# CS231a PSET 2 Review

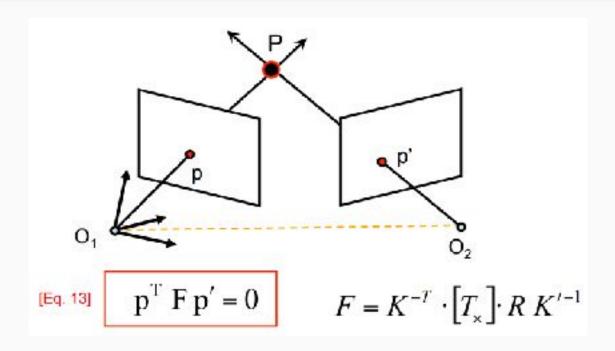
JunYoung Gwak

### Overview

- 1. Problem 1 Fundamental Matrix Estimation From Point Correspondences
- 2. Problem 2 Matching Homographies for Image Rectification
- 3. Problem 3 Factorization Method
- 4. Problem 4 Structure from Motion

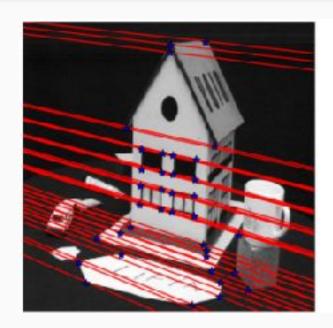
**Fundamental Matrix** 

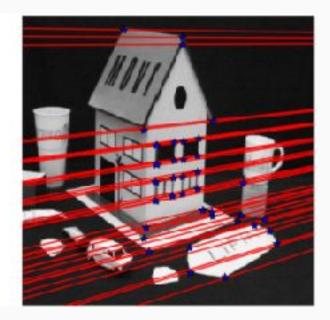
A matrix which maps the relationship of correspondences between stereo images



Ex)

Image and correspondences given in the homework





How to compute F?

Eight point algorithm

Note: in our pset, please review the documentation of the method carefully. Some may define F as p^TFp' or p'^TFp.

[Eq. 13] 
$$\mathbf{p}^{T} \mathbf{F} \mathbf{p'} = \mathbf{0}$$
  $\Rightarrow$   $p = \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} \quad p' = \begin{bmatrix} u' \\ v' \\ 1 \end{bmatrix}$ 

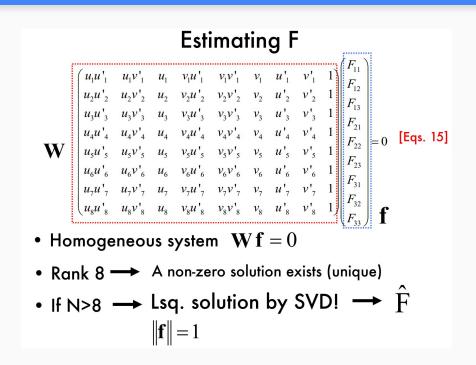
$$(u,v,1) \begin{pmatrix} F_{11} & F_{12} & F_{13} \\ F_{21} & F_{22} & F_{23} \\ F_{31} & F_{32} & F_{33} \end{pmatrix} \begin{pmatrix} u' \\ v' \\ 1 \end{pmatrix} = \mathbf{0} \quad \begin{pmatrix} F_{11} \\ F_{12} \\ F_{13} \\ F_{21} \\ F_{22} \\ F_{23} \\ F_{31} \\ F_{32} \\ F_{33} \end{pmatrix} = \mathbf{0}$$
Let's take 8 corresponding points [Eq. 14]

How to compute F?

Eight point algorithm

Problem?

