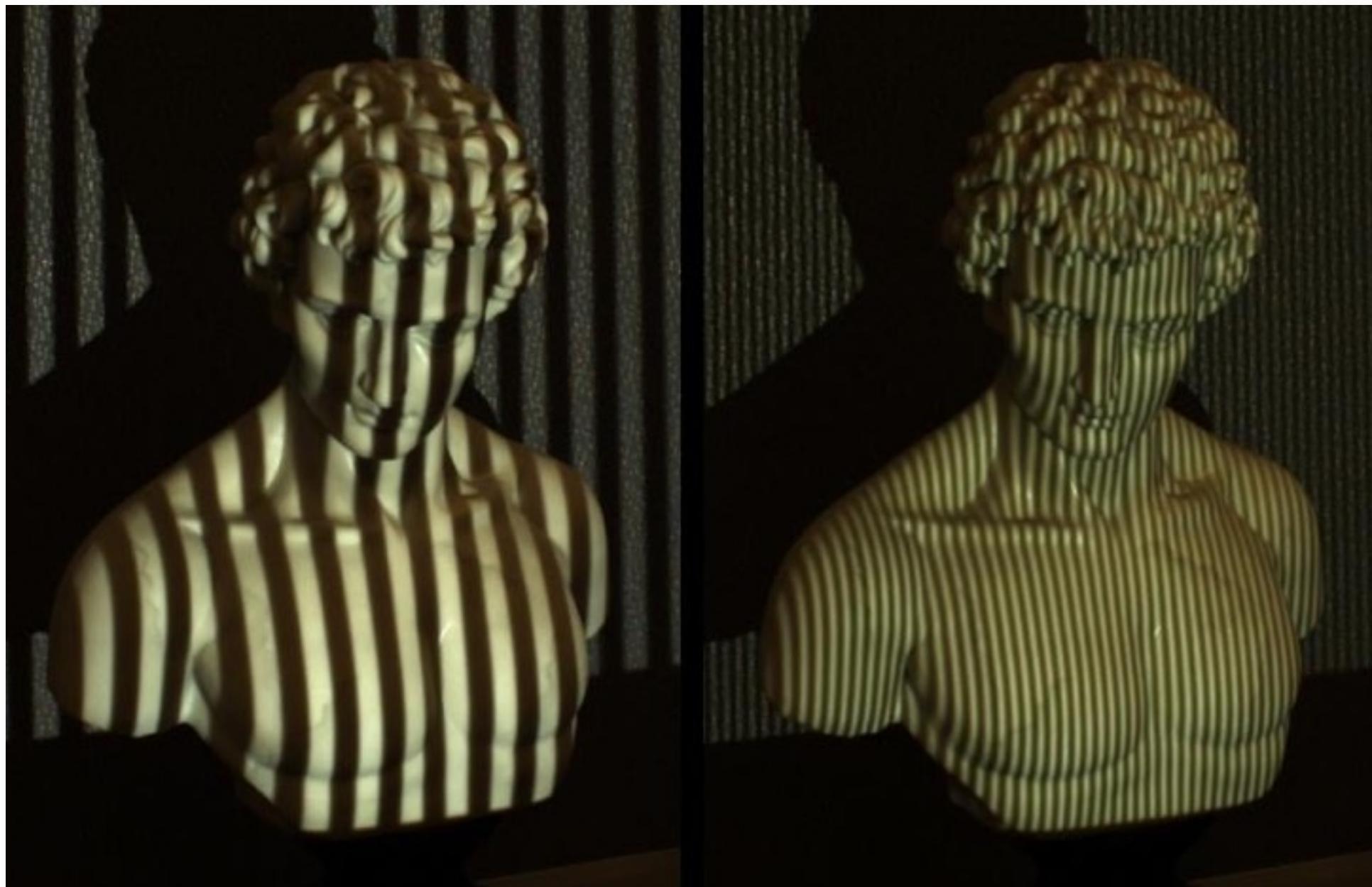


trimensional

3D Scanner for iPhone

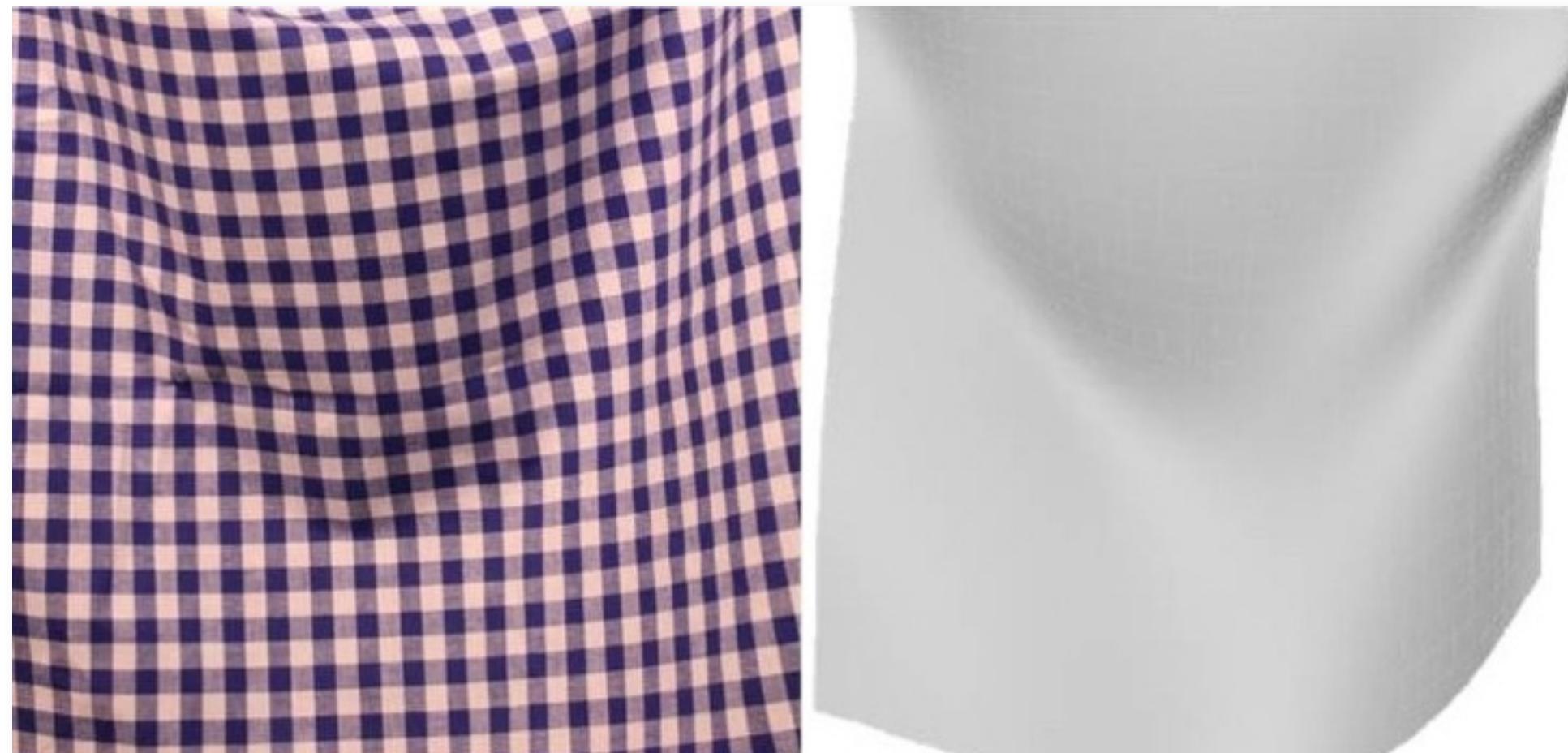
Depth Cues

Shape from
Structured
Light



Depth Cues

Shape from
Texture



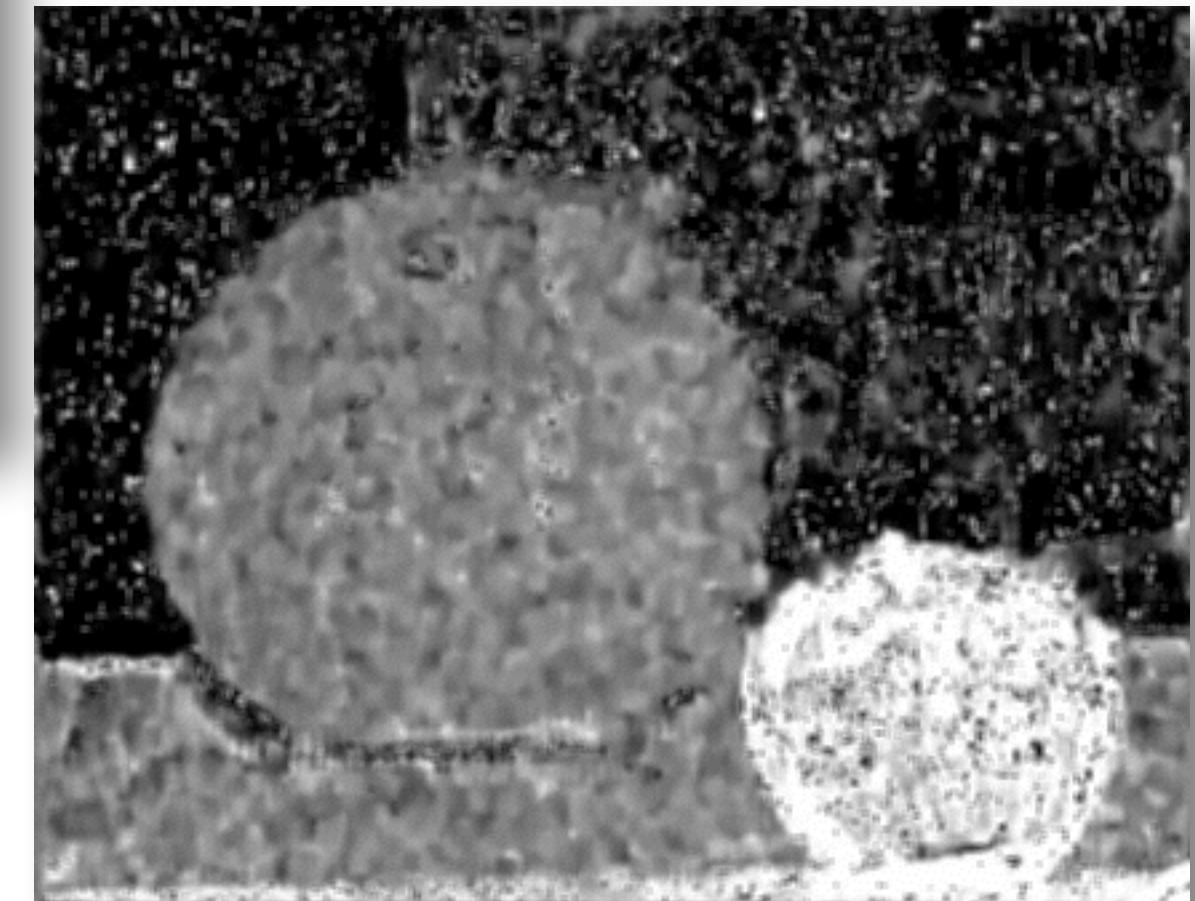
Hilsmann, Schneider, Eisert (2011)

Depth Cues



Shape from
De Focus

Favaro, & Soatto (1999)



Depth Cues

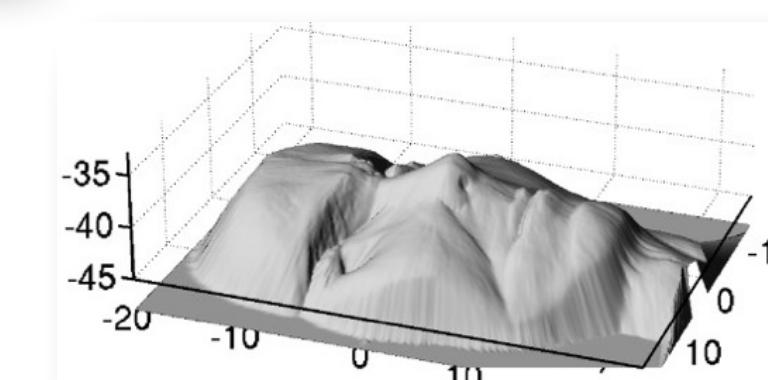
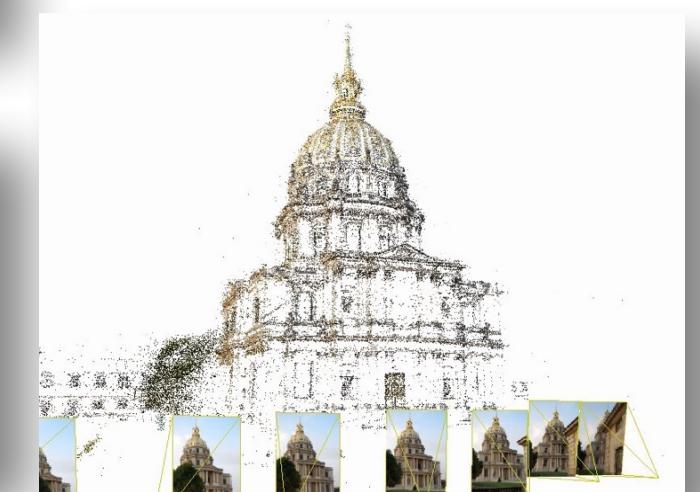
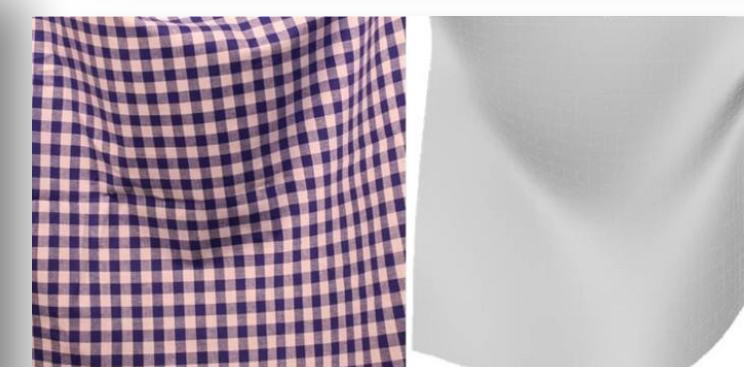
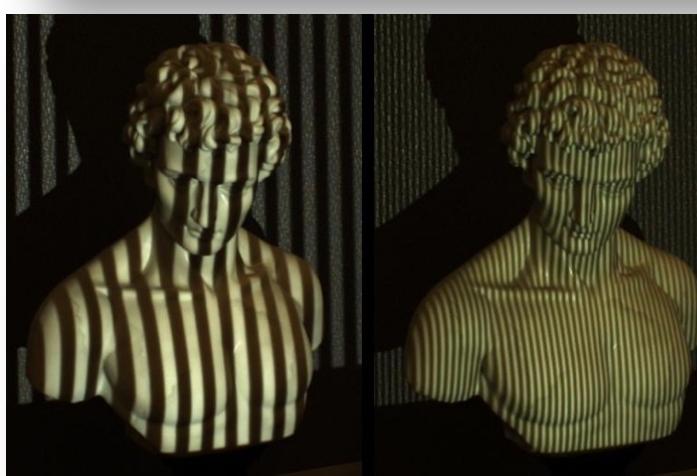
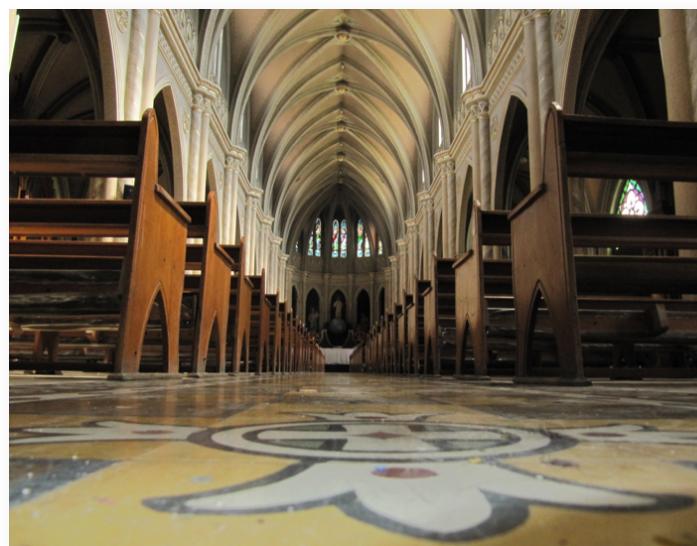


