

# Stereo

Introduction to Computer Vision  
CSE 152  
Lecture 8

CSE 152, Spring 2017

Introduction to Computer Vision

# Announcements

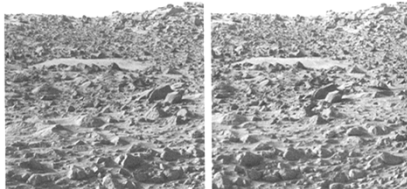
- Homework 2 is due today, 11:59 PM
- Homework 3 will be assigned today
- Reading:
  - Chapter 7: Stereopsis

CSE 152, Spring 2017

Introduction to Computer Vision

## Binocular Stereopsis: Mars

Given two images of a scene where relative locations of cameras are known, estimate depth of all common scene points.



Two images of Mars (Viking Lander)

CSE 152, Spring 2017

Introduction to Computer Vision

## An Application: Mobile Robot Navigation



The Stanford Cart,  
H. Moravec, 1979.

Courtesy O. Faugeras and H. Moravec.

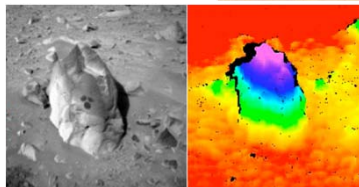
CSE 152, Spring 2017

The INRIA Mobile Robot, 1990.



Introduction to Computer Vision

Mars Exploratory Rovers:  
Spirit and Opportunity



CSE 152, Spring 2017

Introduction to Computer Vision

## Commercial Stereo Heads



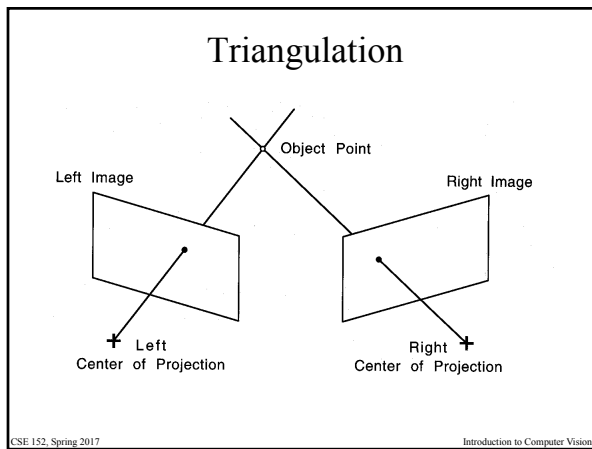
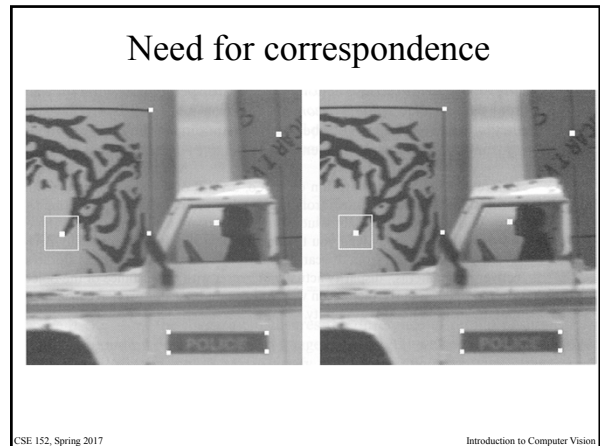
Trinocular  
stereo



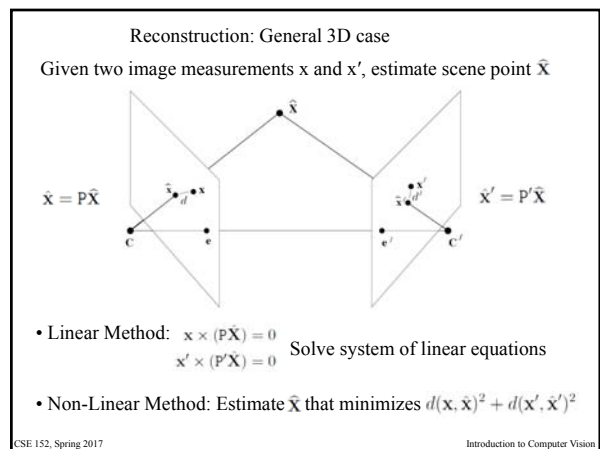
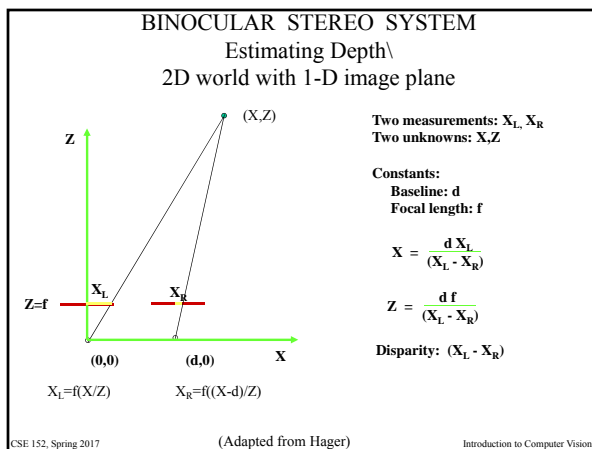
Binocular  
stereo

