

電腦視覺與應用

Computer Vision and Applications

Lecture08-3D reconstruction

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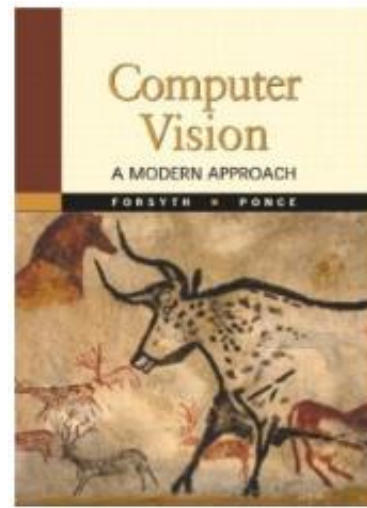
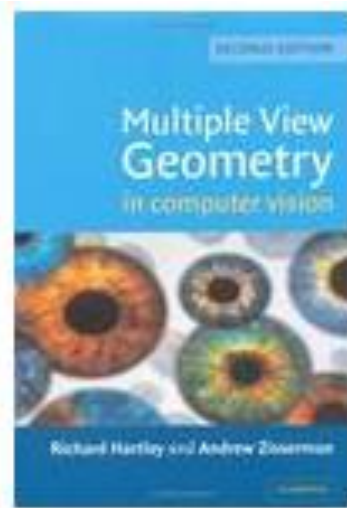
3D reconstruction

- Either from calibrated images
- Or from uncalibrated images



3D reconstruction

- Lecture Reference at:
 - Multiple View Geometry in Computer Vision,
(*Chapter10 and 12)
 - Computer Vision A Modern Approach, Chapter 13





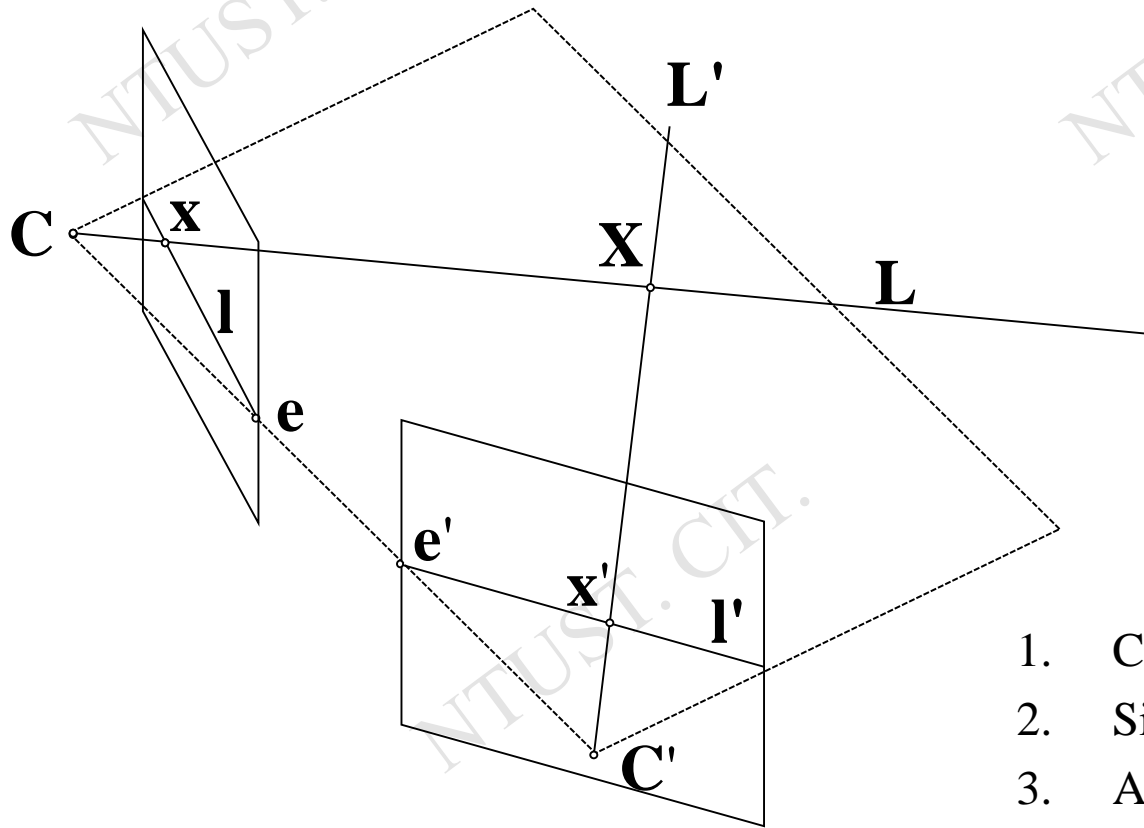
3D reconstruction

- The questions?
 - 1) **Correspondence geometry:** Given an image point \mathbf{x} in the first image, how does this constrain the position of the corresponding point \mathbf{x}' in the second image?
 - 2) **Camera geometry (motion):** Given a set of corresponding image points $\{\mathbf{x}_i \leftrightarrow \mathbf{x}'_i\}$, $i=1, \dots, n$, what are the cameras \mathbf{P} and \mathbf{P}' for the two views?
 - 3) **Scene geometry (structure):** Given corresponding image points $\mathbf{x}_i \leftrightarrow \mathbf{x}'_i$ and cameras \mathbf{P}, \mathbf{P}' , what is the position of \mathbf{X} in space?



3D reconstruction

■ Epipolar geometry



$$\underbrace{\mathbf{x}'^T}_{\mathbf{l}'^T} \underbrace{\mathbf{F}}_{\mathbf{l}'} \mathbf{x} = 0$$

1. Computable from corresponding points
2. Simplifies matching
3. Allows to detect outliers
4. Related to calibration



3D reconstruction of cameras and structure

■ Reconstruction problem:

Given $\mathbf{x}_i \leftrightarrow \mathbf{x}'_i$, compute \mathbf{P} , \mathbf{P}' and \mathbf{X}_i

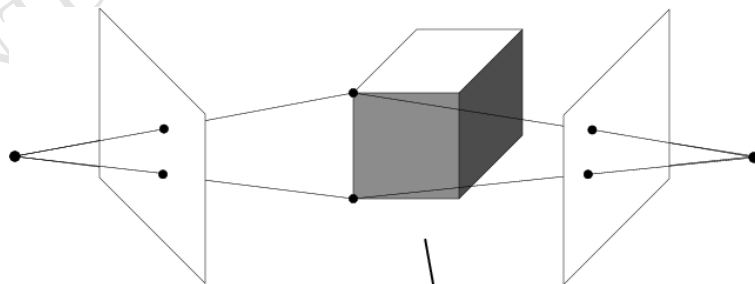
$$\mathbf{x}_i = \mathbf{P}\mathbf{X}_i \text{ and } \mathbf{x}'_i = \mathbf{P}'\mathbf{X}_i \text{ for all } i$$

without additional constraints possible up to projective ambiguity

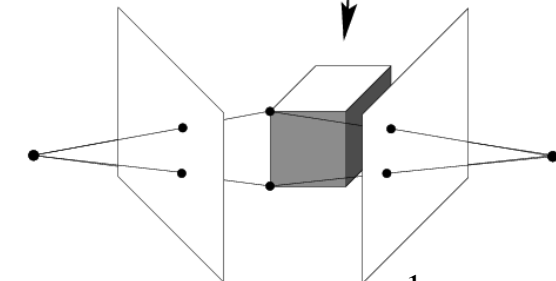


3D reconstruction of cameras and structure

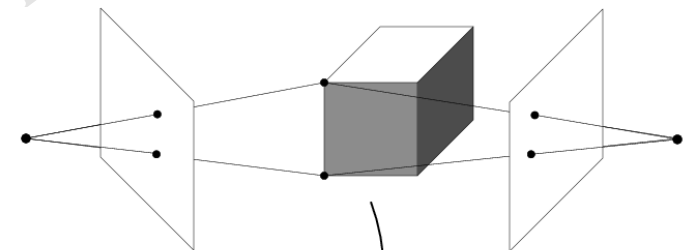
■ Reconstruction ambiguity (projective ambi.)



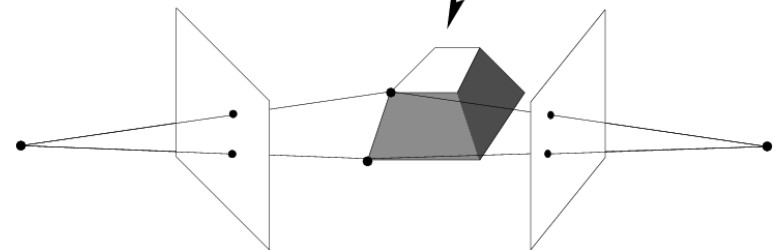
Similarity



$$\mathbf{x}_i = \mathbf{P}\mathbf{X}_i = (\mathbf{P}\mathbf{H}_S^{-1})(\mathbf{H}_S\mathbf{X}_i)$$



Projective



$$\mathbf{x}_i = \mathbf{P}\mathbf{X}_i = (\mathbf{P}\mathbf{H}_P^{-1})(\mathbf{H}_P\mathbf{X}_i)$$

One solution which is projected to \mathbf{x}_i , as well.



Outline of 3D reconstruction

(from uncalibrated images)

- (i) Compute \mathbf{F} from correspondences
- (ii) Compute camera matrices from \mathbf{F}
- (iii) Compute 3D point for each pair of corresponding points

■ computation of \mathbf{F}

use $\mathbf{x}'_i \mathbf{F} \mathbf{x}_i = 0$ equations, linear in coeff. \mathbf{F}

■ computation of camera matrices

use $\mathbf{P} = [\mathbf{I} \mid 0] \quad \mathbf{P}' = [[\mathbf{e}']_{\times} \mathbf{F} + \mathbf{e}' \mathbf{v}^T \mid \lambda \mathbf{e}'] \longrightarrow$ General formula

or $\mathbf{P} = [\mathbf{I} \mid 0] \quad \mathbf{P}' = [[\mathbf{e}']_{\times} \mathbf{F} \mid \mathbf{e}']$

■ triangulation

compute intersection of two backprojected rays

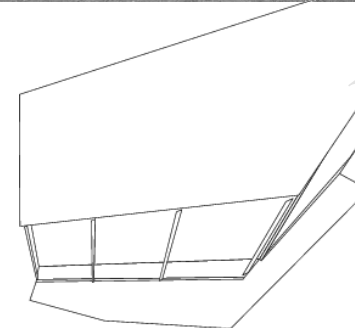
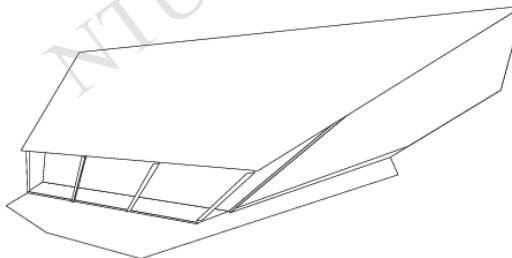
detail at

Q. Luong and T. Vieville, "Canonical representations for the geometries of multiple projective views," *Computer vision and image understanding*, vol. 64, no. 2, pp. 193-229, 1996.



Projective reconstruction

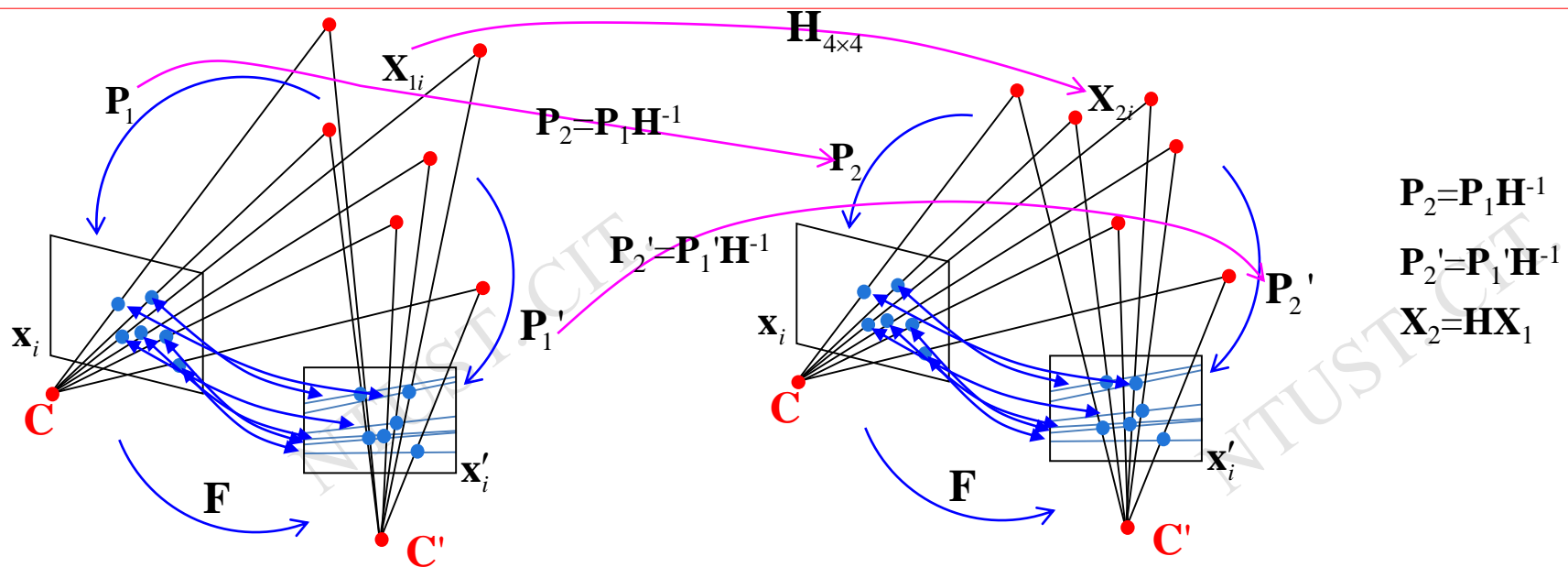
- The reconstruction required no information about camera matrices, or information about the scene geometry. The fundamental matrix F is computed from point correspondences between the images, camera matrices are retrieved from F , and then 3D points are computed by triangulation from the correspondences.





