

# 電腦視覺與應用

# Computer Vision and Applications

## Lecture06-2-Two-views geometry-case study

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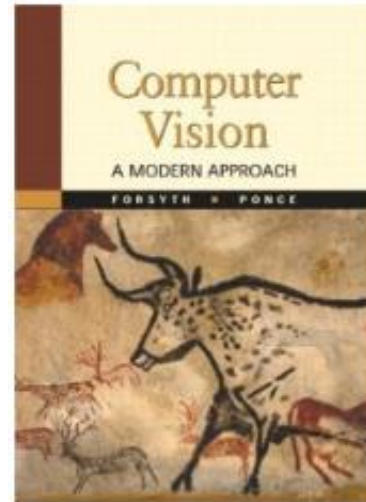
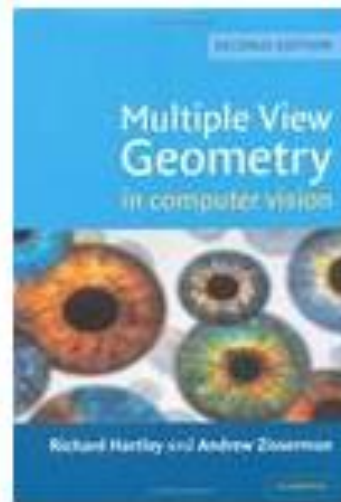
e-mail: [thl@mail.ntust.edu.tw](mailto:thl@mail.ntust.edu.tw)





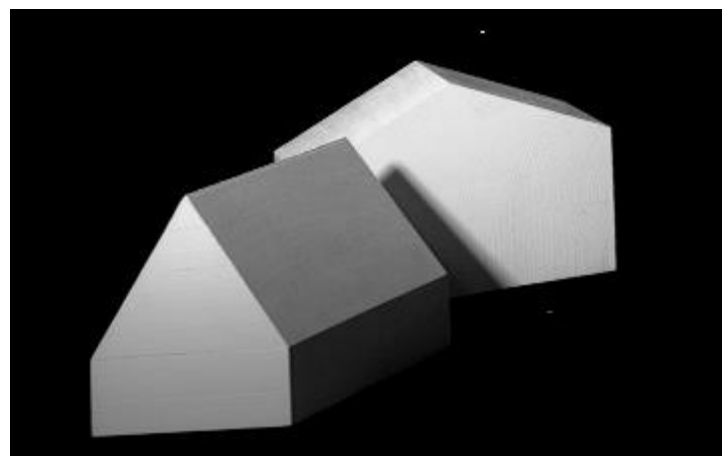
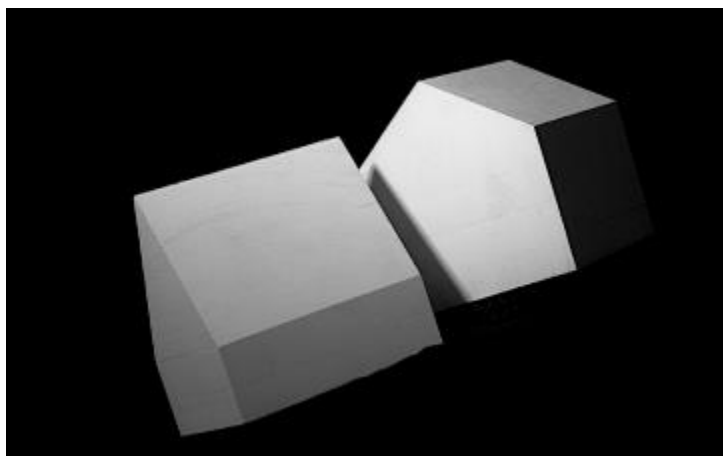
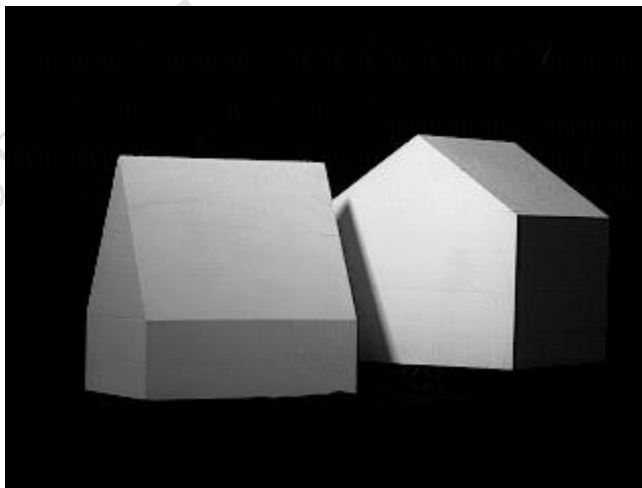
# Two-views geometry

- Case study for stereo-vision & homography
- Lecture Reference at:
  - Multiple View Geometry in Computer Vision, [Chapter 11](#)
  - Computer Vision A Modern Approach, [Chapter 11](#).