CSE 152, Spring 2017

Introduction to Computer Vision



- ### Stereo Vision Outline
- Offline: Calibrate cameras & determine **B** “epipolar geometry”
  - Online
    1. Acquire stereo images
    - C** 2. Rectify images to convenient epipolar geometry
    - D** 3. Establish correspondence
    - A** 4. Estimate depth
- CSE 152, Spring 2017 Introduction to Computer Vision



## Two Approaches

### 1. Feature-Based

- From each image, process "monocular" image to obtain cues (e.g., corners, lines).
- Establish correspondence between the two images.

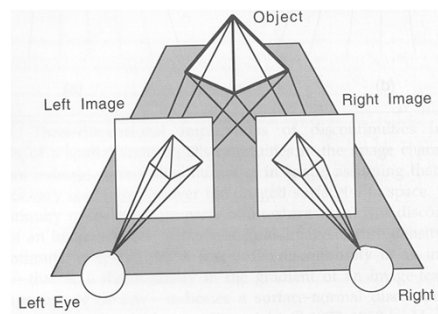
### 2. Area-Based

- Directly compare image regions between the two images.

CSE 152, Spring 2017

Introduction to Computer Vision

## Human Stereopsis



CSE 152, Spring 2017

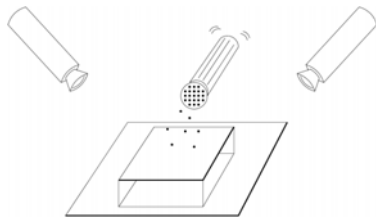
Introduction to Computer Vision

### Human Stereopsis: Binocular Fusion

How are the correspondences established?

Julesz (1971): Is the mechanism for binocular fusion a monocular process or a binocular one??

- There is anecdotal evidence for the latter (camouflage).

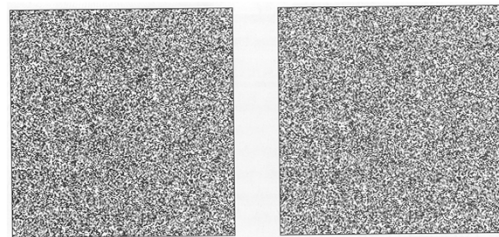


- Random dot stereograms provide an objective answer

CSE 152, Spring 2017

Introduction to Computer Vision

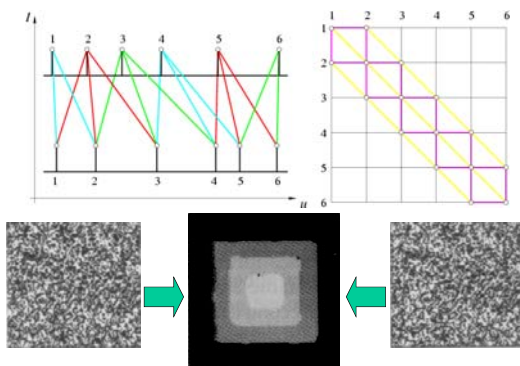
## Random Dot Stereograms



CSE 152, Spring 2017

Introduction to Computer Vision

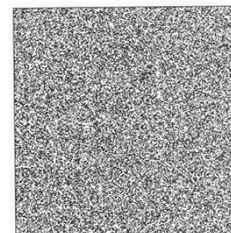
### A Cooperative Model (Marr and Poggio, 1976)