Structure
from
Motion

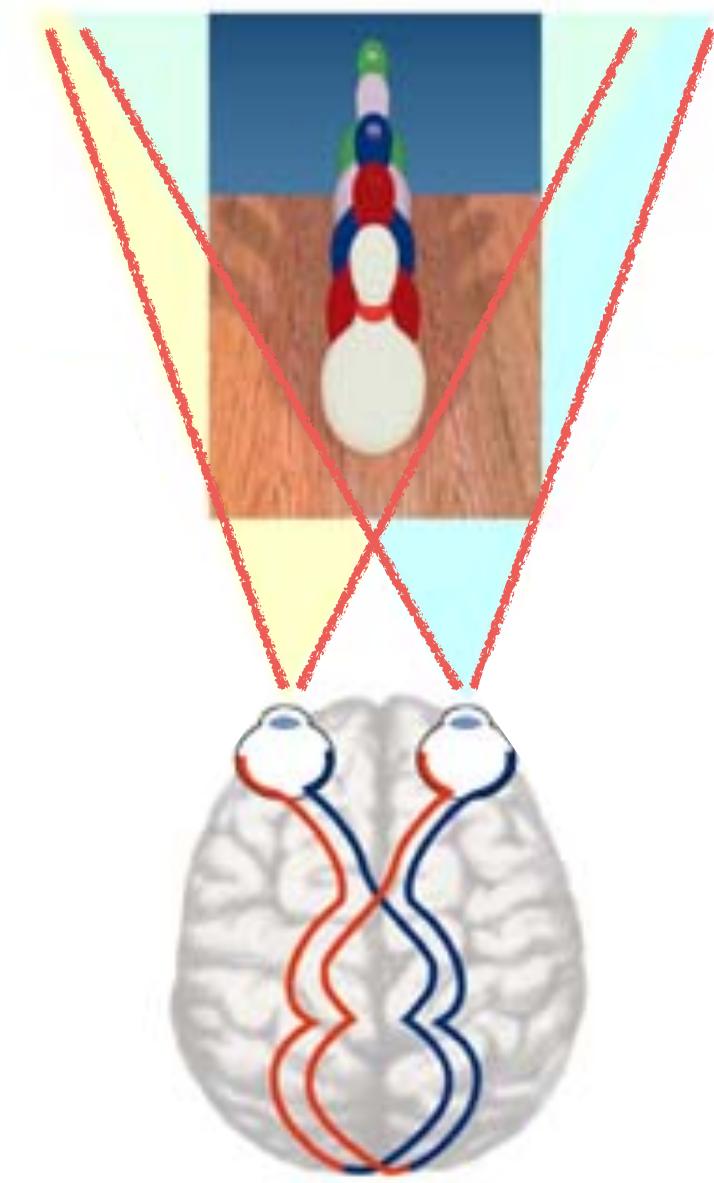
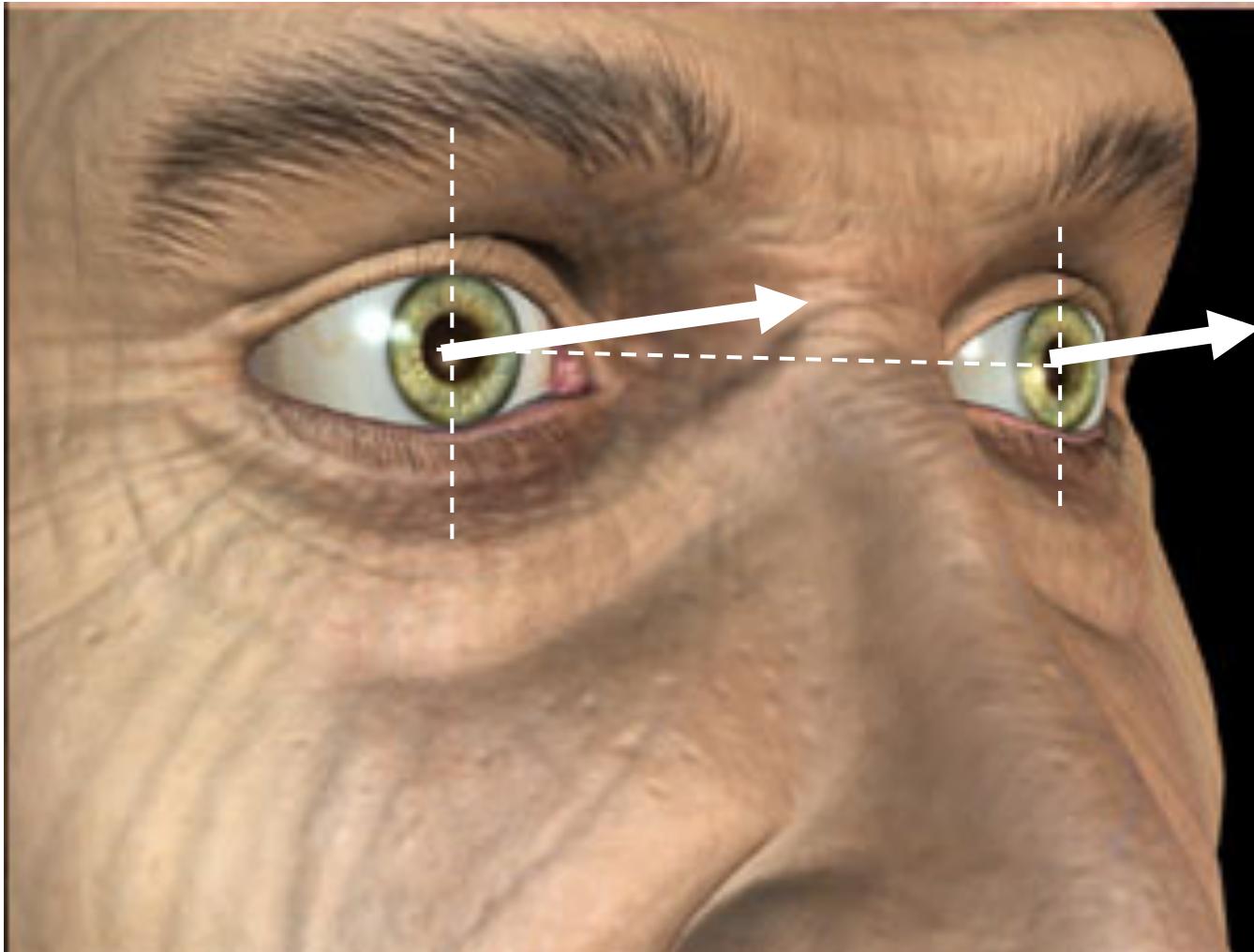
Fredriksson
and Olsson (2012)

Estimating Depth / Shape From One View

- * Shape from X
- * Perspective
- * Shading
- * Motion
- * Focus
- * Occlusions
- * Objects



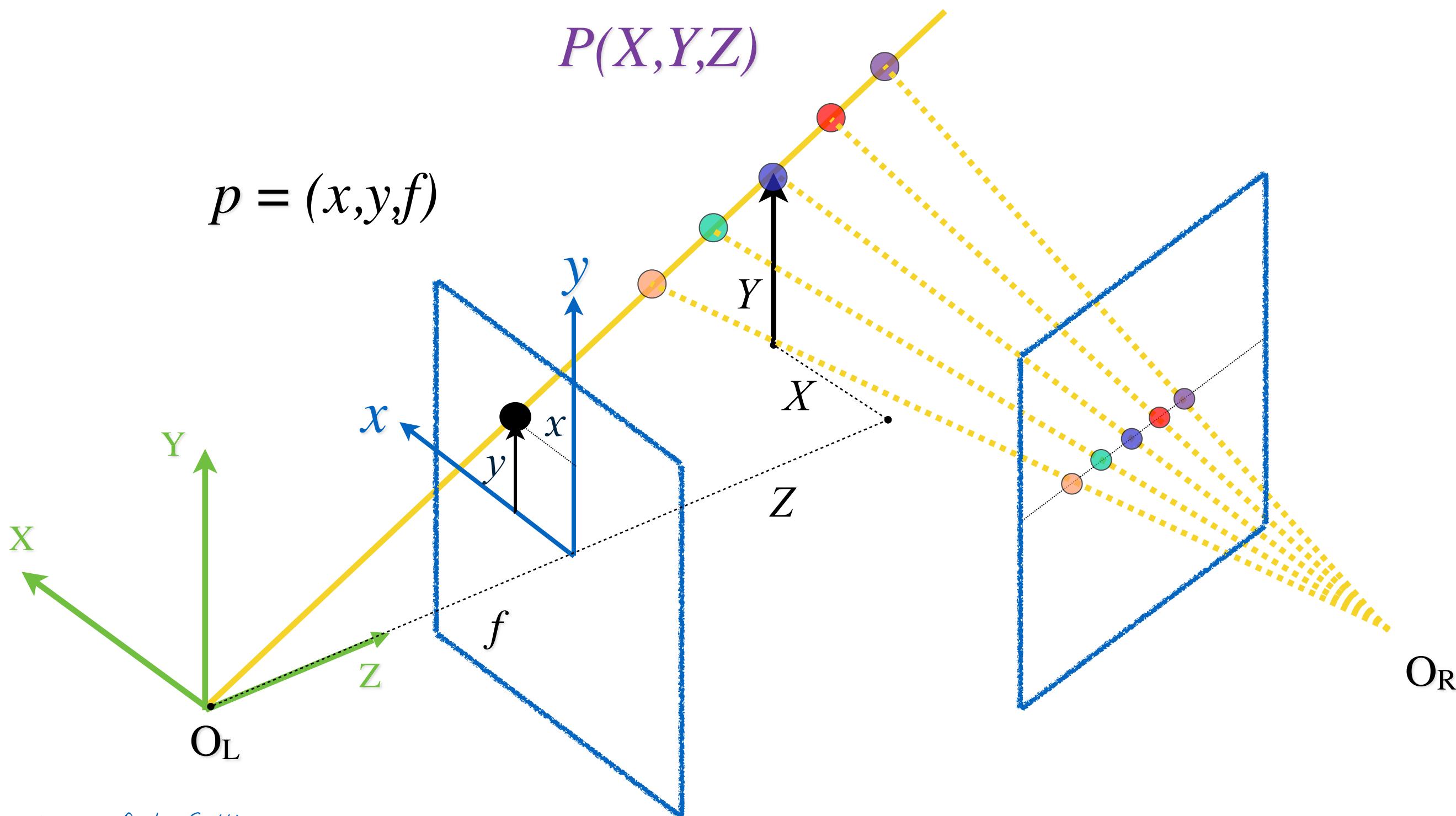
Stereo Vision



Inferring depth from images captured at the same time by two or more cameras

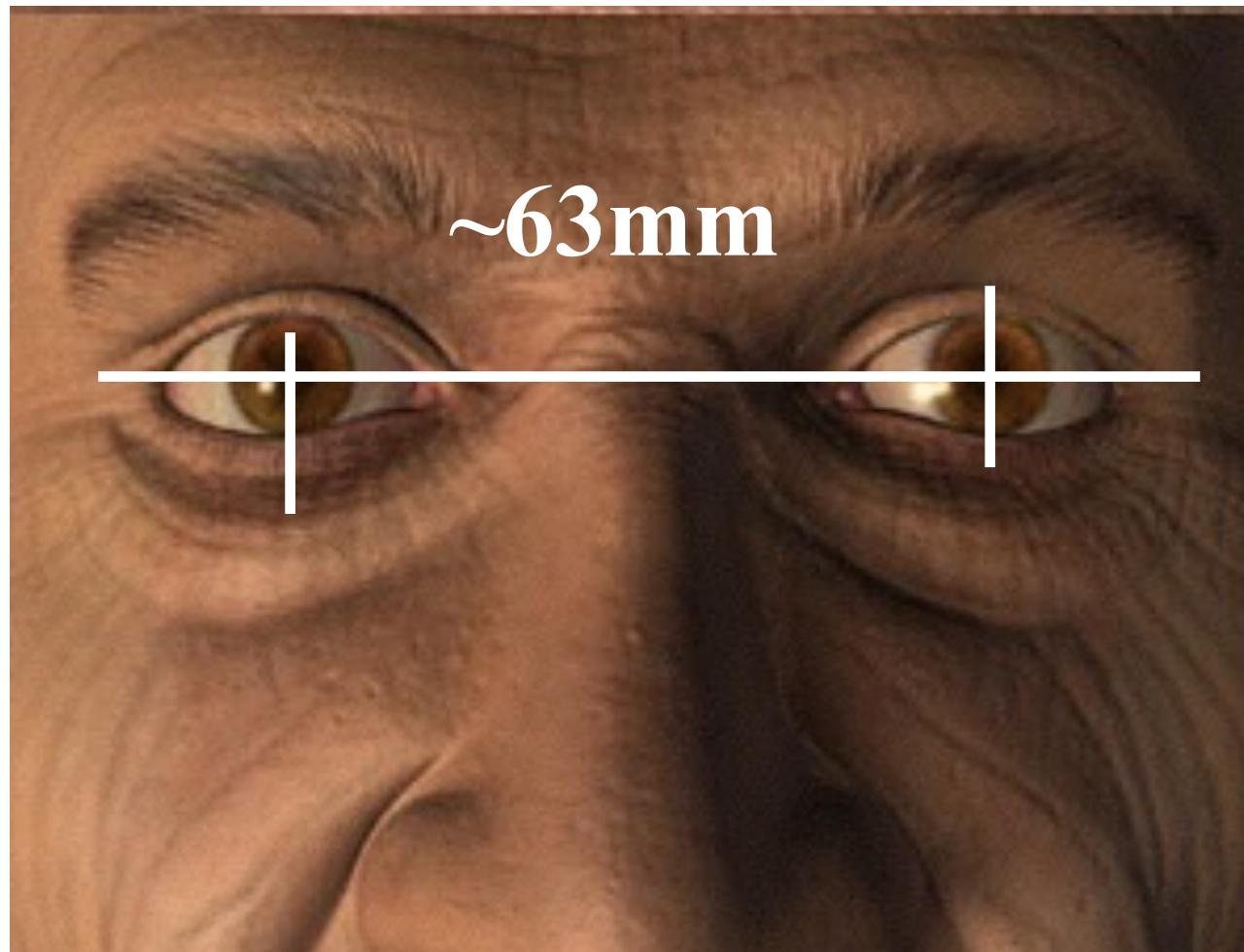
Slide courtesy: Bob Collins

Why Stereo Vision?



Slide courtesy: Bob Collins

Why Stereo Vision?