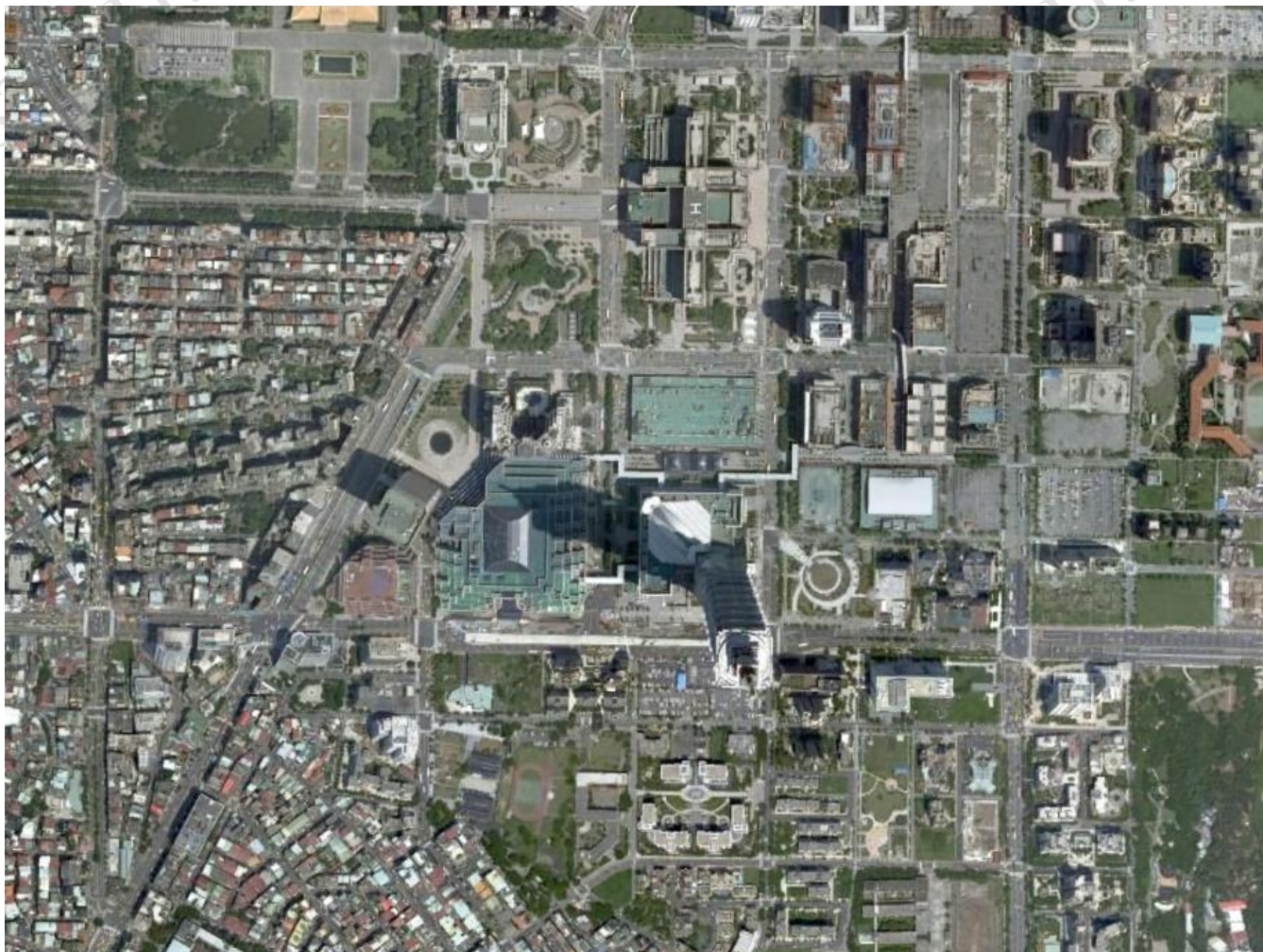


# Stereo-image





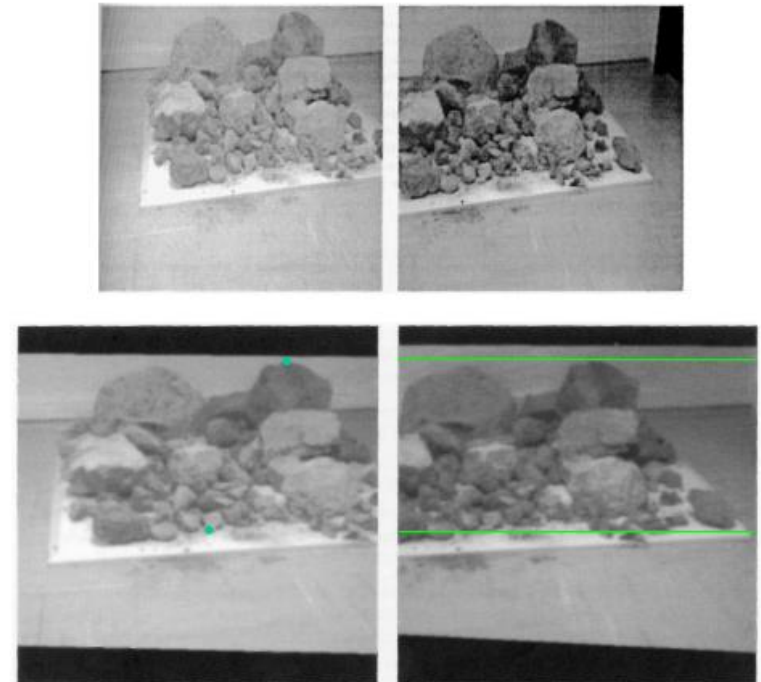
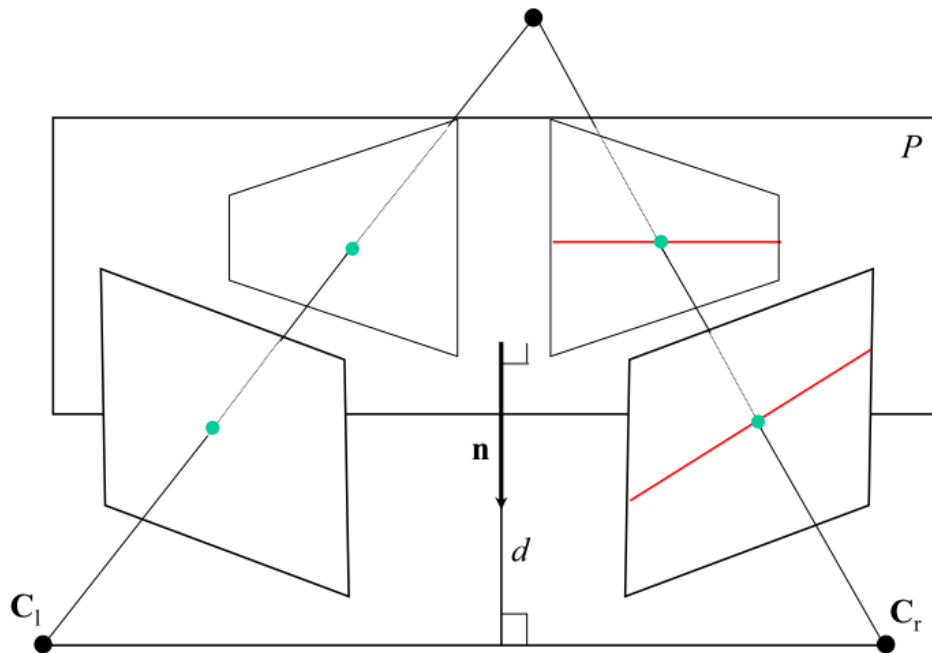
# Rectified images





# Rectification for stereo-image

- To projectively transfer images into a specific position

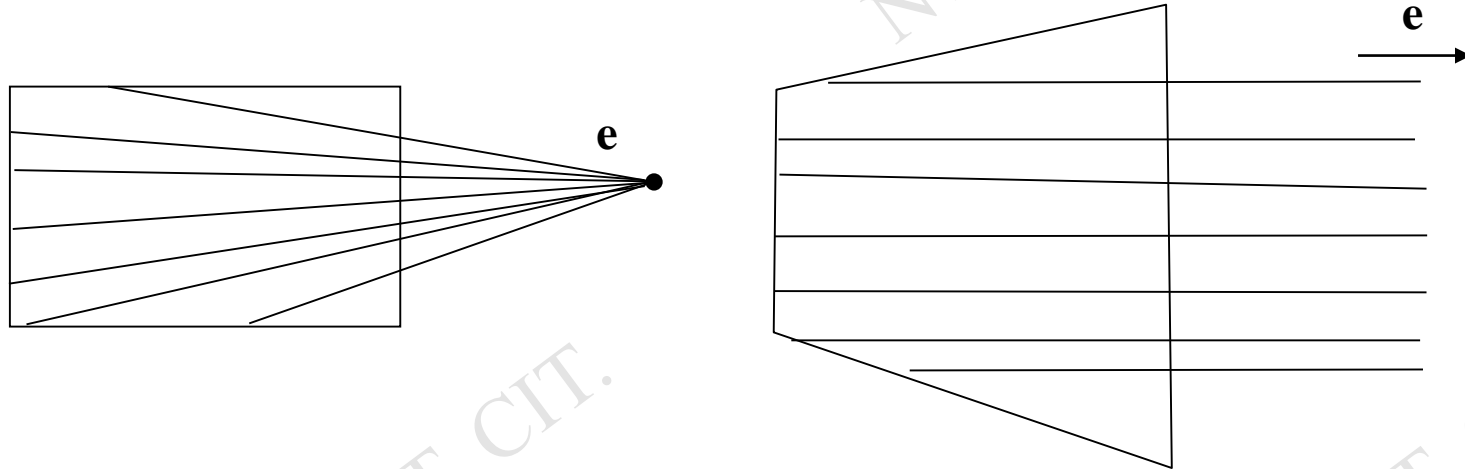






# Image rectification

- Apply projective transformation so that epipolar lines correspond to horizontal line