W is highly unbalanced (not well conditioned)



Final step

Reduce rank(F) to 2

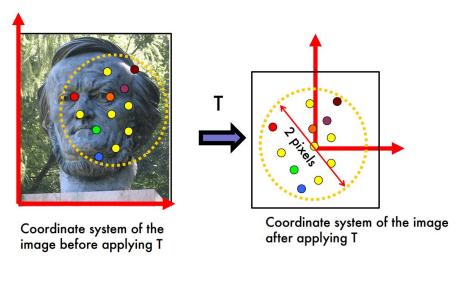
$$F = U \begin{bmatrix} s_1 & 0 & 0 \\ 0 & s_2 & 0 \\ 0 & 0 & 0 \end{bmatrix} V^T \qquad \text{Where:} \\ U \begin{bmatrix} s_1 & 0 & 0 \\ 0 & s_2 & 0 \\ 0 & 0 & s_3 \end{bmatrix} V^T = SVD(\hat{F})$$
 [HZ] pag 281, chapter 11, "Computation of F"

Possible improvement?

Pre-condition our linear system to get more stable result

origin = centroid of the points

mean square distance of the image points from origin is ~2px



• Origin = centroid of image points

• Mean square distance of the image points from origin is ~2 pixels

#### The Normalized Eight-Point Algorithm

- 0. Compute T and T' for image 1 and 2, respectively
- 1. Normalize coordinates in images 1 and 2:

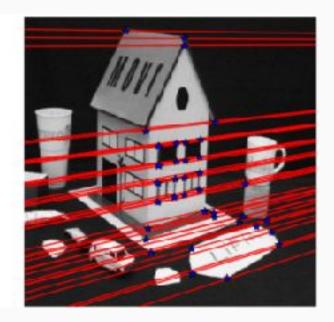
$$q_i = T p_i \qquad q'_i = T' p'_i$$

2. Use the eight-point algorithm to compute  $\hat{F}_q$  from the corresponding points  ${\bf q_i}$  and  ${\bf q_i'}$  .

1. Enforce the rank-2 constraint: 
$$\to$$
  $F_q$  such that: 
$$\begin{cases} q^T F_q \ q' = 0 \end{cases}$$
 2. De-normalize  $F_q$ :  $F = T'^T F_q T$ 

Epipolar lines





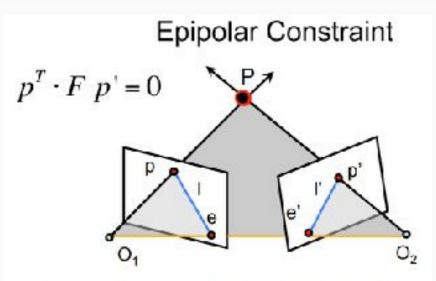
#### Distance to epipolar lines

Computing epipolar lines from F

$$I = Fp'$$

$$I' = F^Tp$$

$$\operatorname{distance}(ax+by+c=0,(x_0,y_0))=rac{|ax_0+by_0+c|}{\sqrt{a^2+b^2}}.$$



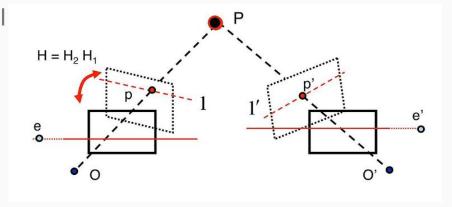
- I = F p' is the epipolar line associated with p'
- I'= F<sup>T</sup>p is the epipolar line associated with p

Make two images parallel to each other ⇒ epipole at infinity along the horizontal axis

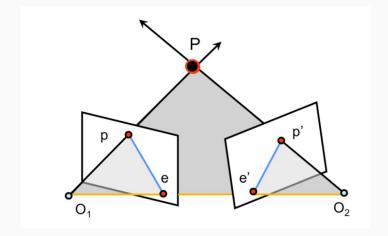


Make two images parallel to each other ⇒ epipole at infinity along the horizontal axis

- 1. Find epipoles
- Find two homographies that shift epipoles to infinity



- 1. Compute epipole
- Epipolar line I = Fp'
- Epipole lies on epipolar lines  $I \cdot x = 0$
- Epipole is an intersection of all epipolar lines



