

# 3D Vision

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# Computer Vision – 3D Vision

❑ What can you tell about the 3D object from their images?

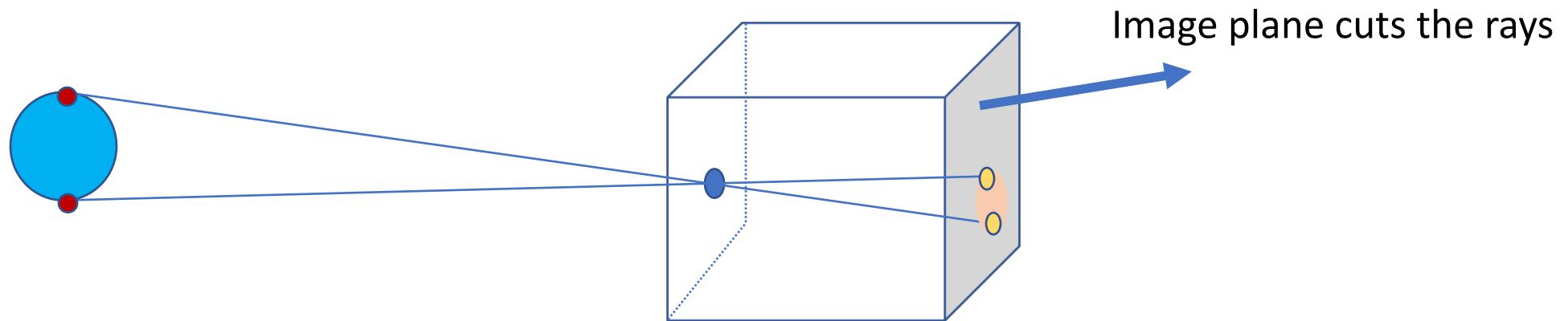
- Pattern Recognition – image pixel colors
- 3D vision – Pixel locations from projection

Main reference: Multiple View Geometry, [Hartley and Zisserman, 2004]

# Pinhole Camera model

## ❑ Camera Projection

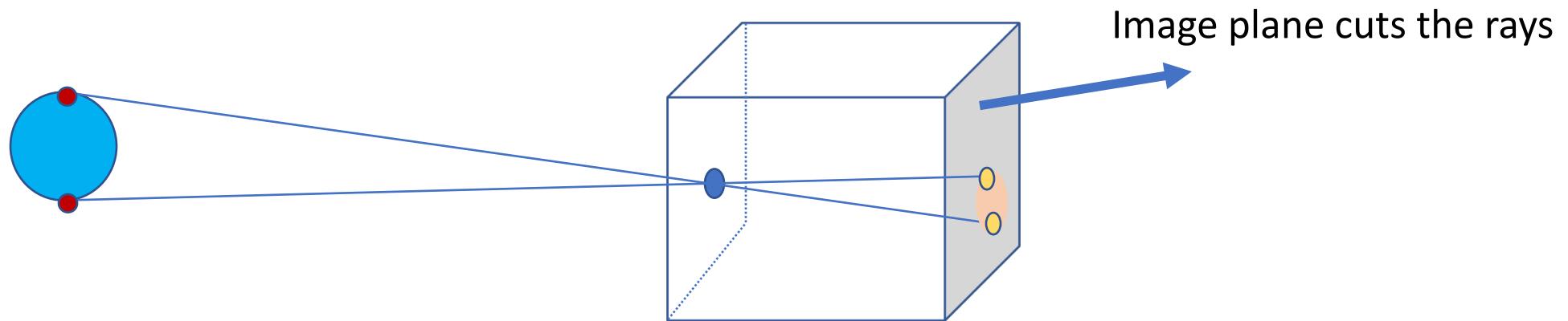
- The pin-hole camera model



# The Image plane

## ❑ Camera Projection

- You can put the image plane anywhere

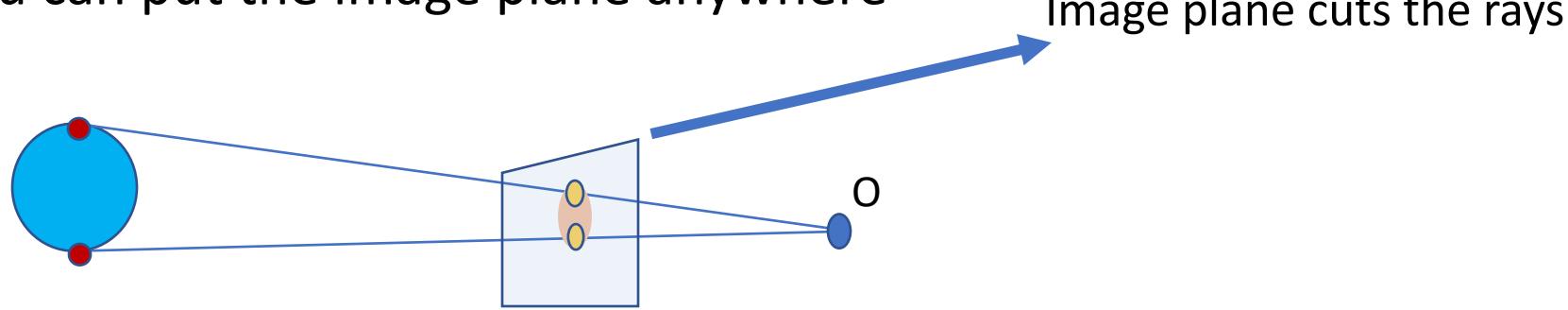


- Simpler to model if it assumed on the front at distance 1.

