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Computer Vision: 7. Camera Models (a)

Objective

- Geometric camera models
  - Intrinsic and extrinsic parameters
  - Projection equations

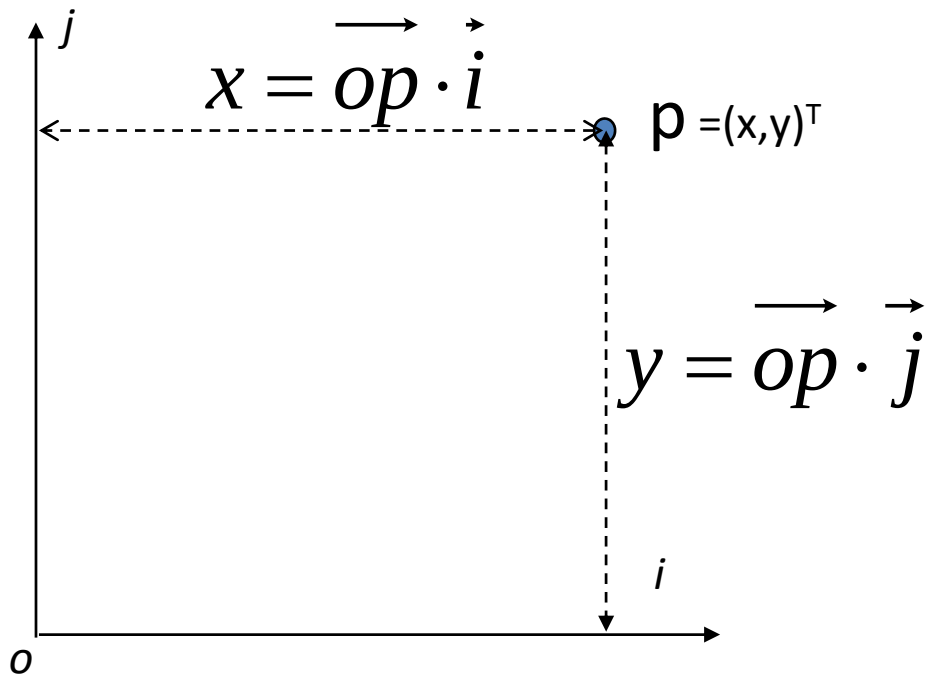
Textbook:
• David A. Forsyth and Jean Ponce, Computer Vision: A Modern Approach, Prentice Hall, New Jersey, 2003.

Plenty of slides are modified from the reference lecture notes or project pages:
• Prof. J. Rehg, Computer Vision, Georgia Inst. of Tech.
• Prof. T. Darrell, Computer Vision and Applications, MIT.

• Prof. D.A. Forsyth, Computer Vision, UIUC.

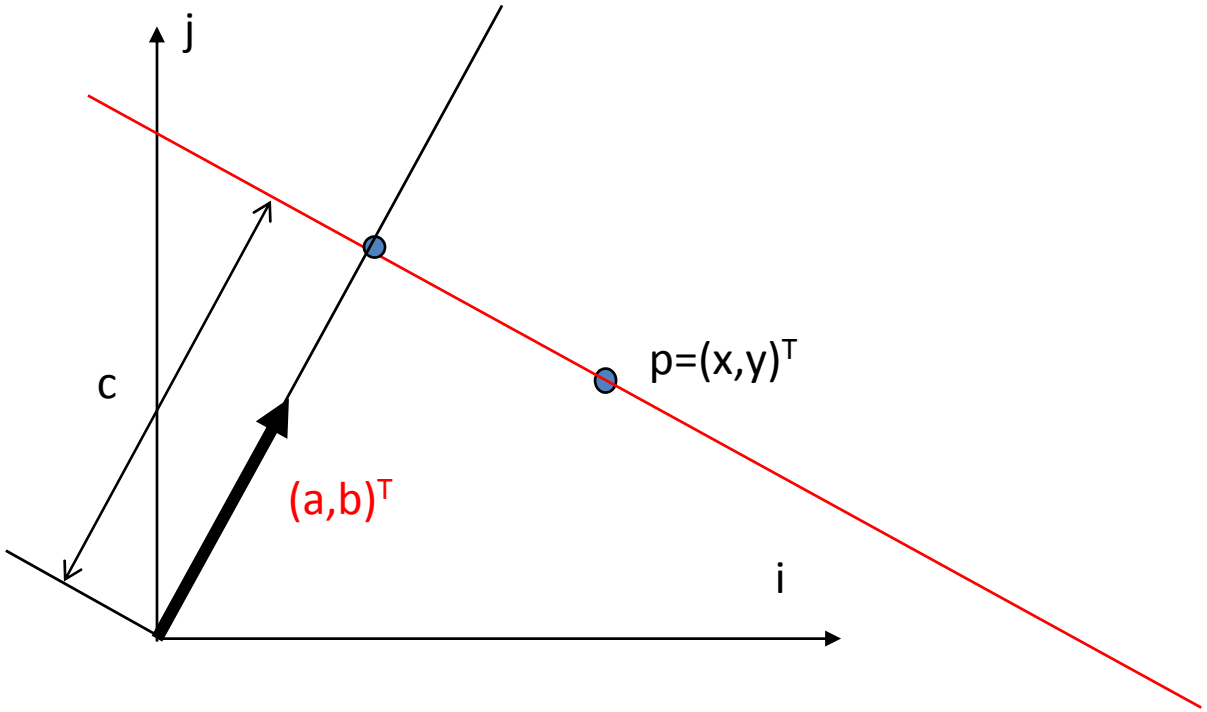
2D coordinate frames & points

- Coordinates x and y
- For a more general coordinate representation, we usually use a vector form.



