# Stereo



Many slides adapted from Steve Seitz

 Given a calibrated binocular stereo pair, fuse it to produce a depth image

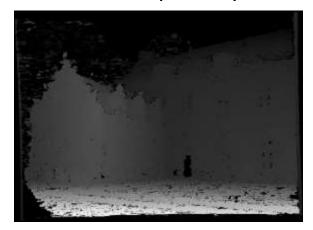
image 1



image 2



Dense depth map

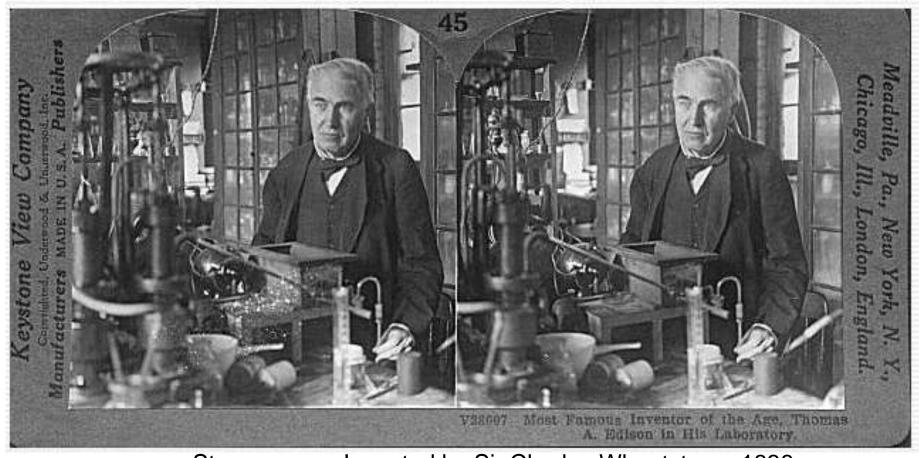


 Given a calibrated binocular stereo pair, fuse it to produce a depth image

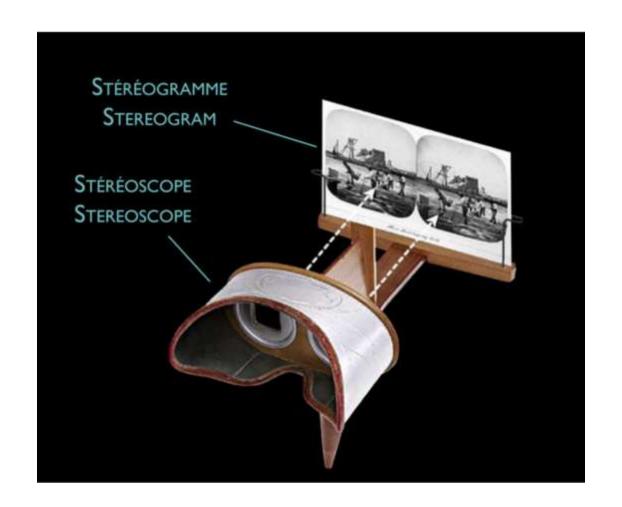


Where does the depth information come from?

- Given a calibrated binocular stereo pair, fuse it to produce a depth image
  - Humans can do it