Theorem for the projective reconstruction

Suppose that $\mathbf{x}_i \leftrightarrow \mathbf{x}'_i$ is a set of correspondences between points in two images and that the fundamental matrix \mathbf{F} is uniquely determined by the condition $\mathbf{x}'_i{}^T \mathbf{F} \mathbf{x}_i = 0$ for all i . Let $(\mathbf{P}_1, \mathbf{P}'_1, \{\mathbf{X}_{1i}\})$ and $(\mathbf{P}_2, \mathbf{P}'_2, \{\mathbf{X}_{2i}\})$ be two reconstructions of the correspondences $\mathbf{x}_i \leftrightarrow \mathbf{x}'_i$. Then there exists a non-singular matrix \mathbf{H} such that $\mathbf{P}_2 = \mathbf{P}_1 \mathbf{H}^{-1}$, $\mathbf{P}'_2 = \mathbf{P}'_1 \mathbf{H}^{-1}$ and $\mathbf{X}_2 = \mathbf{H} \mathbf{X}_1$ for all i , except for those i such that $\mathbf{F} \mathbf{x}_i = \mathbf{x}'_i{}^T \mathbf{F} = 0$



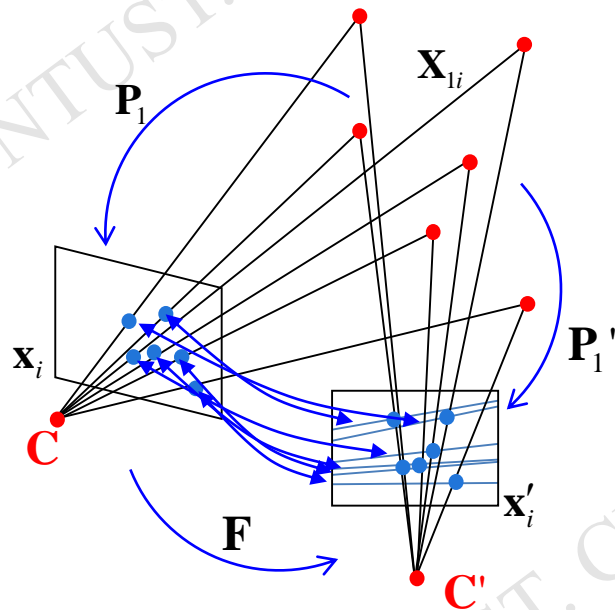
key result:

allows reconstruction from pair of uncalibrated images

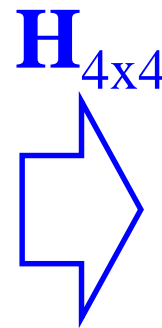


Theorem for the projective reconstruction

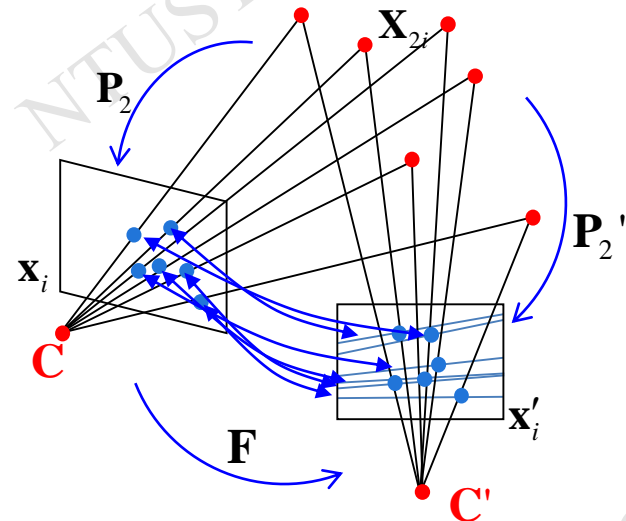
■ In practice



[Projective reconstruction]



How to get \mathbf{H} ?



Affine reconstruction
or Metric reconstruction

Step-1: solve epipolar geometry (\mathbf{F})

Step-2: guess \mathbf{P} & \mathbf{P}' for initialization, then determine \mathbf{X} .

Step-3: determine \mathbf{H} , then, finding new \mathbf{P} & \mathbf{P}' & \mathbf{X} .

$$\mathbf{P}_2 = \mathbf{P}_1 \mathbf{H}^{-1}$$

$$\mathbf{P}_2' = \mathbf{P}_1' \mathbf{H}^{-1}$$

$$\mathbf{X}_2 = \mathbf{H} \mathbf{X}_1$$

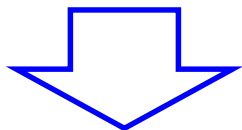


3D reconstruction for projective geometry

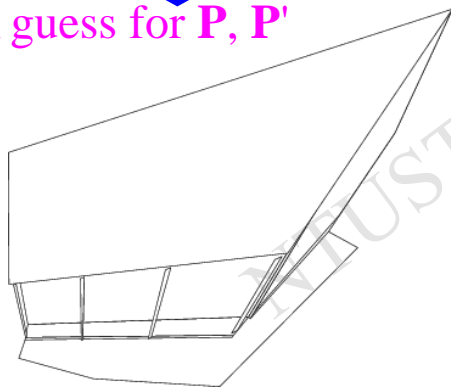
■ (in practice)



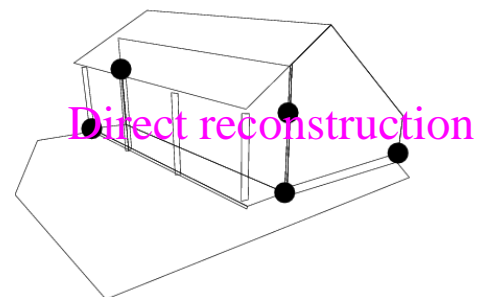
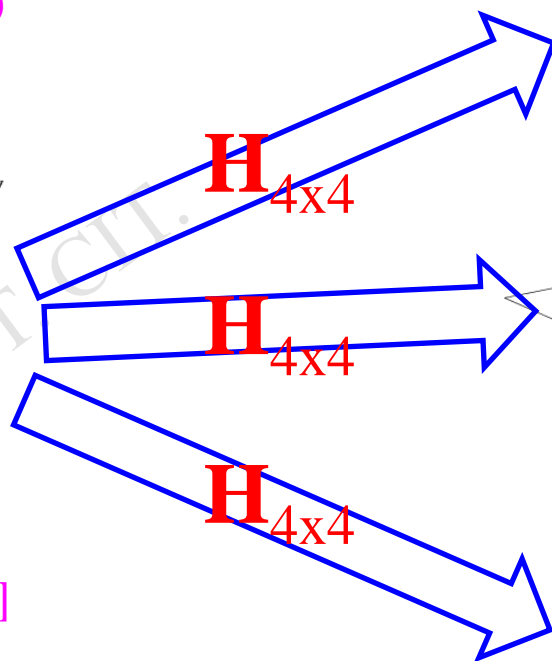
Two views geometry (F)



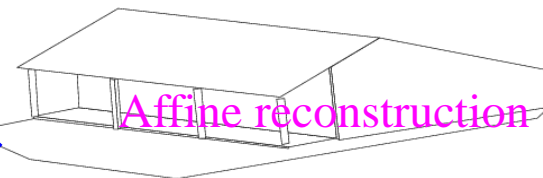
A guess for \mathbf{P} , \mathbf{P}'



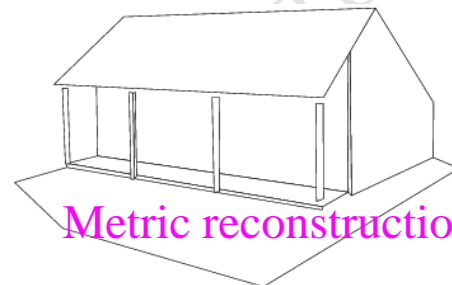
[Projective reconstruction]



Direct reconstruction



Affine reconstruction

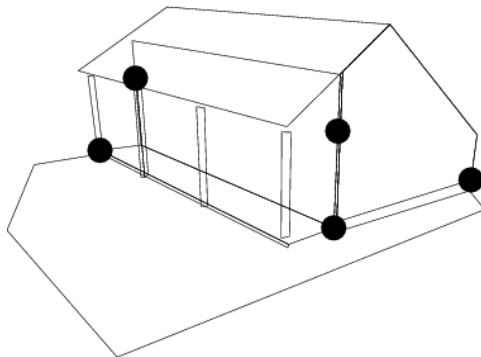


Metric reconstruction



Direct reconstruction using ground truth

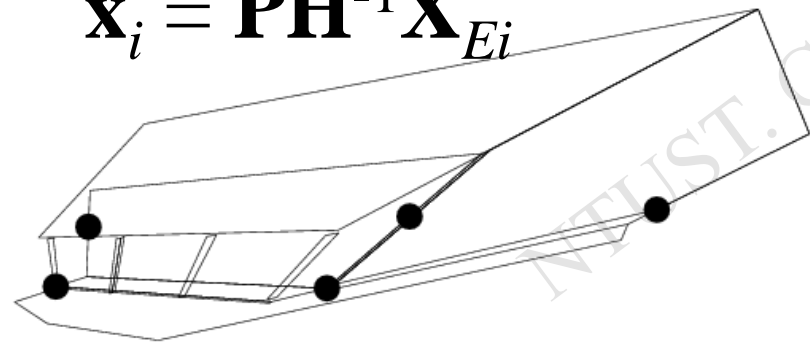
- use control points X_{Ei} with know coordinates, to go from projective to metric



$$\mathbf{X}_{Ei} = \mathbf{H}\mathbf{x}_i$$

\mathbf{X}_{Ei} → Ground truth (measurement in real environment)
 \mathbf{H} → Homography (4x4)
 \mathbf{x}_i → 3D points for an estimated projective reconstruction.

$$\mathbf{x}_i = \mathbf{P}\mathbf{H}^{-1}\mathbf{X}_{Ei}$$

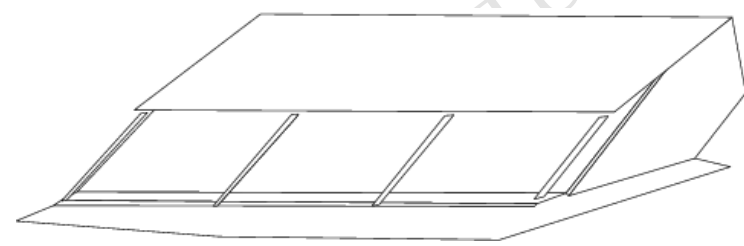
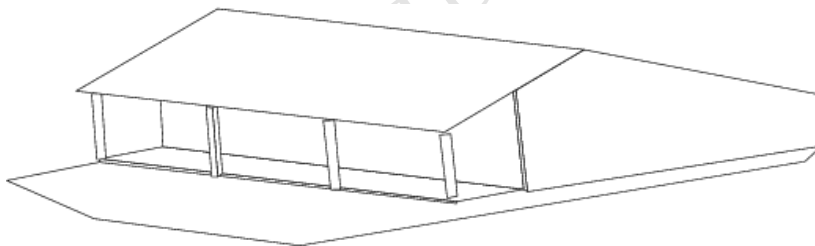


Note! do NOT select degenerated cases,
Ex. 4 points on a plane, 3 points on a line.