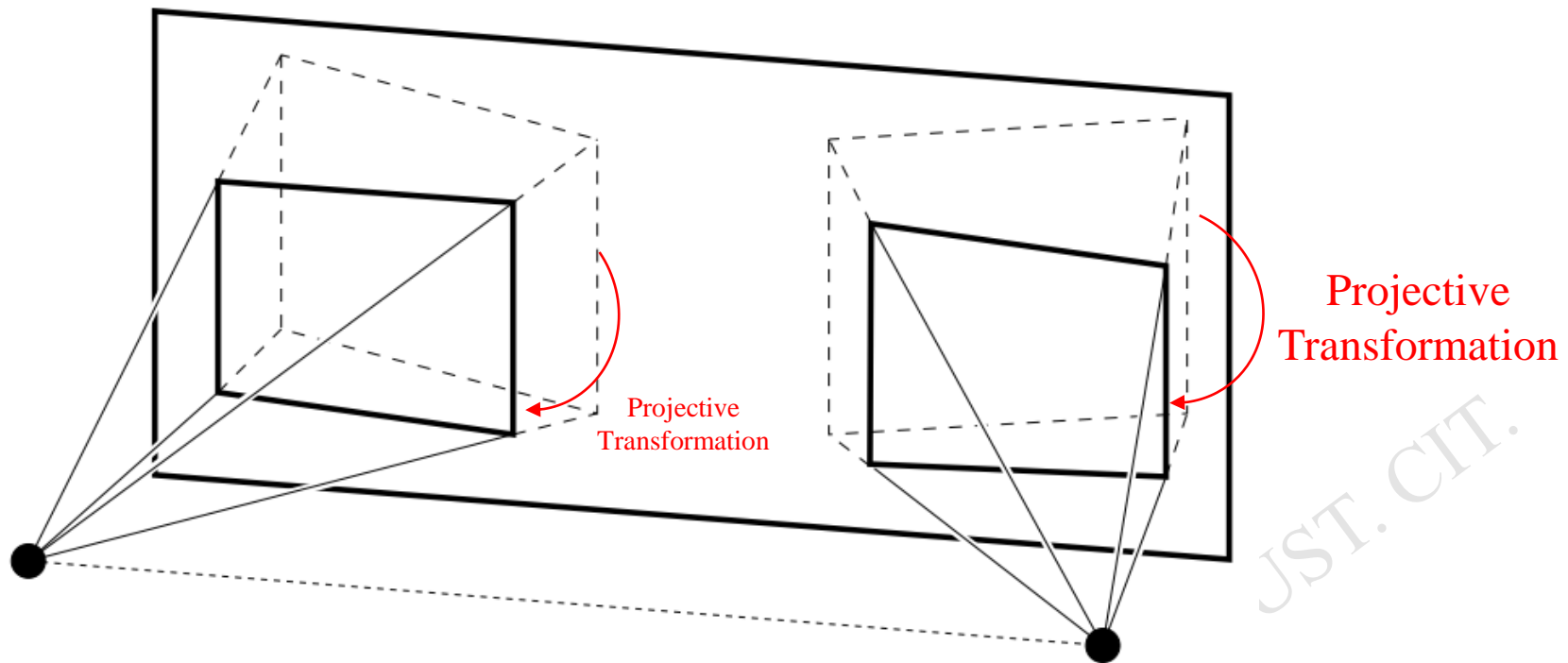


map epipole  $e$  to  $(1,0,0)^T$

try to minimize image distortion



# Image rectification

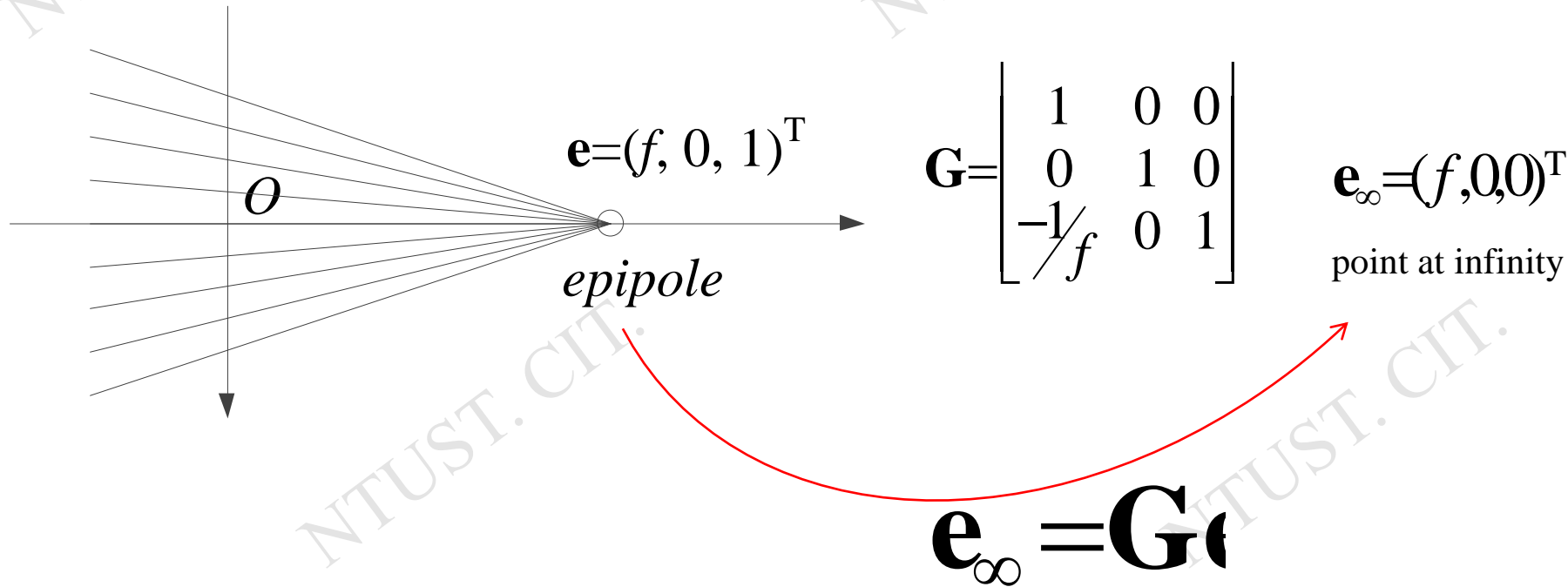


NOTE! This projective transformation is so called **Homography!**



# Image rectification—solution

- To transfer epipole to the point at infinity



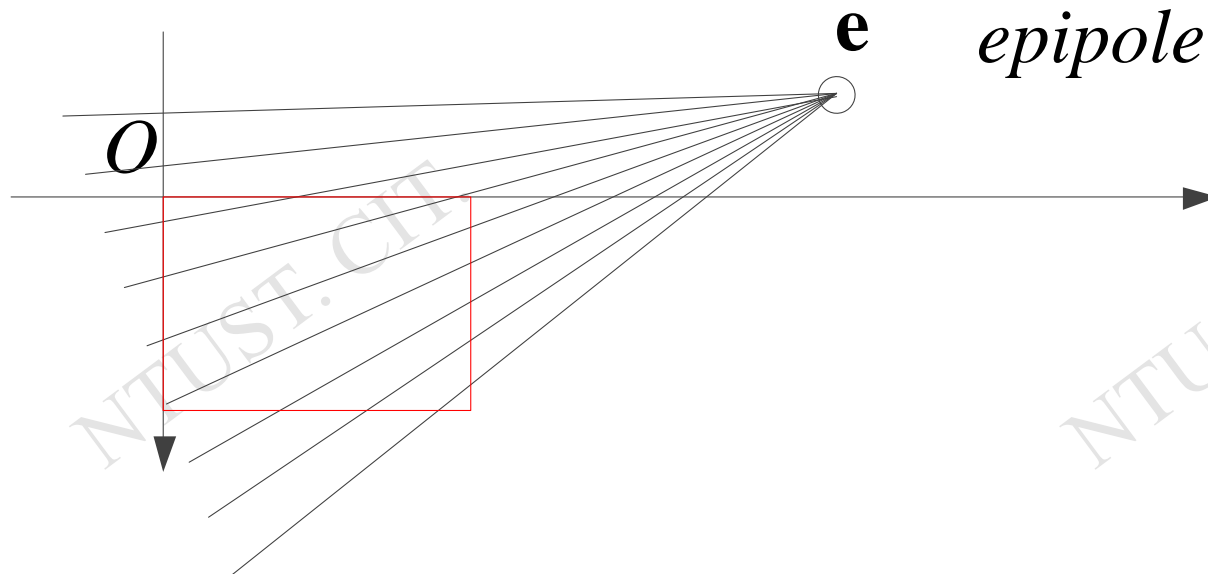
**Note! This is homography**





# Image rectification—solution, cont.

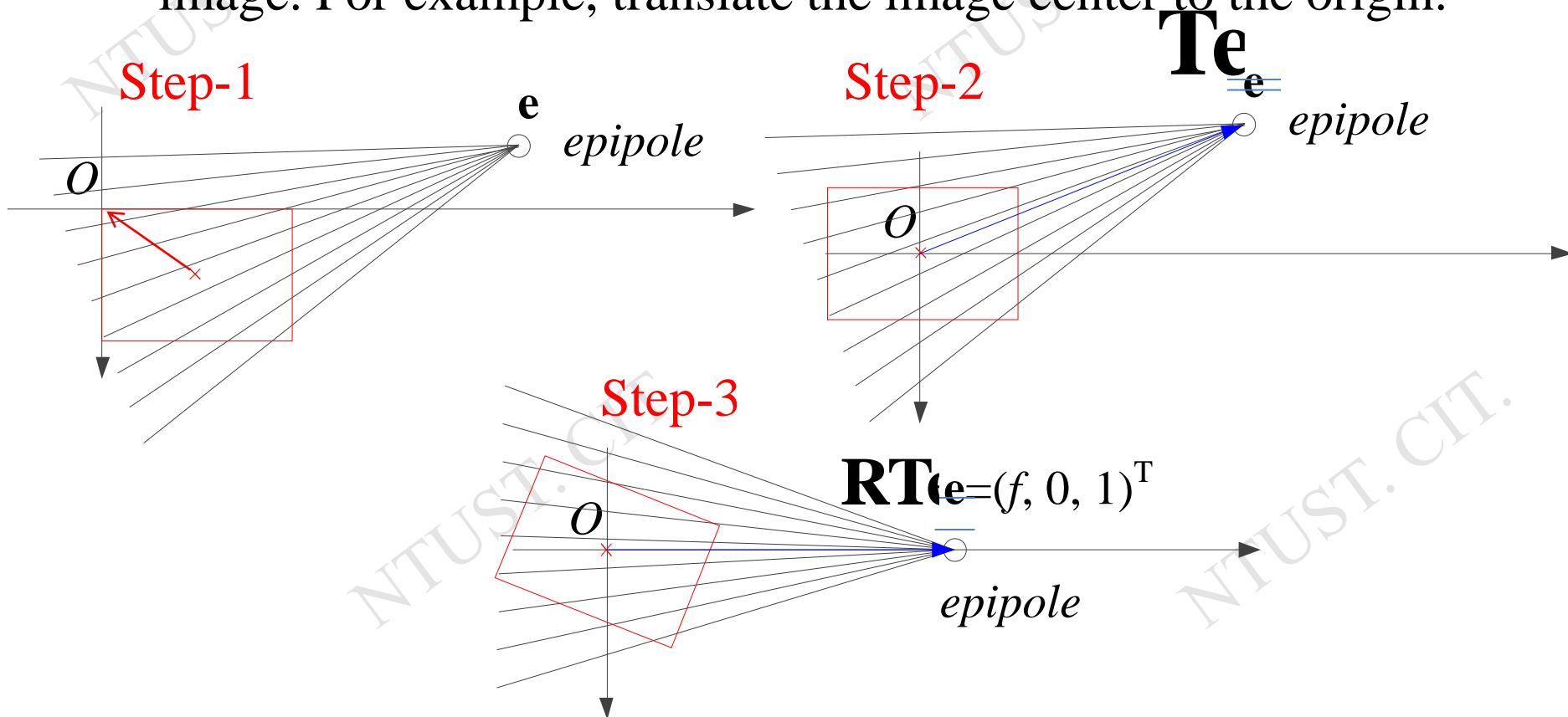
- However, in general, we have the configuration like the following figure.
- So, it needs a translation and a rotation to adjust the epipole on the special condition (on  $x$ -axis), and the homography will be derived as the format in the previous page.





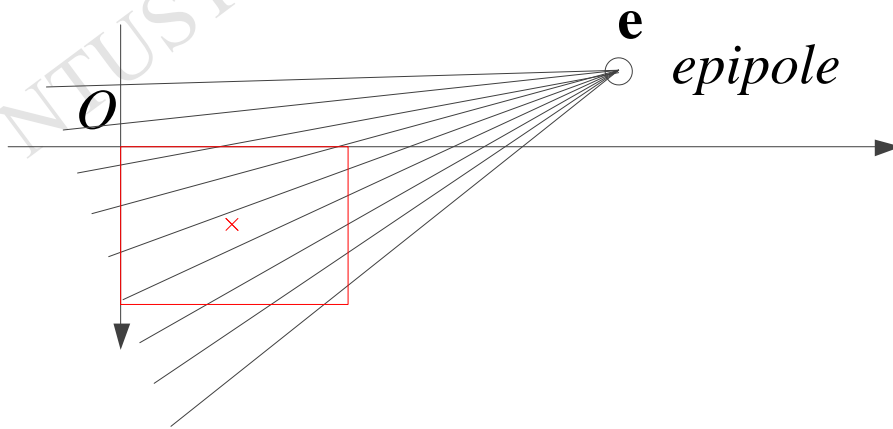
# Image rectification—solution, cont.

- An appropriate choice of translation would be the center of image. For example, translate the image center to the origin.





# Image rectification—solution, cont.



$$\mathbf{e}_{\infty} = \mathbf{H}\boldsymbol{\epsilon}$$

$$\mathbf{e}_{\infty} = (f, 0, 0)^T$$

point at infinity

$$\mathbf{e}_{\infty} = \underbrace{\mathbf{G}}_{\text{homography}} \mathbf{R} \mathbf{T}$$

$$\rightarrow \mathbf{e}_{\infty} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -1/f & 0 & 1 \end{bmatrix} \mathbf{R} \mathbf{T}$$



# Image rectification—solution, cont.

$$\rightarrow \mathbf{e}_{\infty} = \mathbf{H}\mathbf{e} = \mathbf{GRTe} \quad \mathbf{RTe} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -1/f & 0 & 1 \end{bmatrix} \mathbf{Te}$$

- Review the procedure:

