

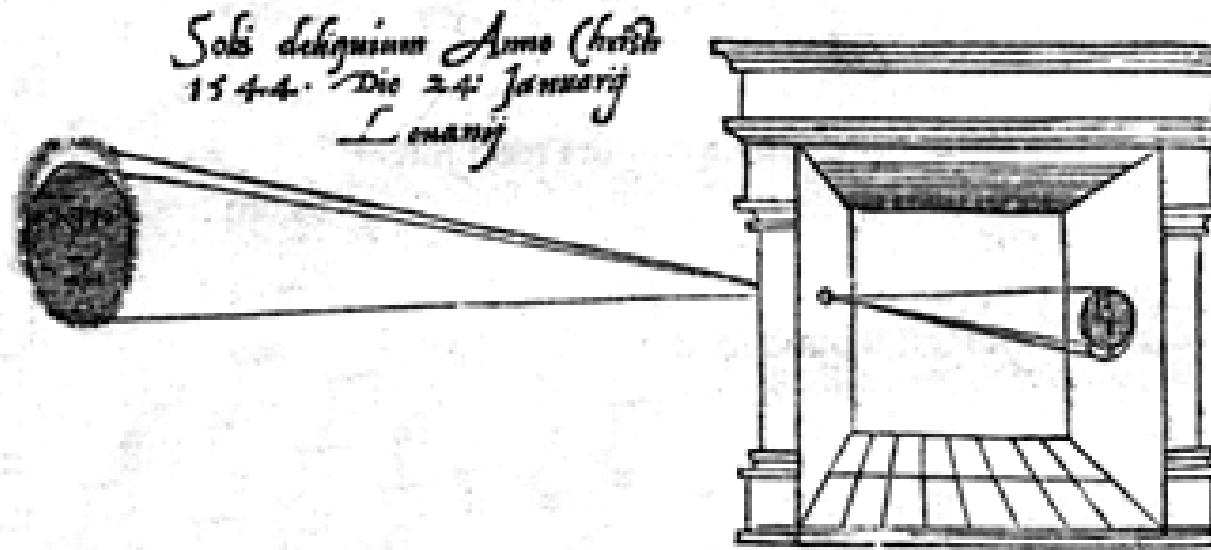
Camera Geometry and single-view Geometry

Camera Geometry

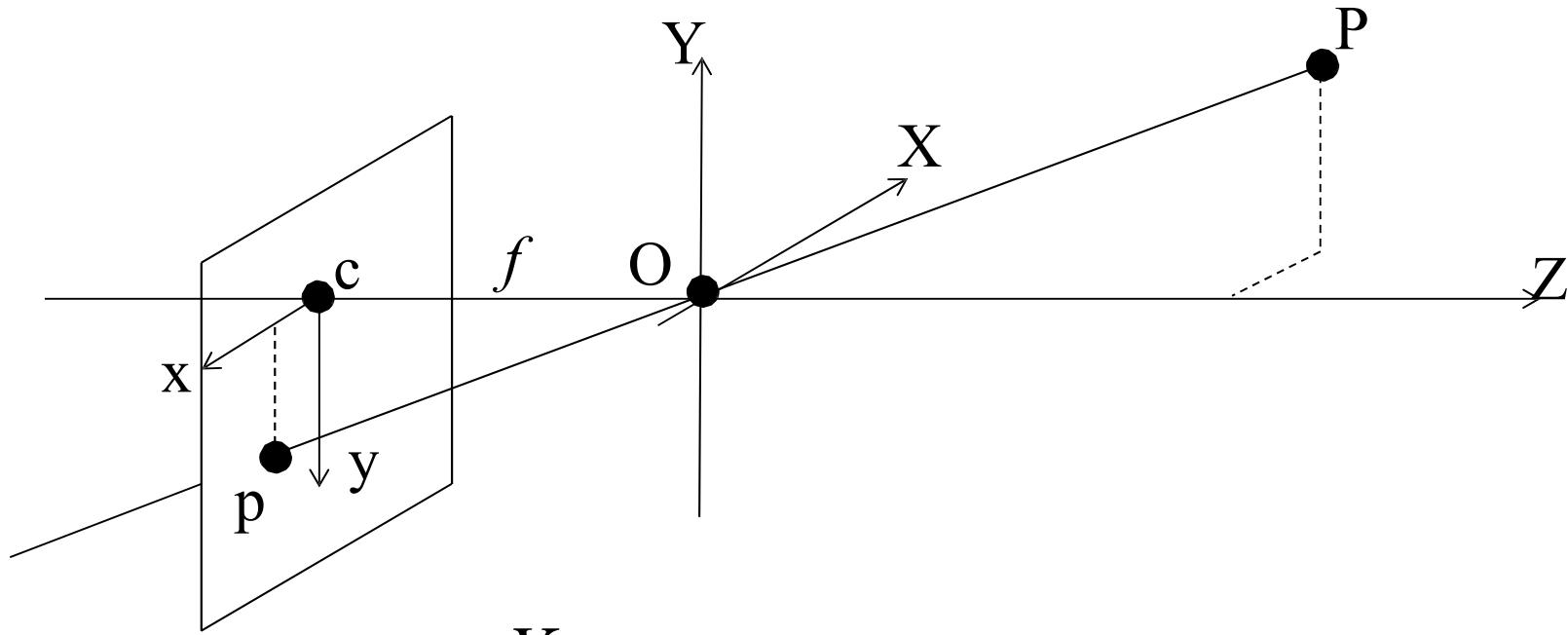
camera projective model

Pinhole camera

illum in tabula per radios Solis , quām in cōelo contin-
git : hoc est , si in cōelo superior pars deliquiū patiatur , in
radiis apparebit inferior deficere , vt ratio exigit optica .



Sic nos exacte Anno . 1544 . Louanii eclipsim Solis
obseruauimus , inuenimusq; deficere paulo plus q̄ dex-



$$x = f \frac{X}{Z}$$

perspective projection

$$y = f \frac{Y}{Z}$$

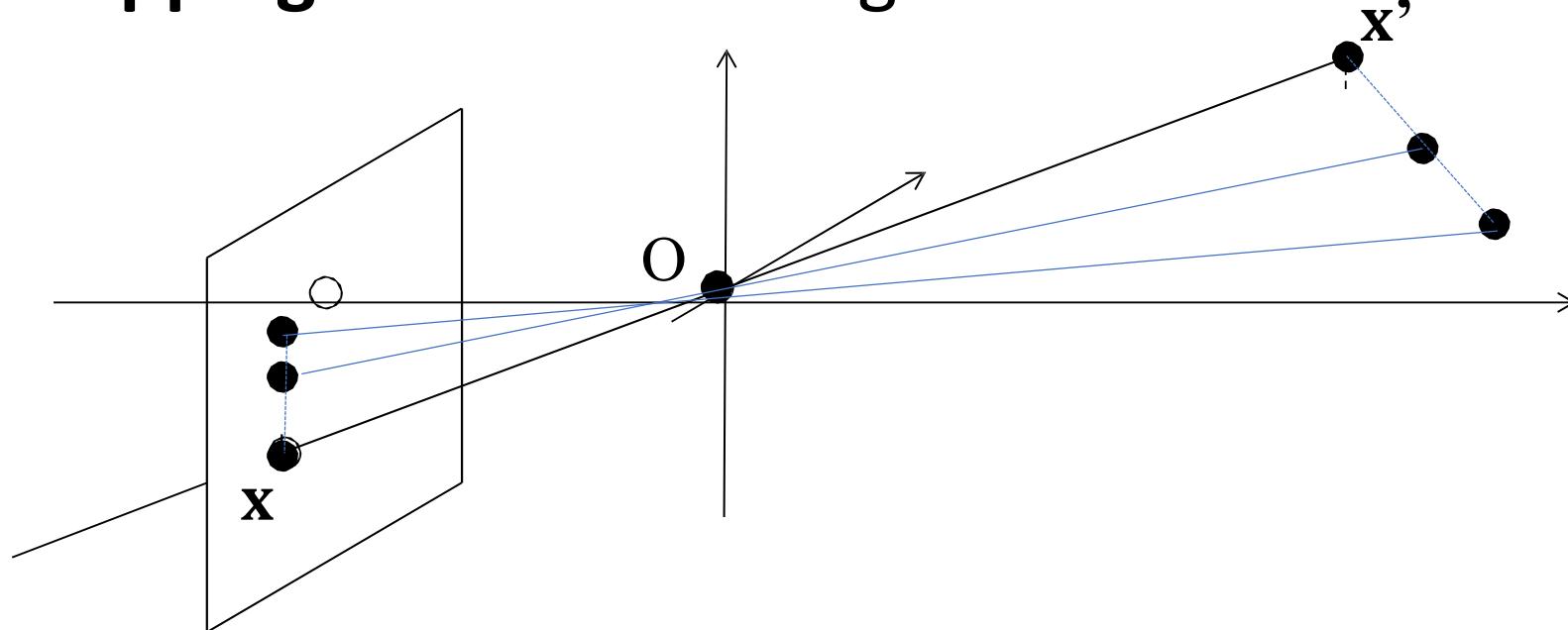
- nonlinear
- not shape-preserving
- not length-ratio preserving

Scene-to-image projection

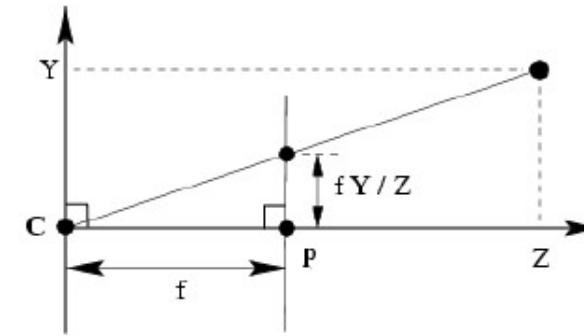
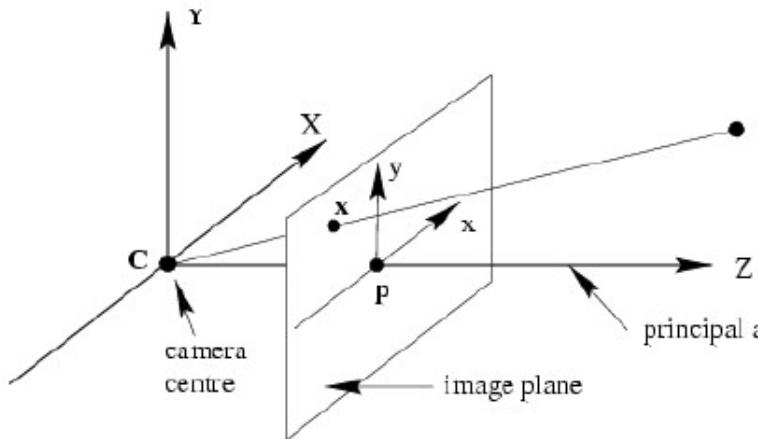
Colinear points are projected onto colinear image points

→ colinearity is preserved

→ **linear mapping** between homogeneous coordinates



CAMERA GEOMETRY



colinearity is preserved → linear relation among homogeneous coords

$$\mathbf{X} = \begin{pmatrix} X \\ Y \\ Z \\ 1 \end{pmatrix} \mapsto \mathbf{u} = \begin{pmatrix} u \\ v \\ w \end{pmatrix} = \mathbf{P}_{3 \times 4} \begin{pmatrix} X \\ Y \\ Z \\ 1 \end{pmatrix} = \mathbf{P}_{3 \times 4} \mathbf{X} = \begin{vmatrix} \mathbf{M}_{3 \times 3} & \mathbf{m}_{3 \times 1} \end{vmatrix} \mathbf{X}$$

3D space image

camera projection matrix

invertible

- SCENE
- CAMERA

SCENE: viewing ray from image point

null-space of camera projection matrix $\mathbf{O} \quad \mathbf{P}\mathbf{O} = 0$

a point \mathbf{Y} on the line \mathbf{X} , $\mathbf{O} \quad \mathbf{Y} = \alpha\mathbf{X} + \beta\mathbf{O}$

its image $\mathbf{u} = \mathbf{P}\mathbf{Y} = \alpha\mathbf{P}\mathbf{X} + \beta\mathbf{P}\mathbf{O} = \mathbf{P}\mathbf{X}$

all points \mathbf{Y} on (\mathbf{X}, \mathbf{O}) project on image of \mathbf{X} ,

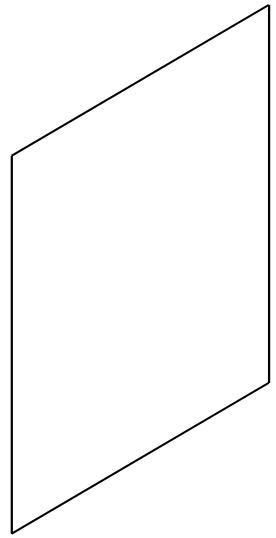
→ $\boxed{\mathbf{O} \text{ is camera center}}$

Image of camera center is $(0,0,0)^T$, i.e. undefined

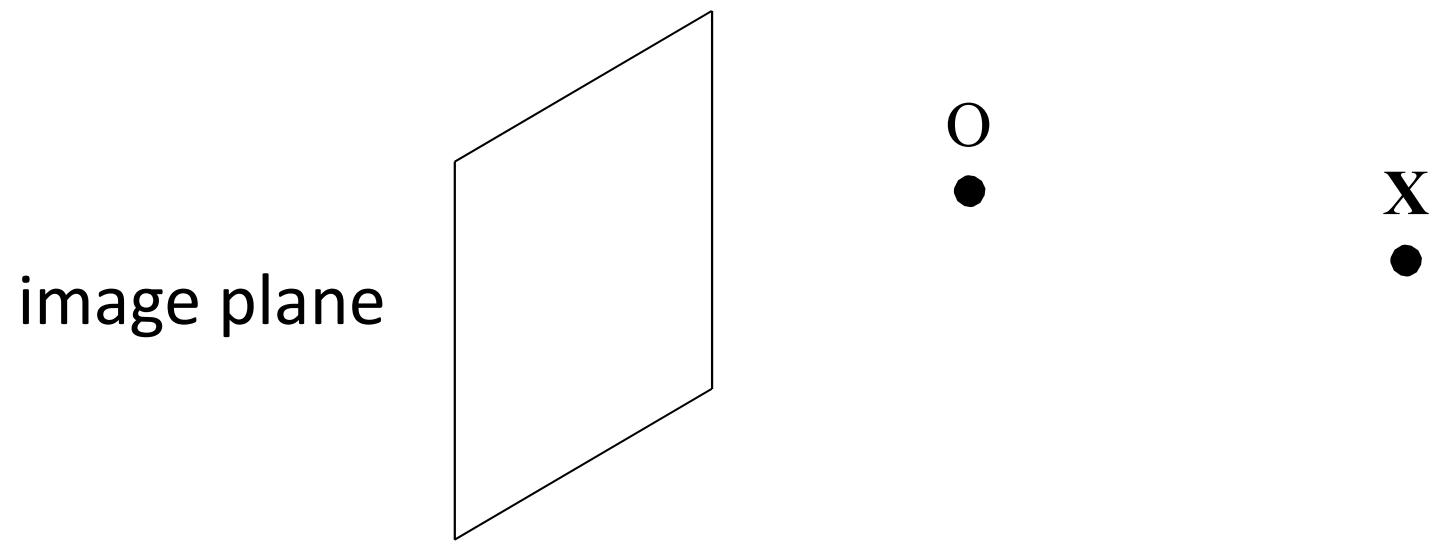
Finite cameras: $\mathbf{O} = \begin{pmatrix} \mathbf{o} \\ 1 \end{pmatrix} = \begin{pmatrix} -\mathbf{M}^{-1}\mathbf{m} \\ 1 \end{pmatrix}$ in fact ...

$$0 = RNS(P) : PO = 0$$

image plane

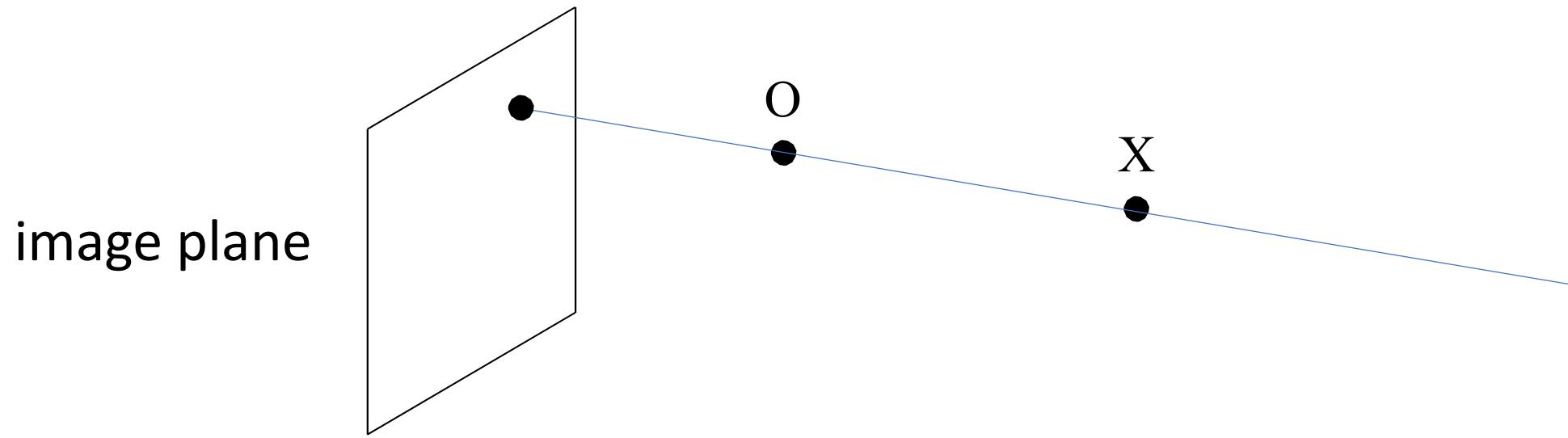


$$0 = RNS(P): PO = 0$$



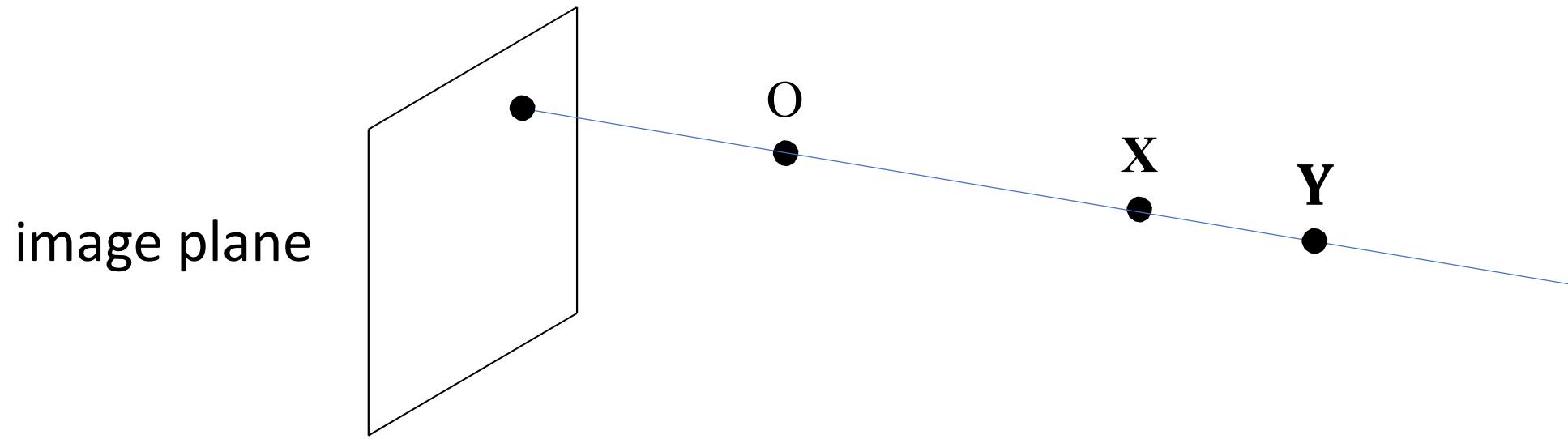
A point **X**

$$0 = RNS(P): PO = 0$$



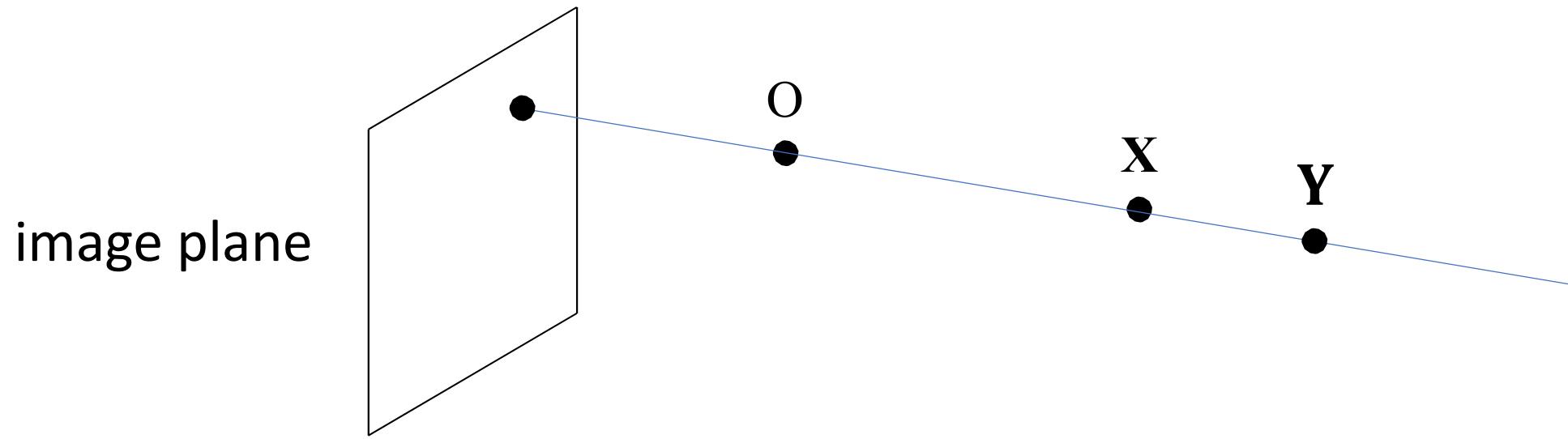
A point **X** and its image

$$0 = RNS(P): PO = 0$$



A point **X** and its image;
any point **Y** on line(**O**, **X**):

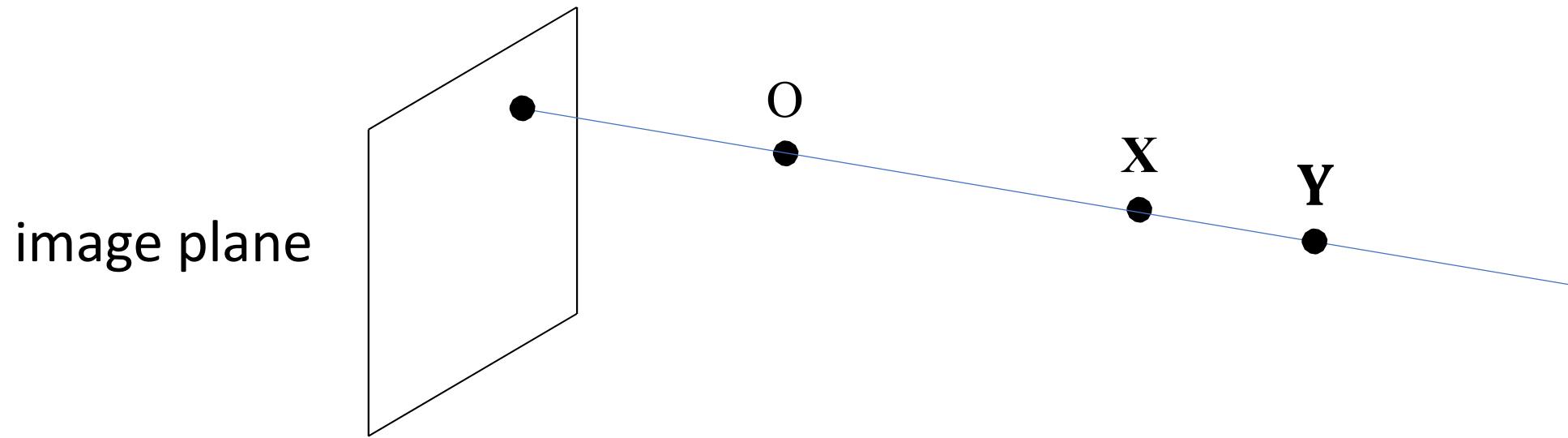
$$0 = RNS(P): PO = 0$$



A point **X** and its image;
any point **Y** on line(**O, X**):

$$\mathbf{Y} = \alpha\mathbf{X} + \beta\mathbf{O}$$

$$0 = RNS(P): PO = 0$$



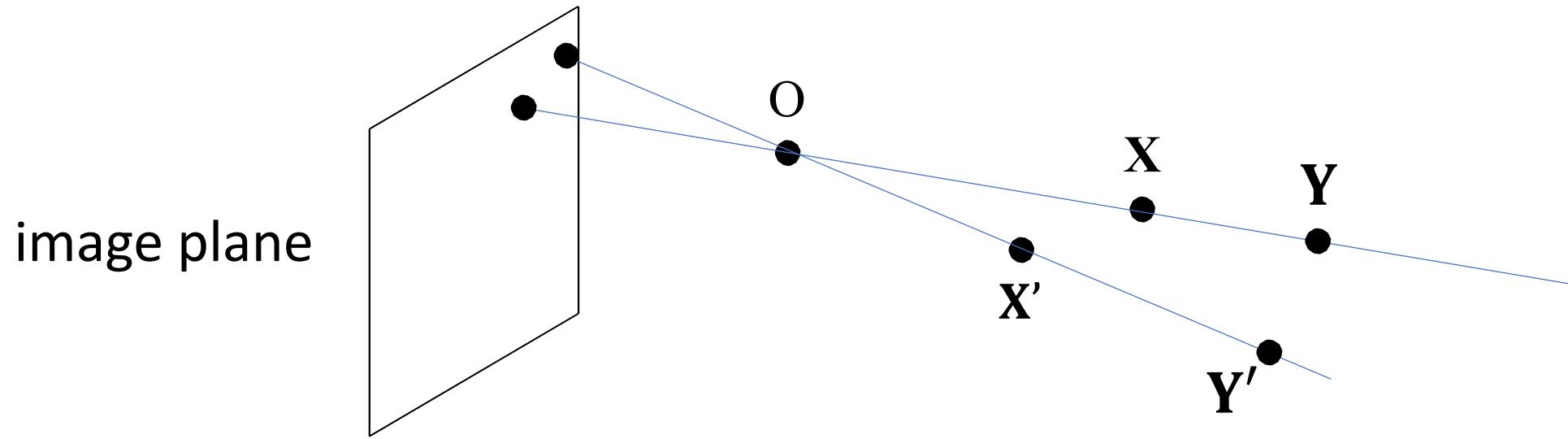
A point \mathbf{X} and its image;
any point \mathbf{Y} on line(\mathbf{O}, \mathbf{X}):

$$\mathbf{Y} = \alpha \mathbf{X} + \beta \mathbf{O}$$

also projects onto the image of \mathbf{X}

$$P\mathbf{Y} = \alpha P\mathbf{X} + \beta PO = \alpha P\mathbf{X} \approx P\mathbf{X}$$

the same for any other point



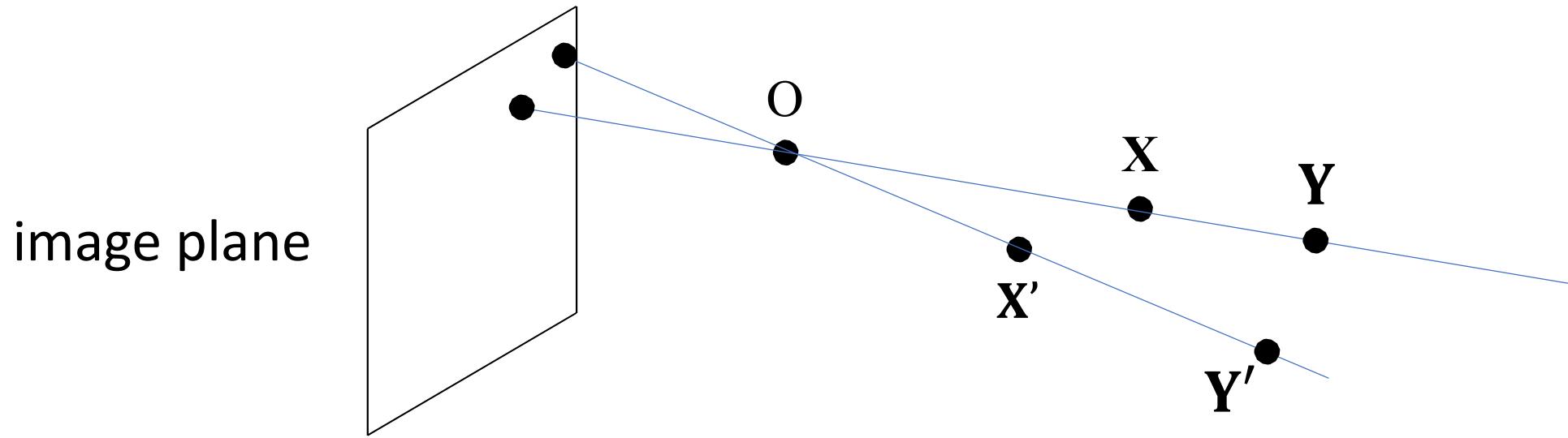
A point **X** and its image;
any point **Y** on line(**O, X**):

$$\mathbf{Y} = \alpha\mathbf{X} + \beta\mathbf{O}$$

also projects onto the image of **X**

$$P\mathbf{Y} = \alpha P\mathbf{X} + \beta P\mathbf{O} = \alpha P\mathbf{X} \approx P\mathbf{X}$$

the same for any other point



A point **X** and its image;
any point **Y** on line(**O, X**):

$$\mathbf{Y} = \alpha\mathbf{X} + \beta\mathbf{O}$$

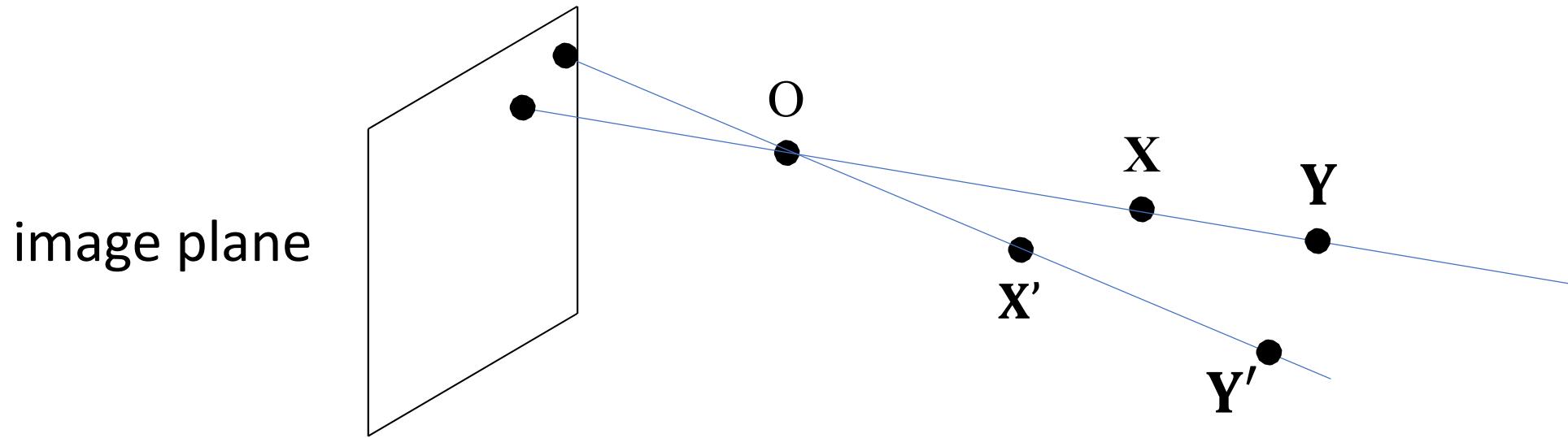
also projects onto the image of **X**

$$P\mathbf{Y} = \alpha P\mathbf{X} + \beta P\mathbf{O} = \alpha P\mathbf{X} \approx P\mathbf{X}$$



all viewing rays go through **O**

the same for any other point



A point \mathbf{X} and its image;
any point \mathbf{Y} on line(\mathbf{O}, \mathbf{X}):

$$\mathbf{Y} = \alpha\mathbf{X} + \beta\mathbf{O}$$

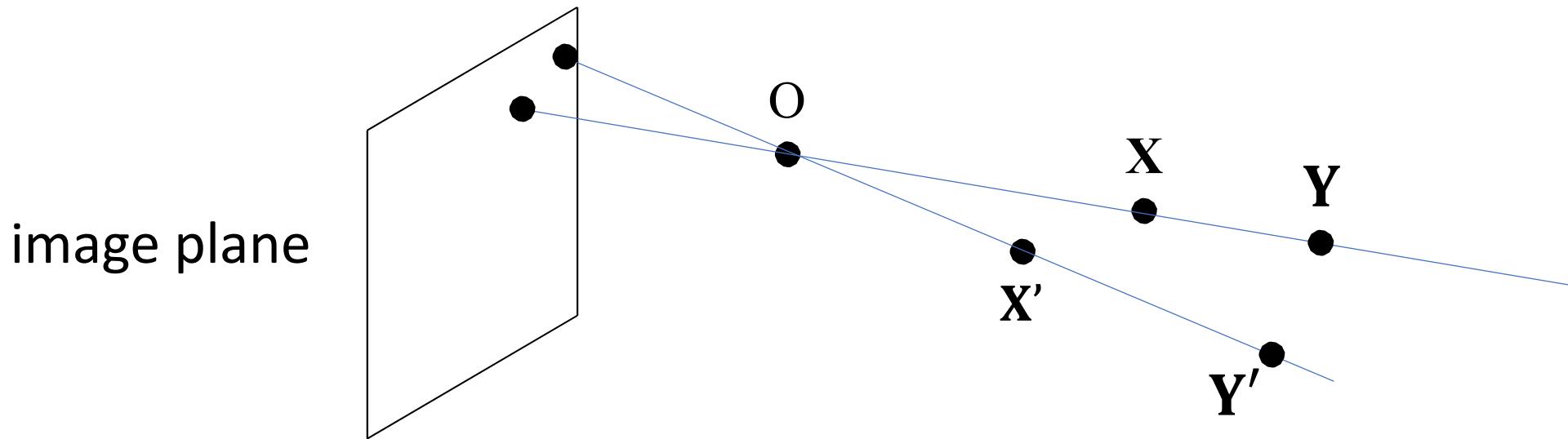
also projects onto the image of \mathbf{X}

$$P\mathbf{Y} = \alpha P\mathbf{X} + \beta P\mathbf{O} = \alpha P\mathbf{X} \approx P\mathbf{X}$$



all viewing rays go through \mathbf{O}
 \mathbf{O} is the camera lens center

the same for any other point



A point \mathbf{X} and its image;
any point \mathbf{Y} on line(\mathbf{O}, \mathbf{X}):

$$\mathbf{Y} = \alpha \mathbf{X} + \beta \mathbf{O}$$

also projects onto the image of \mathbf{X}

$$P\mathbf{Y} = \alpha P\mathbf{X} + \beta PO = \alpha P\mathbf{X} \approx P\mathbf{X}$$



all viewing rays go through \mathbf{O}
 \mathbf{O} is the camera lens center

$$PO = [\mathbf{M} \quad \mathbf{m}] \begin{bmatrix} \mathbf{0} \\ 1 \end{bmatrix} = 0$$

$$\rightarrow \mathbf{o} = -\mathbf{M}^{-1}\mathbf{m}$$

cartesian coordinates of \mathbf{O}

SCENE: viewing ray from image point

null-space of camera projection matrix $\mathbf{O} \quad \mathbf{P}\mathbf{O} = 0$

a point \mathbf{Y} on the line \mathbf{X} , $\mathbf{O} \quad \mathbf{Y} = \alpha\mathbf{X} + \beta\mathbf{O}$

its image $\mathbf{u} = \mathbf{P}\mathbf{Y} = \alpha\mathbf{P}\mathbf{X} + \beta\mathbf{P}\mathbf{O} = \mathbf{P}\mathbf{X}$

all points \mathbf{Y} on (\mathbf{X}, \mathbf{O}) project on image of \mathbf{X} ,

$\rightarrow \boxed{\mathbf{O} \text{ is camera center}}$

all points on a line
through \mathbf{O} share the
same image

Image of camera center is $(0,0,0)^T$, i.e. undefined

Finite cameras: $\mathbf{O} = \begin{pmatrix} \mathbf{o} \\ 1 \end{pmatrix} = \begin{pmatrix} -\mathbf{M}^{-1}\mathbf{m} \\ 1 \end{pmatrix}$

Viewing ray associated to image point \mathbf{u} :

image point \mathbf{u} is image of \mathbf{X} if \mathbf{X} is on a certain line through \mathbf{o}

viewing ray associated to \mathbf{u}

Consider the point at the infinity along this line

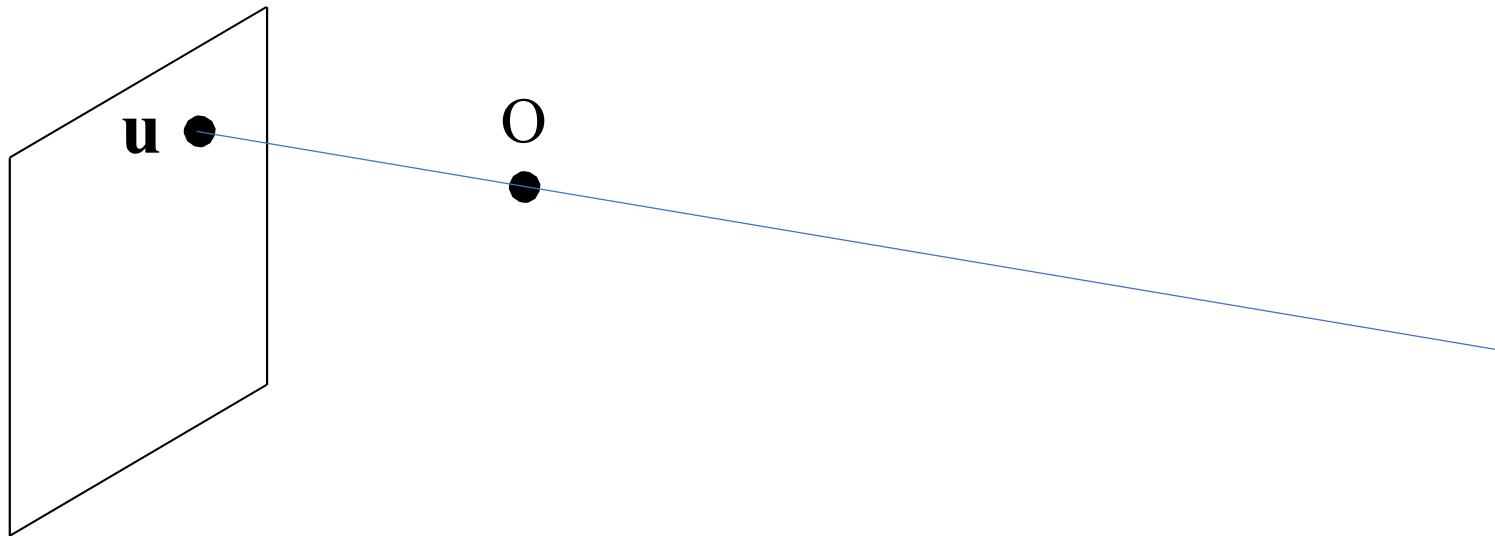
$$\mathbf{X} = \begin{vmatrix} \mathbf{d} \\ 0 \end{vmatrix}$$
$$\mathbf{u} = \begin{vmatrix} \mathbf{M} & \mathbf{m} \end{vmatrix} \begin{vmatrix} \mathbf{d} \\ 0 \end{vmatrix} = \mathbf{M} \cdot \mathbf{d}$$

*The locus of the points \mathbf{x} whose image is \mathbf{u}
is a straight line through \mathbf{o} having direction $\mathbf{d} = \mathbf{M}^{-1} \cdot \mathbf{u}$*

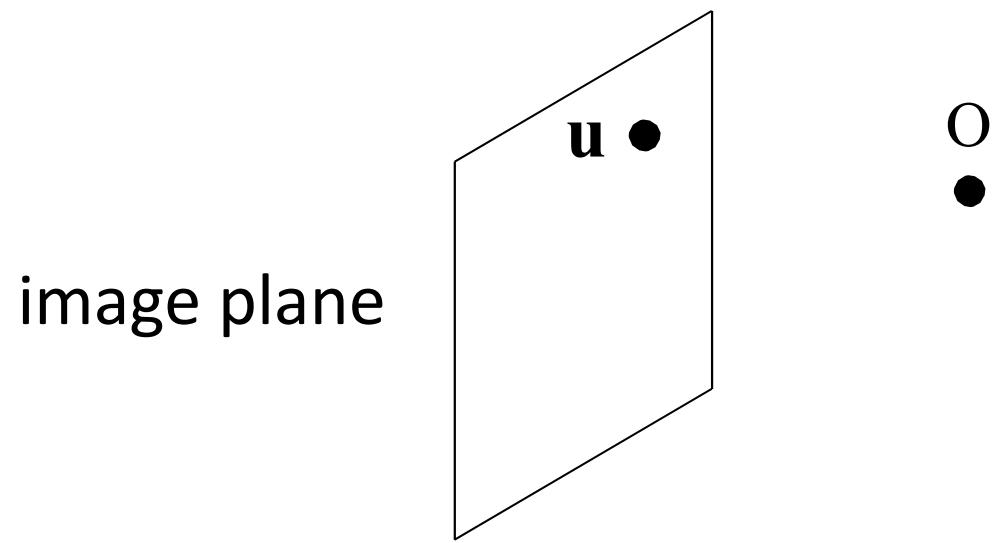
\mathbf{o} is the camera viewpoint (perspective projection center)

$\text{line}(\mathbf{o}, \mathbf{d})$ = viewing ray associated to image point \mathbf{u}

Viewing ray associated to an image point \mathbf{u} :
the «backprojection» of \mathbf{u}

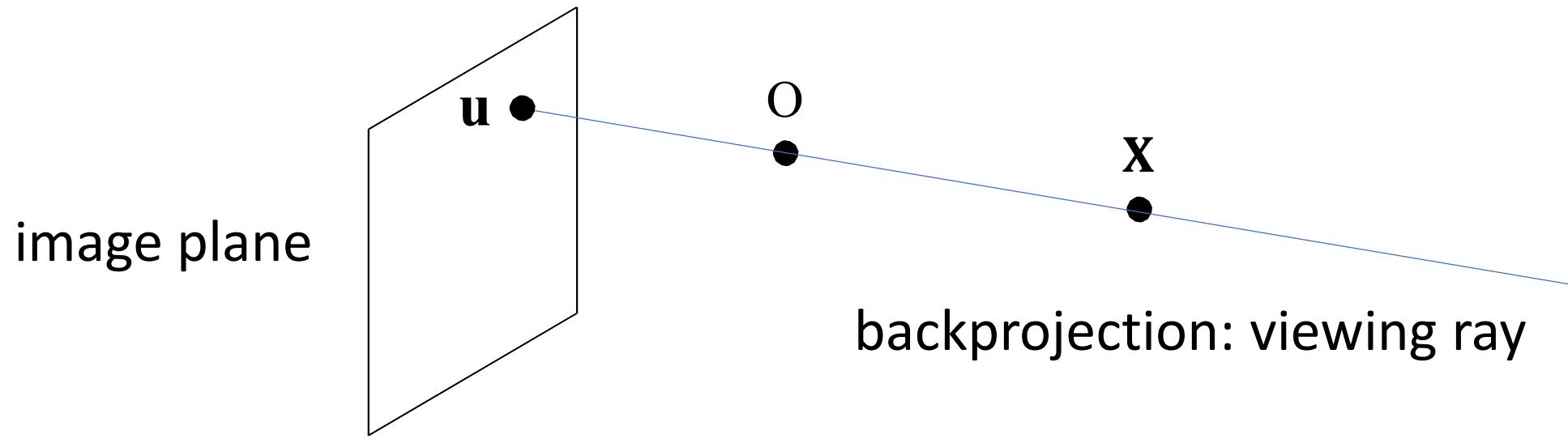


$$0 = RNS(P): PO = 0$$



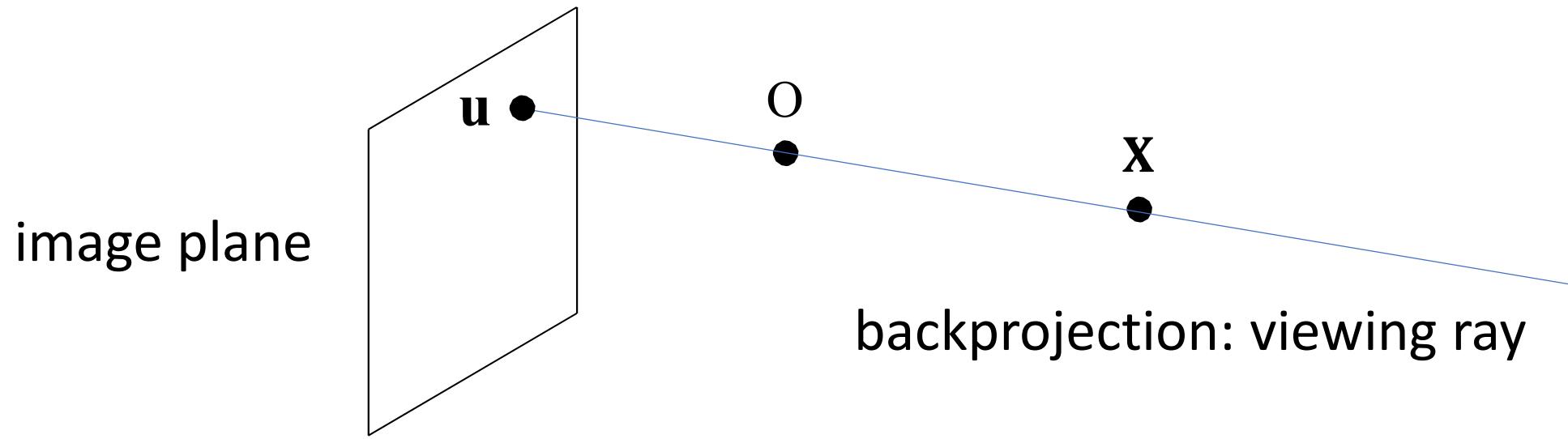
An image point \mathbf{u}

$$0 = RNS(P): PO = 0$$



An image point **u** and its backprojection: set of 3D points **X** projecting onto **u**