Eyes form a stereo system

The right and left eyes see the world from
slightly shifted vantage points

Slide courtesy: Bob Collins

Parallax

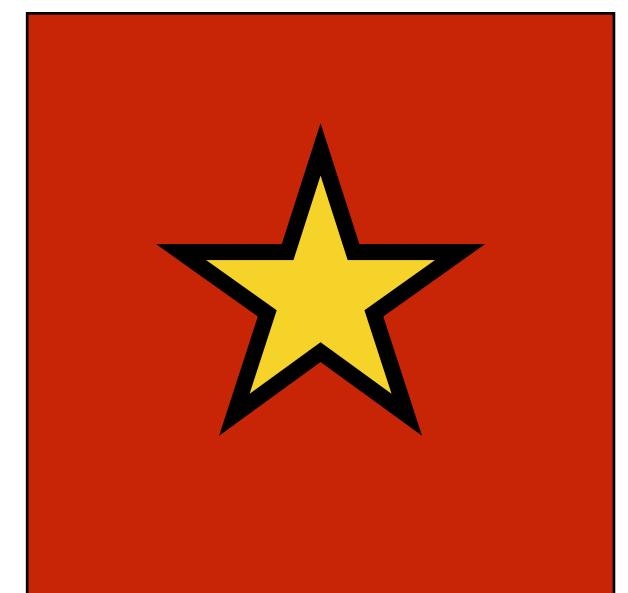
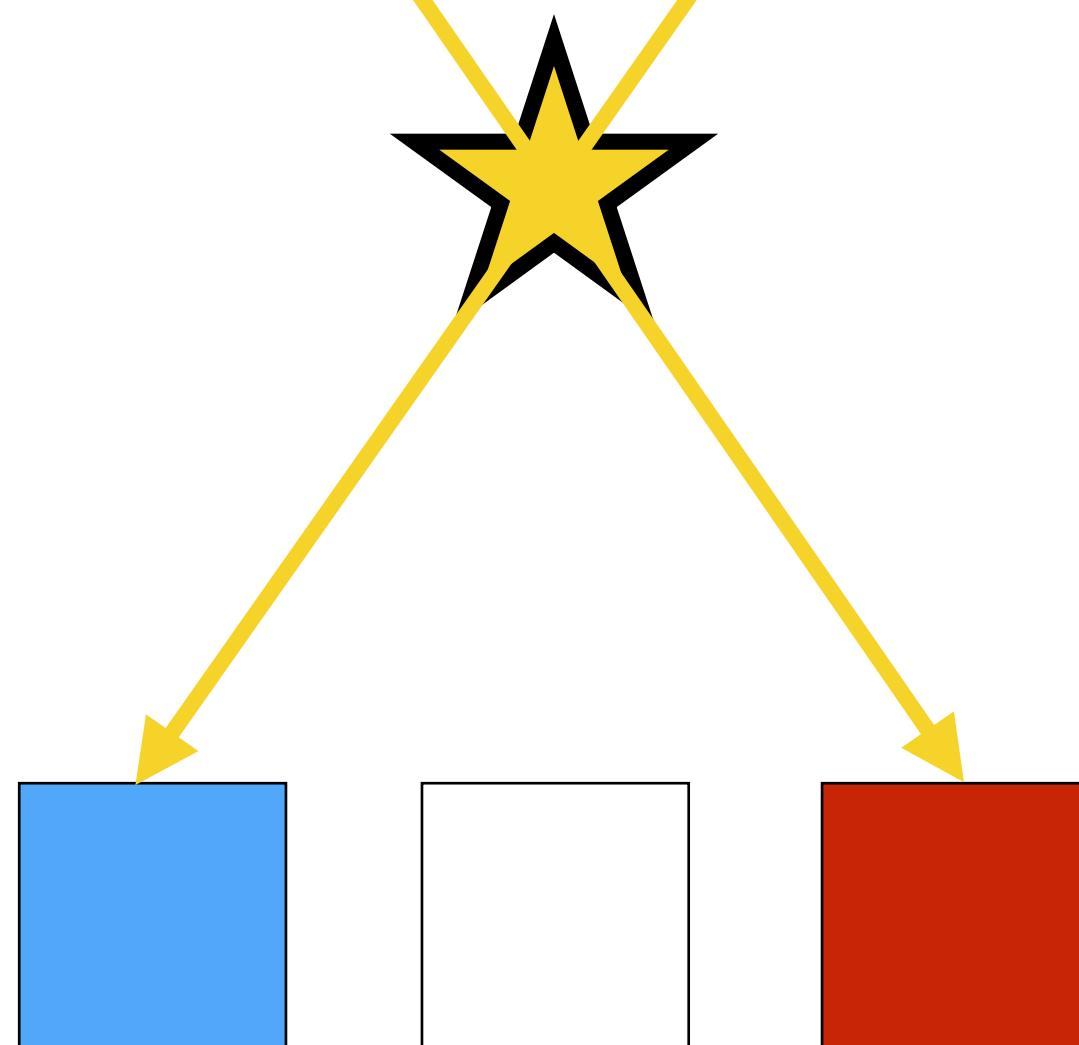
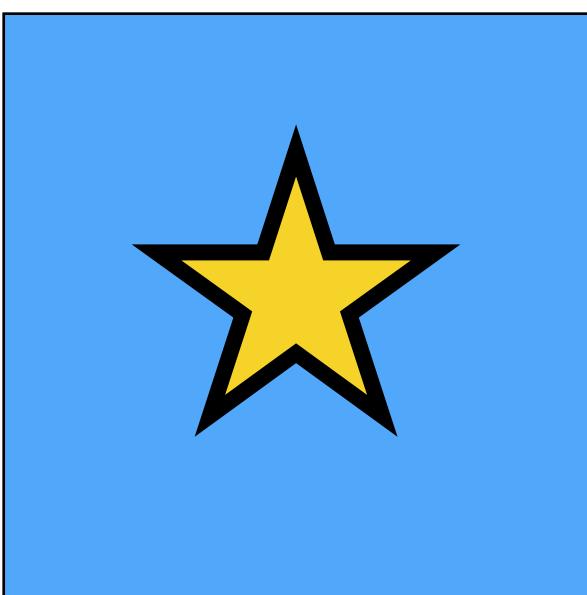


Slide courtesy: Bob Collins

Parallax

Points at different depths displace differently

Nearby points displace more than far ones



Depth via Parallax



Parallax = apparent motion of scene features located at different distances

<http://4hdwallpapers.com/wp-content/uploads/2013/03/California-Mountain.jpg>

Depth via Parallax



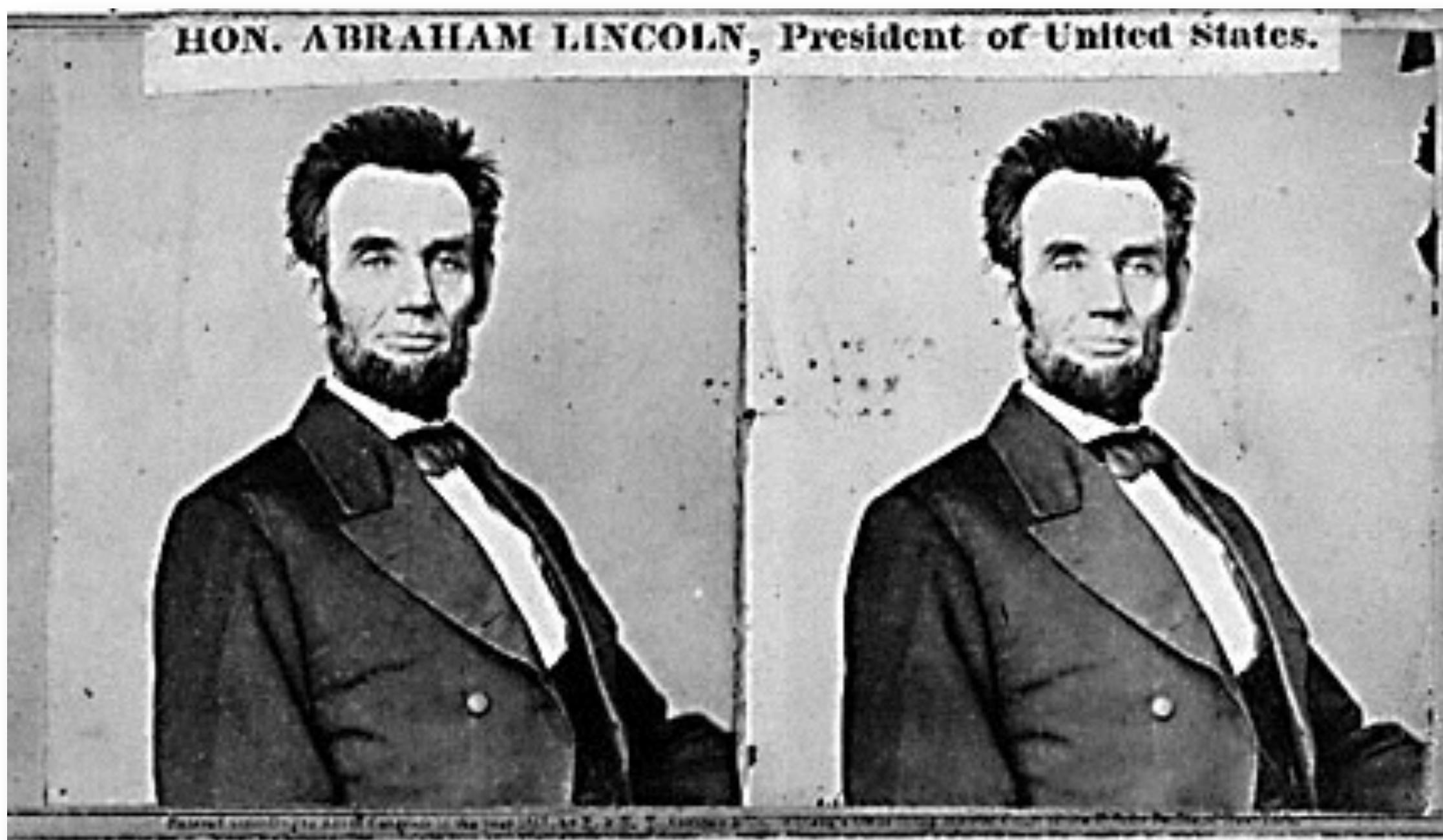
Stereo Photography and Stereo Viewers

Invented by Sir Charles
Wheatstone 1838



Take two pictures of the same subject from two slightly different viewpoints and display so that each eye sees only one of the images