# The Image plane

## ❑ Camera Projection

- You can put the image plane anywhere



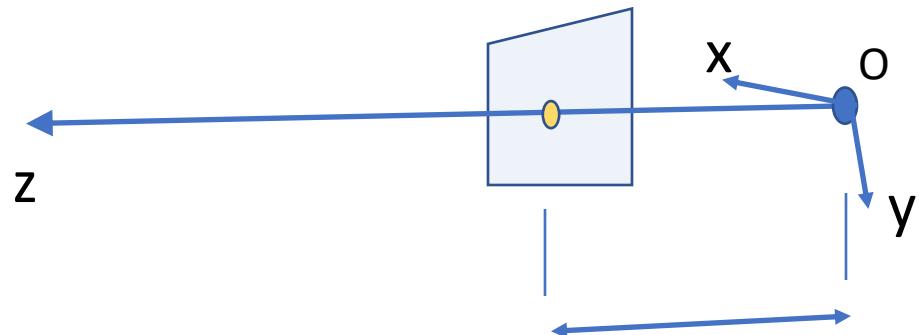
- Simpler to model if it is assumed on the front at distance 1.

Any difference?

# Camera axis

## ❑ Assume orientations

- Z – axis passes perpendicular to the image plane

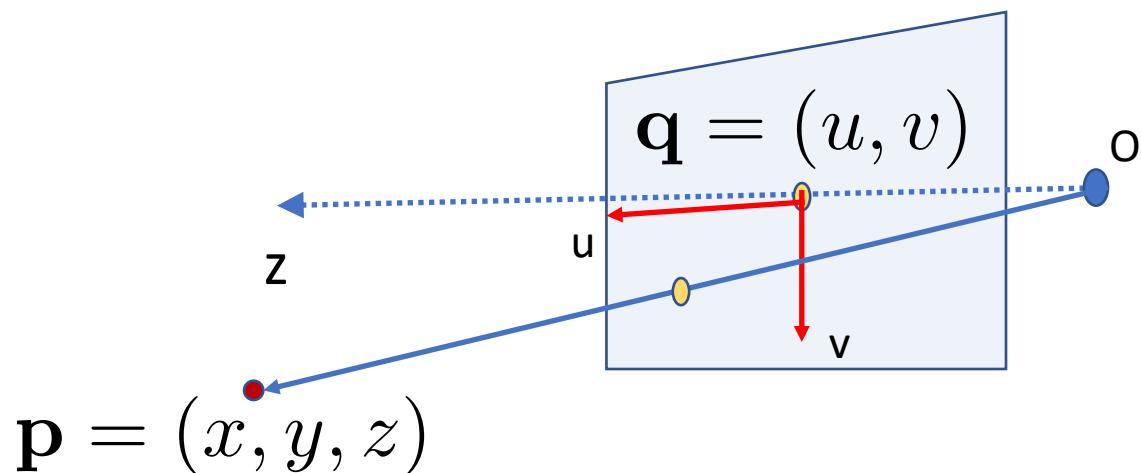


Focal length = 1

# Image coordinates

## ❑ Projection equations

- Image plane coordinate system



$$u = \frac{x}{z}$$

$$v = \frac{y}{z}$$

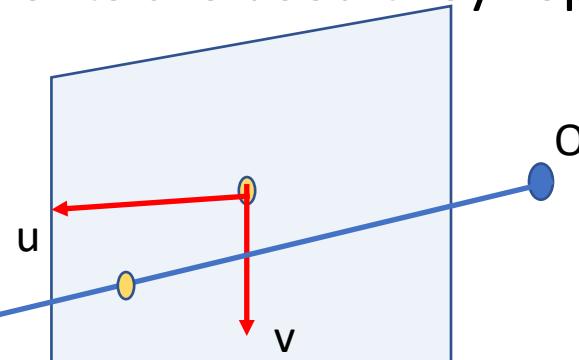
Can you prove them? – Exercise!

What kind of equations are they?

# Image coordinates

## □ Linear form

- Homogeneous coordinates
- Although 3 elements are used they represent projected points in 2D!



$$\mathbf{p} = (x, y, z)$$

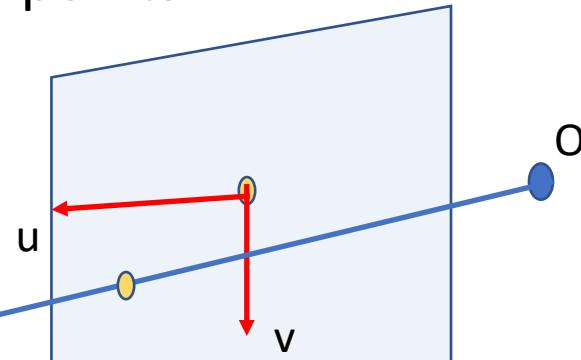
$$\mathbf{q} = (u, v, 1) \in \mathbb{P}^2$$

Homogeneous coordinates!