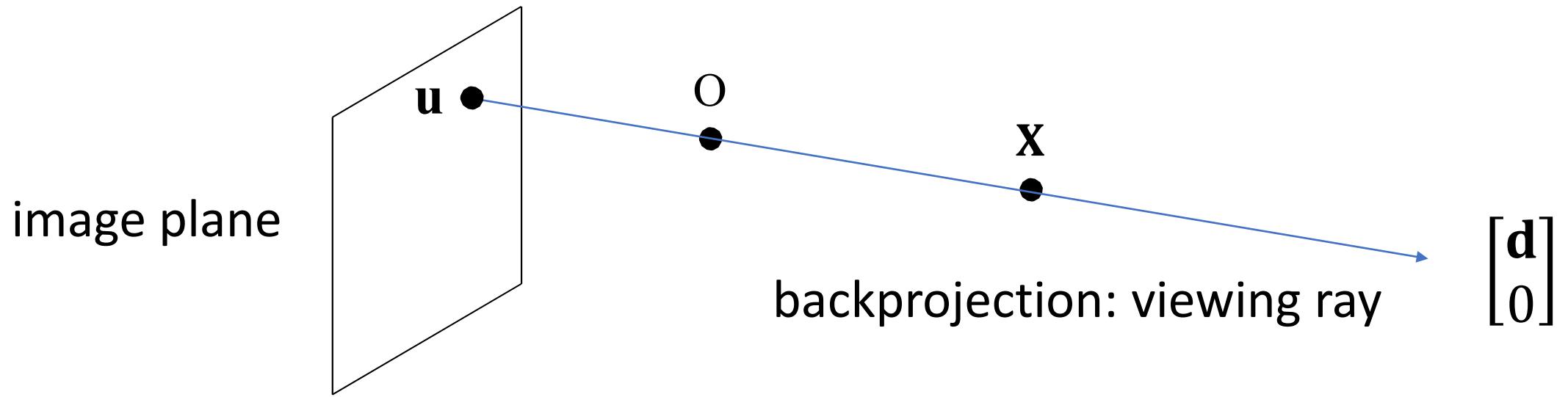
$$O = RNS(P): PO = 0$$



An image point \mathbf{u} and its backprojection: set of 3D points \mathbf{X} projecting onto \mathbf{u}

(i) a line through O

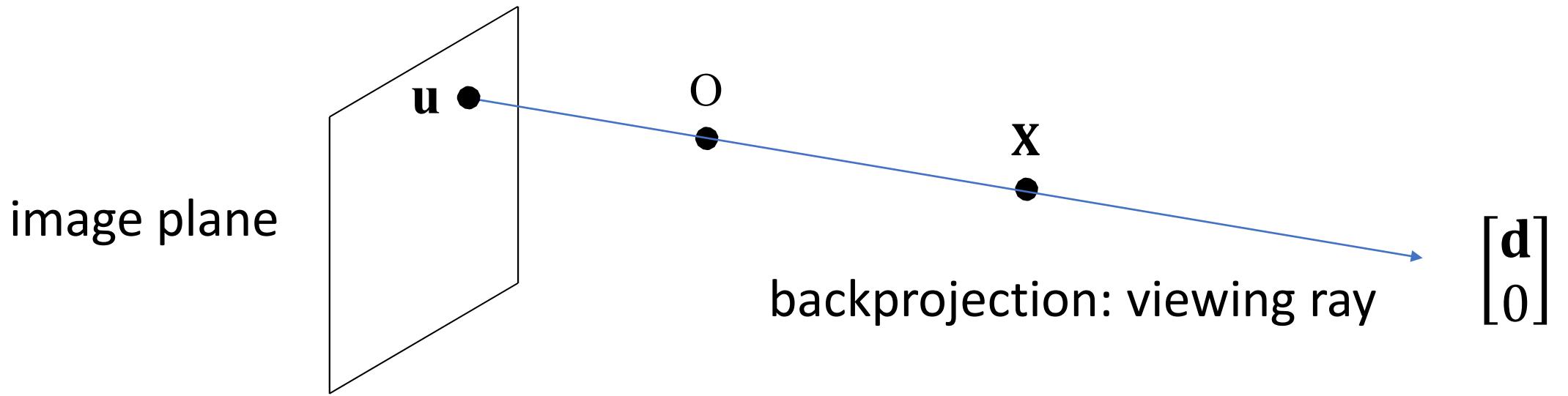
$$0 = RNS(P) : PO = 0$$



An image point u and its backprojection: set of 3D points X projecting onto u

- (i) a line through O
- (ii) whose direction d ...

$$0 = RNS(P) : PO = 0$$

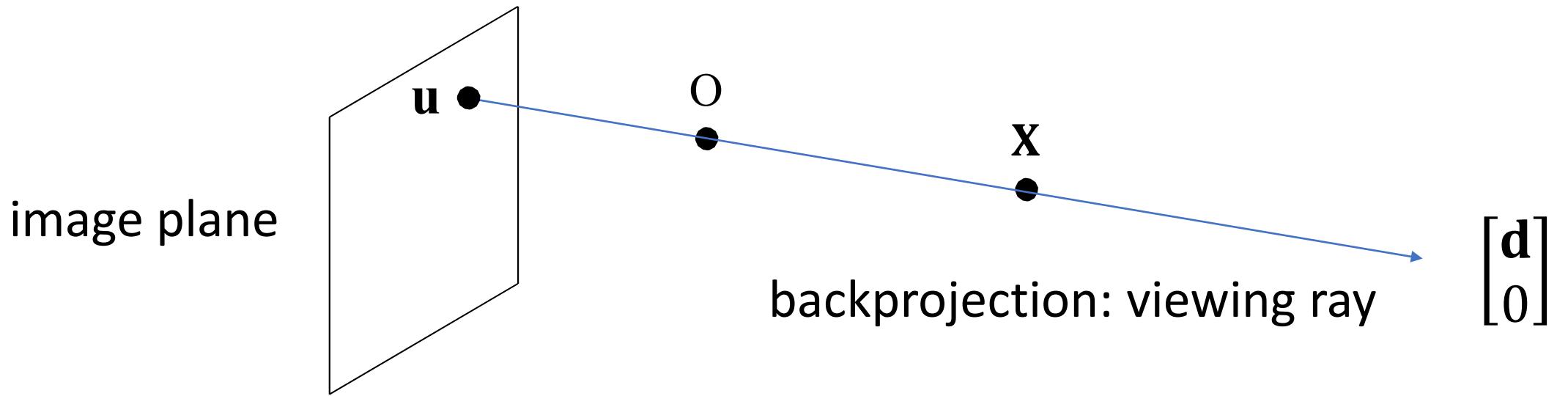


An image point \mathbf{u} and its backprojection: set of 3D points \mathbf{X} projecting onto \mathbf{u}

(i) a line through O

(ii) whose direction \mathbf{d} satisfies $\mathbf{u} = [\mathbf{M} \quad \mathbf{m}] \begin{bmatrix} \mathbf{d} \\ 0 \end{bmatrix} = \mathbf{Md}$

$$0 = RNS(P): PO = 0$$



An image point u and its backprojection: set of 3D points X projecting onto u

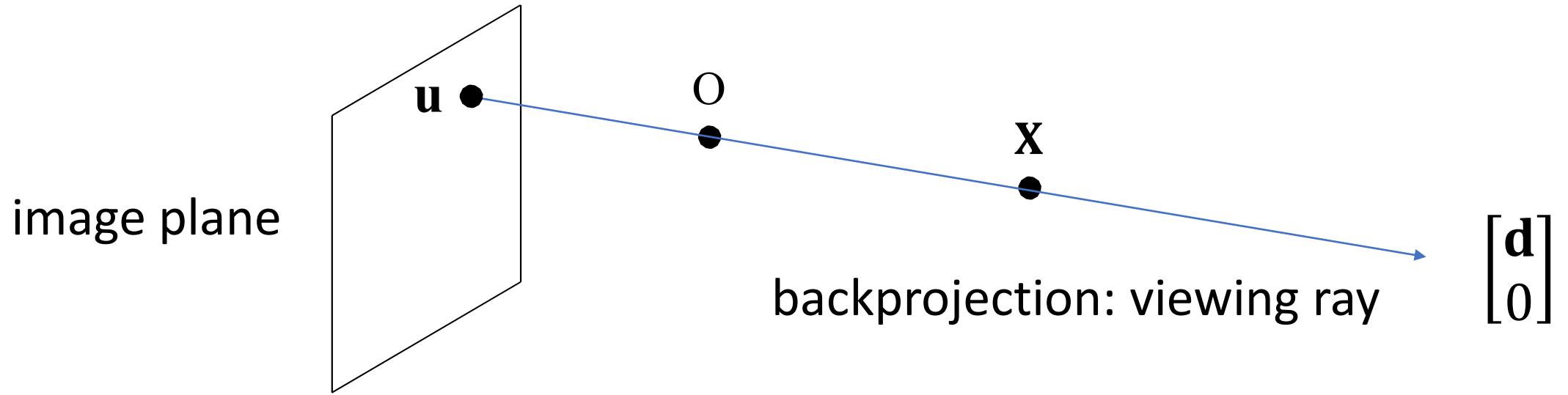
(i) a line through O

(ii) whose direction d satisfies $u = [M \quad m] \begin{bmatrix} d \\ 0 \end{bmatrix} = Md$



$$d = M^{-1}u$$

the viewing ray associated to an image point \mathbf{u}



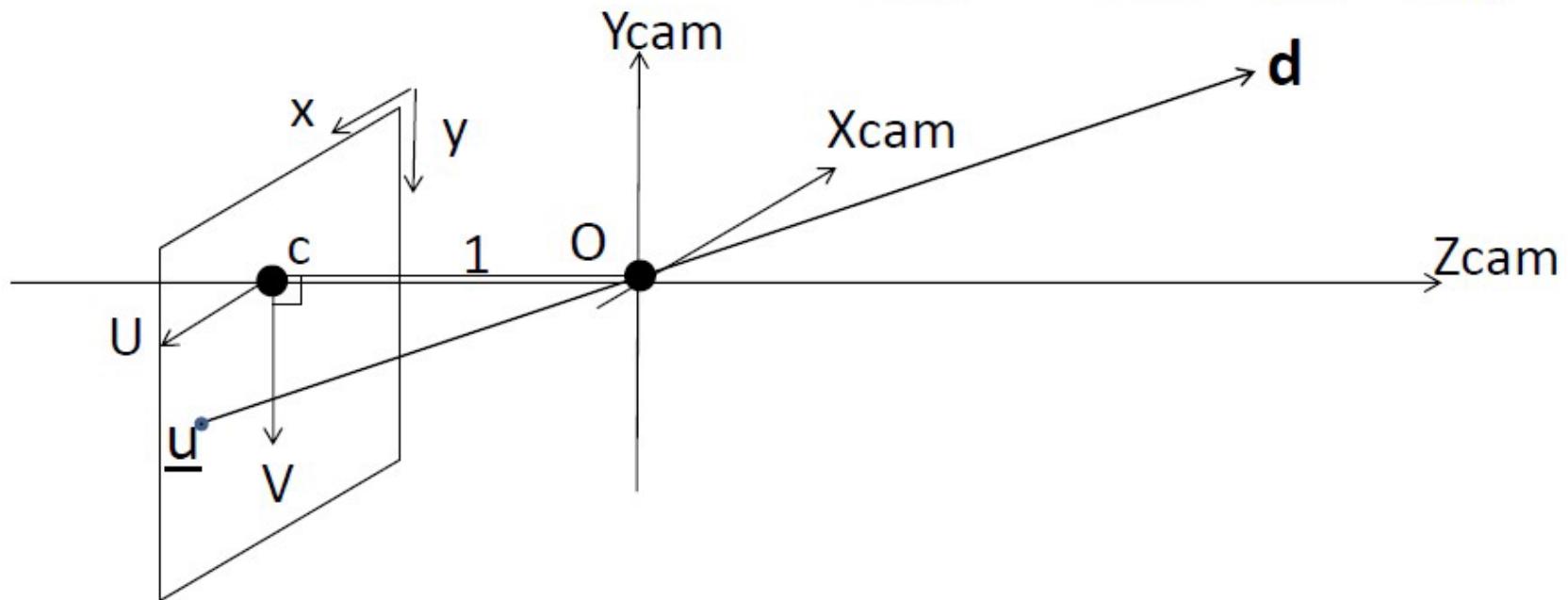
The backprojection of image point \mathbf{u} , i.e., the viewing ray associated to \mathbf{u} , for the camera $P = [\mathbf{M} \quad \mathbf{m}]$, is the straight line

- (i) through $O = RNS(P)$
- (ii) whose direction \mathbf{d} is $\mathbf{d} = \mathbf{M}^{-1}\mathbf{u}$

- SCENE
- CAMERA

CAMERA

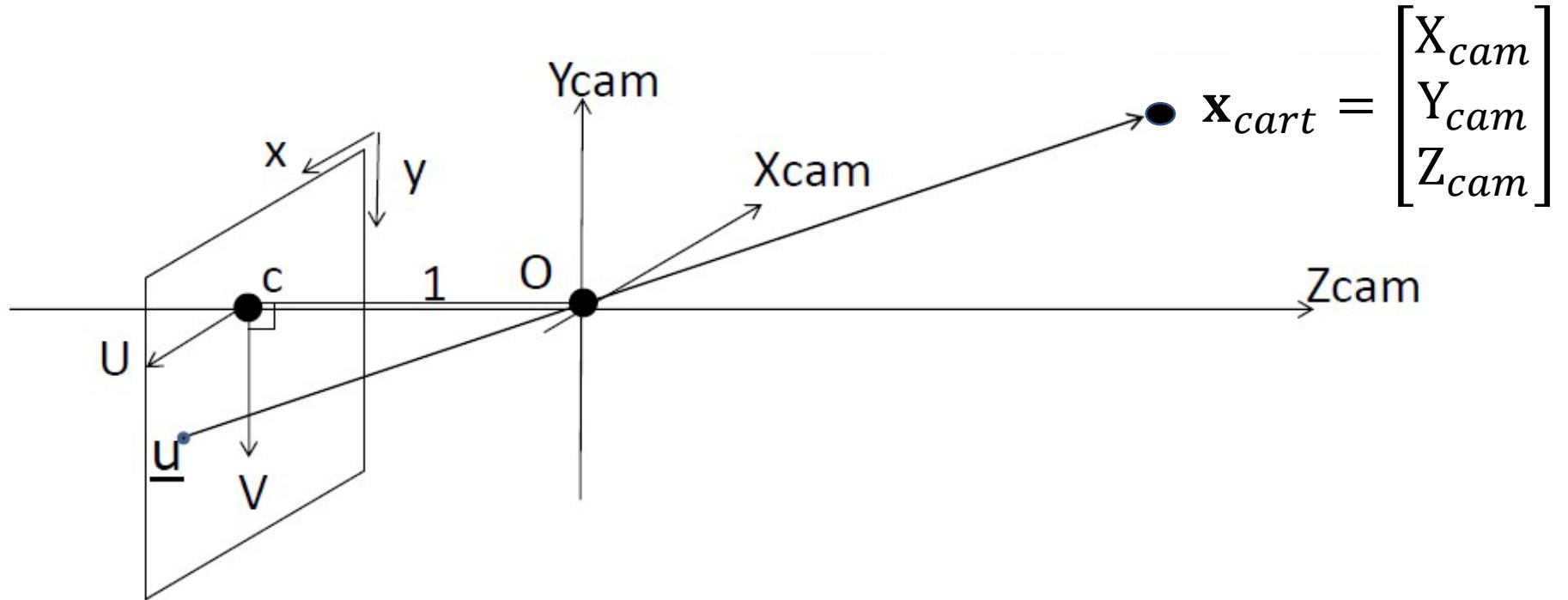
$$\mathbf{d}_{cam} = \mathbf{R}_{cam \rightarrow world} \mathbf{d}_{world}$$



geometric coordinates

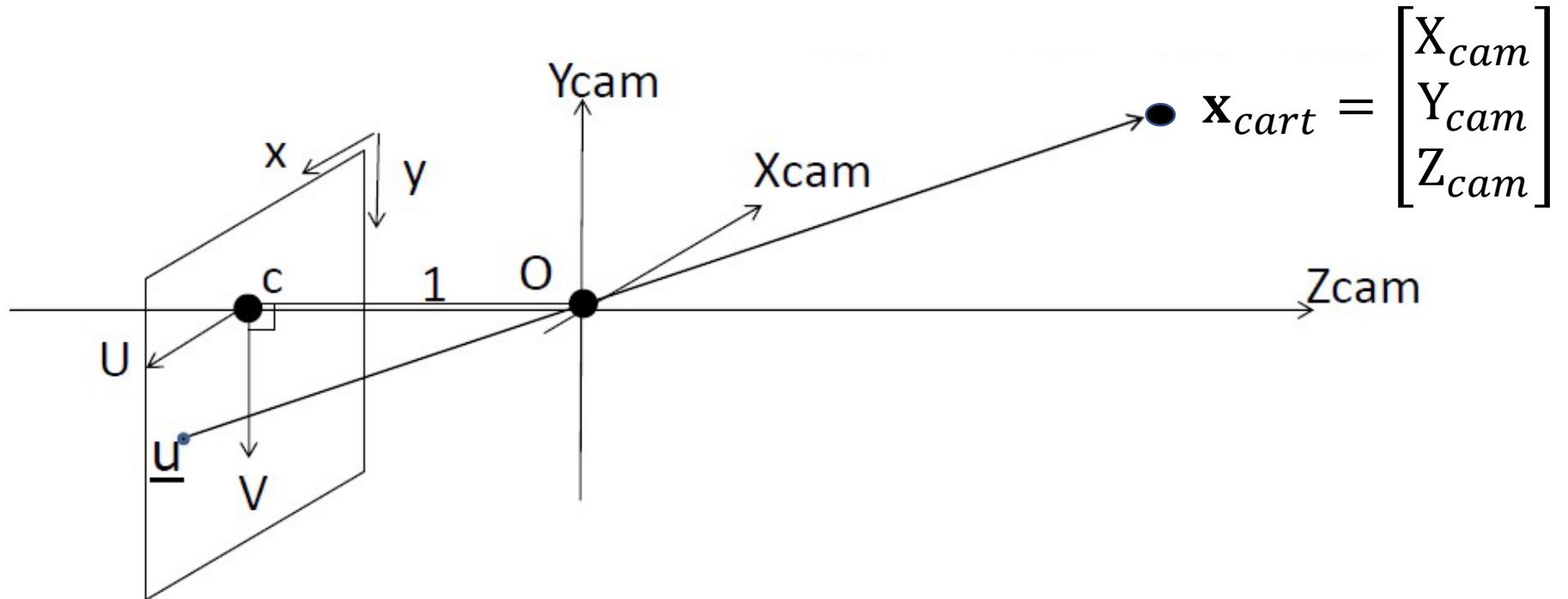
$$\begin{vmatrix} U \\ V \\ 1 \end{vmatrix} = \begin{vmatrix} X \\ Y \\ -1 \end{vmatrix} \propto \mathbf{d}_{cam}$$

camera



geometric coordinates $\begin{bmatrix} U \\ V \\ 1 \end{bmatrix} \propto \begin{bmatrix} X_{cam} \\ Y_{cam} \\ Z_{cam} \end{bmatrix} = \mathbf{x}_{cam}$

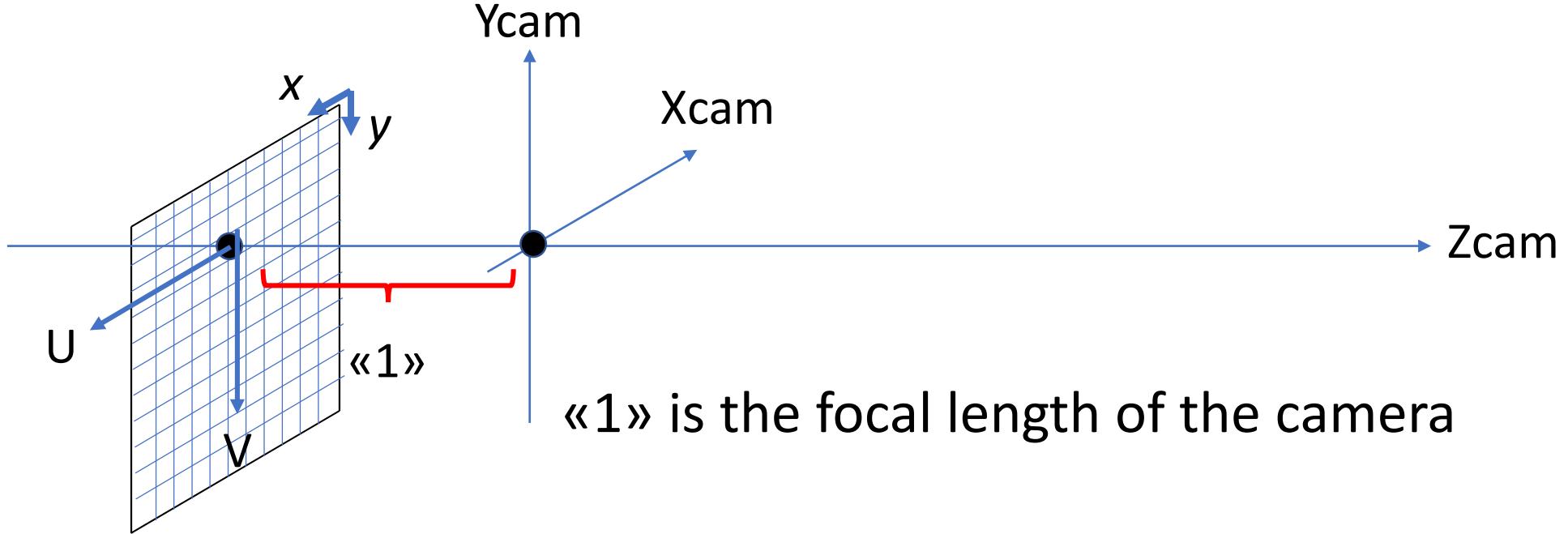
camera



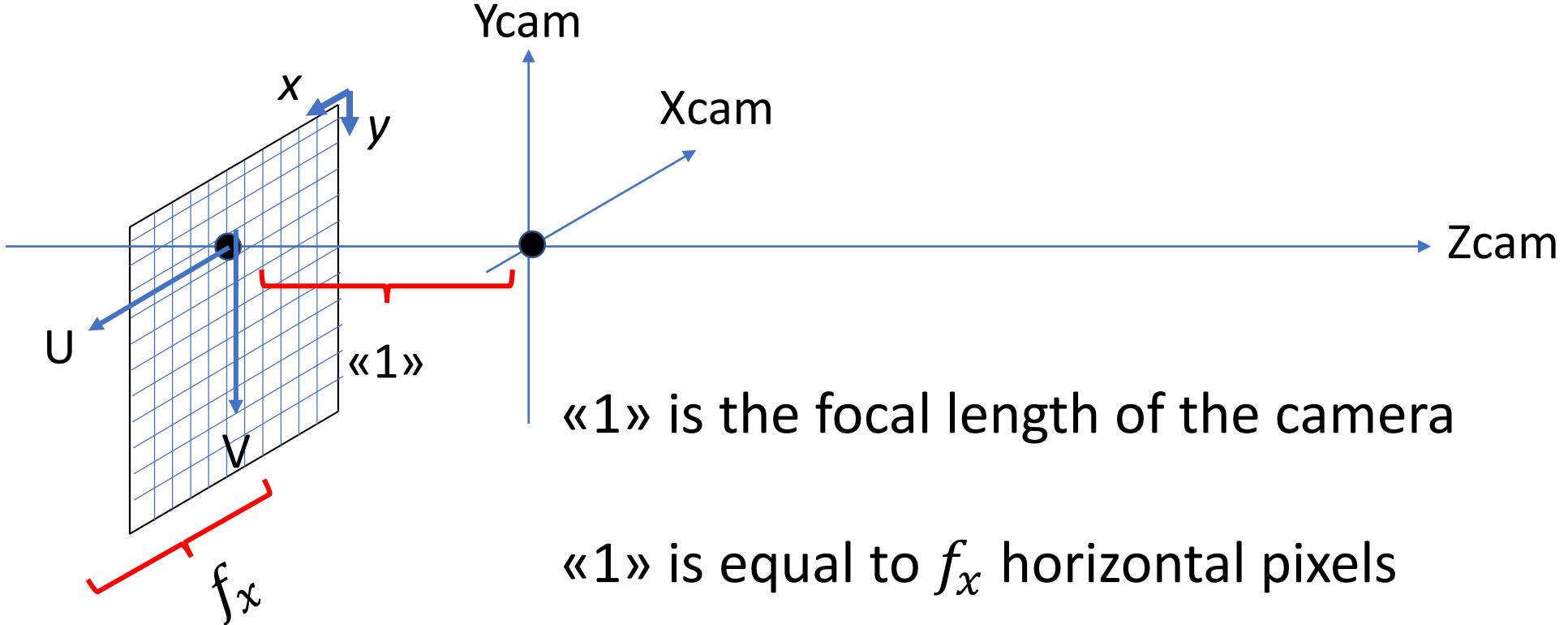
geometric coordinates $\begin{bmatrix} U \\ V \\ 1 \end{bmatrix} \propto \begin{bmatrix} X_{cam} \\ Y_{cam} \\ Z_{cam} \end{bmatrix} = \mathbf{x}_{cart}$

geometric coordinates vs pixel coordinates ??

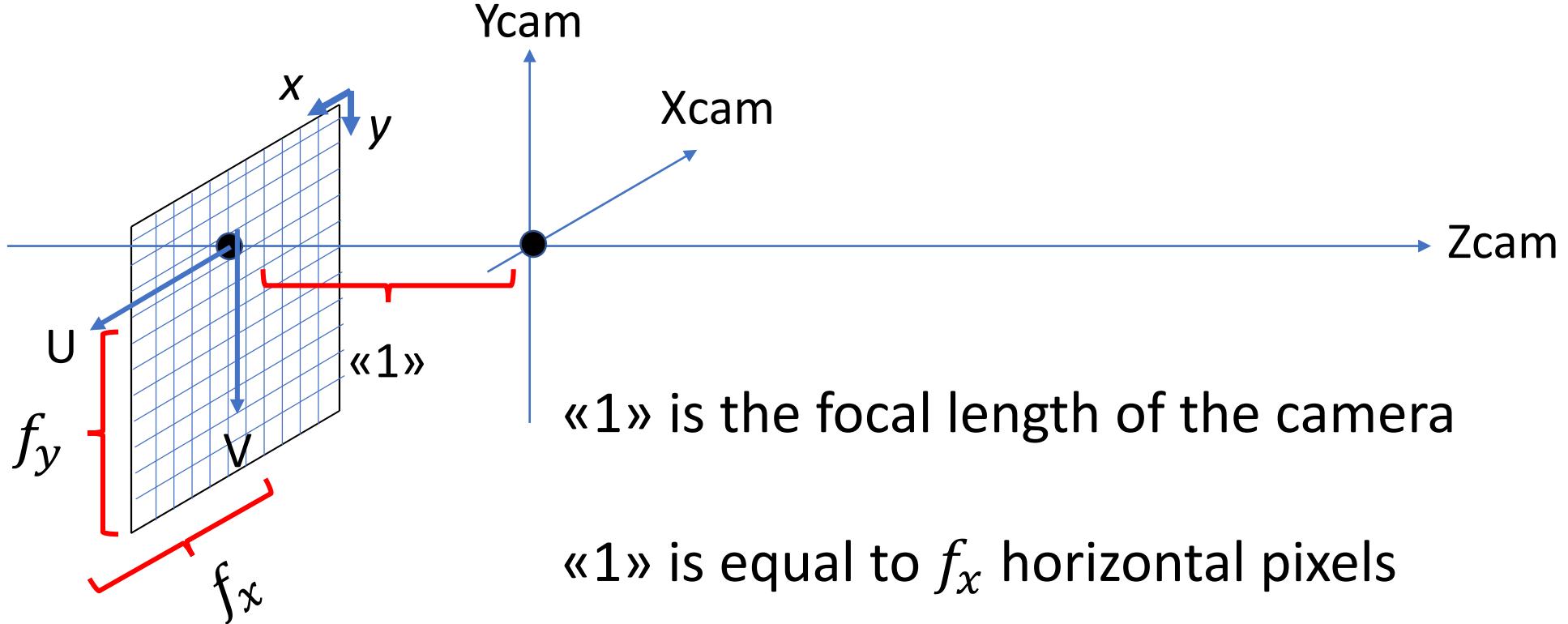
pixel coordinates



pixel coordinates



pixel coordinates

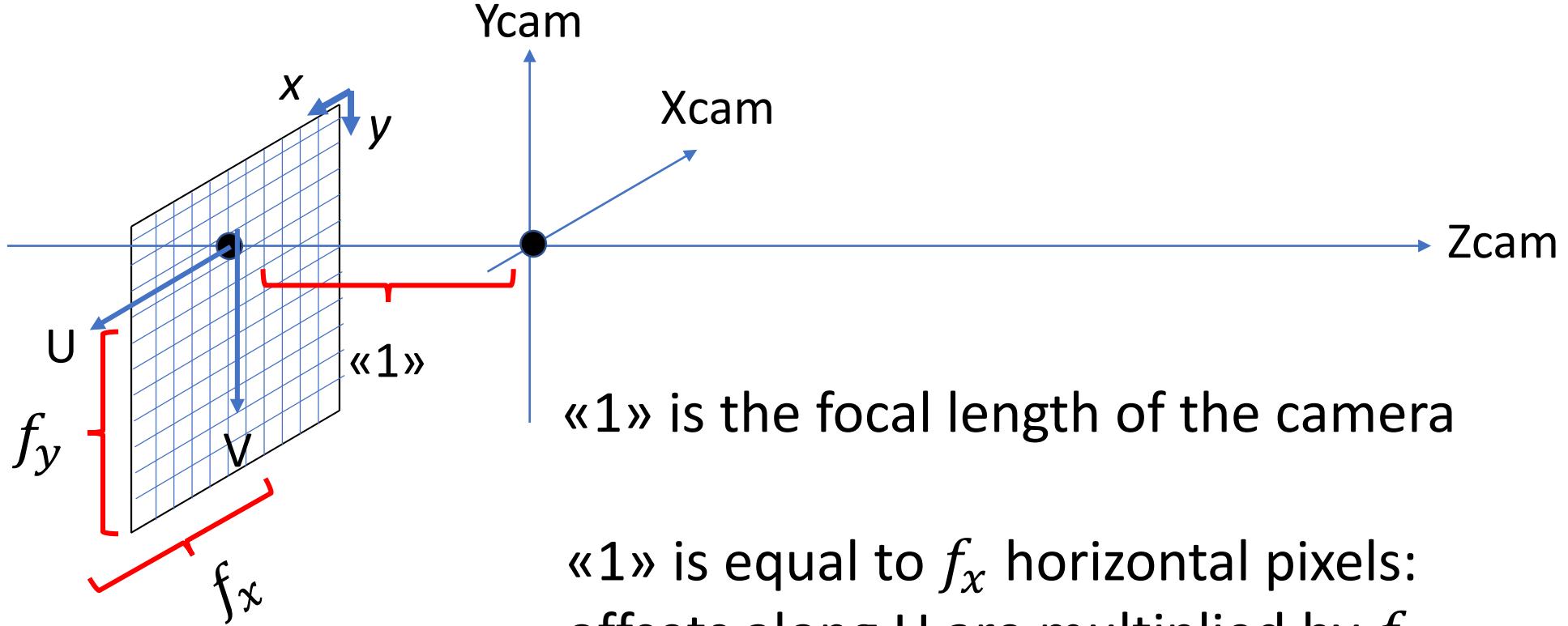


«1» is the focal length of the camera

«1» is equal to f_x horizontal pixels

«1» is equal to f_y vertical pixels

pixel coordinates

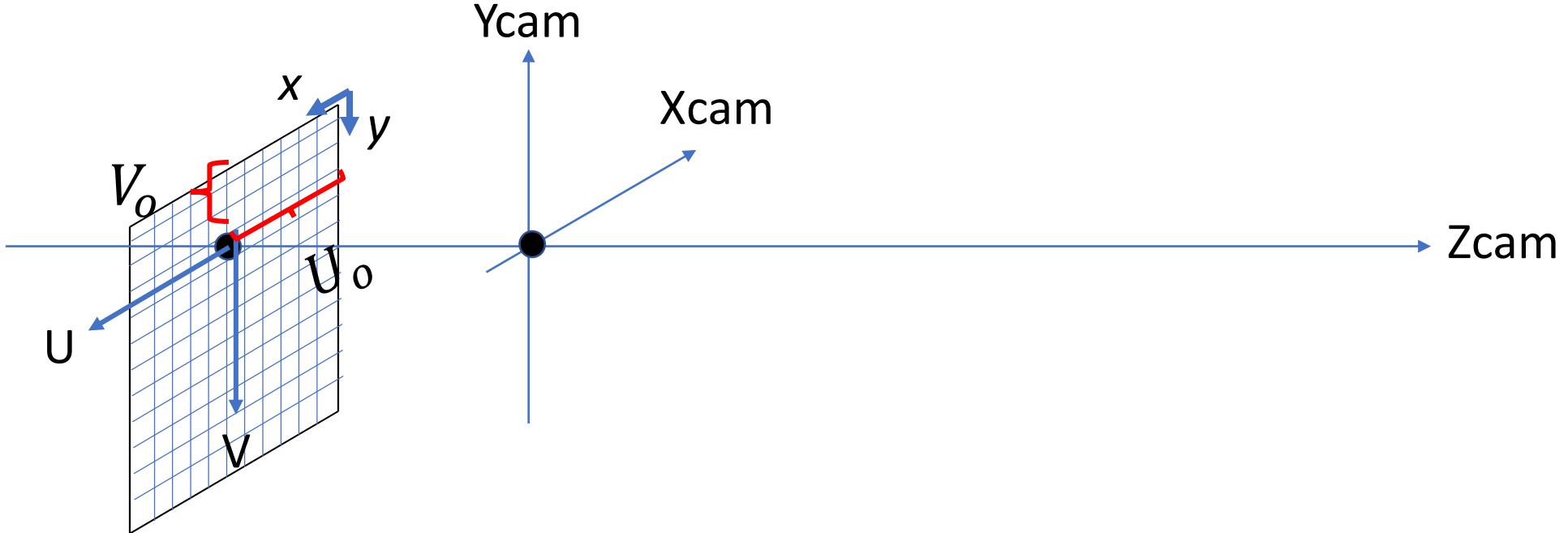


«1» is the focal length of the camera

«1» is equal to f_x horizontal pixels:
offsets along U are multiplied by f_x

«1» is equal to f_y vertical pixels:
offsets along V are multiplied by f_y

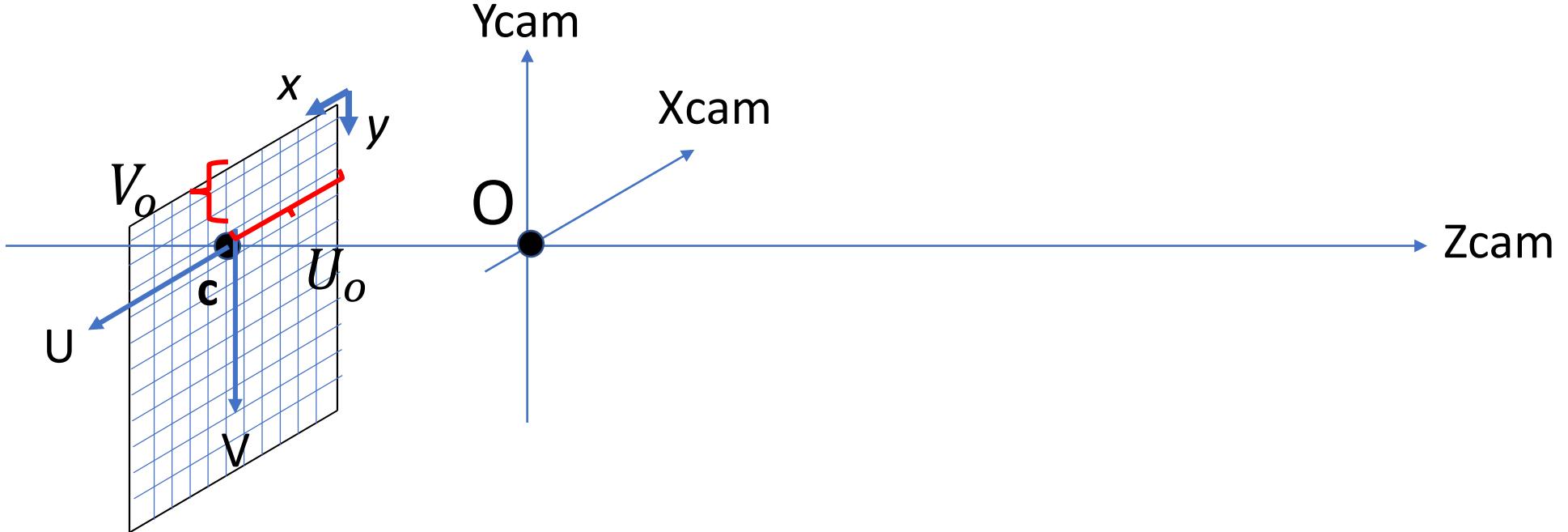
pixel coordinates



Camera principal point: projection of the camera center onto image plane

Pixel coordinates of the principal point:
 (U_o, V_o)

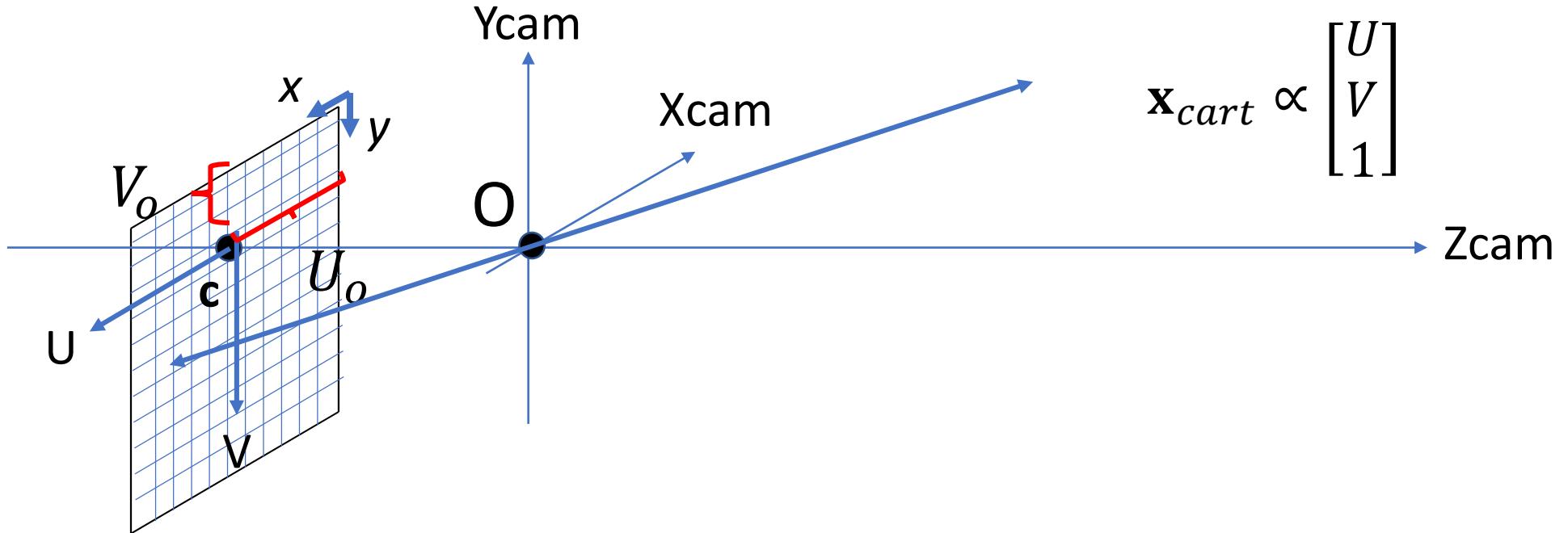
pixel coordinates



Camera principal point: projection of the camera center O onto image plane

Pixel coordinates of the principal point:
 $\mathbf{c} = (U_o, V_o)$

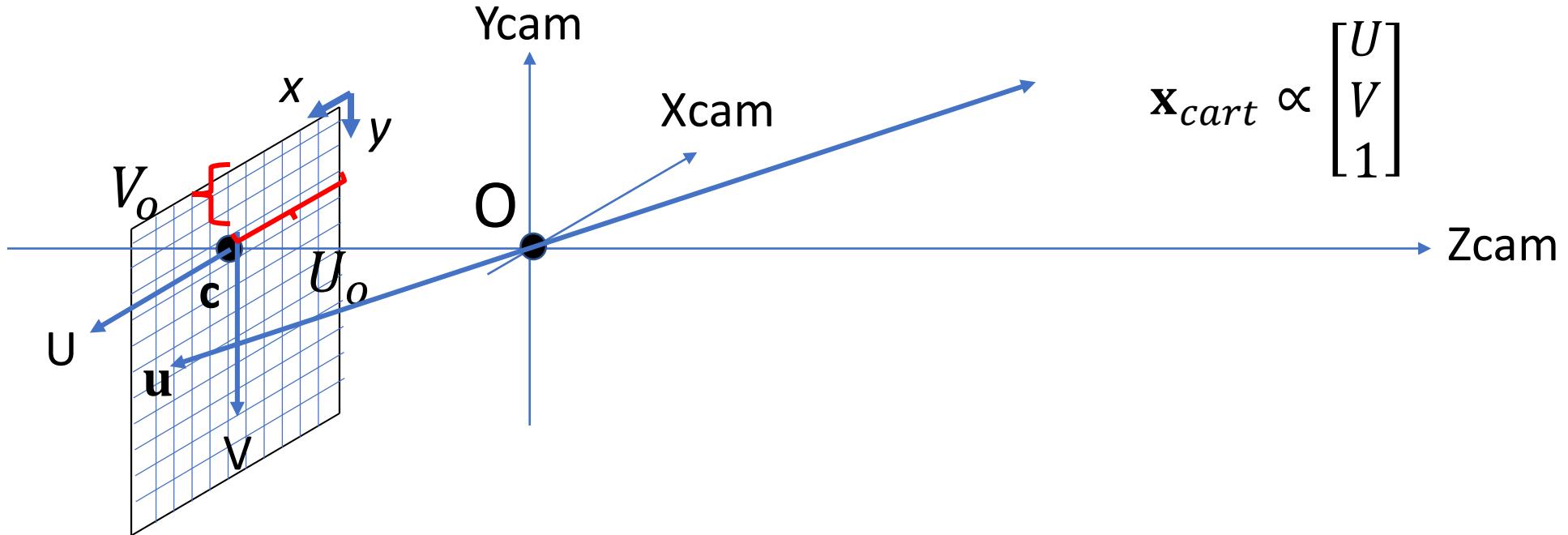
components of \mathbf{d}



Camera principal point: projection of the camera center O onto image plane

Pixel coordinates of the principal point:
 $\mathbf{c} = (U_o, V_o)$

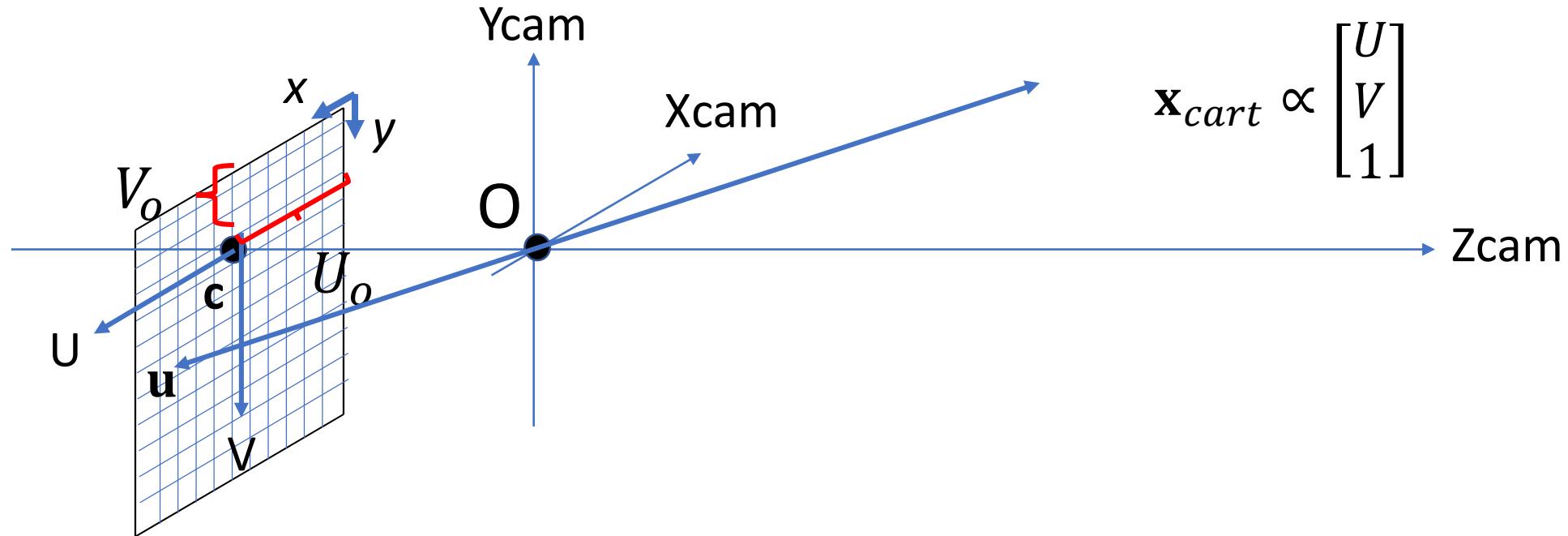
geometric coordinates of the image point: (U, V)