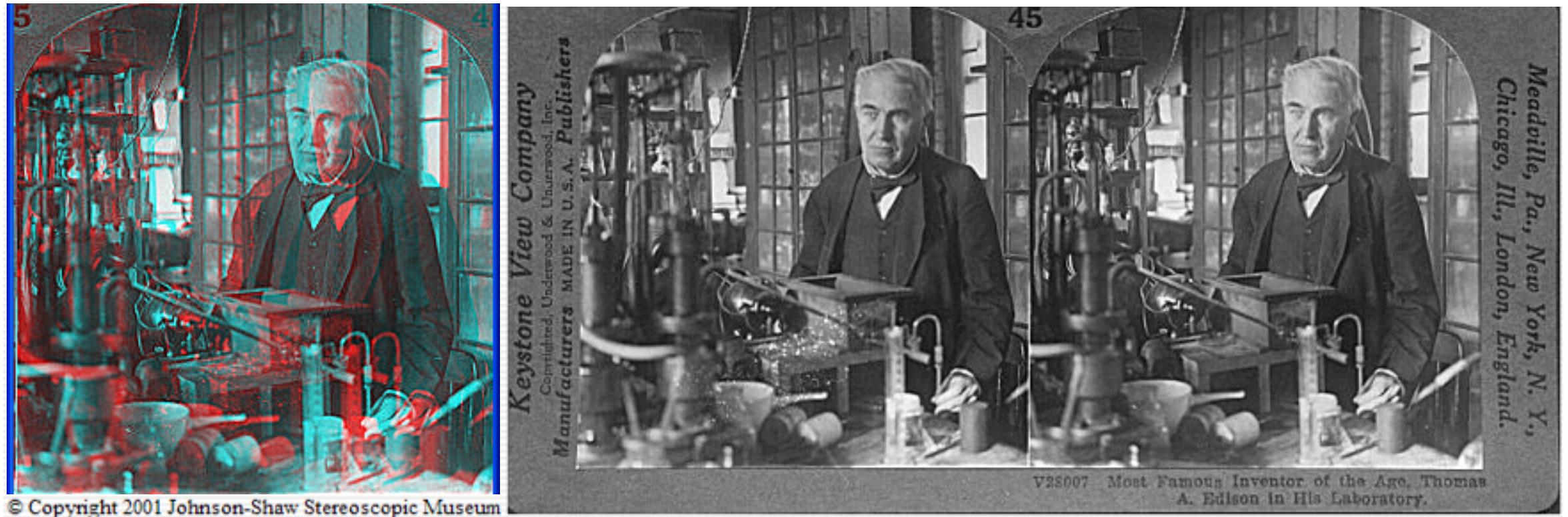
Early 3D photographs



Anaglyph



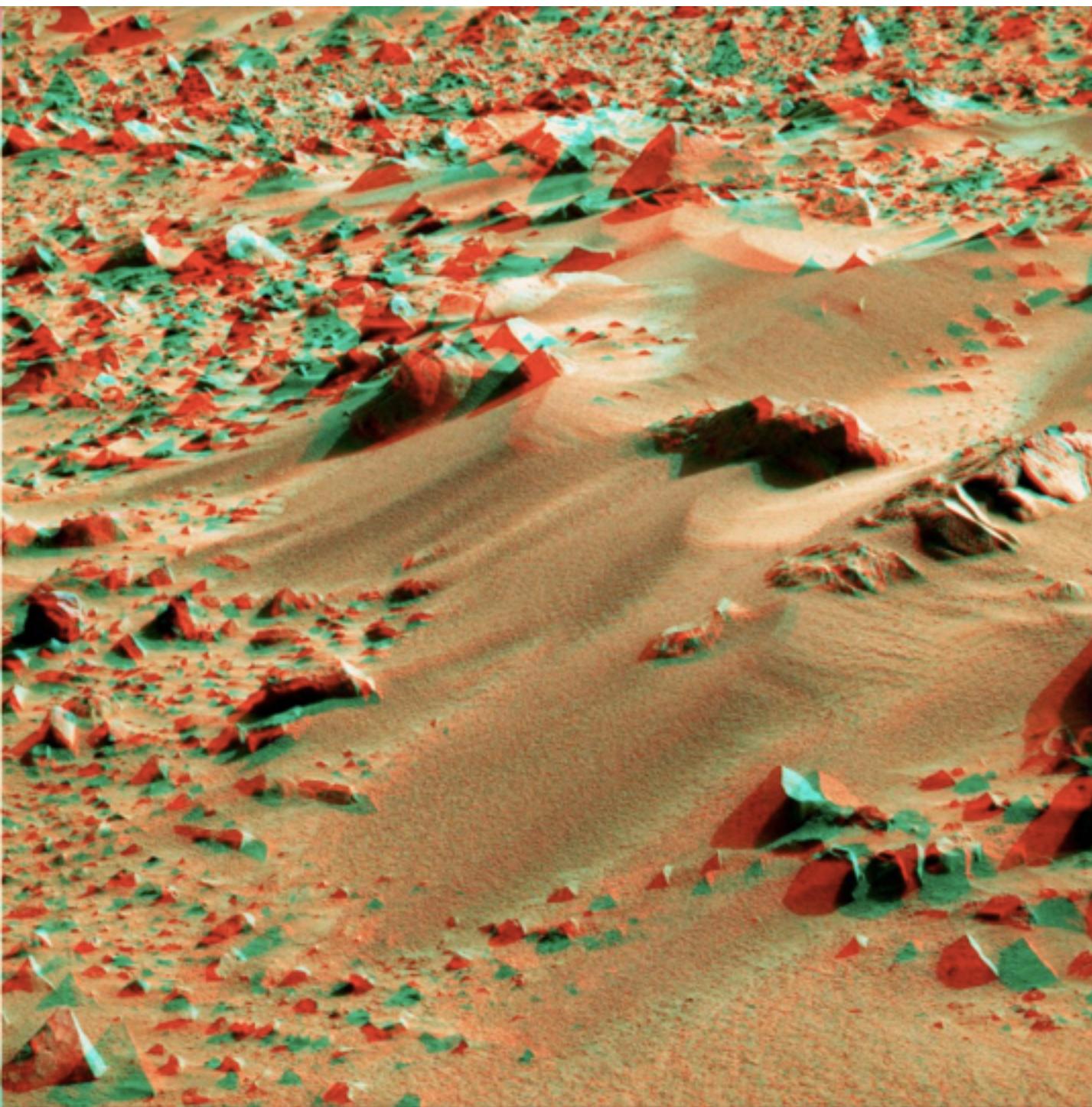
Put red filter
over left eye



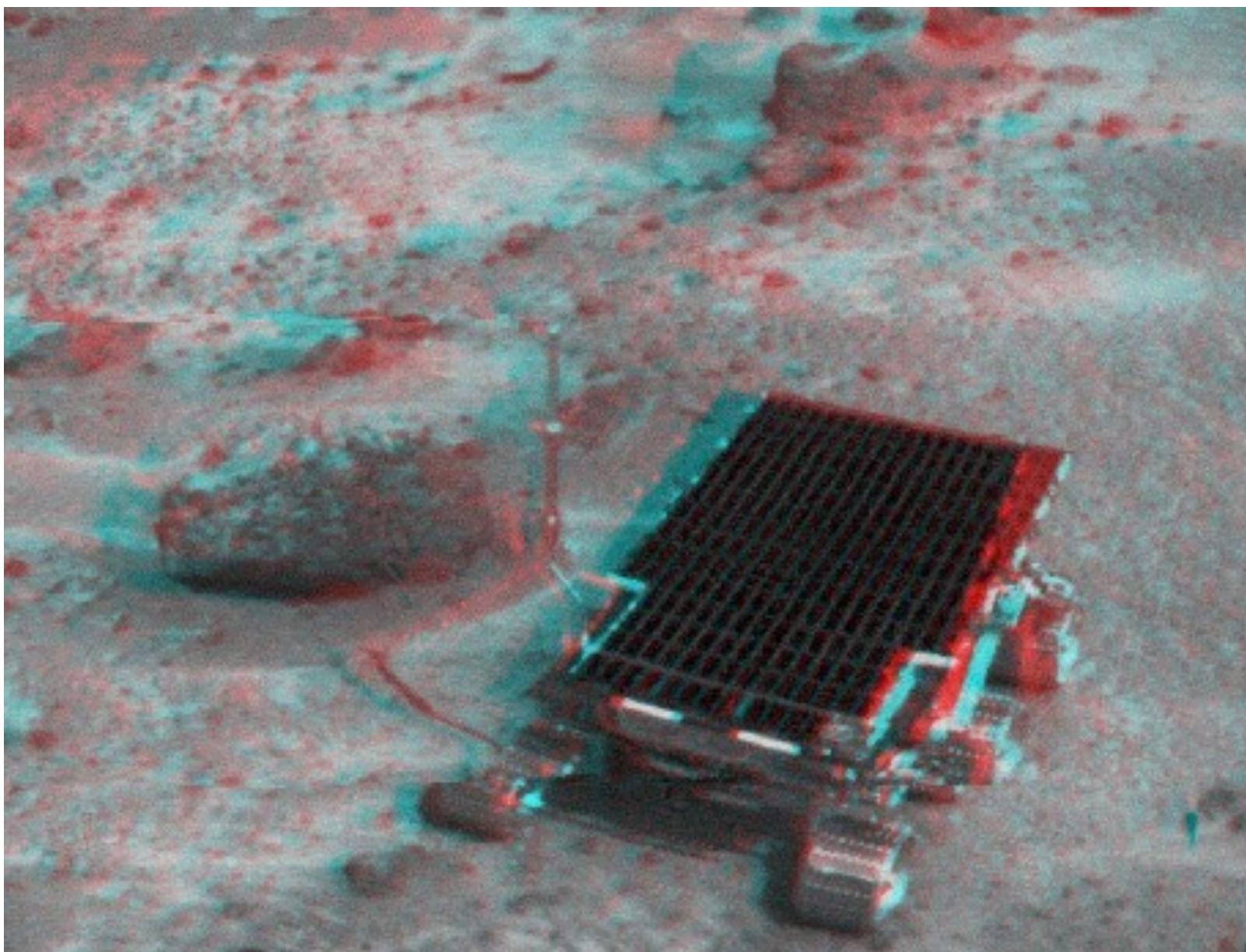
Anaglyphs encode parallax in a single picture.

Two slightly different perspectives of the same subject are superimposed on each other in contrasting colors, producing a three-dimensional effect when viewed through two correspondingly colored filters

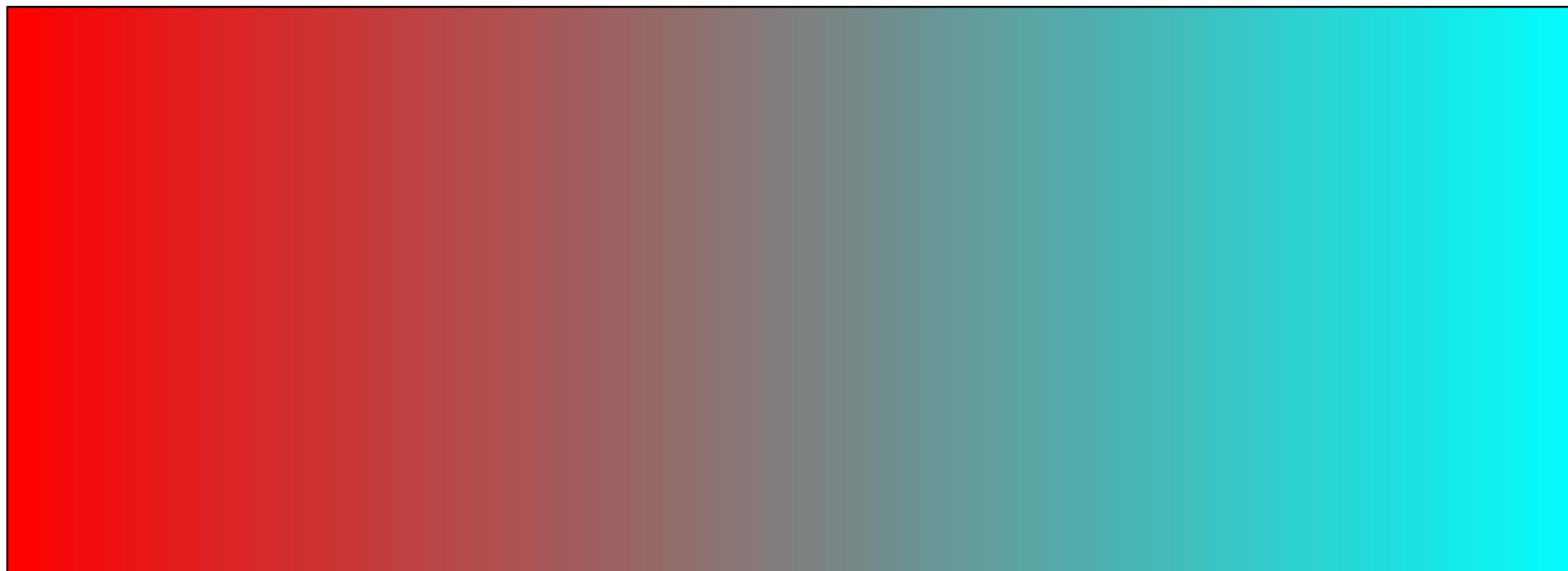
Anaglyph



Anaglyph



How Anaglyphs Work



- * Close right eye, then close left. What do you observe?
- * Red filter selectively passes red color, and similarly for cyan filter and cyan color.

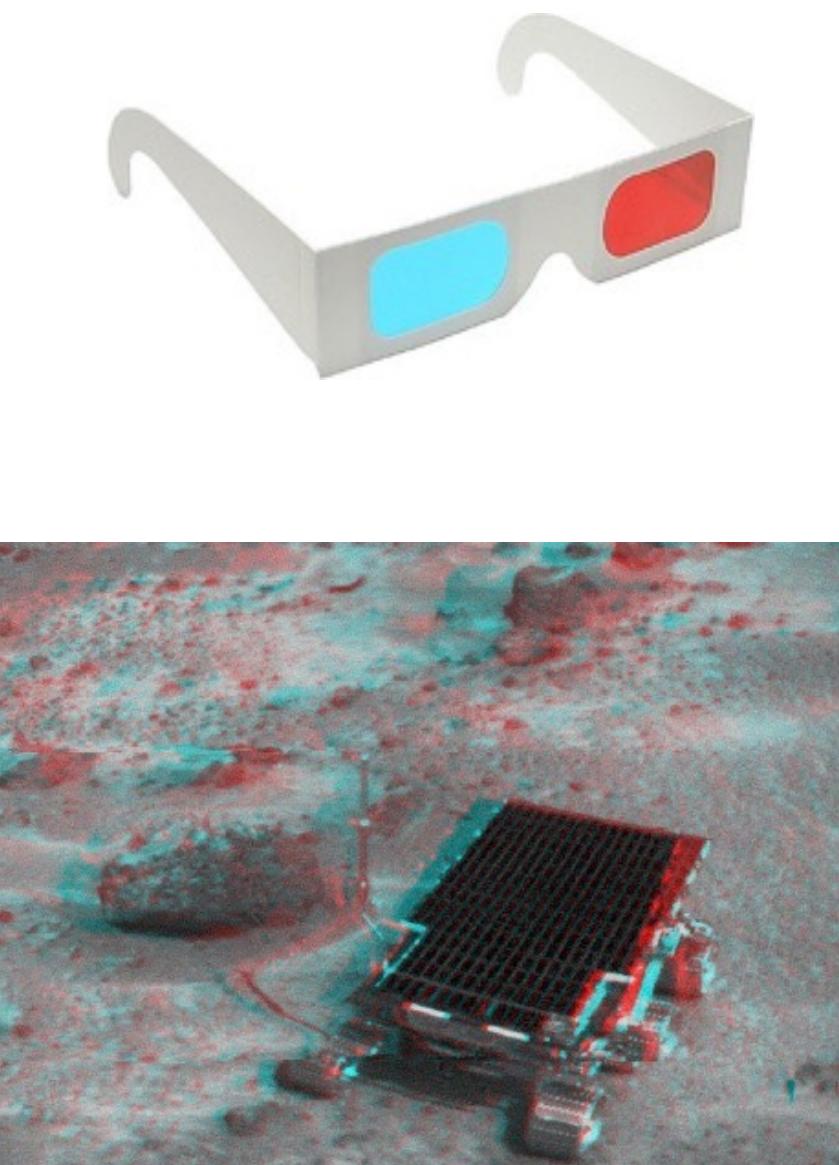


Make an Anaglyph Image

- * Given: Grayscale images `img_left`, `img_right`
- * Output: Red/cyan anaglyph image

Making an Anaglyph

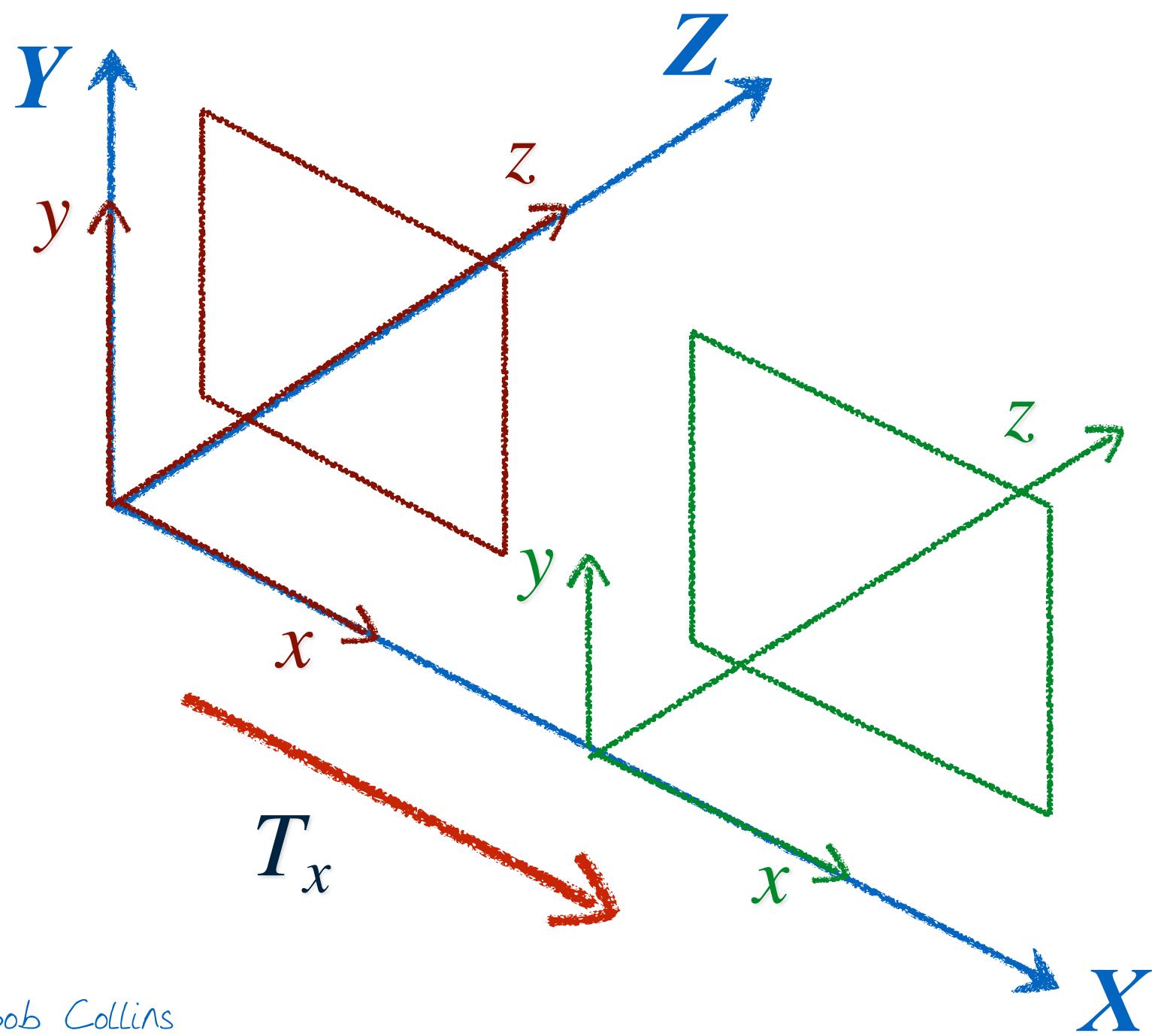
- * Take a greyscale stereo pair
- * Copy the left image to the red channel of a new image (the anaglyph image)
- * Copy the right image to the green and blue channels of the anaglyph image (note: green+blue = cyan)
- * View with red-cyan glasses,
 - * left eye sees only the left image,
 - * and the right eye sees only the right image
 - * The brain fuses to form 3D.