# Image coordinates

## □ Linear form

- Homogeneous coordinates
- Same for the 3D points!



$$\mathbf{p} = (x, y, z, 1)$$

$$\mathbf{q} = (u, v, 1) \in \mathbb{P}^2$$

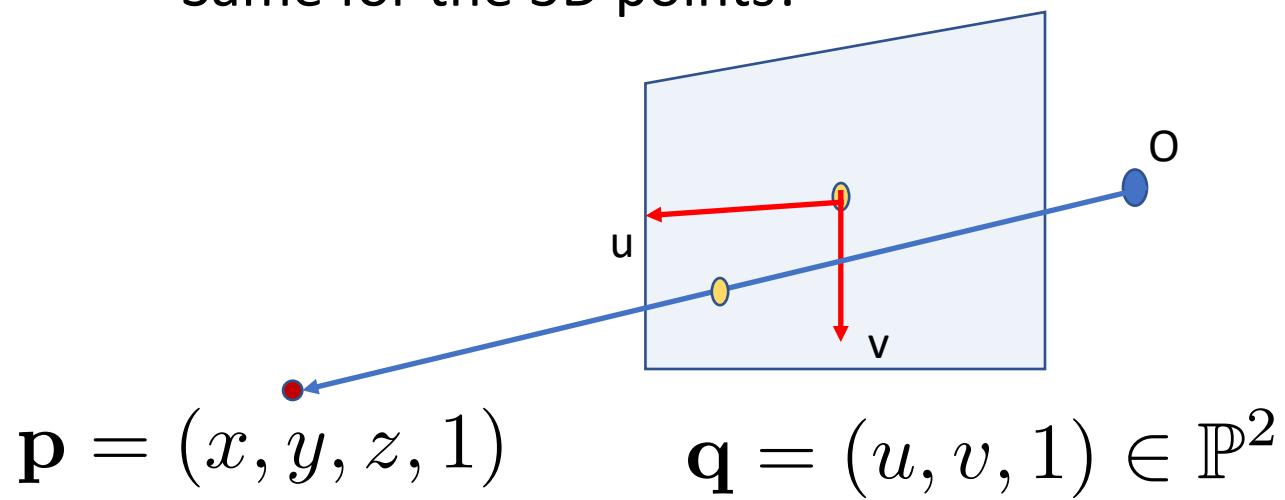
Homogeneous coordinates!

$$(u, v, 1) = (wu, wv, w)$$

# Projection matrix

## □ Linear form

- Homogeneous coordinates
- Same for the 3D points!



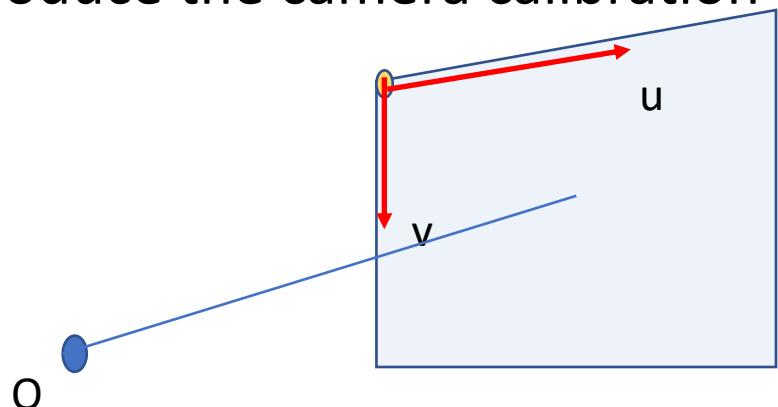
$$M = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}$$

Verify:

$$Mp = q$$

# Your cameras are different!

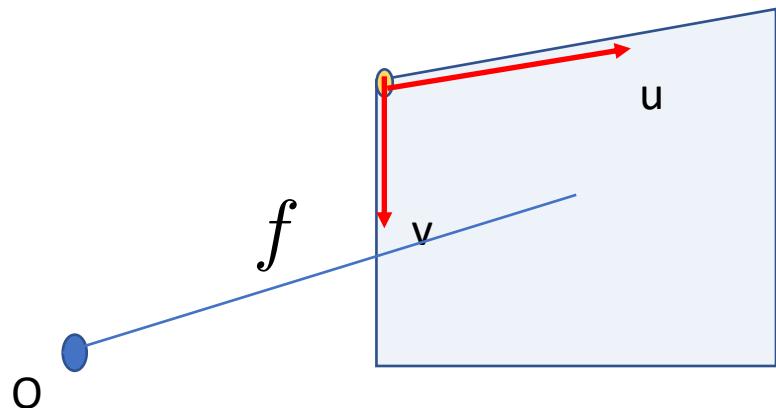
- ❑ Image origin is at the top left
  - This changes the projection equation!
  - Introduce the camera calibration matrix



$$KM\mathbf{p} = \mathbf{q}'$$
$$K = \begin{bmatrix} 1 & 0 & c_1 \\ 0 & 1 & c_2 \\ 0 & 0 & 1 \end{bmatrix}$$

# Your cameras are different!

- ❑ The focal length is much greater than 1 pixel.
  - Scale the image coordinates.