Camera principal point: projection of the camera center O onto image plane

Pixel coordinates of the principal point:
 $\mathbf{c} = (U_o, V_o)$

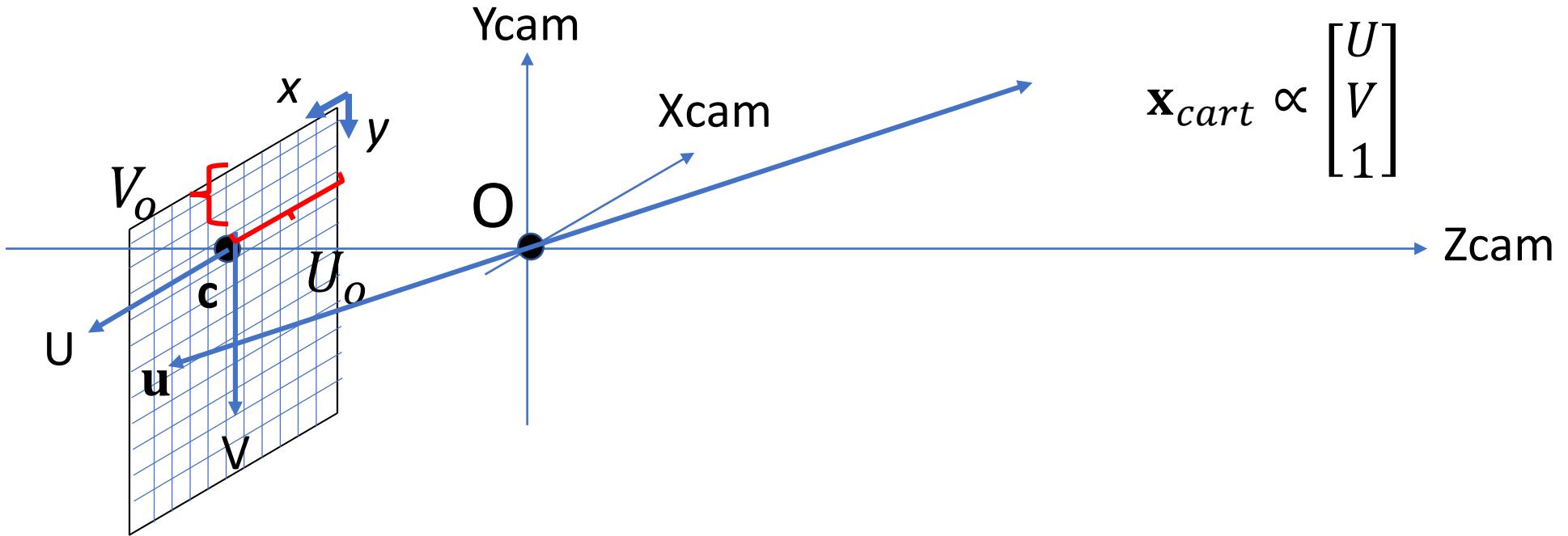
Cartesian pixel coordinates of the image point: (x, y)



Camera **principal point**: projection of the camera center O onto image plane

Pixel coordinates of the principal point:
 $\mathbf{c} = (U_o, V_o)$

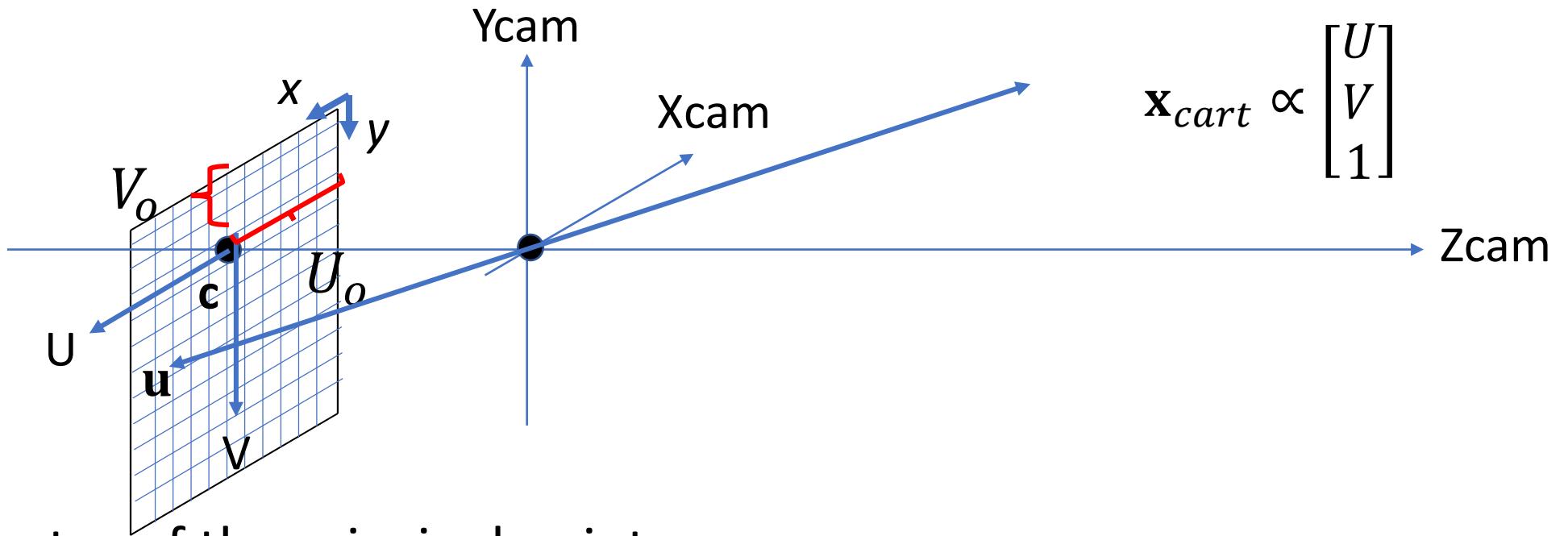
Homogeneous pixel coordinates of image point: $\mathbf{u} = [x, y, 1]^T$



Camera **principal point**: projection of the camera center O onto image plane

Pixel coordinates of the principal point:
 $\mathbf{c} = (U_o, V_o)$

homogeneous pixel coordinates of image point: $\mathbf{u} = [x, y, 1]^T$



Pixel coordinates of the principal point:

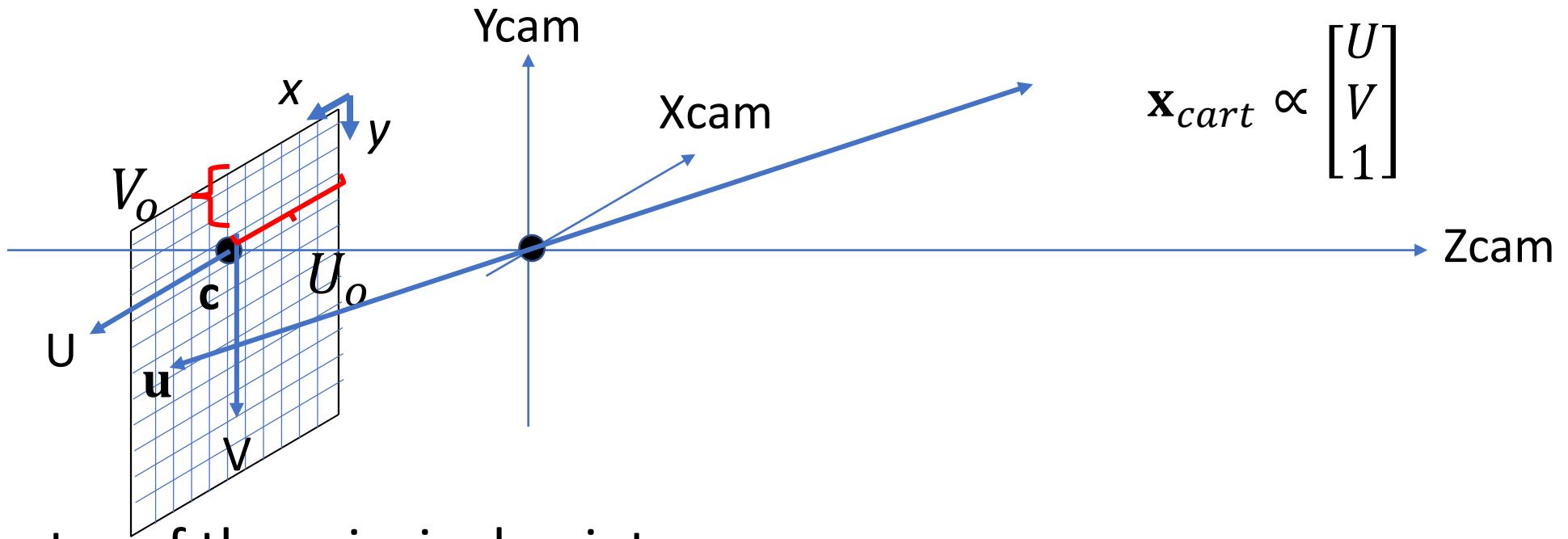
$$\mathbf{c} = (U_o, V_o)$$

«1» is equal to f_x horizontal pixels:
offsets along U are multiplied by f_x

$$\mathbf{u} = \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} f_x U + U_o \\ f_y V + V_o \\ 1 \end{bmatrix}$$

«1» is equal to f_y vertical pixels:
offsets along V are multiplied by f_y

homogeneous pixel coordinates of image point: $\mathbf{u} = [x, y, 1]^T$



Pixel coordinates of the principal point:

$$\mathbf{c} = (U_o, V_o)$$

«1» is equal to f_x horizontal pixels:

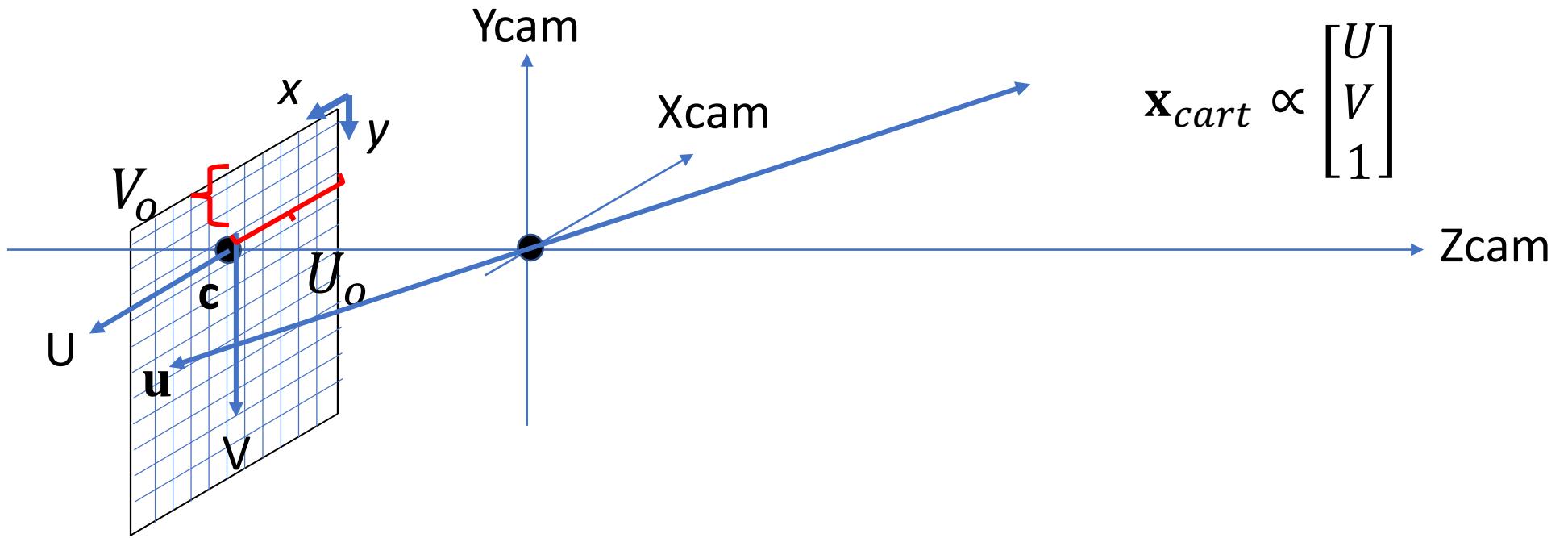
offsets along U are multiplied by f_x

«1» is equal to f_y vertical pixels:

offsets along V are multiplied by f_y

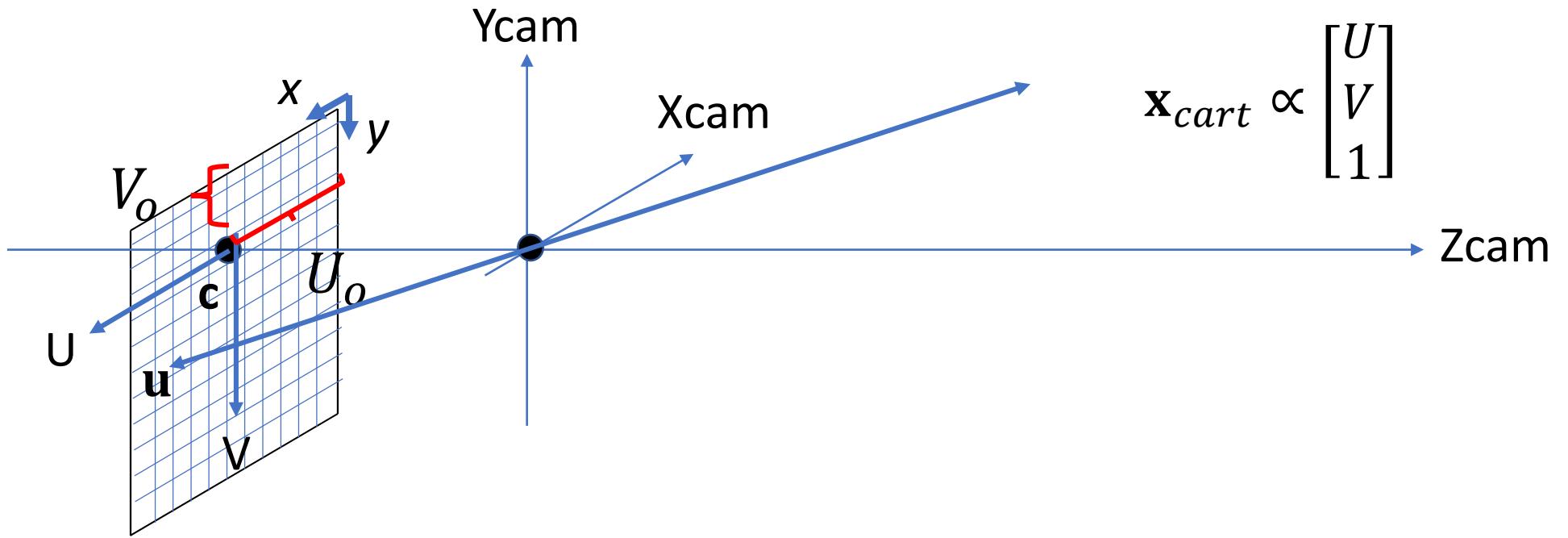
$$\mathbf{u} = \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} f_x U + U_o \\ f_y V + V_o \\ 1 \end{bmatrix} = \begin{bmatrix} f_x & 0 & U_o \\ 0 & f_y & V_o \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} U \\ V \\ 1 \end{bmatrix}$$

homogeneous pixel coordinates of image point: $\mathbf{u} = [x, y, 1]^T$



$$\mathbf{u} = \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} f_x U + U_o \\ f_y V + V_o \\ 1 \end{bmatrix} = \begin{bmatrix} f_x & 0 & U_o \\ 0 & f_y & V_o \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} U \\ V \\ 1 \end{bmatrix} = \begin{bmatrix} f_x & 0 & U_o \\ 0 & f_y & V_o \\ 0 & 0 & 1 \end{bmatrix} \mathbf{x}_{cart}$$

homogeneous pixel coordinates of image point: $\mathbf{u} = [x, y, 1]^T$



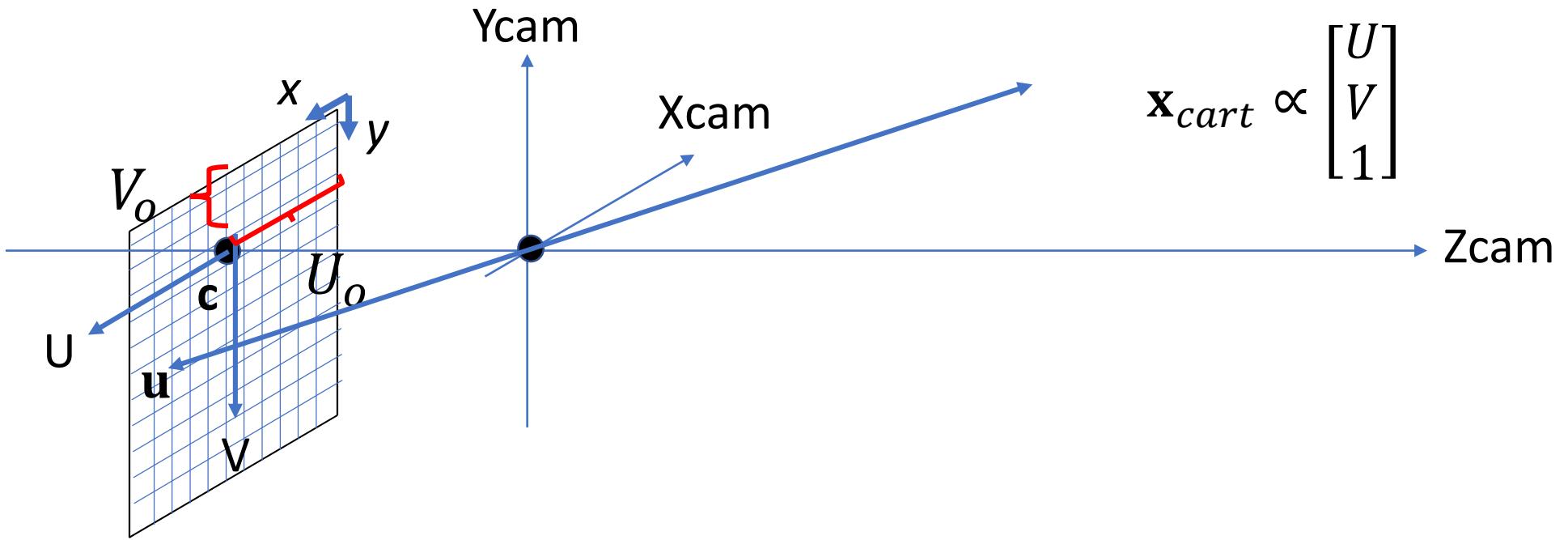
$$\mathbf{x}_{cart} \propto \begin{bmatrix} U \\ V \\ 1 \end{bmatrix}$$

z_{cam}

$$\mathbf{u} = \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} f_x U + U_o \\ f_y V + V_o \\ 1 \end{bmatrix} = \begin{bmatrix} f_x & 0 & U_o \\ 0 & f_y & V_o \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} U \\ V \\ 1 \end{bmatrix} = \begin{bmatrix} f_x & 0 & U_o \\ 0 & f_y & V_o \\ 0 & 0 & 1 \end{bmatrix} \mathbf{x}_{cart} = \mathbf{K} \mathbf{x}_{cart}$$

\mathbf{K}

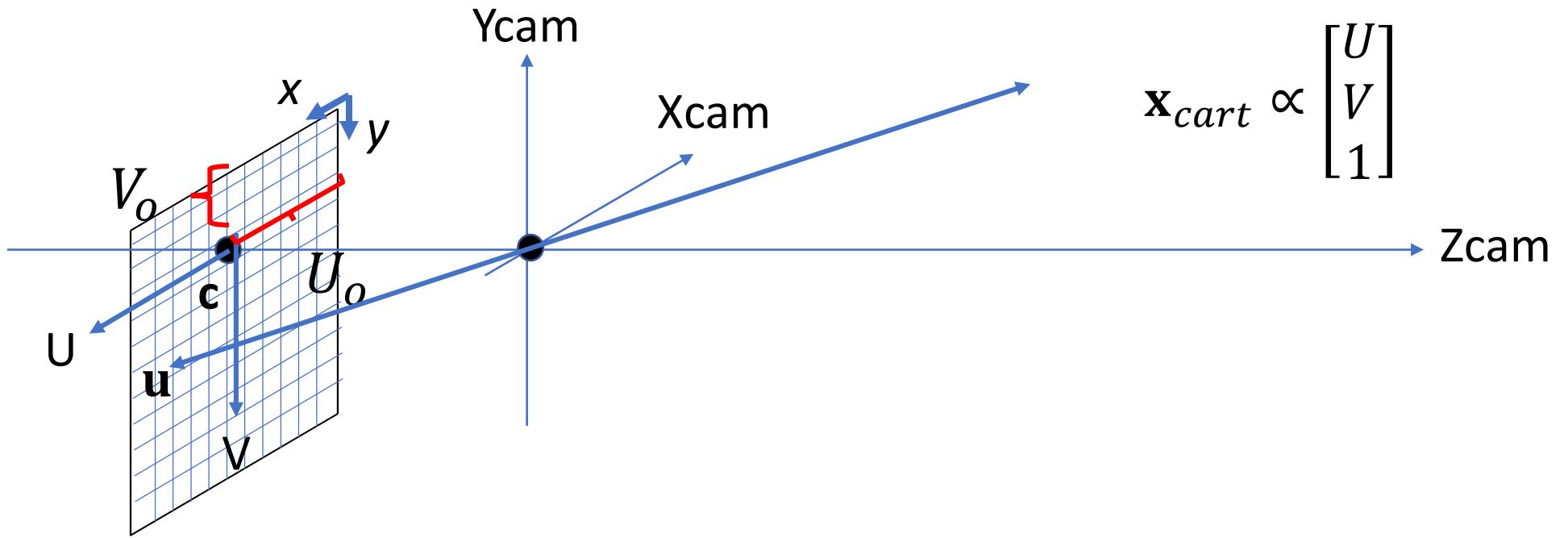
homogeneous pixel coordinates of image point: $\mathbf{u} = [x, y, 1]^T$



$$\mathbf{u} = \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} f_x U + U_o \\ f_y V + V_o \\ 1 \end{bmatrix} = \begin{bmatrix} f_x & 0 & U_o \\ 0 & f_y & V_o \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} U \\ V \\ 1 \end{bmatrix} = \begin{bmatrix} f_x & 0 & U_o \\ 0 & f_y & V_o \\ 0 & 0 & 1 \end{bmatrix} \mathbf{x}_{cart} = \mathbf{K} \begin{bmatrix} X_{cam} \\ Y_{cam} \\ Z_{cam} \end{bmatrix}$$

\mathbf{K}

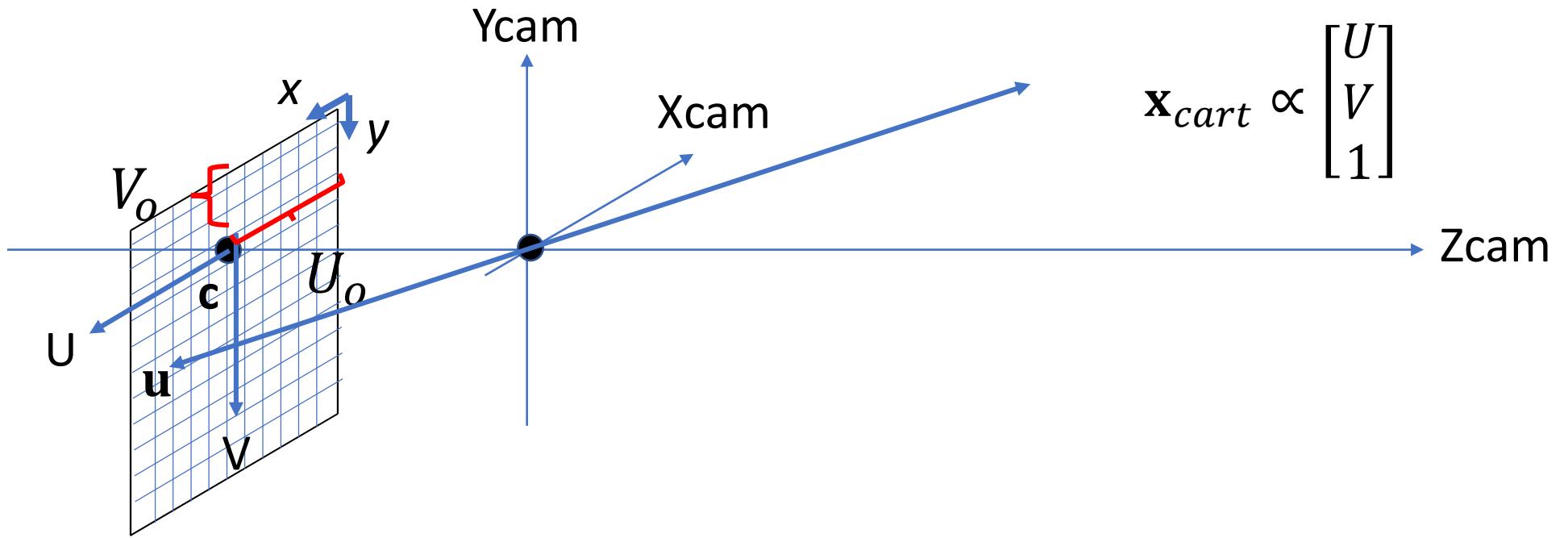
homogeneous pixel coordinates of image point: $\mathbf{u} = [x, y, 1]^T$



$$\mathbf{u} = \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} f_x & 0 & U_o \\ 0 & f_y & V_o \\ 0 & 0 & 1 \end{bmatrix} \mathbf{x}_{cart} = \mathbf{K} \begin{bmatrix} X_{cam} \\ Y_{cam} \\ Z_{cam} \end{bmatrix} = \begin{bmatrix} \mathbf{K} & 0 \\ 0 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} X_{cam} \\ Y_{cam} \\ Z_{cam} \\ 1 \end{bmatrix}$$

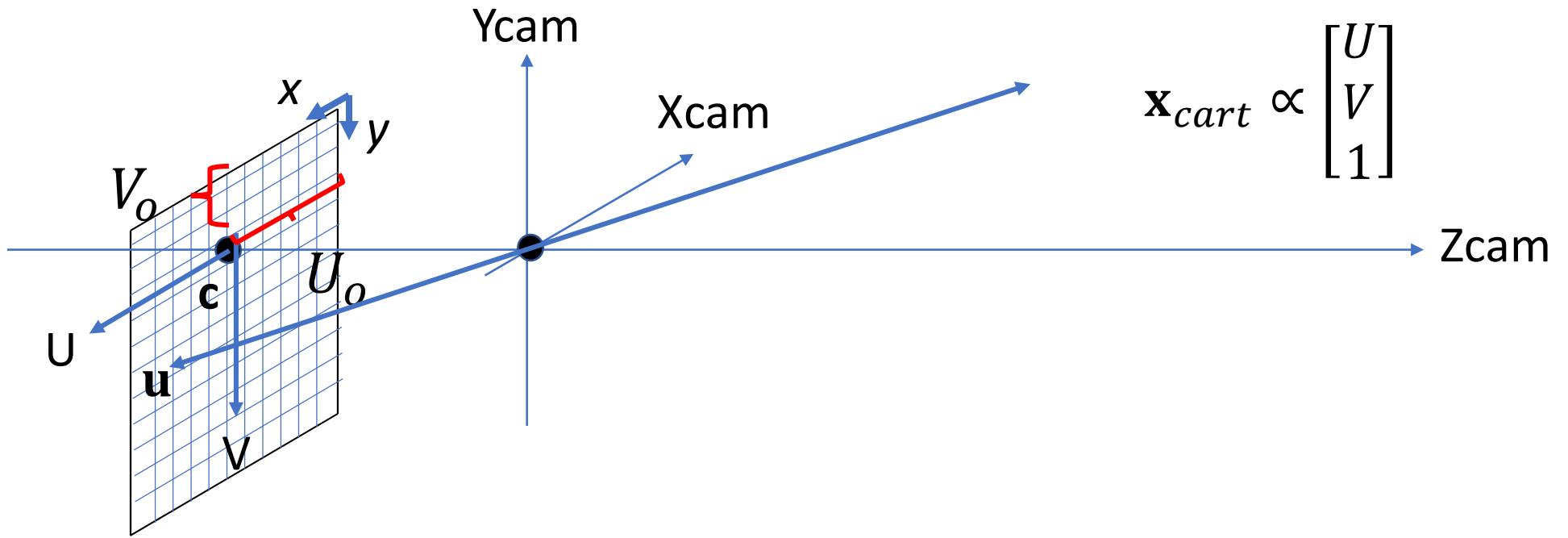
\mathbf{K}

homogeneous pixel coordinates of image point: $\mathbf{u} = [x, y, 1]^T$



$$\mathbf{u} = \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \underbrace{\begin{bmatrix} f_x & 0 & U_o \\ 0 & f_y & V_o \\ 0 & 0 & 1 \end{bmatrix}}_K \mathbf{x}_{cart} = \begin{bmatrix} X_{cam} \\ Y_{cam} \\ Z_{cam} \end{bmatrix} = \begin{bmatrix} K & 0 \\ 0 & 0 \end{bmatrix} \mathbf{X}_{cam} = [K \quad 0] \mathbf{X}_{cam}$$

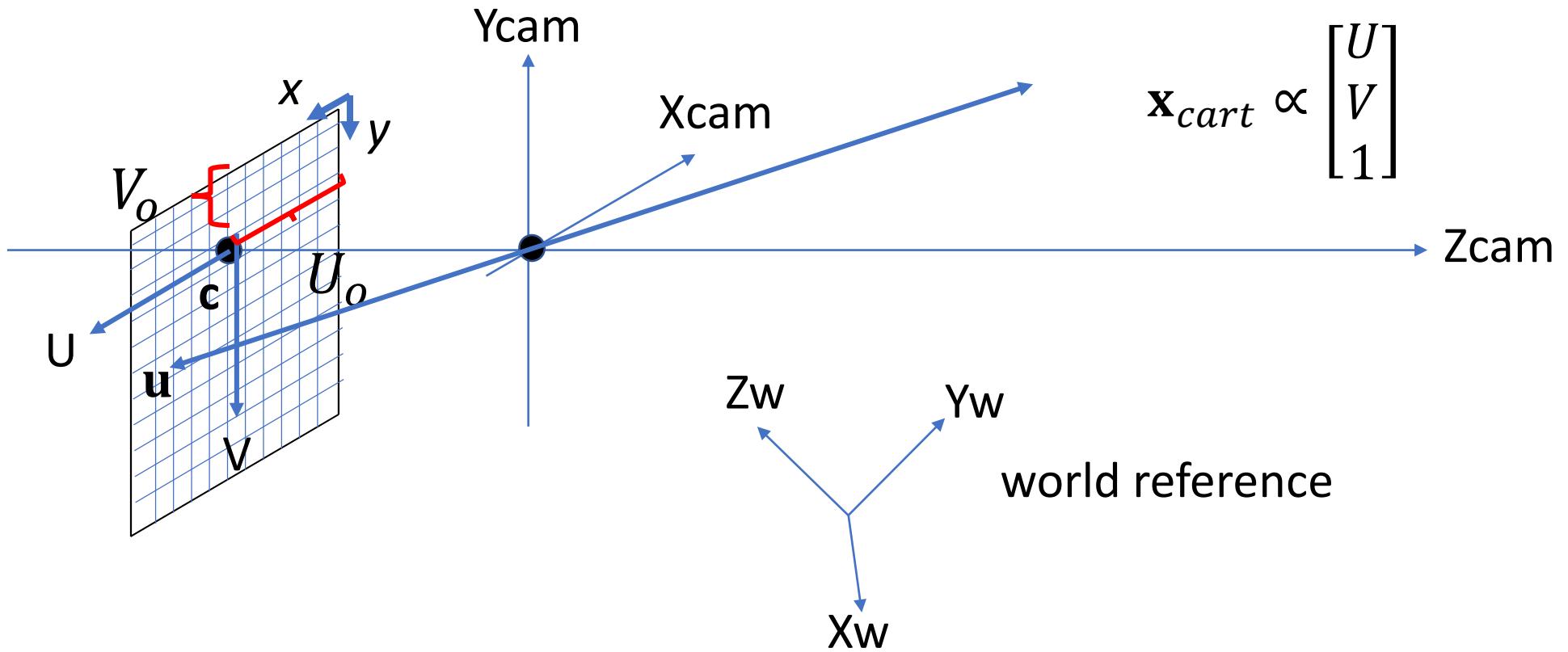
homogeneous pixel coordinates of image point: $\mathbf{u} = [x, y, 1]^T$



$$\mathbf{x}_{cart} \propto \begin{bmatrix} U \\ V \\ 1 \end{bmatrix}$$

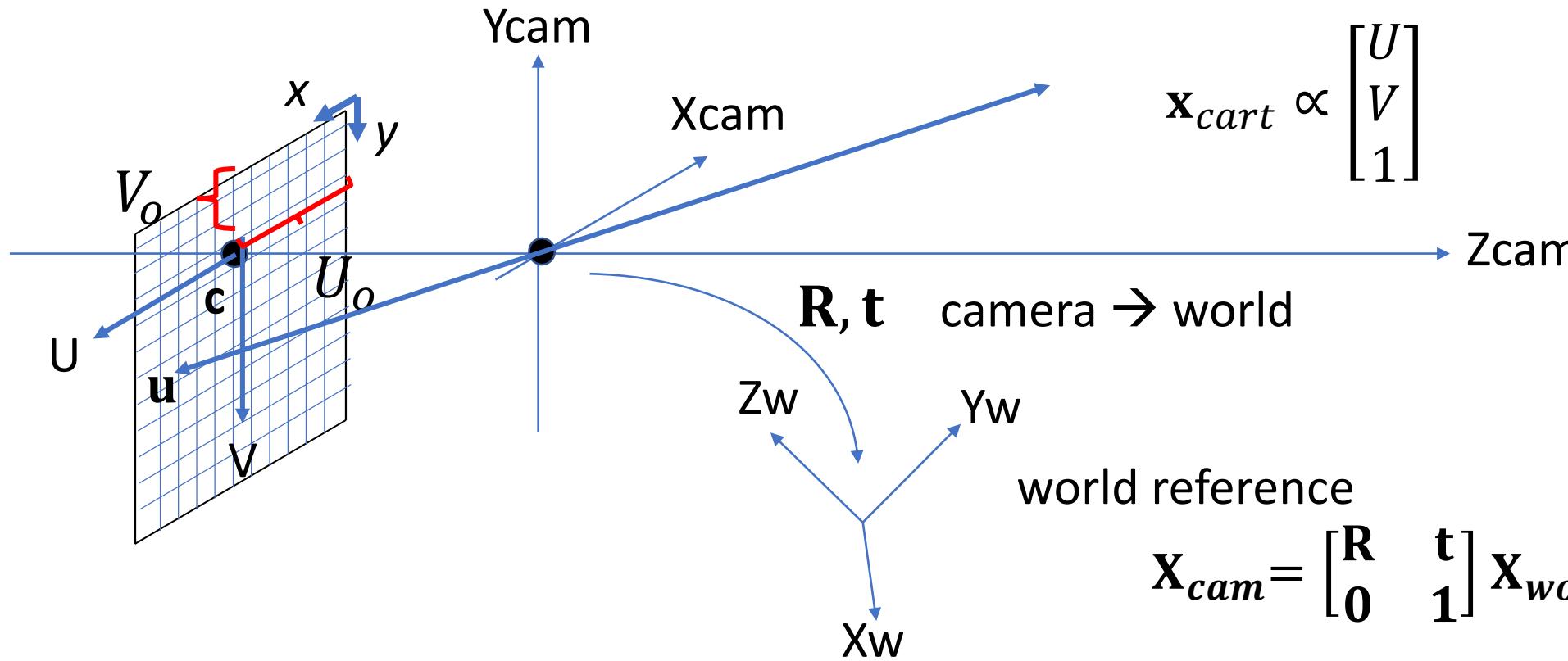
$$\mathbf{u} = \mathbf{K} \mathbf{x}_{cart} = \mathbf{K} \begin{bmatrix} X_{cam} \\ Y_{cam} \\ Z_{cam} \end{bmatrix} = \begin{bmatrix} \mathbf{K} & 0 \\ 0 & 0 \end{bmatrix} \mathbf{X}_{cam} = [\mathbf{K} \quad \mathbf{0}] \mathbf{X}_{cam} = \mathbf{K}[\mathbf{I} \quad \mathbf{0}] \mathbf{X}_{cam}$$

In general, world reference is \neq camera reference



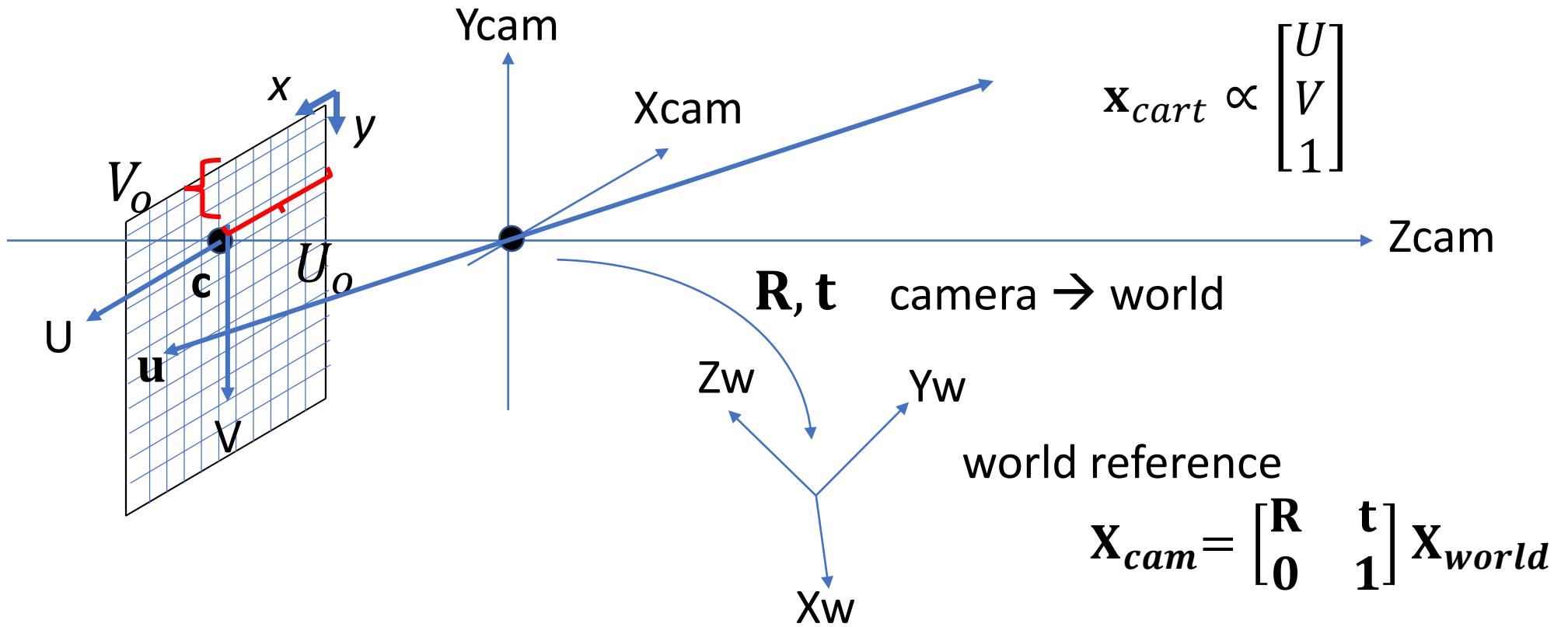
$$\mathbf{u} = \mathbf{K} \mathbf{x}_{cart} = \mathbf{K} \begin{bmatrix} X_{cam} \\ Y_{cam} \\ Z_{cam} \end{bmatrix} = \begin{bmatrix} \mathbf{K} & 0 \\ 0 & 0 \end{bmatrix} \mathbf{x}_{cam} = [\mathbf{K} \quad \mathbf{0}] \mathbf{x}_{cam} = \mathbf{K} [\mathbf{I} \quad \mathbf{0}] \mathbf{x}_{cam}$$

In general, world reference is \neq camera reference



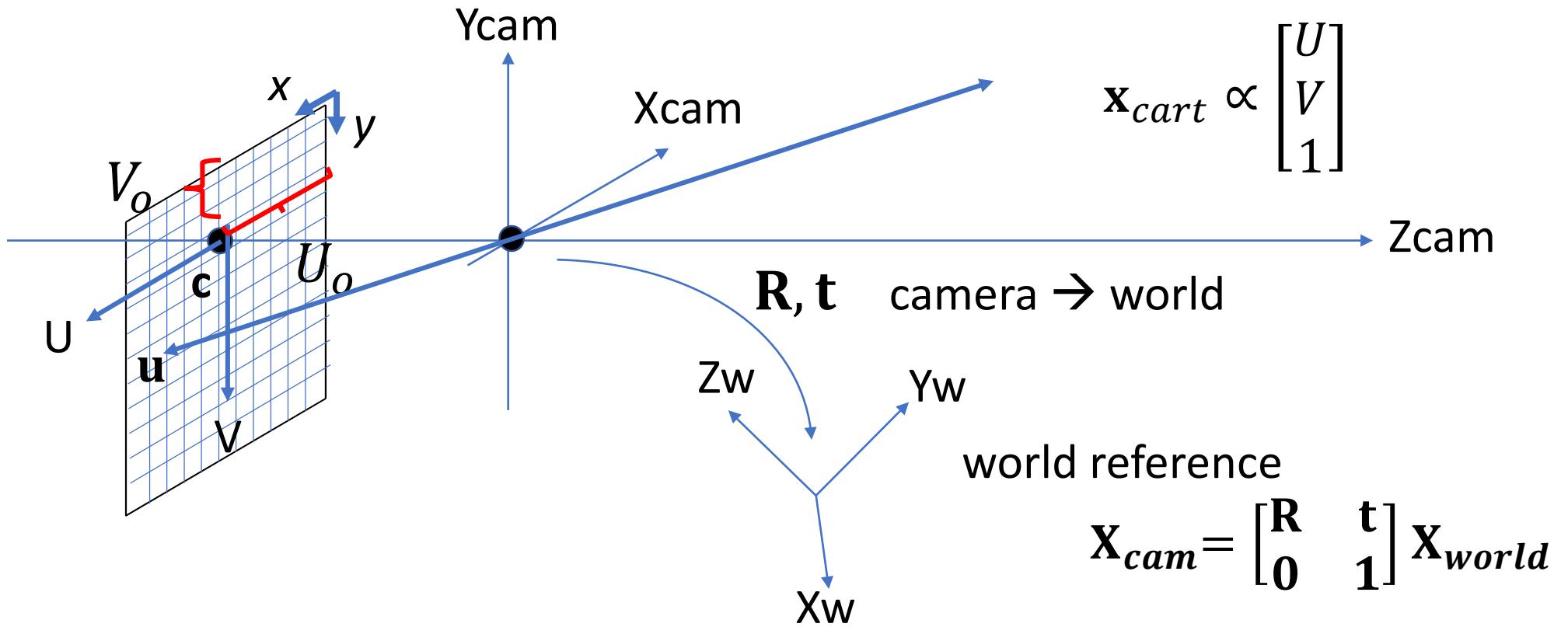
$$\mathbf{u} = \mathbf{K} \mathbf{x}_{cart} = \mathbf{K} \begin{bmatrix} X_{cam} \\ Y_{cam} \\ Z_{cam} \end{bmatrix} = \begin{bmatrix} \mathbf{K} & 0 \\ 0 & 0 \end{bmatrix} \mathbf{x}_{cam} = [\mathbf{K} \quad \mathbf{0}] \mathbf{x}_{cam} = \mathbf{K} [\mathbf{I} \quad \mathbf{0}] \mathbf{x}_{cam}$$

In general, world reference is \neq camera reference



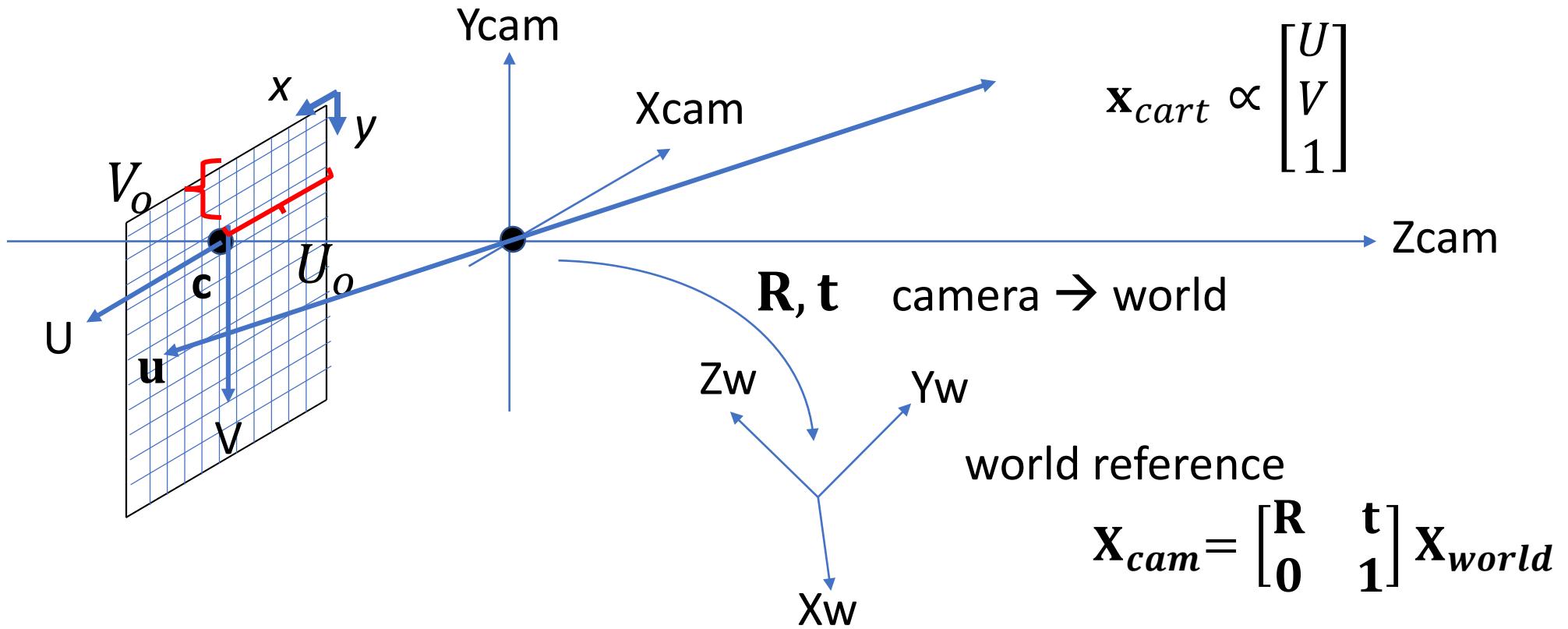
$$\mathbf{u} = \mathbf{K}[\mathbf{I} \quad \mathbf{0}] \mathbf{X}_{cam} = \mathbf{K}[\mathbf{I} \quad \mathbf{0}] \begin{bmatrix} \mathbf{R} & \mathbf{t} \\ \mathbf{0} & 1 \end{bmatrix} \mathbf{X}_{world}$$

In general, world reference is \neq camera reference



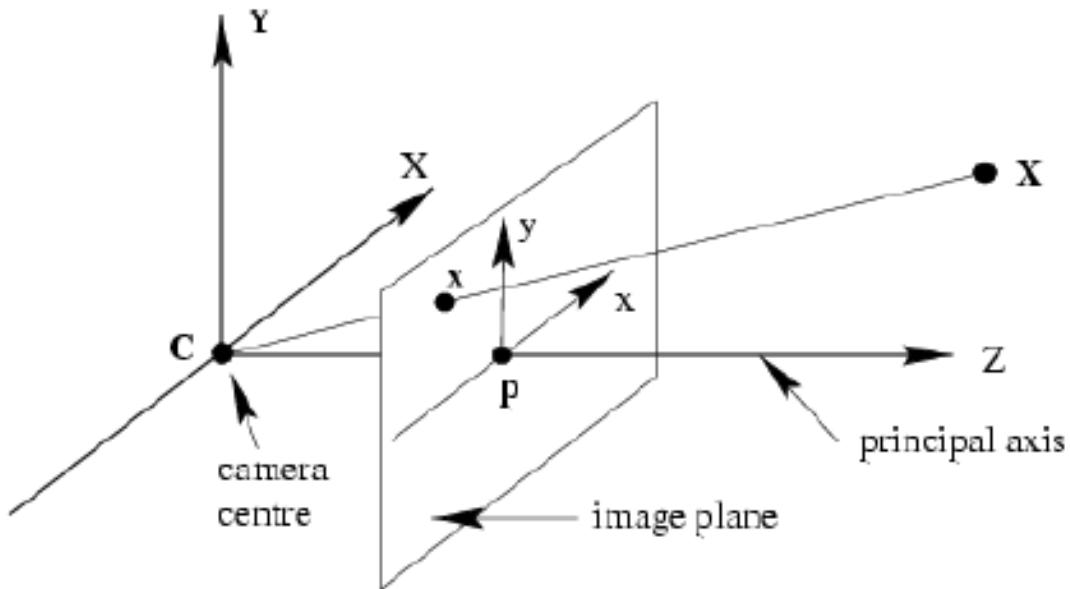
$$u = K[R \quad t] X_{world} = [KR \quad Kt] X_{world}$$

In general, world reference is \neq camera reference



$$\mathbf{u} = \mathbf{K}[\mathbf{R} \ \mathbf{t}] \ \mathbf{X}_{world} = [\mathbf{K}\mathbf{R} \ \ \mathbf{K}\mathbf{t}] \ \mathbf{X}_{world}$$

remember ...