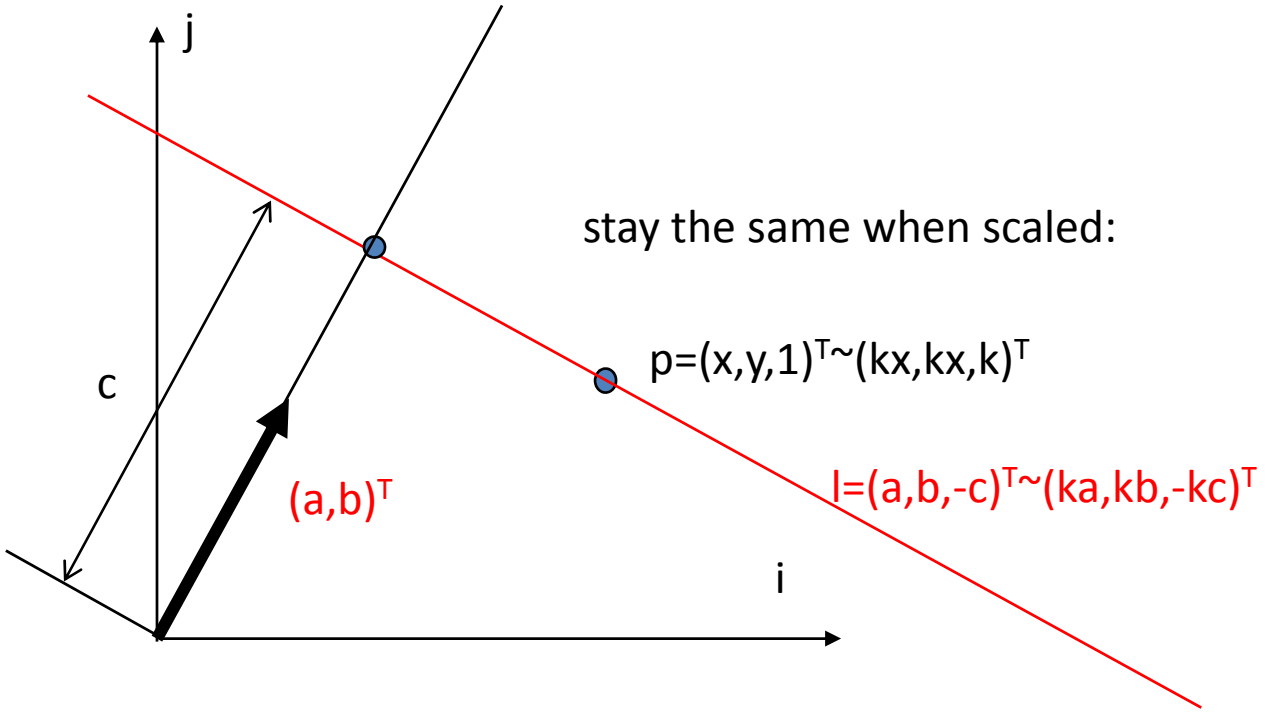
2D lines

- Line *l*: $ax+by=c$ <-> $(a,b)^T(x,y)=c$



Homogeneous coordinates

- Uniform treatment of points and lines
- Line-point incidence: $\boldsymbol{l}^T p=0$

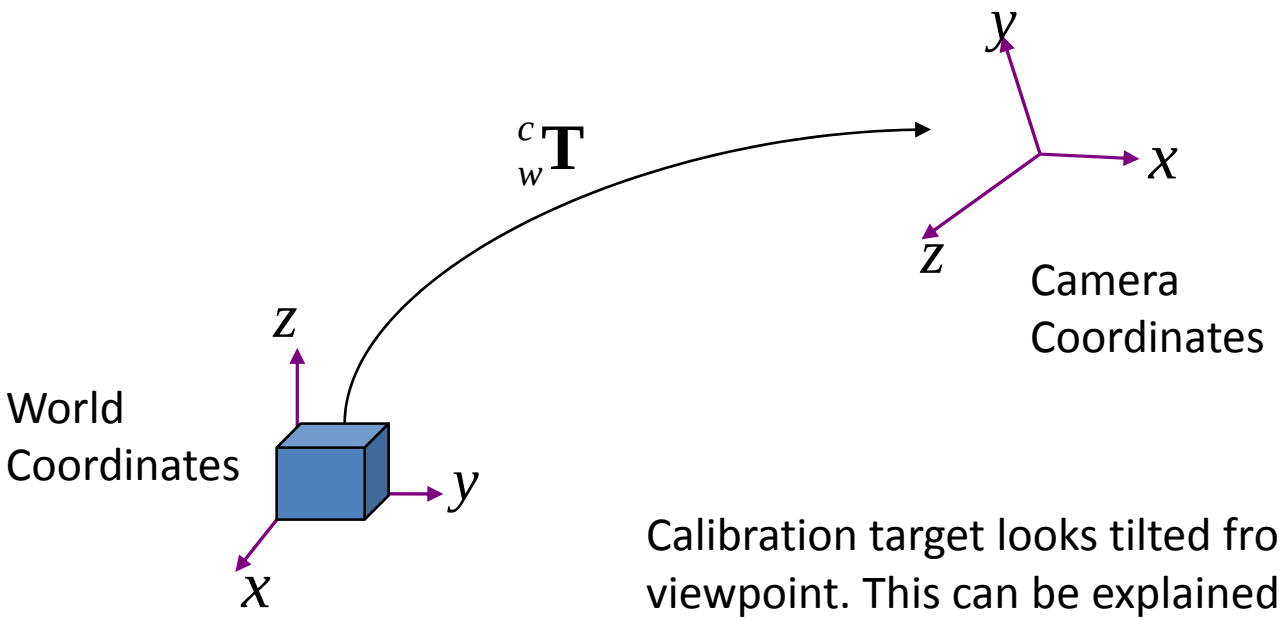


Homogeneous coordinates (cont.)

- Furthermore, …
  - We use homogenous coordinates to combine rotation and translation into same framework: matrix transformation.
  - It allows easy transformation between “frames” – common between computer vision and graphics.