$$KM\mathbf{p} = \mathbf{q}'$$

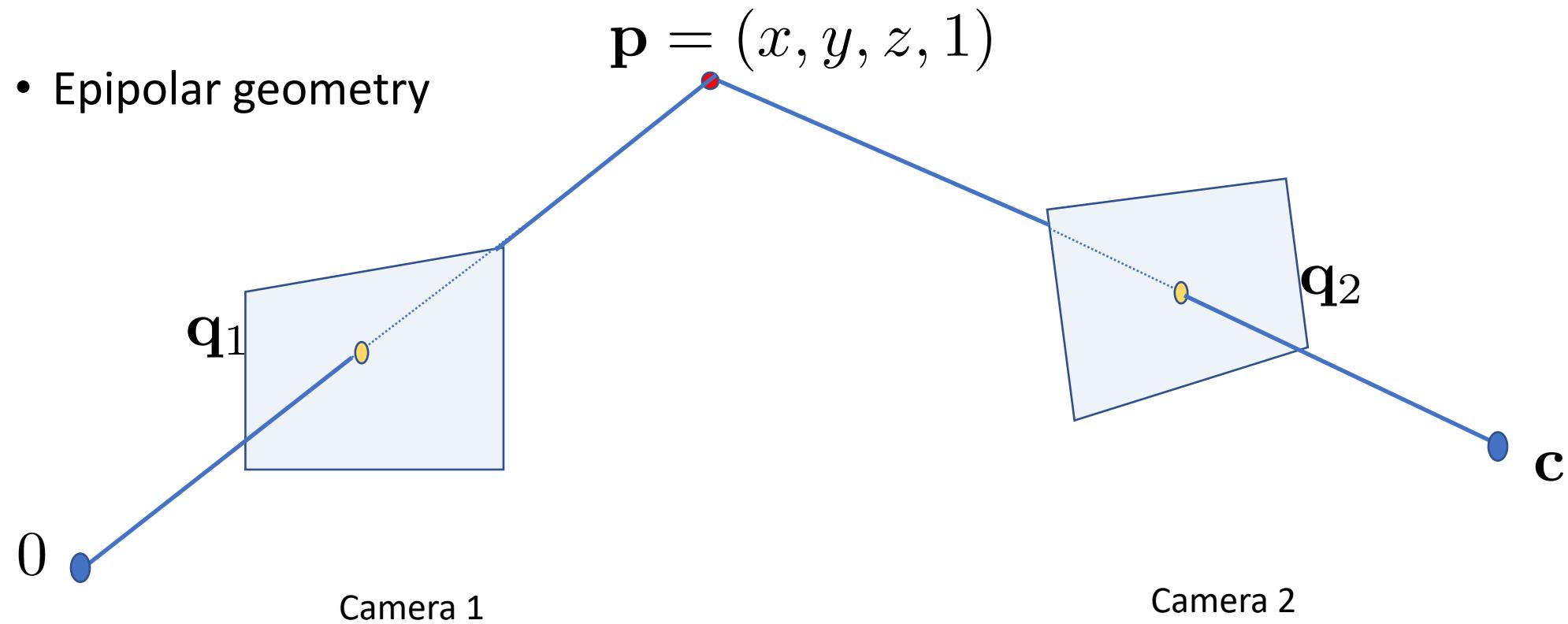
$$K = \begin{bmatrix} f & 0 & c_1 \\ 0 & f & c_2 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\mathbf{q} = K^{-1}\mathbf{q}'$$