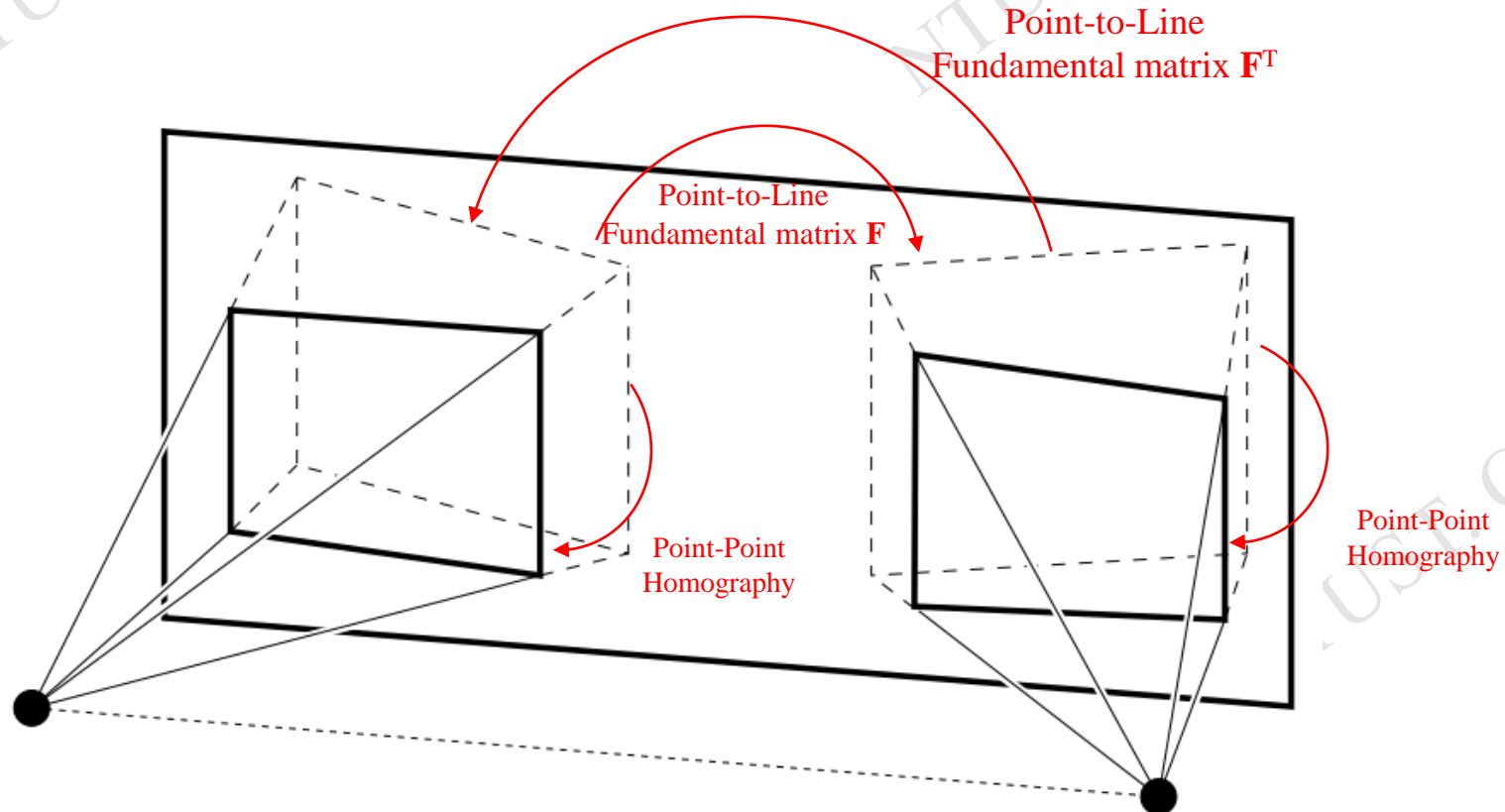
1. Determine one  $\mathbf{F}$  from two image
2. Determine  $\mathbf{e}$  (use cross product of two epipolar lines, line eqs:  $\mathbf{l} = \mathbf{F}^T \mathbf{x}'$ )
3. Determine  $\mathbf{T}$  (of course, you know the image resolution. use its center)
4. Determine  $\mathbf{R}$  (you already have  $\mathbf{Te}$ , rotation angle should be  $-\tan^{-1} \frac{y}{x}$ )
5. Then, get  $f$  from  $\mathbf{RTe}$ .
6. Finally, you have  $\mathbf{H}$ .

Note!  $\mathbf{H}$  is calculated from the projective mapping (homography) of point-point. If you need line mapping according this homography, use  $\mathbf{l}_{rect} = \mathbf{H}^T \mathbf{l}$



# Image rectification—solution, cont.

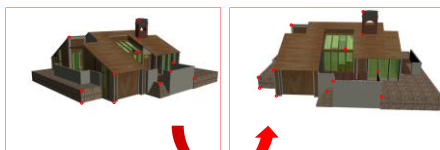
- Call the process, again.





# Image rectification—example

Recall the previous example.  
Image resolution 720x480.



$$\mathbf{F} = \begin{bmatrix} -0.000219 & -0.000913 & 0.292220 \\ 0.000103 & -0.000245 & 0.737529 \\ -0.142952 & -0.450960 & 1.000000 \end{bmatrix}$$

$$\mathbf{e} = \begin{bmatrix} -4089.085693 \\ 1298.432373 \\ 1.000000 \end{bmatrix}$$

$$\mathbf{e}' = \begin{bmatrix} -552.206970 \\ 217.436905 \\ 1.000000 \end{bmatrix}$$

For 1<sup>st</sup> image:

$$\mathbf{T} = \begin{bmatrix} 1 & 0 & -360 \\ 0 & 1 & -240 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\mathbf{T} * \mathbf{e} = \begin{bmatrix} -4449.085693 \\ 1058.432373 \\ 1.0 \end{bmatrix}$$

$$-\tan^{-1}\left(\frac{1058.432}{-4449.086}\right)$$

Rotation angle = 13.38°

$$\mathbf{R} = \begin{bmatrix} \cos(338) & -\sin(338) & 0 \\ \sin(338) & \cos(338) & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

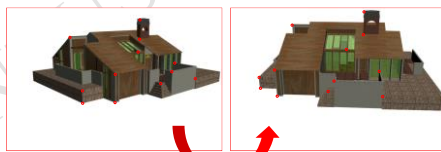
$$\mathbf{RTe} = \begin{bmatrix} -4573 & 0 & 1 \end{bmatrix}^T$$

$$\therefore \mathbf{H} = \mathbf{GRT} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -1/-4573 & 0 & 1 \end{bmatrix} \begin{bmatrix} \cos(338) & -\sin(338) & 0 \\ \sin(338) & \cos(338) & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & -360 \\ 0 & 1 & -240 \\ 0 & 0 & 1 \end{bmatrix}$$





# Image rectification—example, cont.



$$\mathbf{F} = \begin{bmatrix} -0.000219 & -0.000913 & 0.292220 \\ 0.000103 & -0.000245 & 0.737529 \\ -0.142952 & -0.450960 & 1.000000 \end{bmatrix}$$

$$\mathbf{e} = \begin{bmatrix} -4089.085693 \\ 1298.432373 \\ 1.000000 \end{bmatrix}$$

$$\mathbf{e}' = \begin{bmatrix} -552.206970 \\ 217.436905 \\ 1.000000 \end{bmatrix}$$

For 2<sup>nd</sup> image:

$$\mathbf{T} = \begin{bmatrix} 1 & 0 & -360 \\ 0 & 1 & -240 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\mathbf{T}' * \mathbf{e}' = \begin{bmatrix} -912.2070 \\ -22.5631 \\ 1.0000 \end{bmatrix}$$

$$-\tan^{-1}\left(\frac{-2256}{-912}\right)$$

Rotation angle = -1.42

$$\mathbf{R}' = \begin{bmatrix} \cos(1.42) & -\sin(1.42) & 0 \\ \sin(1.42) & \cos(1.42) & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

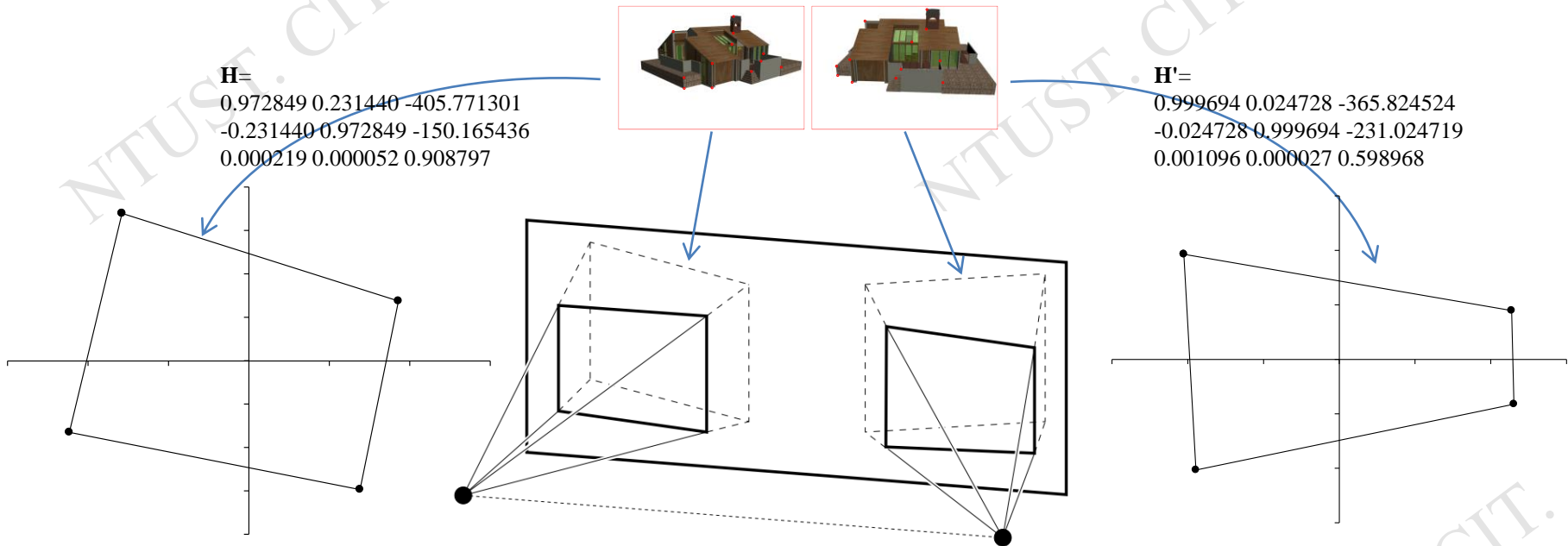
$$\therefore \mathbf{H} = \mathbf{GRT} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -1/912 & 2486 & 0 \end{bmatrix} \begin{bmatrix} \cos(1.42) & -\sin(1.42) & 0 \\ \sin(1.42) & \cos(1.42) & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & -360 \\ 0 & 1 & -240 \\ 0 & 0 & 1 \end{bmatrix}$$



# Image rectification—example, cont.

$\mathbf{H} =$   
0.972849 0.231440 -405.771301  
-0.231440 0.972849 -150.165436  
0.000219 0.000052 0.908797

$\mathbf{H}' =$   
0.999694 0.024728 -365.824524  
-0.024728 0.999694 -231.024719  
0.001096 0.000027 0.598968