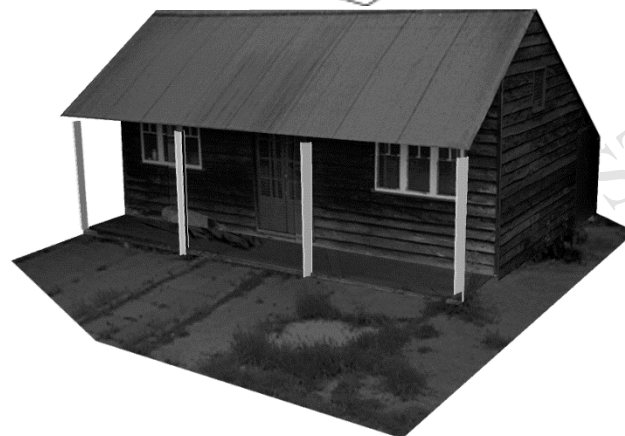
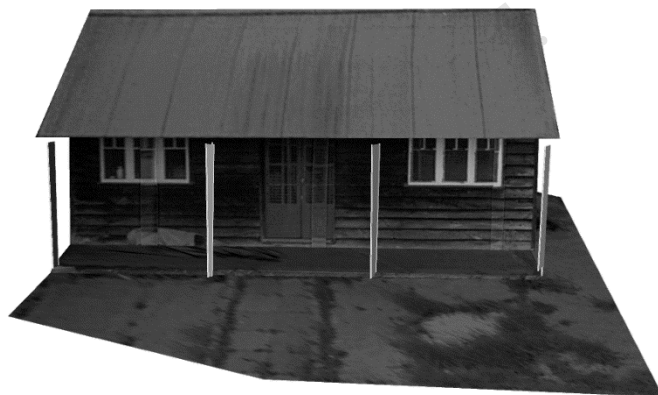
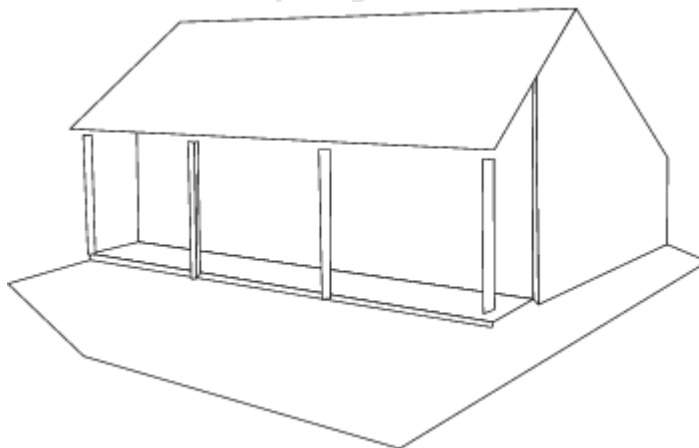
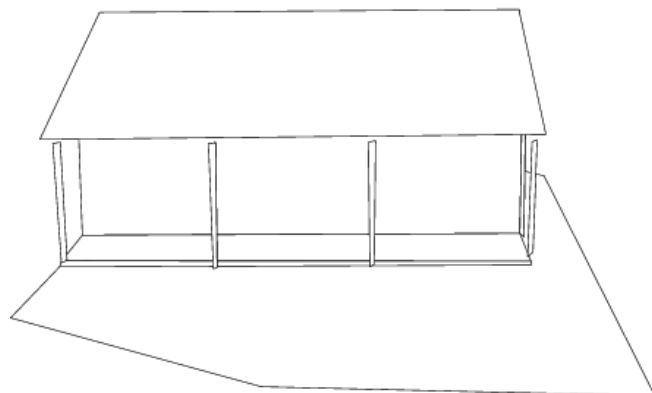
Affine reconstruction

- There are 3 sets of parallel lines in the scene, each set with a different direction. These 3 sets enable the position of the plane at infinity to be computed in the projective reconstruction. Note that parallel scene lines are parallel in reconstruction, but lines are NOT perpendicular in the reconstruction.



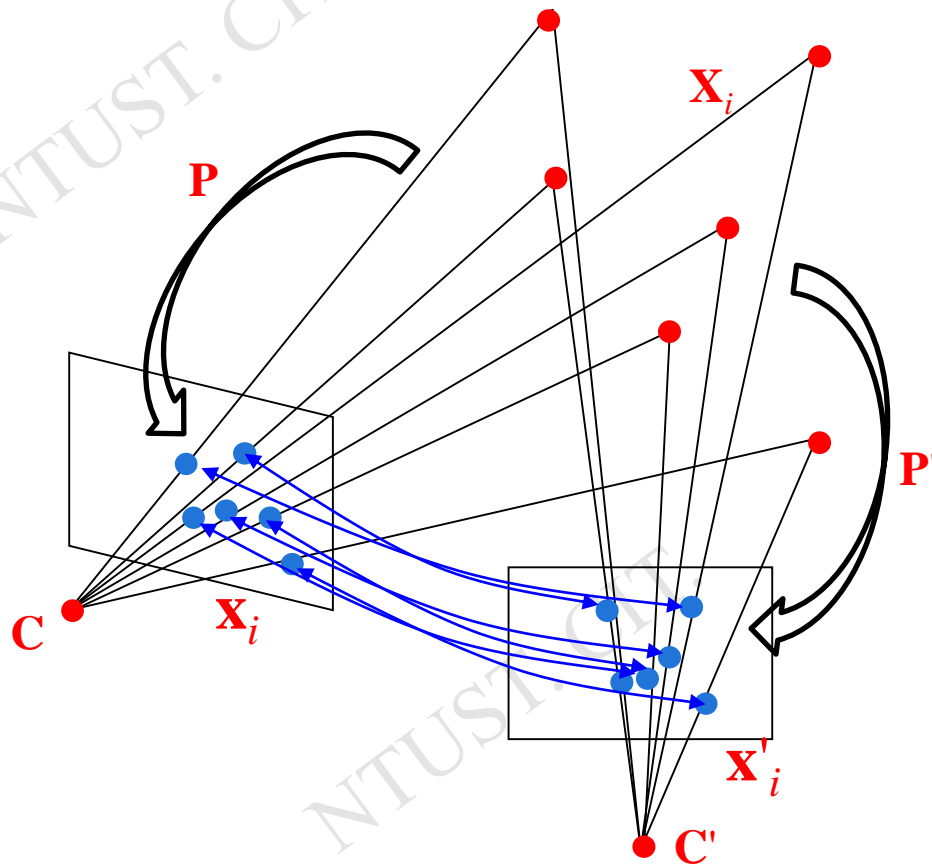


Metric reconstruction





3D estimation Direct Triangulation method



• Condition-1

$\mathbf{P}, \mathbf{P}', \mathbf{x}_i, \mathbf{x}'_i$: known

\mathbf{X} : unknown (to be solved)

• Condition-2

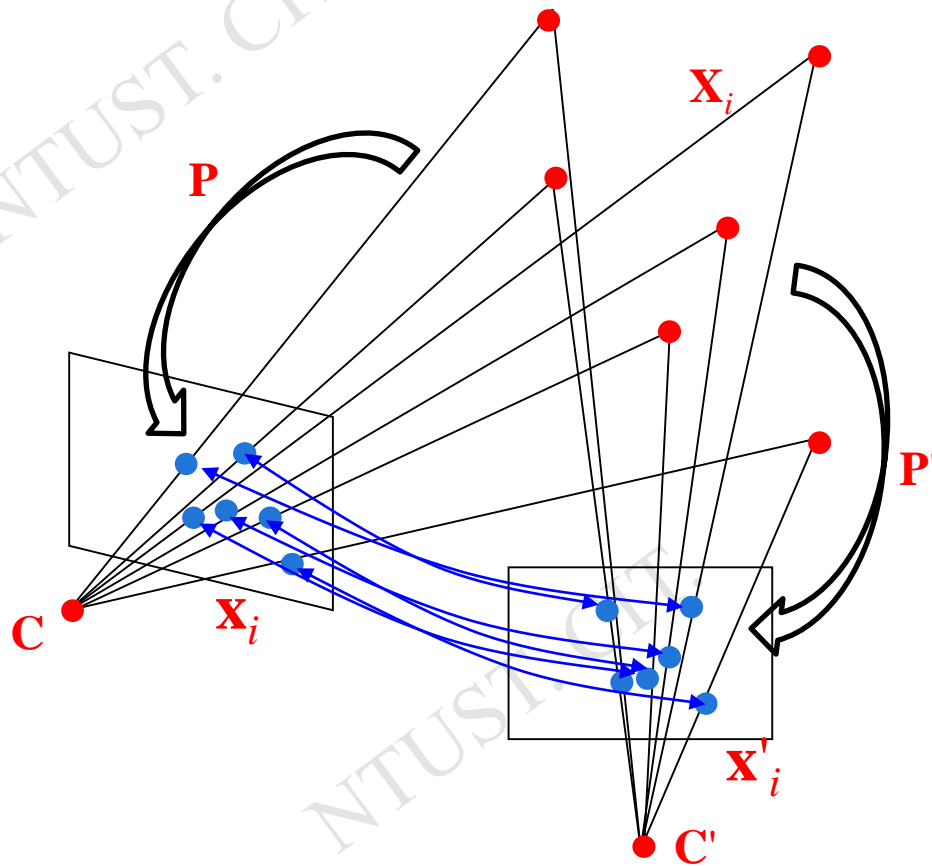
$\mathbf{x}_i, \mathbf{x}'_i, \mathbf{X}$: known

\mathbf{P}, \mathbf{P}' : unknown (to be solved)

→ become a calibration problem



3D estimation Direct Triangulation method



- **Condition-3(direct reconstruction)**

$\mathbf{x}_i, \mathbf{x}'_i$: known

5 points \mathbf{X} : known

\mathbf{P}, \mathbf{P}' : unknown (to be solved)

- **Condition-4(affine reconstruction)**

$\mathbf{x}_i, \mathbf{x}'_i$, other cues: known

\mathbf{P}, \mathbf{P}' : unknown (to be solved)

- **Condition-5(metric reconstruction)**

$\mathbf{x}_i, \mathbf{x}'_i, \mathbf{K}$: known

$\mathbf{X}, \mathbf{P}, \mathbf{P}'$: unknown (to be solved)



Objective

Given two uncalibrated images compute $(\mathbf{P}_M, \mathbf{P}'_M, \{\mathbf{X}_{Mi}\})$
(i.e. within similarity of original scene and cameras)

Algorithm [Textbook: Hartley04, Algorithm 10.1]

(i) Compute projective reconstruction $(\mathbf{P}, \mathbf{P}', \{\mathbf{X}_i\})$

- (a) Compute \mathbf{F} from $\mathbf{x}_i \leftrightarrow \mathbf{x}'_i$
- (b) Compute \mathbf{P}, \mathbf{P}' from \mathbf{F}
- (c) Triangulate \mathbf{X}_i from $\mathbf{x}_i \leftrightarrow \mathbf{x}'_i$

(ii) Rectify reconstruction from projective to metric

(a) Direct method: compute \mathbf{H} from control points \rightarrow (condition-3 in the previous slide)

$$\mathbf{P}_M = \mathbf{P}\mathbf{H}^{-1}, \quad \mathbf{P}'_M = \mathbf{P}'\mathbf{H}^{-1}, \quad \mathbf{X}_{Mi} = \mathbf{H}\mathbf{X}_i$$

Stratified method:

(a) Affine reconstruction: \rightarrow (condition-4 in the previous slide)

$$\mathbf{H} = \begin{bmatrix} \mathbf{I} & 0 \\ \boldsymbol{\pi}_\infty^T & 1 \end{bmatrix}$$

(b) Metric reconstruction: compute IAC $\boldsymbol{\omega} \rightarrow$ (condition-5 in the previous slide)

$$\mathbf{H} = \begin{bmatrix} \mathbf{A}^{-1} & 0 \\ 0 & 1 \end{bmatrix}$$

$$\mathbf{A}\mathbf{A}^T = (\mathbf{M}^T \boldsymbol{\omega} \mathbf{M})^{-1}$$



3D reconstruction

Image information provided	View relations and projective objects	3-space objects	reconstruction ambiguity
point correspondences	\mathbf{F}		projective
point correspondences including vanishing points	$\mathbf{F}, \mathbf{H}_{\infty}$	\mathbf{p}_{∞}	affine
Points correspondences and internal camera calibration	$\mathbf{F}, \mathbf{H}_{\infty}$ ω, ω'	\mathbf{p}_{∞} Ω_{∞}	metric

The two-view relations, image entities, and their 3-space counterpart for various classes of reconstruction ambiguity.



3D estimation Direct Triangulation method

- If the projected point \mathbf{PX} is very close to \mathbf{x} , then

$$\mathbf{x} \times \mathbf{x} = \mathbf{0} \rightarrow \text{zero vector}$$

So, one constraint for \mathbf{X} is

$$\mathbf{x} \times (\mathbf{PX}) = \mathbf{0}$$

For convenience, rewrite \mathbf{P} as

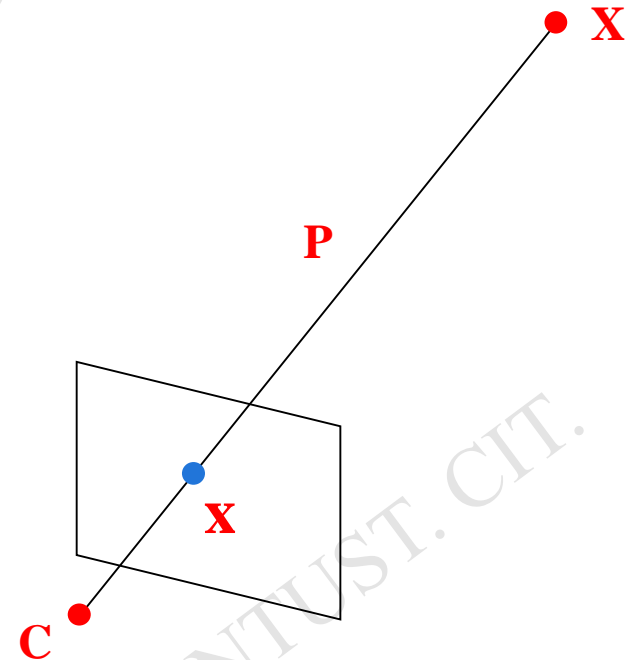
$$\mathbf{P} = \begin{bmatrix} p_{11} & p_{12} & p_{13} & p_{14} \\ p_{21} & p_{22} & p_{23} & p_{24} \\ p_{31} & p_{31} & p_{32} & p_{34} \end{bmatrix} = \begin{bmatrix} \mathbf{p}_1^T \\ \mathbf{p}_2^T \\ \mathbf{p}_3^T \end{bmatrix}_{3 \times 4}$$