# Two-view geometry

- Two pinhole cameras

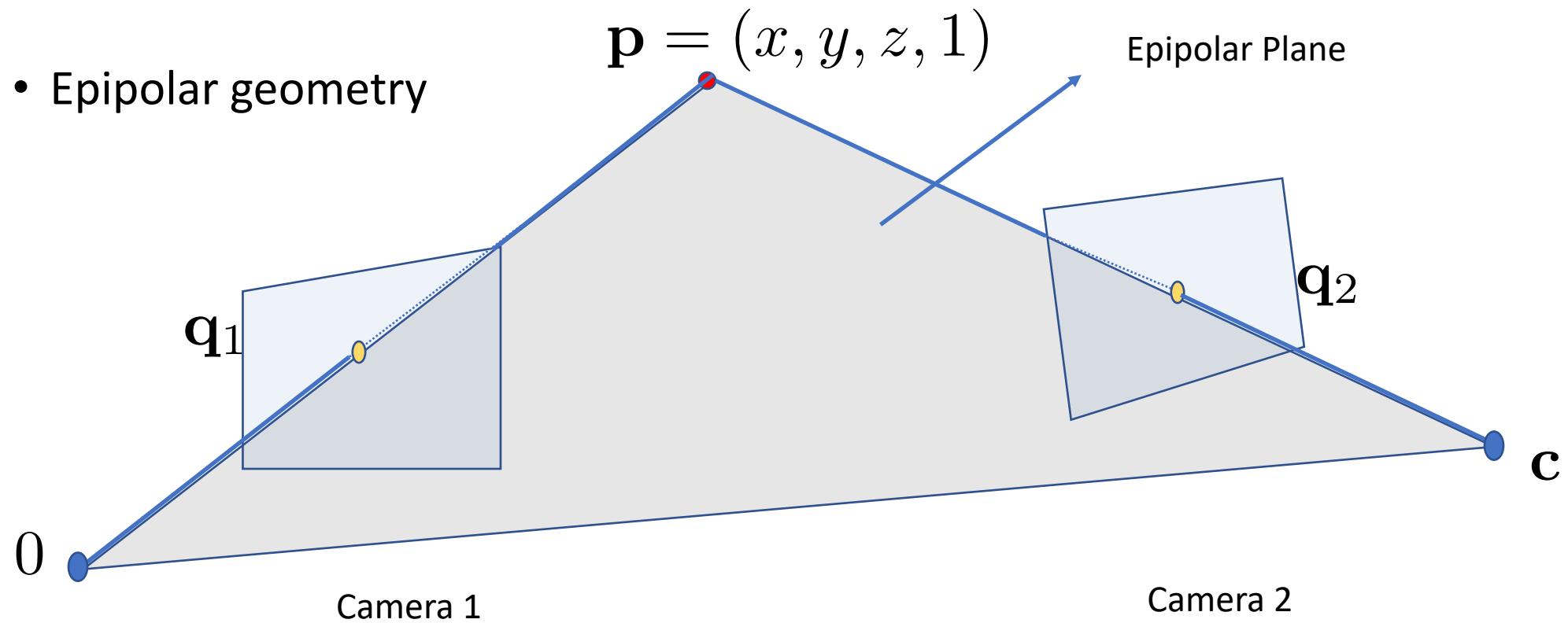
- Epipolar geometry



# Two-view geometry

- Two pinhole cameras

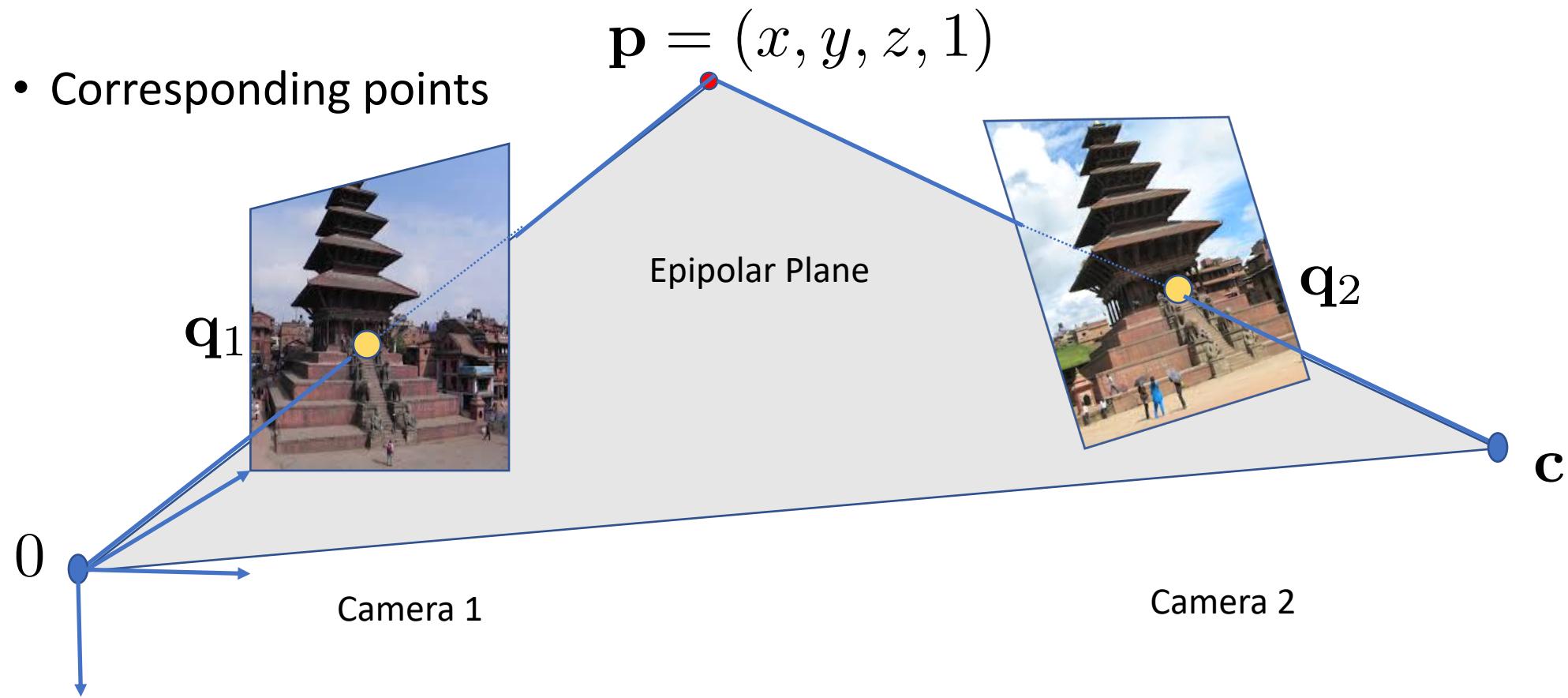
- Epipolar geometry



# Two-view geometry

## ❑ Two pinhole cameras

- Corresponding points



# Two-view geometry

- ❑ Camera 2 has been rotated and translated.