A Simple Stereo System

left
camera
located at
 $(0,0,0)$

right
camera
located at
 $(T_x,0,0)$

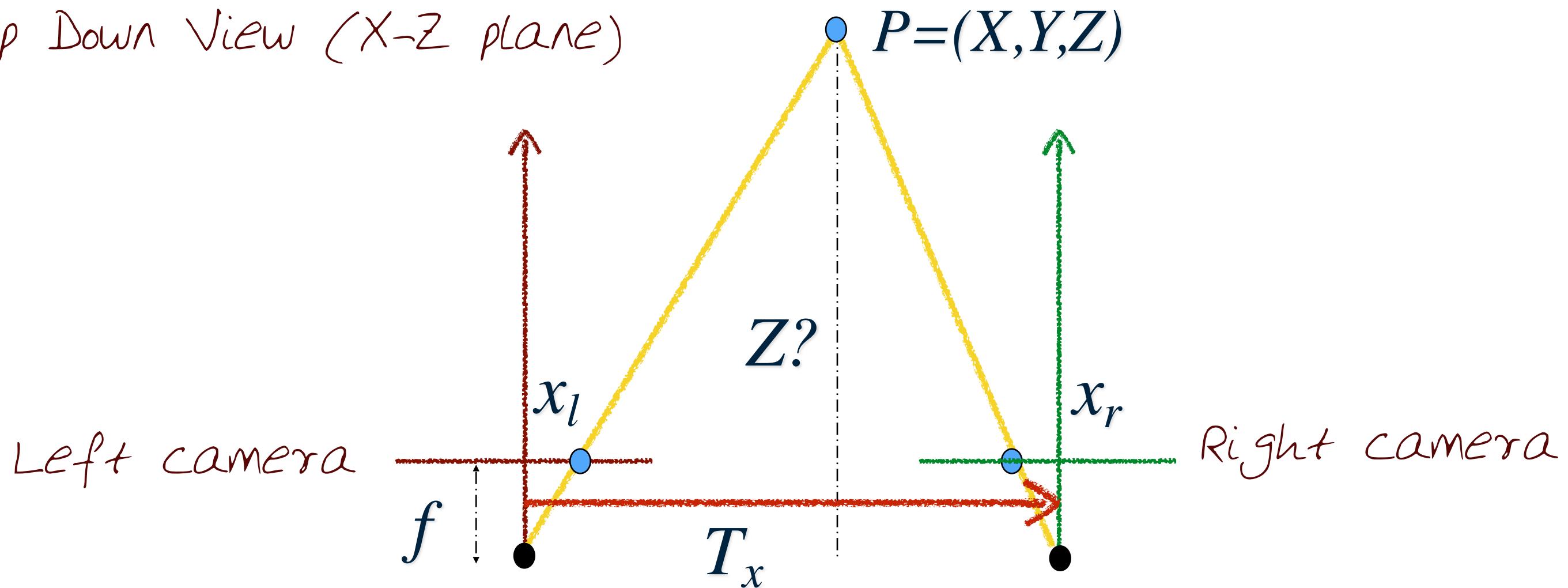


Right camera is simply
shifted by T_x units
along the X axis.
Otherwise, the
cameras
are identical (same
orientation / focal
lengths)

Slide adapted from Bob Collins

A Simple Stereo System

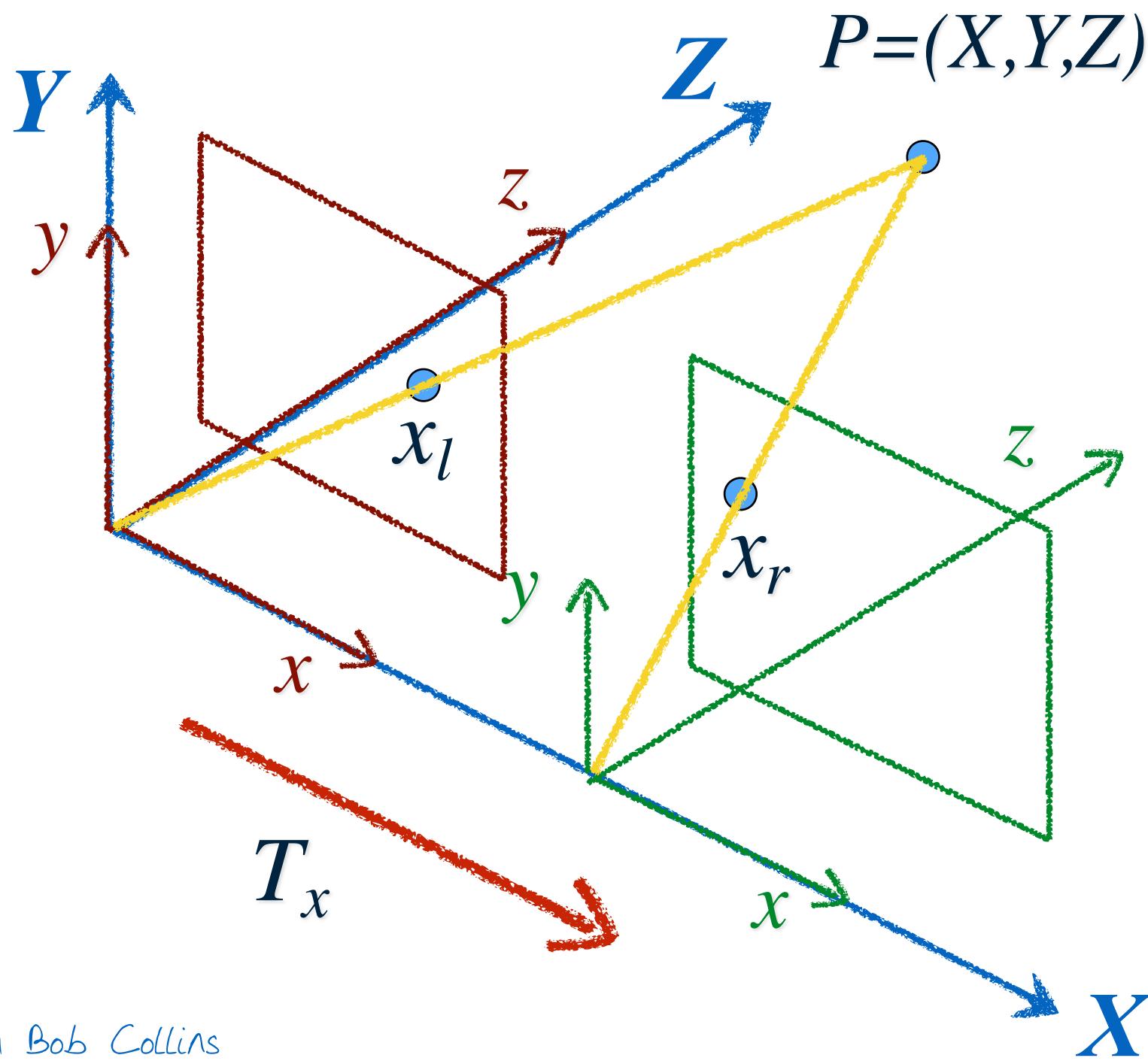
Top Down View (X-Z plane)



Translated by a distance T_x along X axis
(T_x is also called the stereo "baseline")

A Simple Stereo System

left camera located at $(0,0,0)$
right camera located at $(T_x, 0, 0)$



Point P in Left Camera:

$$x_l = f \frac{X}{Z}$$

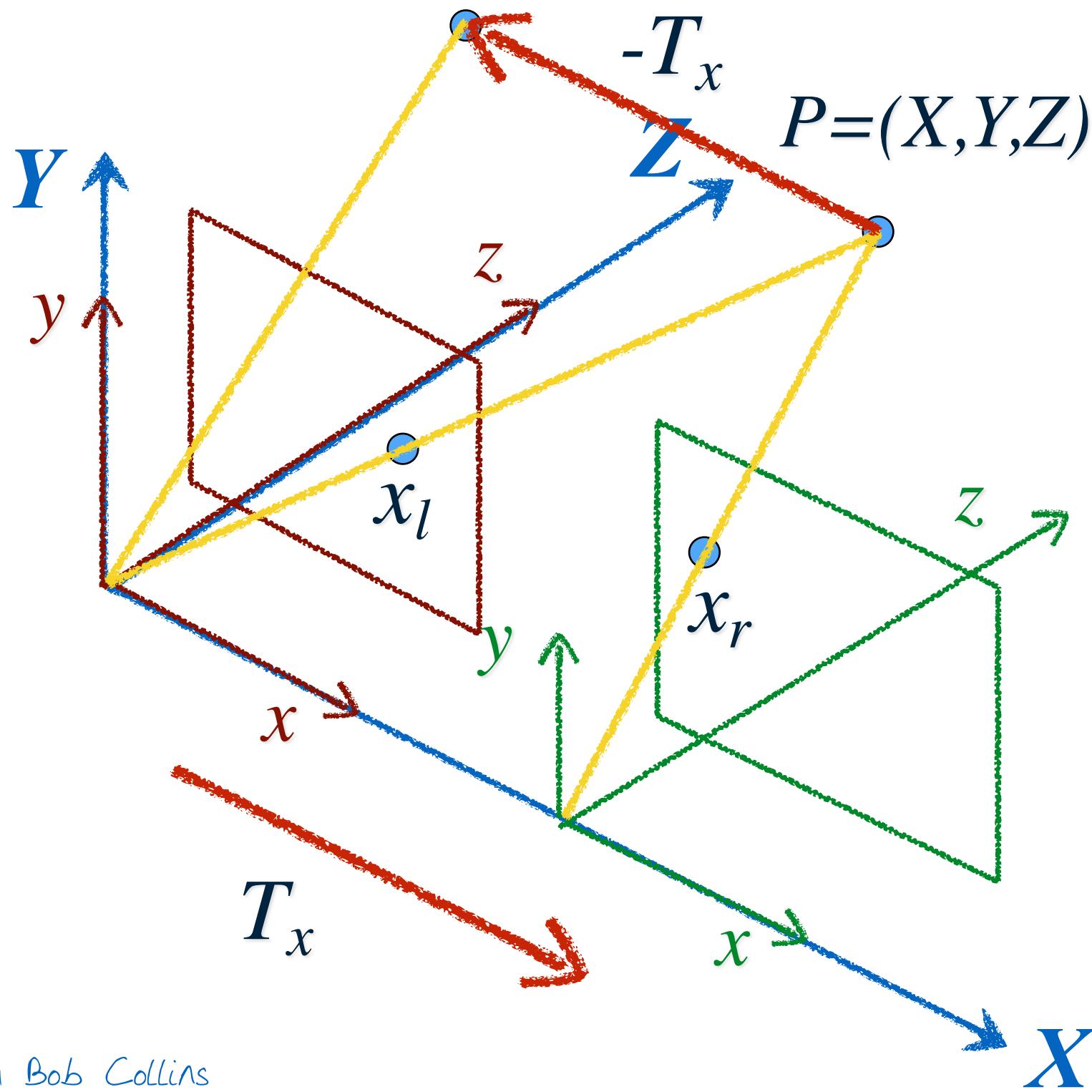
$$y_l = f \frac{Y}{Z}$$

The same point
in the Right Camera?

Slide adapted from Bob Collins

A Simple Stereo System

left camera located at $(0,0,0)$
right camera located at $(T_x,0,0)$



Point P in Left Camera:

$$x_l = f \frac{X}{Z}$$

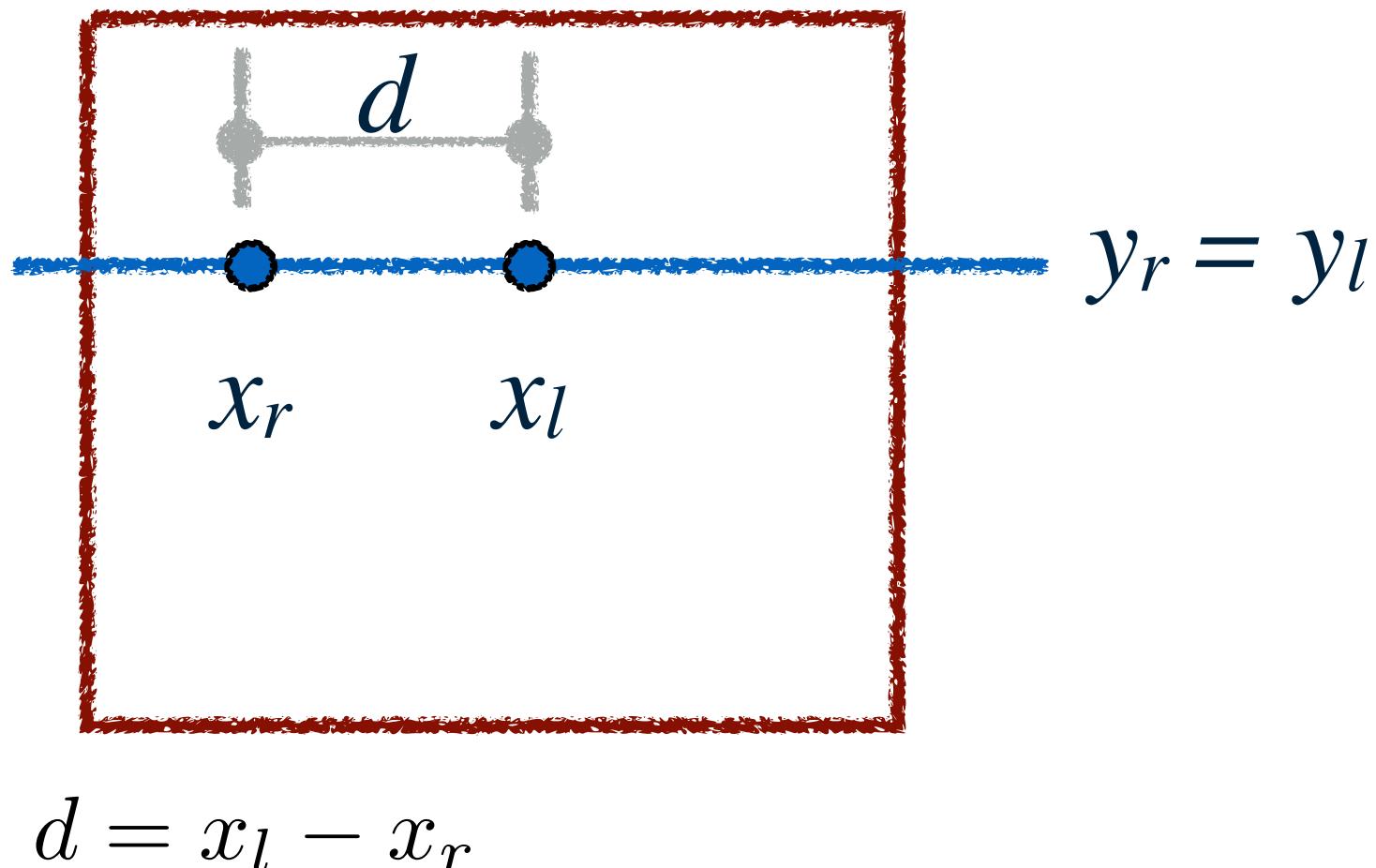
$$y_l = f \frac{Y}{Z}$$

The same point
in the Right Camera?