\swarrow 1x4 vector
 \swarrow 1x4 vector
 \swarrow 1x4 vector

$$\mathbf{p}_1^T = [p_{11} \quad p_{12} \quad p_{13} \quad p_{14}]$$

$$\mathbf{p}_2^T = [p_{21} \quad p_{22} \quad p_{23} \quad p_{24}]$$

$$\mathbf{p}_3^T = [p_{31} \quad p_{32} \quad p_{33} \quad p_{34}]$$





3D estimation Direct Triangulation method-co

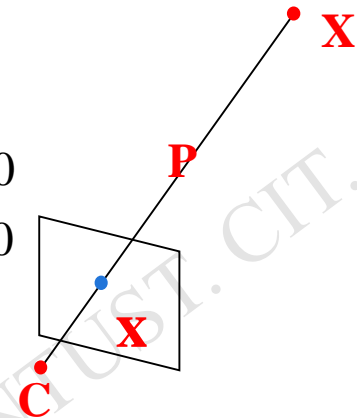
■ To solve the equation: $\mathbf{x} \times (\mathbf{P}\mathbf{X}) = \mathbf{0}$

$$\rightarrow \begin{bmatrix} x \\ y \\ w \end{bmatrix} \times \begin{bmatrix} \mathbf{p}_1^T \mathbf{X} \\ \mathbf{p}_2^T \mathbf{X} \\ \mathbf{p}_3^T \mathbf{X} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \rightarrow \begin{cases} y\mathbf{p}_3^T \mathbf{X} - w\mathbf{p}_2^T \mathbf{X} = 0 \\ -x\mathbf{p}_3^T \mathbf{X} + w\mathbf{p}_1^T \mathbf{X} = 0 \\ x\mathbf{p}_2^T \mathbf{X} - y\mathbf{p}_1^T \mathbf{X} = 0 \end{cases}$$

$$\rightarrow \begin{cases} -\mathbf{p}_2^T \mathbf{X} + \frac{y}{w} \mathbf{p}_3^T \mathbf{X} = 0 \\ -\mathbf{p}_1^T \mathbf{X} + \frac{x}{w} \mathbf{p}_3^T \mathbf{X} = 0 \\ y\mathbf{p}_1^T \mathbf{X} - x\mathbf{p}_2^T \mathbf{X} = 0 \end{cases}$$

Two independent eqs. only. (neglect 3rd eq.)
Let $(u, v) = (x/w, y/w)$ as the image point

$$\begin{cases} (v\mathbf{p}_3^T - \mathbf{p}_2^T) \mathbf{X} = 0 \\ (u\mathbf{p}_3^T - \mathbf{p}_1^T) \mathbf{X} = 0 \end{cases}$$



The unknown \mathbf{X} must satisfy the above eqs. for this image (says \mathbf{C})



3D estimation Direct Triangulation method-co

- Of course, in the second image (says \mathbf{C}'), the point \mathbf{X} will have the same property as:

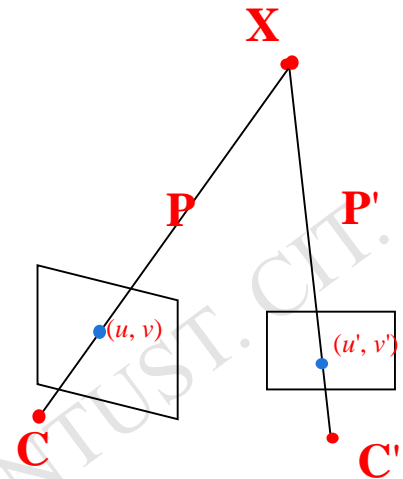
$$\begin{cases} (v'\mathbf{p}'_3 - \mathbf{p}'_2)^T \mathbf{X} = 0 \\ (u'\mathbf{p}'_3 - \mathbf{p}'_1)^T \mathbf{X} = 0 \end{cases}$$

- Finally, from one correspondence (u, v) and (u', v') , we will have

$$\begin{cases} (v\mathbf{p}_3 - \mathbf{p}_2)^T \mathbf{X} = 0 \\ (u\mathbf{p}_3 - \mathbf{p}_1)^T \mathbf{X} = 0 \\ (v'\mathbf{p}'_3 - \mathbf{p}'_2)^T \mathbf{X} = 0 \\ (u'\mathbf{p}'_3 - \mathbf{p}'_1)^T \mathbf{X} = 0 \end{cases} \Rightarrow \begin{bmatrix} u\mathbf{p}_3^T - \mathbf{p}_1^T \\ v\mathbf{p}_3^T - \mathbf{p}_2^T \\ u'\mathbf{p}'_3^T - \mathbf{p}'_1^T \\ v'\mathbf{p}'_3^T - \mathbf{p}'_2^T \end{bmatrix} \mathbf{X}_{4 \times 1} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

Solve it by SVD

remeber: $\mathbf{p}_1^T = [p_{11} \ p_{12} \ p_{13} \ p_{14}] \dots$



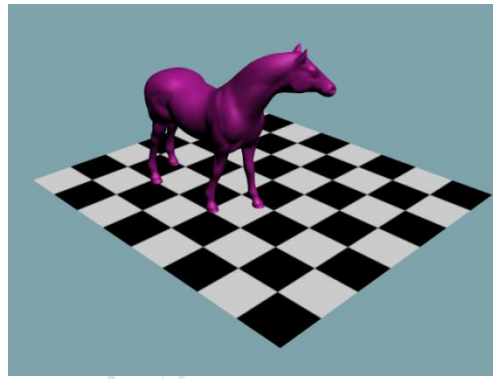
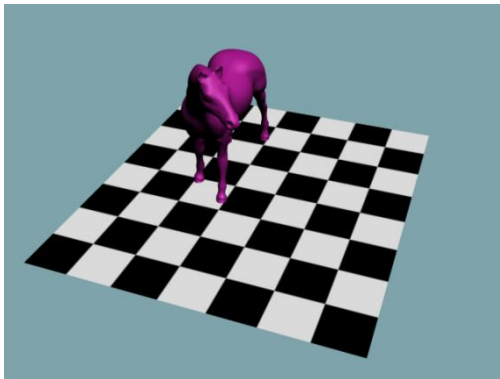


3D estimation Direct Triangulation method

■ Example1 for Condition-1

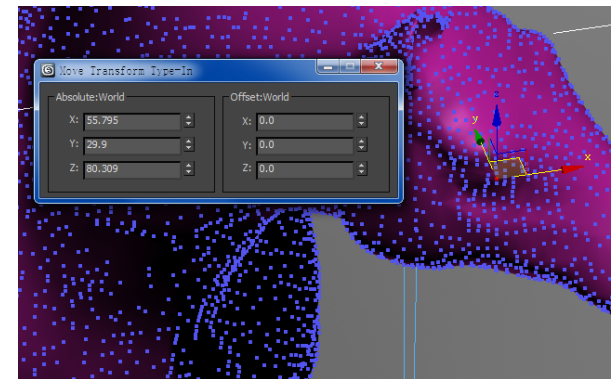
– $\mathbf{P}, \mathbf{P}', \mathbf{x}_i, \mathbf{x}'_i$: known

– \mathbf{X} : unknown (to be solved)



$\mathbf{P} =$

2.0179	1.5967	-0.5695	113.8802	$\mathbf{Pp} =$	2.8143	-1.3450	-0.5673	347.4957
0.2820	-0.7636	-2.4258	305.7125		-0.4439	-0.4444	-3.0134	371.1864
-0.0009	0.0023	-0.0018	1.0000		0.0023	0.0023	-0.0018	1.0000



Ground truth :
[55.795, 29.9, 80.309, 1]^T

To determine the 3D coordinate of the horse's right eye, we pick up this feature on these two images, then we have

$\mathbf{x} = [259, 120, 1]^T$

$\mathbf{x}' = [395, 89, 1]^T$

→ \mathbf{X} ?



3D estimation Direct Triangulation method

■ Example1 for Condition-1—cont.

$\mathbf{P}, \mathbf{P}', \mathbf{x}_i, \mathbf{x}'_i$: known

\mathbf{X} : unknown (to be solved)

$$\begin{bmatrix} u\mathbf{p}_3^T - \mathbf{p}_1^T \\ v\mathbf{p}_3^T - \mathbf{p}_2^T \\ u'\mathbf{p}'_3^T - \mathbf{p}'_1^T \\ v'\mathbf{p}'_3^T - \mathbf{p}'_2^T \end{bmatrix} \mathbf{X}_{4 \times 1} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

\mathbf{p}_1	\mathbf{p}_2	\mathbf{p}_3	$[u \ v]$
p1 =	p2 =	p3 =	u =
2.0179	0.2820	-0.0009	259
1.5967	-0.7636	0.0023	v =
-0.5695	-2.4258	-0.0018	120
113.8802	305.7125	1.0000	

In Matlab (note the notation ' in Matlab means transpose operation)

```
A=[u*p3'-p1';
v*p3'-p2';
up*pp3'-pp1';
vp*pp3'-pp2'];
[U,S,V]=svd(A)
```

\mathbf{p}'_1	\mathbf{p}'_2	\mathbf{p}'_3	$[u' \ v']$
pp1 =	pp2 =	pp3 =	up =
2.8143	-0.4439	0.0023	395
-1.3450	-0.4444	0.0023	vp =
-0.5673	-3.0134	-0.0018	89
347.4957	371.1864	1.0000	

```
...
V=
0.0038  0.7662 -0.3417 -0.5441
0.0030 -0.5582 -0.7735 -0.3002
0.0088 -0.3183  0.5337 -0.7834
-0.9999 -0.0015  0.0010 -0.0099
```

normalize

$\mathbf{X} =$
55.1137
30.4065
79.3529
1.0000

Ground truth :

$[55.795, 29.9, 80.309, 1]^T$

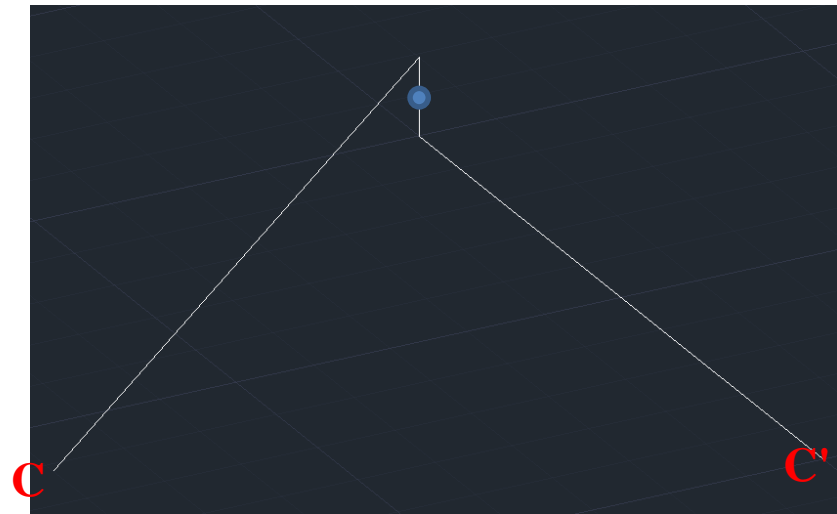
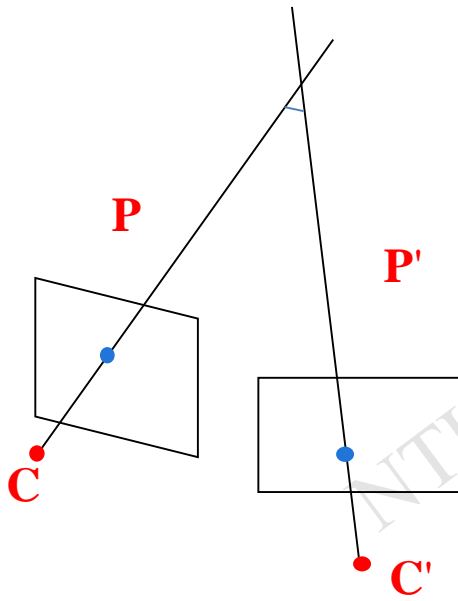


3D estimation Direct Triangulation method

■ Example1 for Condition-1—cont.

Note!

These two spatial lines are not necessary to intersect. The solution is the point whose residual error is smallest.





3D estimation Direct Triangulation method

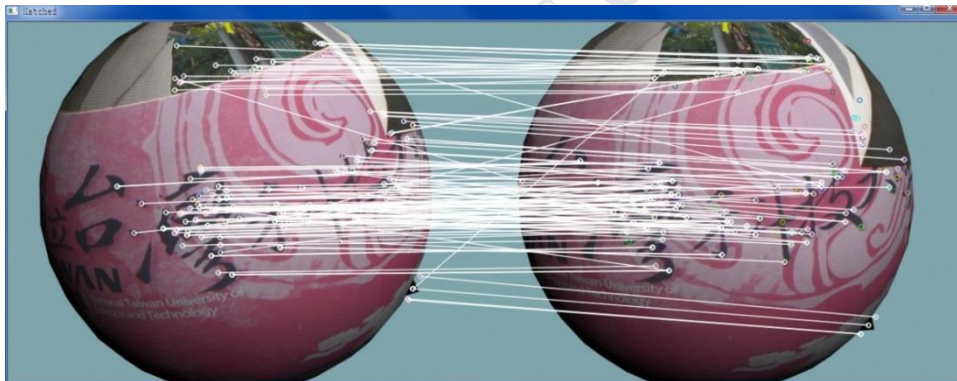
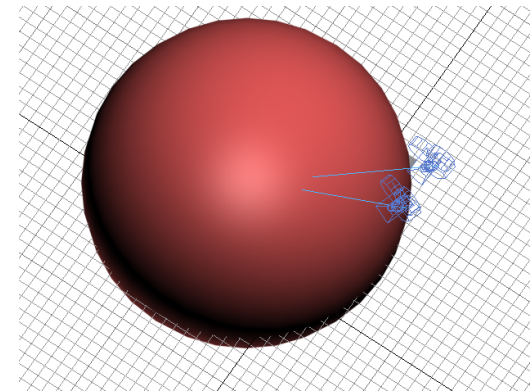
■ Example2 for Condition-1



$P =$
 2.5873 -0.3302 -0.9174 246.8139
 -0.5844 -1.8389 -1.8654 401.7092
 0.0004 0.0014 -0.0029 1.0000



$P_p =$
 2.5737 0.2628 -0.9172 207.1313
 -0.1437 -1.9363 -1.8245 382.3570
 0.0001 0.0014 -0.0029 1.0000



The question is:

P, P' : known

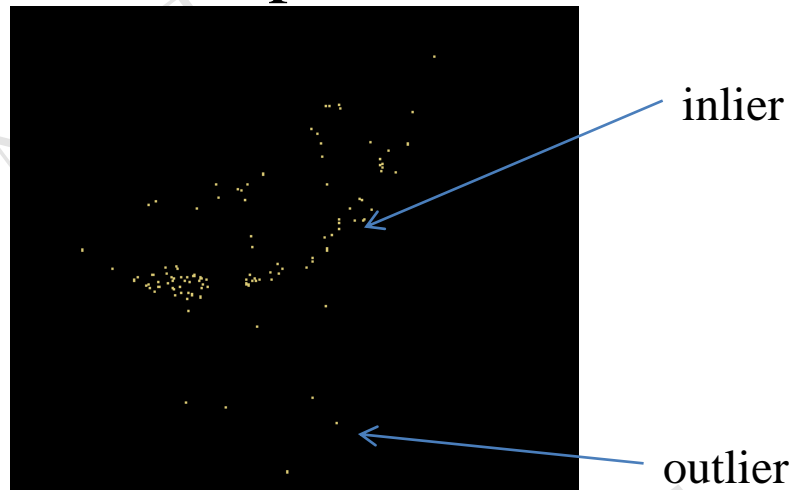
$\mathbf{x}_i, \mathbf{x}'_i$: Generated by Feature
Matching algorithm (including outlier)

$\mathbf{x}?$

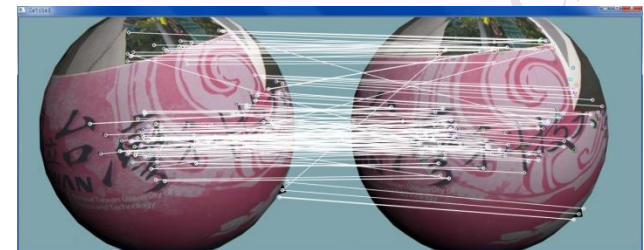
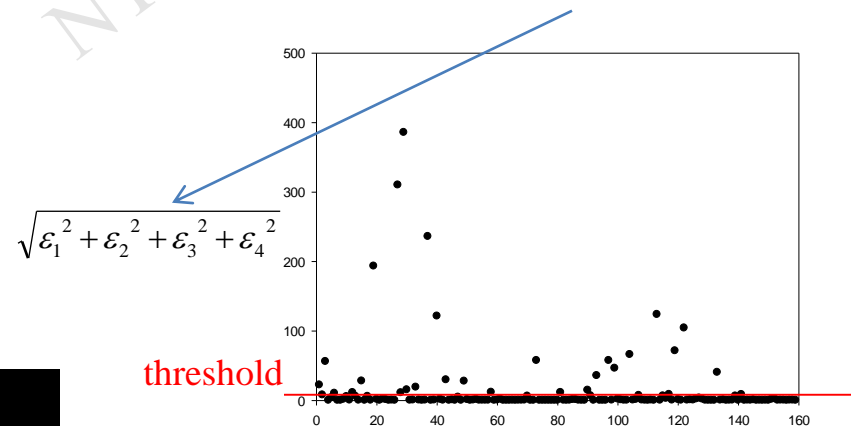


3D estimation Direct Triangulation method

■ Example2 for Condition-1—cont.



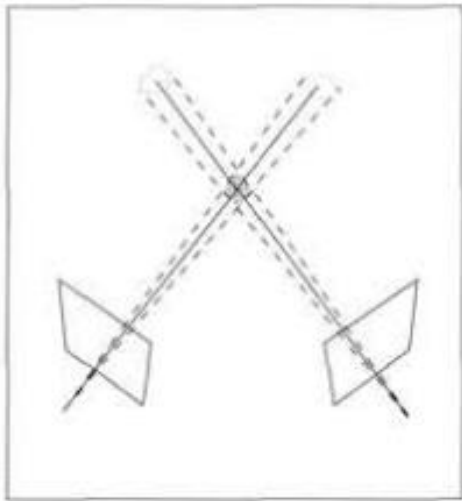
$$\begin{bmatrix} u\mathbf{p}_3^T - \mathbf{p}_1^T \\ v\mathbf{p}_3^T - \mathbf{p}_2^T \\ u'\mathbf{p}_3'^T - \mathbf{p}_1^T \\ v'\mathbf{p}_3'^T - \mathbf{p}_2^T \end{bmatrix} \mathbf{X}_{4 \times 1} = \begin{bmatrix} \varepsilon_1 \\ \varepsilon_2 \\ \varepsilon_3 \\ \varepsilon_4 \end{bmatrix}$$



In practice, the outliers are usually removed by Fundamental Matrix constraints.

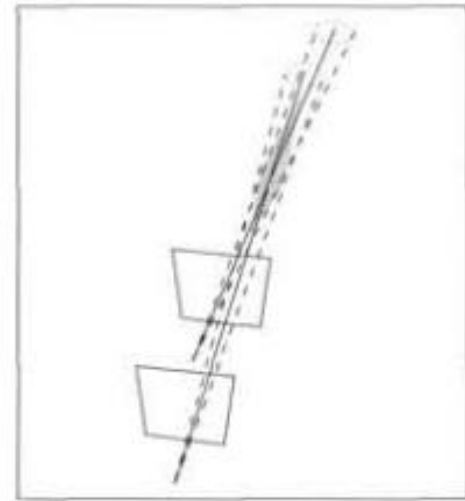
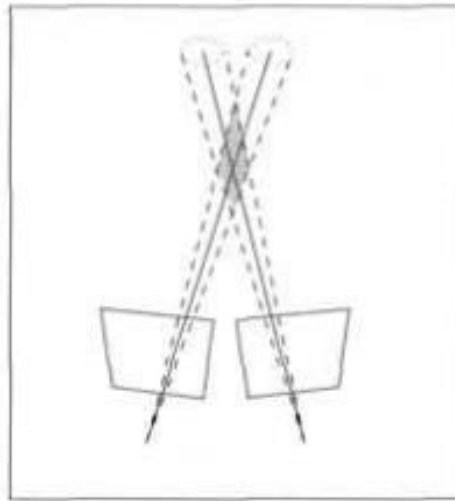


3D estimation Direct Triangulation method



More Difficult to find
correspondences
(feature matching)

Smaller uncertainty
(Better 3D reconstruction)



Easier to find correspondences
(feature matching)

Larger uncertainty
(Poor 3D reconstruction)

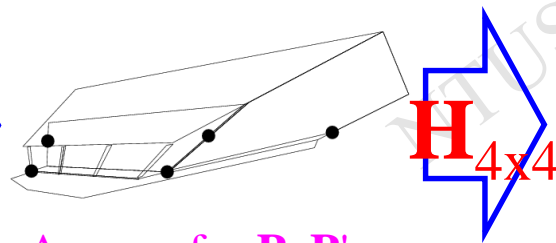
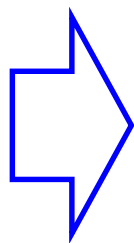


3D reconstruction for projective geometry

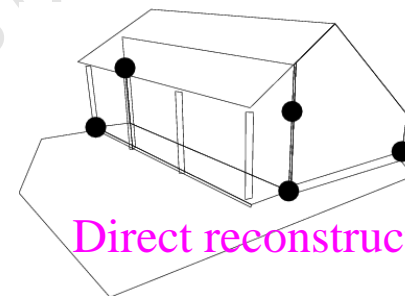
■ Direction reconstruction



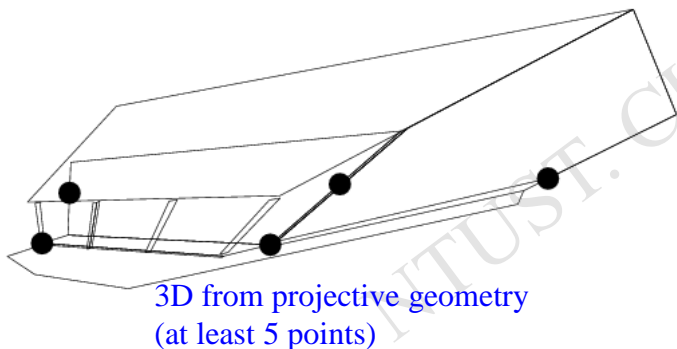
Two views geometry (F)



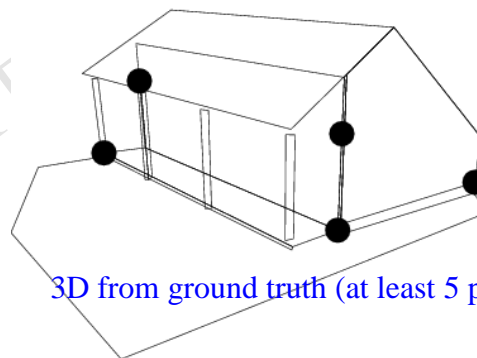
A guess for \mathbf{P}, \mathbf{P}'
[Projective reconstruction]



Direct reconstruction



3D from projective geometry
(at least 5 points)



3D from ground truth (at least 5 points)

Select at least 5 correspondences \rightarrow computer \mathbf{H}
(please review chapter "Projective 3D geometry" for detail)

$$\begin{bmatrix} X_1' \\ X_2' \\ X_3' \\ X_4' \end{bmatrix} = \begin{bmatrix} H_{11} & H_{12} & H_{13} & H_{14} \\ H_{21} & H_{22} & H_{23} & H_{24} \\ H_{31} & H_{32} & H_{33} & H_{34} \\ H_{41} & H_{42} & H_{43} & H_{44} \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \\ X_3 \\ X_4 \end{bmatrix}$$

$$\begin{aligned} \tilde{\mathbf{H}}_1^T &= [H_{11} \ H_{12} \ H_{13} \ H_{14}] \\ \tilde{\mathbf{H}}_2^T &= [H_{21} \ H_{22} \ H_{23} \ H_{24}] \\ \tilde{\mathbf{H}}_3^T &= [H_{31} \ H_{32} \ H_{33} \ H_{34}] \\ \tilde{\mathbf{H}}_4^T &= [H_{41} \ H_{42} \ H_{43} \ H_{44}] \\ \mathbf{X}^T &= [X \ Y \ Z \ 1] \end{aligned}$$

$$\begin{bmatrix} \mathbf{X}^T & 0^T & 0^T & -\mathbf{X}'\mathbf{X}^T \\ 0^T & \mathbf{X}^T & 0^T & -\mathbf{Y}'\mathbf{X}^T \\ 0^T & 0^T & \mathbf{X}^T & -\mathbf{Z}'\mathbf{X}^T \end{bmatrix}_{3 \times 16} \begin{bmatrix} \tilde{\mathbf{H}}_1 \\ \tilde{\mathbf{H}}_2 \\ \tilde{\mathbf{H}}_3 \\ \tilde{\mathbf{H}}_4 \end{bmatrix}_{16 \times 1} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}_{3 \times 1}$$

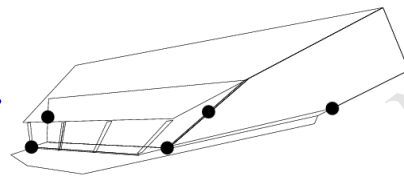
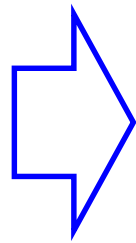


3D reconstruction for projective geometry

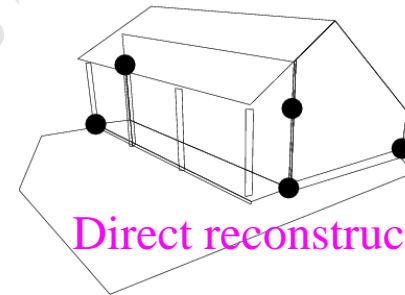
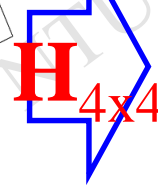
■ Direction reconstruction



Two views geometry (\mathbf{F})



A guess for \mathbf{P}, \mathbf{P}'
[Projective reconstruction]



Direct reconstruction

Once having \mathbf{H} :

$$\mathbf{P}_2 = \mathbf{P}_1 \mathbf{H}^{-1}$$

$$\mathbf{P}_2' = \mathbf{P}_1' \mathbf{H}^{-1}$$

$$\mathbf{X}_2 = \mathbf{H} \mathbf{X}_1$$

Either

Apply homography \mathbf{H} to all old
3D points (\mathbf{X}_{1i}), then get all new
3D points (\mathbf{X}_{2i})

Or

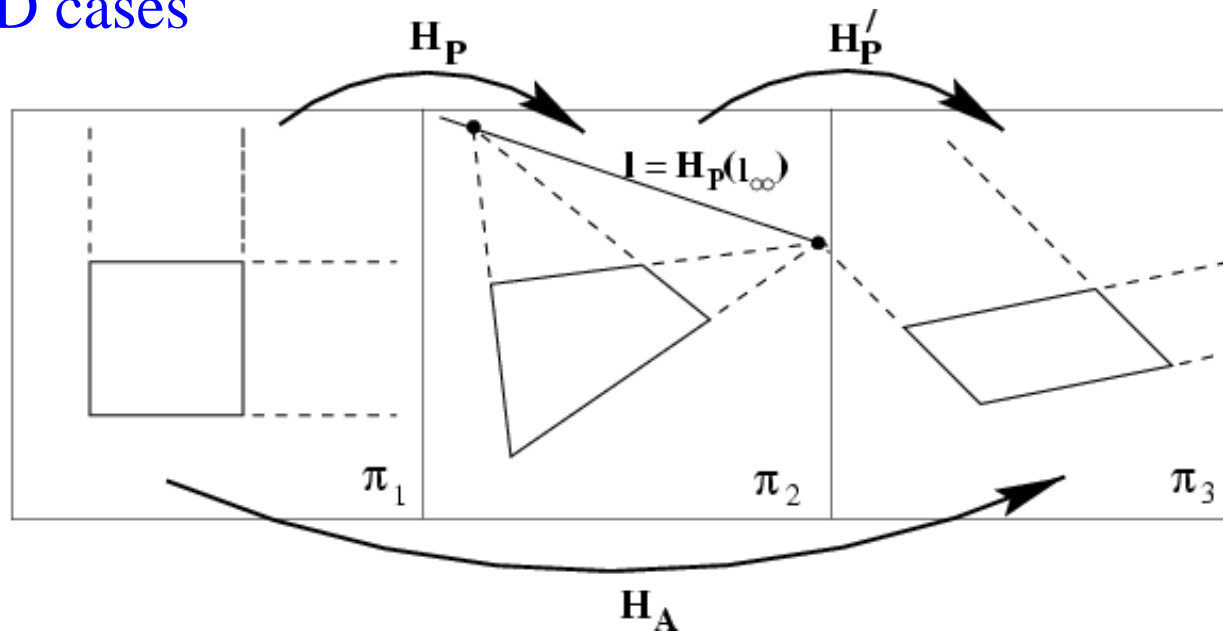
Compute new \mathbf{P} & \mathbf{P}' , and
use triangulation (back-
projection) from all 2D
correspondences, then
generate new 3D points
(\mathbf{X}_{2i})



3D reconstruction for projective geometry

- Affine reconstruction
 - Projective transformation, then affine trans..