Camera pose

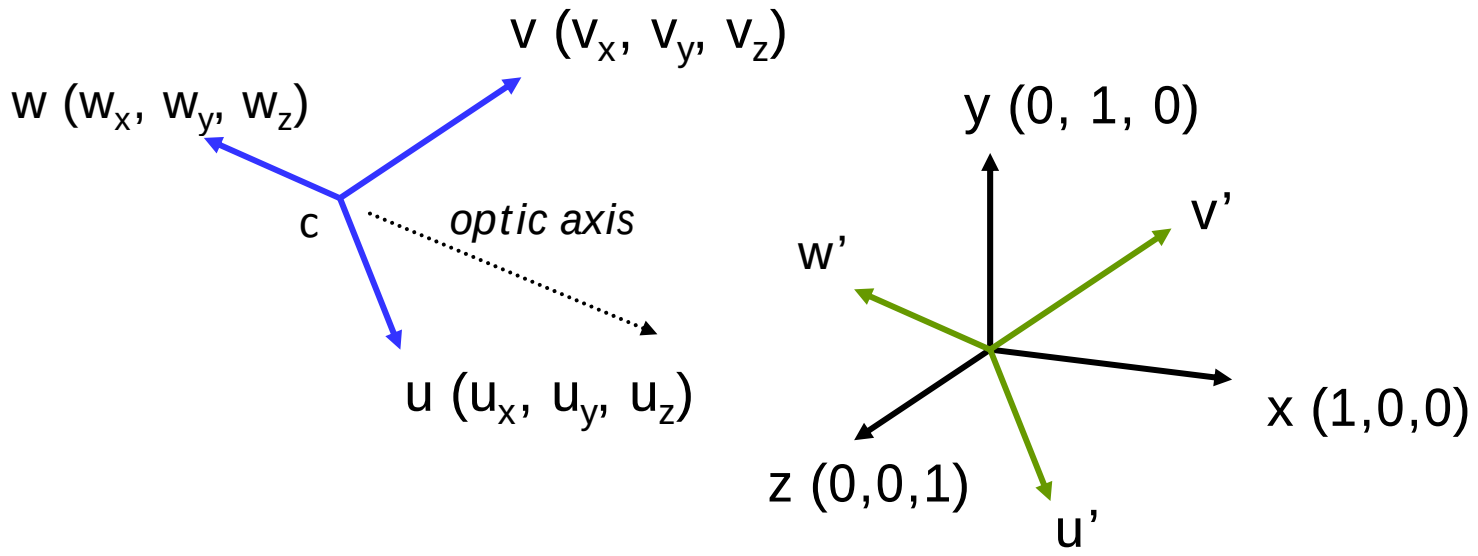
- To apply the camera model, objects in the scene must be expressed in *camera coordinates*.



Calibration target looks tilted from camera viewpoint. This can be explained as a difference in coordinate systems.

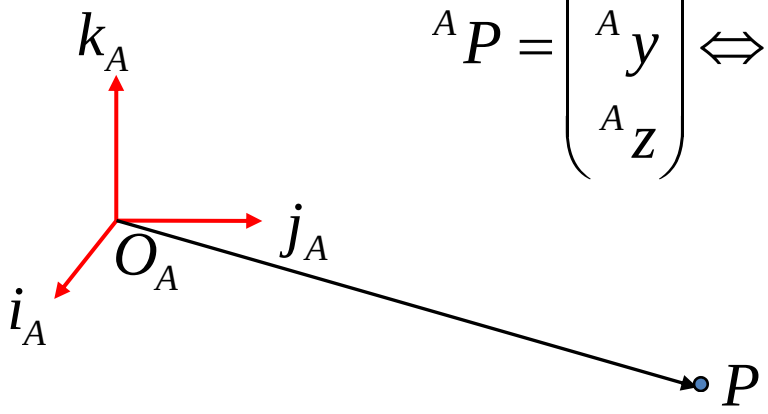
Rigid body transformations

- Need a way to specify the six degrees-of-freedom of a rigid body.
- 3 rotation and 3 translation DOFs.
- *R*, *t*: the extrinsic parameters.



Notations

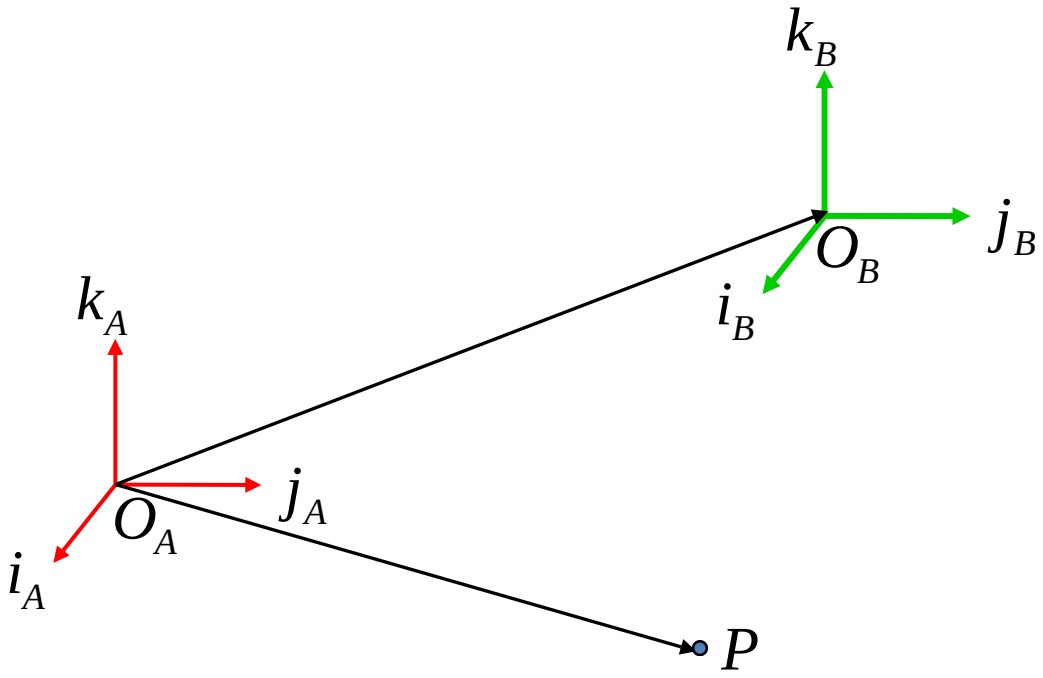
- Superscript references coordinate frame
- ${}^{A}P$ is coordinates of P in frame A
- ${}^{B}P$ is coordinates of P in frame B



$$ {}^{A}P = \begin{pmatrix} {}^{A}x \\ {}^{A}y \\ {}^{A}z \end{pmatrix} \Leftrightarrow \overline{OP} = \left({}^{A}x \bullet \overline{i_A}\right) + \left({}^{A}y \bullet \overline{j_A}\right) + \left({}^{A}z \bullet \overline{k_A}\right) $$

