CS231a HW2

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Overview

Focus on

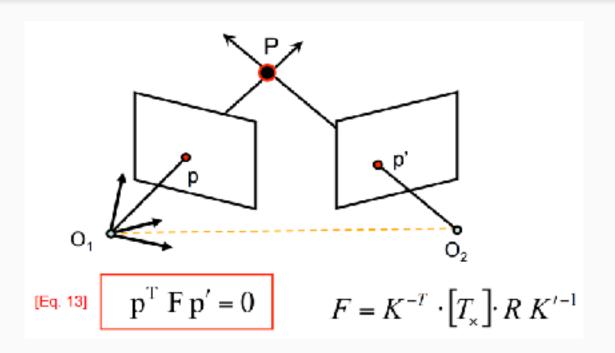
- 1. Problem 2 Image Rectification
- 2. Problem 4 Structure from Motion

Will briefly cover Problem 1 - Fundamental Matrix Estimation

Will not cover Problem 3 - Factorization Method (Enough details covered in class. Please visit our office hour if you need help)

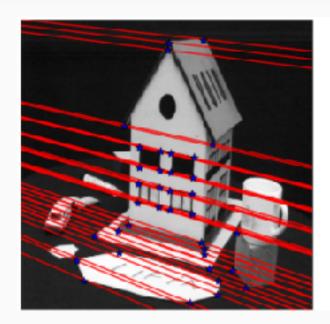
Fundamental Matrix

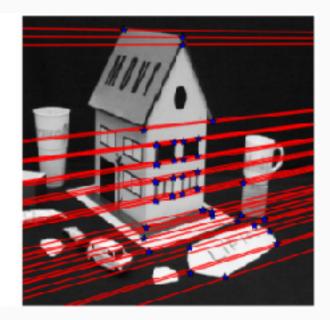
 A matrix which maps the relationship of correspondences between stereo images



Ex)

Image and correspondences given in the homework





How to compute F?

Eight point algorithm

[Eq. 13]
$$\mathbf{p}^{T} \mathbf{F} \mathbf{p}' = \mathbf{0}$$
 \Longrightarrow $p = \begin{bmatrix} u \\ v \\ 1 \end{bmatrix}$ $p' = \begin{bmatrix} u' \\ v' \\ 1 \end{bmatrix}$
$$(u_{1}v_{1}1) \begin{pmatrix} F_{11} & F_{12} & F_{13} \\ F_{21} & F_{22} & F_{23} \\ F_{31} & F_{32} & F_{33} \end{pmatrix} \begin{pmatrix} u' \\ v' \\ 1 \end{pmatrix} = 0$$

$$\begin{pmatrix} F_{11} \\ F_{12} \\ F_{13} \\ F_{21} \\ F_{22} \\ F_{33} \\ F_{31} \\ F_{32} \\ F_{33} \\ F_{33} \end{pmatrix} = 0$$
 Let's take 8 corresponding points
$$\begin{bmatrix} Eq. 14 \end{bmatrix}$$

How to compute F?

Eight point algorithm

Problem?

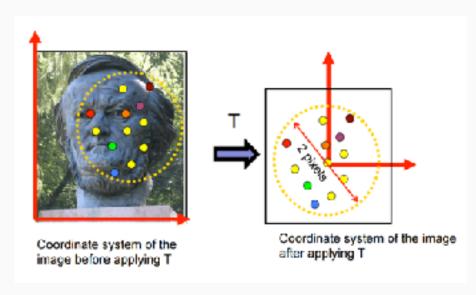
 W is highly unbalanced (not well conditioned)