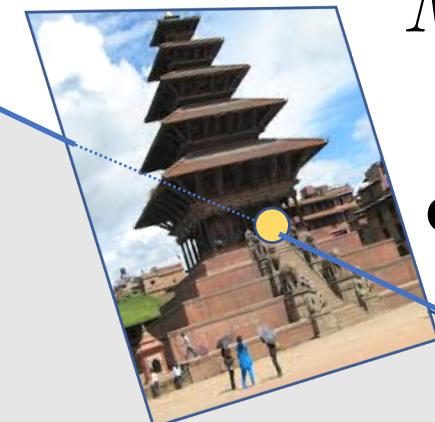
- Corresponding points

$$M_1 = [I \quad \mathbf{0}]$$



$$\mathbf{p} = (x, y, z, 1)$$

$$M_2 = [R \quad \mathbf{t}]$$



0

Camera 1

Camera 2

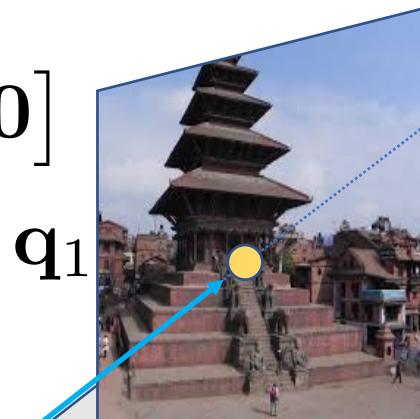
$\mathbf{c}$

# Two-view geometry

$$\mathbf{q}_2 \cdot (\mathbf{t} \times \mathbf{q}_2) = 0$$

$$\mathbf{p} = (x, y, z, 1)$$

$$M_1 = [I \quad \mathbf{0}]$$



$$M_2 = [R \quad \mathbf{t}]$$



0

Camera 1

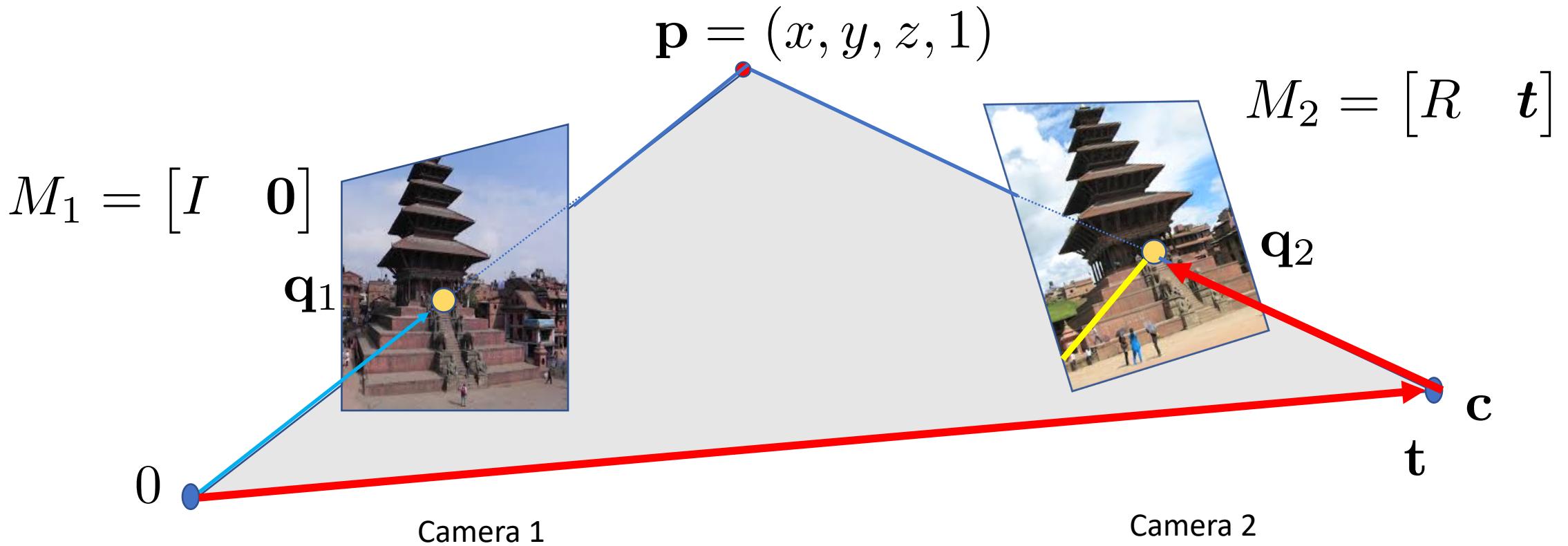
$\mathbf{t}$

Camera 2

$\mathbf{c}$

# Two-view geometry

$$\mathbf{q}_2 \cdot (\mathbf{t} \times R\mathbf{q}_1) = 0 \quad \longrightarrow \quad \mathbf{q}_2^\top [\mathbf{t}]_\times R\mathbf{q}_1 = 0$$



# The Essential Matrix

$$\mathbf{q}_2 \cdot (\mathbf{t} \times R\mathbf{q}_1) = 0 \quad \longrightarrow \quad \mathbf{q}_2^\top [\mathbf{t}]_\times R\mathbf{q}_1 = 0$$

$$E = [t]_\times R$$

❑ Essential matrix is:

- Rank 2 matrix
- 2 equal non-zero singular values
- What about the last singular value?

# The Essential Matrix

- ❑ True for all corresponding points:

$$\mathbf{q}_2^\top E \mathbf{q}_1 = 0$$

- ❑ Mathematically,

$$\text{rank}(E) = 2, \quad \det(E) = 0$$

$$2EE^\top E - \text{trace}(EE^\top)E = 0.$$

# The Fundamental matrix

For cameras with different focal length and principal point

$$\mathbf{q} = K^{-1}\mathbf{q}'$$

$$(K^{-1}\mathbf{q}'_2)^\top [\mathbf{t}]_\times R(K^{-1}\mathbf{q}'_1) = 0$$

$${\mathbf{q}'_2}^\top (K^{-\top} [\mathbf{t}]_\times R K^{-1}) \mathbf{q}'_1 = 0$$

$${\mathbf{q}'_2}^\top F \mathbf{q}'_1 = 0$$

Too many equations!!