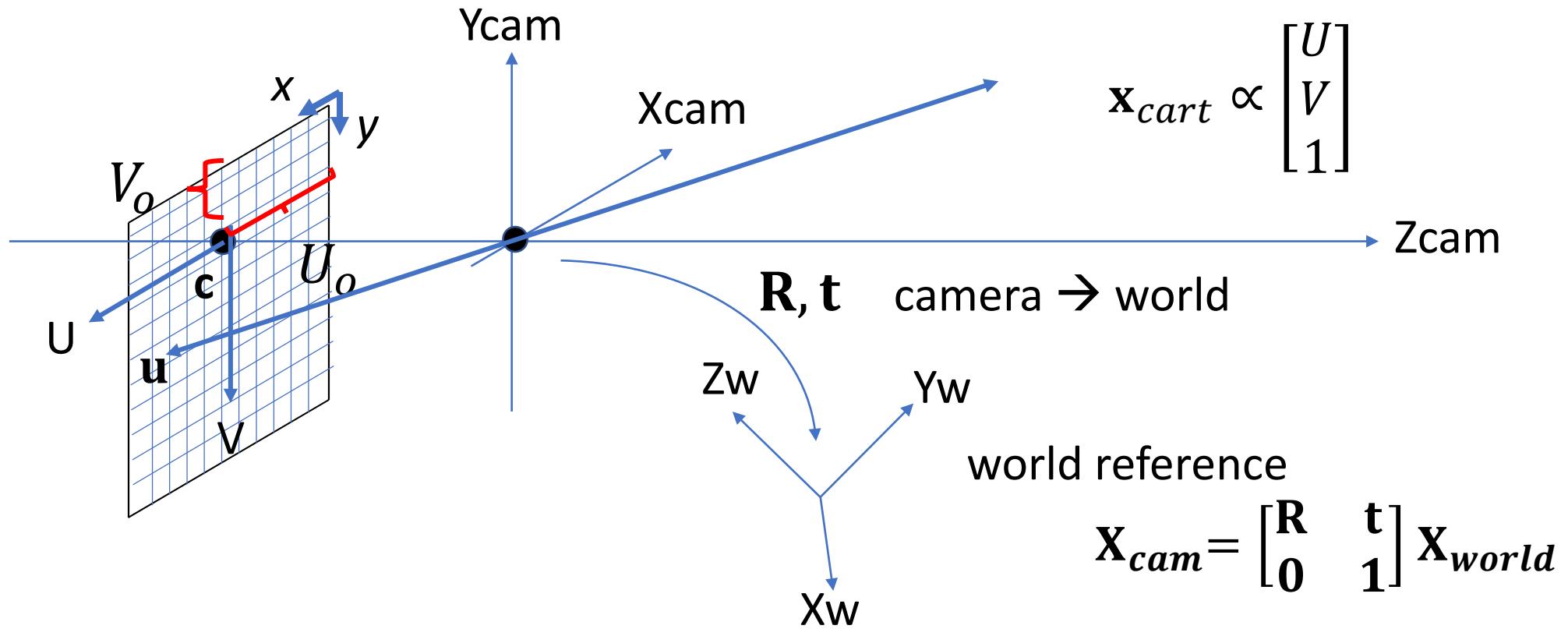
$$\mathbf{X} = \begin{pmatrix} X \\ Y \\ Z \\ 1 \end{pmatrix} \mapsto \mathbf{u} = \begin{pmatrix} u \\ v \\ w \end{pmatrix} = \mathbf{P}_{3 \times 4} \begin{pmatrix} X \\ Y \\ Z \\ 1 \end{pmatrix} = \mathbf{P}_{3 \times 4} \mathbf{X} = \begin{vmatrix} \mathbf{M}_{3 \times 3} & \mathbf{m}_{3 \times 1} \end{vmatrix} \mathbf{X}$$

camera projection matrix

invertible

$$\mathbf{M} = \mathbf{K} \mathbf{R}_{cam \rightarrow world}$$

In general, world reference is \neq camera reference

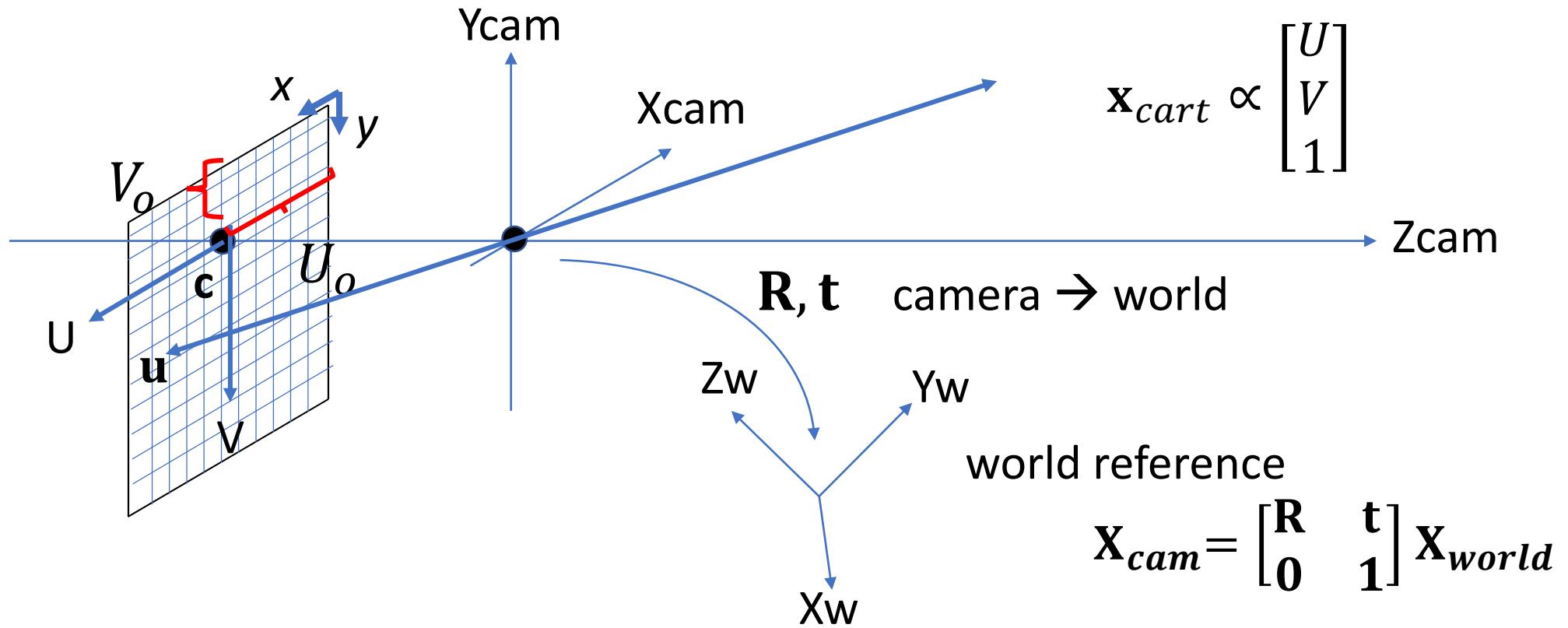


$$u = K[R \ t] X_{world} = [KR \ Kt] X_{world}$$

remember $u = P X_{world}$

$$= [M \ m] X_{world}$$

In general, world reference is \neq camera reference



$$\mathbf{u} = \mathbf{K}[\mathbf{R} \quad \mathbf{t}] \mathbf{X}_{world} = [\mathbf{K}\mathbf{R} \quad \mathbf{K}\mathbf{t}] \mathbf{X}_{world}$$

remember $\mathbf{u} = \mathbf{P} \mathbf{X}_{world}$

$$= [\mathbf{M} \quad \mathbf{m}] \mathbf{X}_{world}$$

$$\rightarrow \boxed{\mathbf{M} = \mathbf{K}\mathbf{R}} \text{ and } \boxed{\mathbf{m} = \mathbf{K}\mathbf{t}}$$

but

$$\begin{aligned} \mathbf{u} &= \mathbf{K}[\mathbf{R} \quad \mathbf{t}] \mathbf{X}_{world} = [\mathbf{KR} \quad \mathbf{Kt}] \mathbf{X}_{world} \\ \mathbf{u} &= \mathbf{P} \mathbf{X}_{world} = [\mathbf{M} \quad \mathbf{m}] \mathbf{X}_{world} \end{aligned}$$

$$\rightarrow \mathbf{M} = \mathbf{KR} \text{ and } \mathbf{m} = \mathbf{Kt}$$

$$\mathbf{M} = \mathbf{KR}_{cam \rightarrow world}$$

given \mathbf{M} , find \mathbf{K} and \mathbf{R} by Q-R matrix decomposition of $\text{inv}(\mathbf{M})$

$$\begin{aligned} \mathbf{u} &= \mathbf{K}[\mathbf{R} \quad \mathbf{t}] \mathbf{X}_{world} = [\mathbf{KR} \quad \mathbf{Kt}] \mathbf{X}_{world} \\ \text{but } \mathbf{u} &= \mathbf{P} \mathbf{X}_{world} = [\mathbf{M} \quad \mathbf{m}] \mathbf{X}_{world} \end{aligned}$$

$$\rightarrow \mathbf{M} = \mathbf{KR} \text{ and } \mathbf{m} = \mathbf{Kt}$$

$$\mathbf{M} = \mathbf{KR}_{cam \rightarrow world}$$

given \mathbf{M} , find \mathbf{K} and \mathbf{R} by Q-R matrix decomposition of $\text{inv}(\mathbf{M})$

$$\text{from } \mathbf{m} = -\mathbf{Mo} \rightarrow \mathbf{u} = [\mathbf{M} \quad \mathbf{m}] \mathbf{X}_{world} = \mathbf{M}[\mathbf{I} \quad -\mathbf{o}] \mathbf{X}_{world}$$

$$\mathbf{u} = \mathbf{KR}[\mathbf{I} \quad -\mathbf{o}] \mathbf{X}_{world}$$

camera calibration

Intrinsic camera calibration:
estimation of matrix \mathbf{K}

- focal distance f_X
- focal distance f_Y
- principal point (U_o, V_o)
- skew factor (in old cameras)

aspect ratio $a = f_X/f_Y$:

ratio between pixel width and height

intrinsic camera parameters don't vary
under camera displacement

Extrinsic camera calibration:
estimation of matrix \mathbf{R} and vector \mathbf{t}

- camera rotation \mathbf{R}
- camera translation \mathbf{t}

extrinsic camera parameters vary
under camera displacement

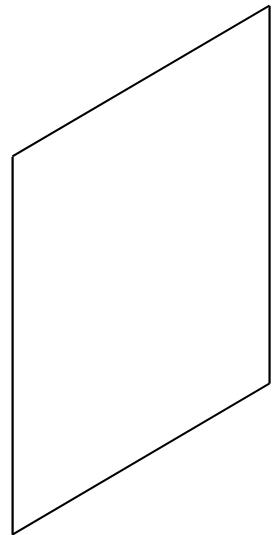


$$x = x_o + (x_o - c_x)(K_1 r^2 + K_2 r^4 + \dots)$$
$$y = y_o + (y_o - c_y)(K_1 r^2 + K_2 r^4 + \dots)$$

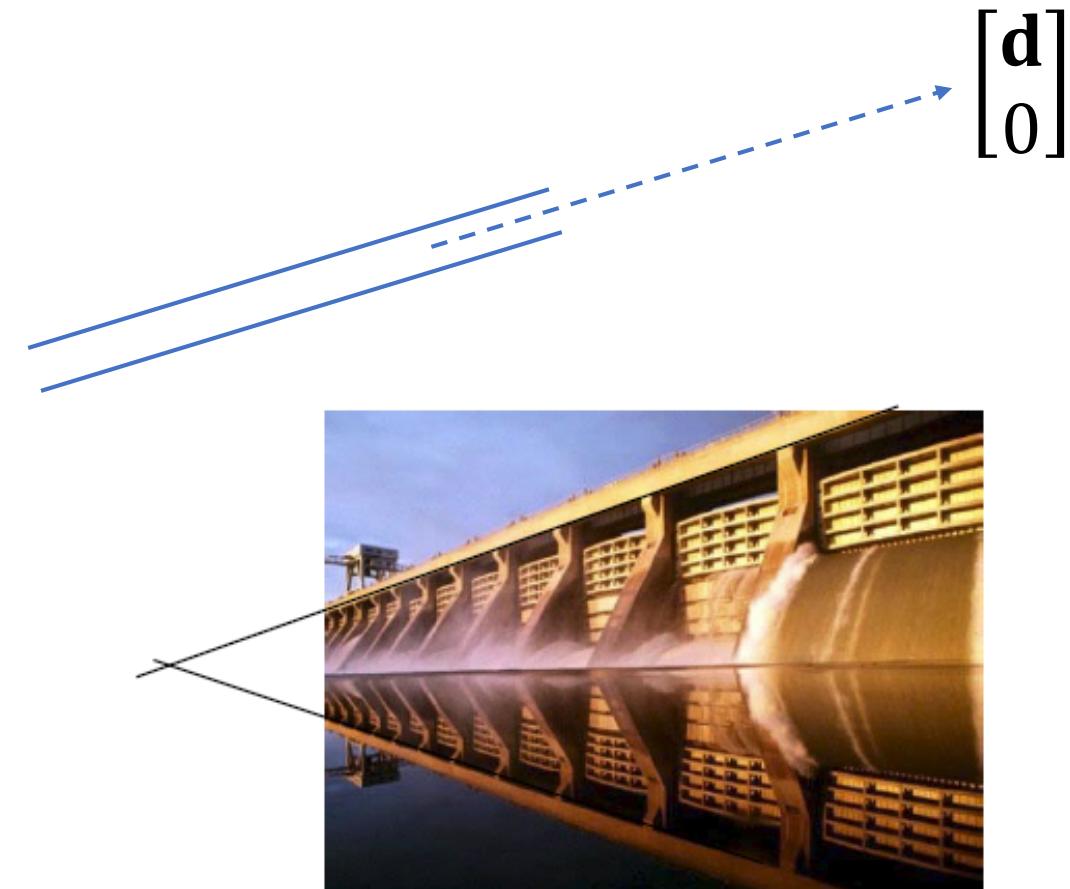
$$r = (x_o - c_x)^2 + (y_o - c_y)^2 .$$

vanishing point of a direction

parallel lines concur at the point at the infinity of
their common direction \mathbf{d}



o



$[d]$
 $[0]$

vanishing points

V image of the point at the ∞ along direction \mathbf{d}

$$\mathbf{u}_v = | \mathbf{M} | \mathbf{m} \cdot \begin{vmatrix} \mathbf{d} \\ \cdots \\ 0 \end{vmatrix} = \mathbf{M} \cdot \mathbf{d}$$

vanishing points

\mathbf{V} image of the point at the ∞ along direction \mathbf{d}

$$\mathbf{u}_v = | \mathbf{M} | \mathbf{m} \cdot \begin{vmatrix} \mathbf{d} \\ \cdots \\ 0 \end{vmatrix} = \mathbf{M} \cdot \mathbf{d}$$

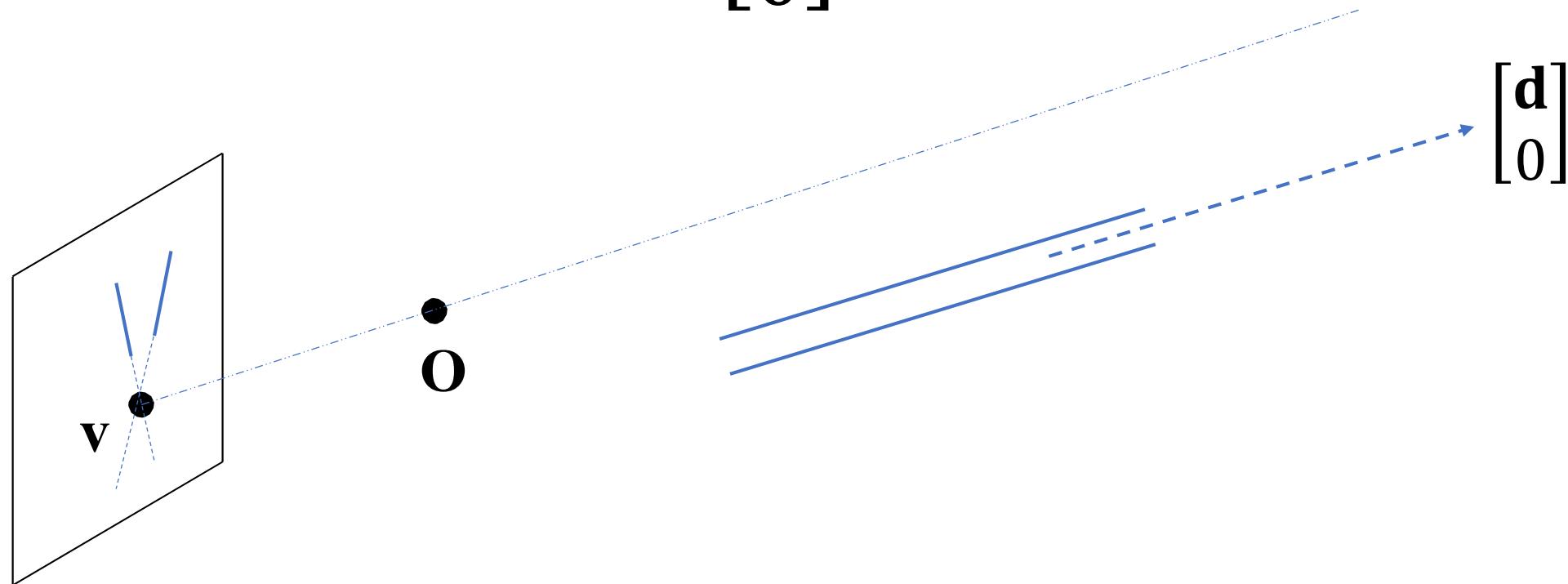
Remember: the direction of the backprojection of image point \mathbf{V} (viewing ray associated to \mathbf{V}) is $\mathbf{M}^{-1}\mathbf{u}_v = \mathbf{M}^{-1}\mathbf{M}\mathbf{d} = \mathbf{d}$



Vanishing Point Theorem:

The viewing ray associated to the vanishing point \mathbf{V} of a direction \mathbf{d} is parallel to \mathbf{d}

The backprojection (viewing ray) of vanishing point v
goes through point at the infinity $\begin{bmatrix} d \\ 0 \end{bmatrix}$ (since v is its image)



→ the viewing ray of v is parallel to the direction d

mobile robot navigation



vehicle driving



If \mathbf{M} is known, find \mathbf{V} and follow the direction of the corridor:

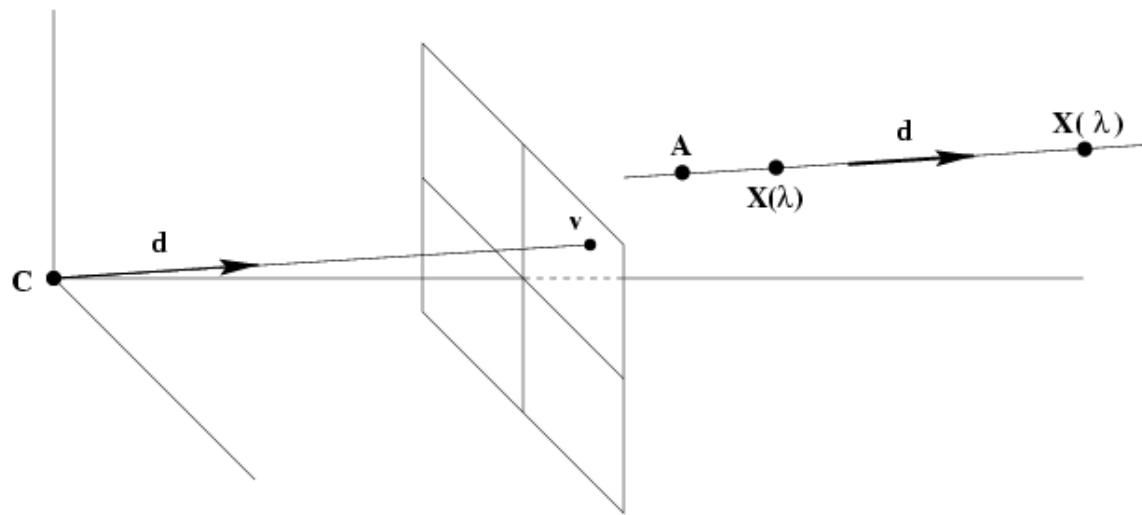
$$\mathbf{d} = \mathbf{M}^{-1}\mathbf{V}$$

Vanishing points (d measured wrt camera reference)

$$x(\lambda) = P\mathbf{X}(\lambda) = PA + \lambda PD = a + \lambda Kd$$

$$v = \lim_{\lambda \rightarrow \infty} x(\lambda) = \lim_{\lambda \rightarrow \infty} (a + \lambda Kd) = Kd$$

$$v = P\mathbf{X}_{\infty} = Kd$$

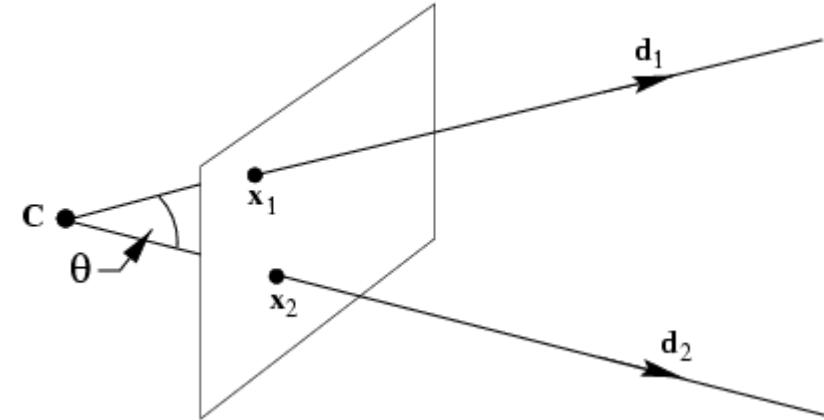


Vanishing points (d measured wrt world reference) $v = P\mathbf{X}_{\infty} = KRd_w$

What does INTRINSIC calibration give?

$$x = K[I \mid 0] \begin{bmatrix} d \\ 0 \end{bmatrix}$$

$$d = K^{-1}x$$



$$\cos \theta = \frac{d_1^T d_2}{\sqrt{(d_1^T d_1)(d_2^T d_2)}} = \frac{x_1^T (K^{-T} K^{-1}) x_2}{\sqrt{(x_1^T (K^{-T} K^{-1}) x_1)(x_2^T (K^{-T} K^{-1}) x_2)}}$$

An image I defines a plane through the camera center with normal $n = K^T I$ measured in the camera's Euclidean frame

→ Relative position of viewing rays associated to different image points

The image ω of the absolute conic Ω_∞ is **independent** of \mathbf{R} and \mathbf{t} : it only depends on intrinsic parameters

$$\mathbf{x} = \mathbf{P}\mathbf{X}_\infty = \mathbf{K}\mathbf{R}[\mathbf{I} | -\tilde{\mathbf{C}}] \begin{pmatrix} \mathbf{d} \\ 0 \end{pmatrix} = \mathbf{K}\mathbf{R}\mathbf{d}$$

mapping between π_∞ to an image is given by the planar homography $\mathbf{u} = \mathbf{H}\mathbf{d}$, with $\mathbf{H} = \mathbf{M} = \mathbf{K}\mathbf{R}$

image of the absolute conic (IAC) $\omega = \mathbf{H}^{-T}\Omega_\infty\mathbf{H}^{-1} = \mathbf{K}^{-T}\mathbf{R}\mathbf{I}_3\mathbf{R}^{-1}\mathbf{K}^{-1}$

$$\omega = (\mathbf{K}\mathbf{K}^T)^{-1} = \mathbf{K}^{-T}\mathbf{K}^{-1} \quad (\mathbf{C} \mapsto \mathbf{H}^{-T}\mathbf{C}\mathbf{H}^{-1})$$

- (i) IAC depends only on intrinsics
- (ii) angle between two rays
- (iii) DIAC = $\omega^* = \mathbf{K}\mathbf{K}^T$
- (iv) $\omega \Leftrightarrow \mathbf{K}$ (cholesky factorisation)
- (v) IAC contains image of circular points

$$\cos \theta = \frac{\mathbf{x}_1^T \omega \mathbf{x}_2}{\sqrt{(\mathbf{x}_1^T \omega \mathbf{x}_1)(\mathbf{x}_2^T \omega \mathbf{x}_2)}}$$

summary

homography from π_∞ to image plane: $\mathbf{H} = \mathbf{M} = \mathbf{K} \mathbf{R}$

→ vanishing points = image of points at the ∞ : $\mathbf{v}_d = \mathbf{H} \mathbf{d} = \mathbf{K} \mathbf{R} \mathbf{d}$

image of the absolute conic $\boldsymbol{\omega} = \mathbf{H}^{-T} \Omega_\infty \mathbf{H}^{-1} = \mathbf{K}^{-T} \mathbf{R}^{-T} \mathbf{I}_3 \mathbf{R}^{-1} \mathbf{K}^{-1} = \mathbf{K}^{-T} \mathbf{K}^{-1}$

→ IAC $\boldsymbol{\omega}$ independent of \mathbf{R} and t ; it only depends on intrinsic \mathbf{K}

from IAC $\boldsymbol{\omega}$ to calibration matrix \mathbf{K} : Cholesky factorisation of DIAC $\boldsymbol{\omega}^{-1} = \mathbf{K} \mathbf{K}^T$

absolute conic $\Omega_\infty = \bigcup_{\pi} \{\mathbf{I}_{\pi}, \mathbf{J}_{\pi}\}$ made of circular points

→ IAC $\boldsymbol{\omega} = \bigcup_{\pi} \{\mathbf{I}'_{\pi}, \mathbf{J}'_{\pi}\}$ made of imaged circular points $\mathbf{I}'_{\pi} = \mathbf{H} \mathbf{I}_{\pi}, \mathbf{J}'_{\pi} = \mathbf{H} \mathbf{J}_{\pi}$

constraint on $\boldsymbol{\omega}$ from known angle btw. two directions, and their vanishing points

$$\cos \theta = \frac{\mathbf{v}_1^T \boldsymbol{\omega} \mathbf{v}_2}{\sqrt{(\mathbf{v}_1^T \boldsymbol{\omega} \mathbf{v}_1)(\mathbf{v}_2^T \boldsymbol{\omega} \mathbf{v}_2)}}$$

Scenario n. 1

Known: vanishing points and angle θ between directions

$$\cos \theta = \frac{\mathbf{d}_1^T \mathbf{d}_2}{\sqrt{(\mathbf{d}_1^T \mathbf{d}_1)(\mathbf{d}_2^T \mathbf{d}_2)}} = \frac{\mathbf{v}_1^T \omega \mathbf{v}_2}{\sqrt{(\mathbf{v}_1^T \omega \mathbf{v}_1)(\mathbf{v}_2^T \omega \mathbf{v}_2)}}$$



constraint on the IAC ω
(linear if directions are orthogonal)

Scenario n. 2

Known: vanishing points and IAC ω

$$\cos \theta = \frac{\mathbf{d}_1^T \mathbf{d}_2}{\sqrt{(\mathbf{d}_1^T \mathbf{d}_1)(\mathbf{d}_2^T \mathbf{d}_2)}} = \frac{\mathbf{v}_1^T \omega \mathbf{v}_2}{\sqrt{(\mathbf{v}_1^T \omega \mathbf{v}_1)(\mathbf{v}_2^T \omega \mathbf{v}_2)}}$$



compute angle θ between directions

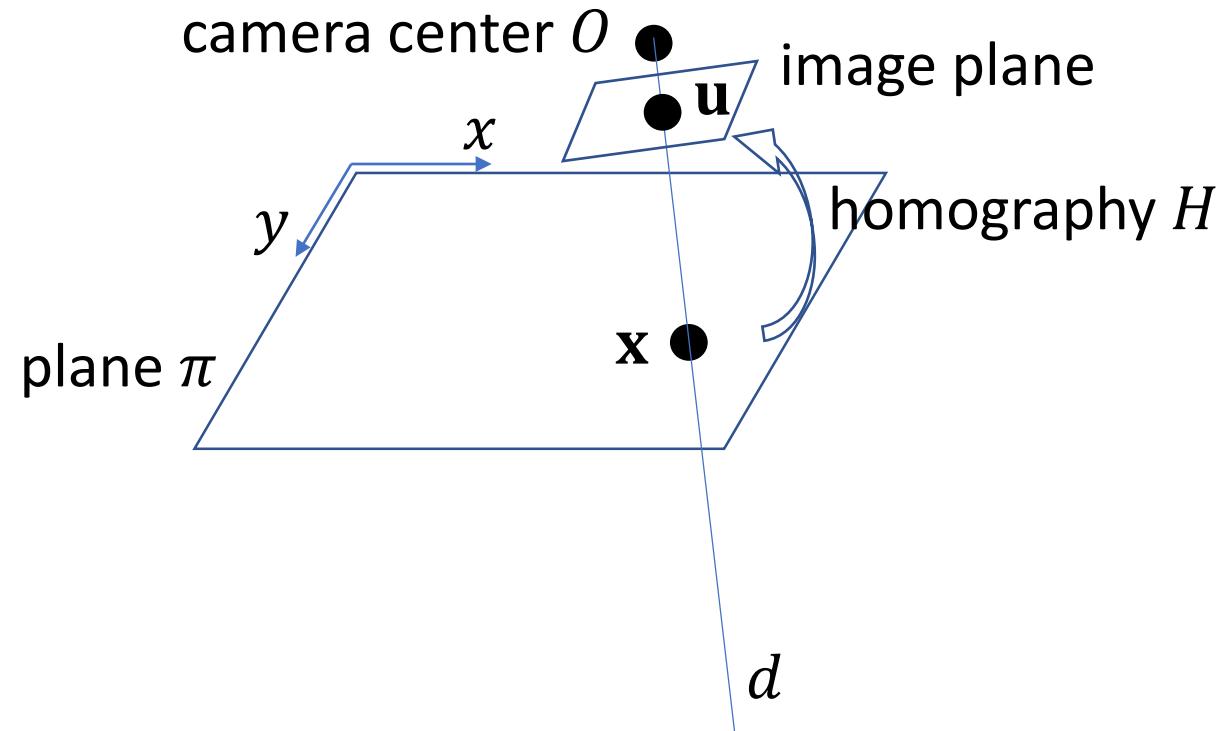
→ reconstruction of the shape

Camera calibration
from a single image of a known planar shape
and known camera center position

Calibration from known plane π and camera center O

Let us refer the coordinates to a reference attached to a known plane π : a generic point on this plane has homogeneous coordinates

$\mathbf{x} = \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$, while $O = \begin{bmatrix} x_o \\ y_o \\ z_o \end{bmatrix}$ are the known cartesian coordinates of the camera center,



Calibration from known plane π and camera center O

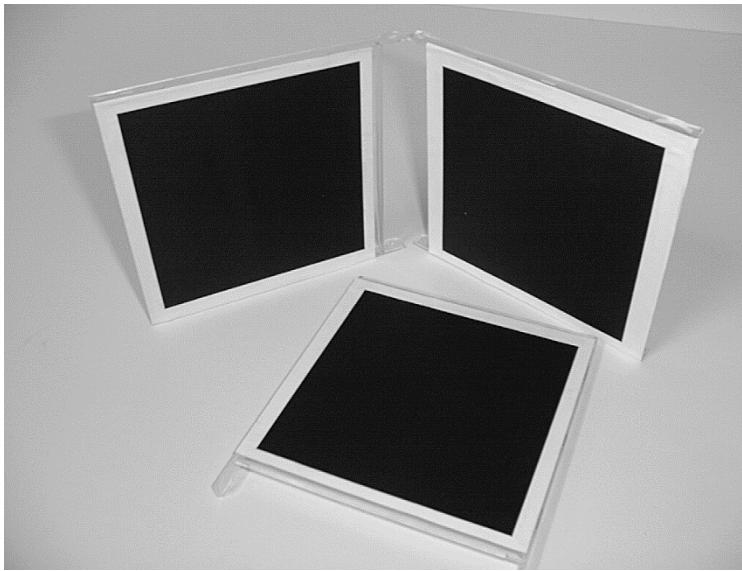
- 1) Estimate homography H from known points on the plane π and their images
- 2) Call M_o the matrix relating any point \mathbf{x} on π to the direction \mathbf{d} of a ray from O to \mathbf{x} :

$$\mathbf{d} = \begin{bmatrix} x - x_o \\ y - y_o \\ -z_o \end{bmatrix} = \begin{bmatrix} 1 & 0 & -x_o \\ 0 & 1 & -y_o \\ 0 & 0 & -z_o \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} \rightarrow M_o^{-1} = \begin{bmatrix} 1 & 0 & -x_o \\ 0 & 1 & -y_o \\ 0 & 0 & -z_o \end{bmatrix}$$

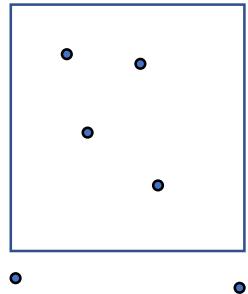
- 3) Compute the matrix M relating any image point \mathbf{u} to the direction \mathbf{d} of its viewing ray from $\mathbf{u} = H\mathbf{x}$, is $\mathbf{d} = M_o^{-1}\mathbf{x} = M_o^{-1}H^{-1}\mathbf{u} \rightarrow M^{-1} = M_o^{-1}H^{-1}$
- 4) Q-R decompose matrix M^{-1} as $M^{-1} = R^{-1}K^{-1}$, where K is the camera intrinsic calibration matrix and R^{-1} is the rotation matrix from the world (reference attached to π) to the camera

Camera calibration from images of known planar shapes (Zhang method)

A simple calibration device (Zhang 2000)

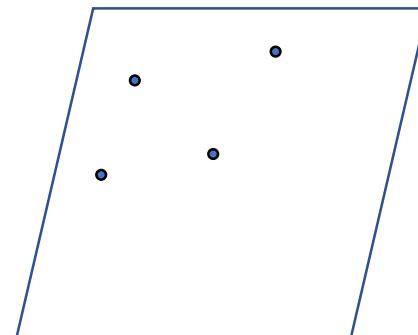
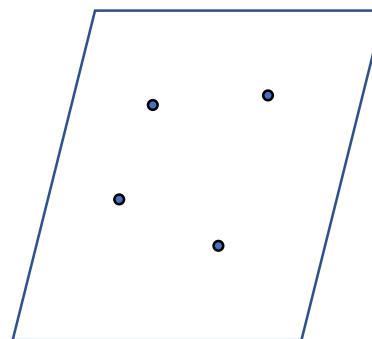
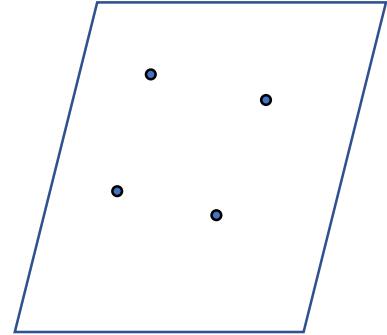
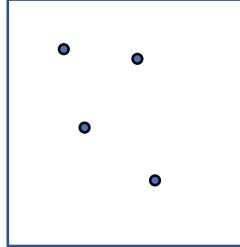


- (i) compute H for each square
(corners $\square (0,0), (1,0), (0,1), (1,1)$)
- (ii) compute the imaged circular points $H(1, \pm i, 0)^T$
- (iii) fit a conic to 6 imaged circular points
- (iv) compute K from ω through Cholesky factorization

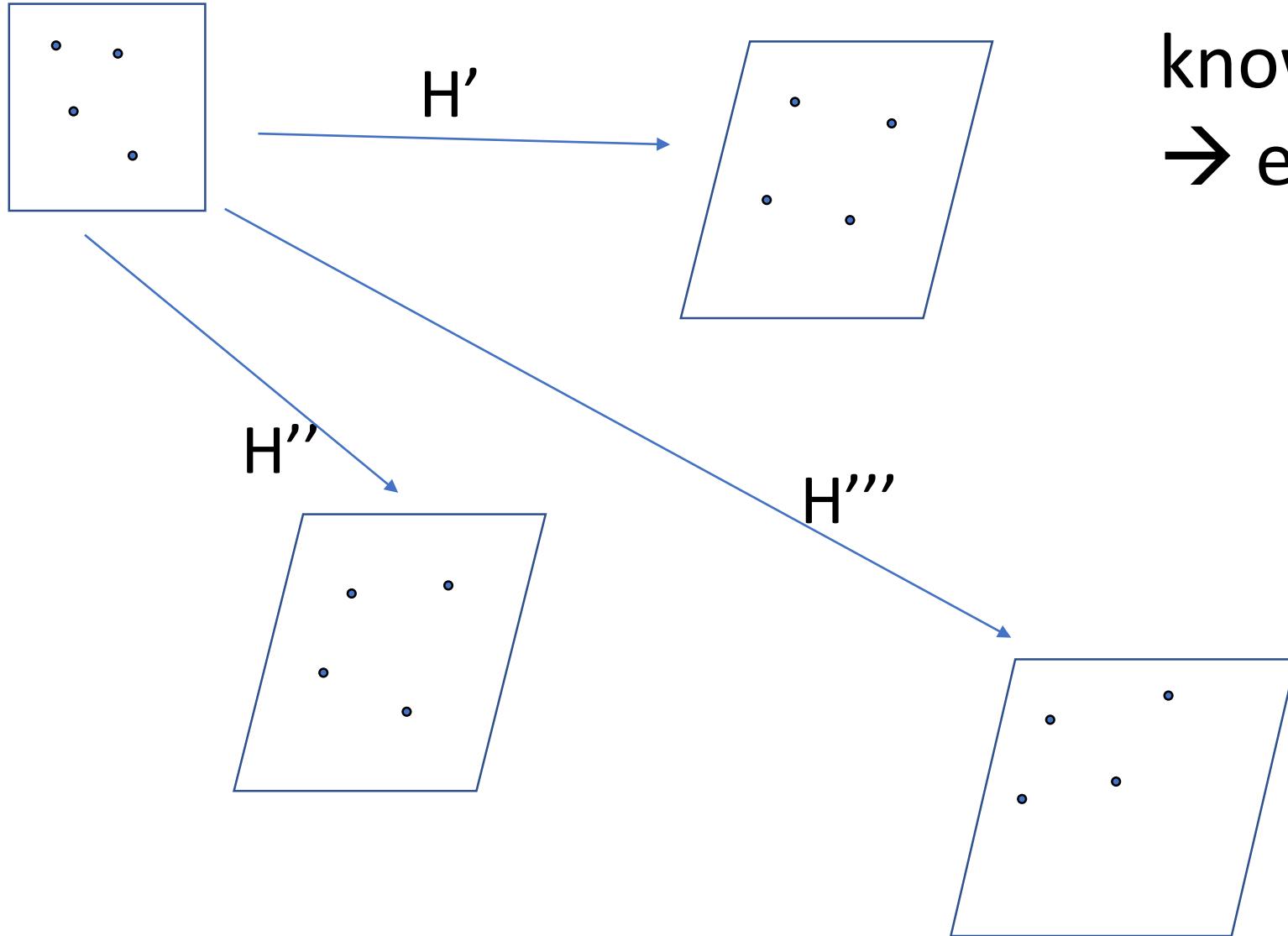


known planar scene

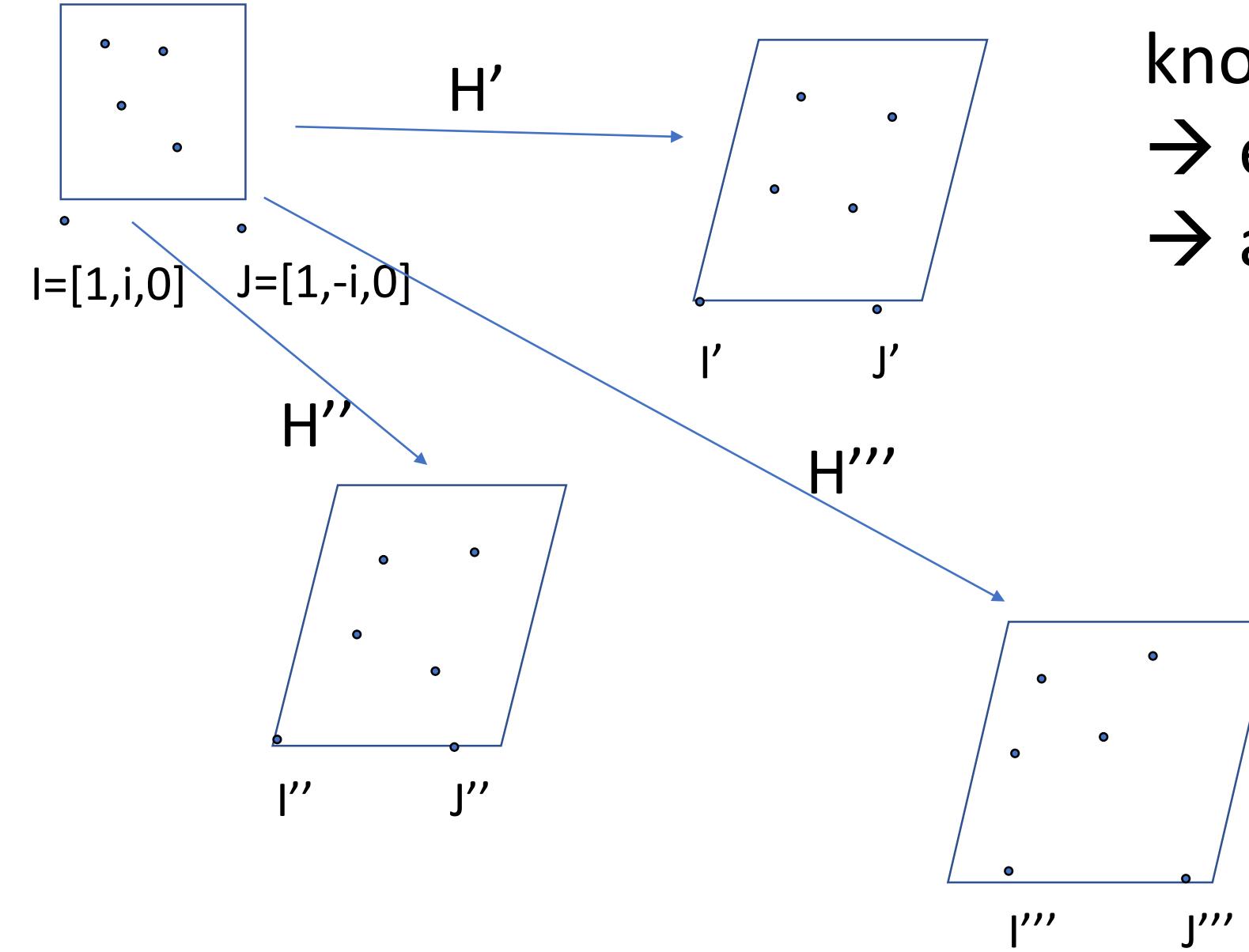
$$I = [1, i, 0] \quad J = [1, -i, 0]$$



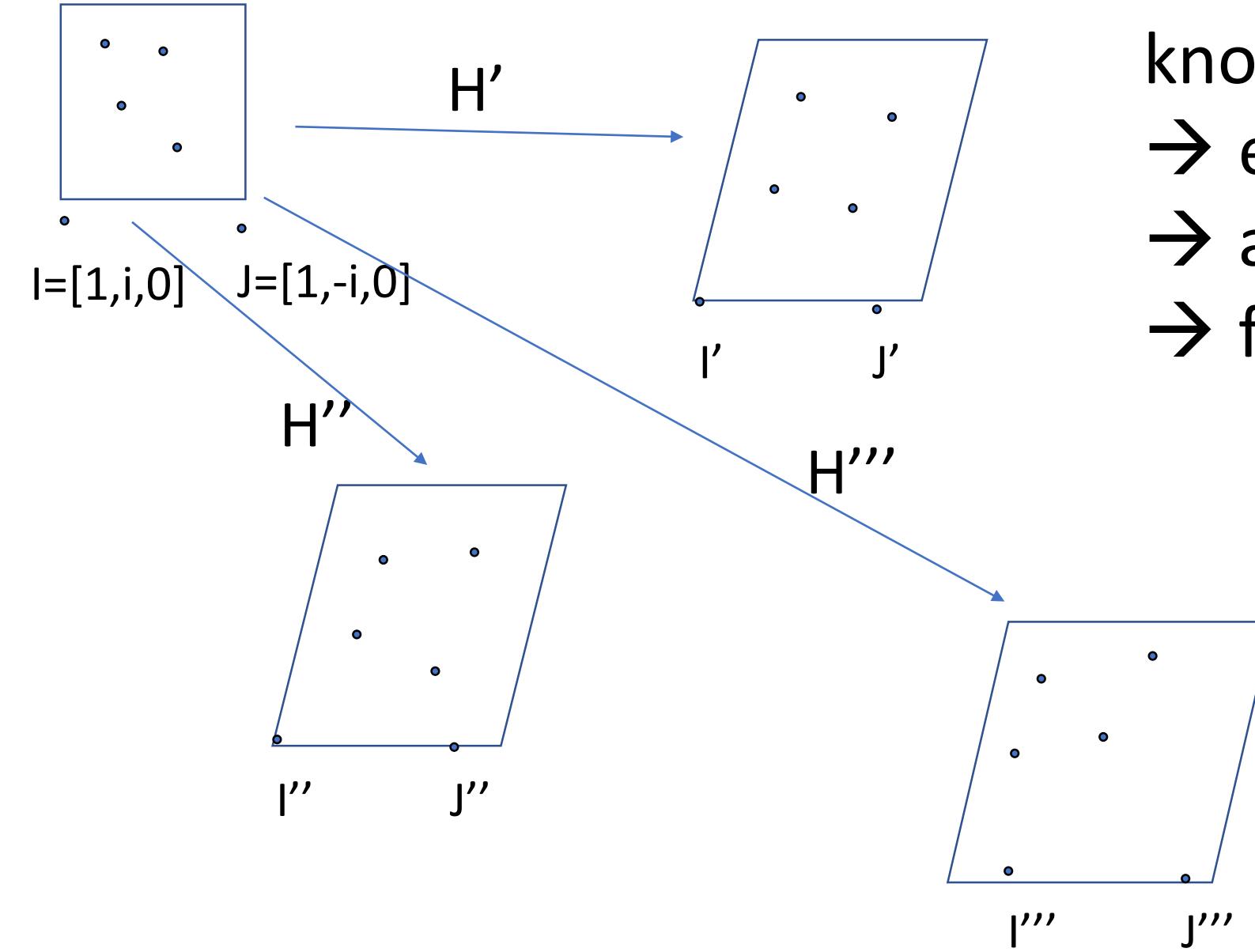
known planar scene:
take at least 3 images with
the same camera (constant \mathbf{K})