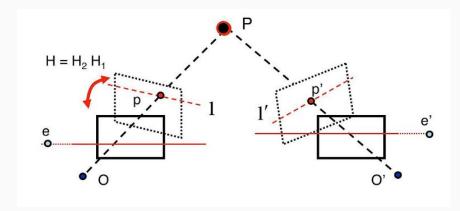
- 1. Compute epipole
   Due to noisy measurement, not all epipolar lines intersect in a single point
   ⇒ Find a point that minimizes least
- ⇒ Find a point that minimizes least square error of fitting a point to all the epipolar lines
- ⇒ Solve least square by SVD





$$\begin{bmatrix} \ell_1^T \\ \vdots \\ \ell_n^T \end{bmatrix} e = 0$$

- 2. Find two homographies that shift epipoles to infinity
  - a. Find homography H\_2 that maps the second epipole e' to a horizontal axis at infinity (f, 0, 0)
- Find the matching homography H\_1for the first image



Find homography H\_2 that maps the second epipole e' to a horizontal axis at infinity (f, 0, 0)

- i. Translate the second image s.t. the center is at (0, 0, 1) in homogeneous coord (T)
- ii. Apply rotation to place the epipole on the horizontal axis (f, 0, 1) (R)
- iii. Bring epipole at infinity on the horizontal axis (f, 0, 0) (G)

$$H_2 = T^{-1}GRT$$

Find homography H\_2 that maps the second epipole e' to a horizontal axis at infinity (f, 0, 0)

i. Translate the second image s.t. the center is at (0, 0, 1) in homogeneous coord (T)

$$T = egin{bmatrix} 1 & 0 & -rac{ ext{width}}{2} \ 0 & 1 & -rac{ ext{height}}{2} \ 0 & 0 & 1 \end{bmatrix}$$

Find homography H\_2 that maps the second epipole e' to a horizontal axis at infinity (f, 0, 0)

ii. Apply rotation to place the epipole on the horizontal axis (f, 0, 1) ( $\bf R$ ) The translated epipole  $\bf T$ e' =  $(e_1', e_2', 1)$ 

$$R = \begin{bmatrix} \alpha \frac{e_1'}{\sqrt{e_1'^2 + e_2'^2}} & \alpha \frac{e_2'}{\sqrt{e_1'^2 + e_2'^2}} & 0\\ -\alpha \frac{e_2'}{\sqrt{e_1'^2 + e_2'^2}} & \alpha \frac{e_1'}{\sqrt{e_1'^2 + e_2'^2}} & 0\\ 0 & 0 & 1 \end{bmatrix}$$

where  $\alpha = 1$  if  $e'_1 \geq 0$  and  $\alpha = -1$  otherwise.

Find homography H\_2 that maps the second epipole e' to a horizontal axis at infinity (f, 0, 0)

iii. Bring epipole (f, 0, 1) at infinity on the horizontal axis (f, 0, 0) (*G*)

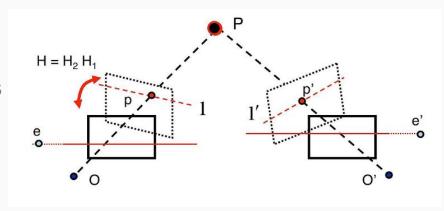
$$G = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -\frac{1}{f} & 0 & 1 \end{bmatrix}$$

Find homography H\_2 that maps the second epipole e' to a horizontal axis at infinity (f, 0, 0)

- i. Translate the second image s.t. the center is at (0, 0, 1) in homogeneous coord (T)
- ii. Apply rotation to place the epipole on the horizontal axis (f, 0, 1) (R)
- iii. Bring epipole at infinity on the horizontal axis (f, 0, 0) (**G**)