Translation

$$^{B}P = {}^{A}P + {}^{B}(O_A)$$



Translation

- Using homogeneous coordinates, translation can be expressed as a matrix multiplication.

$$ {}^{B}P = {}^{A}P + {}^{B}O_{A} $$

$$ \begin{bmatrix} {}^{B}P \\ 1 \end{bmatrix} = \begin{bmatrix} I & {}^{B}O_{A} \\ 0 & 1 \end{bmatrix} \begin{bmatrix} {}^{A}P \\ 1 \end{bmatrix} $$

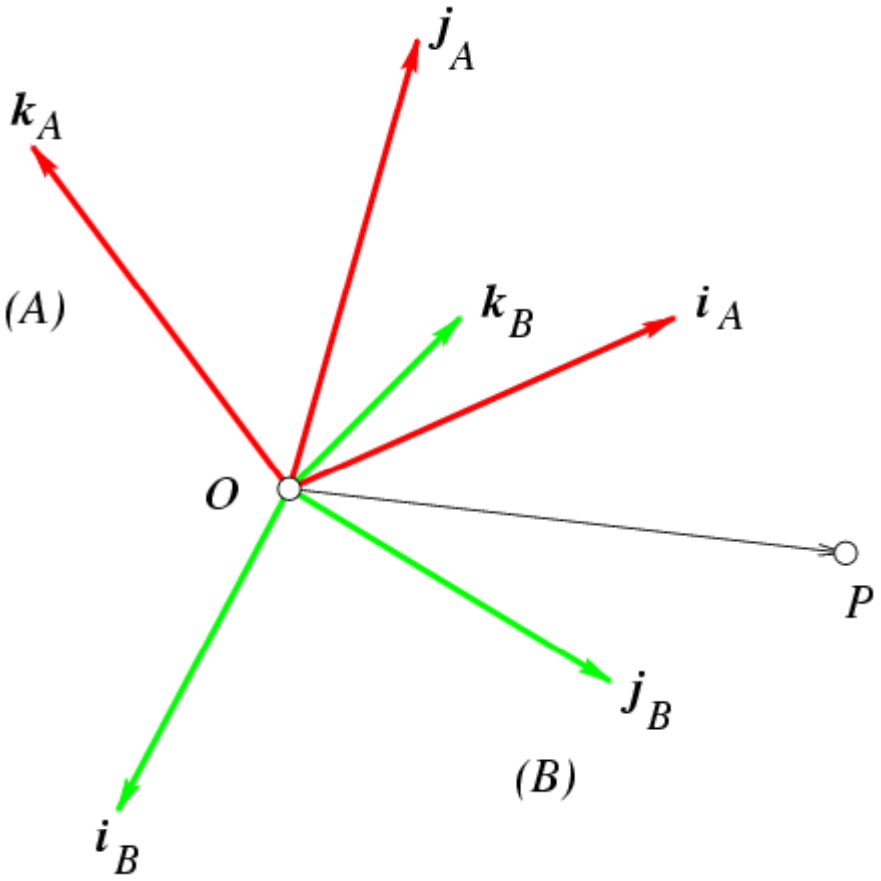
- Translation is commutative

Rotation

$$\overrightarrow{OP} = \begin{pmatrix} i_A & j_A & k_A \end{pmatrix} \begin{pmatrix} {}^A x \\ {}^A y \\ {}^A z \end{pmatrix} = \begin{pmatrix} i_B & j_B & k_B \end{pmatrix} \begin{pmatrix} {}^B x \\ {}^B y \\ {}^B z \end{pmatrix}$$

$$^B P = {}^B_A R \, {}^A P$$

${}^B_A R$ means describing frame A in The coordinate system of frame B

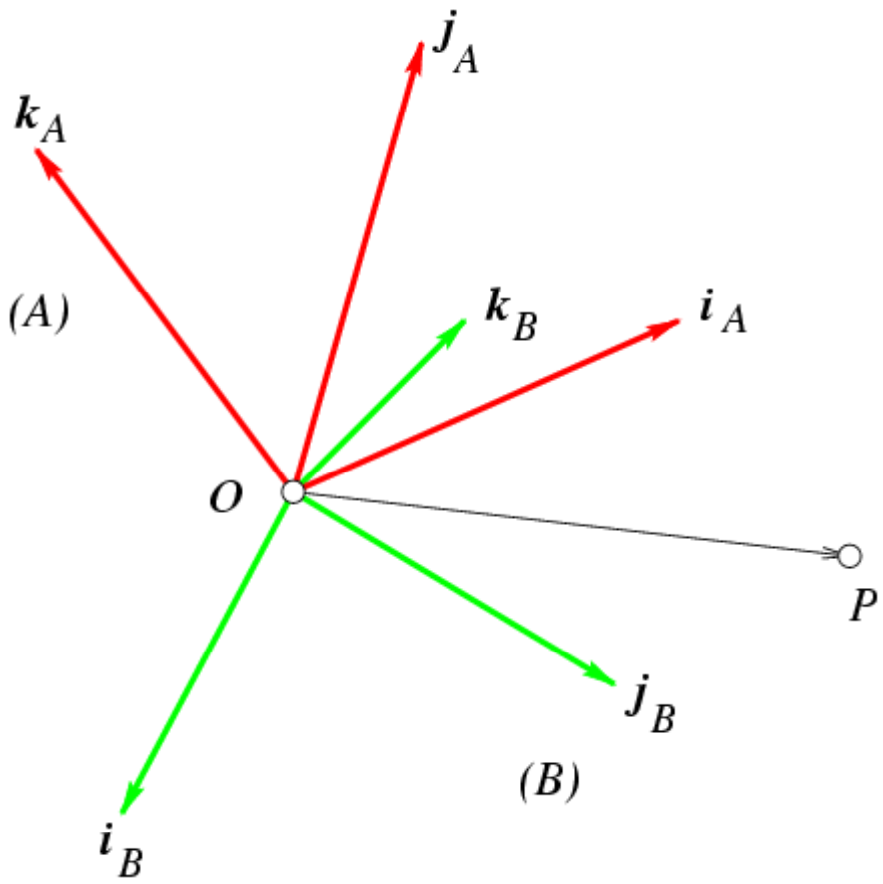


Rotation (from frame A to B)