$$\begin{aligned} \mathbf{Wf} &= 0 \\ \begin{pmatrix} u_1u'_1 & u_1v'_1 & u_1 & v_1u'_1 & v_1v'_1 & v_1 & u'_1 & v'_1 & 1 \\ u_2u'_2 & u_2v'_2 & u_2 & v_2u'_2 & v_2v'_2 & v_2 & u'_2 & v'_2 & 1 \\ u_3u'_1 & u_3v'_3 & u_3 & v_3u'_3 & v_3v'_3 & v_3 & u'_3 & v'_3 & 1 \\ u_4u'_4 & u_4v'_4 & u_4 & v_4u'_4 & v_4v'_4 & v_4 & u'_4 & v'_4 & 1 \\ u_5u'_5 & u_5v'_5 & u_5 & v_5u'_5 & v_5v'_5 & v_5 & u'_5 & v'_5 & 1 \\ u_6u'_6 & u_5v'_6 & u_6 & v_6u'_6 & v_6v'_6 & v_6 & u'_6 & v'_5 & 1 \\ u_7u'_7 & u_7v'_7 & u_7 & v_7u'_7 & v_7v'_7 & v_7 & u'_7 & v'_7 & 1 \\ u_8u'_8 & u_8v'_8 & u_8 & v_8u'_8 & v_8v'_8 & v_8 & u'_8 & v'_8 & 1 \end{pmatrix} \begin{bmatrix} F_{11} \\ F_{12} \\ F_{13} \\ F_{21} \\ F_{22} \\ F_{23} \\ F_{31} \\ F_{32} \\ F_{33} \end{bmatrix}$$

Possible improvement?

Pre-condition our linear system to get more stable result

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



Final step

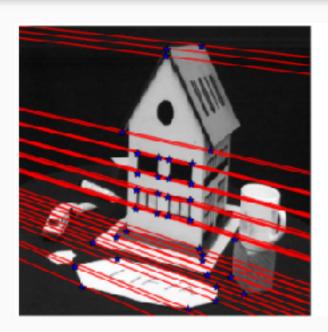
Reduce rank(F) to 2

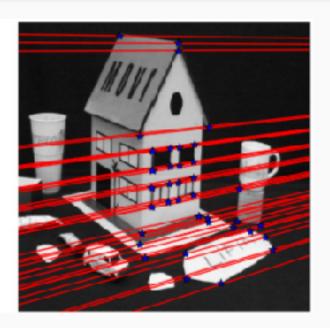
Find F that minimizes
$$\left\|F-\hat{F}\right\|=0$$
 Frobenius norm (*) Subject to $\text{det}(\text{F})\text{=}0$

$$F = U \begin{bmatrix} s_1 & 0 & 0 \\ 0 & s_2 & 0 \\ 0 & 0 & 0 \end{bmatrix} V^T \qquad \text{Where:}$$

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

Epipolar lines

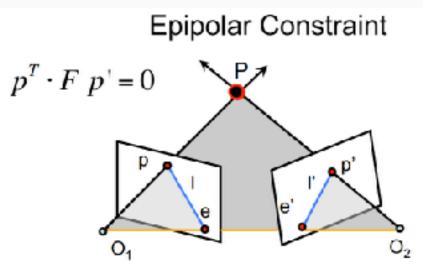




Computing epipolar lines from F

$$I = Fp'$$

$$I' = F^Tp$$

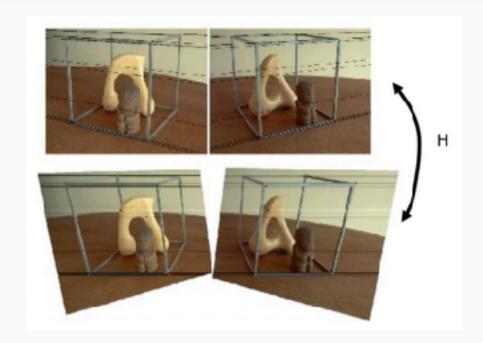


- I = F p' is the epipolar line associated with p'
- I'= F^Tp is the epipolar line associated with p

Image Rectification

Making two images "parallel"