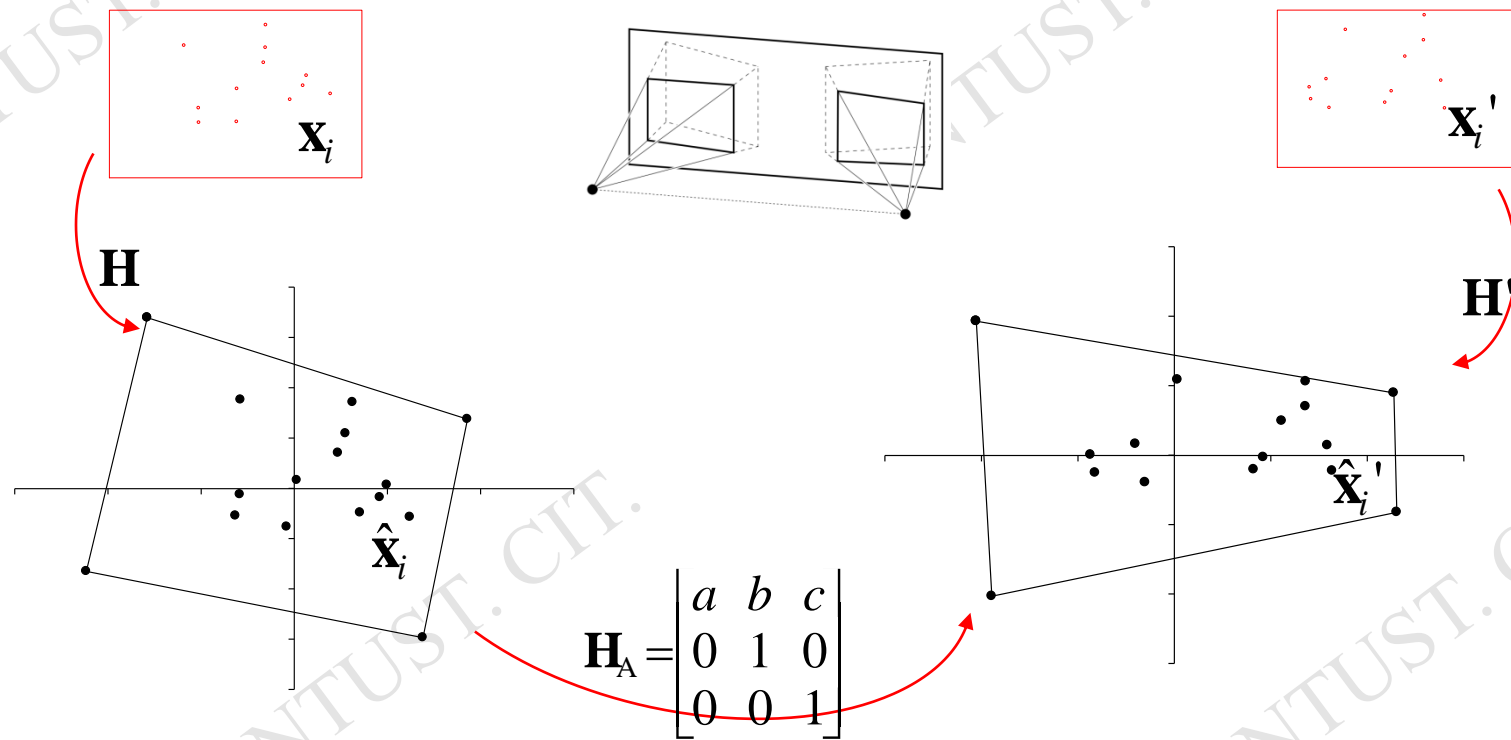
After rectification adjustment, two problems remain

1. Correspondences in two image may have disparity on y direction.
2. Pixel coordinates may not fall in positive region.



# Image rectification—example, cont.

- Minimize the horizontal disparity (minimize dist.)



Minimize  $\sum_i d(\mathbf{H}_A \hat{\mathbf{x}}_i, \hat{\mathbf{x}}'_i)^2$

Note! The minimization is to minimize “distance” of correspondences (that are  $\hat{\mathbf{x}}'_i, \mathbf{H}_A \hat{\mathbf{x}}_i$ )



# Image rectification—example, cont.

## ■ Minimize the horizontal disparity—cont.

$$\begin{aligned} \text{Minimize } & \sum_i d(\mathbf{H}_A \hat{\mathbf{x}}_i, \hat{\mathbf{x}}'_i)^2 \\ \rightarrow & \sum_i [(a\hat{x}_i + b\hat{y}_i + c - \hat{x}'_i)^2 + (\hat{y}_i - \hat{y}'_i)^2] \end{aligned}$$

$(\hat{y}_i - \hat{y}'_i)^2$  is a constant, and will not affect the chosen of solution for  $(a, b, c)$

So, minimization terms are reduced...

$$\text{Minimize } \sum_i (a\hat{x}_i + b\hat{y}_i + c - \hat{x}'_i)^2$$



# Image rectification—example, cont.

## ■ Minimize the horizontal disparity—cont.

$$\text{Minimize } \sum_i (a\hat{x}_i + b\hat{y}_i + c - \hat{x}_i')^2$$

For convenience

$$\text{let } g = \sum_i (a\hat{x}_i + b\hat{y}_i + c - \hat{x}_i')^2$$

To determine extreme value of  $g$ , take partial difference for  $g$ . (偏微分)

$$\begin{cases} \frac{\partial g}{\partial a} = 0 \\ \frac{\partial g}{\partial b} = 0 \\ \frac{\partial g}{\partial c} = 0 \end{cases} \left\{ \begin{array}{l} 2 \sum_i (a\hat{x}_i + b\hat{y}_i + c - \hat{x}_i') \hat{x}_i = 0 \\ 2 \sum_i (a\hat{x}_i + b\hat{y}_i + c - \hat{x}_i') \hat{y}_i = 0 \\ 2 \sum_i (a\hat{x}_i + b\hat{y}_i + c - \hat{x}_i') = 0 \end{array} \right\} \left\{ \begin{array}{l} (\sum_i \hat{x}_i^2) a + (\sum_i \hat{x}_i \hat{y}_i) b + (\sum_i \hat{x}_i) c = \sum_i \hat{x}_i \hat{x}_i' \\ (\sum_i \hat{x}_i \hat{y}_i) a + (\sum_i \hat{y}_i^2) b + (\sum_i \hat{y}_i) c = \sum_i \hat{y}_i \hat{x}_i' \\ (\sum_i \hat{x}_i) a + (\sum_i \hat{y}_i) b + (\sum_i 1) c = \sum_i \hat{x}_i' \end{array} \right.$$



# Image rectification—example, cont.

- Minimize the horizontal disparity—cont.

$$\begin{bmatrix} (\sum_i \hat{x}_i^2) & (\sum_i \hat{x}_i \hat{y}_i) & (\sum_i \hat{x}_i) \\ (\sum_i \hat{x}_i \hat{y}_i) & (\sum_i \hat{y}_i^2) & (\sum_i \hat{y}_i) \\ (\sum_i \hat{x}_i) & (\sum_i \hat{y}_i) & (\sum_i 1) \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} \sum_i \hat{x}_i \hat{x}_i' \\ \sum_i \hat{y}_i \hat{x}_i' \\ \sum_i \hat{x}_i' \end{bmatrix}$$

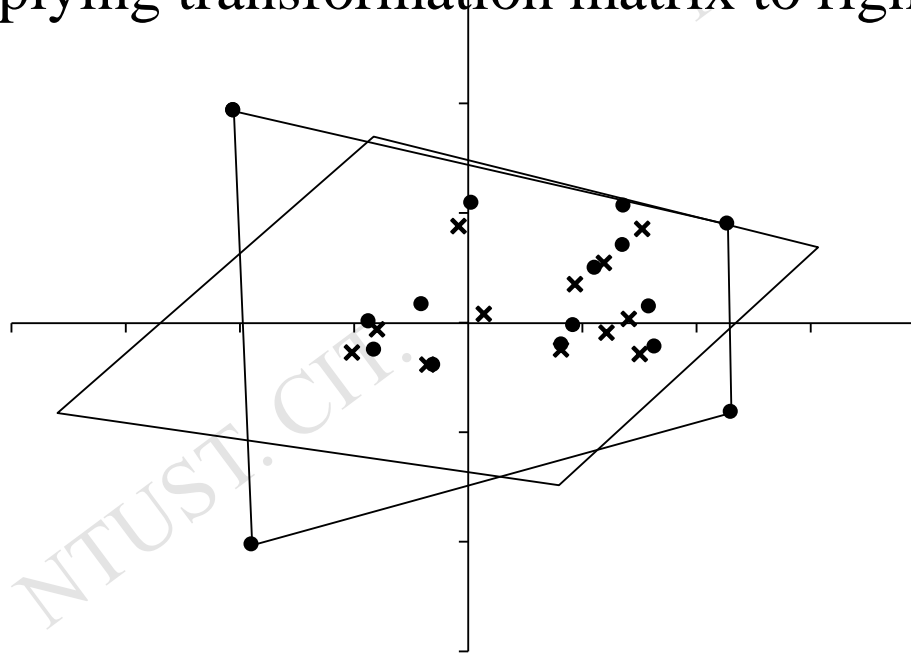
$$\begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} (\sum_i \hat{x}_i^2) & (\sum_i \hat{x}_i \hat{y}_i) & (\sum_i \hat{x}_i) \\ (\sum_i \hat{x}_i \hat{y}_i) & (\sum_i \hat{y}_i^2) & (\sum_i \hat{y}_i) \\ (\sum_i \hat{x}_i) & (\sum_i \hat{y}_i) & (\sum_i 1) \end{bmatrix}^{-1} \begin{bmatrix} \sum_i \hat{x}_i \hat{x}_i' \\ \sum_i \hat{y}_i \hat{x}_i' \\ \sum_i \hat{x}_i' \end{bmatrix}$$





# Image rectification—example, cont.

- Minimize the horizontal disparity—cont.
- After applying transformation matrix to right image:





# Image rectification—example, cont.

- Minimize the vertical disparity, as well as horizontal distance.

$$\text{Minimize } \sum_i d(\mathbf{H}_A \hat{\mathbf{x}}_i, \hat{\mathbf{x}}'_i)^2$$

$$\rightarrow \sum_i [(a\hat{x}_i + b\hat{y}_i + c - \hat{x}'_i)^2 + (d\hat{y}_i + e - \hat{y}'_i)^2]$$

$$\mathbf{H}_A = \begin{bmatrix} a & b & c \\ 0 & d & e \\ 0 & 0 & 1 \end{bmatrix}$$

$$\begin{cases} \frac{\partial}{\partial a} = 0 \\ \frac{\partial}{\partial b} = 0 \\ \frac{\partial}{\partial c} = 0 \\ \frac{\partial}{\partial d} = 0 \\ \frac{\partial}{\partial e} = 0 \end{cases} \rightarrow \begin{cases} 2 \sum_i (a\hat{x}_i + b\hat{y}_i + c - \hat{x}'_i) \hat{x}_i = 0 \\ 2 \sum_i (a\hat{x}_i + b\hat{y}_i + c - \hat{x}'_i) \hat{y}_i = 0 \\ 2 \sum_i (a\hat{x}_i + b\hat{y}_i + c - \hat{x}'_i) = 0 \\ 2 \sum_i (d\hat{y}_i + e - \hat{y}'_i) \hat{y}_i = 0 \\ 2 \sum_i (d\hat{y}_i + e - \hat{y}'_i) = 0 \end{cases} \rightarrow \begin{cases} (\sum_i \hat{x}_i^2) a + (\sum_i \hat{x}_i \hat{y}_i) b + (\sum_i \hat{x}_i) c = \sum_i \hat{x}_i \hat{x}'_i \\ (\sum_i \hat{x}_i \hat{y}_i) a + (\sum_i \hat{y}_i^2) b + (\sum_i \hat{y}_i) c = \sum_i \hat{y}_i \hat{x}'_i \\ (\sum_i \hat{x}_i) a + (\sum_i \hat{y}_i) b + (\sum_i 1) c = \sum_i \hat{x}'_i \\ (\sum_i \hat{y}_i^2) d + (\sum_i \hat{y}_i) e = \sum_i \hat{y}_i \hat{y}'_i \\ (\sum_i \hat{y}_i) d + (\sum_i 1) e = \sum_i \hat{y}'_i \end{cases}$$



# Image rectification—example, cont.

- Minimize the vertical disparity, as well as horizontal distance.—cont.