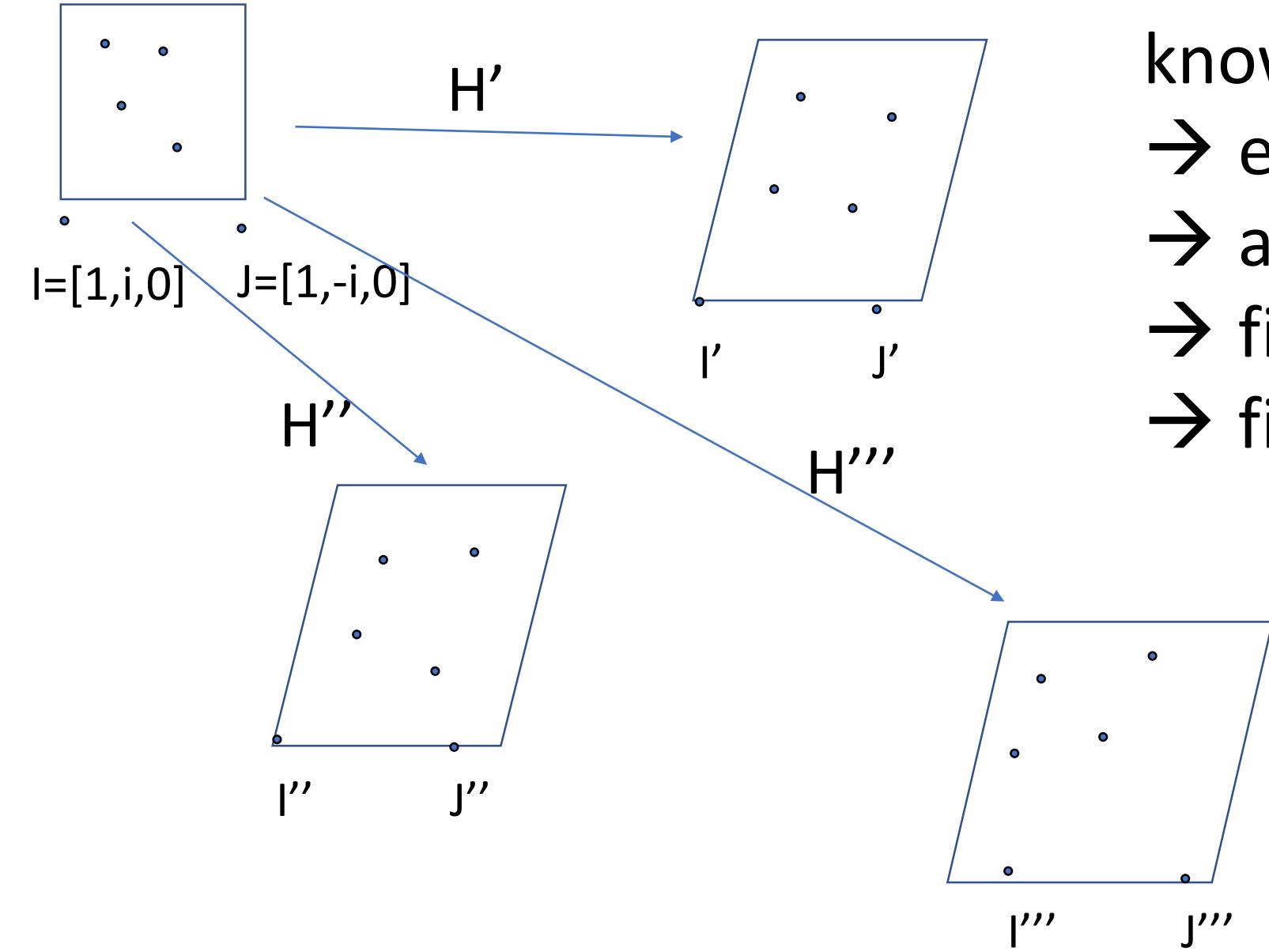
known planar scene
→ estimate homographies



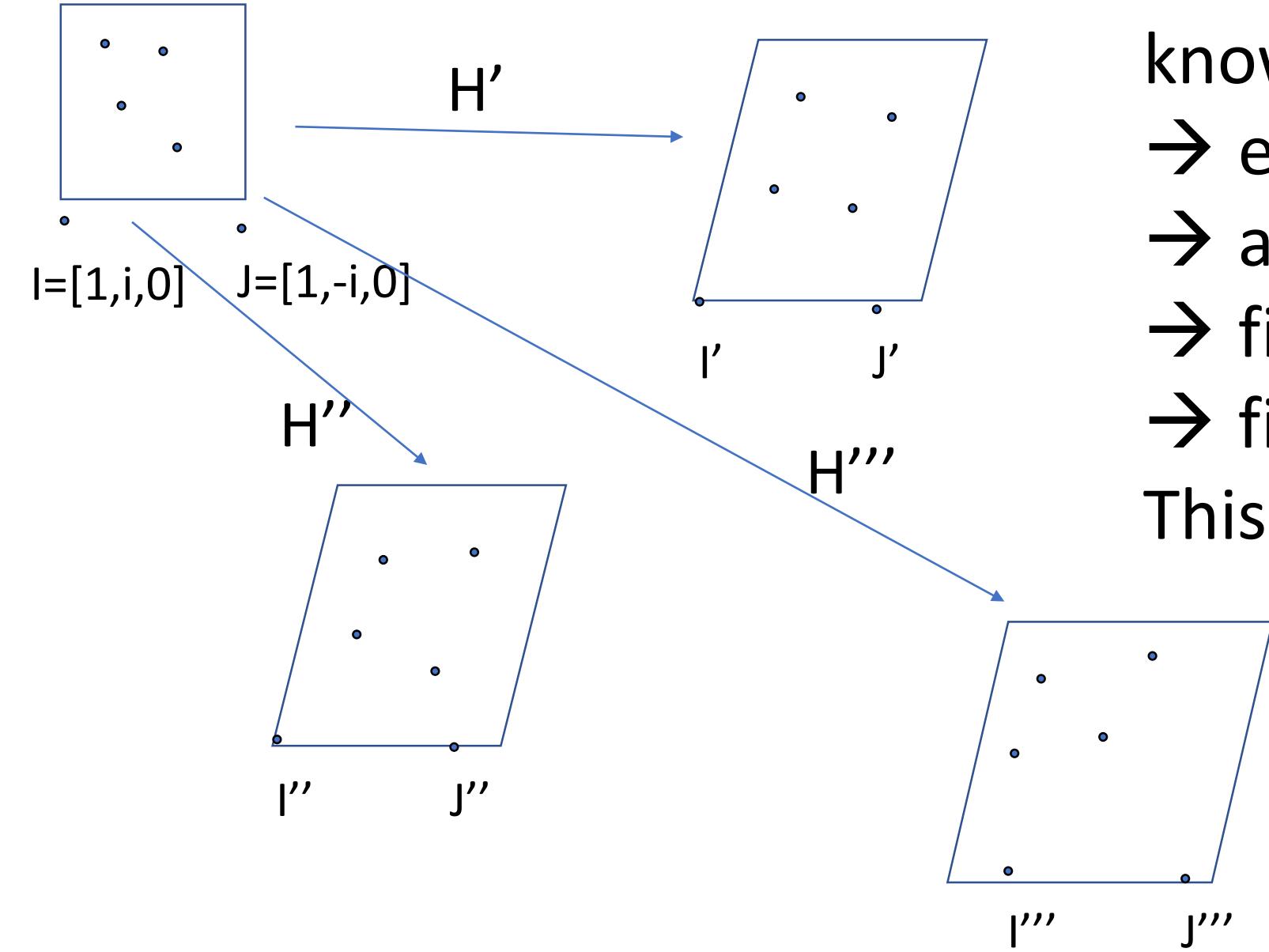
known planar scene
 → estimate homographies
 → apply them to I and J



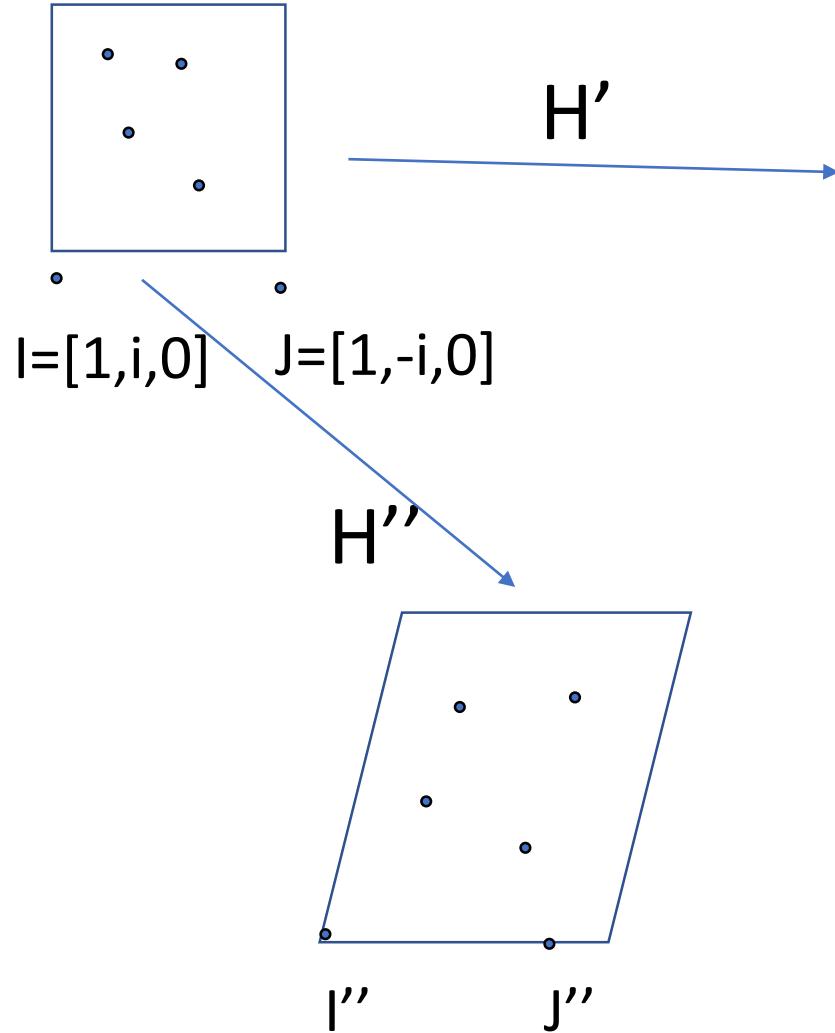
known planar scene
 → estimate homographies
 → apply them to I and J
 → find $I', J', I'', J'', I''', J'''$



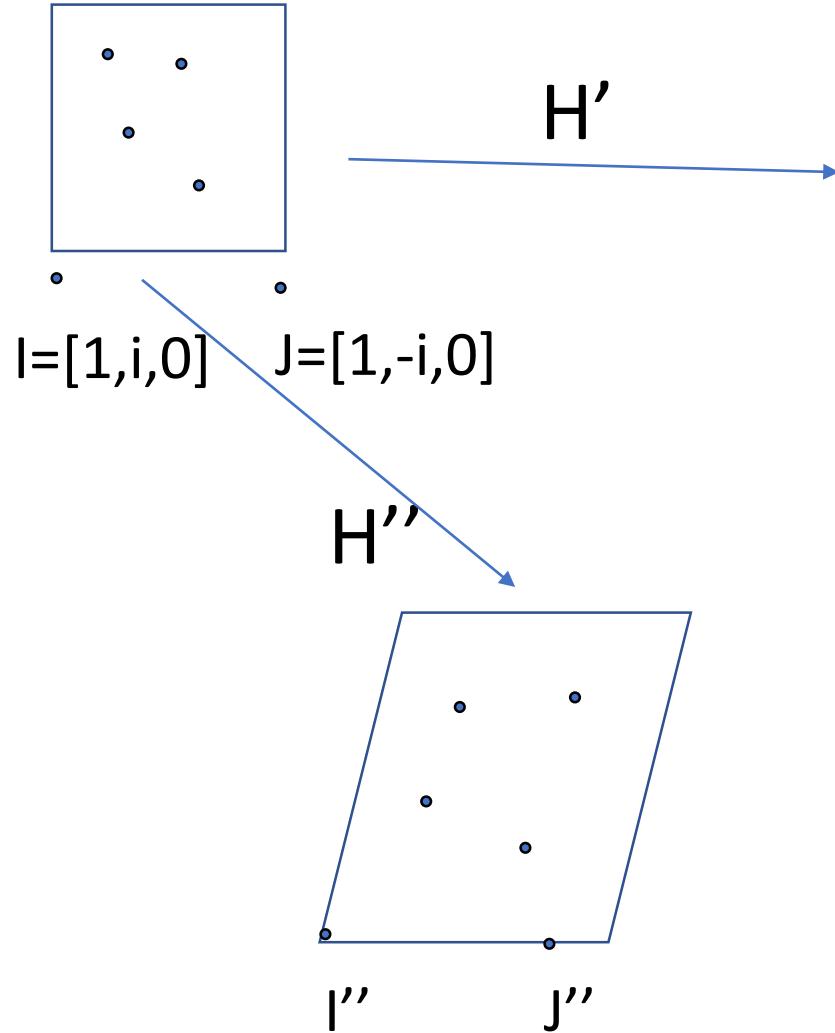
known planar scene
 → estimate homographies
 → apply them to I and J
 → find I' , J' , I'' , J'' , I''' , J'''
 → fit a conic ω to them:



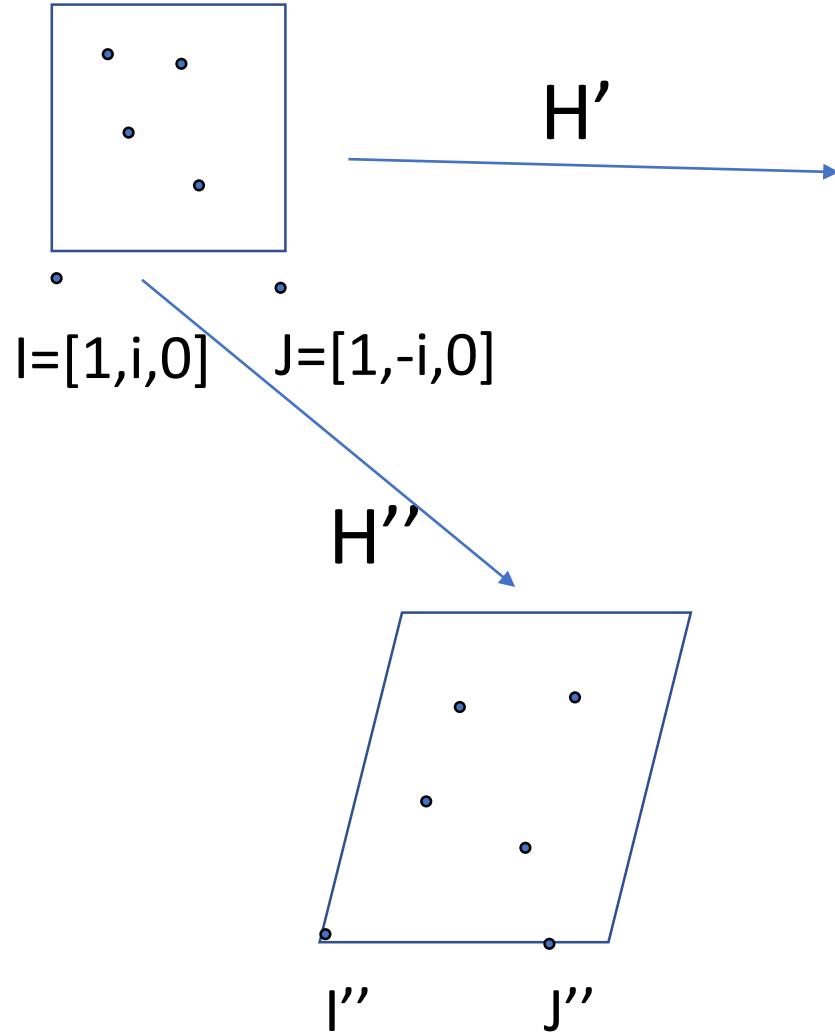
known planar scene
 → estimate homographies
 → apply them to I and J
 → find I' , J' , I''' , J'''
 → fit a conic to them:
 This conic is the IAC ω !!



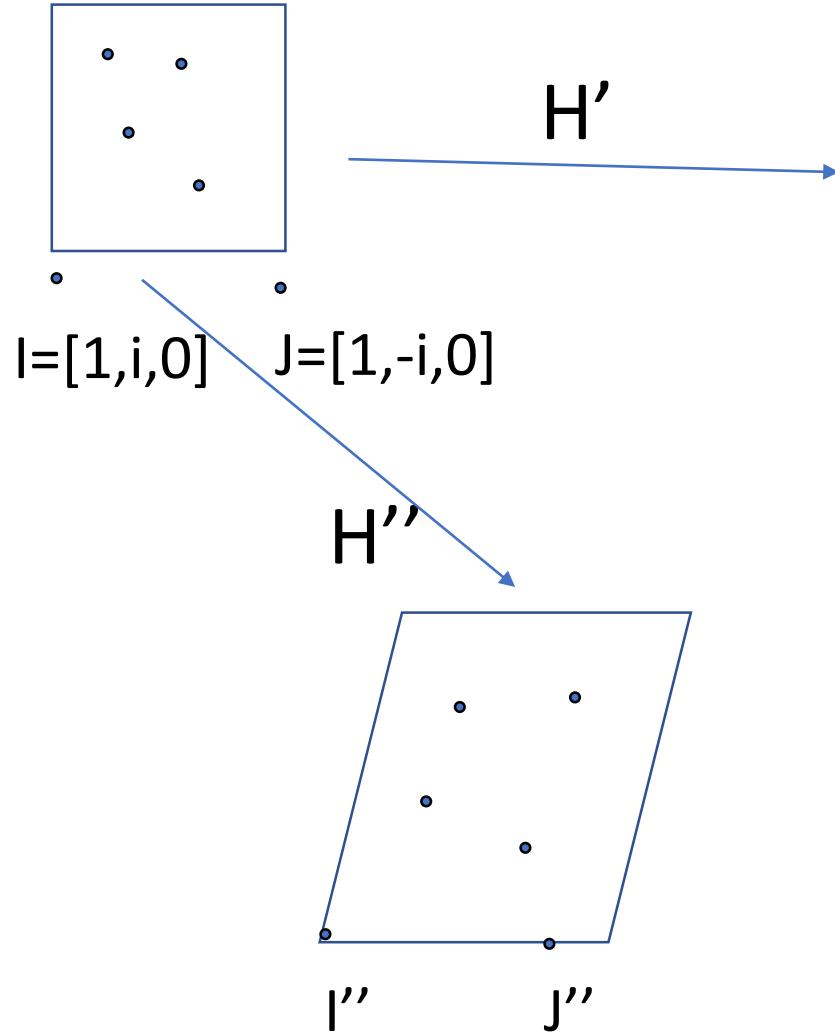
- known planar scene
- estimate homographies
- apply them to I and J
- find $I', J', I'', J'', I''', J'''$
- fit a conic to them:
- This conic is the IAC ω !!
- then take the DIAC ω^{-1}



- known planar scene
- estimate homographies
- apply them to I and J
- find $I', J', I'', J'', I''', J'''$
- fit a conic to them:
- This conic is the IAC ω !!
- then take the DIAC ω^{-1}
- find K by Cholesky factorisation of
 $\omega^{-1} = KK^T$



- known planar scene
- estimate homographies
- apply them to I and J
- find $I', J', I'', J'', I''', J'''$
- fit a conic to them: (HOW?)
- This conic is the IAC ω !!
- then take DIAC ω^{-1}
- find K by Cholesky factorisation of
 $\omega^{-1} = KK^T$



- known planar scene
- estimate homographies
 - apply them to I and J
 - find $I', J', I'', J'', I''', J'''$
 - fit a conic to them: (HOW?)
 - This conic is the IAC ω !! (WHY?)
 - then take DIAC ω^{-1}
 - find K from Cholesky factorisation
of $\omega^{-1} = KK^T$

HOW? Fitting ω to images of circular points

$$I' = H'I = [h_1, h_2, h_3]I = [h_1, h_2, h_3] \begin{bmatrix} 1 \\ i \\ 0 \end{bmatrix} = h_1 + ih_2 \in \omega$$
$$(h_1 + ih_2)^T \omega (h_1 + ih_2) = 0$$

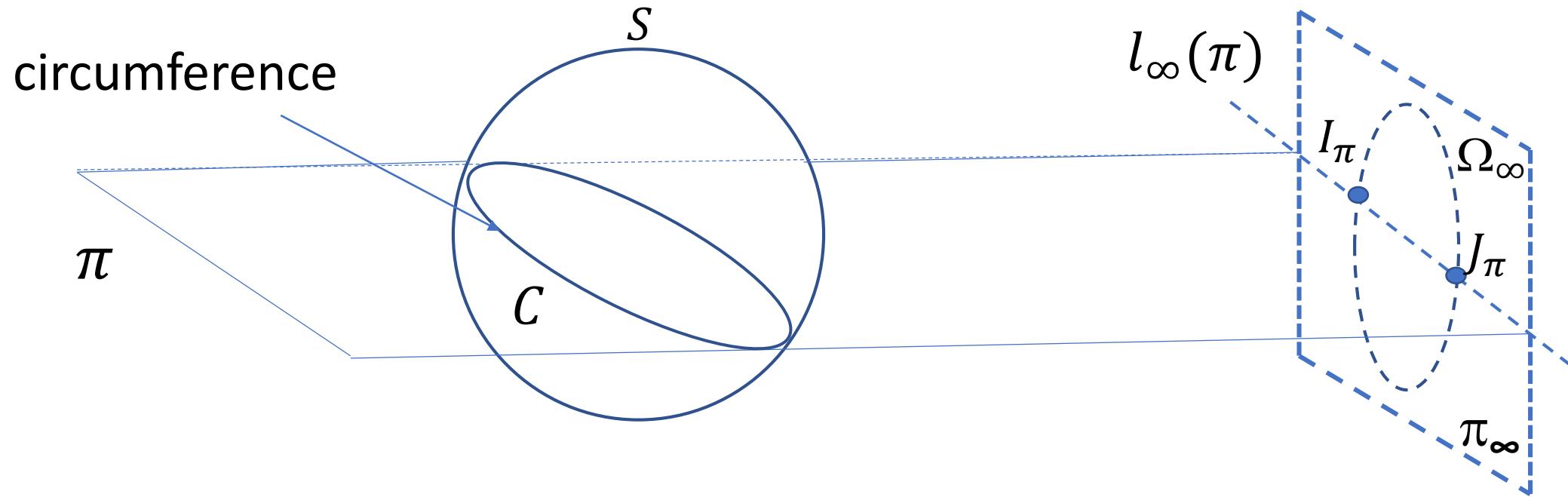
Real part = 0 and imaginary part = 0

$${h_1}^T \omega h_2 = 0$$
$${h_1}^T \omega h_1 - {h_2}^T \omega h_2 = 0$$

2 **real** eqns for each image:
→ at least 3 images needed

WHY does IAC ω contain images of circular points?

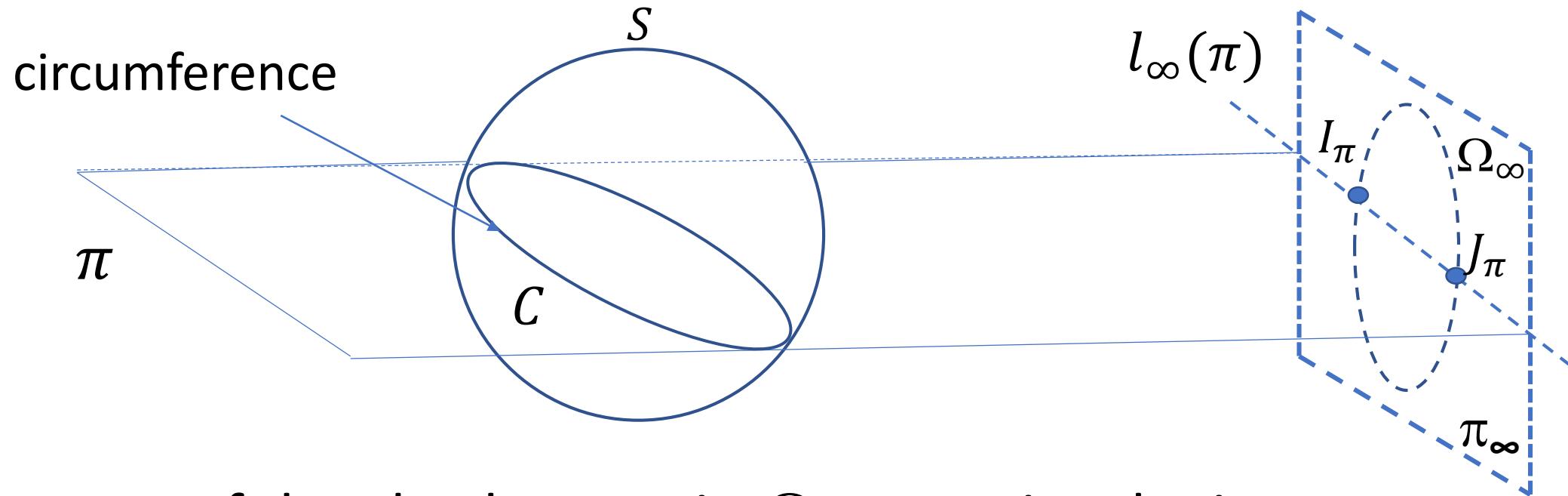
1. Does the absolute conic Ω_∞ contain all the circular points (i.e. the circular points I_π, J_π of any plane belong to absolute conic Ω_∞) ?



C crosses $l_\infty(\pi)$ at circular points I_π, J_π but $C \subset S$, thus $I_\pi, J_\pi \in S$
In addition, $I_\pi, J_\pi \in \pi_\infty \rightarrow I_\pi, J_\pi \in \pi_\infty \cap S = \Omega_\infty$

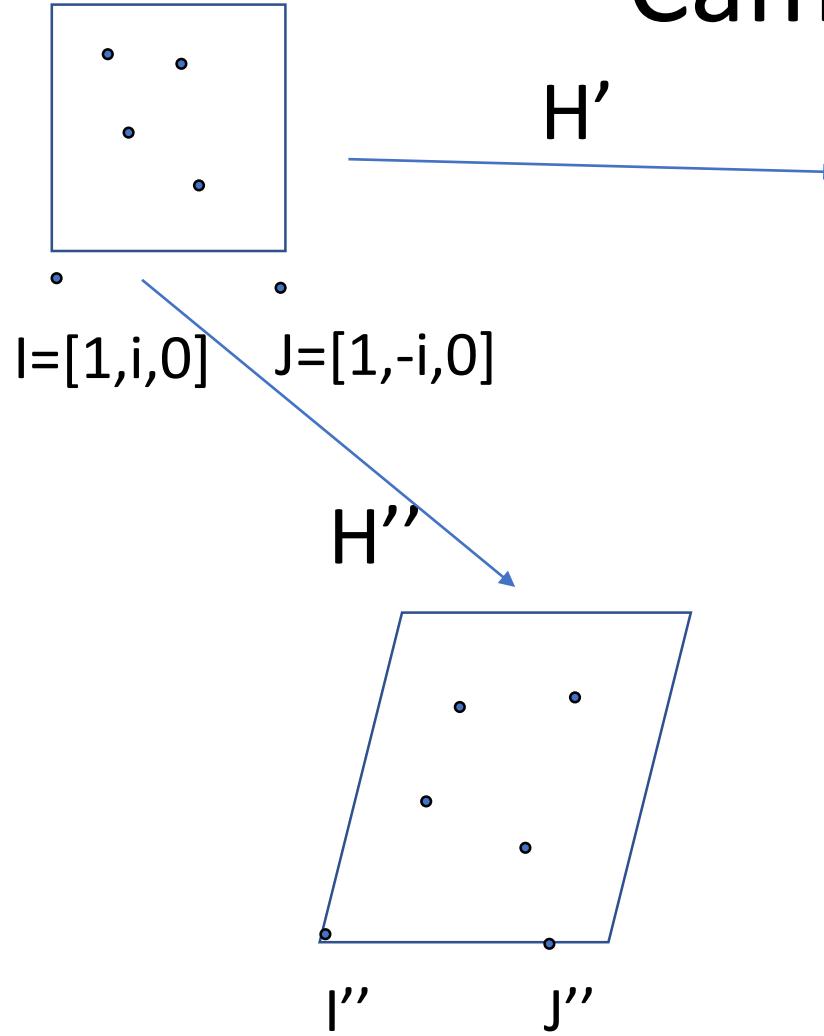
WHY does IAC ω contain images of circular points?

1. Yes: the absolute conic Ω_∞ contains all the circular points (i.e. the circular points I_π, J_π of any plane belong to absolute conic Ω_∞) !!



→ The image ω of the absolute conic Ω_∞ contains the image I'_π, J'_π of all the circular points I_π, J_π

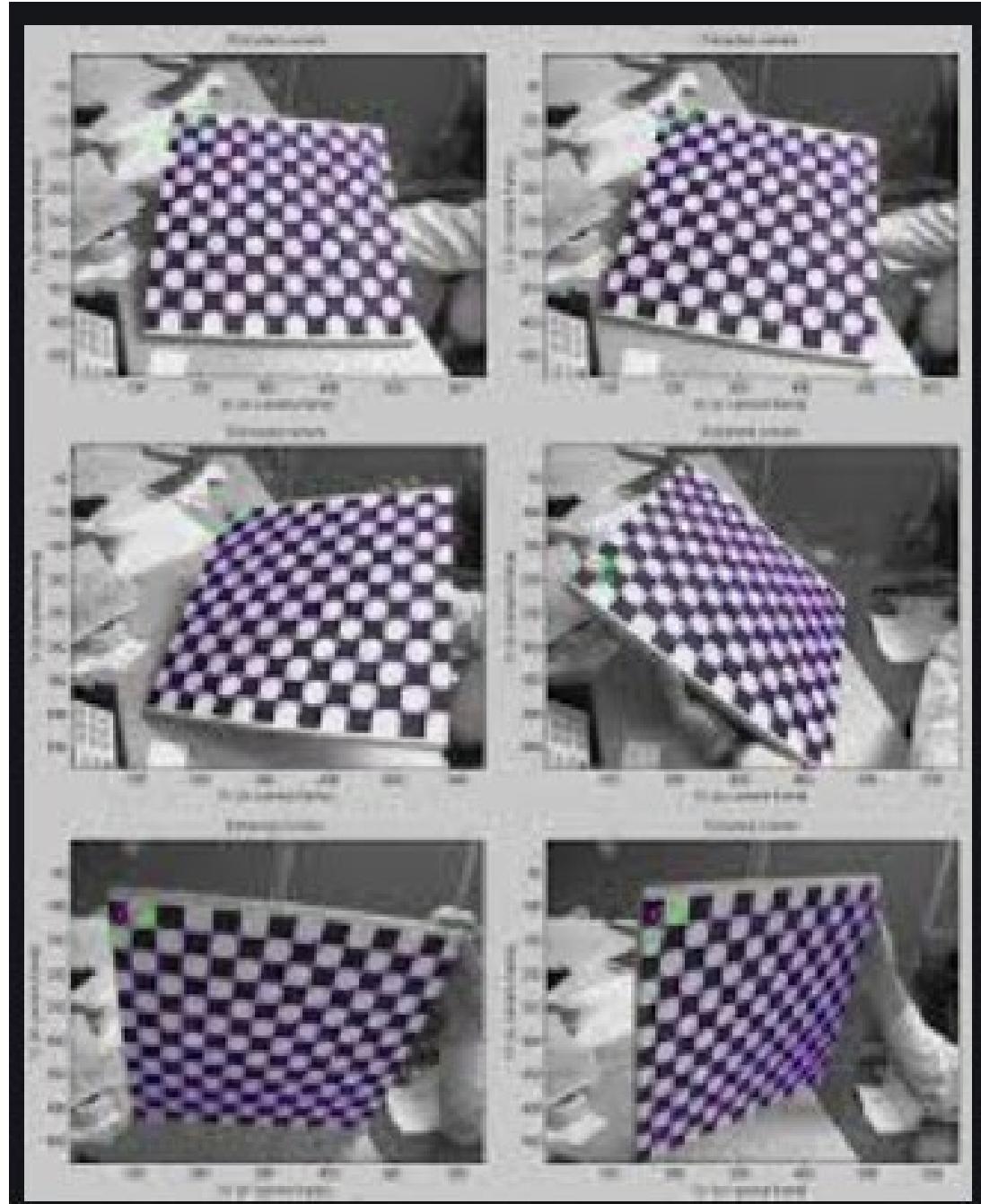
Camera calibration from images of a known scene



- Take ≥ 3 images
 - estimate homographies
 - apply them to I and J
 - find $I', J', I'', J'', I''', J'''$
 - fit a conic to them:
 - This conic is the IAC ω !!
 - then take the DIAC ω^{-1}
 - find K by Cholesky factorisation of $\omega^{-1} = KK^T$

Camera calibration toolbox

- implements Zhang method
- based on the IAC ω
- planar target (easily printable)
- several images (~ 20) to cope with noise
- also estimates distortion param.
- provides accurate calibration





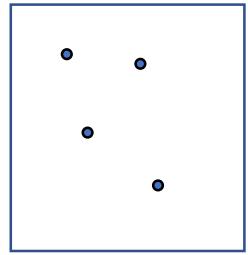
Matlab Calibration
Toolbox also contains
estimation of
distortion parameters

$$x = x_o + (x_o - c_x)(K_1 r^2 + K_2 r^4 + \dots)$$

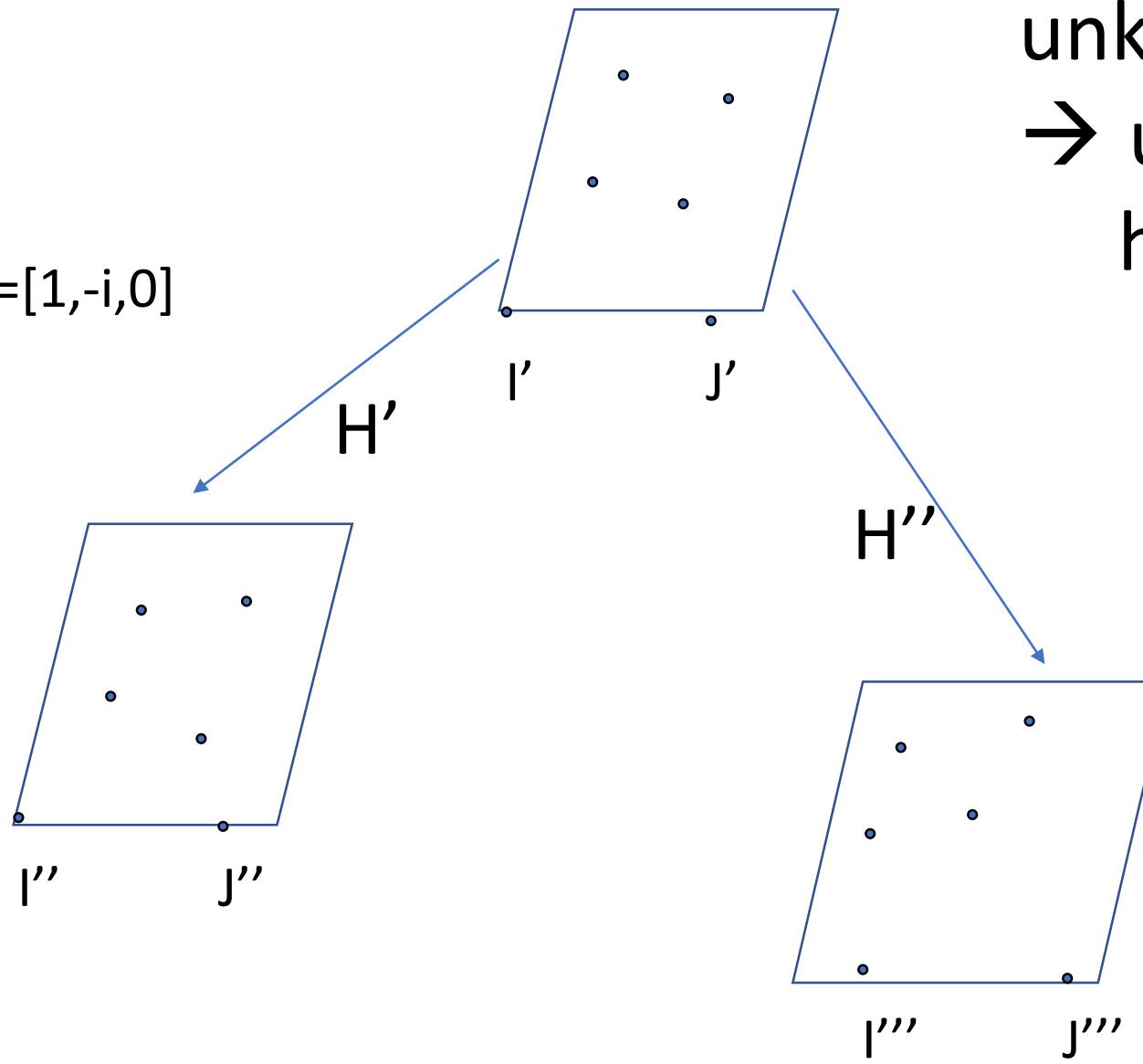
$$y = y_o + (y_o - c_y)(K_1 r^2 + K_2 r^4 + \dots)$$

$$r = (x_o - c_x)^2 + (y_o - c_y)^2 .$$

Camera calibration from images of an **unknown** planar scene



$$I = [1, i, 0] \quad J = [1, -i, 0]$$



unkown planar scene
→ use image-to-image
homographies

$$I'' = H'I'$$

$$I'^T \omega I' = 0$$

$$I''^T \omega I'' = 0$$

$$I'^T H'^T \omega H'I' = 0$$

Unknowns: I' and J' and ω \rightarrow at least 5 images
(each nonlinear eqn leads to 2 constraints: Re and Im part)

$$I'' = H'I'$$

$$I'^T \omega I' = 0$$

$$I''^T \omega I'' = 0$$

$$I'^T H'^T \omega H'I' = 0$$

Unknowns: I' and J' and ω \rightarrow at least 5 images
(each nonlinear eqn leads to 2 constraints: Re and Im part)

Camera calibration from images of an **unknown** planar scene

- take images of a planar scene with camera with constant \mathbf{K}
- estimate image-image homographies
- formulate equations

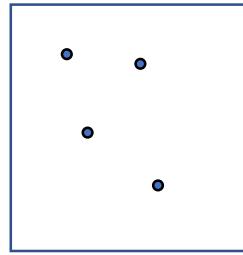
$$\mathbf{I}'^T \boldsymbol{\omega} \mathbf{I}' = 0$$

$$\mathbf{I}'^T \mathbf{H}'^T \boldsymbol{\omega} \mathbf{H}' \mathbf{I}' = 0$$

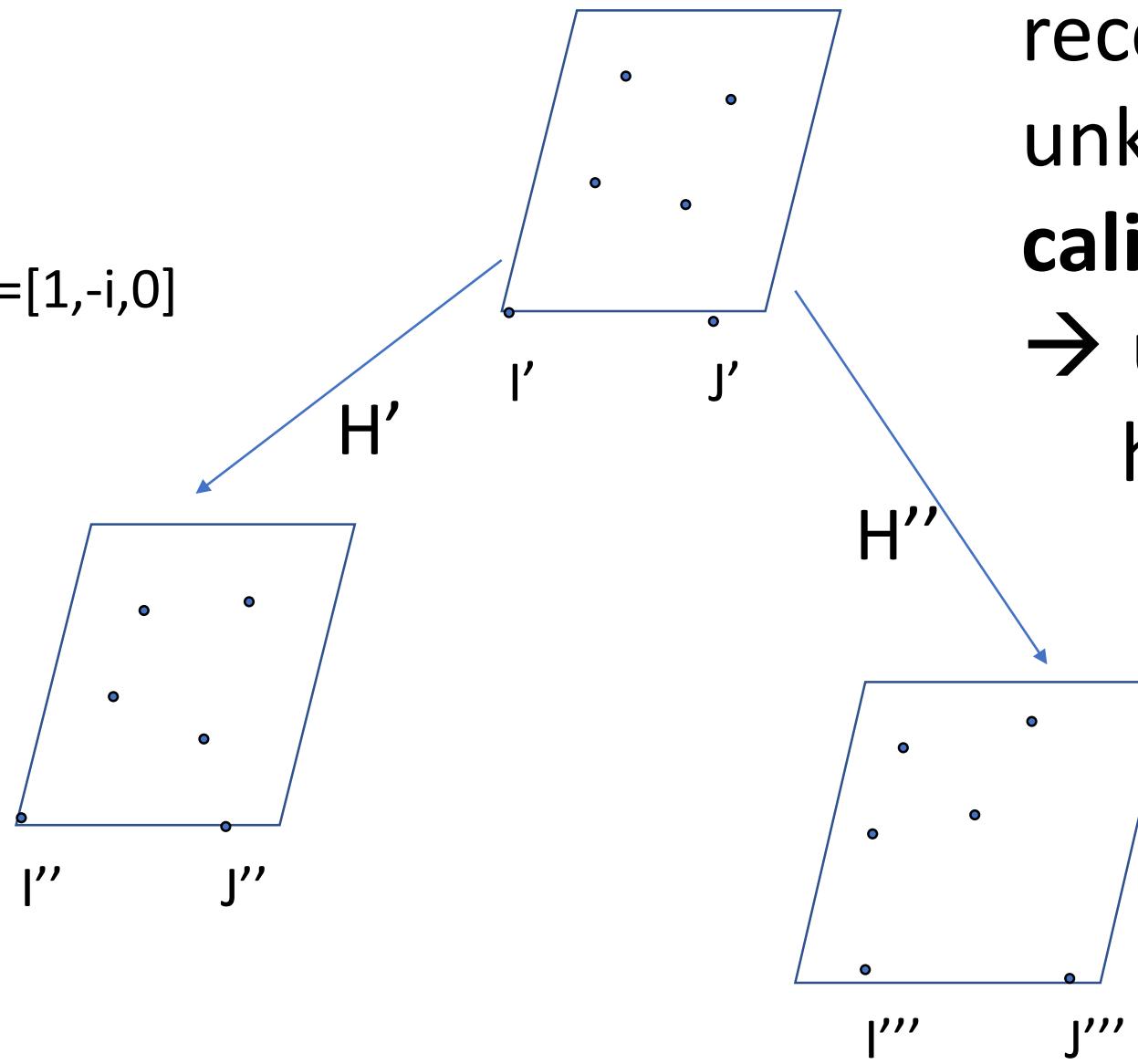
- solve them for $\boldsymbol{\omega}$ and \mathbf{I}'
- then take the DIAC $\boldsymbol{\omega}^{-1}$
- find \mathbf{K} by Cholesky factorisation of $\boldsymbol{\omega}^{-1} = \mathbf{K} \mathbf{K}^T$

USUALLY LESS ACCURATE THAN ZHANG METHOD

Reconstruction of an unknown planar scene
from calibrated images (i.e. images taken by
calibrated cameras)



$$I=[1,i,0] \quad J=[1,-i,0]$$



reconstruction of an
unkown planar scene, but
calibrated camera
→ use image-to-image
homographies

same calibrated camera

$$I'' = H'I'$$

$$I'^T \omega I' = 0$$

$$I''^T \omega I'' = 0$$

$$I'^T H'^T \omega H'I' = 0$$

$\omega = (KK^T)^{-1} = K^{-T}K^{-1}$ is known after calibration

just 4 unknowns I' complex coordinates \rightarrow at least 2 images
(each eqn leads to 2 constraints: Re and Im part)

two different calibrated cameras

$$I'' = H'I'$$

$$I'^T \omega' I' = 0$$

$$I''^T \omega'' I'' = 0$$

$$I'^T H'^T \omega'' H'I' = 0$$

ω', ω'' known after calibration

just 4 unknowns I' complex coordinates → at least 2 images
(each eqn leads to 2 constraints: Re and Im part)

$$I'^T \omega' I' = 0$$
$$I'^T H'^T \omega'' H' I' = 0$$

reduces to intersection of two conics:

ω' and $H'^T \omega'' H'$



two resulting pairs of
imaged circular points



selection based on reprojection
or on an additional (third) image

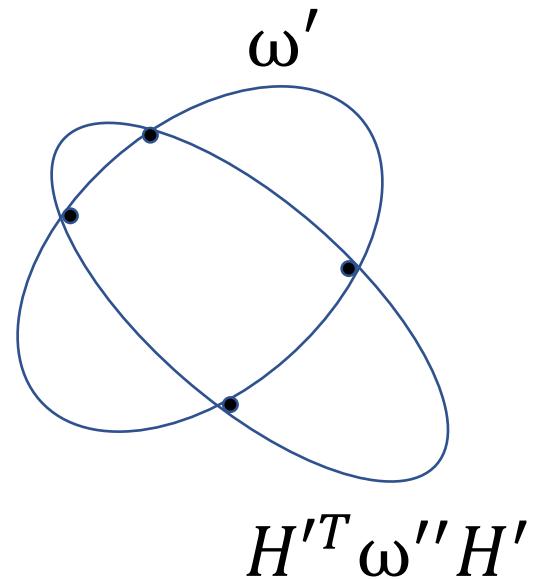


image of circular points: $\{l', J'\} = l'_\infty \cap \omega$

- Image of the circular points \rightarrow Image of the conic dual to the circular points

$$C_\infty^{*'} = I'J'^T + J'I'^T$$

- Singular value decomposition

$$\text{svd}(C_\infty^{*'}) = U \begin{vmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{vmatrix} U^T = H_{rect}^{-1} C_\infty^* H_{rect}^{-T}$$

- Rectifying homography (from svd output U)

$$H_{rect} = U^T$$

In practice ...