2-D cases





3D reconstruction for projective geometry

■ Affine reconstruction

- Projective transformation, then affine trans..

$$(\mathbf{P}, \mathbf{P}', \{\mathbf{X}_i\})$$

$$\boldsymbol{\pi}_{\infty} = (A, B, C, D)^T \rightarrow (0, 0, 0, 1)^T$$

$$\mathbf{H}^{-T} \boldsymbol{\pi}_{\infty} = (0, 0, 0, 1)^T \rightarrow \text{to find a } \boldsymbol{\pi}_{\infty} \text{ under this constraint, but how?}$$

$$\mathbf{H} = \begin{bmatrix} \mathbf{I} & | & 0 \\ \boldsymbol{\pi}_{\infty}^T & & \end{bmatrix}$$

(if determinant $\neq 0$)

A desired solution if $\boldsymbol{\pi}_{\infty}$ can be determined, the mapped 3D points will be an affine-mapping

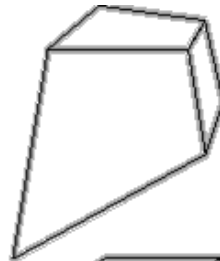


3D reconstruction for projective geometry

■ Hierarchy of transformations

Projective
15dof

$$\begin{bmatrix} A & t \\ v^T & v \end{bmatrix}$$



Intersection and tangency

Affine
12dof

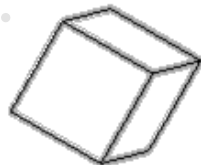
$$\begin{bmatrix} A & t \\ 0^T & 1 \end{bmatrix}$$



Parallellism of planes,
Volume ratios, centroids,
The plane at infinity π_∞

Similarity
7dof

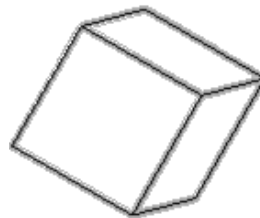
$$\begin{bmatrix} sR & t \\ 0^T & 1 \end{bmatrix}$$



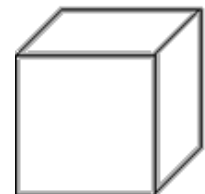
The absolute conic Ω_∞

Euclidean
6dof

$$\begin{bmatrix} R & t \\ 0^T & 1 \end{bmatrix}$$



Volume





3D reconstruction for projective geometry

■ Affine reconstruction

- Projective transformation, then affine trans..

How to determine π_{∞} ?

Here are various cases:

$$\mathbf{H}^{-T} \pi_{\infty} = (0,0,0,1)^T$$

3D plane at infinity

3D plane: the scene in the projective geometry
(need to be determined, ex. use 3 points to find out)

Translation motion

Parallel lines

Distance ratios on a lines (angle)

The infinite homography

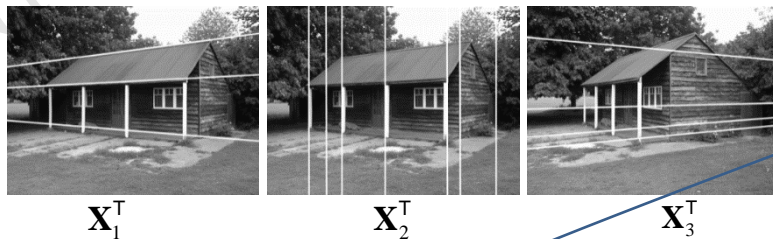
One of the cameras is affine



3D reconstruction for projective geometry

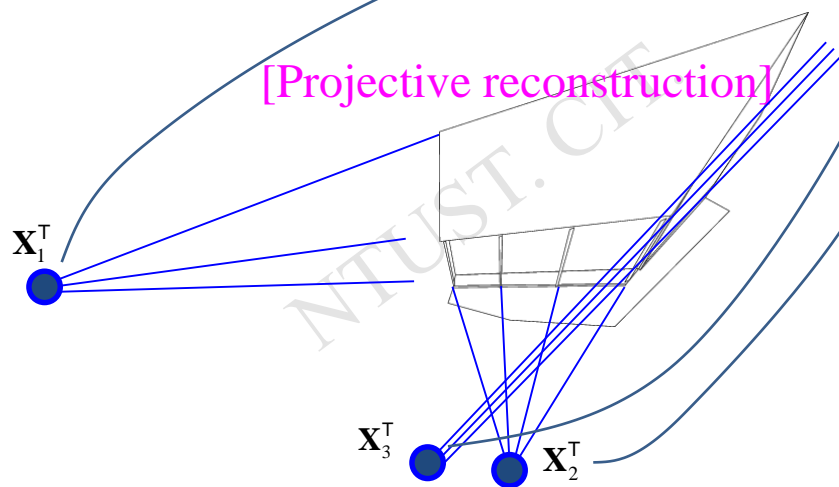
■ Affine reconstruction

■ Projective transformation, then affine trans..



Recall slide: "Projective 3D geometry"
Determine a 3D plane

$$\begin{bmatrix} \mathbf{X}_1^T \\ \mathbf{X}_2^T \\ \mathbf{X}_3^T \end{bmatrix} \pi = 0 \quad \begin{bmatrix} (X_1)_1 & (X_1)_2 & (X_1)_3 & (X_1)_4 \\ (X_2)_1 & (X_2)_2 & (X_2)_3 & (X_2)_4 \\ (X_3)_1 & (X_3)_2 & (X_3)_3 & (X_3)_4 \end{bmatrix} \begin{bmatrix} \pi_1 \\ \pi_2 \\ \pi_3 \\ \pi_4 \end{bmatrix} = 0$$



Find π_∞

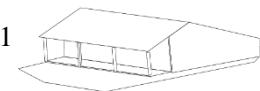
Then, $\mathbf{H} = \begin{bmatrix} \mathbf{I} & | & 0 \\ \hline & & \pi_\infty^T \end{bmatrix}$

Again, calculate final \mathbf{X}
(here is affine reconstruction)

$$\mathbf{P}_2 = \mathbf{P}_1 \mathbf{H}^{-1}$$

$$\mathbf{P}_2' = \mathbf{P}_1' \mathbf{H}^{-1}$$

$$\mathbf{X}_2 = \mathbf{H} \mathbf{X}_1$$





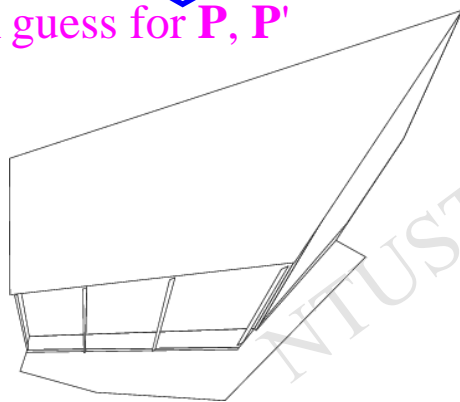
3D reconstruction for projective geometry

■ Metric reconstruction



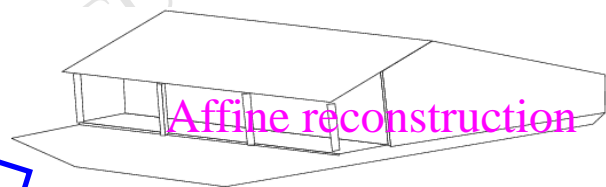
Two views geometry (\mathbf{F})

A guess for \mathbf{P}, \mathbf{P}'



[Projective reconstruction]

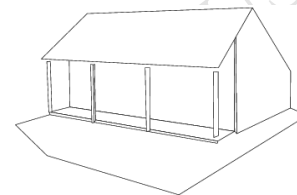
$$\mathbf{H} = \begin{bmatrix} \mathbf{I} & | & 0 \\ \boldsymbol{\pi}_{\infty}^T & & \end{bmatrix}$$



Affine reconstruction

$$\mathbf{H} = \begin{bmatrix} \mathbf{A}^{-1} & 0 \\ 0 & 1 \end{bmatrix}$$

$$\mathbf{A}\mathbf{A}^T = (\mathbf{M}^T \boldsymbol{\omega} \mathbf{M})^{-1}$$

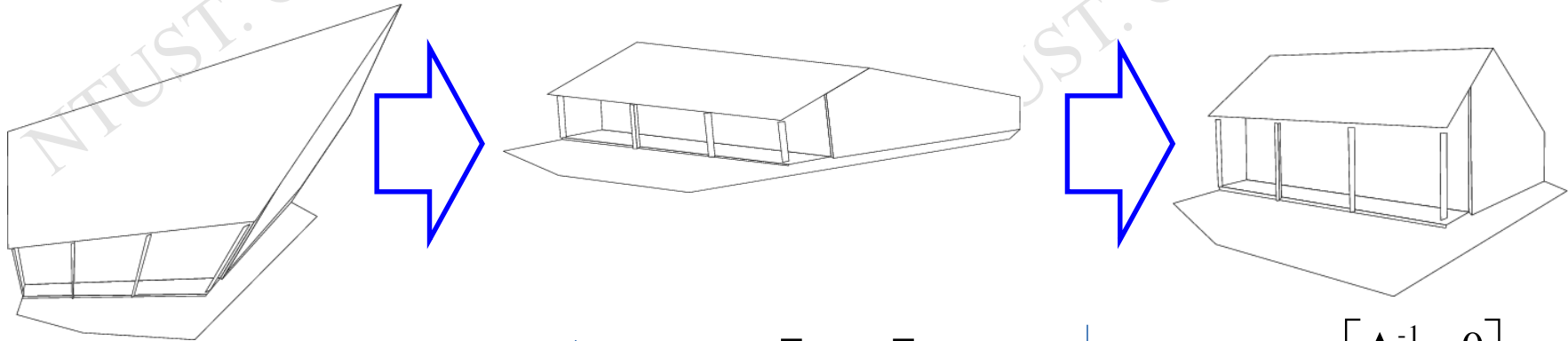


Metric reconstruction



3D reconstruction for projective geometry

■ Metric reconstruction—cont.



Either

$$\mathbf{P}_1 = [\mathbf{I} \mid 0] \quad \mathbf{P}_1' = [[\mathbf{e}']_{\times} \mathbf{F} + \mathbf{e}' \mathbf{v}^T \mid \lambda \mathbf{e}']$$

Or

$$\mathbf{P}_1 = [\mathbf{I} \mid 0] \quad \mathbf{P}_1' = [[\mathbf{e}']_{\times} \mathbf{F} \mid \mathbf{e}']$$

Triangulation for all 3D points (\mathbf{X}_{1i}),

$$\text{ex } \begin{bmatrix} u\mathbf{p}_3^T - \mathbf{p}_1^T \\ v\mathbf{p}_3^T - \mathbf{p}_2^T \\ u'\mathbf{p}_3^T - \mathbf{p}_1'^T \\ v'\mathbf{p}_3^T - \mathbf{p}_2'^T \end{bmatrix} \mathbf{X}_{4 \times 1} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\mathbf{H} = \begin{bmatrix} \mathbf{I} \mid 0 \\ \boldsymbol{\pi}_{\infty}^T \end{bmatrix}$$

$$\mathbf{P}_2 = \mathbf{P}_1 \mathbf{H}^{-1}$$

$$\mathbf{P}_2' = \mathbf{P}_1' \mathbf{H}^{-1}$$

$$\mathbf{X}_{2i} = \mathbf{H} \mathbf{X}_{1i}$$

$$\mathbf{H} = \begin{bmatrix} \mathbf{A}^{-1} & 0 \\ 0 & 1 \end{bmatrix}$$

$$\mathbf{A} \mathbf{A}^T = (\mathbf{M}^T \boldsymbol{\omega} \mathbf{M})^{-1}$$

$$\mathbf{P}_3 = \mathbf{P}_2 \mathbf{H}^{-1}$$

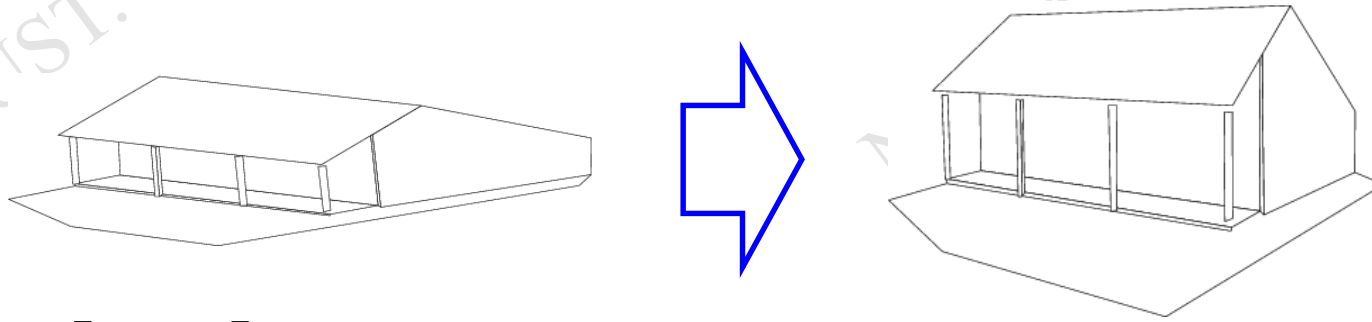
$$\mathbf{P}_3' = \mathbf{P}_2' \mathbf{H}^{-1}$$

$$\mathbf{X}_{3i} = \mathbf{H} \mathbf{X}_{2i}$$



3D reconstruction for projective geometry

■ Metric reconstruction—cont.