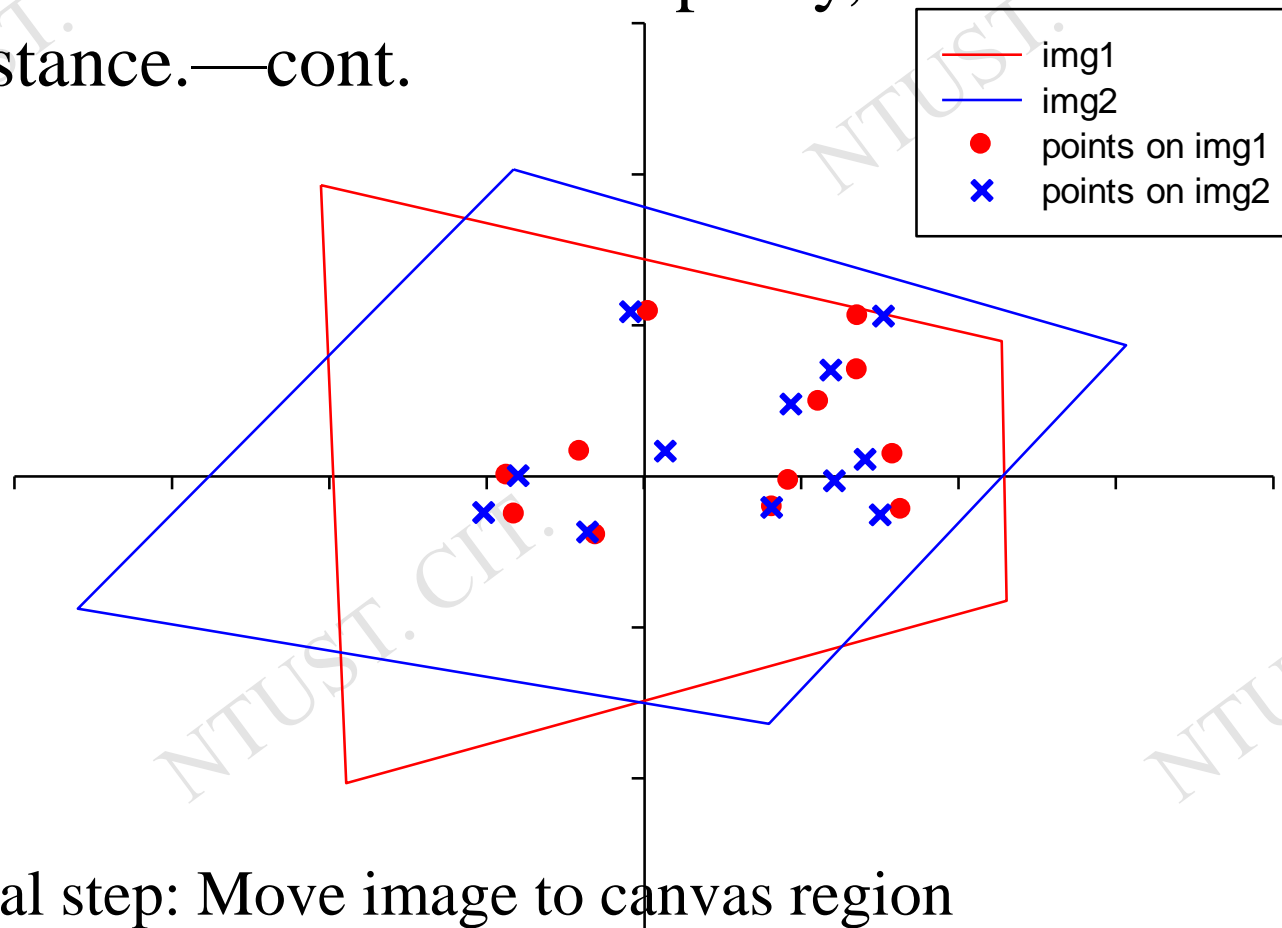
$$\begin{bmatrix} \sum_i \hat{x}_i^2 & \sum_i \hat{x}_i \hat{y}_i & \sum_i \hat{x}_i & 0 & 0 \\ \sum_i \hat{x}_i \hat{y}_i & \sum_i \hat{y}_i^2 & \sum_i \hat{y}_i & 0 & 0 \\ \sum_i \hat{x}_i & \sum_i \hat{y}_i & \sum_i 1 & 0 & 0 \\ 0 & 0 & 0 & \sum_i \hat{y}_i^2 & \sum_i \hat{y}_i \\ 0 & 0 & 0 & \sum_i \hat{y}_i & \sum_i 1 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \\ d \\ e \end{bmatrix} = \begin{bmatrix} \sum_i \hat{x}_i \hat{x}_i' \\ \sum_i \hat{y}_i \hat{x}_i' \\ \sum_i \hat{x}_i' \\ \sum_i \hat{y}_i \hat{y}_i' \\ \sum_i \hat{y}_i' \end{bmatrix} \Rightarrow \begin{bmatrix} a \\ b \\ c \\ d \\ e \end{bmatrix} = \begin{bmatrix} \sum_i \hat{x}_i^2 & \sum_i \hat{x}_i \hat{y}_i & \sum_i \hat{x}_i & 0 & 0 \\ \sum_i \hat{x}_i \hat{y}_i & \sum_i \hat{y}_i^2 & \sum_i \hat{y}_i & 0 & 0 \\ \sum_i \hat{x}_i & \sum_i \hat{y}_i & \sum_i 1 & 0 & 0 \\ 0 & 0 & 0 & \sum_i \hat{y}_i^2 & \sum_i \hat{y}_i \\ 0 & 0 & 0 & \sum_i \hat{y}_i & \sum_i 1 \end{bmatrix}^{-1} \begin{bmatrix} \sum_i \hat{x}_i \hat{x}_i' \\ \sum_i \hat{y}_i \hat{x}_i' \\ \sum_i \hat{x}_i' \\ \sum_i \hat{y}_i \hat{y}_i' \\ \sum_i \hat{y}_i' \end{bmatrix}$$

Finally, recover  $\mathbf{H}_A = \begin{bmatrix} a & b & c \\ 0 & d & e \\ 0 & 0 & 1 \end{bmatrix}$



# Image rectification—example, cont.

- Minimize the vertical disparity, as well as horizontal distance.—cont.

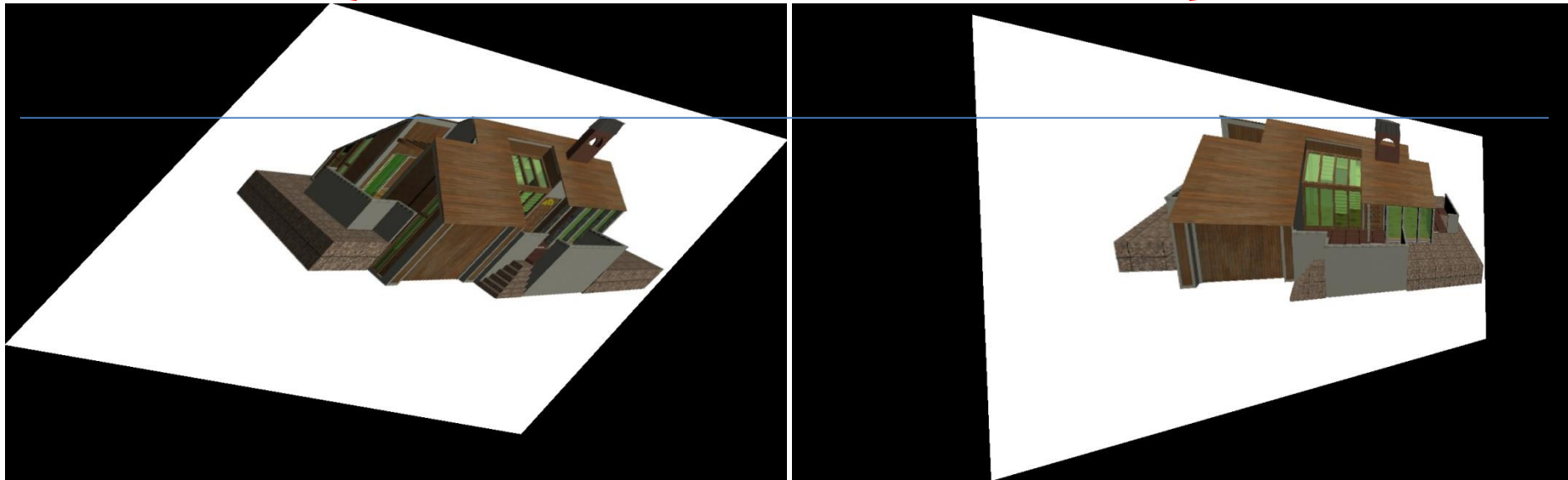
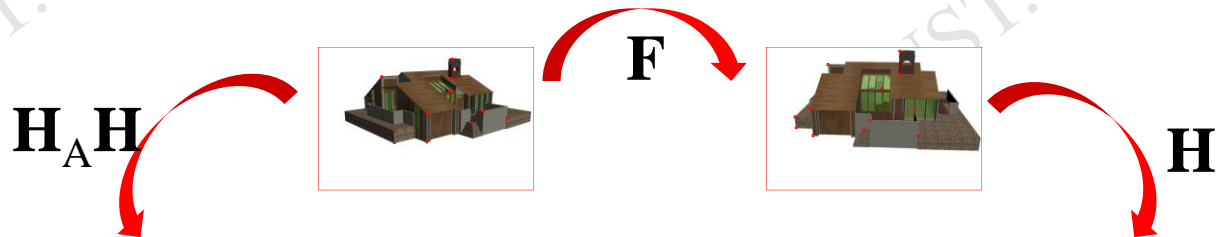