CSE 152, Spring 2017

Introduction to Computer Vision

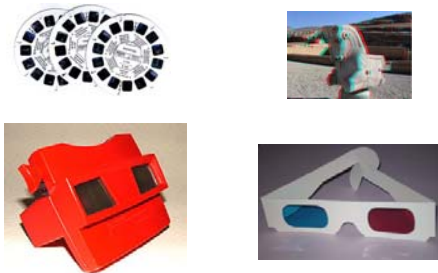
## Random Dot Stereograms



CSE 152, Spring 2017

Introduction to Computer Vision

## Stereoscopic 3D



CSE 152, Spring 2017

Introduction to Computer Vision

## Stereoscopic 3D



CSE 152, Spring 2017

Introduction to Computer Vision

## Was Rembrandt Stereo Blind?

- Detail of a 1639 etching.



CSE 152, Spring 2017

Introduction to Computer Vision



- In Rembrandt's painted self-portraits (left panel) in which the eyes are clearly visible, his left eye frequently looks straight out and the right off to the side. It is the opposite in his etchings (right panel).

CSE 152, Spring 2017

Introduction to Computer Vision

## Need for correspondence

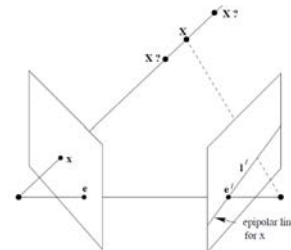


CSE 152, Spring 2017

Introduction to Computer Vision

## Point correspondence geometry

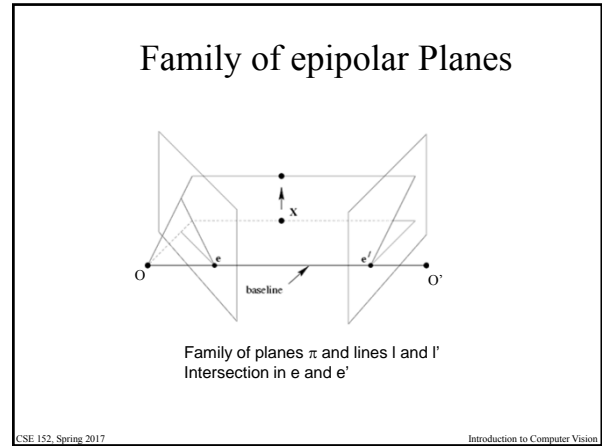
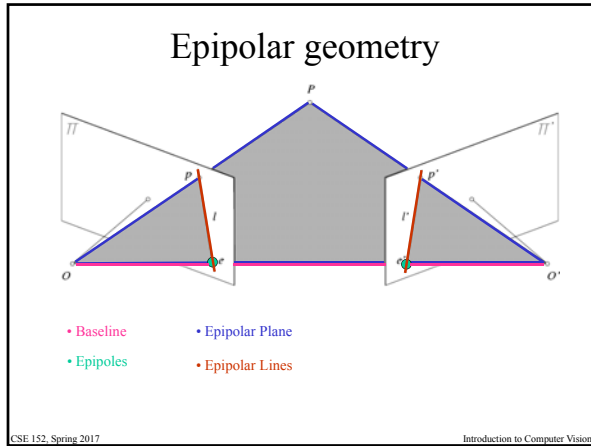
Where does a point  $x$  in the left image match a point in the right image?



- Potential matches for  $x$  have to lie on the corresponding epipolar line  $l'$

CSE 152, Spring 2017

Introduction to Computer Vision



### Epipolar Constraint: Calibrated Case

The vectors  $O_p$ ,  $OO'$ , and  $O'p'$  are coplanar

$$\vec{O}p \cdot [\vec{OO'} \times \vec{O'p'}] = 0 \Rightarrow p \cdot [t \times (\mathcal{R}p')] = 0 \text{ with } \begin{cases} p = (u, v, 1)^T \\ p' = (u', v', 1)^T \\ \mathcal{M} = (\text{Id } \mathbf{0}) \\ \mathcal{M}' = (\mathcal{R}^T, -\mathcal{R}^T t) \end{cases}$$

↓

Essential Matrix  
(Longuet-Higgins, 1981)

$$\leftarrow p^T \mathcal{E} p' = 0 \text{ with } \mathcal{E} = [t_x] \mathcal{R}$$