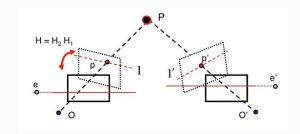
$$H_2 = T^{-1}GRT$$

- 2. Find two homographies that shift epipoles to infinity
  - a. Find homography H\_2 that maps the second epipole e' to a horizontal axis at infinity (f, 0, 0)
- b. Find the matching homographyH\_1 for the first image



Find the matching homography H\_1 for the first image

$$\arg\min_{H_1} \sum_i \|H_1 p_i - H_2 p_i'\|^2$$



Although the derivation is out of the scope of this class,

$$H_1 = H_A H_2 M$$
 
$$M = [e]_{\times} F + ev^T$$
 
$$v^T = \begin{bmatrix} 1 & 1 & 1 \end{bmatrix}$$

$$M = [e]_{\times} F + ev^{T}$$

$$v^{T} = \begin{bmatrix} 1 & 1 & 1 \end{bmatrix}$$

$$H_{A} = \begin{bmatrix} a_{1} & a_{2} & a_{3} \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Find the matching homography H\_1 for the first image

$$\arg\min_{H_1} \sum_i \|H_1 p_i - H_2 p_i'\|^2$$

$$H_1 = H_A H_2 M$$

$$\hat{p}_i = H_2 M p_i$$

$$\hat{p}_i' = H_2 p_i'$$

$$\arg\min_{H_A} \sum_i \|H_A \hat{p}_i - \hat{p}_i'\|^2$$

Find the matching homography H\_1 for the first image

$$\hat{p}_i = H_2 M p_i$$

$$\hat{p}'_i = H_2 p'_i$$

$$\hat{p}_i = (\hat{x}_i, \hat{y}_i, 1) \text{ and } \hat{p}'_i = (\hat{x}'_i, \hat{y}'_i, 1)$$

$$\arg\min_{\mathbf{a}} \sum_{i} (a_1 \hat{x}_i + a_2 \hat{y}_i + a_3 - \hat{x}'_i)^2$$

#### Solving least-square $W\mathbf{a} = b$

$$W = \begin{bmatrix} x_1 & y_1 & 1 \\ & \vdots & \\ \hat{x}_n & \hat{y}_n & 1 \end{bmatrix} \qquad b = \begin{bmatrix} x'_1 \\ \vdots \\ \hat{x}'_n \end{bmatrix}$$

$$b = \begin{bmatrix} x_1 \\ \vdots \\ \hat{x}'_n \end{bmatrix}$$

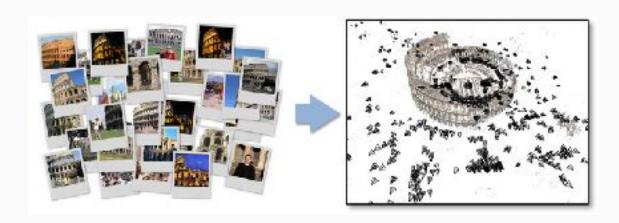
Structure from Motion (SfM)

Estimating 3D structure from 2D images that may be coupled with local motions

Input: 2D images

Output: 3D structure

(+ camera extrinsic)



In this homework, we explore two different approaches for SfM

- Factorization Method (problem 3) Tomasi & Kanade algorithm
- Bundle Adjustment (problem 4)

### Problem 3 - Factorization Method

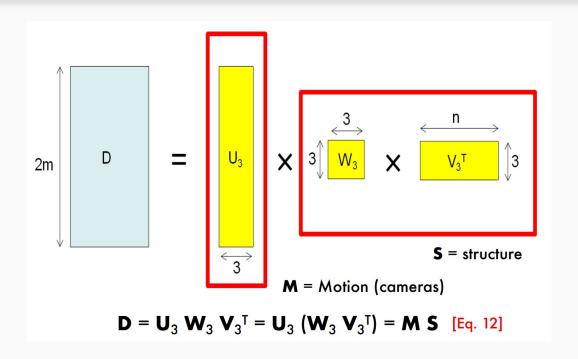
Centering: subtract the centroid of the image points