# The Fundamental matrix

- ❑ For all corresponding points:

$$\mathbf{q}'_2^\top F \mathbf{q}'_1 = 0$$

- ❑ Properties:

$$\text{rank}(F) = 2, \quad \det(F) = 0$$

- ❑ Lets discuss!! What does it mean by the rank being 2?

# The 8-point Method

- Given many correspondences, compute  $F$  or  $E$

$$A$$

$$\mathbf{f} = 0$$

$$\begin{bmatrix} u_2u_1 & u_2 & v_1 & u_2 & v_2u_1 & v_2v_1 & v_2 & u_1v_1 & 1 \end{bmatrix}$$

$$\begin{bmatrix} f_{11} \\ f_{12} \\ f_{13} \\ f_{21} \\ f_{22} \\ f_{23} \\ f_{31} \\ f_{32} \\ f_{33} \end{bmatrix} = 0 \quad \|\mathbf{f}\| = 1$$

Requires at least 8 points to solve!

# The 8-point Method

- Given many correspondences, compute  $F$  or  $E$

$$\begin{bmatrix} u_2 & v_2 & 1 \end{bmatrix} \begin{bmatrix} f_{11} & f_{12} & f_{13} \\ f_{21} & f_{22} & f_{23} \\ f_{31} & f_{32} & f_{33} \end{bmatrix} \begin{bmatrix} u_1 \\ v_1 \\ 1 \end{bmatrix} = 0$$

$$u_2u_1f_{11} + u_2v_1f_{12} + u_2f_{13} + v_2u_1f_{21} + v_2v_1f_{22} + v_2f_{23} + u_1f_{31} + v_1f_{32} + f_{33} = 0$$

# The 8-point Method

- Given many correspondences, compute  $F$  or  $E$

$$u_2 u_1 f_{11} + u_2 v_1 f_{12} + u_2 f_{13} + v_2 u_1 f_{21} + v_2 v_1 f_{22} + v_2 f_{23} + u_1 f_{31} + v_1 f_{32} + f_{33} = 0$$

$$\begin{bmatrix} u_2 u_1 & u_2 v_1 & u_2 & v_2 u_1 & v_2 v_1 & v_2 & u_1 & v_1 & 1 \end{bmatrix} \begin{bmatrix} f_{11} \\ f_{12} \\ f_{13} \\ f_{21} \\ f_{22} \\ f_{23} \\ f_{31} \\ f_{32} \\ f_{33} \end{bmatrix} = 0$$

# The 5-point Method

- ❑ Can you do better for the Essential matrix?

$$A\mathbf{f} = 0, \quad \|\mathbf{f}\| = 1$$

- ❑ Matrix constraint

$$2EE^\top E - \text{trace}(EE^T)E = 0.$$

# Triangulation

- ❑ Obtain 3D from 2D images

$$\mathbf{p} = (x, y, z, 1)$$

$$M_1 = [I \quad \mathbf{0}]$$

$\mathbf{q}_1$

0



$$M_2 = [R \quad \mathbf{t}]$$

$\mathbf{c}$

$$E = [\mathbf{t}] \times R$$

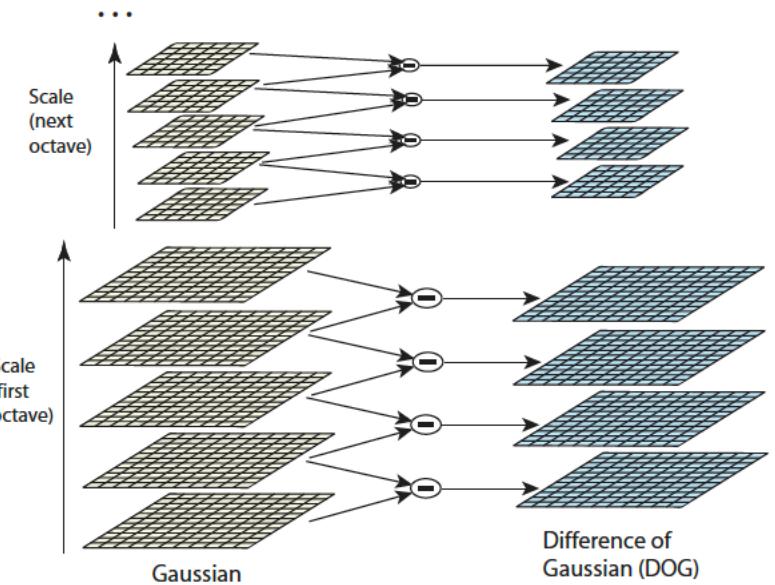
# Corresponding points

- Find “keypoints” in images
  - Scale Invariant Feature Transform (SIFT) – [Lowe, 2004]