CSE 152, Spring 2017 Introduction to Computer Vision

### Skew Symmetric Matrix & Cross Product

- The cross product  $\mathbf{a} \times \mathbf{b}$  of two vectors  $\mathbf{a}$  and  $\mathbf{b}$  can be expressed as a matrix vector product  $[\mathbf{a}_x] \mathbf{b}$  where  $[\mathbf{a}_x]$  is the skew symmetric matrix:
 
$$[\mathbf{a}_x] = \begin{bmatrix} 0 & -a_3 & a_2 \\ a_3 & 0 & -a_1 \\ -a_2 & a_1 & 0 \end{bmatrix}$$
- A matrix  $S$  is skew symmetric if and only if
 
$$S = -S^T$$

CSE 152, Spring 2017 Introduction to Computer Vision

### Properties of the Essential Matrix

$p^T \mathcal{E} p' = 0 \text{ with } \mathcal{E} = [t_x] \mathcal{R}$

- $\mathcal{E} p'$  is the epipolar line associated with  $p'$ .
- $\mathcal{E}^T p$  is the epipolar line associated with  $p$ .
- $\mathcal{E} e' = 0$  and  $\mathcal{E}^T e = 0$ .
- $\mathcal{E}$  is singular (rank 2).
- $\mathcal{E}$  has two equal non-zero singular values (Huang and Faugeras, 1989).

CSE 152, Spring 2017 Introduction to Computer Vision

### Calibration

Determine intrinsic parameters and extrinsic relation of two cameras

CSE 152, Spring 2017 Introduction to Computer Vision

### The Eight-Point Algorithm (Longuet-Higgins, 1981)

$$(u, v, 1) \begin{pmatrix} F_{11} & F_{12} & F_{13} \\ F_{21} & F_{22} & F_{23} \\ F_{31} & F_{32} & F_{33} \end{pmatrix} \begin{pmatrix} u' \\ v' \\ 1 \end{pmatrix} = 0 \Rightarrow (uu', uv', u, vu', vv', v, u', v', 1) \begin{pmatrix} F_{11} \\ F_{12} \\ F_{13} \\ F_{21} \\ F_{22} \\ F_{23} \\ F_{31} \\ F_{32} \\ F_{33} \end{pmatrix} = 0$$

- Consider 8 points  $(u_i, v_i), (u'_i, v'_i)$
- Set  $F_{33}$  to 1

$$\begin{pmatrix} u_1 u'_1 & u_1 v'_1 & u_1 & v_1 u'_1 & v_1 v'_1 & v_1 & u'_1 & v'_1 \\ u_2 u'_2 & u_2 v'_2 & u_2 & v_2 u'_2 & v_2 v'_2 & v_2 & u'_2 & v'_2 \\ u_3 u'_3 & u_3 v'_3 & u_3 & v_3 u'_3 & v_3 v'_3 & v_3 & u'_3 & v'_3 \\ u_4 u'_4 & u_4 v'_4 & u_4 & v_4 u'_4 & v_4 v'_4 & v_4 & u'_4 & v'_4 \\ u_5 u'_5 & u_5 v'_5 & u_5 & v_5 u'_5 & v_5 v'_5 & v_5 & u'_5 & v'_5 \\ u_6 u'_6 & u_6 v'_6 & u_6 & v_6 u'_6 & v_6 v'_6 & v_6 & u'_6 & v'_6 \\ u_7 u'_7 & u_7 v'_7 & u_7 & v_7 u'_7 & v_7 v'_7 & v_7 & u'_7 & v'_7 \\ u_8 u'_8 & u_8 v'_8 & u_8 & v_8 u'_8 & v_8 v'_8 & v_8 & u'_8 & v'_8 \end{pmatrix} \begin{pmatrix} F_{11} \\ F_{12} \\ F_{13} \\ F_{21} \\ F_{22} \\ F_{23} \\ F_{31} \\ F_{32} \end{pmatrix} = - \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{pmatrix}$$

Solve for  $F_{11}$  to  $F_{32}$   
For more than 8 points, solve using linear least squares

### The Eight-Point Algorithm (Longuet-Higgins, 1981)

$$(u, v, 1) \begin{pmatrix} F_{11} & F_{12} & F_{13} \\ F_{21} & F_{22} & F_{23} \\ F_{31} & F_{32} & F_{33} \end{pmatrix} \begin{pmatrix} u' \\ v' \\ 1 \end{pmatrix} = 0 \Rightarrow (uu', uv', u, vu', vv', v, u', v', 1) \begin{pmatrix} F_{11} \\ F_{12} \\ F_{13} \\ F_{21} \\ F_{22} \\ F_{23} \\ F_{31} \\ F_{32} \\ F_{33} \end{pmatrix} = 0$$

- Alternatively, view this as system of homogenous equations in  $F_{11}$  to  $F_{33}$
- Solve as Eigenvector corresponding to the smallest Eigenvalue of matrix created from the image data.