$$
{}^{B}_{A}R = \begin{bmatrix} \mathbf{i}_A.\mathbf{i}_B & \mathbf{j}_A.\mathbf{i}_B & \mathbf{k}_A.\mathbf{i}_B \\ \mathbf{i}_A.\mathbf{j}_B & \mathbf{j}_A.\mathbf{j}_B & \mathbf{k}_A.\mathbf{j}_B \\ \mathbf{i}_A.\mathbf{k}_B & \mathbf{j}_A.\mathbf{k}_B & \mathbf{k}_A.\mathbf{k}_B \end{bmatrix}
$$

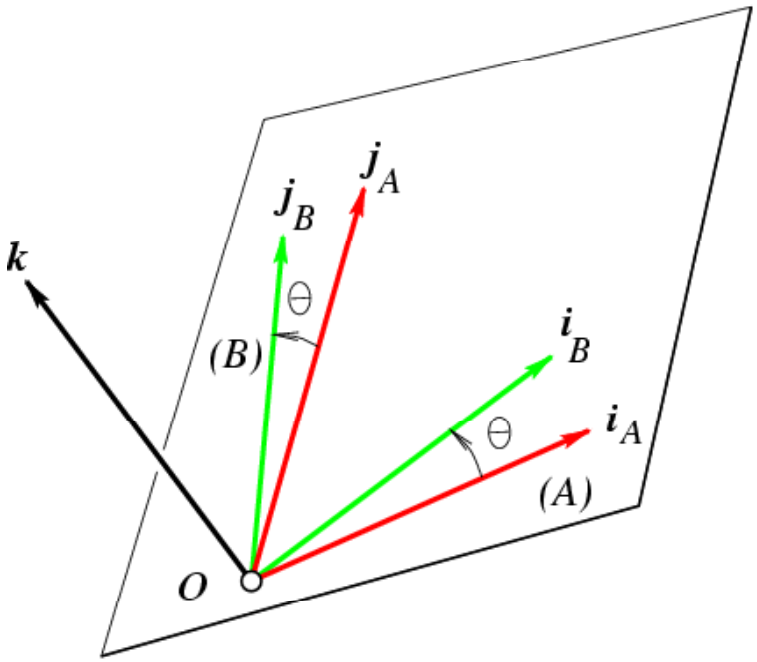
$$
= \begin{bmatrix} {}^{B}\mathbf{i}_A & {}^{B}\mathbf{j}_A & {}^{B}\mathbf{k}_A \end{bmatrix}
$$

$$
= \begin{bmatrix} {}^{A}\mathbf{i}_B^T \\ {}^{A}\mathbf{j}_B^T \\ {}^{A}\mathbf{k}_B^T \end{bmatrix}
$$

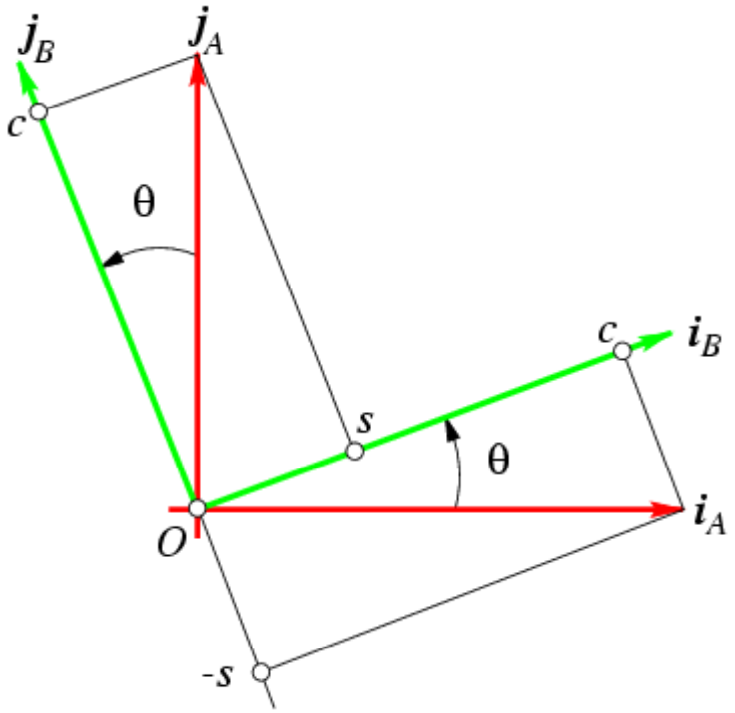


Orthogonal matrix: $R^{-1} = R^T$

Example: Rotation about z axis



What is the rotation matrix?



Combine 3 to get arbitrary rotation

- Euler angles: Z, X', Y''
- Heading, pitch roll: world Z, new X, new Y
- Three basic matrices: order matters, but we'll probably not focus on that

$$R_Z(\theta) = \begin{bmatrix} \cos(\theta) & -\sin(\theta) & 0 \\ \sin(\theta) & \cos(\theta) & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$R_X(\phi) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos(\phi) & -\sin(\phi) \\ 0 & \sin(\phi) & \cos(\phi) \end{bmatrix}$$

$$R_Y(\kappa) = \begin{bmatrix} \cos(\kappa) & 0 & \sin(\kappa) \\ 0 & 1 & 0 \\ -\sin(\kappa) & 0 & \cos(\kappa) \end{bmatrix}$$

Remind: applying coordinate rotation φ is equal to applying rotation – φ to objects.