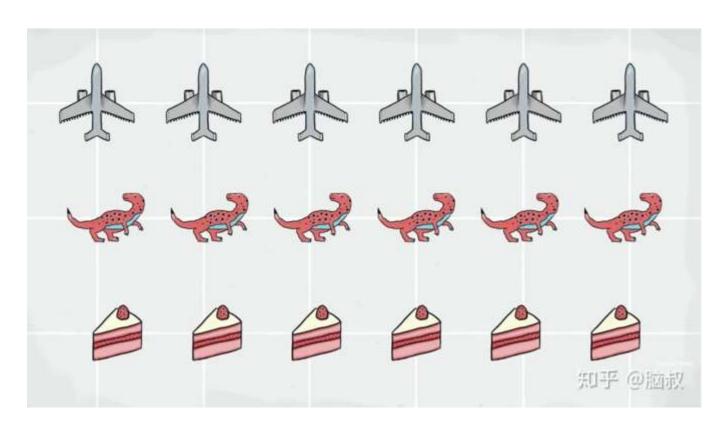


Stereograms: Invented by Sir Charles Wheatstone, 1838

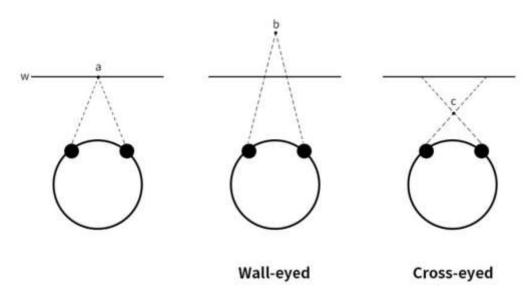


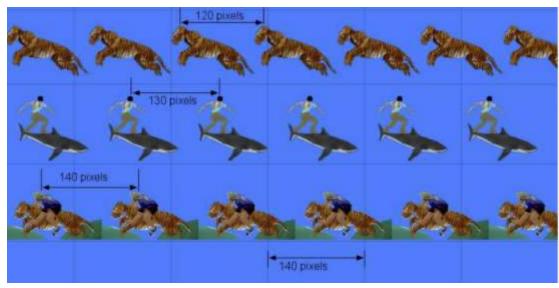


- Given a calibrated binocular stereo pair, fuse it to produce a depth image
  - Humans can do it

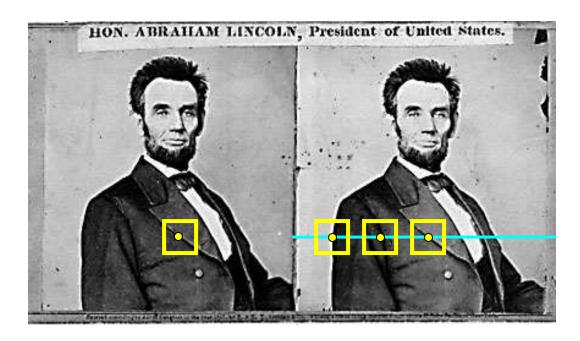


Autostereograms: www.magiceye.com



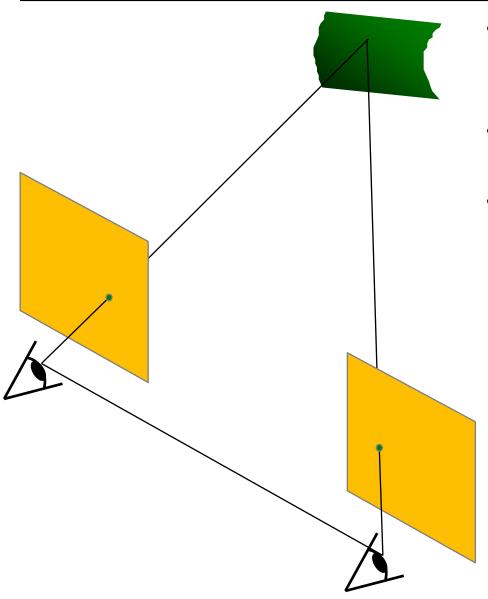


# Basic stereo matching algorithm



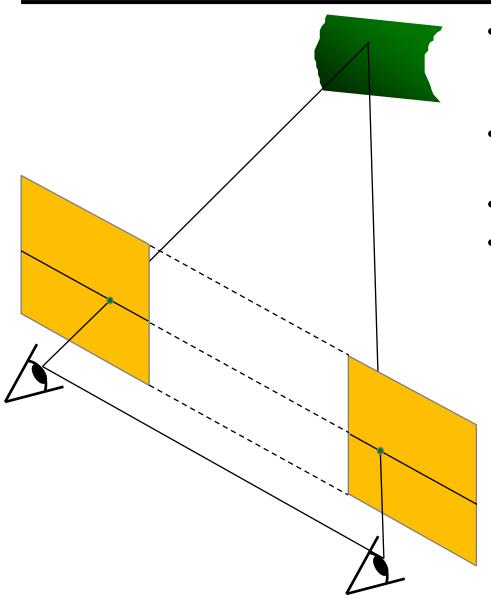
- For each pixel in the first image
  - Find corresponding epipolar line in the right image
  - Examine all pixels on the epipolar line and pick the best match
  - Triangulate the matches to get depth information
- Simplest case: epipolar lines are corresponding scanlines
  - When does this happen?