$$\mathbf{H} = \begin{bmatrix} \mathbf{A}^{-1} & 0 \\ 0 & 1 \end{bmatrix}$$

$$\mathbf{A}\mathbf{A}^T = (\mathbf{M}^T \boldsymbol{\omega} \mathbf{M})^{-1}$$

Cholesky factorization

Here

$$\boldsymbol{\omega} = (\mathbf{K}\mathbf{K}^T)^{-1}$$

→ you may assume intrinsic parameter \mathbf{K} is known, or under some constraints.

$$\mathbf{P}_3 = \mathbf{P}_2 \mathbf{H}^{-1}$$

$$\mathbf{P}_3' = \mathbf{P}_2' \mathbf{H}^{-1}$$

$$\mathbf{X}_{3i} = \mathbf{H} \mathbf{X}_{2i}$$

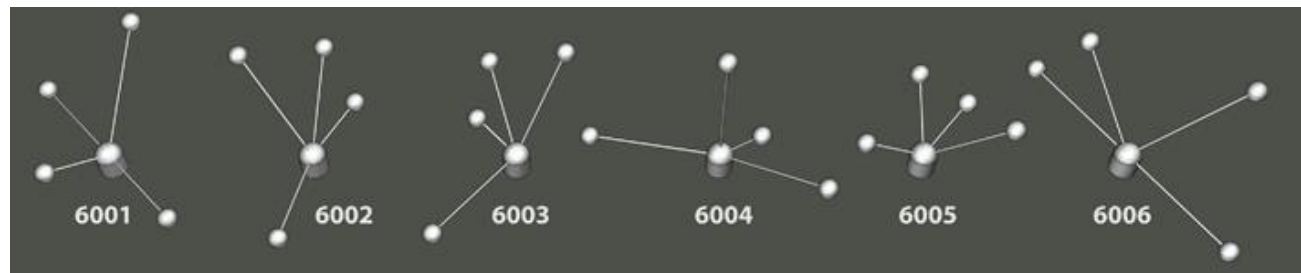
$$\mathbf{P}_2 = [\mathbf{M} | \mathbf{m}]$$

The projection matrix
of affine reconstruction



3D reconstruction–Applications

- Motion Tracking (other tracking issue)
- How to? 3D→2D (handling features)



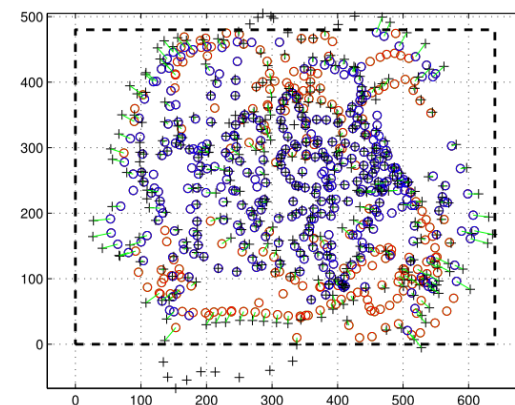
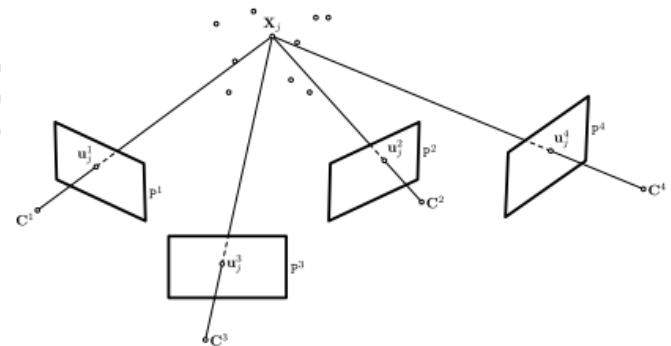
iotracker
advanced optical motion tracking





3D reconstruction–Applications

■ Multi-camera calibration





3D reconstruction–Applications

■ Other issues

- 1. Bundle adjustment?
- 2. Structure From motion?
- 3. Factorization?

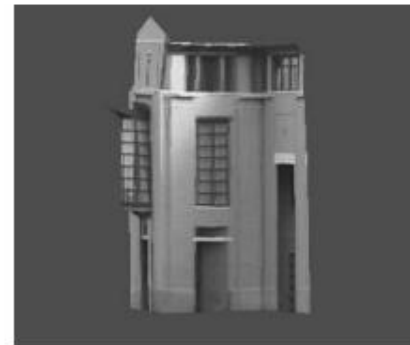
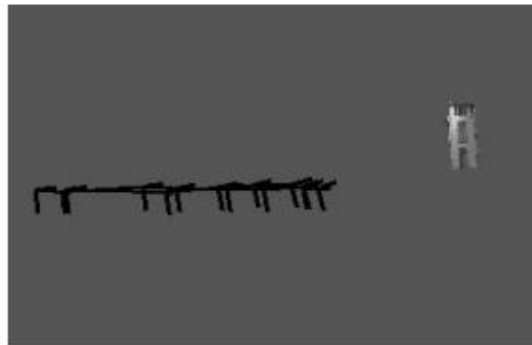
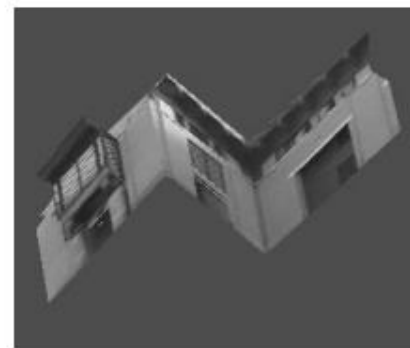
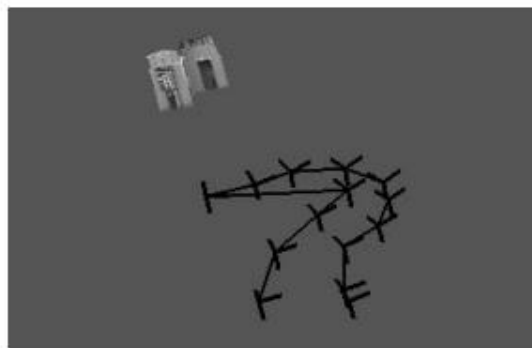
Select paper:

- M. Han and T. Kanade, “Creating 3D models with uncalibrated cameras,” in *IEEE Workshop on Applications of Computer Vision*, 2000, pp. 178-185.



3D reconstruction–Applications

■ Factorization



(a)

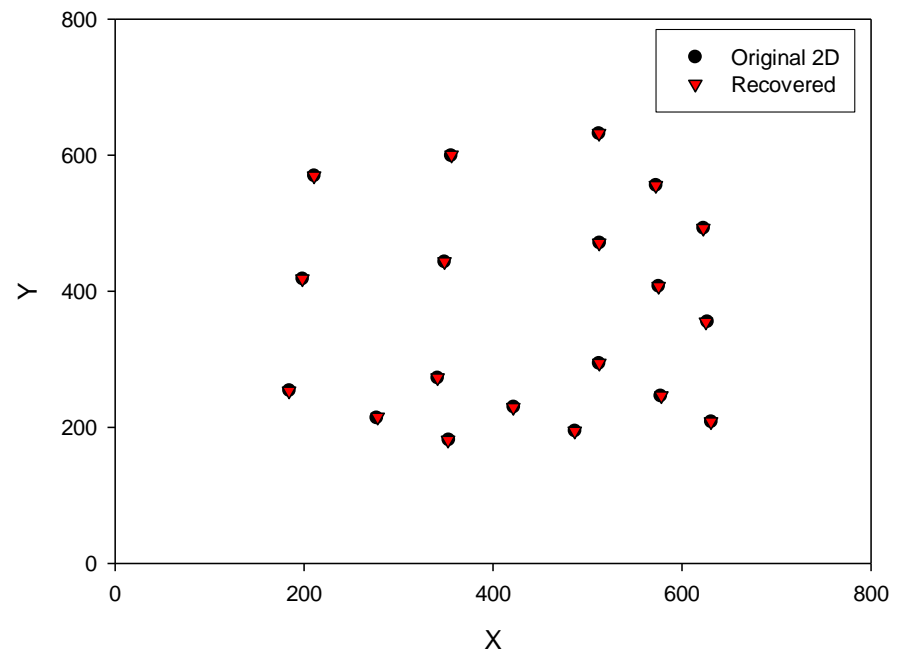
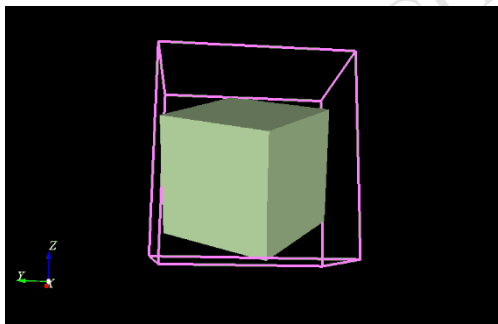
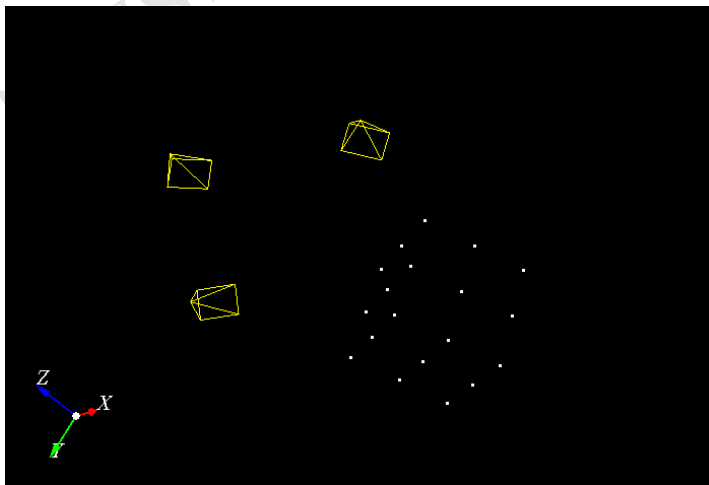
(b)

(c)



3D reconstruction—Applications

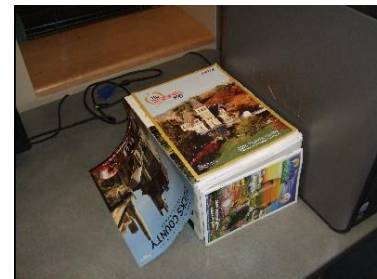
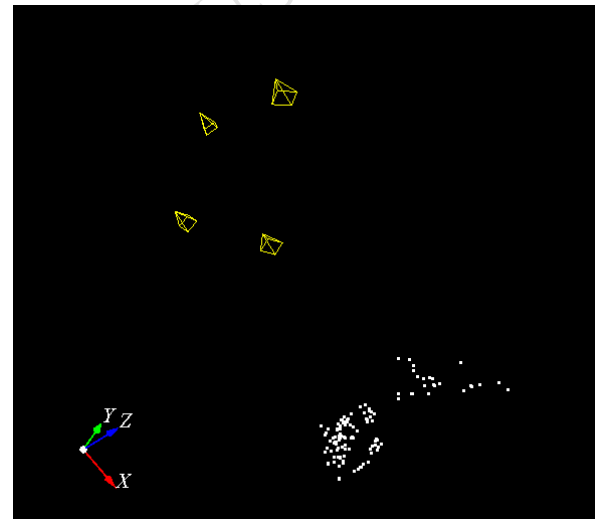
■ Factorization—Implementation result