Rotation in homogeneous coordinates

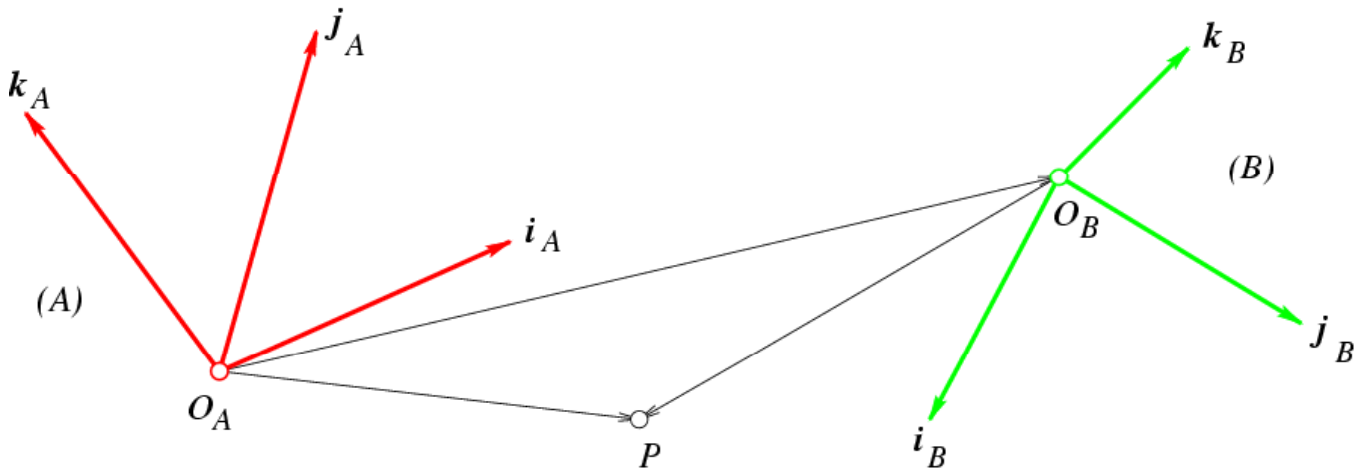
- Using homogeneous coordinates, rotation can be expressed as a matrix multiplication.

$$ {}^{B}P = {}^{B}_{A}R\,{}^{A}P $$

$$ \begin{bmatrix} {}^{B}P \\ 1 \end{bmatrix} = \begin{bmatrix} {}^{B}_{A}R & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} {}^{A}P \\ 1 \end{bmatrix} $$

- Rotation is not commutative

Rigid transformations



$$^{B}P = {}^{B}_{A}R\,{}^{A}P + {}^{B}O_A$$

Rigid transformations (cont.)

- Unified treatment using homogeneous coordinates.

$$\begin{bmatrix} {}^{B}P \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & {}^{B}O_{A} \\ 0 & 1 \end{bmatrix} \begin{bmatrix} {}^{B}_{A}R & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} {}^{A}P \\ 1 \end{bmatrix}$$

$$= \begin{bmatrix} {}^{B}_{A}R & {}^{B}O_{A} \\ \mathbf{0}^{T} & 1 \end{bmatrix} \begin{bmatrix} {}^{A}P \\ 1 \end{bmatrix}$$

Invertible!

$$\begin{bmatrix} {}^{B}P \\ 1 \end{bmatrix} = {}^{B}_{A}T \begin{bmatrix} {}^{A}P \\ 1 \end{bmatrix}$$

Perspective camera model

- Linear transformation of perspective projection coordinate.

$$p = \begin{bmatrix} u \\ v \\ w \end{bmatrix} = [I \quad 0]P = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

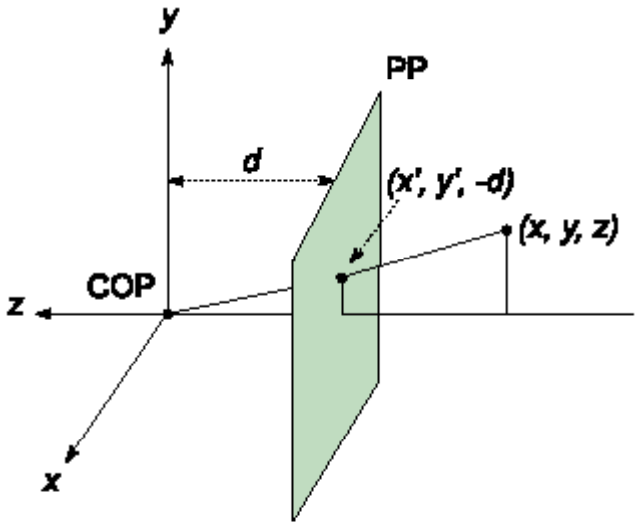
- Recover image (normalized) coordinate by projection.

$$\hat{u} = \frac{u}{w} = \frac{X}{Z}$$

$$\hat{v} = \frac{v}{w} = \frac{Y}{Z}$$

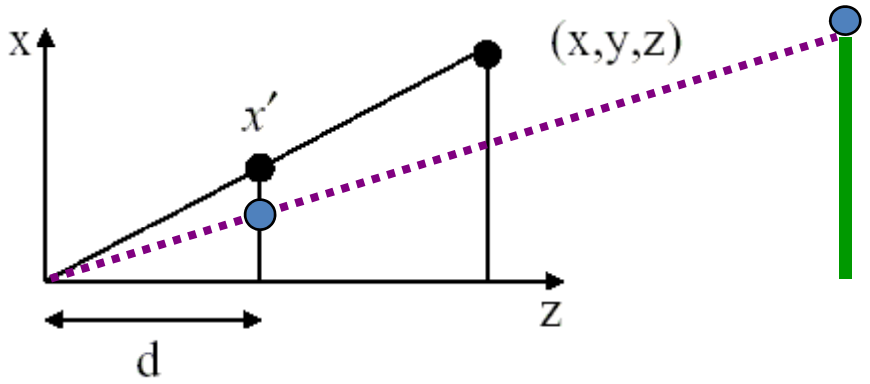
Perspective projection

- Recall perspective projection



Using similar triangles gives:

http://commons.wikimedia.org/wiki/File:Taiwan_HighSpeedRail_Train_Business_Class_Car.JPG



Affine camera model

Pretend depth is constant (often OK !)