- Image of the circular points → Image of the conic dual to the circular points

$$C_{\infty}' = I'J'^T + J'I'^T$$

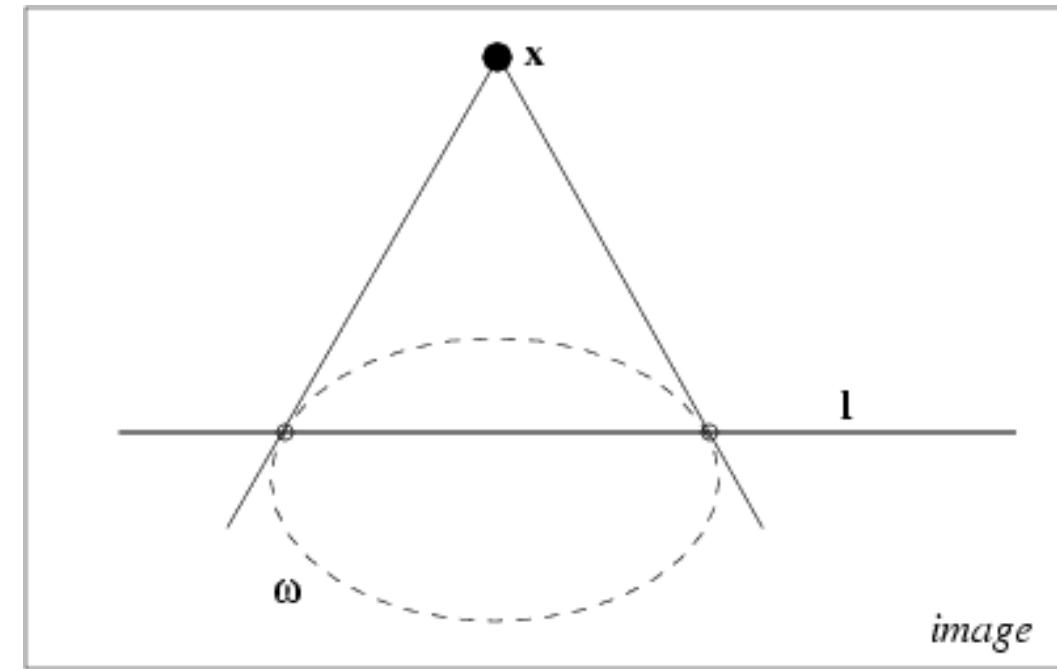
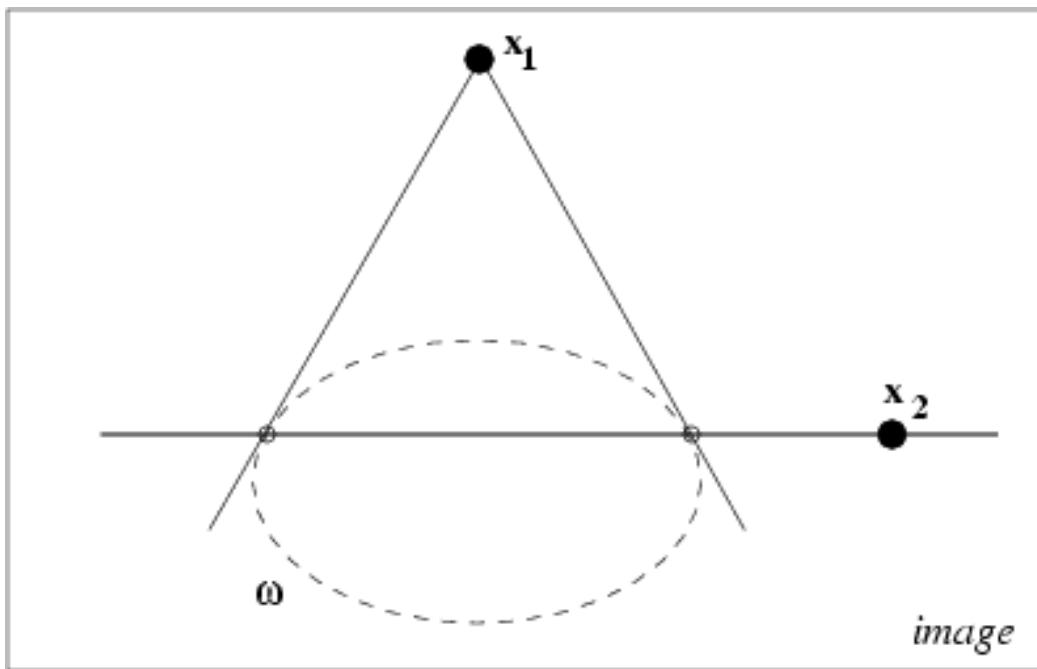
- Singular value decomposition

$$\text{svd}(C_{\infty}') = U \begin{bmatrix} s_1 & 0 & 0 \\ 0 & s_2 & 0 \\ 0 & 0 & \varepsilon \end{bmatrix} U^T$$

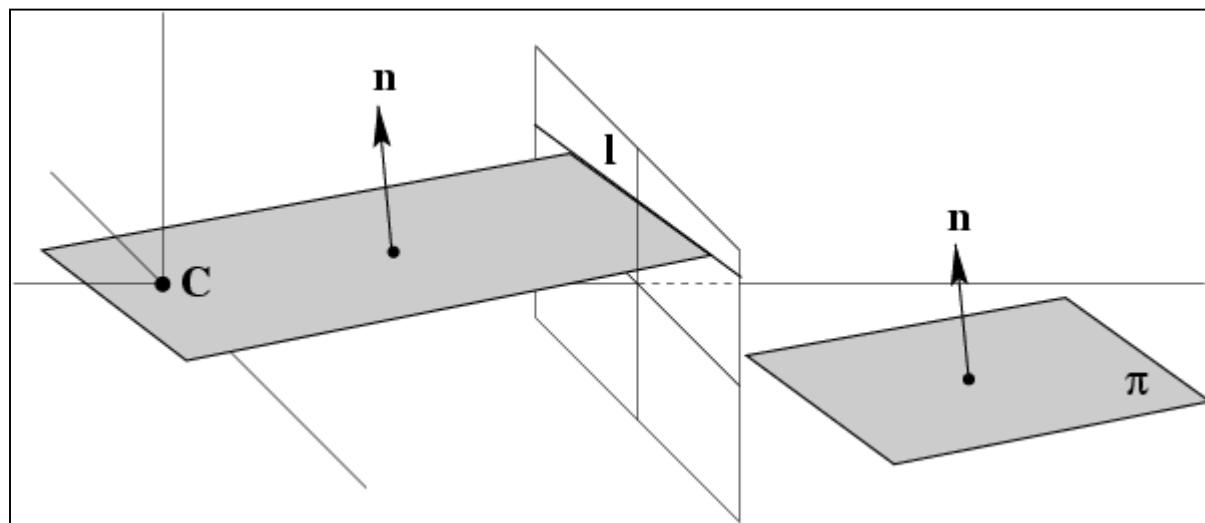
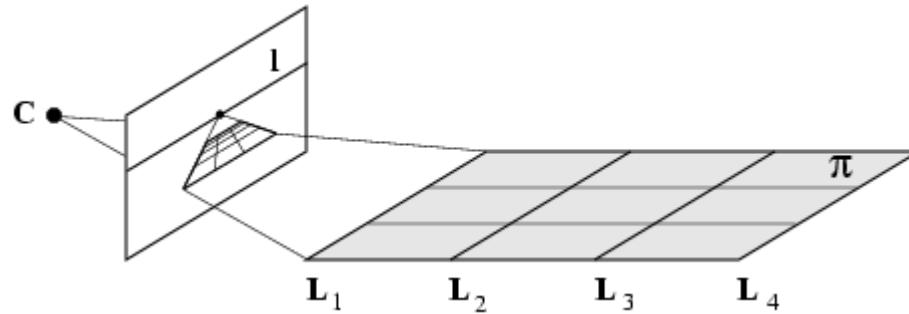
- Rectifying homography (from svd output U)

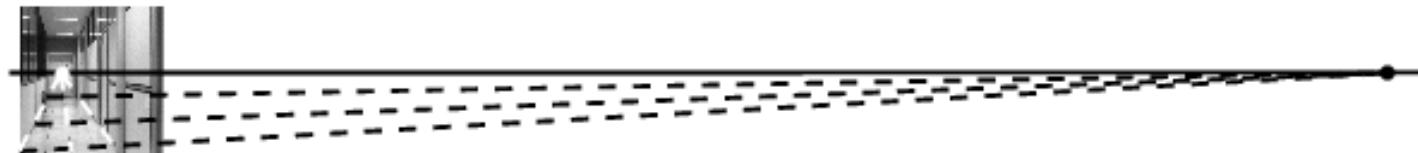
$$H_{rect} = (U \begin{bmatrix} \sqrt{s_1} & 0 & 0 \\ 0 & \sqrt{s_2} & 0 \\ 0 & 0 & 1 \end{bmatrix})^{-1} = \begin{bmatrix} \sqrt{s_1^{-1}} & 0 & 0 \\ 0 & \sqrt{s_2^{-1}} & 0 \\ 0 & 0 & 1 \end{bmatrix} U^T$$

Orthogonality = pole-polar w.r.t. IAC



Vanishing lines





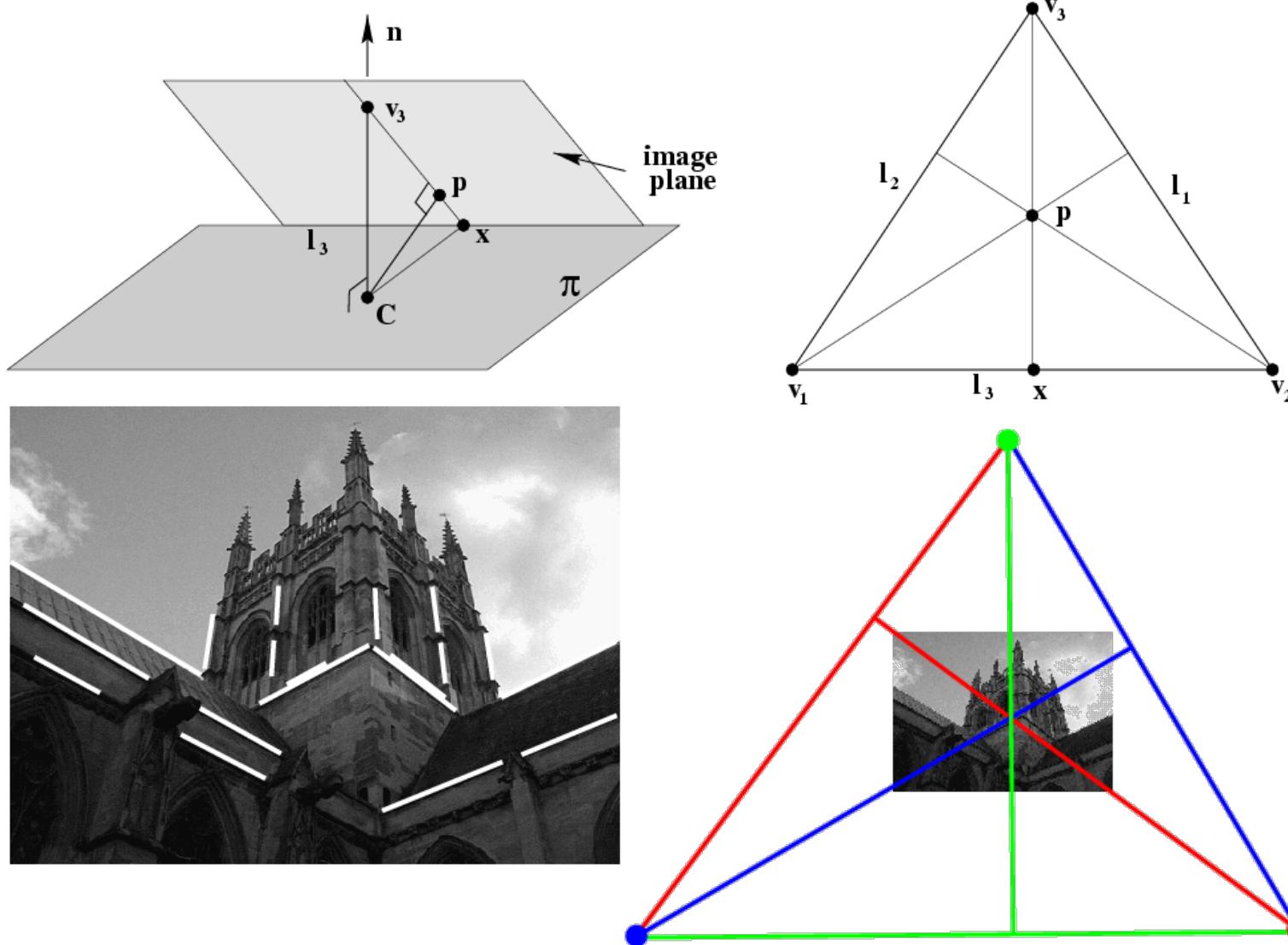
vanishing line: line through two vanishing points

Orthogonality relation: nonlinear equation on ω
reduces to linear for orthogonal directions

$$\cos \theta = \frac{\mathbf{v}_1^T \boldsymbol{\omega} \mathbf{v}_2}{\sqrt{(\mathbf{v}_1^T \boldsymbol{\omega} \mathbf{v}_1)(\mathbf{v}_2^T \boldsymbol{\omega} \mathbf{v}_2)}}$$

$$\mathbf{v}_1^T \boldsymbol{\omega} \mathbf{v}_2 = 0$$

Calibration of NATURAL CAMERAS from vanishing points of orthogonal directions



Orthogonality relation

$$\cos \theta = \frac{\mathbf{v}_1^T \boldsymbol{\omega} \mathbf{v}_2}{\sqrt{(\mathbf{v}_1^T \boldsymbol{\omega} \mathbf{v}_1)(\mathbf{v}_2^T \boldsymbol{\omega} \mathbf{v}_2)}} = 0$$

$$\mathbf{v}_1^T \boldsymbol{\omega} \mathbf{v}_2 = 0$$

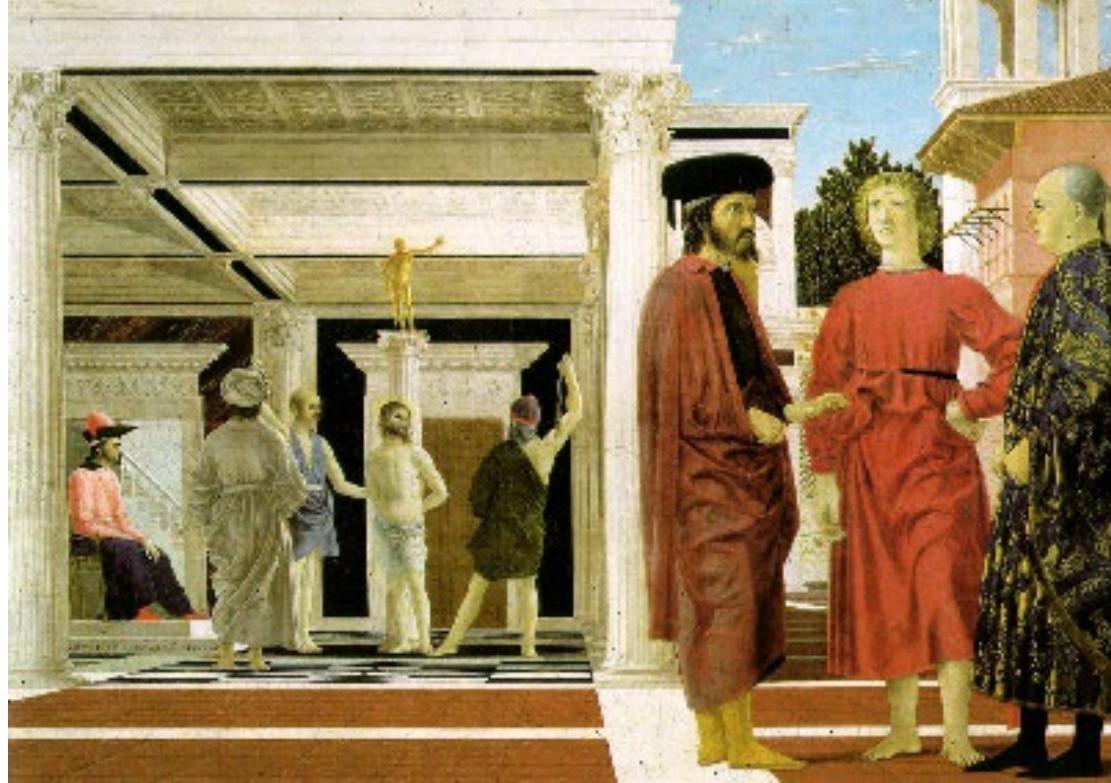
$$\mathbf{v}_2^T \boldsymbol{\omega} \mathbf{v}_3 = 0$$

$$\mathbf{v}_3^T \boldsymbol{\omega} \mathbf{v}_1 = 0$$

The painter's eye

An Example of Camera calibration

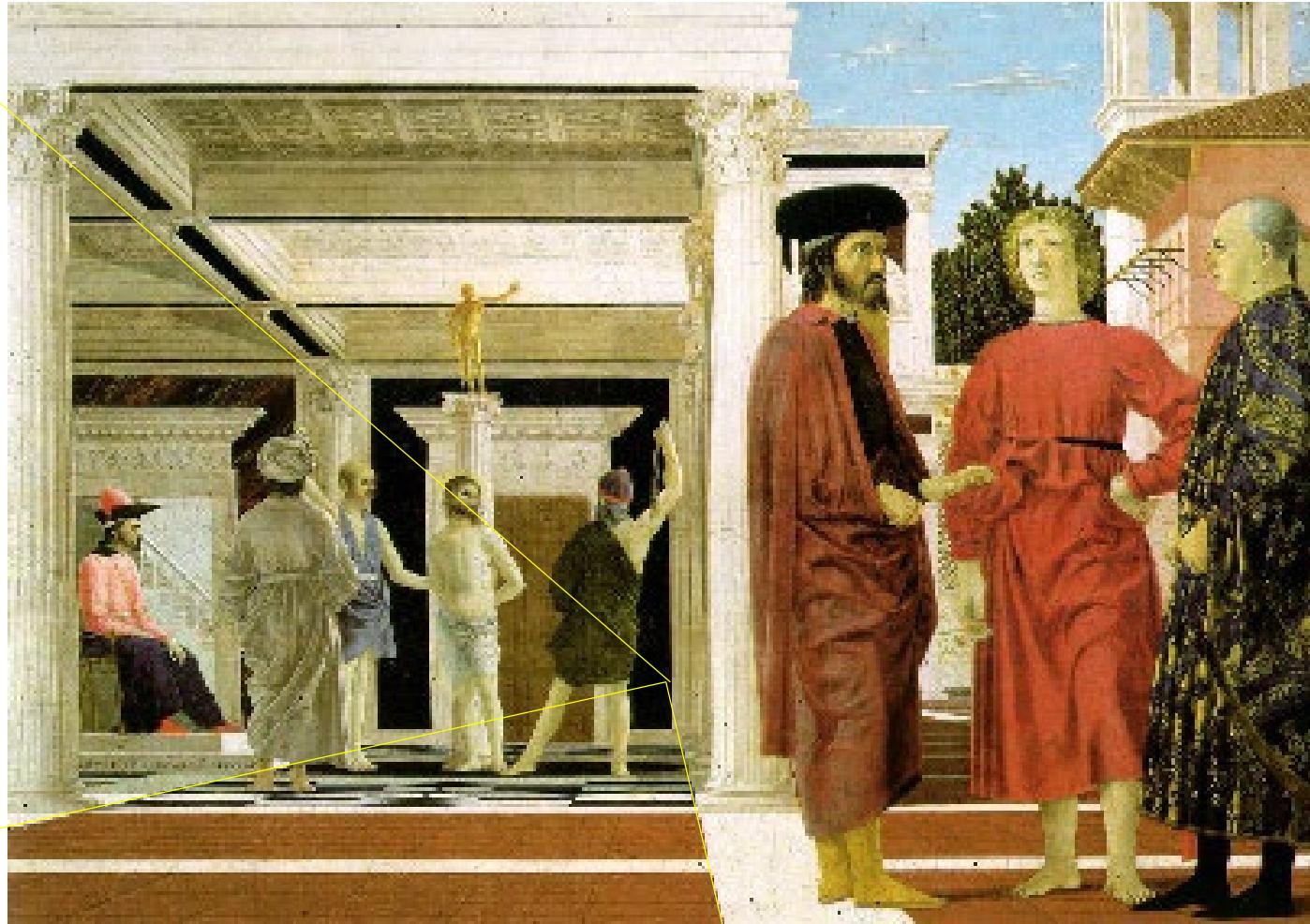
where was the painter's eye (wrt to the painting)?



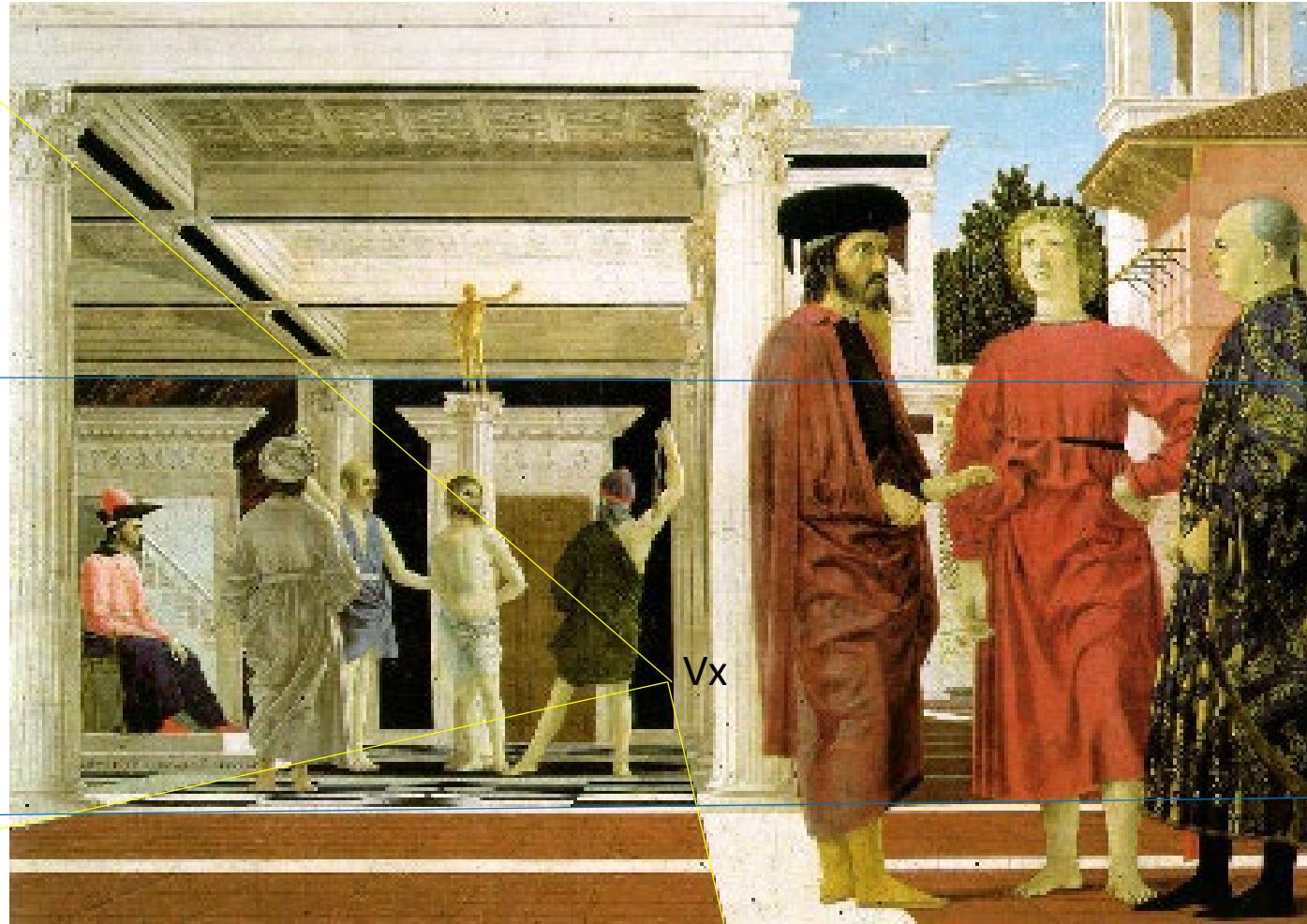
Assumptions:

- the floor is horizontal
- floor tiles are square

Vanishing point of «x» direction: V_x



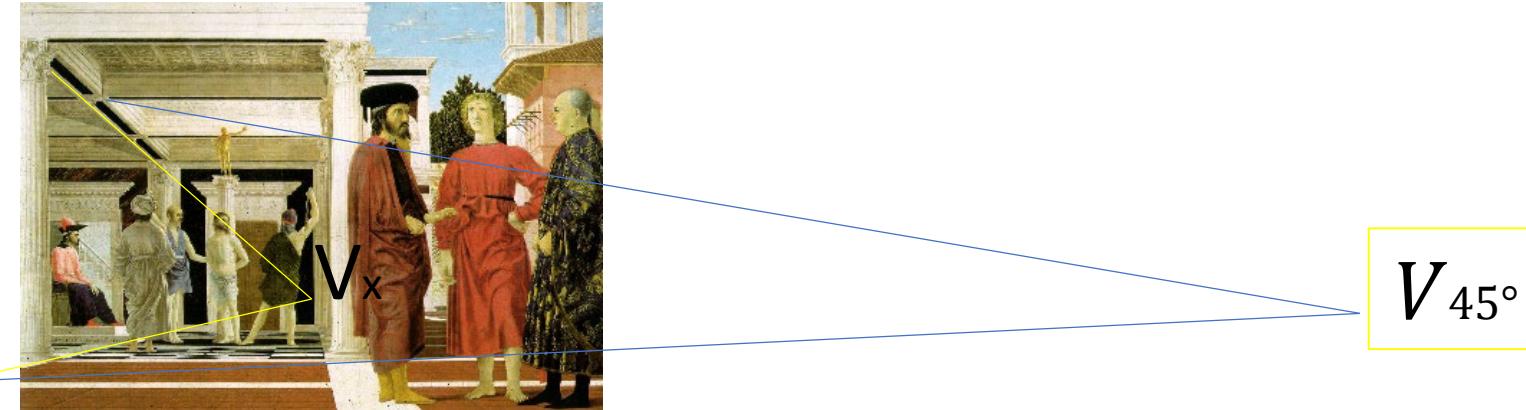
Vanishing point of «y» direction

 V_∞

Theorem: viewing ray associated to the vanishing point of a direction d is parallel to the direction d

- Floor is horizontal \rightarrow line (V_x, eye) is horizontal
- V_y is at the infinity \rightarrow line (eye, V_∞) is parallel to the painting
- Theorem \rightarrow line (V_∞, eye) is orthogonal to line (V_x, eye)
- \rightarrow line (eye, V_x) is horizontal and perpendicular to y axis
- ... eye is on the normal to the y-axis through the point V_x
- How far?

Tiles are square → diagonals form a 45° angle with x-direction



Theorem → line (eye, V_{45°) forms a 45° angle with line (eye, V_x)
floor horizontal → line (eye, V_{45°) is also horizontal



V_x

V_{45°

eye

Camera calibration from

- a planar object of known (reconstructed) shape and
- a vanishing point of the normal to the plane

calibration after reconstruction of a planar face



Shape of the face:
known after metric rectification



How much additional information is needed to calibrate the camera?

- Vanishing point of the direction normal to the reconstructed face



v

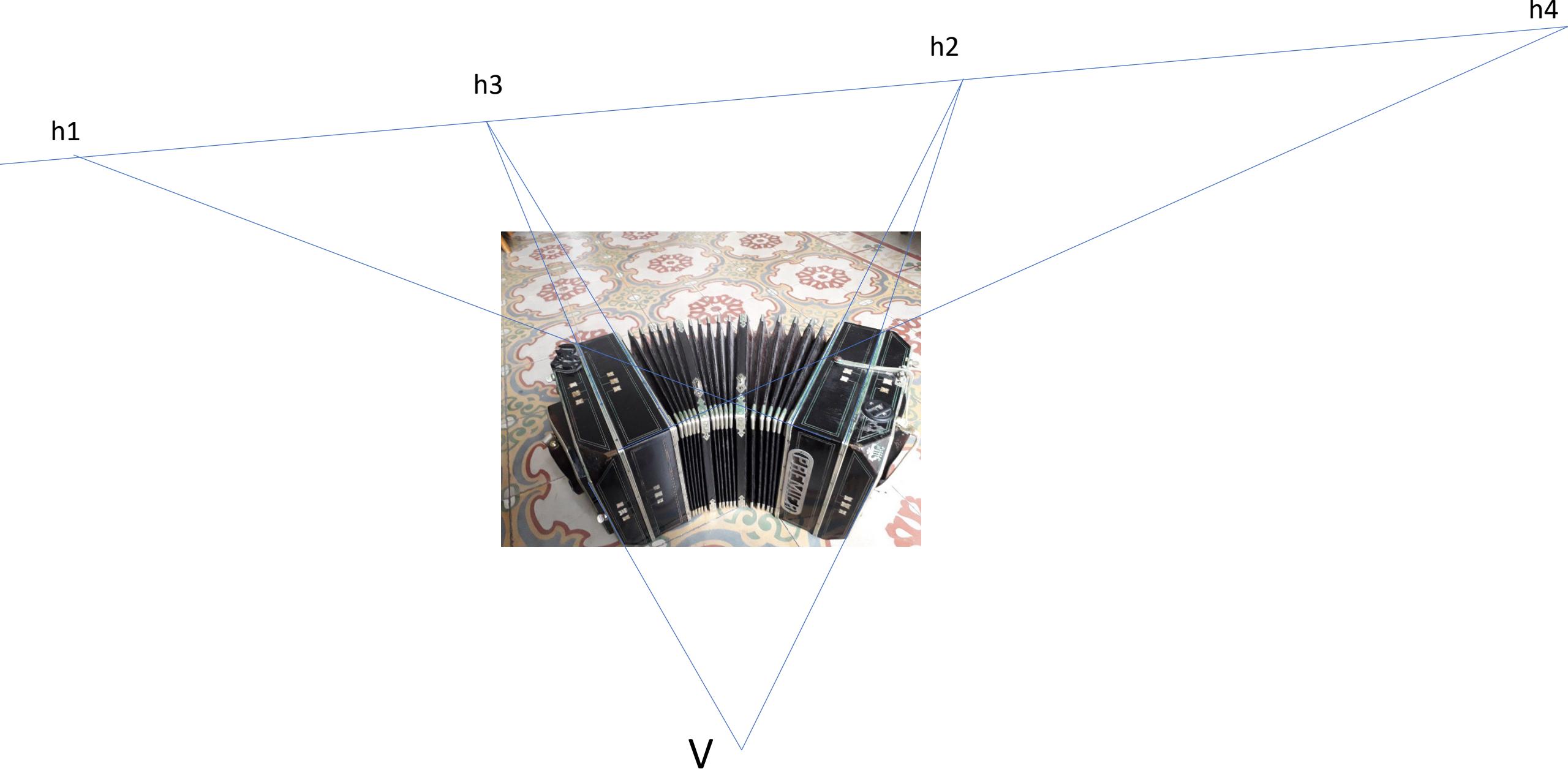
Vanishing point of vertical direction

Camera calibration (assume skew = 0)

- Only four unknowns:

$$\omega = (KK^T)^{-1} = \begin{vmatrix} a^2 & 0 & -u_0 a^2 \\ * & 1 & -v_0 \\ * & * & f_Y^2 + a^2 u_0^2 + v_0^2 \end{vmatrix}$$

→ only four equations are needed



h1

h3

h2

h4

V

Calibration: rectified planar face plus vanishing point of the direction normal to the face

$$\begin{aligned} h_1^T \omega h_2 &= 0 \\ h_3^T \omega h_4 &= 0 \\ v^T \omega h_1 &= 0 \\ v^T \omega h_2 &= 0 \end{aligned}$$

3° and 4° equations are linearly independent,
but there are no further ones (why?)

- → solve for ω
- → find \mathbf{K} (by Cholesky factorisation of $\omega^{-1} = \mathbf{K}\mathbf{K}^T$)

Calibration from rectified face plus orthogonal vanishing point

direct method:

independent of the chosen pairs of
mutually orthogonal vanishing points

from

- the reconstructing homography H_R from given img to rectified img
- the image of the line at the infinity l'_∞
- the vanishing point v along the direction orthogonal to the face

Calibration from rectified face plus orthogonal vanishing points

- from $\mathbf{h}_1^T \omega \mathbf{v} = 0$ and $\mathbf{h}_2^T \omega \mathbf{v} = 0$, and $\mathbf{l}'_\infty = \mathbf{h}_1 \times \mathbf{h}_2 \rightarrow$

$$\mathbf{l}'_\infty = \omega \mathbf{v} \text{ (2 eqns)}$$

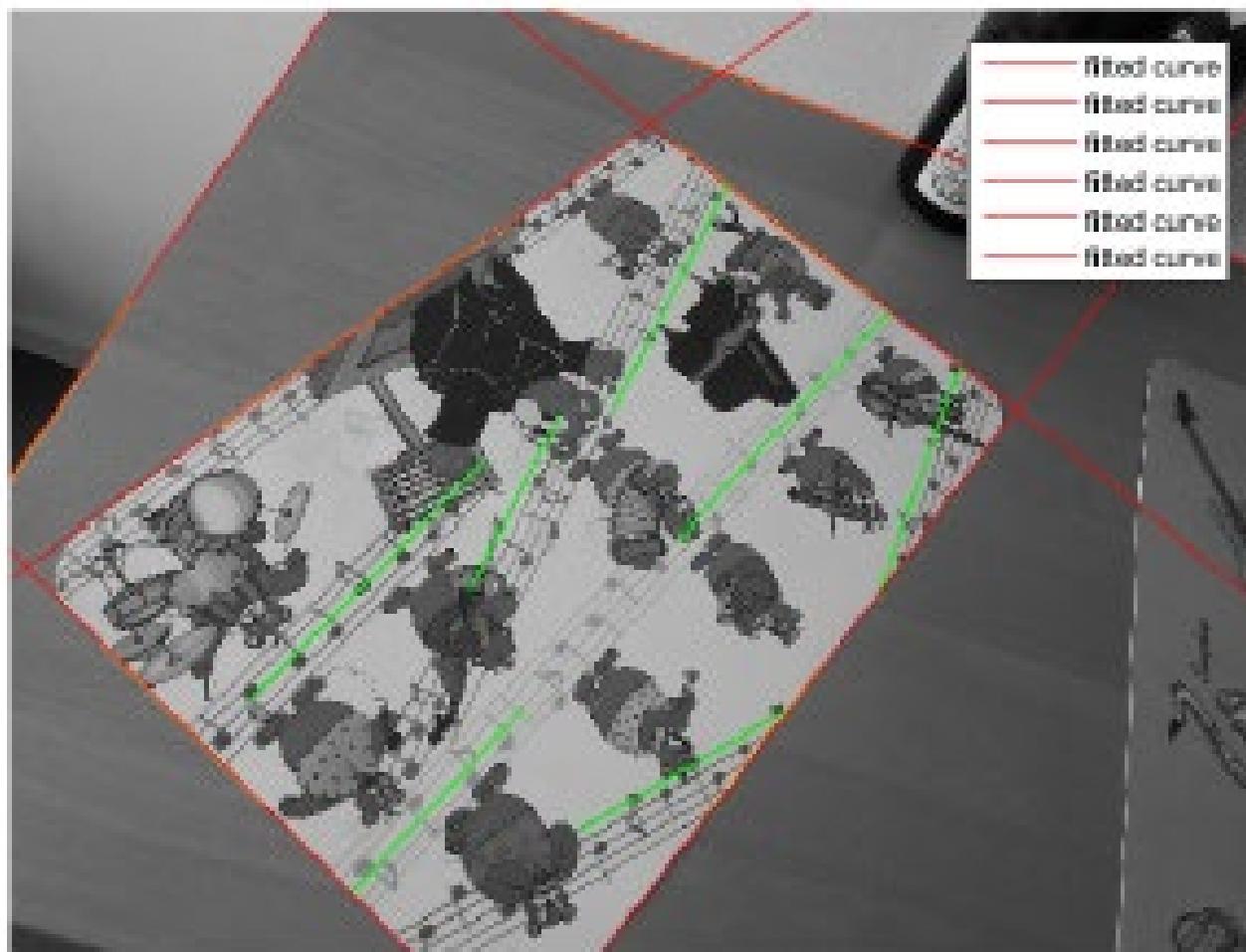
- from $I' = H_R^{-1} I = H_R^{-1} \begin{bmatrix} 1 \\ i \\ 0 \end{bmatrix} = [h_1 \ h_2 \ h_3] \begin{bmatrix} 1 \\ i \\ 0 \end{bmatrix} = h_1 + ih_2$,
and $(h_1 + ih_2)^T \omega (h_1 + ih_2) = 0 \rightarrow$

$$\begin{aligned} h_1^T \omega h_2 &= 0 \\ h_1^T \omega h_1 - h_2^T \omega h_2 &= 0 \end{aligned}$$

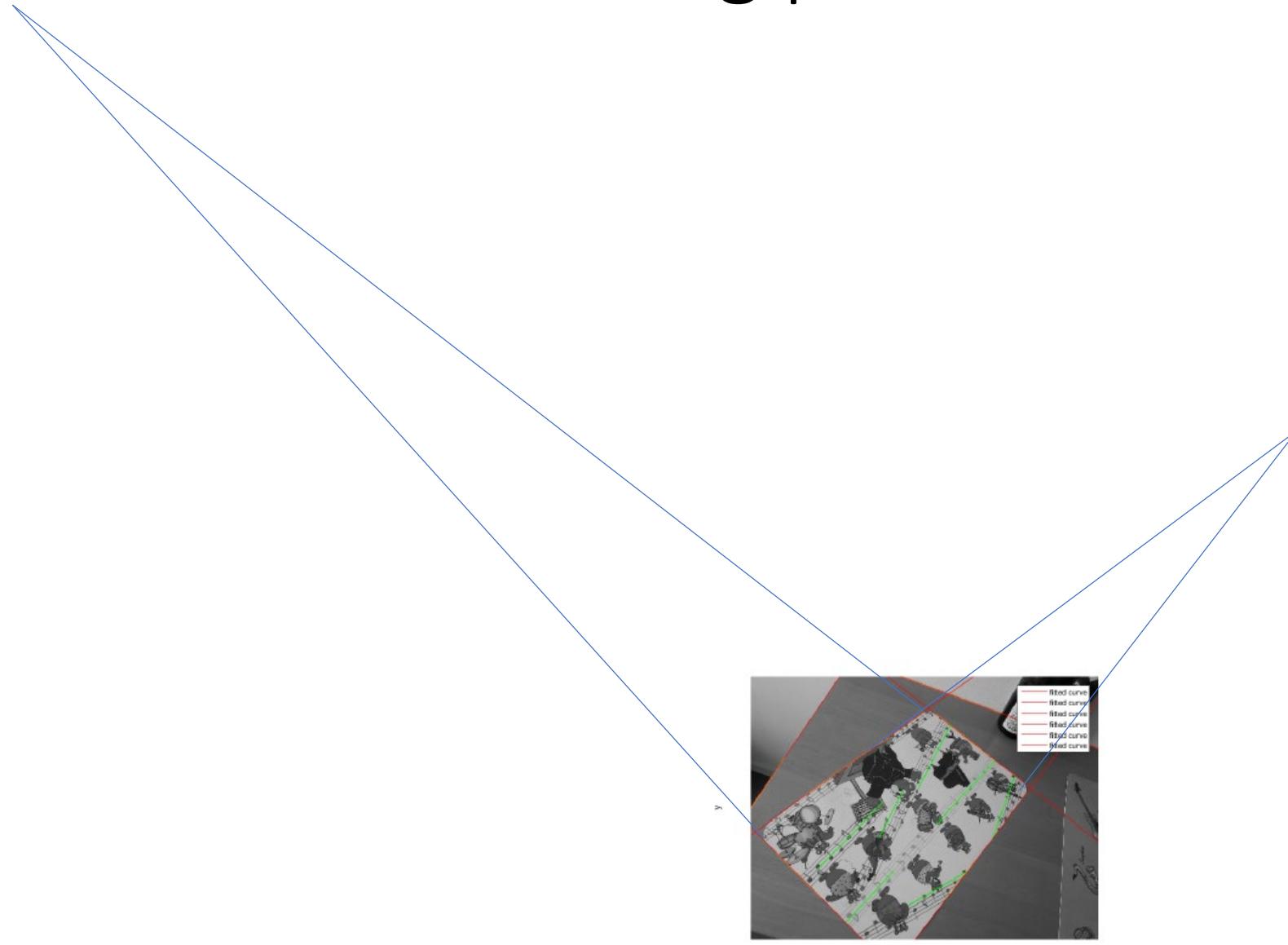
Rectification of a single calibrated
image from vanishing points