[Eq. 6] 
$$\hat{\mathbf{x}}_{ij} = \mathbf{x}_{ij} - \frac{1}{n} \sum_{k=1}^{n} \mathbf{x}_{ik}$$

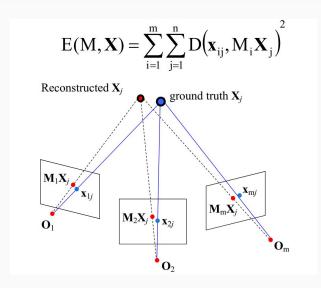
[Eq. 6] 
$$\hat{\mathbf{x}}_{ij} = \mathbf{x}_{ij} - \frac{1}{n} \sum_{k=1}^{n} \mathbf{x}_{ik}$$

$$\mathbf{D} = \begin{bmatrix} \hat{\mathbf{x}}_{11} & \hat{\mathbf{x}}_{12} & \cdots & \hat{\mathbf{x}}_{1n} \\ \hat{\mathbf{x}}_{21} & \hat{\mathbf{x}}_{22} & \cdots & \hat{\mathbf{x}}_{2n} \\ & \ddots & & \\ \hat{\mathbf{x}}_{m1} & \hat{\mathbf{x}}_{m2} & \cdots & \hat{\mathbf{x}}_{mn} \end{bmatrix}$$

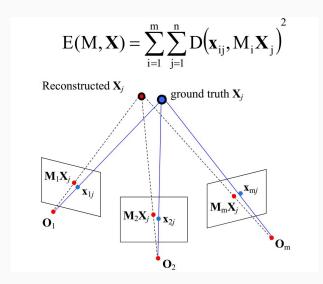
### Problem 3 - Factorization Method



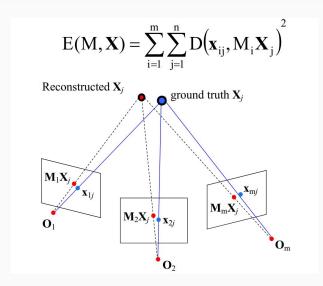
- 1. Compute essential matrix E from two views
- Use E to make initial estimate of relative rotation R and translation T
- 3. Estimate 3D location of the reconstruction given RT
- 4. Optimize (bundle adjustment)
  - Jointly optimize all relative camera motions (R's and T's)
  - Minimize total reprojection error with respect to all 3D point and camera parameters
- 5. Repeat 3 and 4 for pairs of frames



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1. To compute R: Given the singular value decomposition  $E = UDV^T$ 

$$Q = UWV^T$$
 or  $UW^TV^T$ , where

$$W = \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

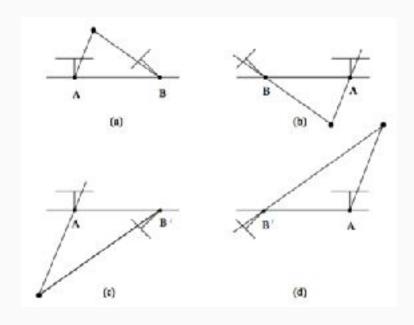
Note that this factorization of E only guarantees that Q is orthogonal. To find a rotation, we simply compute  $R = (\det Q)Q$ .

2. To compute T: Given that  $E = U\Sigma V^T$ , T is simply either  $u_3$  or  $-u_3$ , where  $u_3$  is the third column vector of U.

Use E to make initial estimate of relative rotation R and translation T

However, this gives four pairs of rotation and translation, (R<sub>1</sub>, R<sub>2</sub>) x (T, -T)

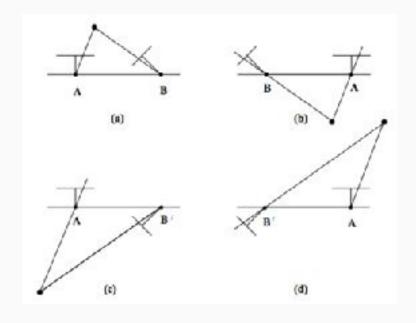
How do we find out which R and T is the correct one?