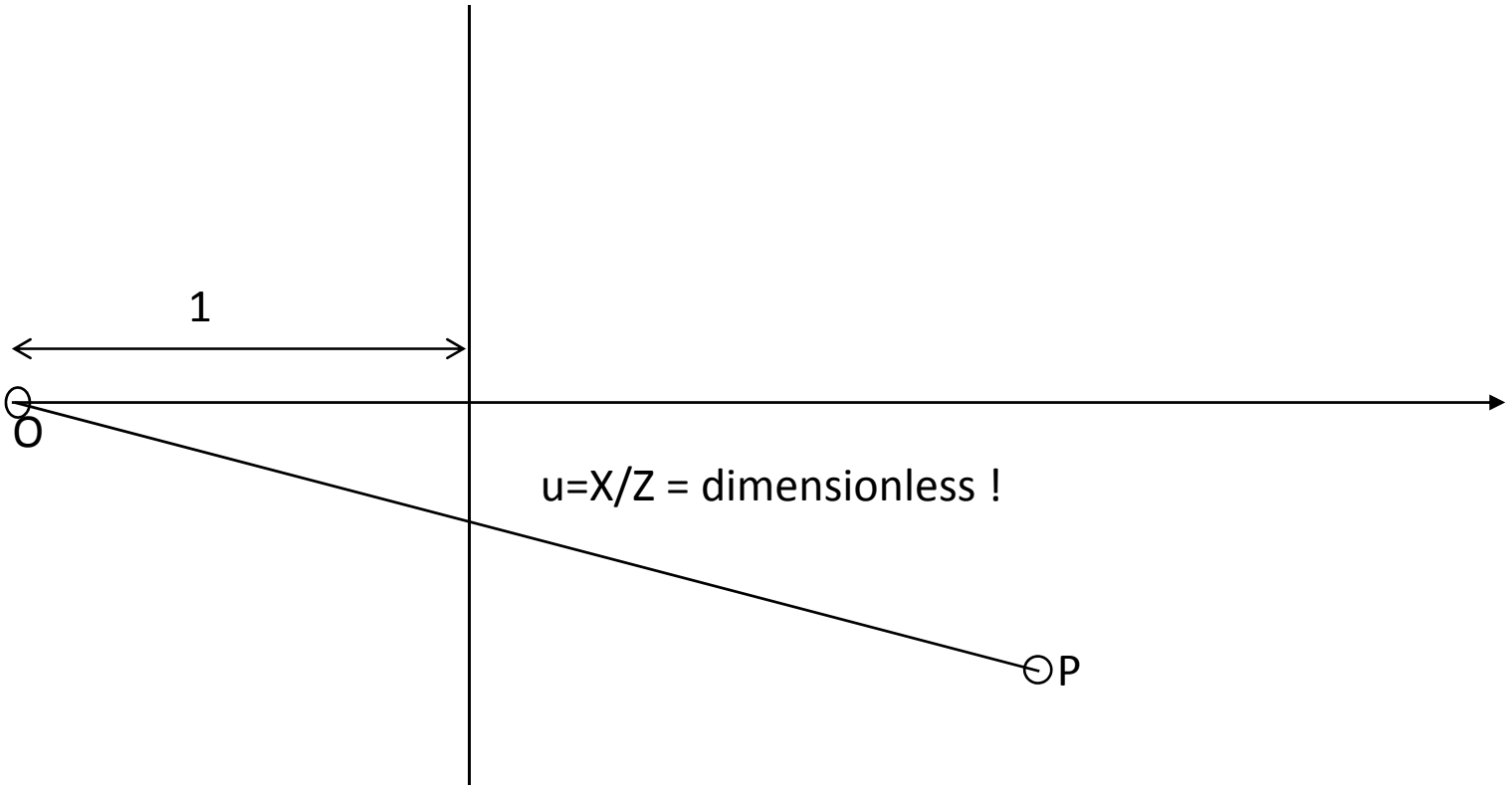
$$\hat{u}=\frac{X}{Z_r}$$

$$\hat{v}=\frac{Y}{Z_r}$$

Can also be written as a linear transformation :