# Simplest Case: Parallel images



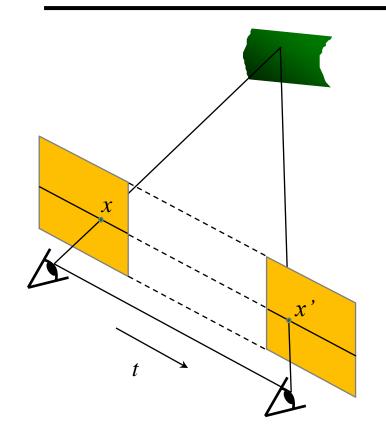
- Image planes of cameras are parallel to each other and to the baseline
- Camera centers are at same height
- Focal lengths are the same

# Simplest Case: Parallel images



- Image planes of cameras are parallel to each other and to the baseline
- Camera centers are at same height
- Focal lengths are the same
- Then epipolar lines fall along the horizontal scan lines of the images

# Essential matrix for parallel images



Epipolar constraint:

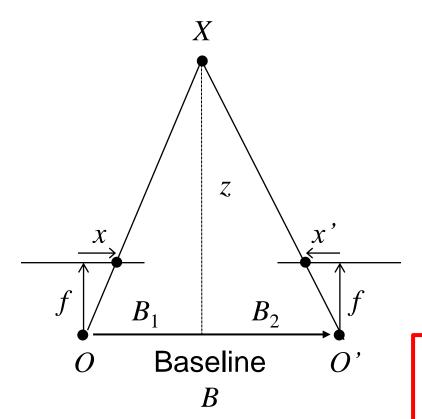
$$\boldsymbol{x}'^T \boldsymbol{E} \boldsymbol{x} = 0, \quad \boldsymbol{E} = [\boldsymbol{t}_{\times}] \boldsymbol{R}$$

$$R = I$$
  $t = (T, 0, 0)$ 

$$\boldsymbol{E} = [\boldsymbol{t}_{\times}]\boldsymbol{R} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & -T \\ 0 & T & 0 \end{bmatrix}$$

The y-coordinates of corresponding points are the same!

# Depth from disparity



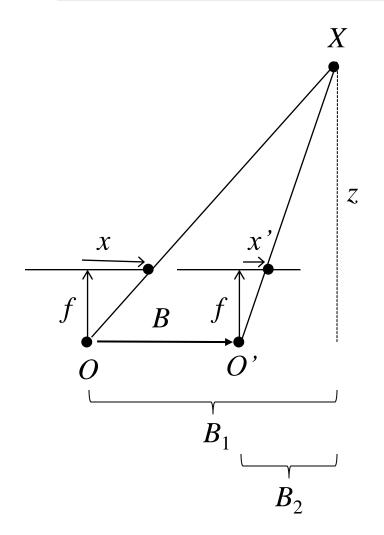
$$\frac{x}{f} = \frac{B_1}{z} \qquad \frac{-x^{\sharp}}{f} = \frac{B_2}{z}$$

$$\frac{x - x^{\complement}}{f} = \frac{B_1 + B_2}{z}$$

$$disparity = x - x' = \frac{B \cdot f}{z}$$

Disparity is inversely proportional to depth!

# Depth from disparity



$$\frac{x}{f} = \frac{B_1}{z} \qquad \frac{x^{\complement}}{f} = \frac{B_2}{z}$$

$$\frac{x - x^{\complement}}{f} = \frac{B_1 - B_2}{z}$$

$$disparity = x - x' = \frac{B \cdot f}{z}$$

# Stereo image rectification

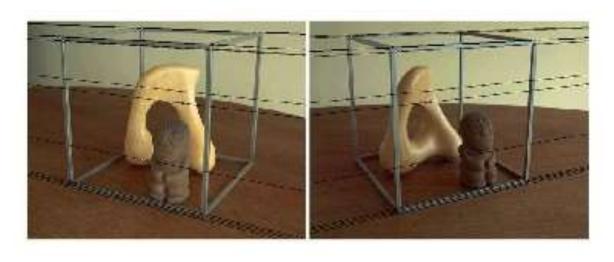
Stereo image rectification Reproject image planes onto a common

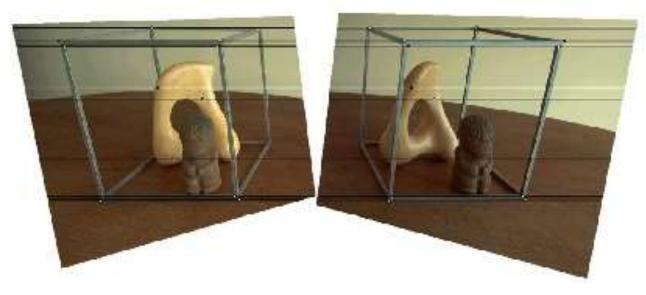
plane parallel to the line between optical centers

- Pixel motion is horizontal after this transformation
- Two homographies (3x3 transform), one for each input image reprojection