Equivalent to solving

Minimize:

$$\sum_{i=1}^n (\mathbf{p}_i^T \mathcal{F} \mathbf{p}'_i)^2$$

under the constraint

$$|\mathcal{F}|^2 = 1.$$

### Epipolar geometry example



### The Fundamental matrix

The epipolar constraint is given by:

where  $\mathbf{p}$  and  $\mathbf{p}'$  are called homogeneous normalized image coordinates of points in the two images.

Without calibration, we can still identify corresponding points in two images, but we cannot convert to 3D coordinates. However, the relationship between the calibrated coordinates  $\mathbf{p}$  and  $\mathbf{p}'$ , and uncalibrated coordinates  $\mathbf{x}$  and  $\mathbf{x}'$  can be expressed as  $\mathbf{p} = \mathbf{K}^{-1} \mathbf{x}$  and  $\mathbf{p}' = \mathbf{K}'^{-1} \mathbf{x}'$

Therefore, we can express the epipolar constraint as:

$$\mathbf{p}'^T \mathbf{E} \mathbf{p} = (\mathbf{K}'^{-1} \mathbf{x}')^T \mathbf{E} (\mathbf{K}^{-1} \mathbf{x}) = \mathbf{x}'^T \mathbf{K}'^{-T} \mathbf{E} \mathbf{K}^{-1} \mathbf{x} = \mathbf{x}'^T \mathbf{F} \mathbf{x} = 0$$

where  $\mathbf{F} = \mathbf{K}'^{-T} \mathbf{E} \mathbf{K}^{-1}$  is called the Fundamental Matrix.

Can be solved using 8 point algorithm WITHOUT CALIBRATION

### Two-View Geometry

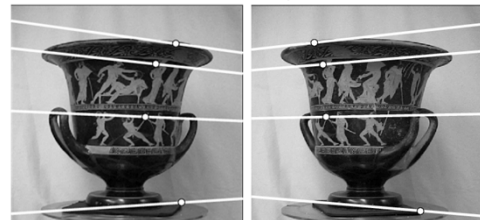
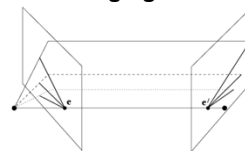
#### Essential Matrix E

- Calibrated
- Normalized coordinates
- Rank 2
  - Two nonzero singular values are equal
- 5 degrees of freedom
  - Camera rotation
  - Direction of camera translation
- Similarity reconstruction

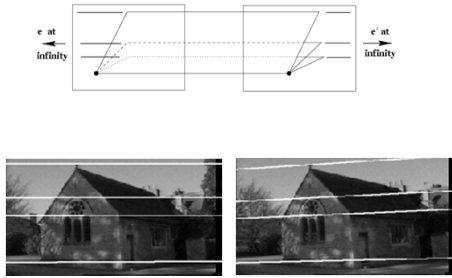
#### Fundamental Matrix F

- Uncalibrated
- Image coordinates
- Rank 2
- 7 degrees of freedom
  - Homogeneous matrix to scale
  - $\det \mathbf{F} = 0$
- Projective reconstruction

### Example: converging cameras



### Example: motion parallel with image plane

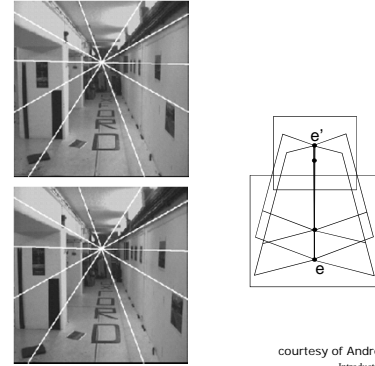


(simple for stereo  $\rightarrow$  rectification)

CSE 152, Spring 2017

courtesy of Andrew Zisserman  
Introduction to Computer Vision

### Example: forward motion



CSE 152, Spring 2017

courtesy of Andrew Zisserman  
Introduction to Computer Vision

## Next Lecture

- Early vision: multiple images
  - Stereo
- Reading:
  - Chapter 7: Stereopsis

CSE 152, Spring 2017

Introduction to Computer Vision