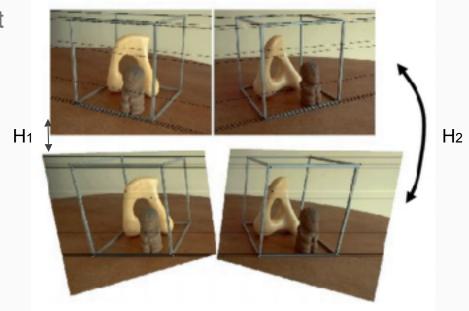
All correspondences lie on the same y axis



Application

Novel view synthesis (view morphing) based on rectified images

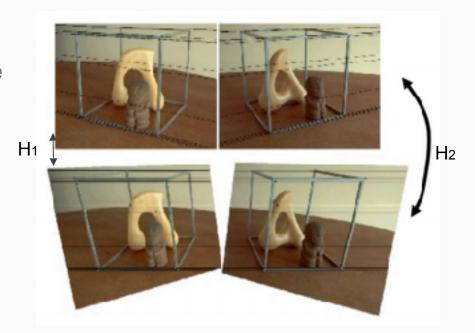




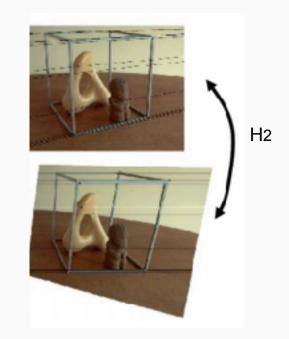
Overview

- Find H_2: morph image 2 to make all epipolar lines parallel
 - Find epipole of image 2 (e₂)
 - Find homography H₂ which brings e₂ to a point a infinity
- Find H_1: minimize square distance from rectified image 2 to image 1

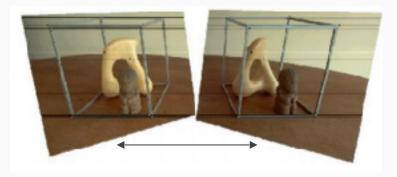
- 1. Find H₂ such that epipolar lines are all horizontally aligned (parallel)
- 2. Find H₁ which minimizes square distance between corresponding points of the image



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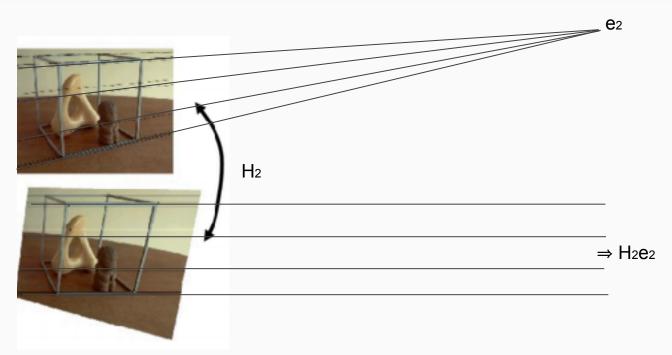
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- 1. Find epipole of image 2, e₂
- 2. FindhomographyH₂ whichbrings e₂ to apoint at infinity

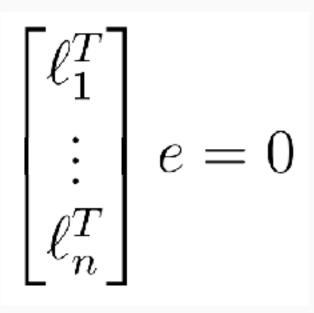


Overview

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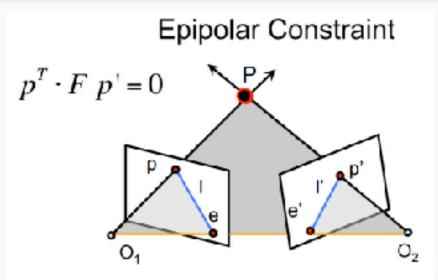
Find epipole of image 2 (e₂)

- Epipole is an intersection of epipolar lines
 - ⇔ epipole lies on every epipolar lines
 - o defining epipolar line ℓ such that all points on the line are in set $\{x|\ell^Tx=0\}$, formulate a linear system of equations on the right
 - Solve using SVD