Final step: Move image to canvas region



# Image rectification—example, cont.

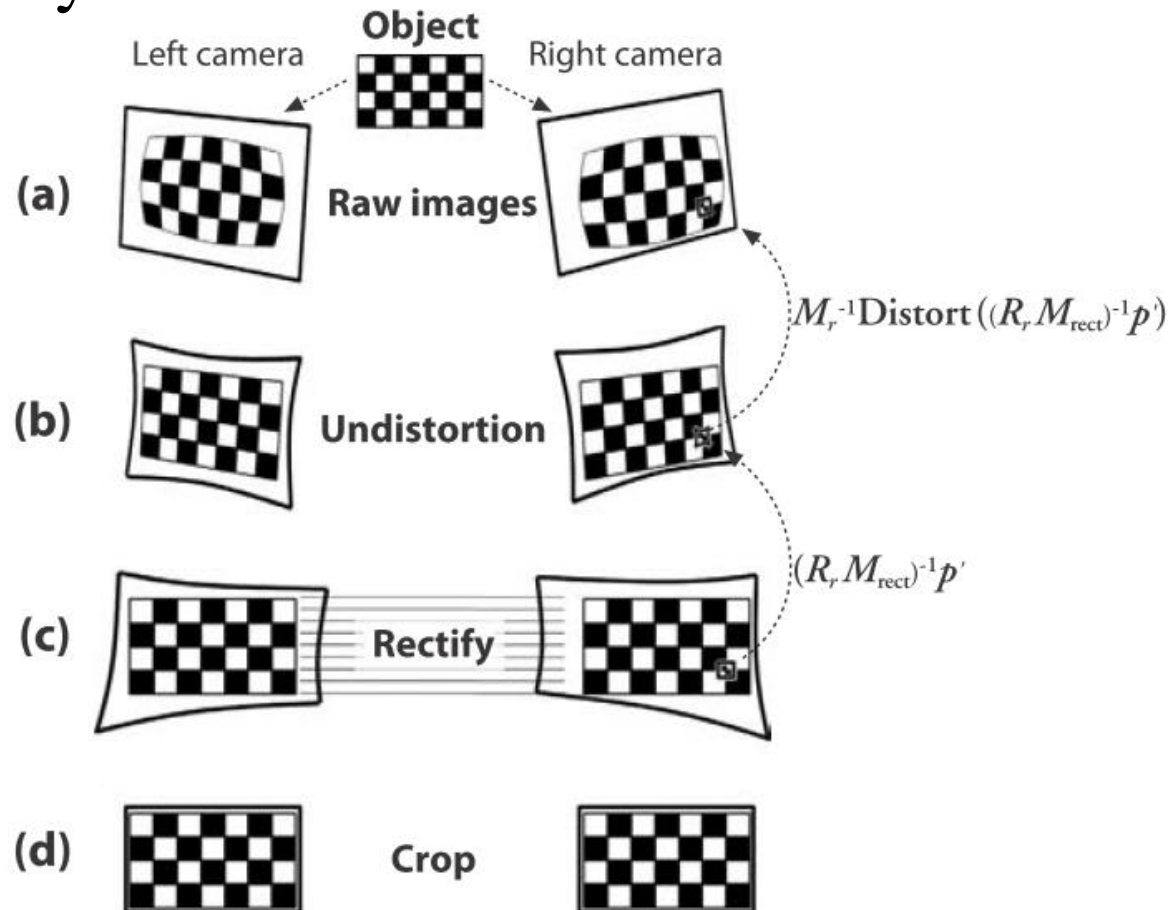
## ■ Unique solution?





# Image rectification—openCV

## ■ StereoRectifyUncalibrated







# Image rectification—openCV

## ■ The same operation in OpenCV

*cvStereoRectifyUncalibrated*

```
int cvStereoRectifyUncalibrated(  
    const CvMat* points1,   
    const CvMat* points2,   
    const CvMat* F,   
    CvSize imageSize,   
    CvMat* Hl,   
    CvMat* Hr,   
    double threshold  
);
```

→ The 2 arrays of corresponding 2D points. (input)

→ Fundamental Matrix (input)

→ (input)

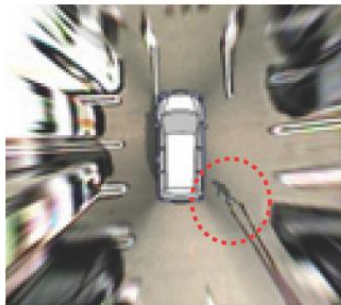
→ Homography Matrix (output for Left, Right Images)

→ (input) for rejecting the outliers  
for which  $|\mathbf{x}'^T \mathbf{F} \mathbf{x}| > \text{threshold}$

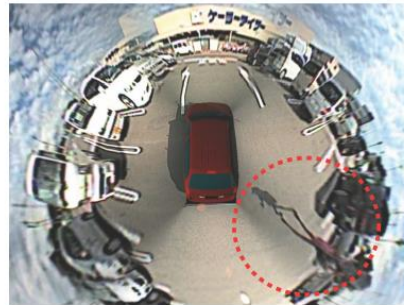


# Homography—Applications

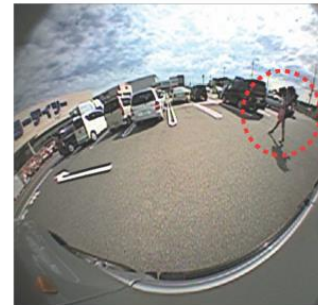
## ■ Intelligent automobile



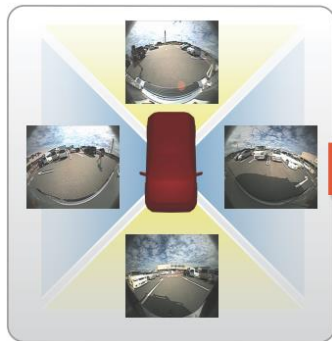
Conventional technology-based image, vehicles and people are not visible.



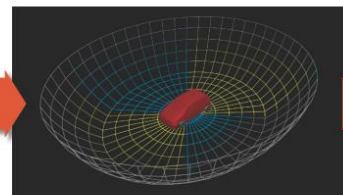
Sample view using Fujitsu Laboratories' new technology perspective from above-rear (pedestrian is visible)



Sample view using Fujitsu's new technology, perspective from front facing vehicle (rearview pedestrian is visible)



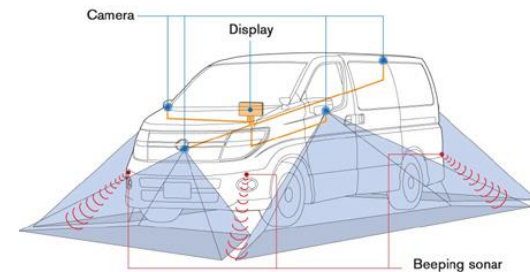
4 cameras capture video images of 4 different views (perspectives)



Virtual 3D model for projection of video images is synthesized (scene is projected virtually onto a 3D curved plane). Image is changed to the desired perspective



Desired view (perspective) is displayed



NISSAN





# Homography—Applications

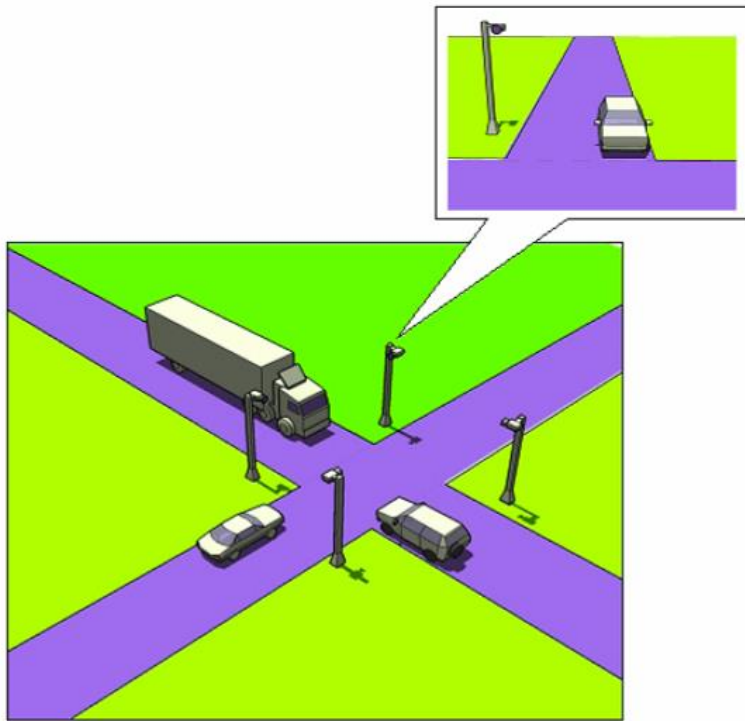
## ■ Intelligent automobile—cont.



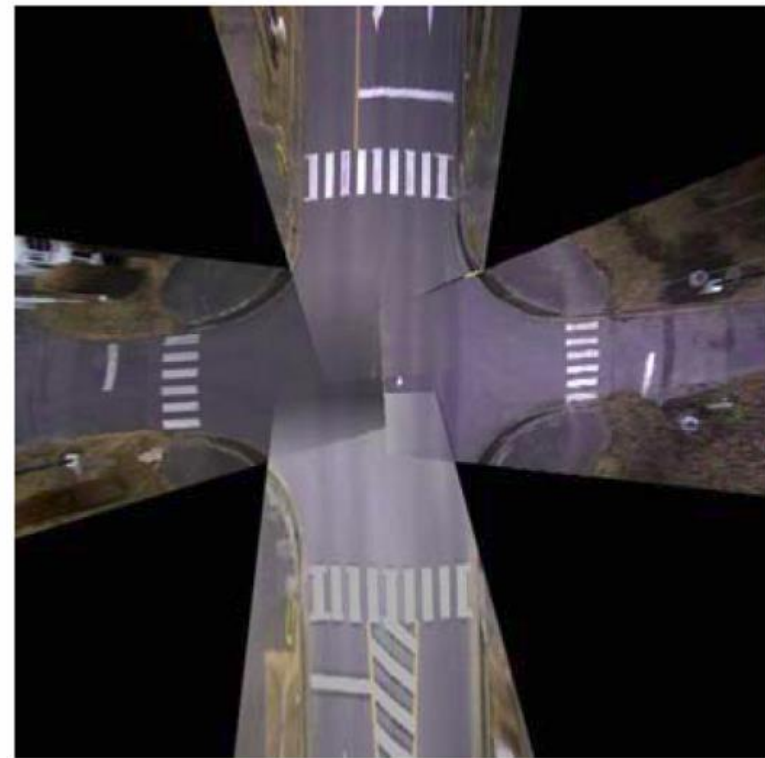


# Homography—Applications

- Multi-camera surveillance
- 物聯網



筑波大學(JP)



筑波大學(JP)



# Homography—Applications

## ■ Application scenarios

**Image-  
Undistortion**



**Homography**



**Image stitching**



**Driving / navigation**  
(Auto or assistance)

**Safety driving**

.....