Images of pairs of parallel lines



vanishing points



Vanishing line l'_∞

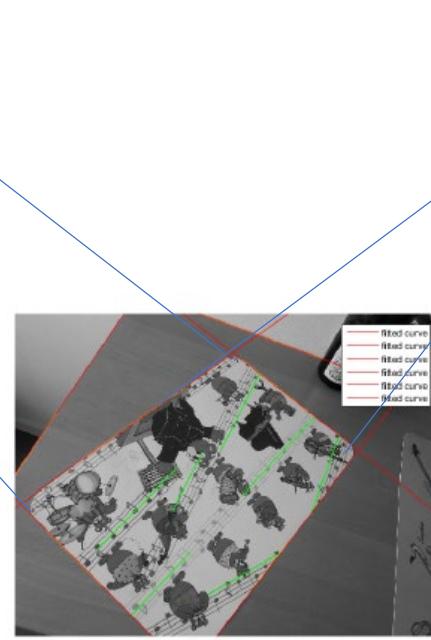


Image of the absolute conic $\omega = (KK^T)^{-1}$

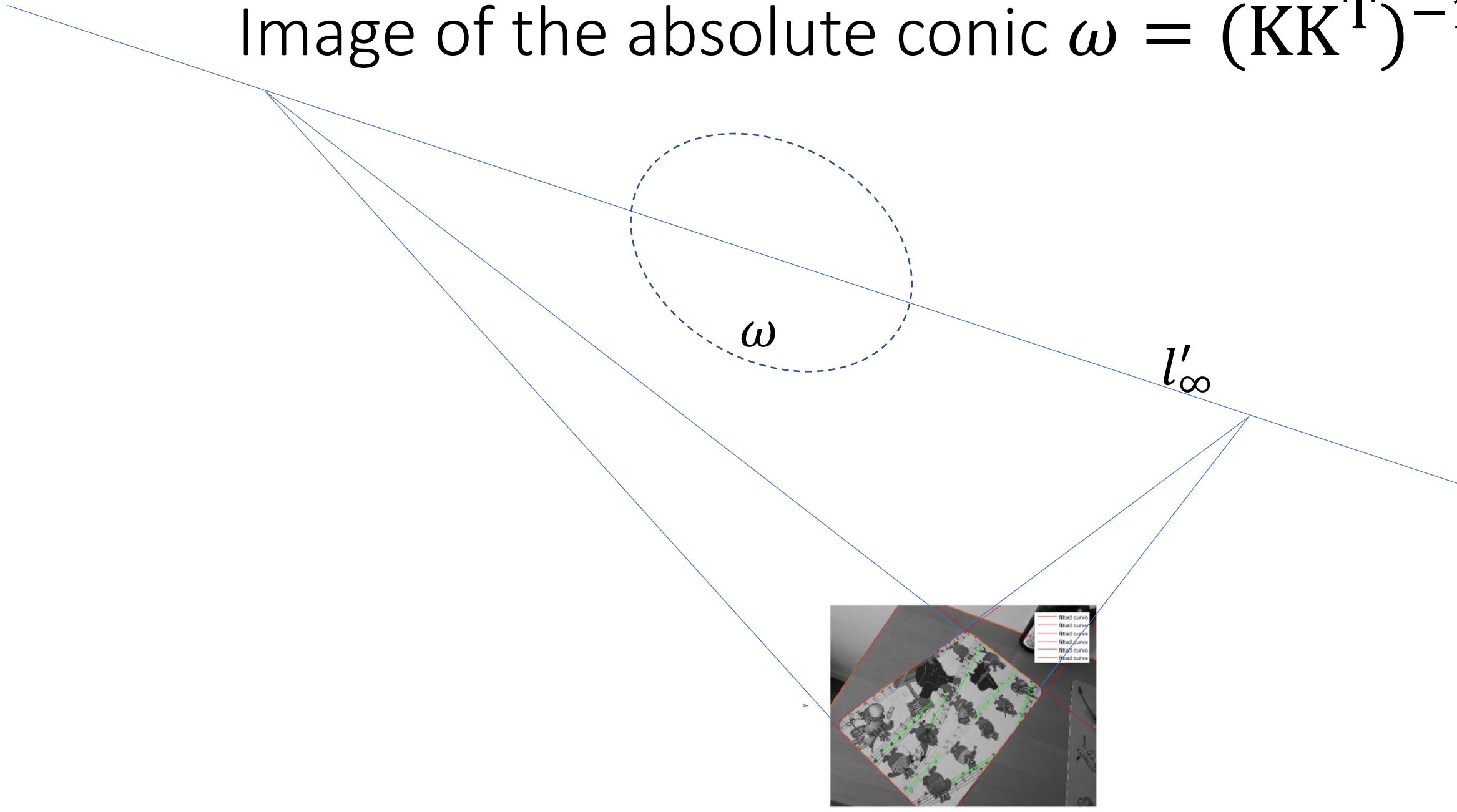


Image of the circular points $l'_\infty \cap \omega$

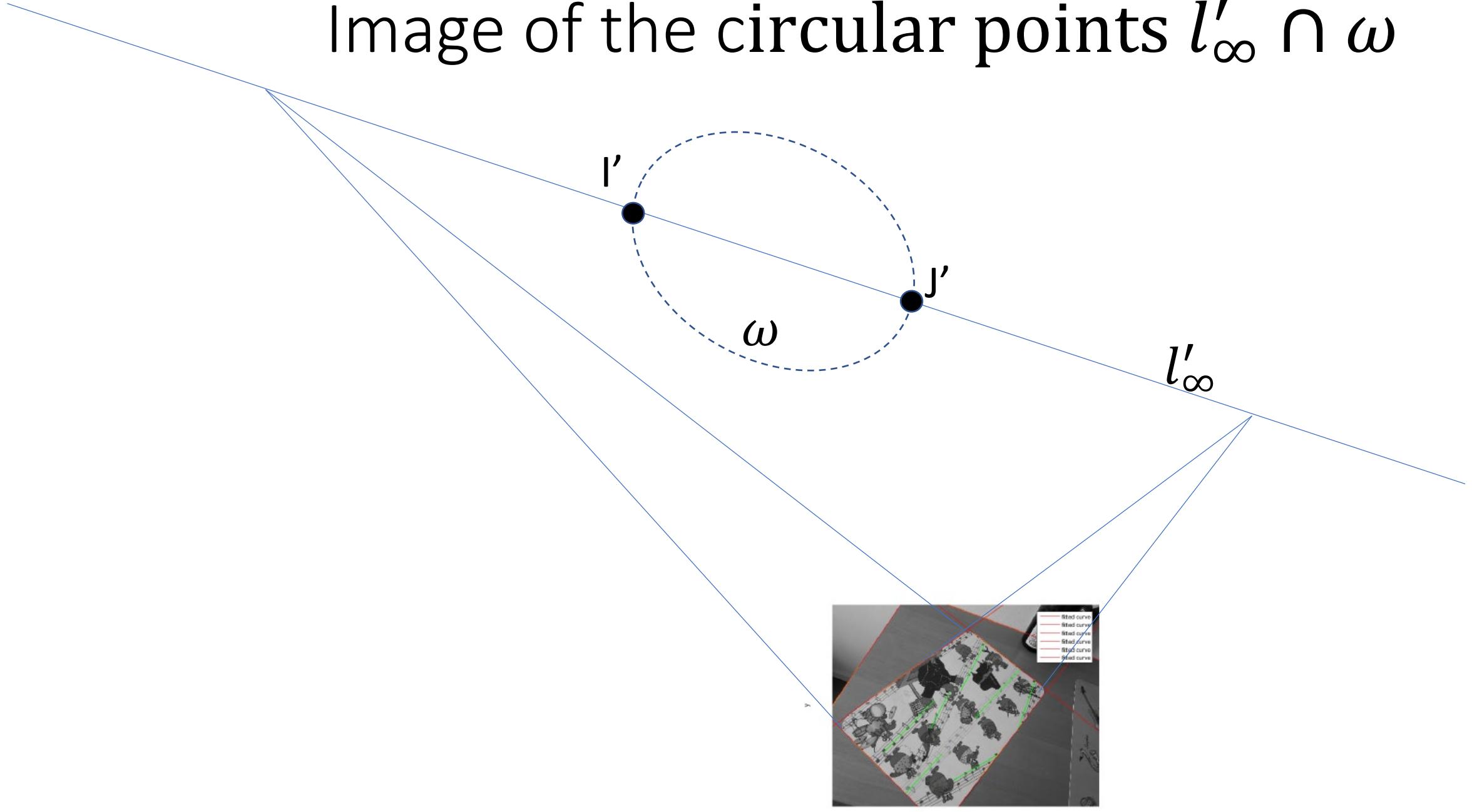
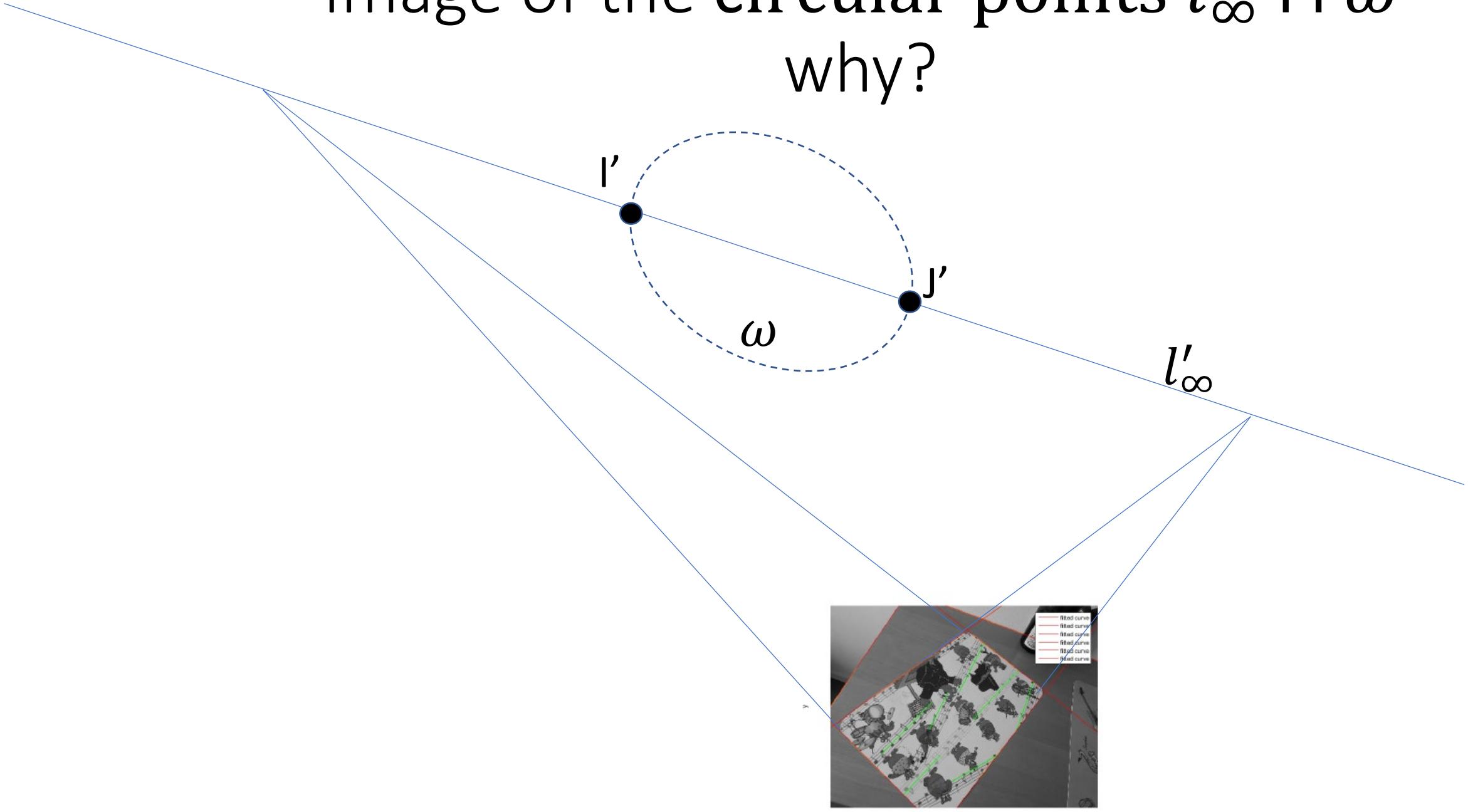
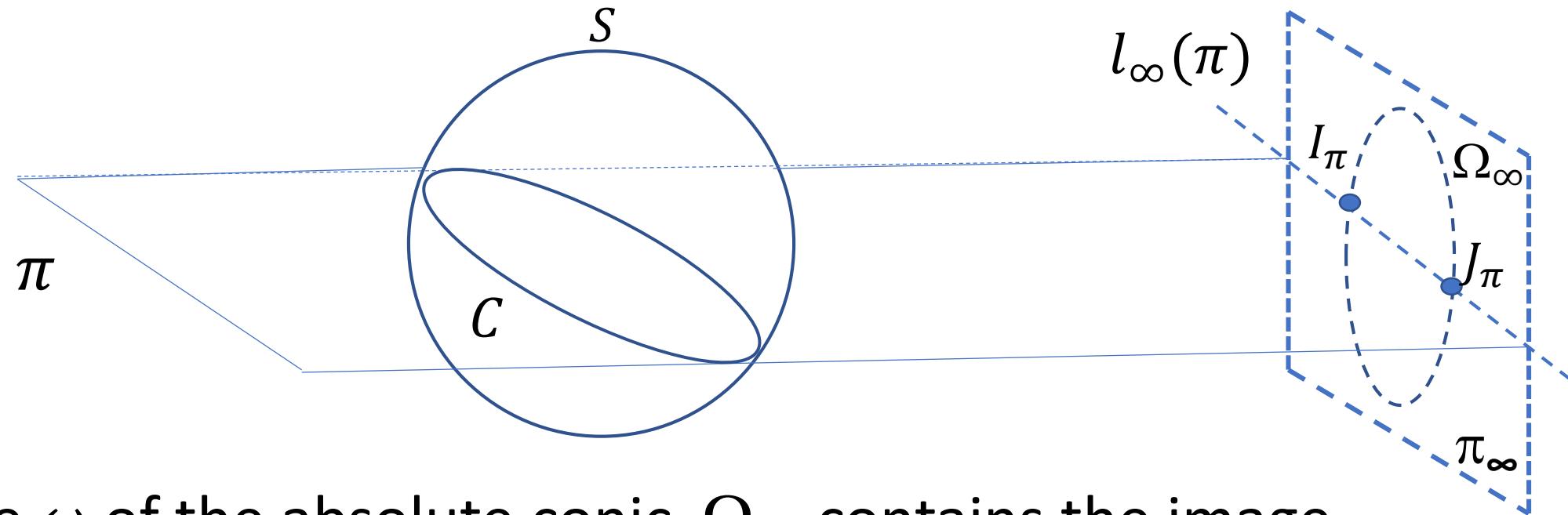


Image of the circular points $l'_\infty \cap \omega$
why?



remember: IAC ω contain images of circular points

1. The absolute conic Ω_∞ contains all the circular points (i.e. the circular points I_π, J_π of any plane π belong to absolute conic Ω_∞) !!



2. The image ω of the absolute conic Ω_∞ contains the image I'_π, J'_π of all the circular points I_π, J_π

reconstruction of the planar scene

image of circular points: $\{l', J'\} = l'_\infty \cap \omega$

- Image of the circular points \rightarrow Image of the conic dual to the circular points

$$C_\infty^{*'} = I'J'^T + J'I'^T$$

- Singular value decomposition

$$\text{svd}(C_\infty^{*'}) = U \begin{vmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{vmatrix} U^T = H_{rect}^{-1} C_\infty^* H_{rect}^{-T}$$

- Rectifying homography (from svd output U)

$$H_{rect} = U^T$$

In practice ...

- Image of the circular points → Image of the conic dual to the circular points

$$C_{\infty}' = I'J'^T + J'I'^T$$

- Singular value decomposition

$$\text{svd}(C_{\infty}') = U \begin{bmatrix} s_1 & 0 & 0 \\ 0 & s_2 & 0 \\ 0 & 0 & \varepsilon \end{bmatrix} U^T$$

- Rectifying homography (from svd output U)

$$H_{rect} = (U \begin{bmatrix} \sqrt{s_1} & 0 & 0 \\ 0 & \sqrt{s_2} & 0 \\ 0 & 0 & 1 \end{bmatrix})^{-1} = \begin{bmatrix} \sqrt{s_1^{-1}} & 0 & 0 \\ 0 & \sqrt{s_2^{-1}} & 0 \\ 0 & 0 & 1 \end{bmatrix} U^T$$

Euclidean reconstruction = Image rectification



Localization of known
(e.g. reconstructed) planar objects
from a single calibrated image

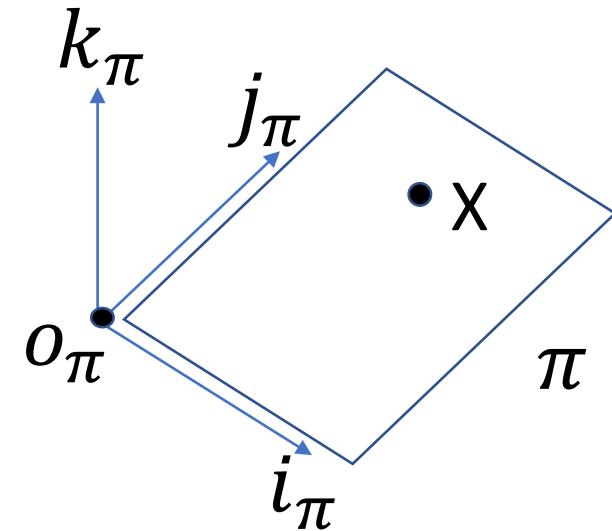
Image of the upper face



known shape of the upper face
(e.g. after reconstruction)

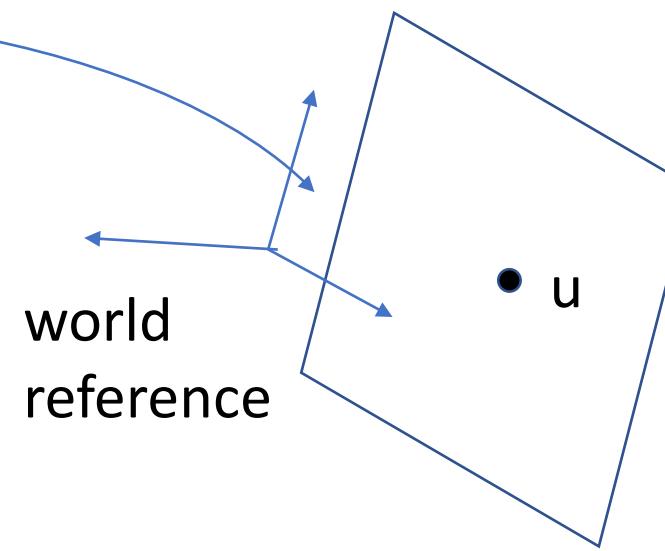


known planar object



object reference

H



camera reference

world
reference

world reference \equiv camera reference

$$\mathbf{x}_\pi = \begin{bmatrix} x \\ y \\ 0 \\ w \end{bmatrix}$$

$$\mathbf{x}_w = \begin{bmatrix} i_\pi & j_\pi & k_\pi & o_\pi \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 0 \\ w \end{bmatrix}$$

$$\mathbf{X}_w = \begin{bmatrix} \mathbf{i}_\pi & \mathbf{j}_\pi & \mathbf{k}_\pi & \mathbf{o}_\pi \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 0 \\ w \end{bmatrix} = \begin{bmatrix} \mathbf{i}_\pi & \mathbf{j}_\pi & \mathbf{o}_\pi \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ w \end{bmatrix} = \begin{bmatrix} \mathbf{i}_\pi & \mathbf{j}_\pi & \mathbf{o}_\pi \\ 0 & 0 & 1 \end{bmatrix} \mathbf{x}_\pi$$

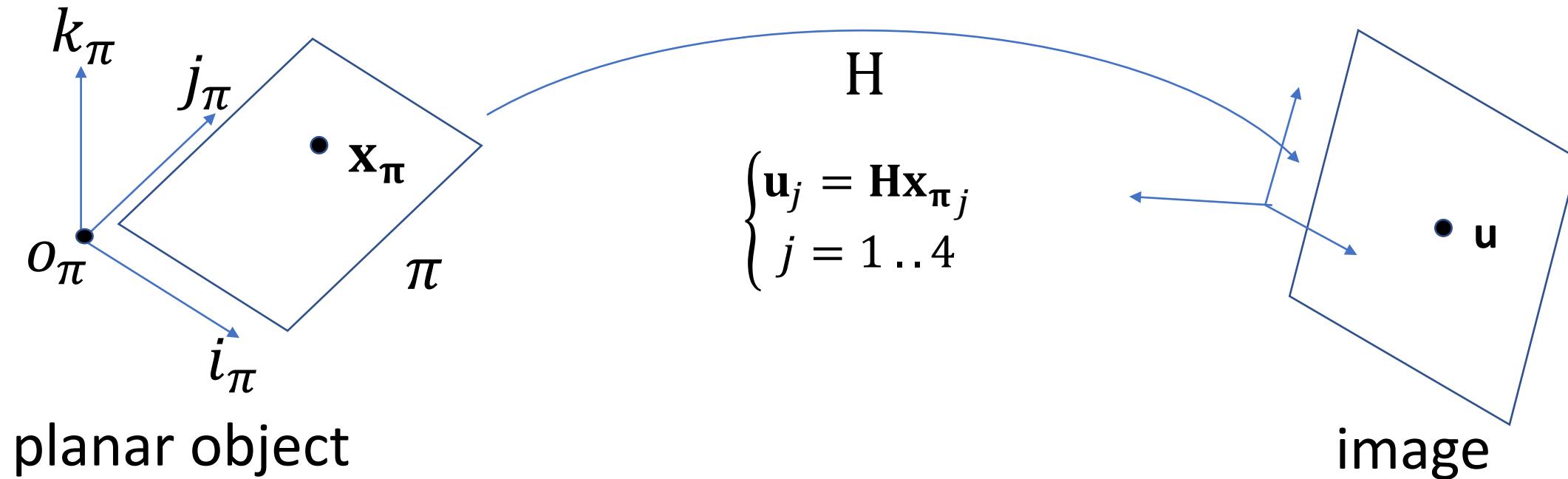
world reference on camera reference: $\mathbf{R} = \mathbf{I}_3, \mathbf{t} = \mathbf{0}$

$$\mathbf{P} = [\mathbf{KR} \quad \mathbf{KRt}] = [\mathbf{K} \quad \mathbf{0}]$$

$$\mathbf{u} = \mathbf{PX}_w = [\mathbf{K} \quad \mathbf{0}] \begin{bmatrix} \mathbf{i}_\pi & \mathbf{j}_\pi & \mathbf{o}_\pi \\ 0 & 0 & 1 \end{bmatrix} \mathbf{x}_\pi = \mathbf{K}[\mathbf{i}_\pi \quad \mathbf{j}_\pi \quad \mathbf{o}_\pi] \mathbf{x}_\pi$$

plane π to image homography $\mathbf{H} = \mathbf{K}[\mathbf{i}_\pi \quad \mathbf{j}_\pi \quad \mathbf{o}_\pi]$

\mathbf{H} known from image and object model
(4 pairs of corresponding points)



pose of the planar object relative to the camera

$$[\mathbf{i}_\pi \quad \mathbf{j}_\pi \quad \mathbf{o}_\pi] = \mathbf{K}^{-1}\mathbf{H}$$

$$(\mathbf{k}_\pi = \mathbf{i}_\pi \times \mathbf{j}_\pi)$$

Localization of a reconstructed face wrt camera

Localization from homography H
between the planar face and its image

Localization localization of the upper face

Planar object (upper planar face) wrt camera

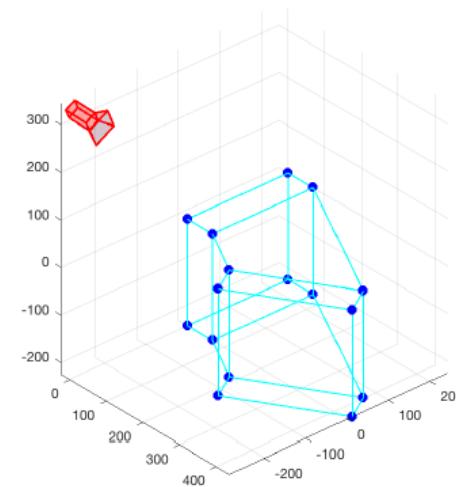
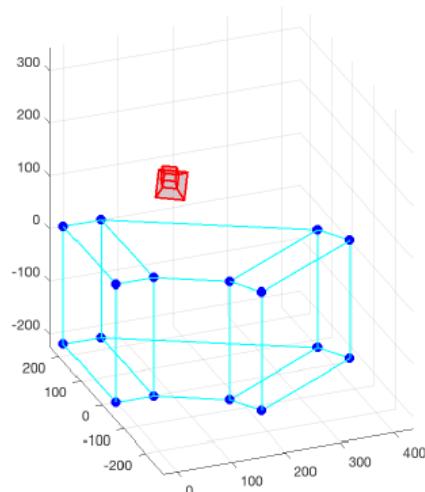
$$[i_\pi \quad j_\pi \quad o_\pi] = \mathbf{K}^{-1} \mathbf{H}$$

where \mathbf{K} is the calibration matrix of the camera,
while

\mathbf{H} is the homography from the planar face to its image

Localize camera wrt upper planar face:

$$\begin{bmatrix} i_\pi & j_\pi & i_\pi \times j_\pi & 0_\pi \\ 0 & 0 & 0 & 1 \end{bmatrix}^{-1}$$



remember

$$\begin{aligned} \mathbf{u} &= \mathbf{K}[\mathbf{R} \quad \mathbf{t}] \mathbf{X}_{world} = [\mathbf{KR} \quad \mathbf{Kt}] \mathbf{X}_{world} \\ \mathbf{u} &= \mathbf{P} \mathbf{X}_{world} = [\mathbf{M} \quad \mathbf{m}] \mathbf{X}_{world} \end{aligned}$$

$$\rightarrow \mathbf{M} = \mathbf{KR} \text{ and } \mathbf{m} = \mathbf{Kt}$$

$$\mathbf{M} = \mathbf{KR}_{cam \rightarrow world}$$

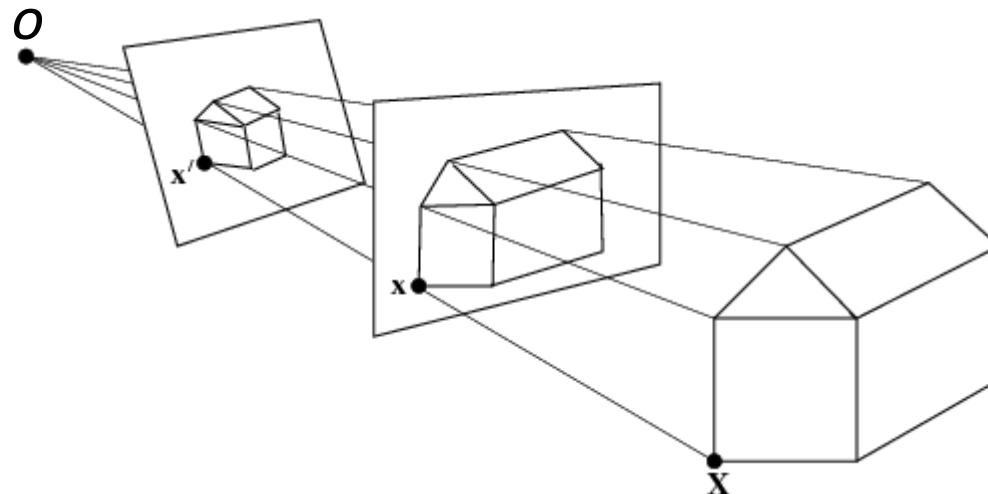
given \mathbf{M} , find \mathbf{K} and \mathbf{R} by Q-R matrix decomposition of $\text{inv}(\mathbf{M})$

$$\text{from } \mathbf{m} = -\mathbf{Mo} \rightarrow \mathbf{u} = [\mathbf{M} \quad \mathbf{m}] \mathbf{X}_{world} = \mathbf{M}[\mathbf{I} \quad -\mathbf{o}] \mathbf{X}_{world}$$

$$\mathbf{u} = \mathbf{KR}[\mathbf{I} \quad -\mathbf{o}] \mathbf{X}_{world}$$

$$\rightarrow \mathbf{P} = \mathbf{KR}[\mathbf{I} \quad -\mathbf{o}]$$

Image transformations with fixed camera center



$$P = KR[I] - o \quad , \quad P' = K'R'[I] - o$$

$$P' = K'R'(KR)^{-1}P$$

$$x' = P'X = K'R'(KR)^{-1}PX = K'R'(KR)^{-1}x$$

$$x' = Hx \text{ with } H = K'R'(KR)^{-1} = K'R'R^{-1}K^{-1}$$

Translating the image plane: zooming (assuming fixed principal point)

$$x' = Hx = K'R'R^{-1}K^{-1}x \text{ but, since } R' = R \\ x' = K'(K)^{-1}x$$

$$H = K'(K)^{-1} = \begin{bmatrix} kI & (1-k)\tilde{x}_0 \\ 0^T & 1 \end{bmatrix} \quad k = f / f'$$

$$K' = \begin{bmatrix} kI & (1 - k)\tilde{x}_0 \\ 0^T & 1 \end{bmatrix} K$$
$$\xrightarrow{\quad\quad\quad}$$
$$K' = \begin{bmatrix} kI & (1 - k)\tilde{x}_0 \\ 0^T & 1 \end{bmatrix} \begin{bmatrix} A & \tilde{x}_0 \\ 0^T & 1 \end{bmatrix} = \begin{bmatrix} kA & \tilde{x}_0 \\ 0^T & 1 \end{bmatrix} = K \begin{bmatrix} kI & 0 \\ 0^T & 1 \end{bmatrix}$$

Camera rotation around camera center

$$x = K[I|0] X$$

$$x' = K'[R|0] X = K'R [I|0] X = K' R K^{-1} K[I|0] X = K' R K^{-1} x$$

$$H = K' R K^{-1}$$



Camera rotation around camera center with fixed intrinsic parameters

A camera is rotated about its centre with no change in the internal parameters.. Algebraically, if x, x' are the images of a point X before and after the pure rotation

$$x = K [I | 0] X$$

$$x' = K [R | 0] X = KR [I | 0] X = KRK^{-1} K[I | 0] X = KRK^{-1} x$$

so that $x' = H x$ with $H = KRK^{-1}$: conjugate rotation

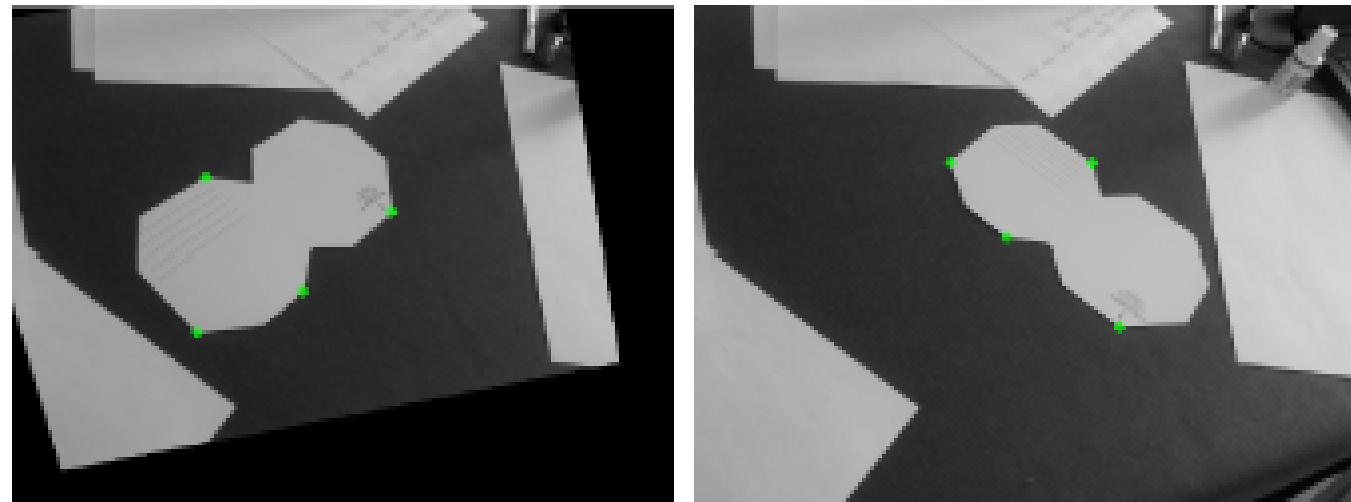
eigenvalues of $H = KRK^{-1}$: $\{\mu, \mu e^{i\theta}, \mu e^{-i\theta}\}$ = the same eigenvalues of R

where μ is an unknown scale factor (if H scaled such that $\det H = 1$, then $\mu = 1$). Hence, the rotation angle between views may be computed directly from the phase of the complex eigenvalues of H .

Similarly, it can be shown (exercise) that the eigenvector of H corresponding to the real eigenvalue is the vanishing point of the rotation axis.

Reminds to planar motion

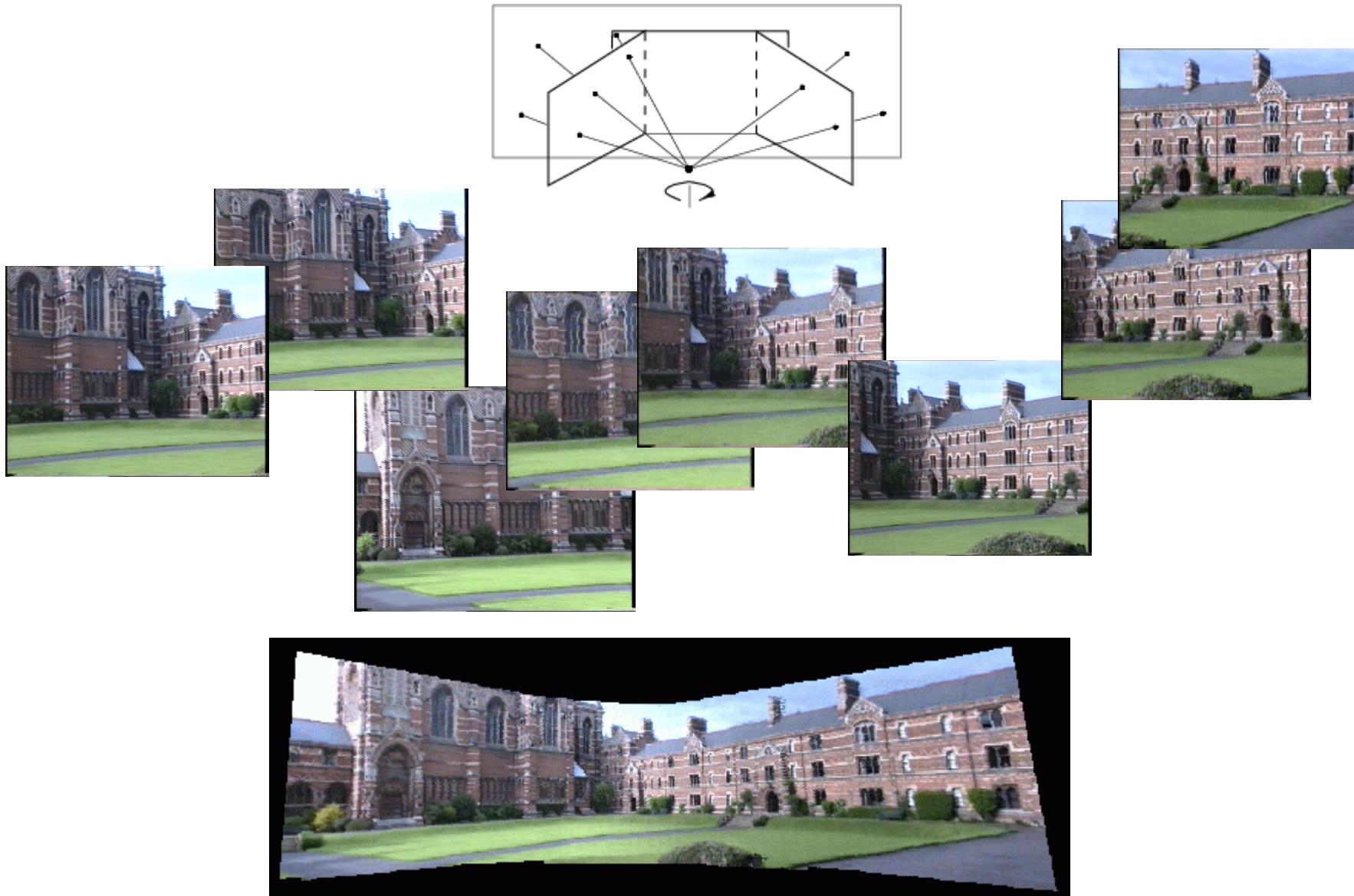
- find eigenvector-eigenvalues of H :
- eigenvalues are proportional to $\lambda' = 1, \lambda'' = e^{i\theta}, \lambda''' = e^{-i\theta}$
- eigenvector e' associated to $\lambda' = 1$ is the image of the C.O.R. O
- angle θ is the rotation angle



Common aspects and differences between them

- in planar motion, rotation does not need to be about the camera center
- in planar motion, rotation axis is perpendicular to the plane

Planar homography mosaicing



Planar homography mosaicing

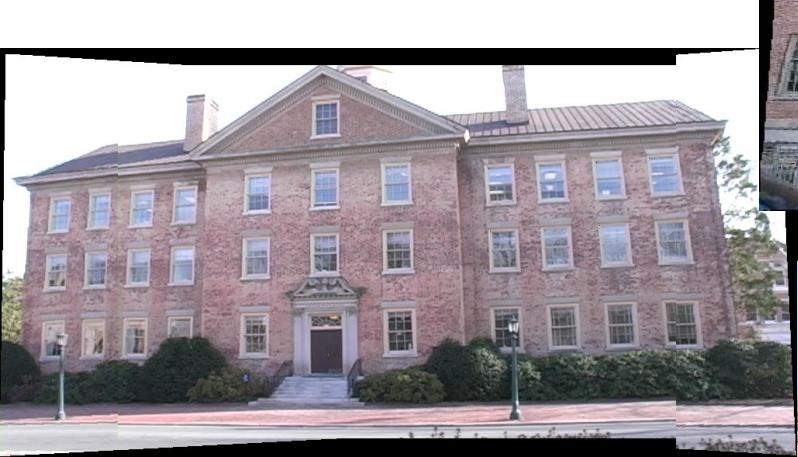
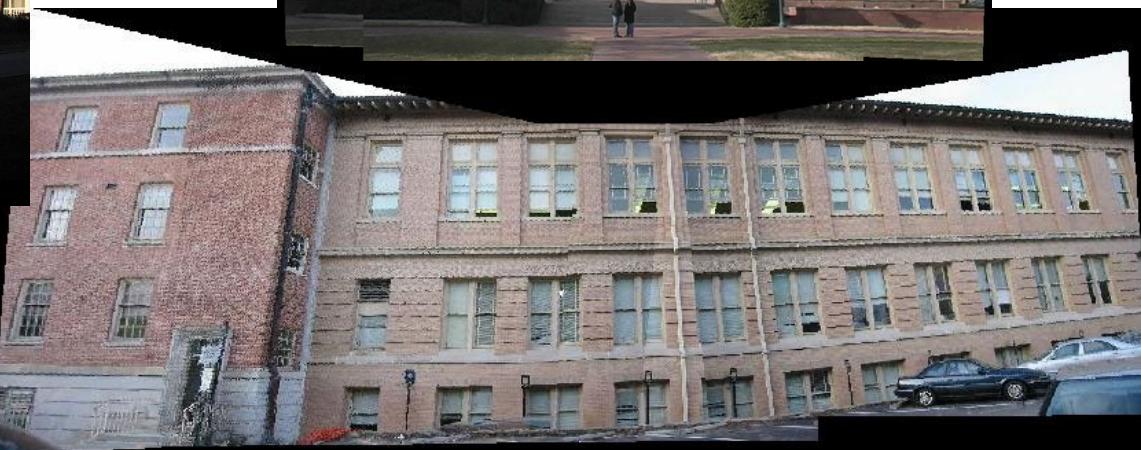
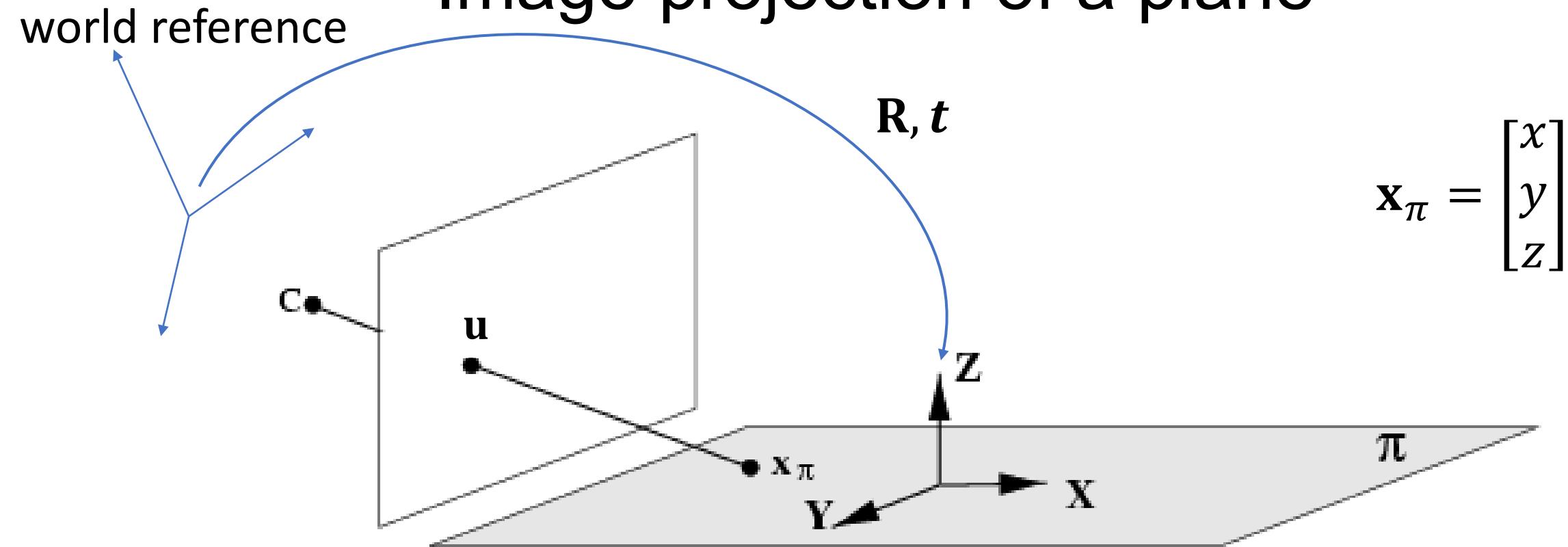


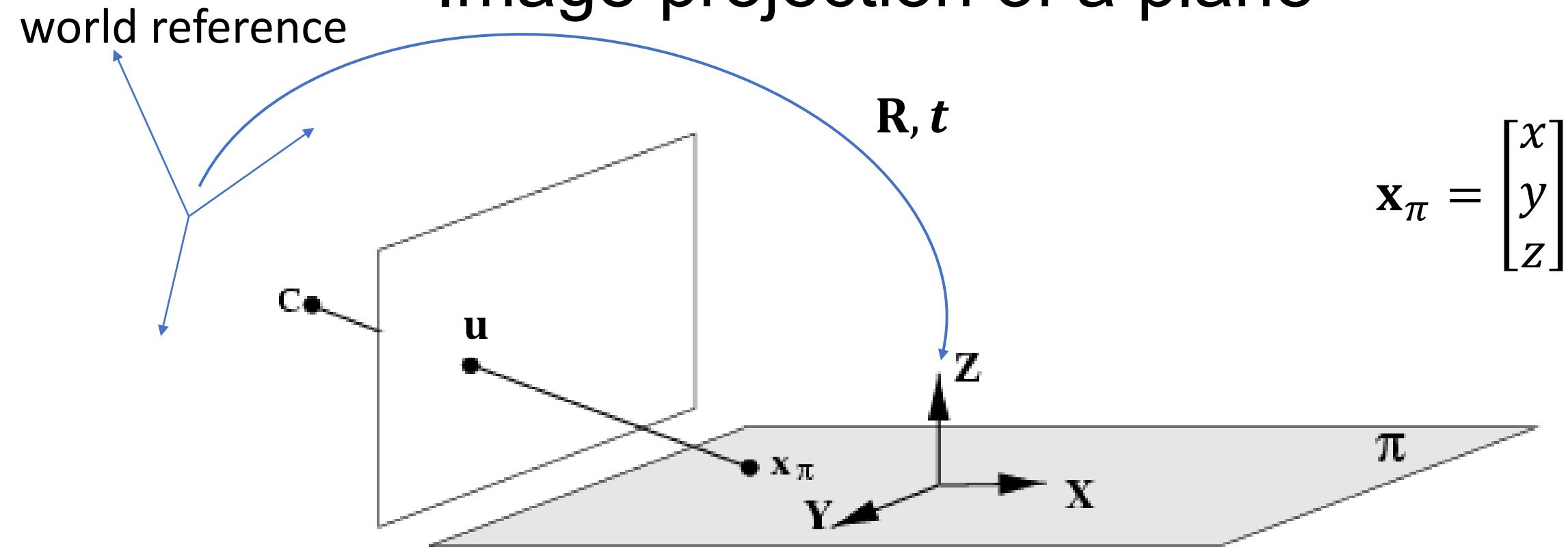
Image: projection and back-projection

Image projection of a plane



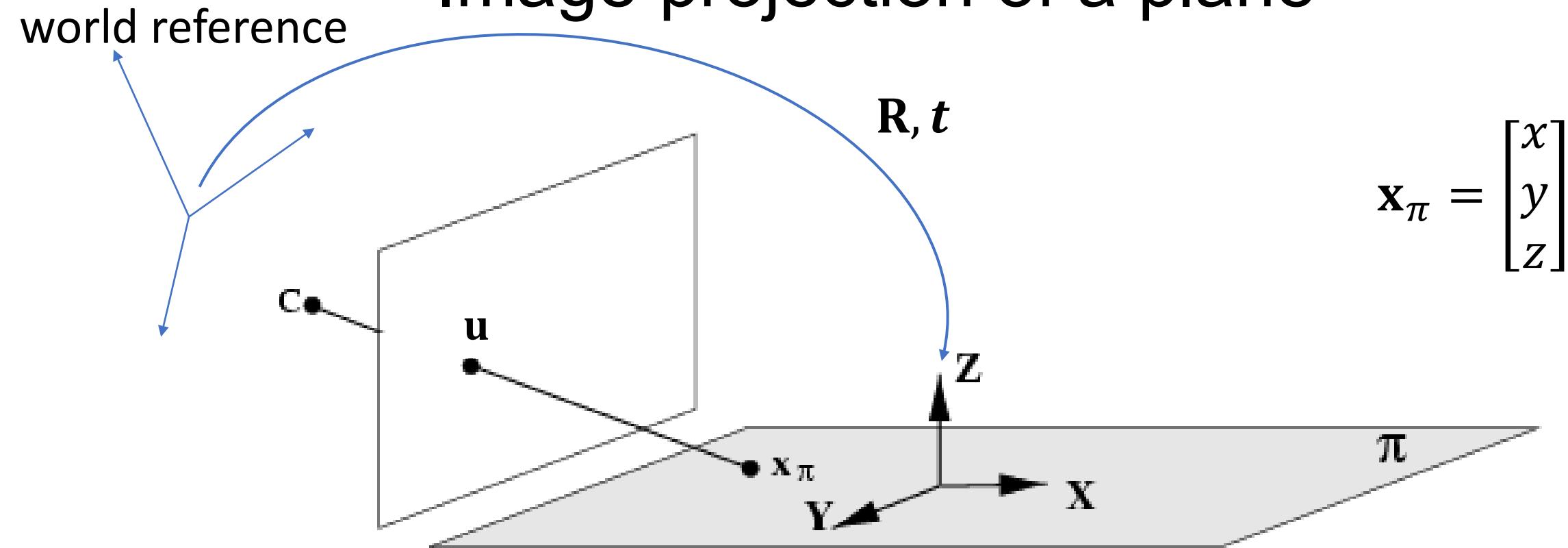
$$X_w = \begin{bmatrix} \mathbf{R} & \mathbf{t} \\ 0 & 1 \end{bmatrix} X_\pi$$

Image projection of a plane



$$\mathbf{u} = \mathbf{P}X_w = [\mathbf{M} \quad \mathbf{m}]X_w = [\mathbf{M} \quad \mathbf{m}] \begin{bmatrix} \mathbf{R} & \mathbf{t} \\ 0 & 1 \end{bmatrix} X_\pi = [\mathbf{M}\mathbf{R} \quad \mathbf{M}\mathbf{t} + \mathbf{m}]X_\pi$$

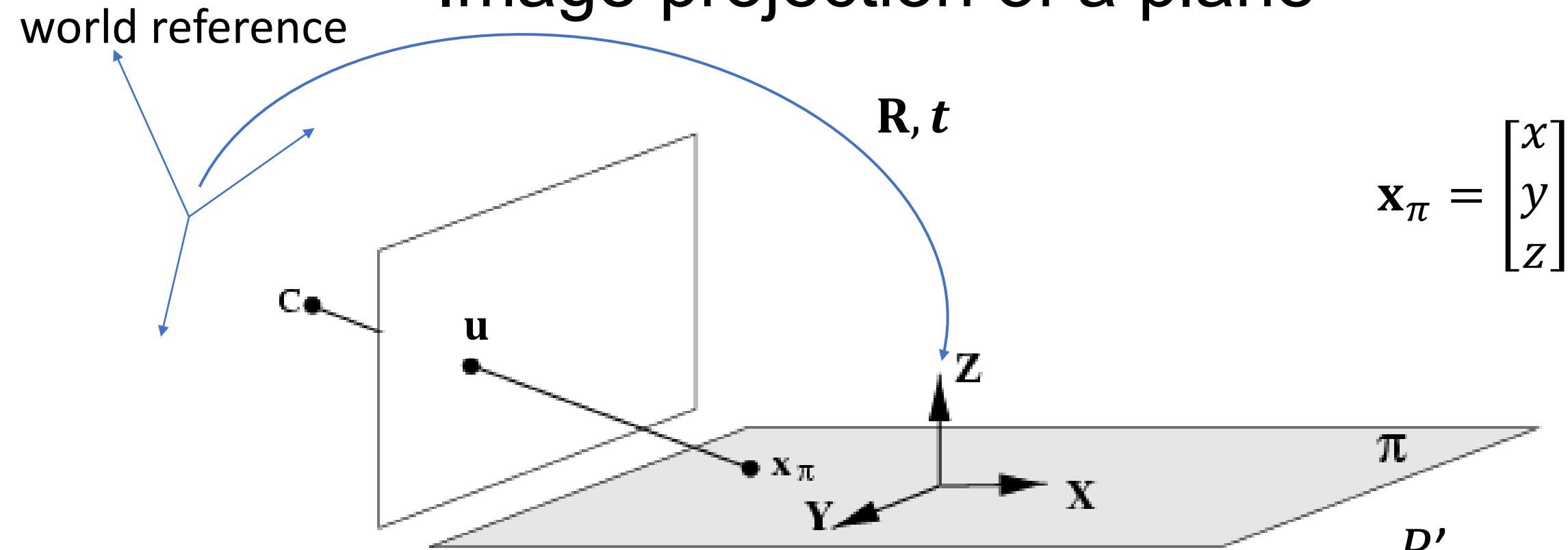
Image projection of a plane



$$\mathbf{x}_\pi = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

$$\mathbf{u} = \mathbf{P}X_w = [\mathbf{M} \quad \mathbf{m}]X_w = [\mathbf{M} \quad \mathbf{m}] \begin{bmatrix} \mathbf{R} & \mathbf{t} \\ 0 & 1 \end{bmatrix} X_\pi = \underbrace{[\mathbf{M}\mathbf{R} \quad \mathbf{M}\mathbf{t} + \mathbf{m}]}_{\mathbf{P}'} \begin{bmatrix} x \\ y \\ 0 \\ w \end{bmatrix}$$