$$x_r = f \frac{X - T_x}{Z} \quad y_r = f \frac{Y}{Z}$$

Slide adapted from Bob Collins

Stereo Disparity



$$d = x_l - x_r$$

$$d = f \frac{X}{Z} - f \frac{X - T_x}{Z}$$

depth

left camera

$$x_l = f \frac{X}{Z} \quad y_l = f \frac{Y}{Z}$$

right camera

$$x_r = f \frac{X - T_x}{Z} \quad y_r = f \frac{Y}{Z}$$

baseline

$$Z = f \frac{T_x}{d}$$

disparity

Stereo Example



Left Image



Right Image

From middlebury stereo evaluation page <http://www.middlebury.edu/stereo/>

Stereo Example



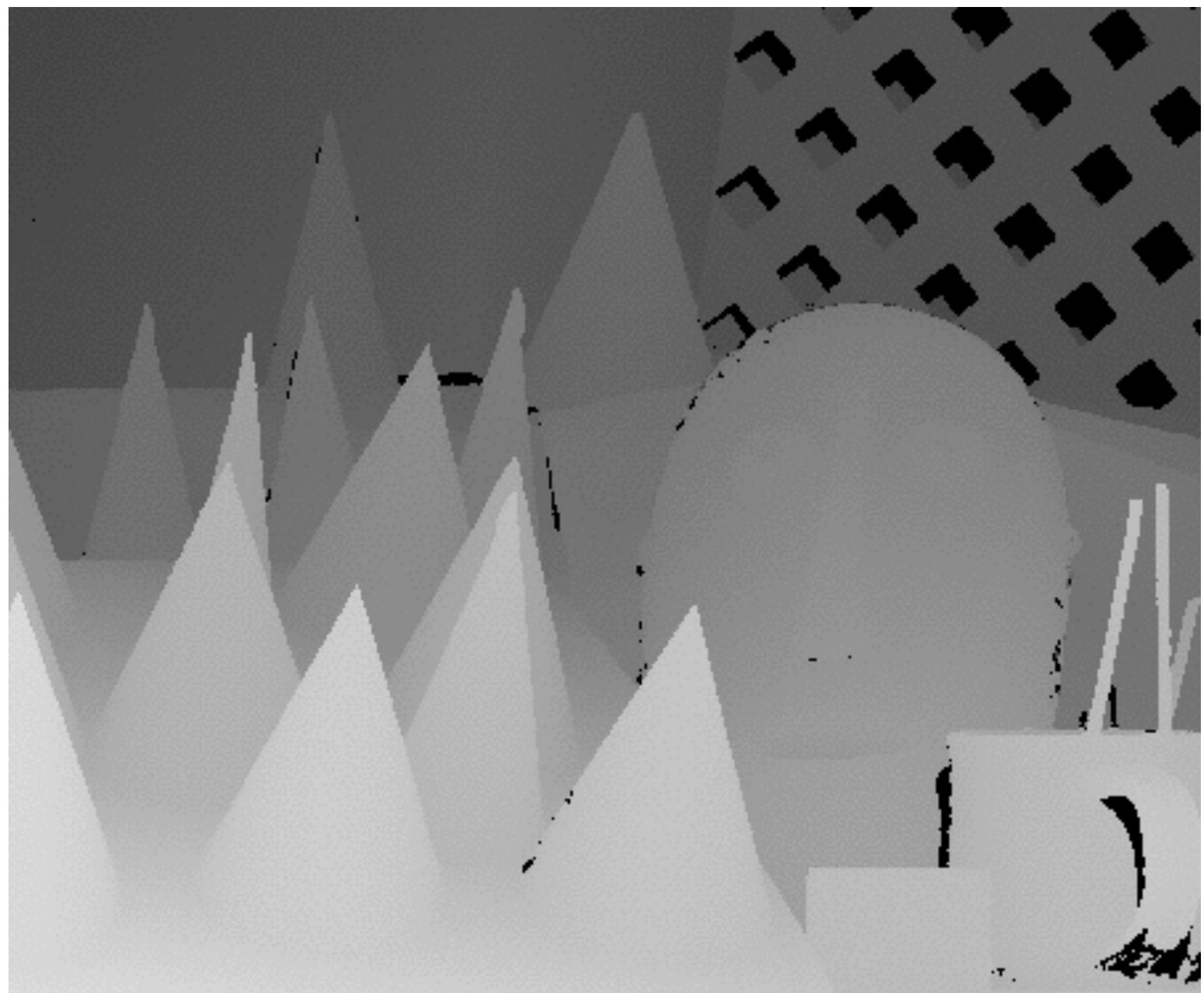
Left
image



Right
image

Slide adapted from Bob Collins

Disparity values (0-64)

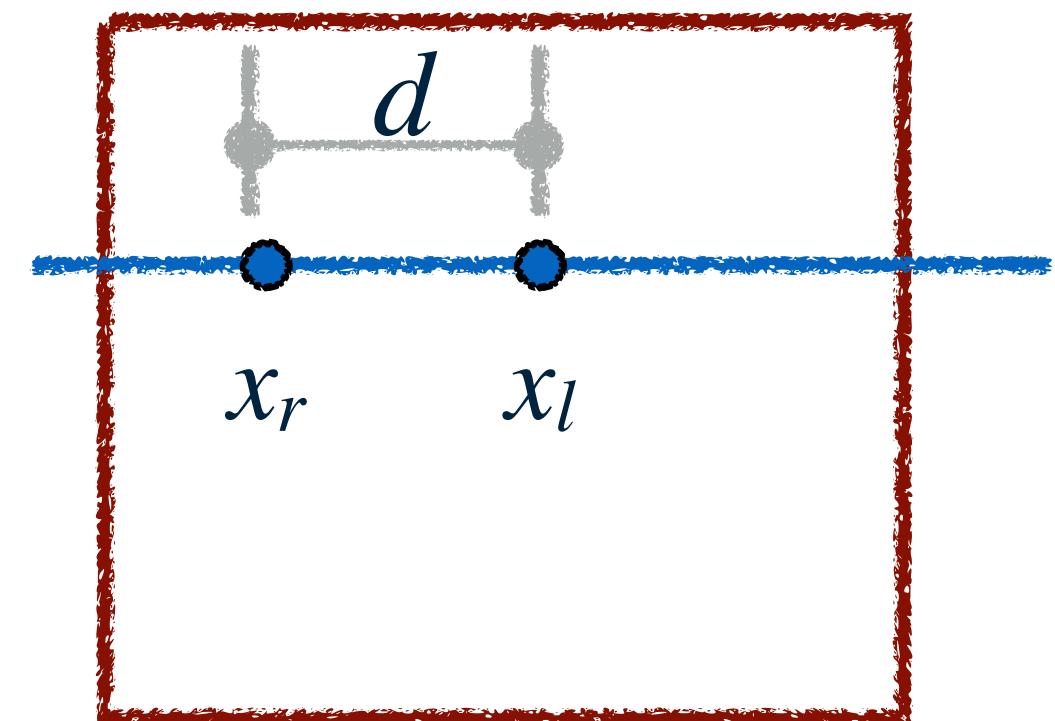


Disparity is larger for closer surfaces

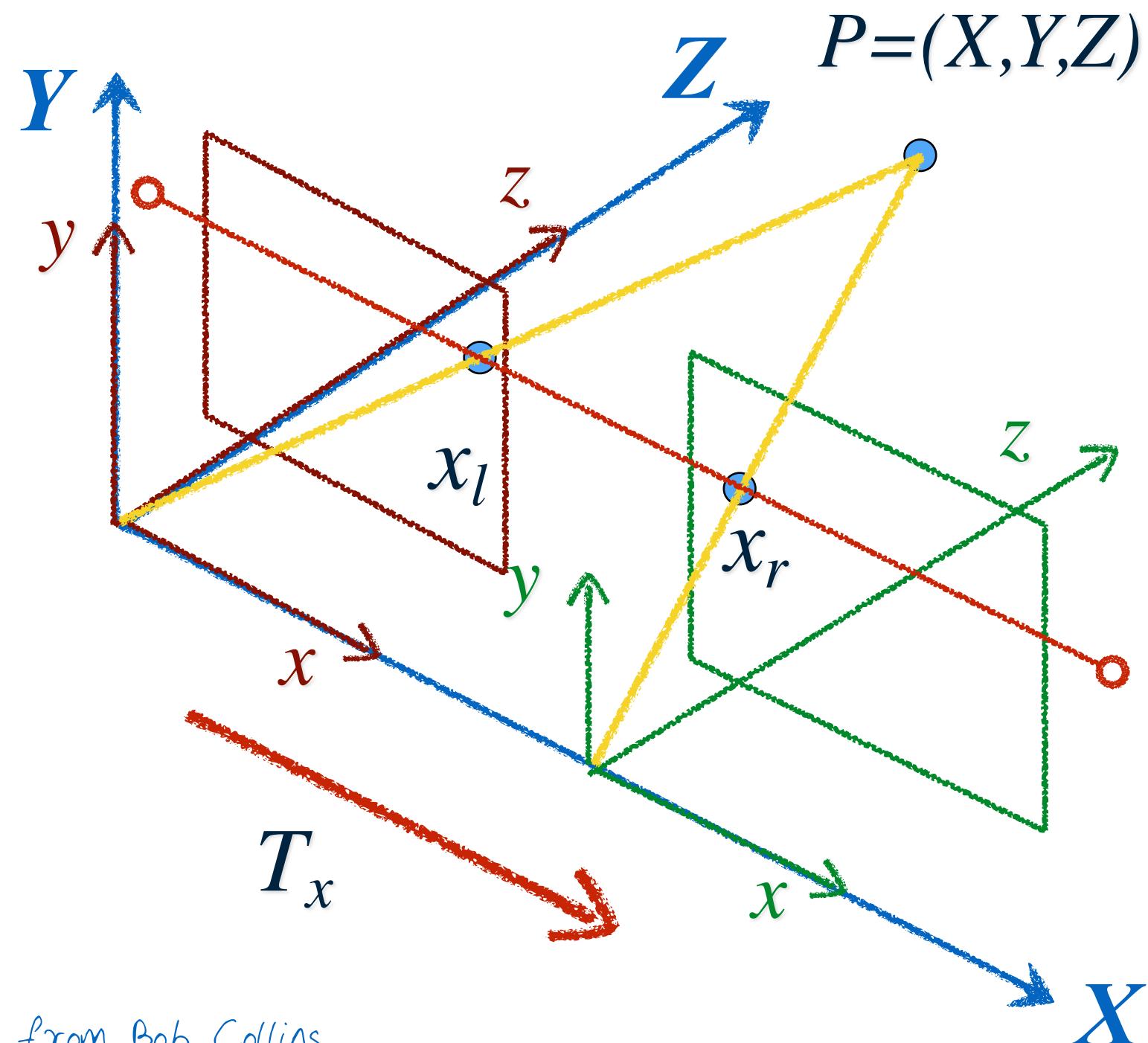
Computing Disparity

$$d = x_l - x_r$$

- * Correspondence
 - * Which pixel in the right image corresponds to each pixel in the left
- * Epipolar Constraint
 - * No need to search the whole 2D right image
 - * The epipolar constraint reduces the search space to a one-dimensional line