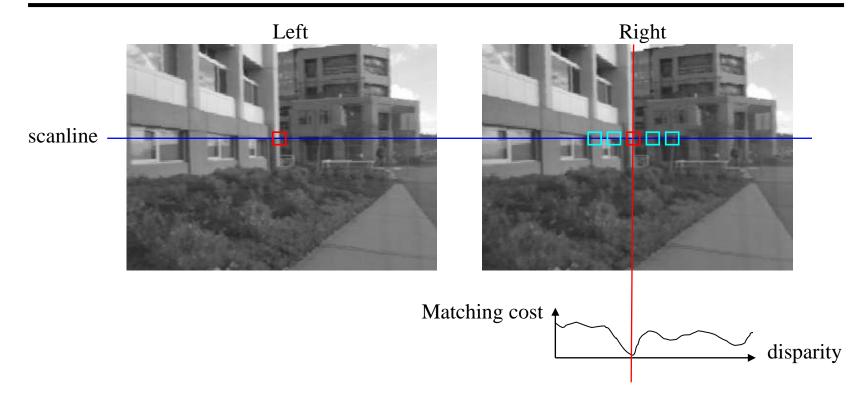
C. Loop and Z. Zhang. Computing Rectifying Homographies for Stereo Vision. IEEE Conf. Computer Vision and Pattern Recognition, 1999.

# Rectification example



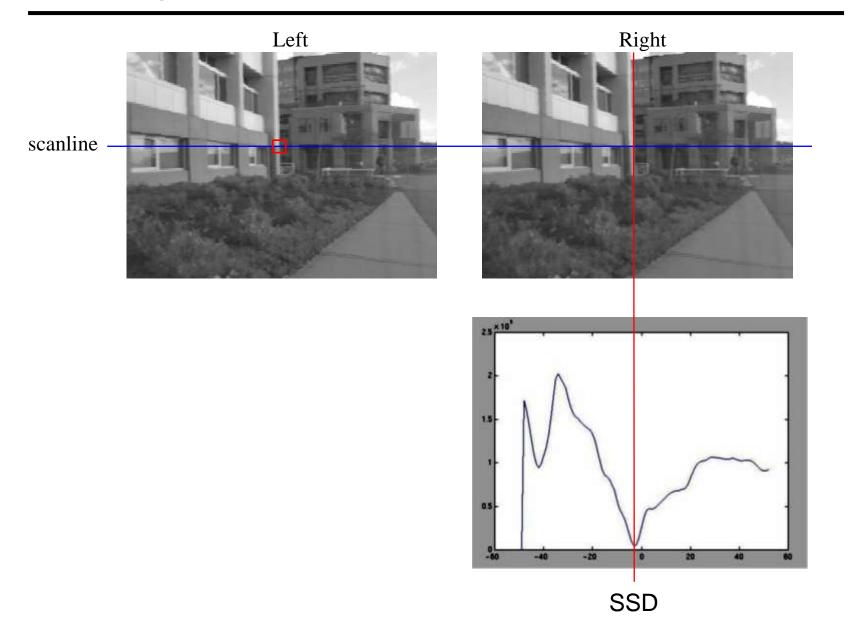


## Correspondence search

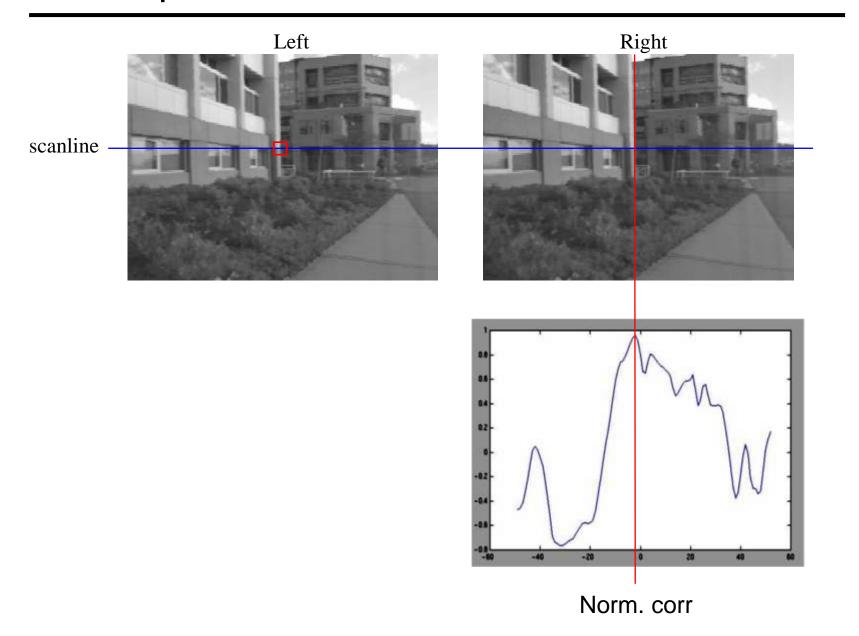


- Slide a window along the right scanline and compare contents of that window with the reference window in the left image
- Matching cost: SSD or normalized correlation