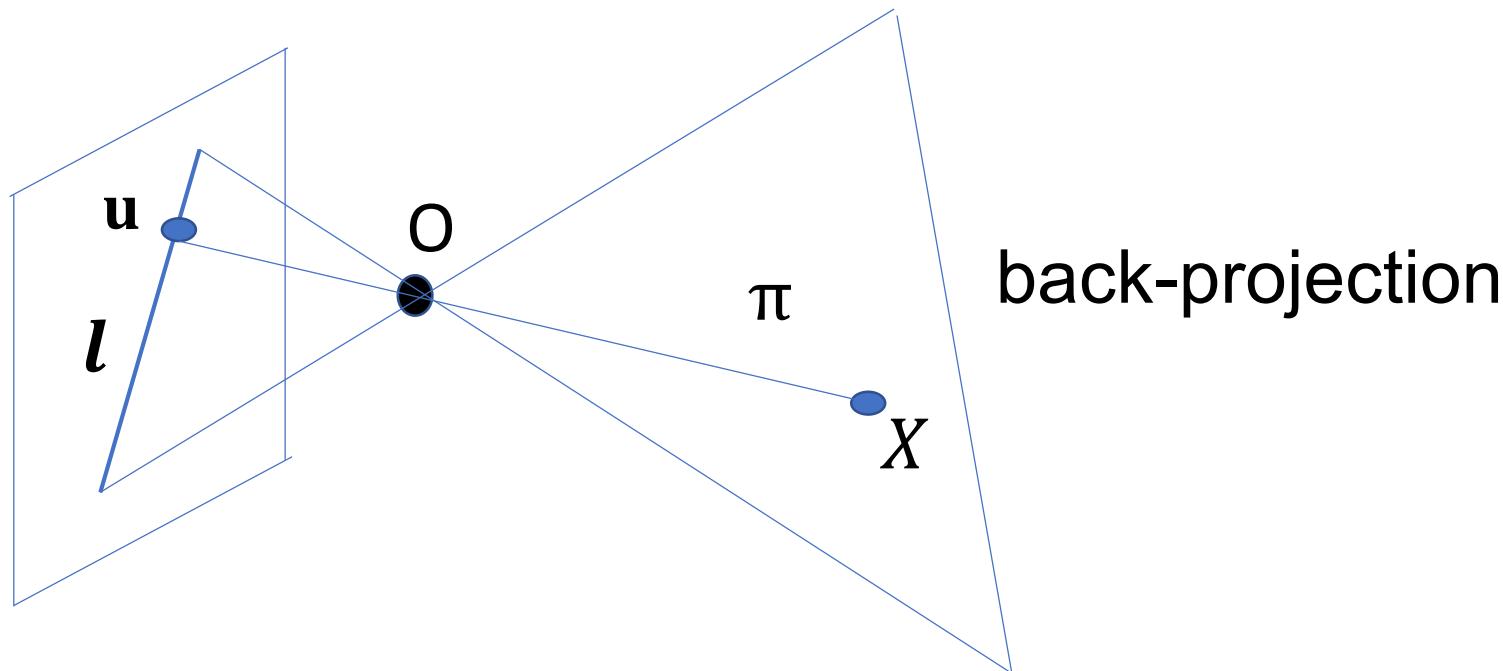
Image projection of a plane



$$\begin{aligned}
 \mathbf{u} &= P\mathbf{X}_w = [\mathbf{M} \quad \mathbf{m}] \mathbf{X}_w = [\mathbf{M} \quad \mathbf{m}] \begin{bmatrix} \mathbf{R} & \mathbf{t} \\ \mathbf{0} & 1 \end{bmatrix} \mathbf{x}_\pi = [\mathbf{MR} \quad \mathbf{Mt} + \mathbf{m}] \begin{bmatrix} \mathbf{x} \\ \mathbf{y} \\ 0 \\ \mathbf{w} \end{bmatrix} \\
 &= [p'_1 \quad p'_2 \quad p'_3 \quad p'_4] [x \quad y \quad 0 \quad w]^T = [p'_1 \quad p'_2 \quad p'_4] \mathbf{x}_\pi
 \end{aligned}$$

Back-projection of an image line

set of space points X , whose image projection $\mathbf{u} = \mathbf{P}X$ is on image line \mathbf{l}



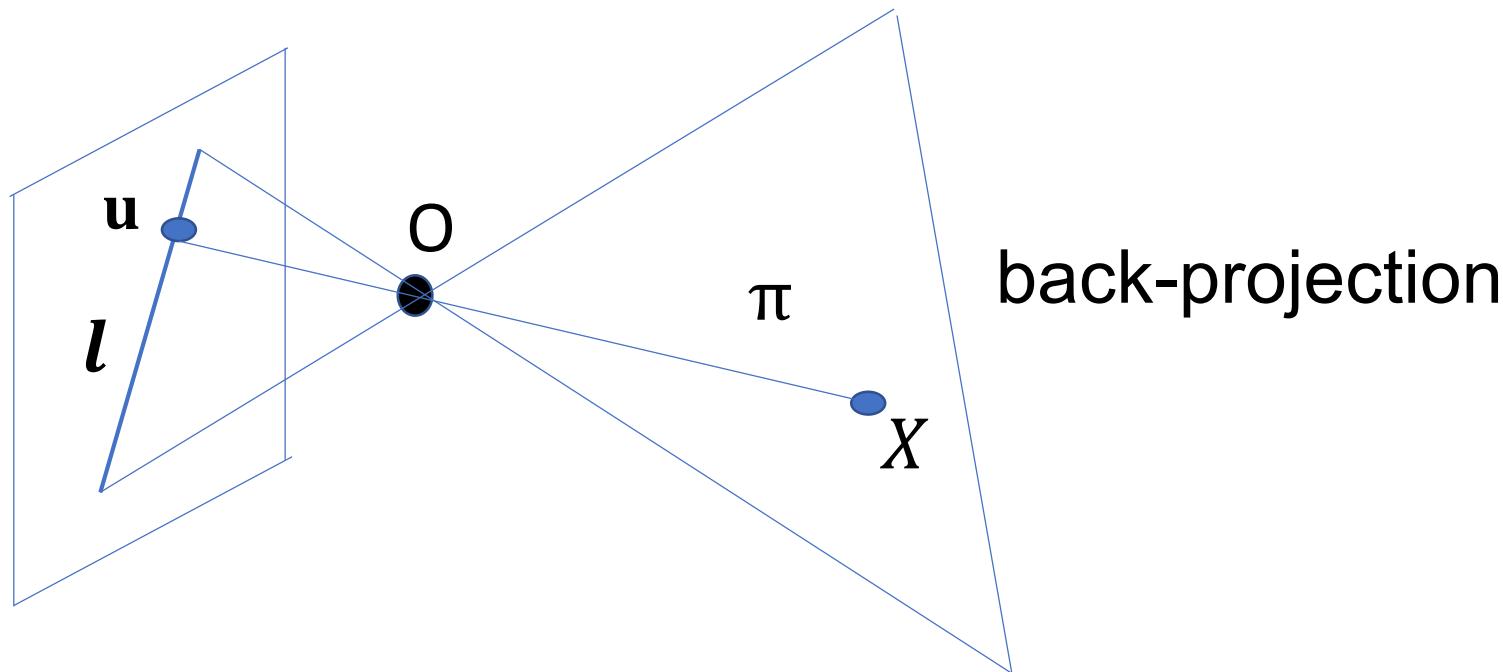
$$\mathbf{l}^T \mathbf{u} = \mathbf{l}^T \mathbf{P}X = 0$$

$$\boldsymbol{\pi}^T X = 0$$

back-projection of \mathbf{l} is plane $\boldsymbol{\pi} = \mathbf{P}^T \mathbf{l}$

Back-projection of an image line

set of space points X , whose image projection $\mathbf{u} = \mathbf{P}X$ is on image line \mathbf{l}



$$\mathbf{l}^T \mathbf{u} = \mathbf{l}^T \mathbf{P}X = 0$$

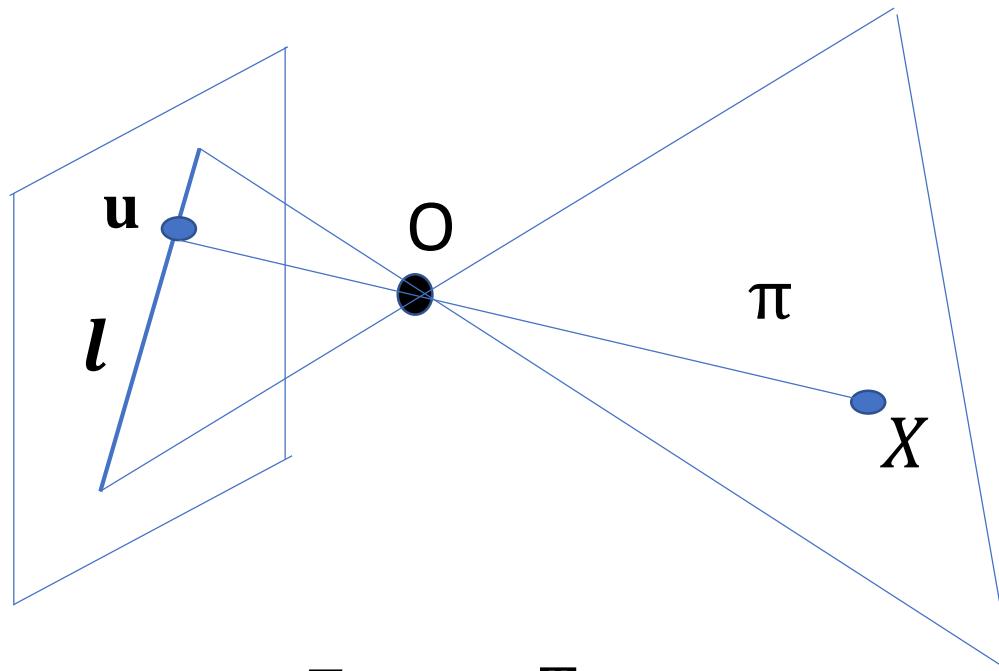
$$\pi^T X = 0$$

back-projection of \mathbf{l} is plane $\pi = \mathbf{P}^T \mathbf{l}$ through \mathbf{O} , in fact, since $\mathbf{O} = \text{RNS}(\mathbf{P})$

$$\pi^T \mathbf{O} = \mathbf{l}^T \mathbf{P} \mathbf{O} = \mathbf{l}^T \mathbf{0} = 0$$

Back-projection of an image line

set of space points X , whose image projection is on image line l



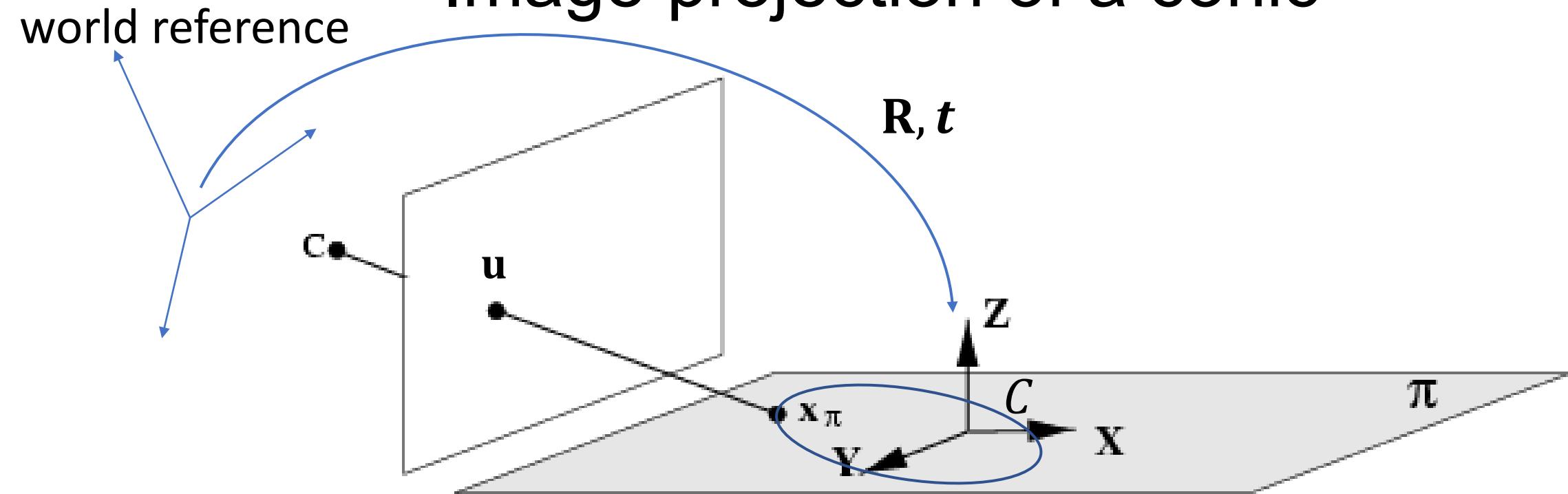
back-projection of l is $\pi = P^T l$

$$\pi^T X = l^T P X = 0$$

try with $X = O = \text{RNS}(P)$:
 $\pi^T O = l^T P O = l^T 0 = 0$

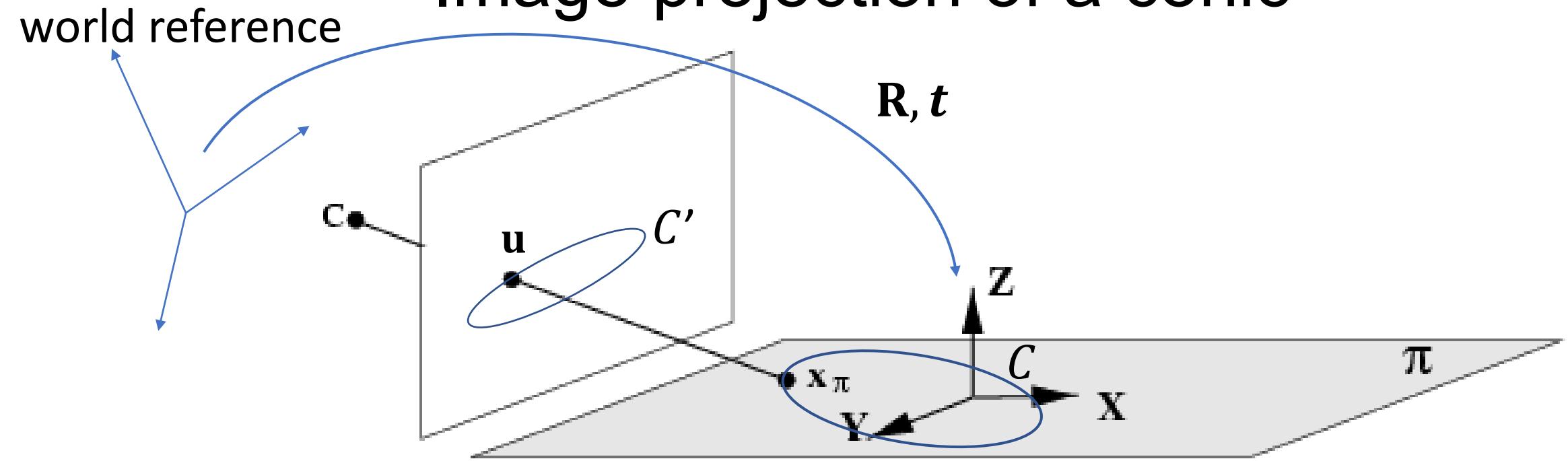
The backprojection of l is a plane $\pi = P^T l$ through O

Image projection of a conic



$$\begin{aligned}
 \mathbf{u} &= P X_w = [\mathbf{M} \quad \mathbf{m}] X_w = [\mathbf{M} \quad \mathbf{m}] \begin{bmatrix} \mathbf{R} & \mathbf{t} \\ 0 & 1 \end{bmatrix} X_\pi = [\underbrace{\mathbf{M}\mathbf{R}}_{\mathbf{P}'}, \underbrace{\mathbf{M}\mathbf{t} + \mathbf{m}}_{\mathbf{w}}] \begin{bmatrix} x \\ y \\ 0 \\ w \end{bmatrix} \\
 &= [p'_1 \quad p'_2 \quad p'_3 \quad p'_4] [x \quad y \quad 0 \quad w]^T = [p'_1 \quad p'_2 \quad p'_4] \mathbf{x}_\pi = P' \pi \mathbf{x}_\pi
 \end{aligned}$$

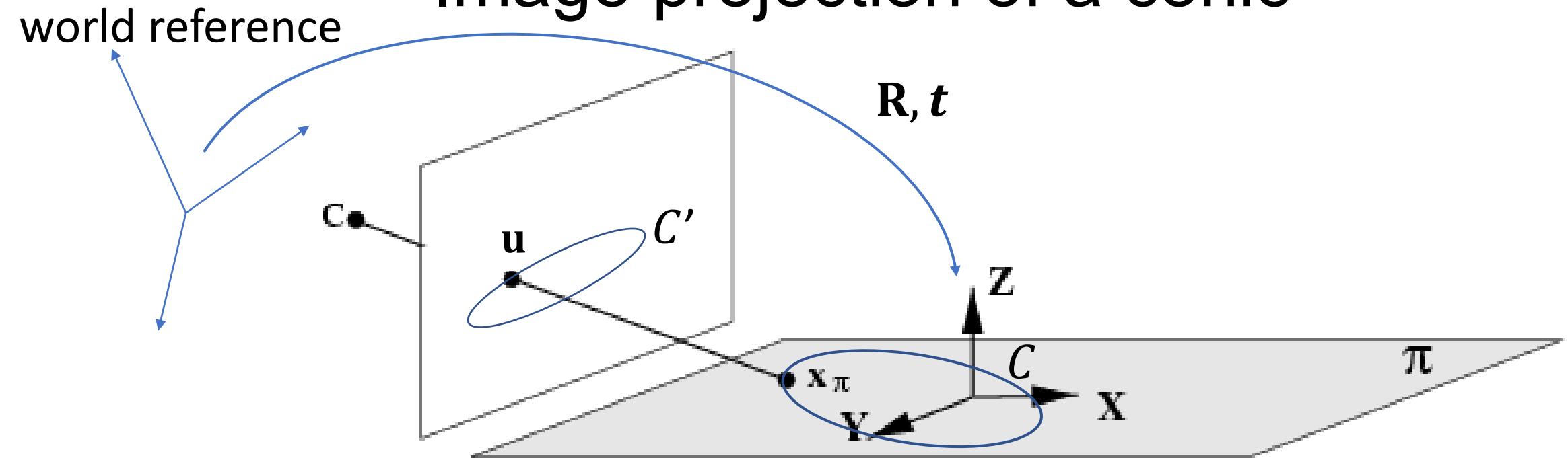
Image projection of a conic



$$\mathbf{x}_\pi^T C \mathbf{x}_\pi = 0$$

$$\mathbf{u} = [p'_1 \quad p'_2 \quad p'_4] \quad \mathbf{x}_\pi = P'_\pi \mathbf{x}_\pi \rightarrow \mathbf{x}_\pi = P'^{-1}_\pi \mathbf{u}$$

Image projection of a conic



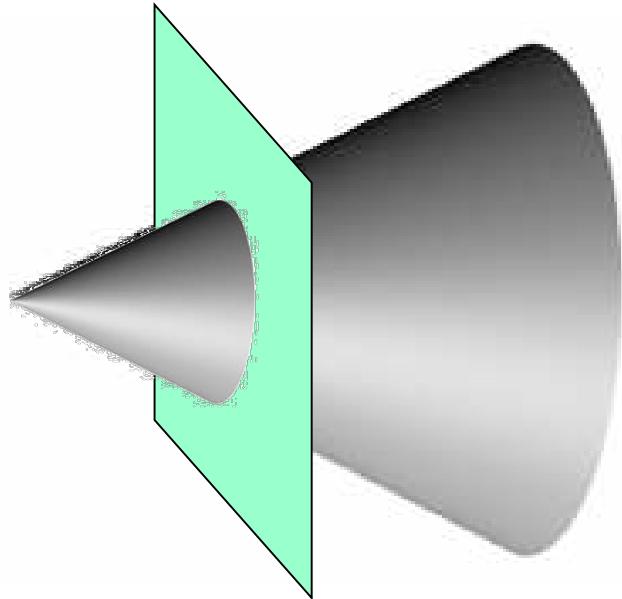
$$\mathbf{x}_\pi^T C \mathbf{x}_\pi = 0$$

$$\mathbf{u} = [p'_1 \quad p'_2 \quad p'_4] \quad \mathbf{x}_\pi = P'_\pi \mathbf{x}_\pi \rightarrow \mathbf{x}_\pi = P'^{-1}_\pi \mathbf{u}$$
$$\rightarrow$$

$$\mathbf{u}^T P'^{-T}_\pi C P'^{-1}_\pi \mathbf{u} = 0$$

$$\mathbf{u}^T C' \mathbf{u} = 0 \text{ with } C' = P'^{-T}_\pi C P'^{-1}_\pi$$

Back-projection of an image conic



$$u = PX$$

$$u^T Cu = X^T P^T C P X = X^T Q_{co} X = 0$$

back-projection of an image
conic is a rank-3 quadric, $Q_{co} = P^T C P$
 \rightarrow a **cone**

example: $Q_{co} = \begin{bmatrix} K^T \\ 0 \end{bmatrix} C^T [K | 0] = \begin{bmatrix} K^T C K & 0 \\ 0 & 0 \end{bmatrix}$

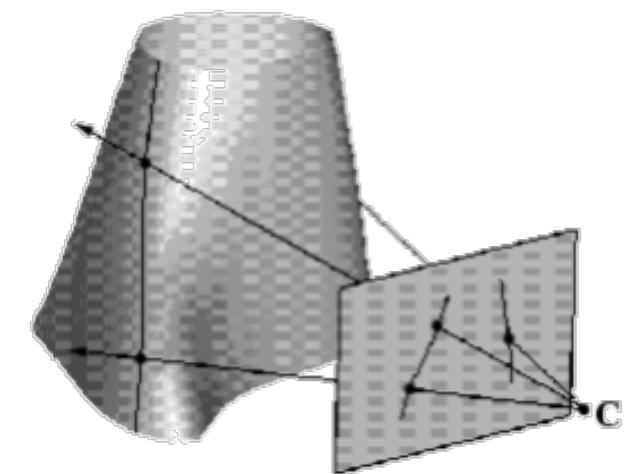
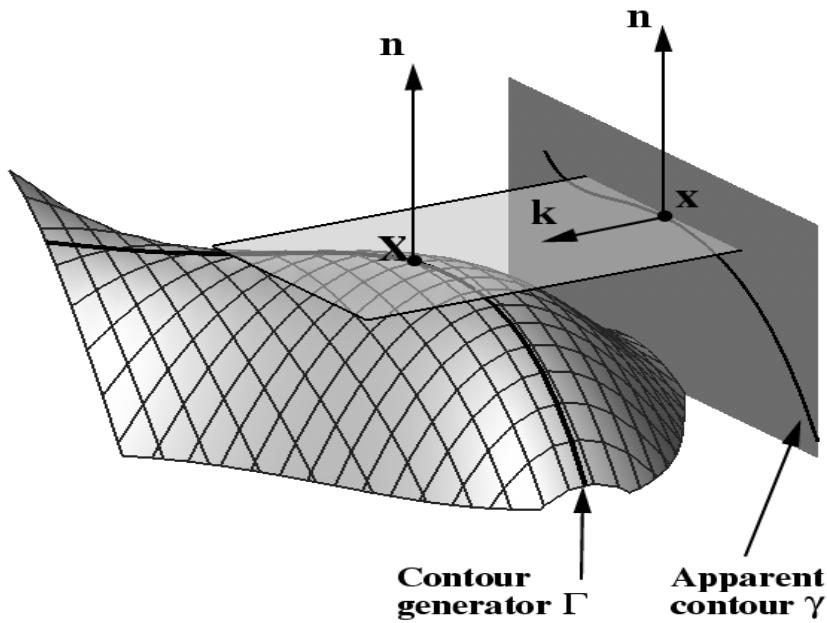
Apparent contour of a smooth surface

Image projection of smooth surfaces

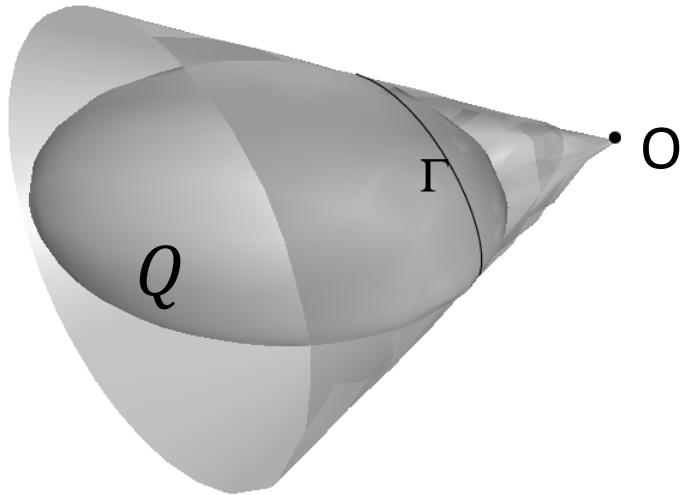
The **contour generator** Γ is the set of points X on a surface S , whose viewing rays are tangent to S .

The corresponding **apparent contour** is the image of Γ

- Γ depends only on position of projection center O ,
- Γ depends also on rest of the projection matrix P

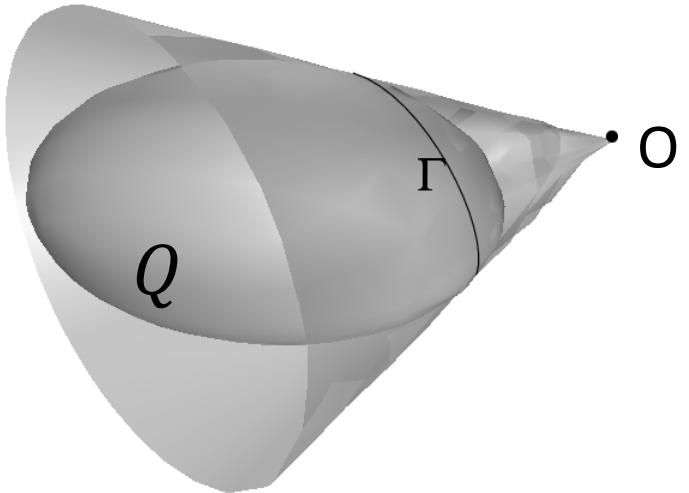


The apparent contour of a quadric



select those planes, tangent to the quadric Q ,
which go through camera center O

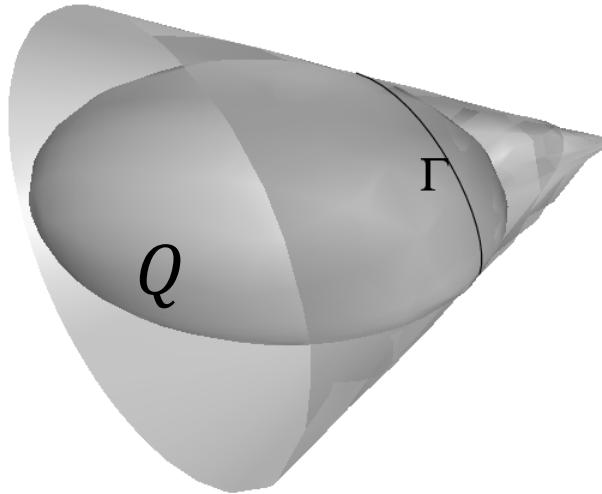
The apparent contour of a quadric



- o select those planes, tangent to the quadric Q ,
i.e. belonging to dual quadric $Q^* = Q^{-1}$
which go through camera center O

$$\Pi^T Q^* \Pi = 0$$

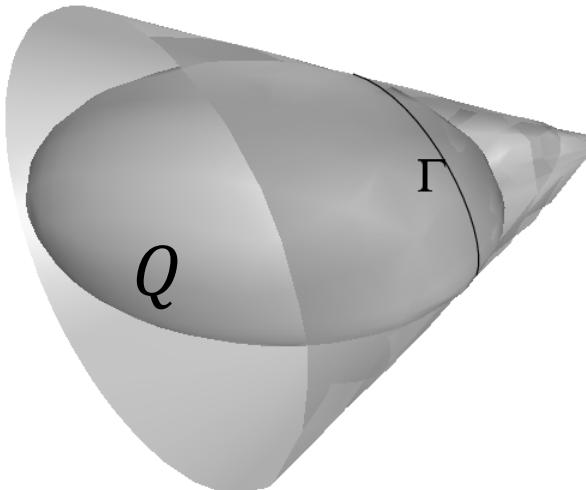
The apparent contour of a quadric



- o select those planes, tangent to the quadric Q ,
i.e. belonging to dual quadric $Q^* = Q^{-1}$
which go through camera center O
.e., that are backprojections $P^T l$ of some image lines l

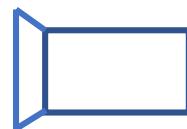
$$\Pi^T Q^* \Pi = l^T P Q^* P^T l = 0$$

The apparent contour of a quadric



- o select those planes, tangent to the quadric Q ,
i.e. belonging to dual quadric $Q^* = Q^{-1}$
which go through camera center O
.e., that are backprojections $P^T l$ of some image lines l

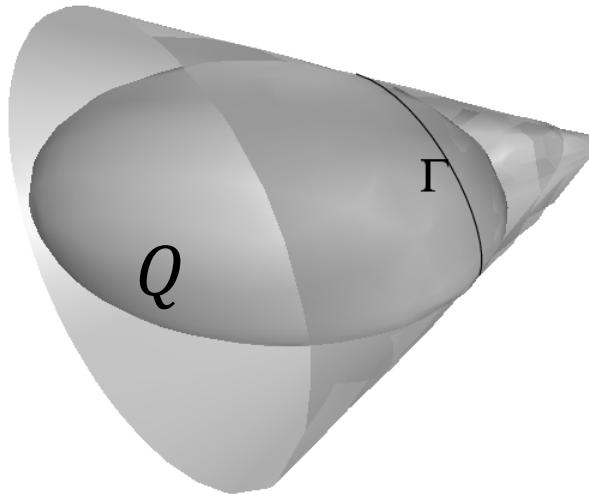
these planes Π are tangent
to the contour generator Γ



$$\Pi^T Q^* \Pi = l^T P Q^* P^T l = 0$$

these lines l are image of planes Π
 $\rightarrow l$ are tangent to image of Γ
 $\rightarrow l$ are tangent to apparent contour γ

The apparent contour of a quadric

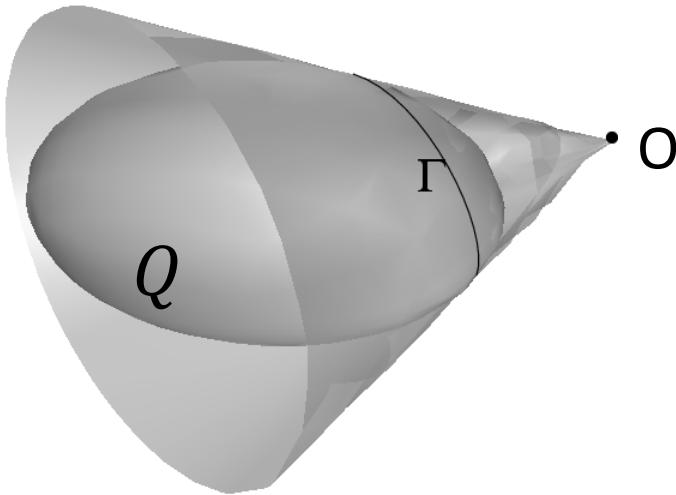


- o select those planes, tangent to the quadric Q ,
i.e. belonging to dual quadric $Q^* = Q^{-1}$
which go through camera center O
i.e., that are backprojections $P^T l$ of some image lines l

$$\Pi^T Q^* \Pi = l^T P Q^* P^T l = 0$$

these image lines l satisfy a quadratic equation
→ they belong to a dual conic $C^* = P Q^* P^T$

The apparent contour of a quadric

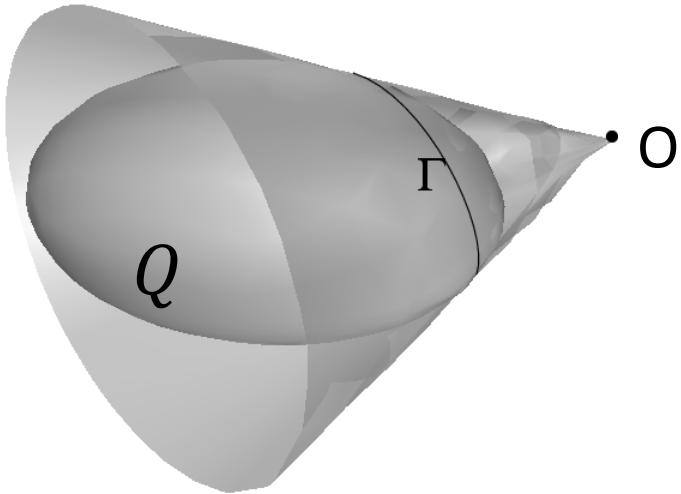


- o select those planes, tangent to the quadric Q ,
i.e. belonging to dual quadric $Q^* = Q^{-1}$
which go through camera center O
i.e., that are backprojections of some image lines \mathbf{l}

$$\Pi^T Q^* \Pi = \mathbf{l}^T P Q^* P^T \mathbf{l} = 0$$

these image lines \mathbf{l} satisfy a quadratic equation
→ they belong to a dual conic $C^* = P Q^* P^T$
→ they are tangent to a conic $C = C^{*-1} = (P Q^* P^T)^{-1}$

The apparent contour of a quadric



- o select those planes, tangent to the quadric Q ,
i.e. belonging to dual quadric $Q^* = Q^{-1}$
which go through camera center O
i.e., that are backprojections of some image lines \mathbf{l}

$$\Pi^T Q^* \Pi = \mathbf{l}^T P Q^* P^T \mathbf{l} = 0$$

these image lines \mathbf{l} satisfy a quadratic equation

→ they belong to a dual conic $C^* = P Q^* P^T$

→ they are tangent to a conic $C = C^{*-1} = (P Q^* P^T)^{-1}$

But \mathbf{l} are tangent to conic $C = C^{*-1} = (P Q^* P^T)^{-1} \rightarrow$ a.c γ is the conic C

Exercise: apparent contour of a cone?

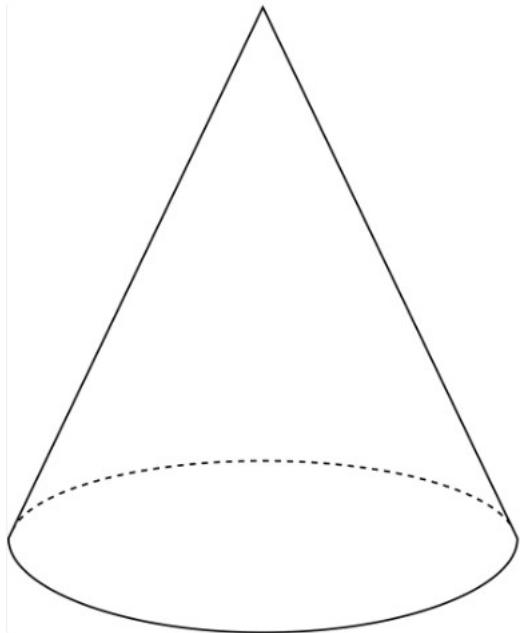
Exercise: apparent contour of a cone?

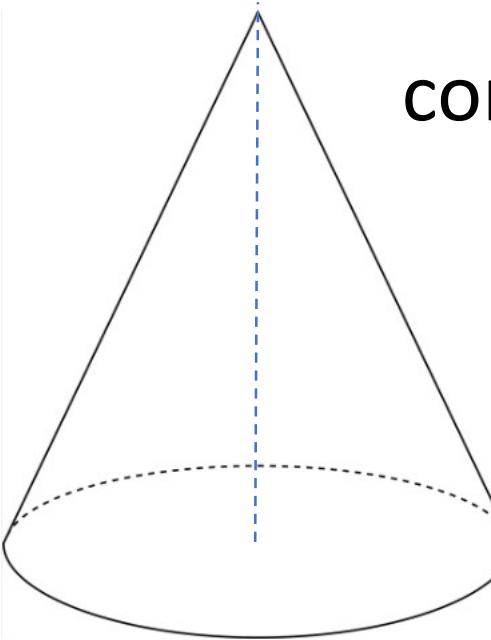
- contour generator Γ : set of tangency points from O
→ two straight lines through the vertex V
- apparent contour γ : image of Γ
→ two image lines: i.e, a degenerate conic

Example: contour generator of a **right** cone?

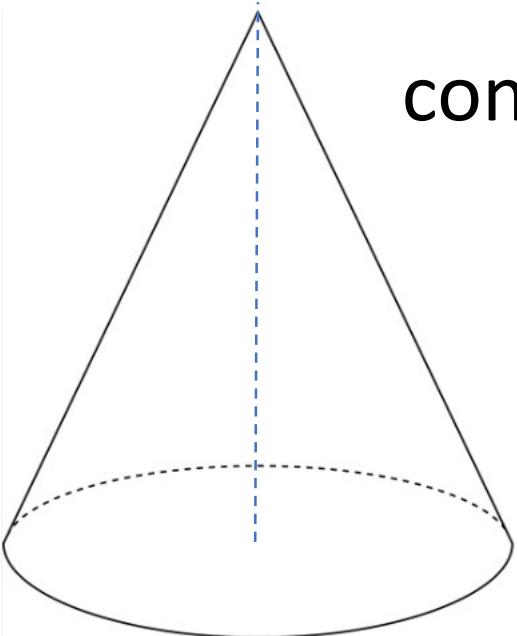
- contour generator Γ : set of tangency points from O
→ two straight lines through the vertex V
- **right** cone alone is **symmetric** wrt to its axis
→ right cone + viewpoint O : «less» symmetry,
which symmetry?

a right cone



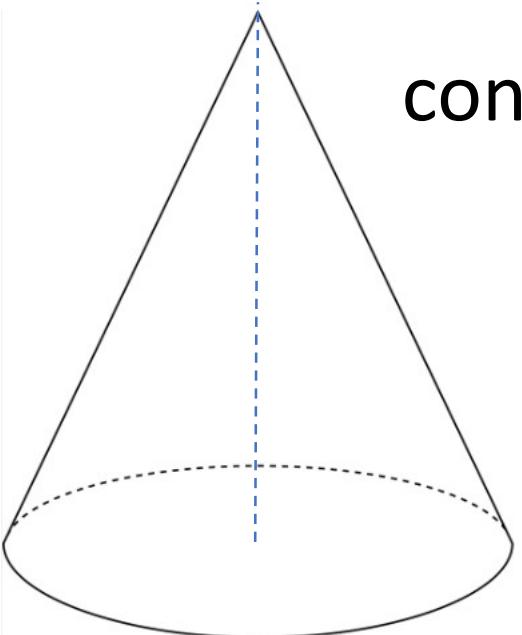


cone axis



cone axis

a circular cross section

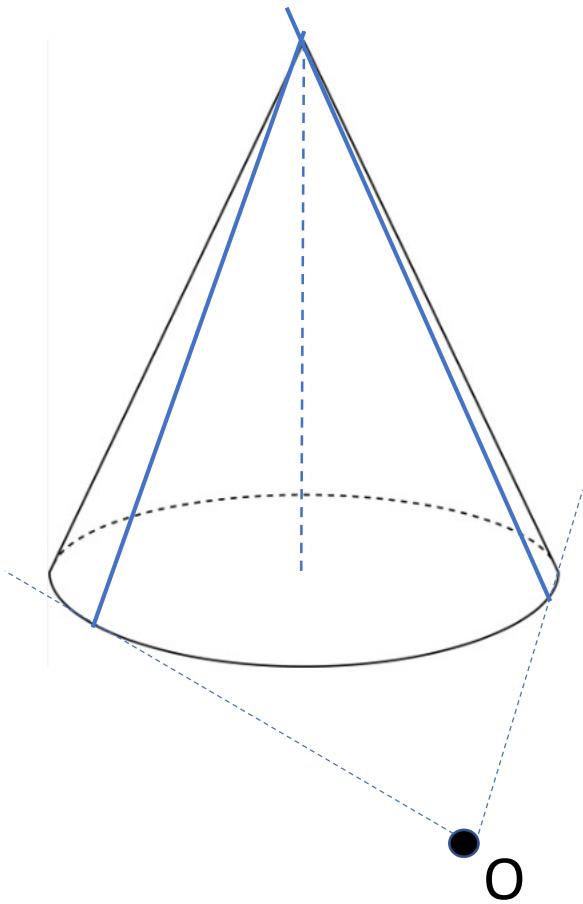


cone axis

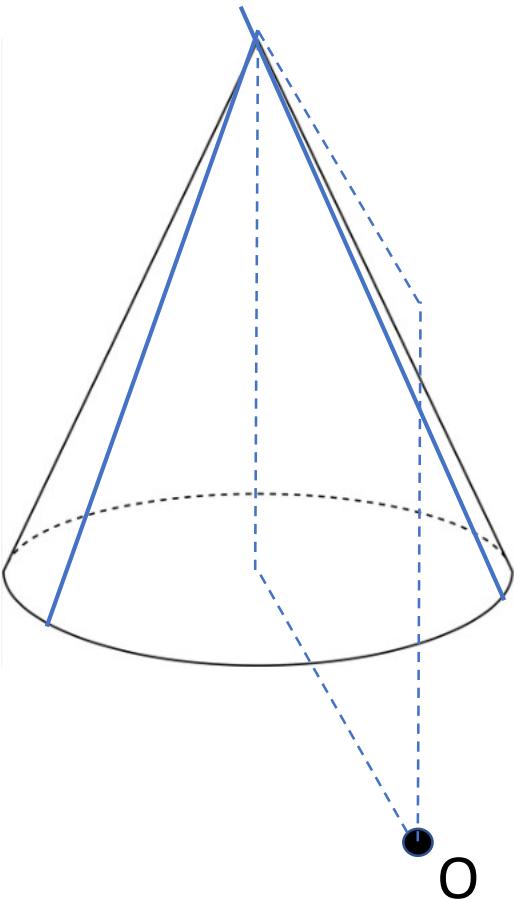
a circular cross section



camera viewpoint

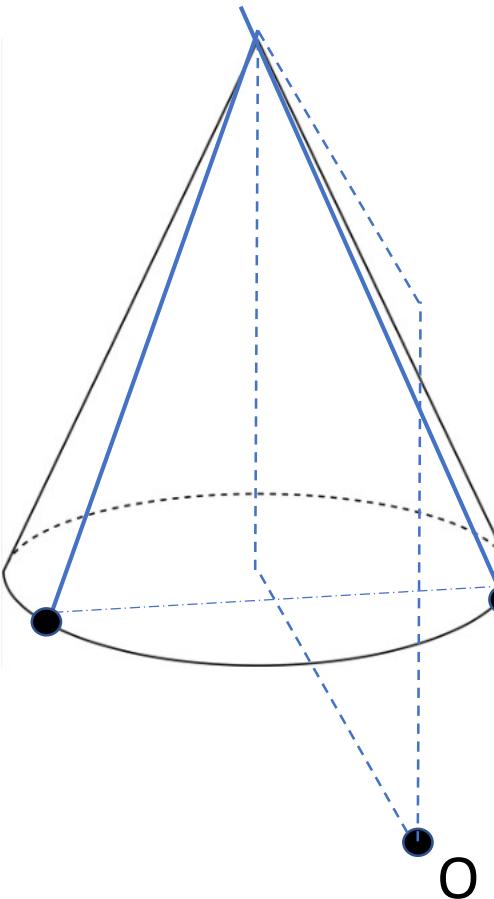


contour generator



contour generator

symmetry plane



contour generator

points are symmetric wrt symmetry plane

symmetry plane

Example: contour generator of a **right** cone?

- contour generator Γ : set of tangency points from O
→ two straight lines through the vertex V
- **right** cone e alone is **symmetric** wrt to its axis
→ right cone + viewpoint O : «less» symmetry,
which symmetry?
PLANAR SYMMETRY wrt plane through axis and O

Example: contour generator of a **right** cone?

- contour generator Γ : set of tangency points from O
→ two straight lines through the vertex V
- **right** cone alone is **symmetric** wrt to its axis
→ right cone + viewpoint O : «less» symmetry,
which symmetry?
PLANAR SYMMETRY wrt plane through axis and O
namely, wrt backprojection plane of the imaged axis

natural camera calibration
from 3 images of a rectangle

Classical approach

Three equations from vanishing points of three mutually orthogonal directions:

Drawbacks: estimation of vanishing points might be inaccurate

Alternative approach

- Motivated from Zhang: images of a planar object of known shape



- Here the object shape has a unique unknown parameter: the aspect ratio a , i.e., the ratio between «horizontal» and «vertical» length

Given a rectangle R of unknown aspect ratio a

- Take a first image of the rectangle: let r_1 be the extracted image
- Rectify this image by mapping it onto a rectangle R_1 by means of a homography H_1
- The aspect ratio of this reconstructed rectangle is $a_1 = sa$, where s is an unknown scale factor
- Therefore, the circular points of the plane containing the real rectangle R have been mapped onto two points (I_1, J_1) given by

$$(I_1, J_1) = \begin{bmatrix} s & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} (I, J)$$

- The first image of these circular points is therefore

$$(I'_1, J'_1) = H_1^{-1}(I_1, J_1) = H_1^{-1} \begin{bmatrix} s & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} (I, J)$$

- Consider now the second image R_2 of the rectangle: once it is partially rectified to a rectangle r_2 , compute its aspect ratio $a_2 = s_2 a$. This time, the scale factor s_2 can be expressed in terms of s as

$$s_2 = \lambda s$$

with $\lambda = \frac{a_2}{a_1}$ known after the observation of the two aspect ratios

- If H_2 is the homography used to rectify the second image R_2 of the rectangle to the partially rectified rectangle r_2 , then the second image of the circular points (I, J) is given by

$$(I'_2, J'_2) = H_2^{-1}(I_2, J_2) = H_2^{-1} \begin{bmatrix} \lambda s & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} (I, J)$$

and, similarly, the third image of the same circular points is

$$(I'_3, J'_3) = H_3^{-1}(I_3, J_3) = H_3^{-1} \begin{bmatrix} \mu s & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} (I, J)$$

Now we have three equations like in Zhang,
but also an additional unknown scaling factor s

A further equation comes from the natural camera constraint:

An additional pair of imaged circular points is

$$(I, J) = \left(\begin{bmatrix} 1 \\ i \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ -i \\ 0 \end{bmatrix} \right)$$

in fact,

they are images of circular points of any plane parallel to the image plane

Therefore we have to solve

$$I_1'^T \omega I_1' = 0$$

$$I_2'^T \omega I_2' = 0$$

$$I_3'^T \omega I_3' = 0$$

and

$$I^T \omega I = 0$$

for both ω and s , with

$$\lambda = \frac{a_2}{a_1}$$

and

$$\mu = \frac{a_3}{a_1}$$