**Auto parking /  
Parking assistant  
system**

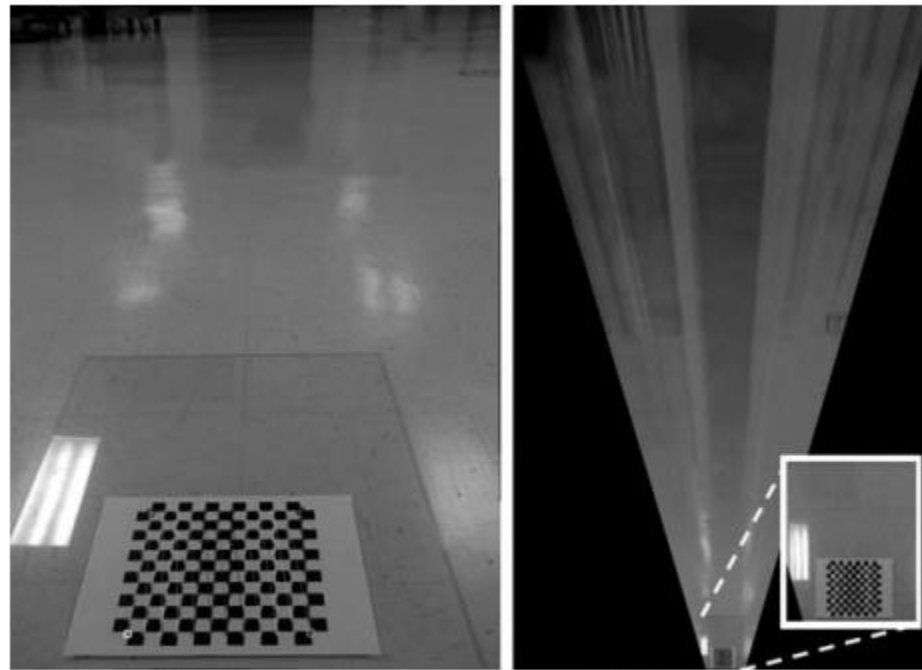
**Obstacle detection**  
(distance measurement)

Note! In many application, wide-angle lens (fish-eye) cameras are used. In these situations, the UN-distortion processing is important.



# Homography—Applications

- Once again, define your problem first!



View of a planar surface

Bird's-eye view

Homography

1. Pre-processing or post-processing for  $\mathbf{H}$
2. Constant  $\mathbf{H}$  or various  $\mathbf{H}$

Solution: (homography)

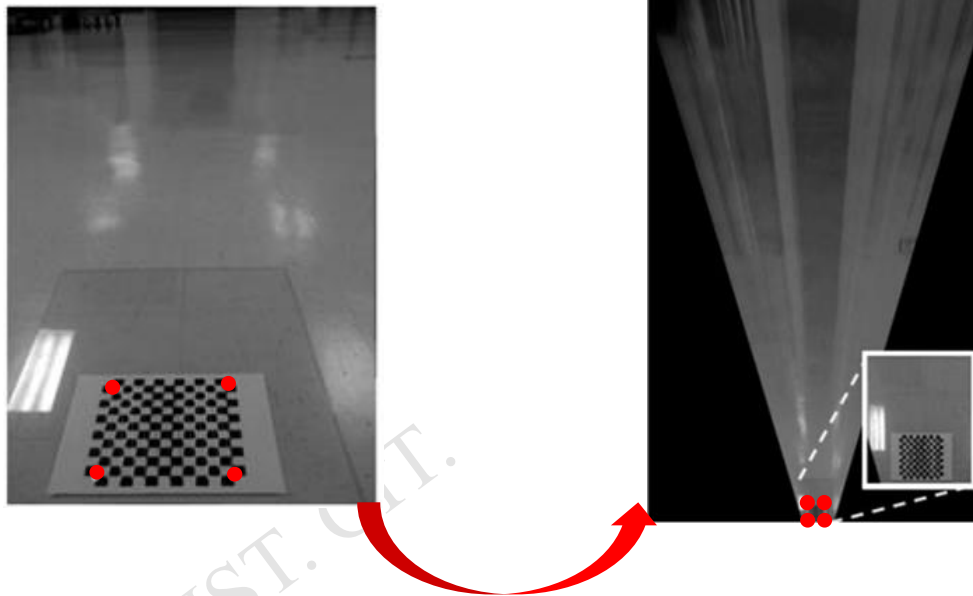
1. Line cue ? Point cue ?
2. Correspondence
3. Scale issue





# Homography—Applications

## ■ Solution for finding $\mathbf{H}$



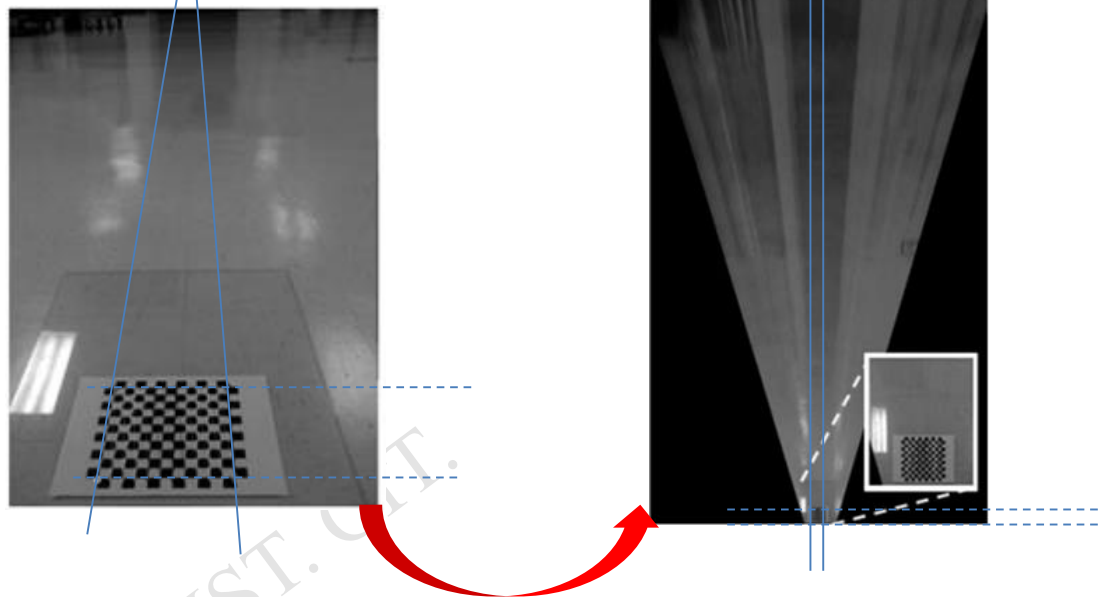
Possible method

Manually assign at least 4 correspondences (define your new image-resolution well)



# Homography—Applications

## ■ Solution for finding $H$ —cont.



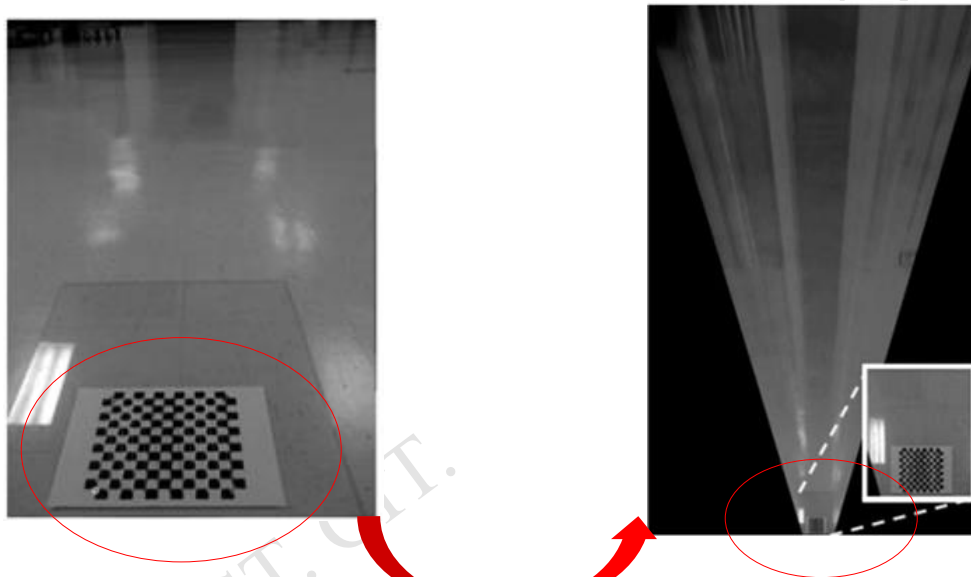
Possible method

Find vanishing points from parallel structure → can be an automatic method based on the line detection algorithm (line fitting, as well)



# Homography—Applications

## ■ Solution for finding $\mathbf{H}$ —cont.



Pattern/object recognition

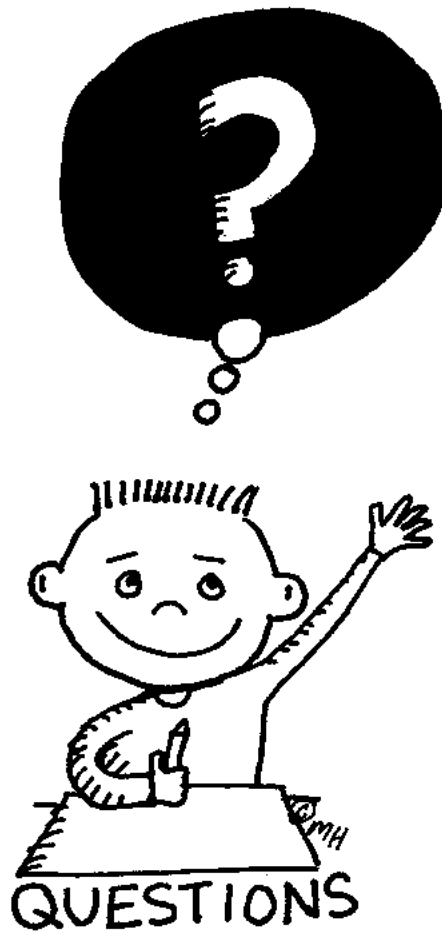
For a very simple example, the checkerboard has the fixed size and parallel grids. If you have an algorithm to determine the corners on the checkerboard, it will be easy to have correspondences for finding  $\mathbf{H}$ .



# Homography—Applications

## ■ Dynamic seethroughs (homography 應用)





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