Computing epipolar lines from F

$$l = Fp'$$
$$l' = F^T p$$



- I = F p' is the epipolar line associated with p'
- I'= F^Tp is the epipolar line associated with p

Overview

- Find H_2: morph image 2 to make all epipolar lines parallel
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Find homography H₂ which brings e₂ to a point at infinity

Intuition (from Hartley, Zisserman):

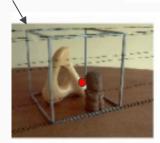
To keep image as realistic as possible after transformation, we want to keep H₂ to act as a rigid transformation in the neighborhood of a given selected point x₀ of the image

Center of the image is often a good choice for xo

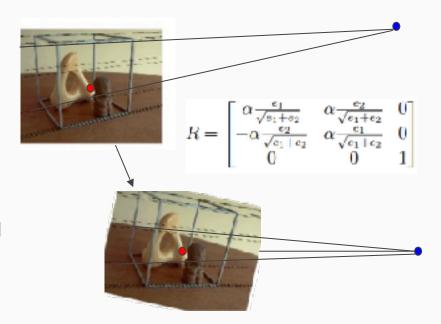
- 1. Translate image coordinate so that the origin will be at the image center
- 2. Rotate the image so that e₂ will lie horizontal axis at some point (f, 0, 1)
- 3. Bring e₂ to infinity on (f, 0, 0)
- 4. Translate image coordinate back to the original origin



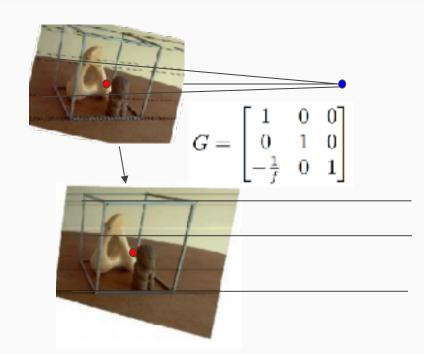
$$T = \begin{bmatrix} 1 & 0 & -\frac{\text{width}}{2} \\ 0 & 1 & -\frac{\text{height}}{2} \\ 0 & 0 & 1 \end{bmatrix}$$



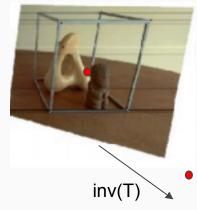
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$$H_2 = T^{-1}GRT$$

Overview

- Find H₂: morph image 2 to make all epipolar lines parallel
 - Find epipole of image 2 (e₂)
 - Find homography H₂ which brings e₂ to a point a infinity
- Find H₁: minimize square distance from rectified image 2 to image 1

Find H₁: minimize square distance from rectified image 2 to image 1

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

Why this? $H_1^{-T}\ell_i = H_2^{-T}\ell_i'$ Reduce matching lines to matching points

We will take a simpler approach for this problem

Find H₁: minimize square distance from rectified image 2 to image 1

We can show that
$$H_1 = H_A H_2 M$$

where
$$H_A = egin{bmatrix} a_1 & a_2 & a_3 \ 0 & 1 & 0 \ 0 & 0 & 1 \end{bmatrix}$$

This form of H_A allows a transformation of H_1 where its epipole is at (1, 0, 0) - a page long proof of this in Chapter 11 of Hartley & Zisserman's textbook

Find H₁: minimize square distance from rectified image 2 to image 1

We can also show that
$$M = [e]_{\times}F + ev^T$$

because 1)
$$M = [e]_{\times} F$$

- we know F up to scale
- ullet any skew-symmetric matrix X (including [e]x) is $A=A^3$ up to scale

$$F = [e]_{\times} M = [e]_{\times} [e]_{\times} [e]_{\times} M = [e]_{\times} [e]_{\times} F$$