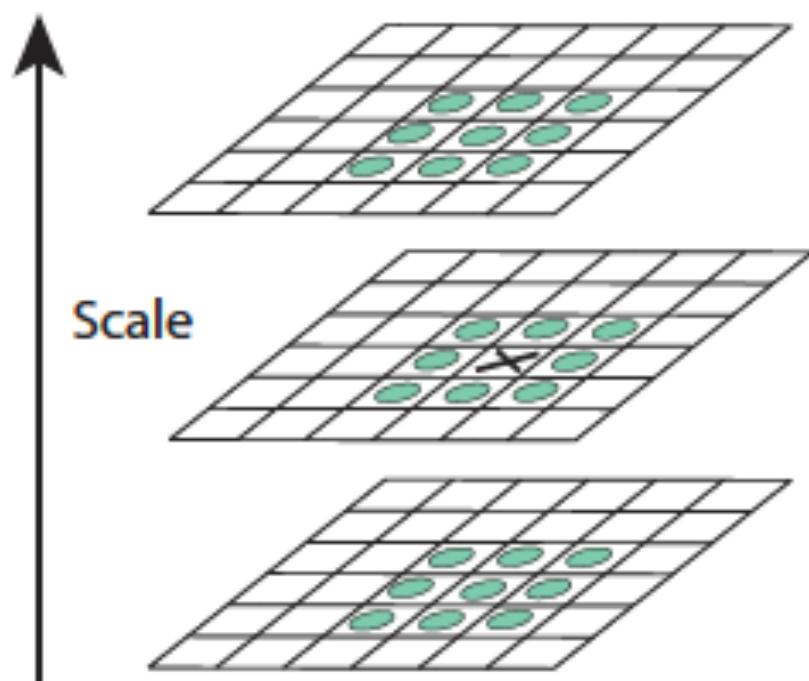
# SIFT

- ☐ Scale Invariant Feature Transform [Lowe 2004]



# SIFT

- ❑ Scale Invariant Feature Transform



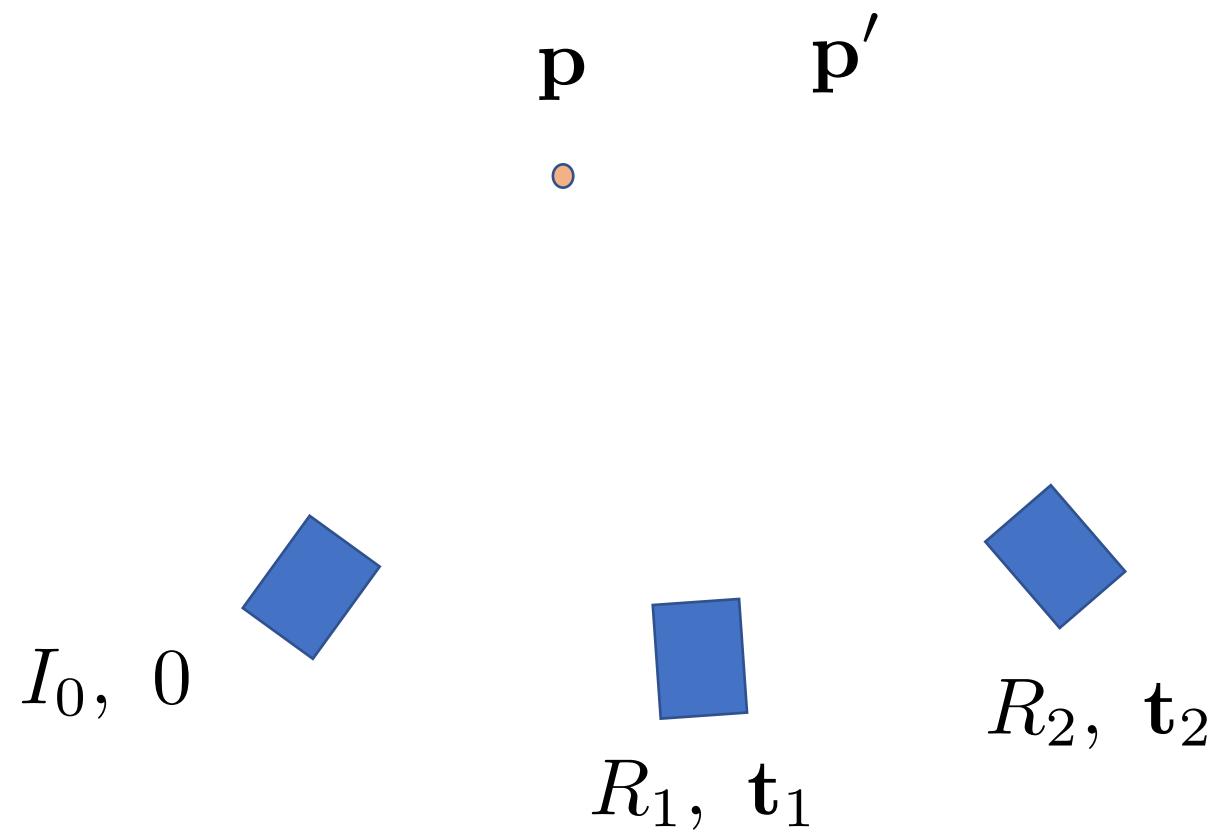
# Structure-from-Motion

## ❑ Multi-view triangulation

- Triangulate with two views
- Transfer points to common reference
- Minimize triangulation error – Bundle adjustment
- Sparse 3D points
- Compute dense 3D points

# Structure-from-Motion

- ❑ Pose computation



# Structure-from-Motion

## ❑ Multi-view triangulation