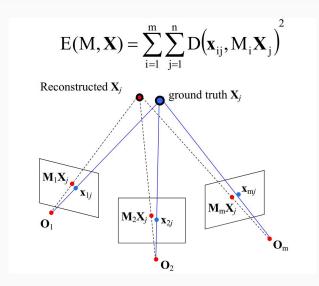
There exists only **one** solution that will consistently produce 3D points which are both in front of camera

Compute 3D point's location in the RT frame!

- Find 3D location of the image points given RT frame
- Chose the one which has the most 3D points with positive depth (z-coordinate) with respect to both camera frame



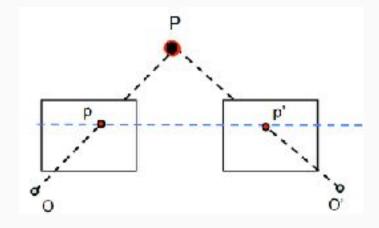
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Estimate 3D location of the reconstruction given 1. projective camera matrix 2. their image coordinates

Two different possible approaches:

- 1. Formulating a linear equation to solve
- Nonlinear optimization to minimize reprojection error



1. For each image i, we have  $p_i = M_i P$ , where P is the 3D point,  $p_i$  is the homogenous image coordinate of that point, and  $M_i$  is the projective camera matrix.

2. Formulate matrix

$$A = \begin{bmatrix} p_{1,1}m^{3\top} - m^{1\top} \\ p_{1,2}m^{3\top} - m^{2\top} \\ \vdots \\ p_{n,1}m^{3\top} - m^{1\top} \\ p_{n,2}m^{3\top} - m^{2\top} \end{bmatrix}$$

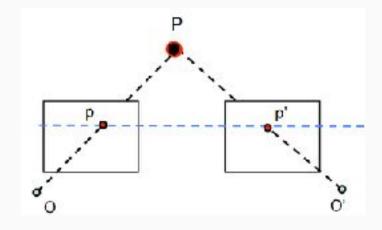
where  $p_{i,1}$  and  $p_{i,2}$  are the xy coordinates in image i and  $m^{k\top}$  is the k-th row of M.

3. The 3D point can be solved for by using the singular value decomposition.

Estimate 3D location of the reconstruction given 1. projective camera matrix 2. their image coordinates

Two different possible approaches:

- 1. Formulating a linear equation to solve
- 2. Nonlinear optimization to minimize reprojection error



Estimate 3D location of the reconstruction given 1. projective camera matrix 2. their image coordinates

Nonlinear optimization to minimize reprojection error

Gauss-Newton algorithm

$$\hat{P} = \hat{P} - (J^T J)^{-1} J^T e$$

Begin from linear estimation for better initialization

(reprojection) error: difference between the projected point (MiP) and ground-

truth image coordinate pi

Jacobian:

$$e = \begin{bmatrix} e_1 \\ \vdots \\ e_N \end{bmatrix} = \begin{bmatrix} p_1 - M_1 \hat{P} \\ \vdots \\ p_n - M_n \hat{P} \end{bmatrix}$$

$$J = \begin{bmatrix} \frac{\partial e_1}{\partial \hat{P}_1} & \frac{\partial e_1}{\partial \hat{P}_2} & \frac{\partial e_1}{\partial \hat{P}_3} \\ \vdots & \vdots & \vdots \\ \frac{\partial e_N}{\partial \hat{P}_1} & \frac{\partial e_N}{\partial \hat{P}_2} & \frac{\partial e_N}{\partial \hat{P}_3} \end{bmatrix}$$

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