Find H₁: minimize square distance from rectified image 2 to image 1

We can also show that
$$M = [e]_{ imes} F + ev^T$$

because 2) if columns of M are added by any scalar multiple of e,

up to scale,
$$F=[e]_{ imes}M$$

Therefore, $M=[e]_{\times}F+ev^T$ is more general case of defining M Where in practice, we use $v=\begin{bmatrix}1&1&1\end{bmatrix}$

Find H₁: minimize square distance from rectified image 2 to image 1

Which reduces to

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

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

Find H₁: minimize square distance from rectified image 2 to image 1

Which reduces down to solving least squares $W[a_1, a_2, a_3]^T = b$

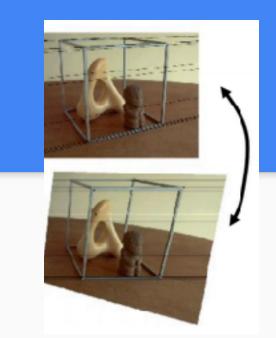
Where
$$W = \begin{bmatrix} \hat{x}_1 & \hat{y}_1 & 1 \\ & \vdots & \\ \hat{x}_n & \hat{y}_n & 1 \end{bmatrix} \qquad b = \begin{bmatrix} \hat{x}_1' \\ \vdots \\ \hat{x}_n' \end{bmatrix}$$

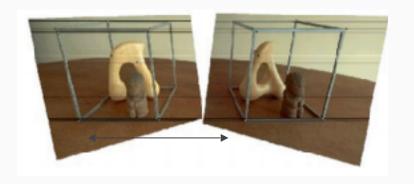
and
$$\hat{p}_i = H_2 M p_i$$
 $\hat{p}'_i = H_2 p_i$

Problem 2 - Image Rectification

Find a homography H₁ and H₂ such that corresponding points of image 1 and image 2 lie on the same y axis

- 1. Find H₂ such that epipolar lines are all horizontally aligned (parallel)
- 2. Find H₁ which minimizes square distance between corresponding points of the image





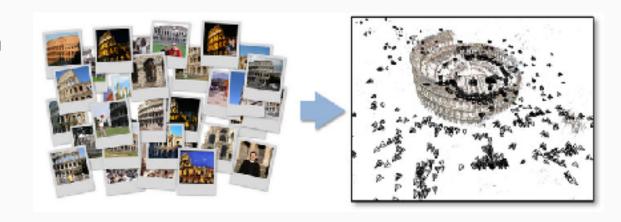
Structure from Motion (SfM)

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

Input: 2D images

Output: 3D structure

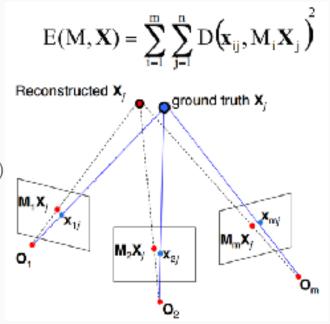
(+ camera extrinsic)



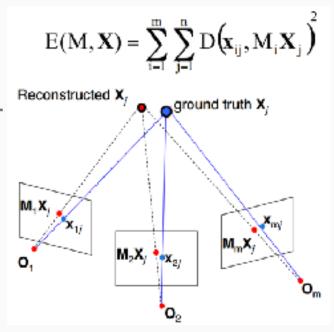
In this homework, we explore two different approaches for SfM

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

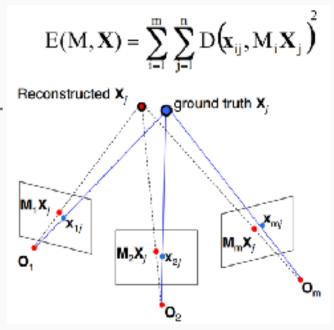
- 1. Compute essential matrix E from two views
- Use E to make initial estimate of relative rotation R and translation T
- 3. Estimate 3D location of the reconstruction given RT
- 4. Optimize (bundle adjustment)
 - Jointly optimize all relative camera motions (R's and T's)
 - Minimize total reprojection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