Recall: A Simple Stereo System



left camera

$$x_l = f \frac{X}{Z} \quad y_l = f \frac{Y}{Z}$$

right camera

$$x_r = f \frac{X - T_x}{Z} \quad y_r = f \frac{Y}{Z}$$

Slide adapted from Bob Collins

Matching Using Epipolar Lines

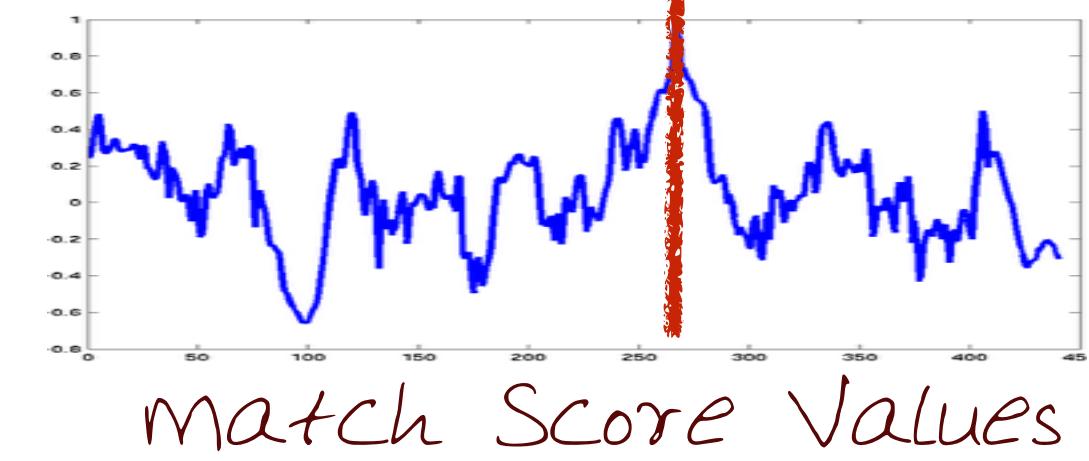
Left
Image



Right
Image

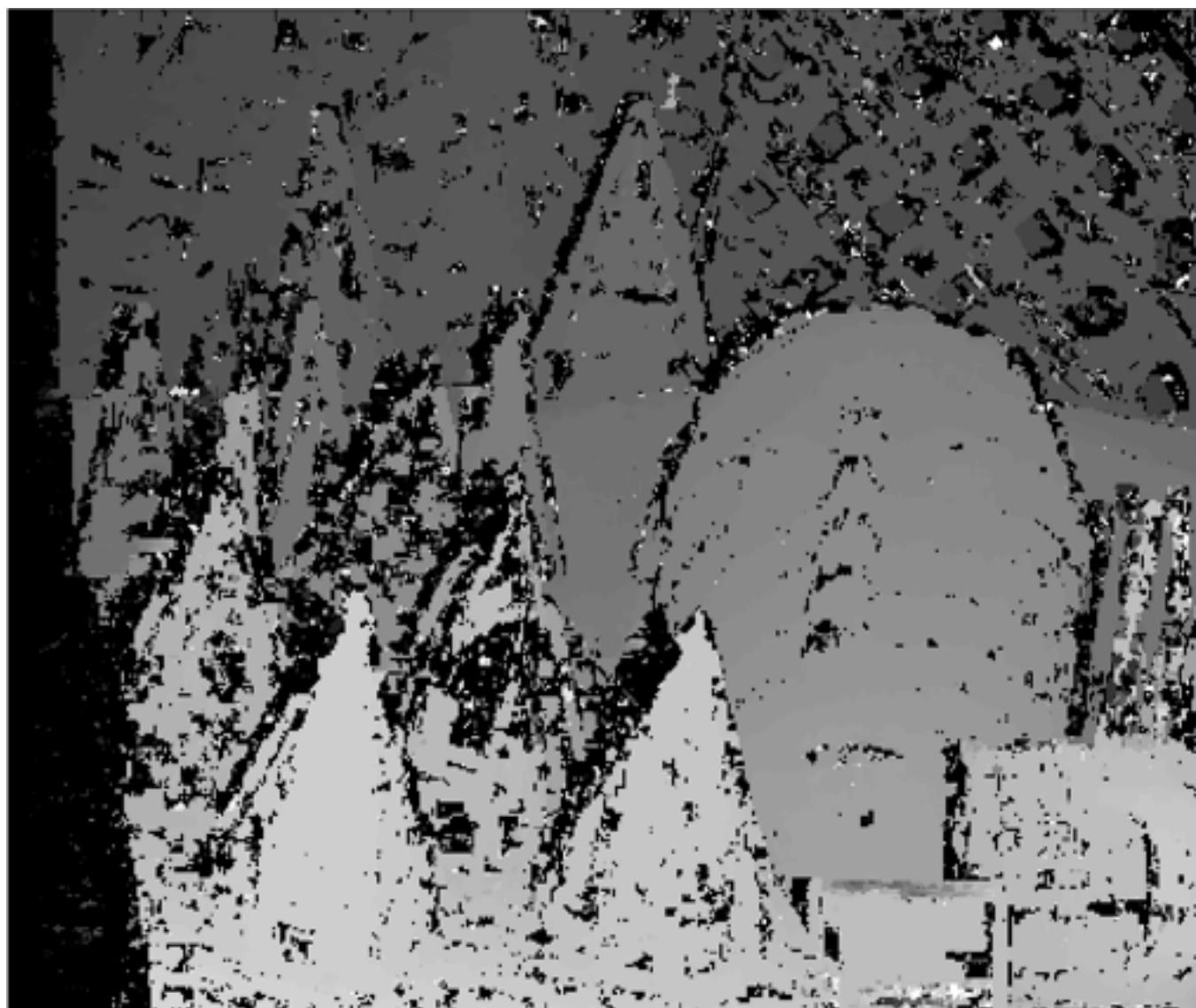


For a patch in left image
Compare with patches along
same row in right image

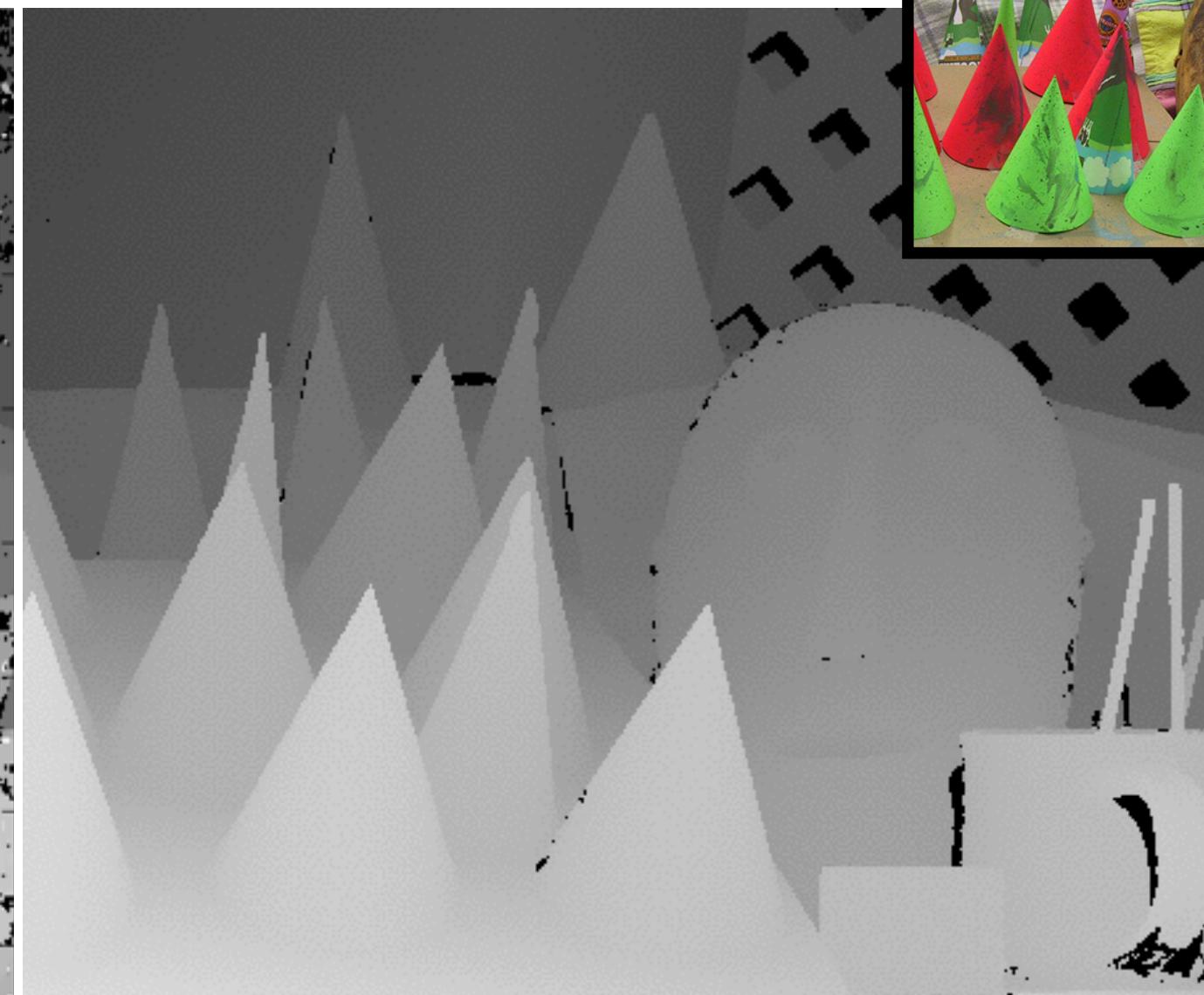


match Score Values

No Matches



Computed disparities

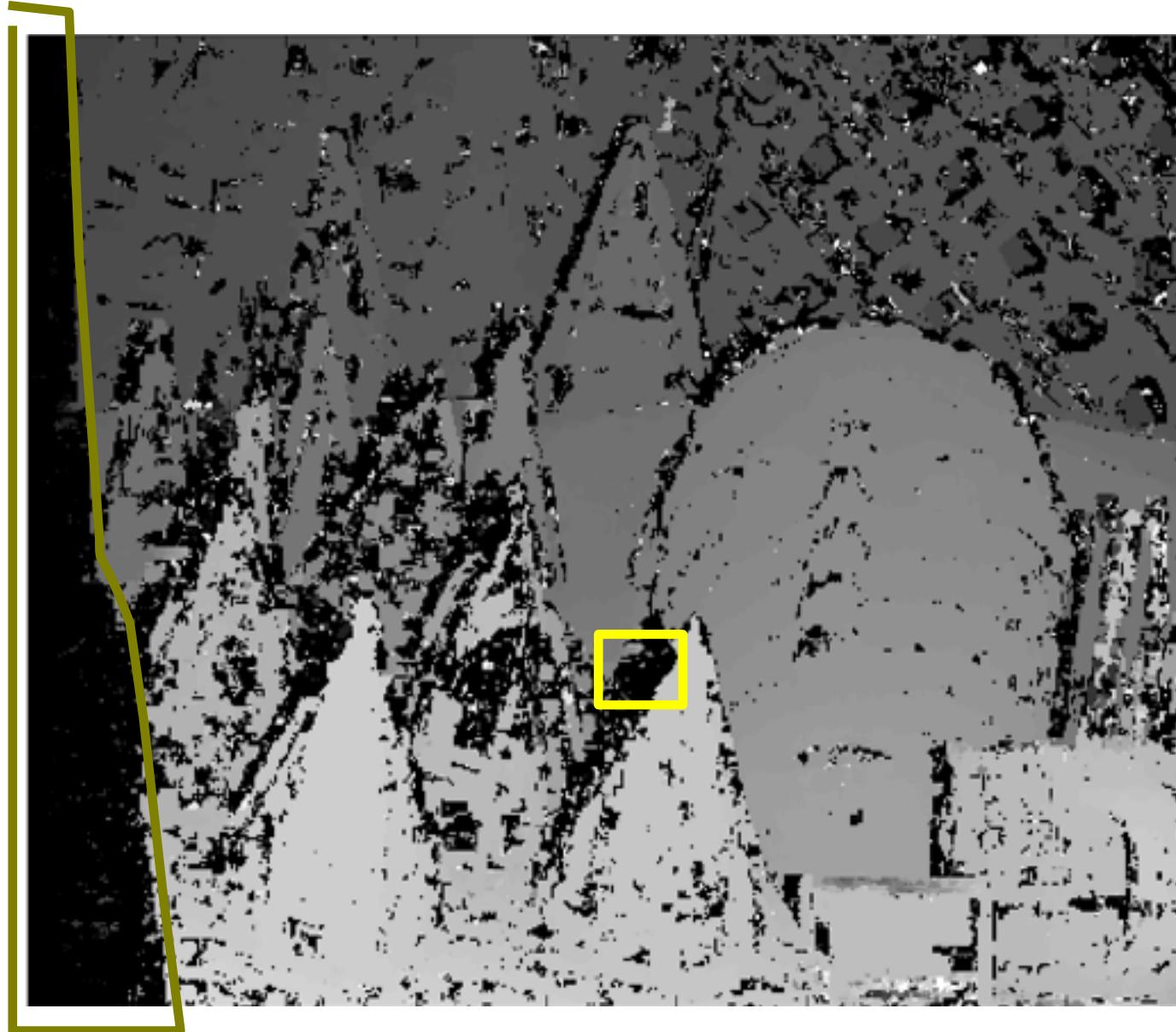


Ground truth

Black pixels: bad disparity values, or no matching patch in right image



Occlusions: No Matches



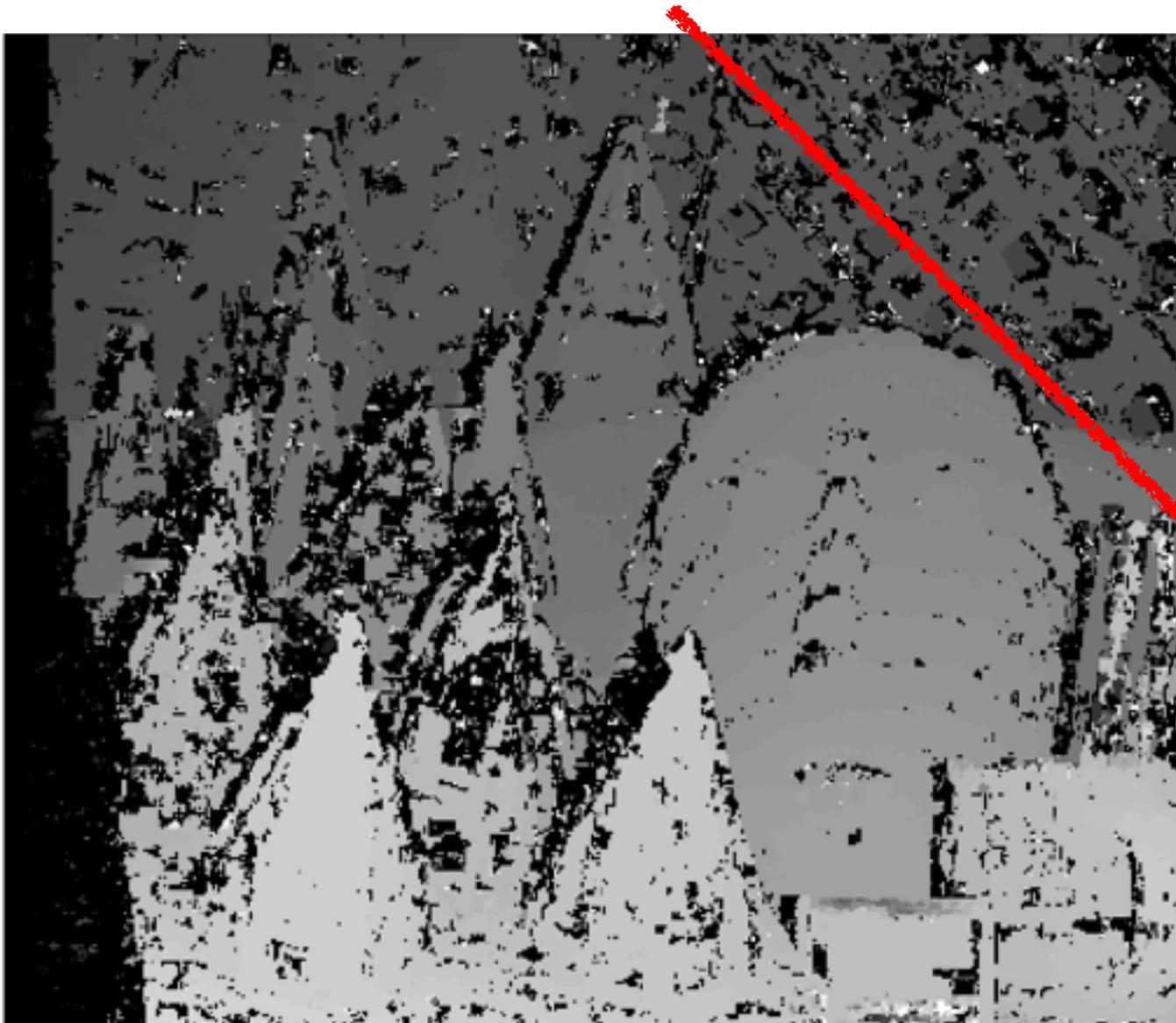
Left
Image



Right
Image

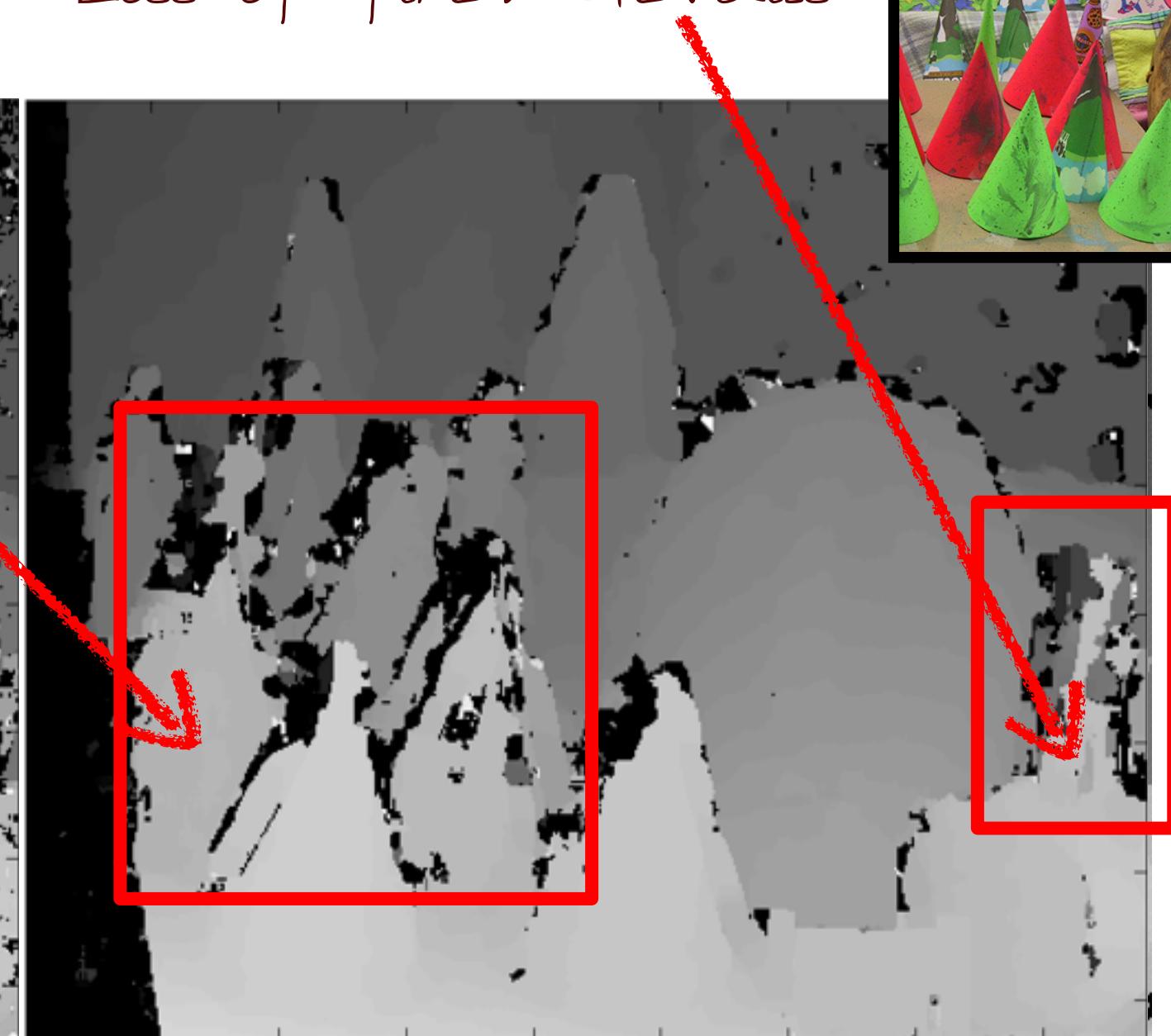
Effects of Patch Size

Smoother in some areas



5×5 patches

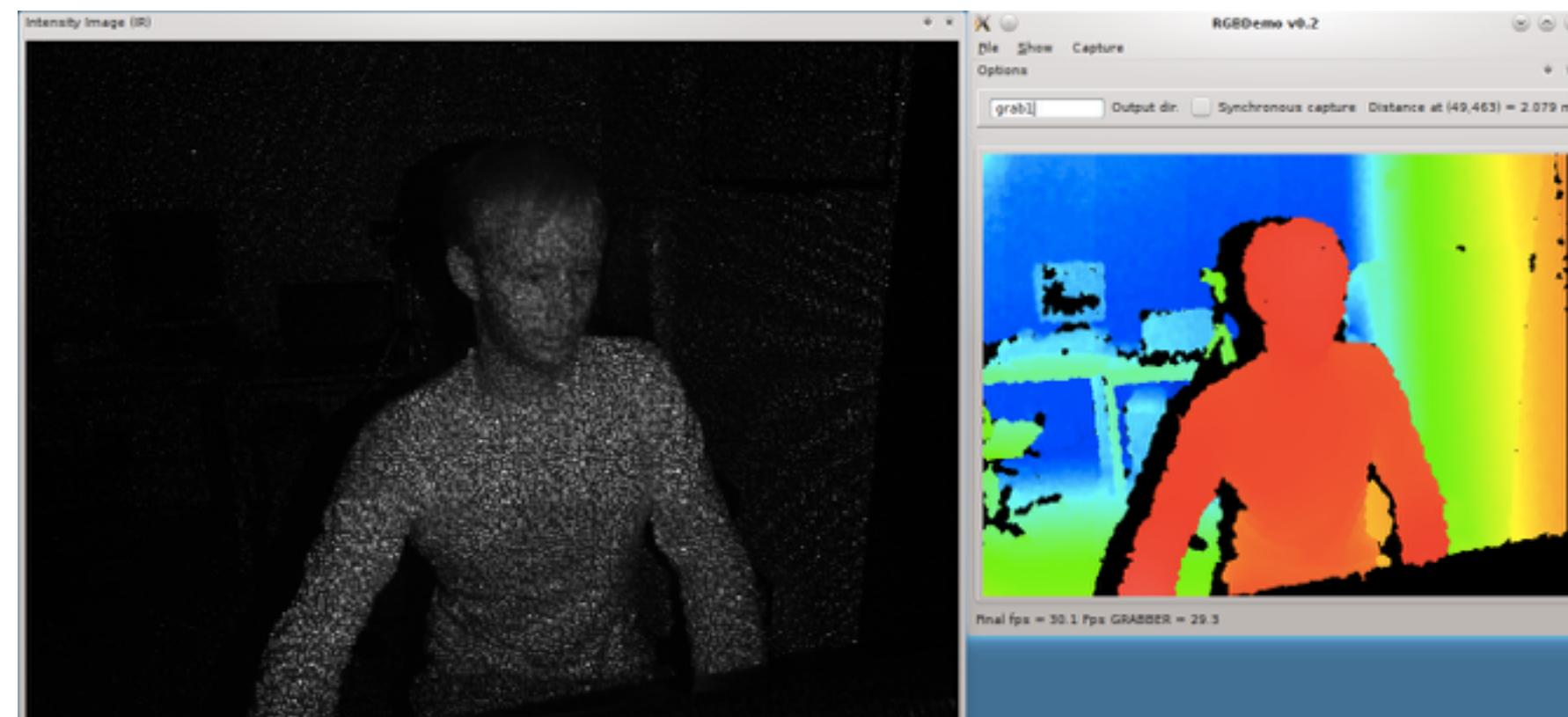
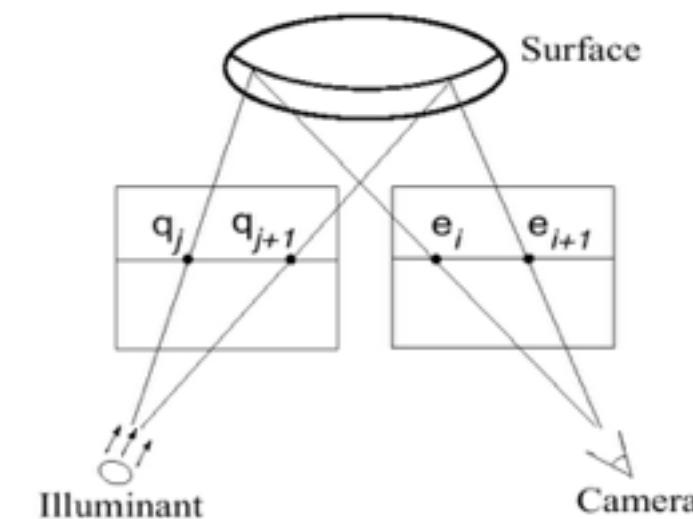
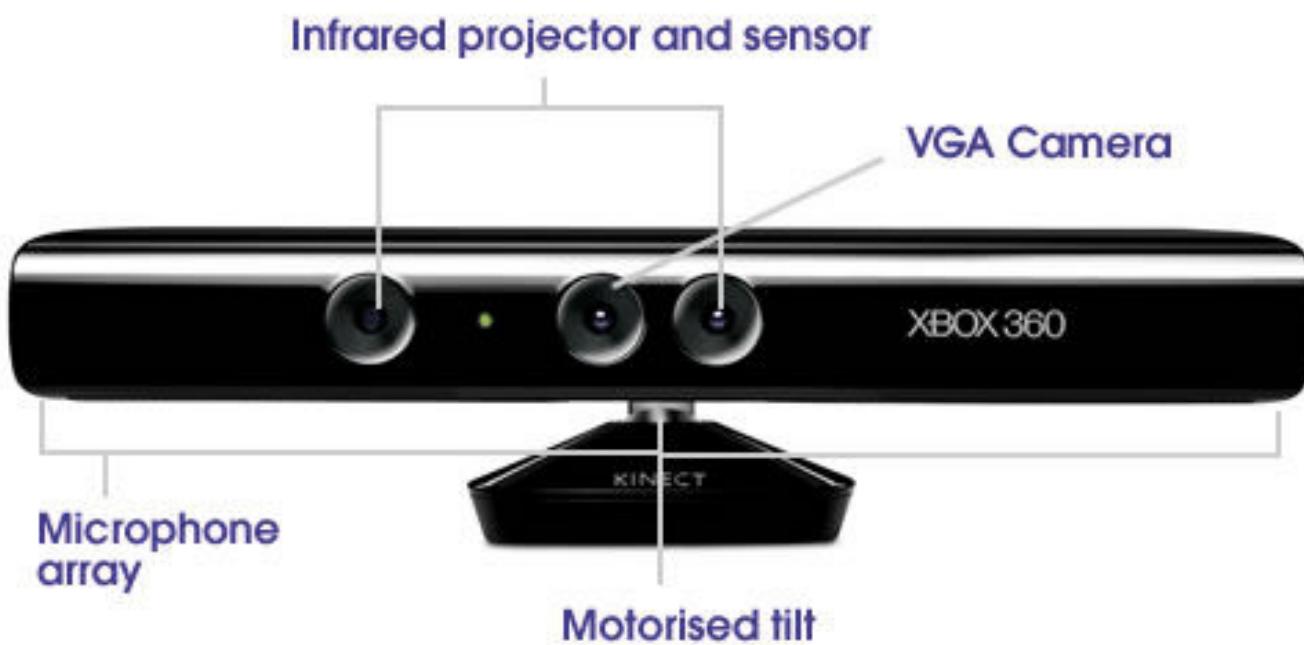
Loss of finer details

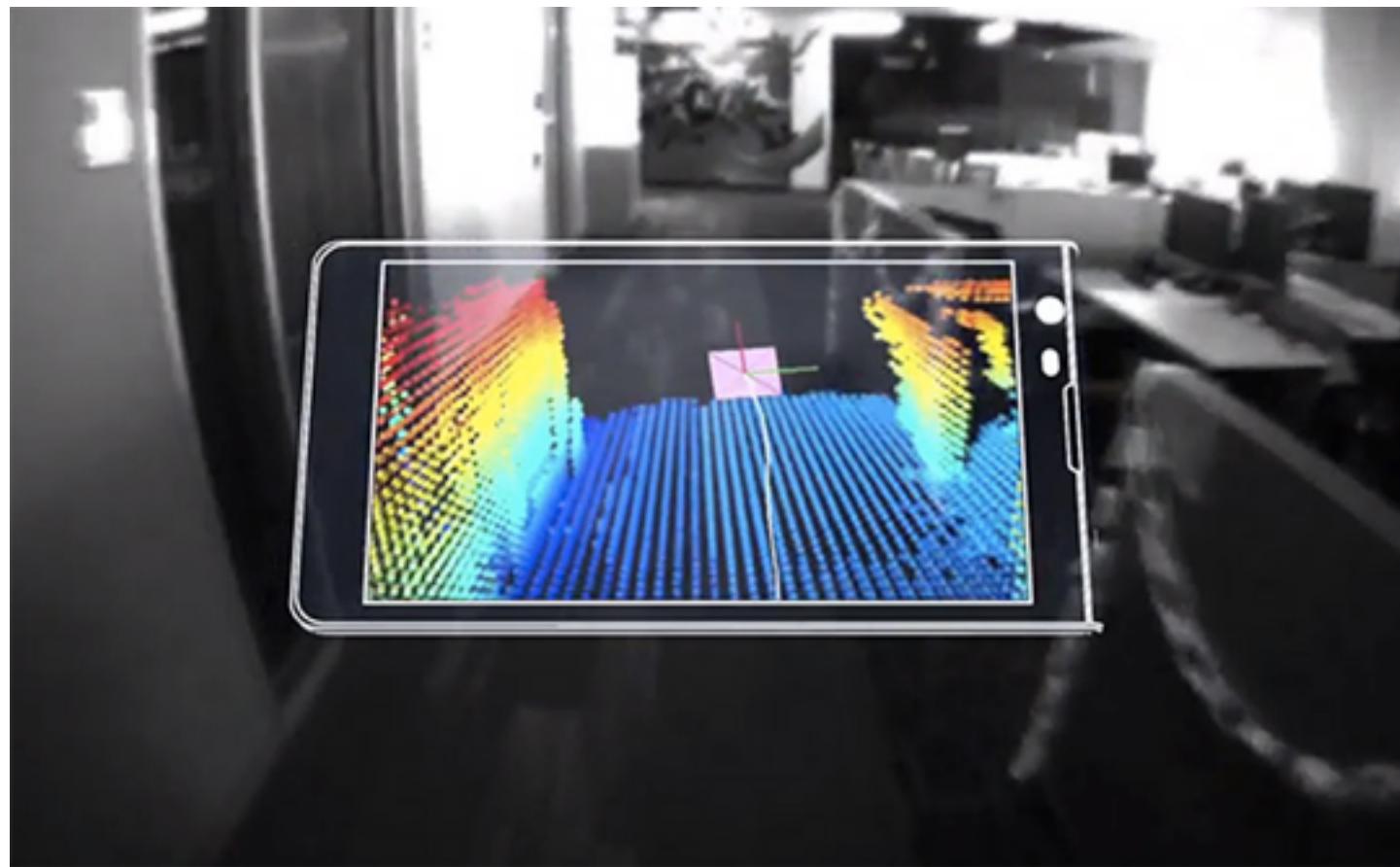


11×11 patches



Some well known RGBD Cameras





Google Project Tango



Amazon Fire
5 Cameras



Processed Image From Kinect



Image via <http://blogs.msdn.com>

zugara