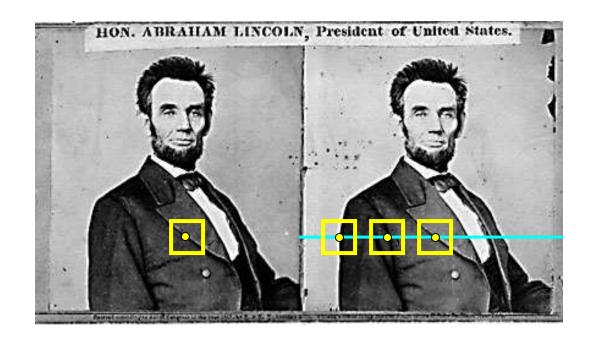
# Correspondence search



# Correspondence search

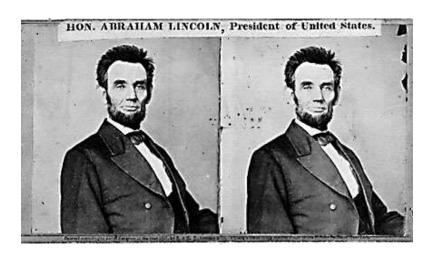


## Basic stereo matching algorithm



- If necessary, rectify the two stereo images to transform epipolar lines into scanlines
- For each pixel x in the first image
  - Find corresponding epipolar scanline in the right image
  - Examine all pixels on the scanline and pick the best match x'
  - Compute disparity x-x' and set  $depth(x) = B^*f/(x-x')$

# Failures of correspondence search



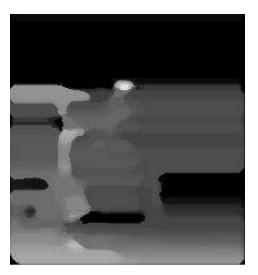
Textureless surfaces



Occlusions, repetition







Non-Lambertian surfaces, specularities

## Effect of window size







W = 3

W = 20

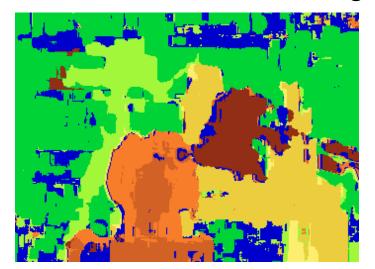
- Smaller window
  - + More detail
  - More noise
- Larger window
  - + Smoother disparity maps
  - Less detail

## Results with window search

#### Data



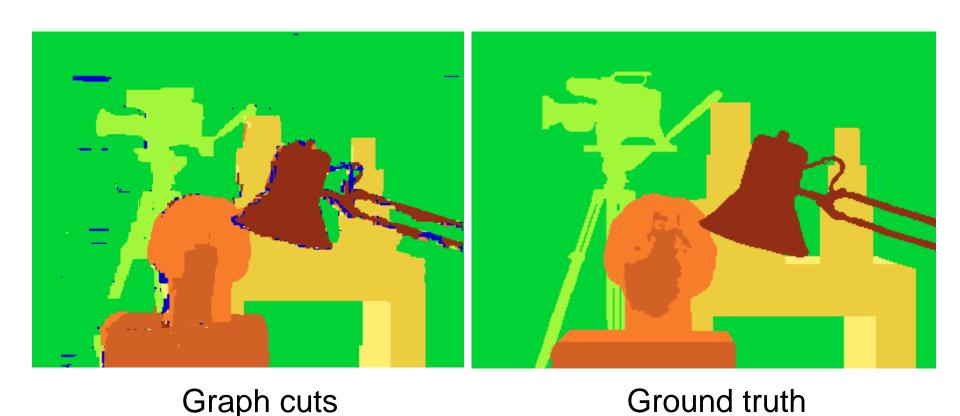
Window-based matching



Ground truth



## Better methods exist...



Y. Boykov, O. Veksler, and R. Zabih, <u>Fast Approximate Energy</u> <u>Minimization via Graph Cuts</u>, PAMI 2001

For the latest and greatest: <a href="http://www.middlebury.edu/stereo/">http://www.middlebury.edu/stereo/</a>