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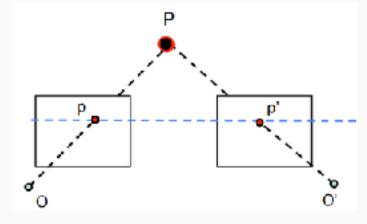
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Estimate 3D location of the reconstruction given 1. projective camera matrix 2. their image coordinates

Two different possible approaches:

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



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

Formulating a linear equation to solve, using [p=MP] Solve for AP=0 (using SVD) where:

$$A = \begin{bmatrix} p_{1,1}m^{3\top} - m^{1\top} \\ p_{1,2}m^{3\top} - m^{2\top} \\ \vdots \\ p_{n,1}m^{3\top} - m^{1\top} \\ p_{n,2}m^{3\top} - m^{2\top} \end{bmatrix} \quad \begin{array}{l} \text{p}_{\text{i,j:}} \; (\mathbf{x}, \, \mathbf{y})[\mathbf{j}] \; \text{coordinate of } \textit{ith image} \\ \text{m}^{\text{kT}} : \textit{k-th row of M} \\ \end{array}$$

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

Nonlinear optimization to minimize reprojection error

Gauss-Newton algorithm

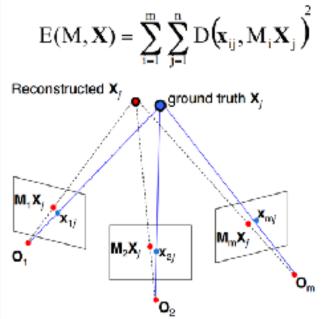
$$\hat{P} = \hat{P} - (J^T J)^{-1} J^T e$$

Begin from linear estimation for better initialization How to define error and Jacobian?

(reprojection) error: difference between the projected point (MiP) and groundtruth image coordinate pi

Jacobian:
$$J = \begin{bmatrix} \frac{\partial e_1}{\partial P_1} & \frac{\partial e_1}{\partial P_2} & \frac{\partial e_1}{\partial P_3} \\ \vdots & \vdots & \vdots \\ \frac{\partial e_m}{\partial P_1} & \frac{\partial e_m}{\partial P_2} & \frac{\partial e_m}{\partial P_3} \end{bmatrix}$$

- Compute essential matrix E from two views
- Use E to make initial estimate of relative rotation R and translation T
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Use E to make initial estimate of relative rotation R and translation T

$$E = \left[T_{\times}\right] \cdot R$$

1. To compute R: Given the singular value decomposition $E = UDV^T$, we can rewrite E = MQ where $M = UZU^T$ and $Q = UWV^T$ or UW^TV^T , where

$$Z = \begin{bmatrix} 0 & 1 & 0 \\ -1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \quad \text{and} \quad W = \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

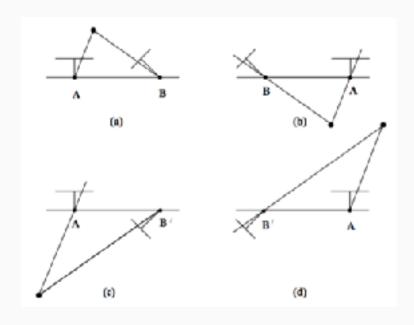
Note that this factorization of E only guarantees that Q is orthogonal. To find a rotation, we simply compute $R = (\det Q)Q$.

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, $(R_1, R_2) \times (T, -T)$

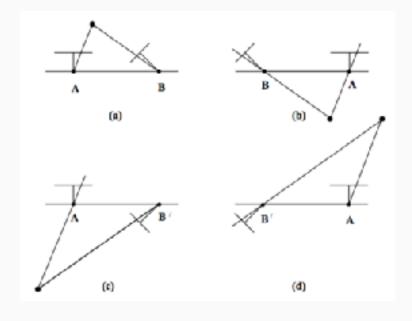
How do we find out which R and T is the correct one?



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



- 1. Compute fundamental matrix F from two views
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