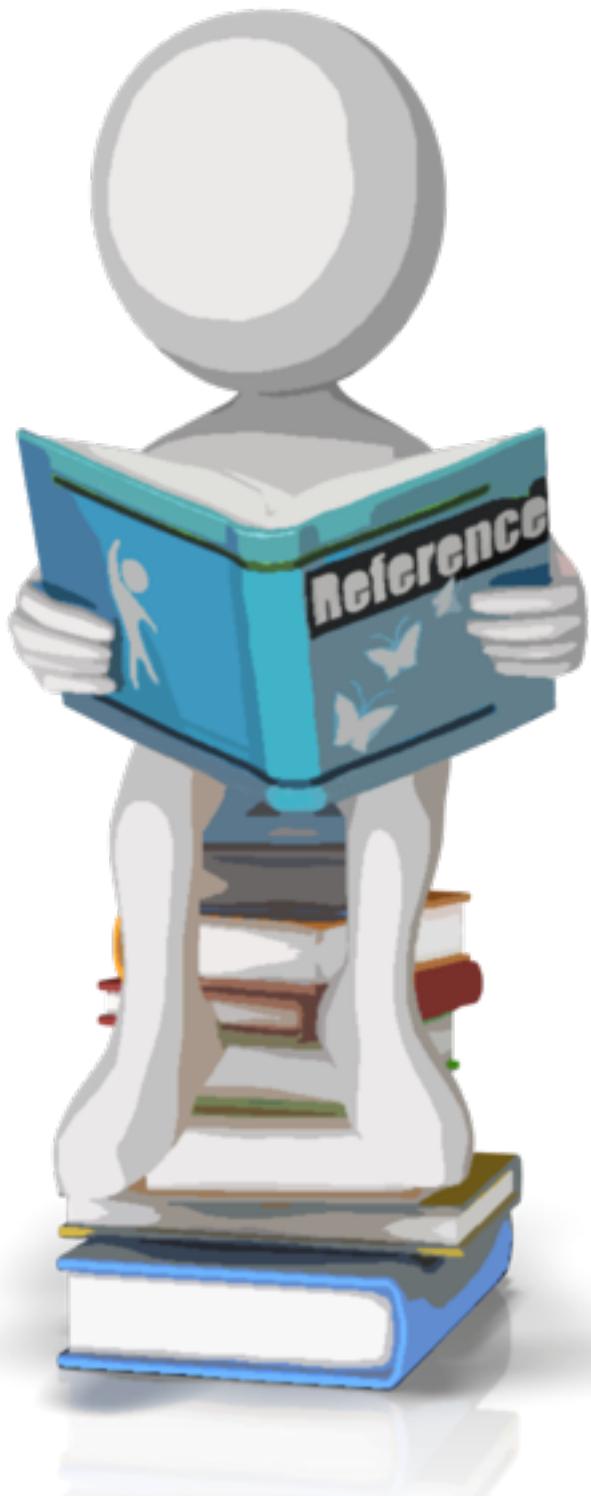
Kinect 1 to Kinect 2: Time of Flight Camera

Summary



- * Shape/Structure/
Geometry from X
methods
- * Stereo Imaging
- * Compute depth from a
Stereo image pair

Further Reading



- * McMillan & Gortler (1999) "Image-Based Rendering: A New Interface Between Computer Vision and Computer Graphics" Applications of Computer Vision to Computer Graphics Vol 33, No 2 1999 [[Link](#)]
- * Szeliski (2010)
- * Hartley & Zisserman (2004)



To-Do List

- Rewrite
- Finish
- Write
- Write
- Make
- Move
- Rewrite AIs
- Finish grad
- Write rubri
- Write feat
- Make anna
- Move
- Move tes

Credits



- * For more information, see:
 - * Richard Szeliski (2010) Computer Vision: Algorithms and Applications, Springer (Chapter 3)
- * Some concepts in slides motivated by similar slides by James Hays, Alyosha Efros, Svetlana Lazebnik, Bob Collins, and Greg Turk
- * Grant Schindler,
 - * <http://www.trimensional.com/>
- * Additional list will be available on website

Computational Photography

- * Study the basics of computation and its impact on the entire workflow of photography, from capturing, manipulating and collaborating on, and sharing photographs.



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