3D reconstruction—Applications

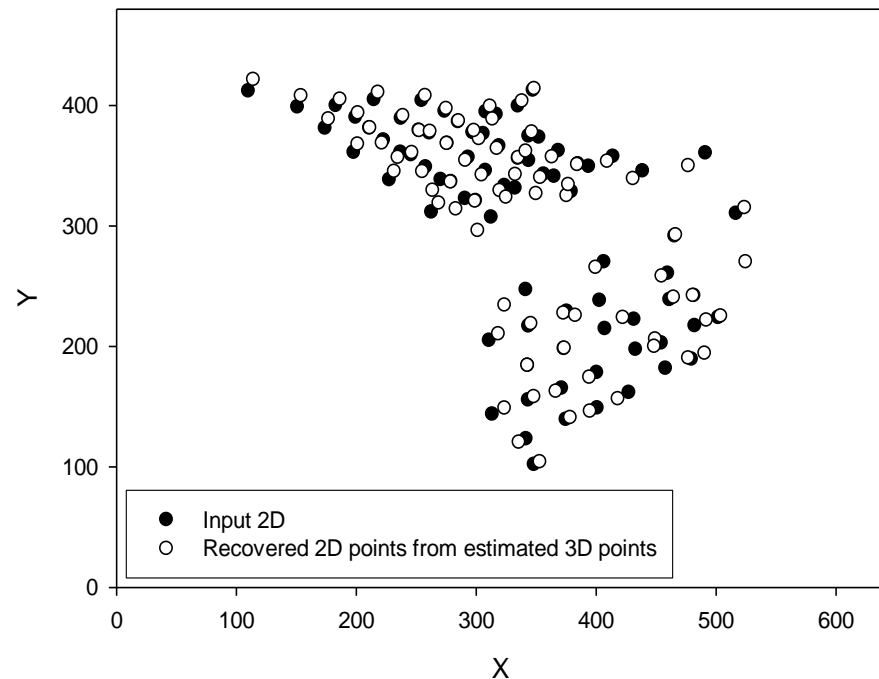
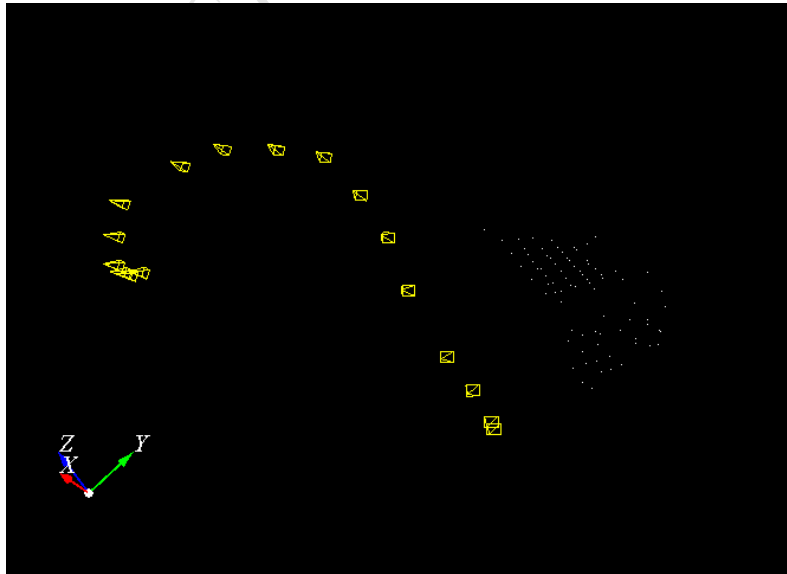
■ Factorization—Implementation result—cont.





3D reconstruction—Applications

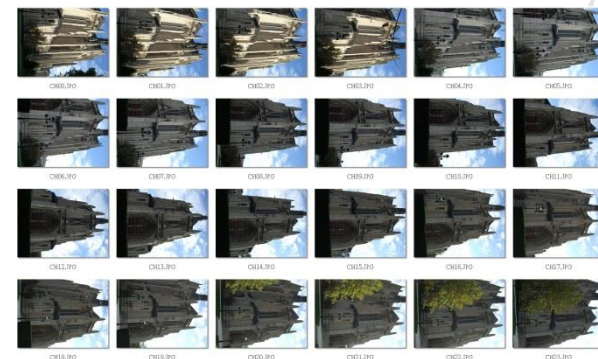
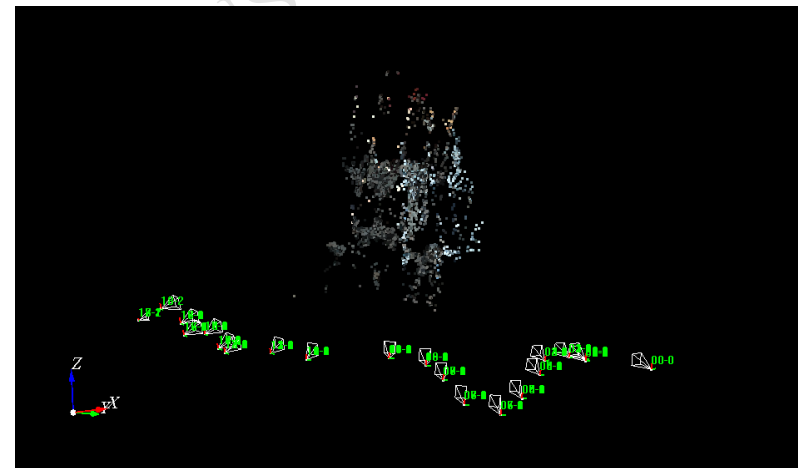
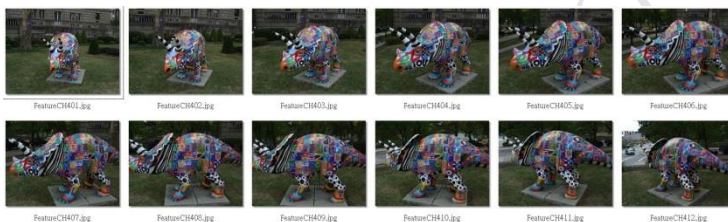
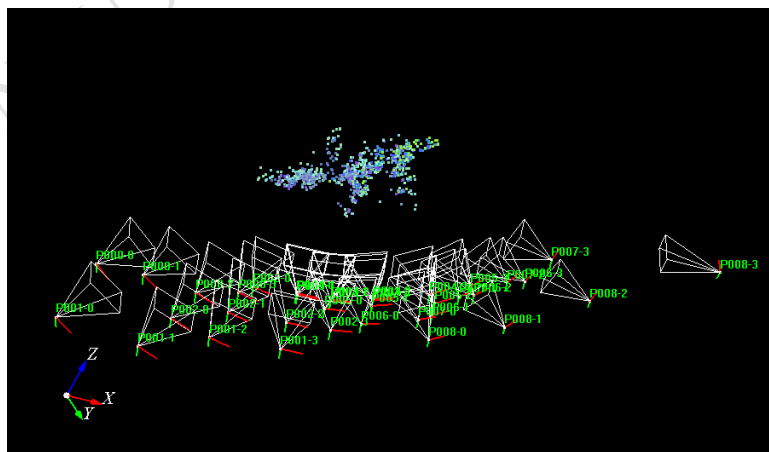
■ Factorization—Implementation result—cont.





3D reconstruction—Applications

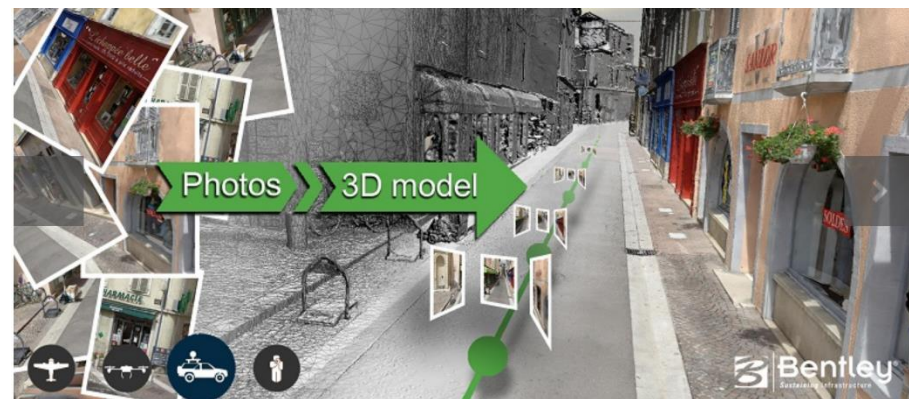
■ Factorization—Implementation result—cont.





Commercial / Noncommercial Tools

- Agisoft PhotoScan
- RealityCapture
- PhotoModeler
- Autodesk Remake
- Strata Foto 3D CX2



<http://www.agisoft.com/>

<https://www.capturingreality.com/>

<http://www.photomodeler.com/index.html>

<https://remake.autodesk.com/about>

<https://www.strata.com/foto-3d-cx-create-textured-3d-models-from-your-digital-camera/>



Commercial / Noncommercial Tools

- 3DF Zephyr Pro
- PIX4D
- DroneDeploy
- senseFly

<http://www.3dflow.net/3df-zephyr-pro-3d-models-from-photos/>

<https://pix4d.com/>

<https://www.dronedeploy.com/>

<https://www.sensefly.com/drones/ebee.html>



Structure from motion

CMPMVS

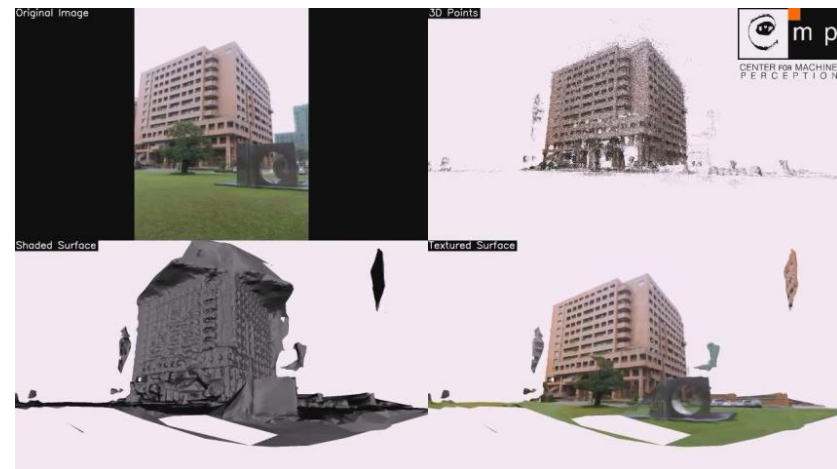
- Image-based 3D reconstruction (freeware)

CMPMVS - Multi-View Reconstruction Software



Authors: Michal Jancosek & Tomas Pajdla
Software written by: Michal Jancosek
Latest version: 0.6.0
Release date: September 28, 2012
Reference to cite:
 [1] M. Jancosek, T. Pajdla. *Multi-View Reconstruction Preserving Weakly-Supported Surfaces*, CVPR 2011 - IEEE Conference on Computer Vision and Pattern Recognition 2011 (pdf).

Introduction
 CMPMVS is a multi-view reconstruction software. The input to our software is a set of perspective images and camera parameters (internal and external camera calibrations). The output is a textured mesh of the rigid scene visible in the images. Non-rigid objects are implicitly ignored.
 For discussion on the software please visit our Google group at
<http://groups.google.com/group/cmpmvs>

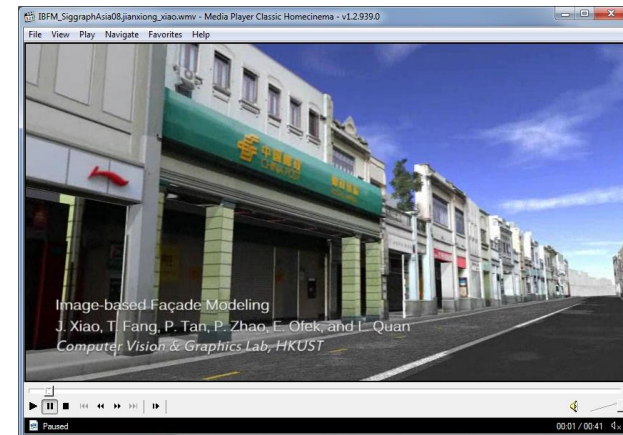
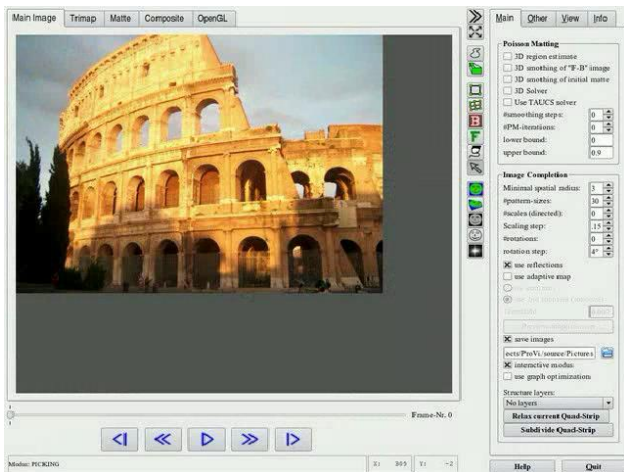


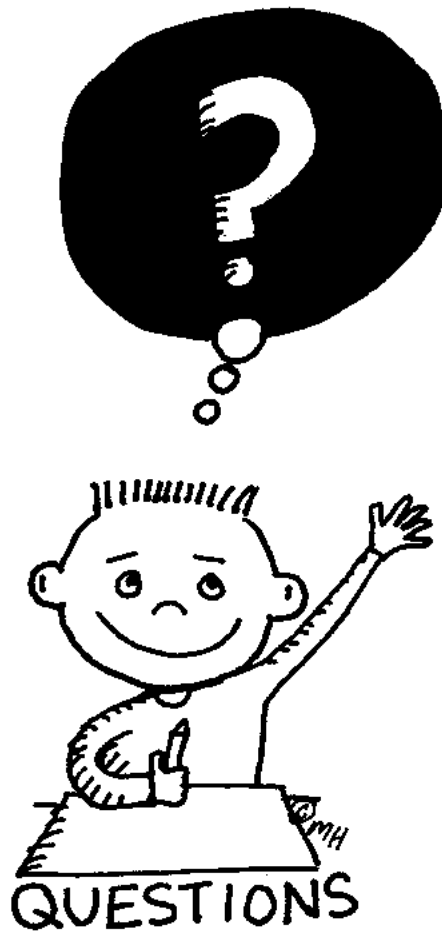


Structure from motion (software pre-product)

Applications

- Video trace: image synthesis
- Authoring tools to 3D (static object)
- Reconstruct urban 3D: Deal with frontal textures





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