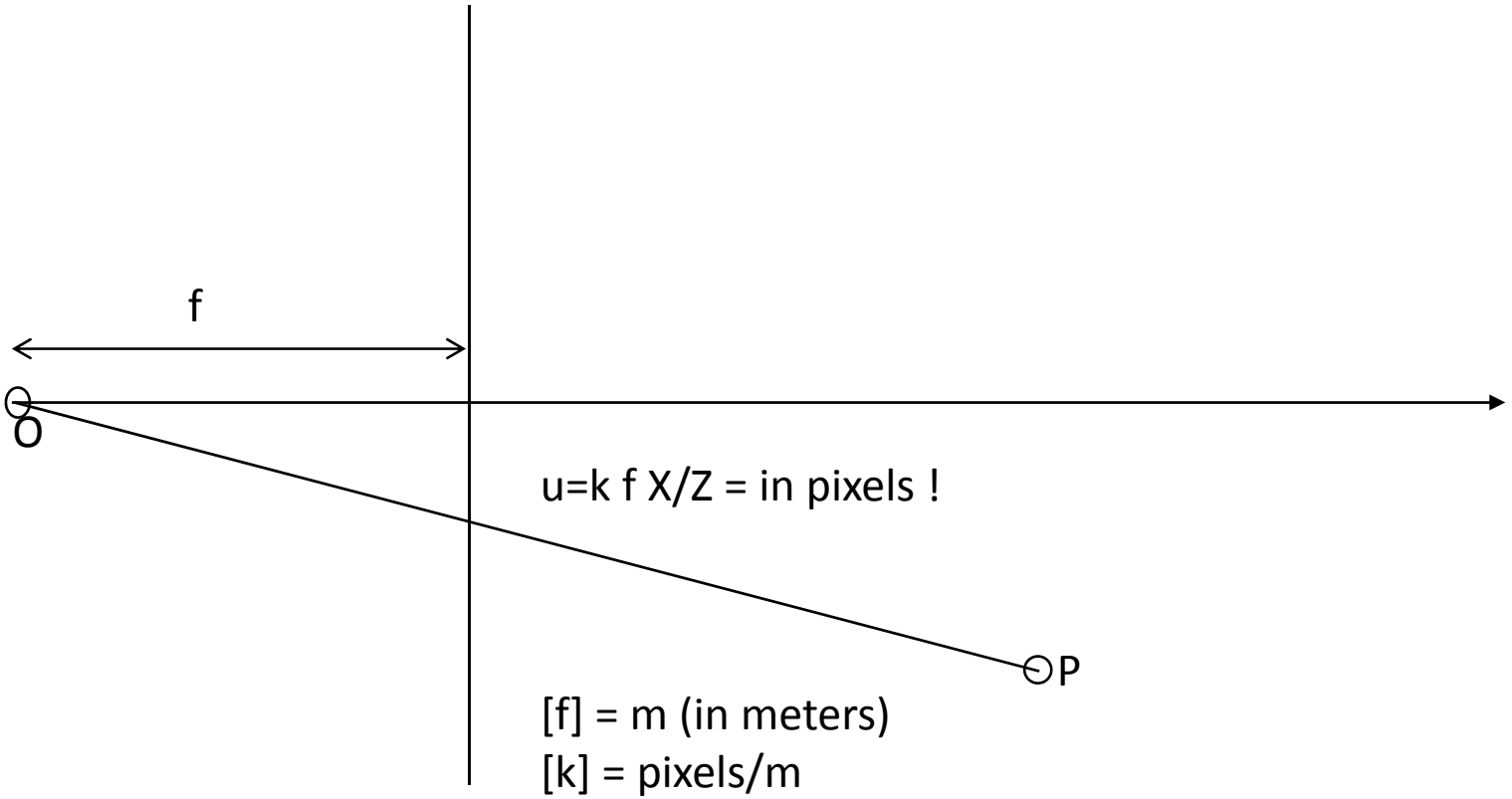
$$\begin{bmatrix} u \\ v \\ w \end{bmatrix} = \frac{1}{Z_r}\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & Z_r \end{bmatrix}\begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

Normalized Image coordinates



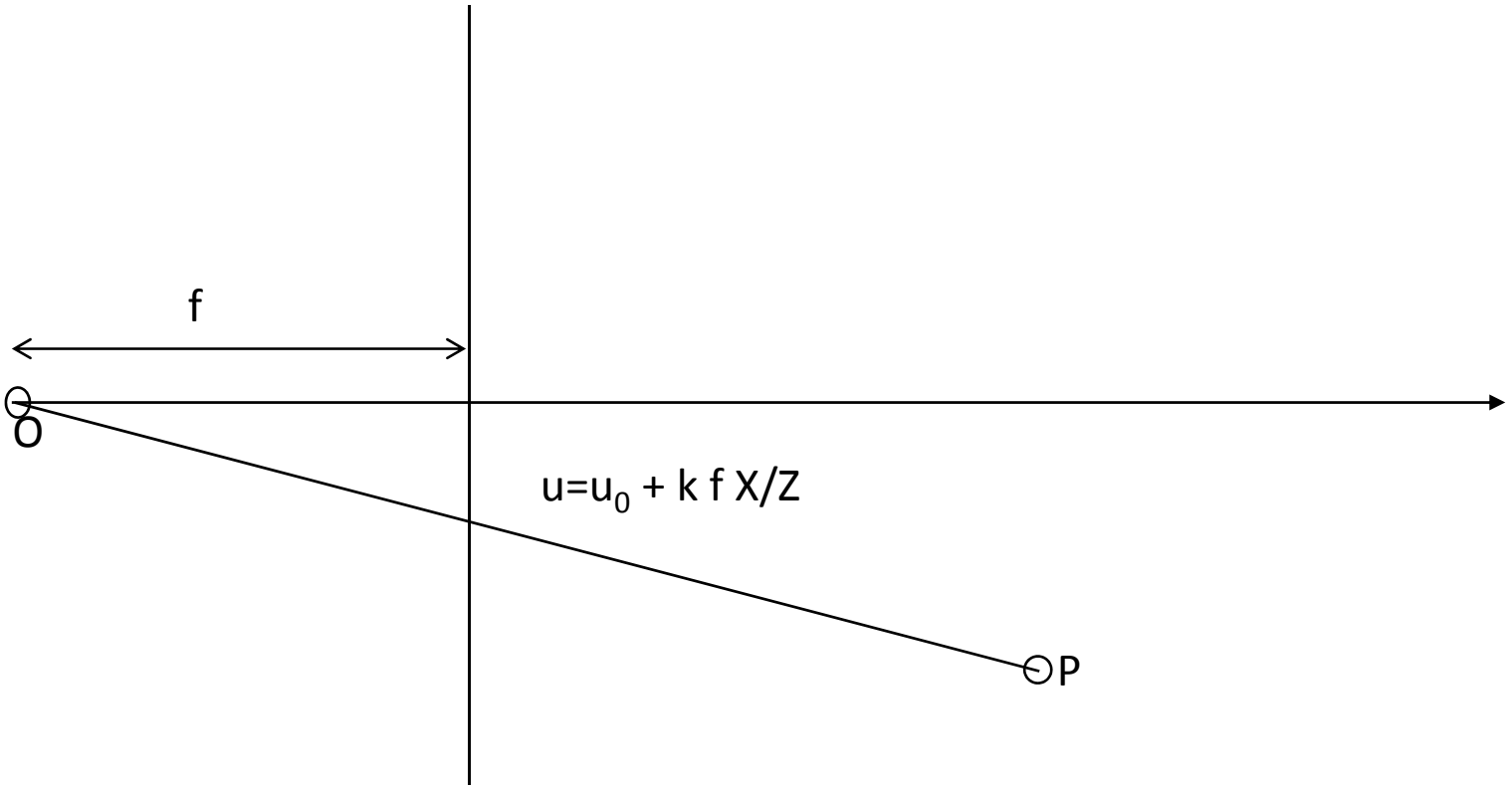
Pixel units

Pixels are on a grid of a certain dimension

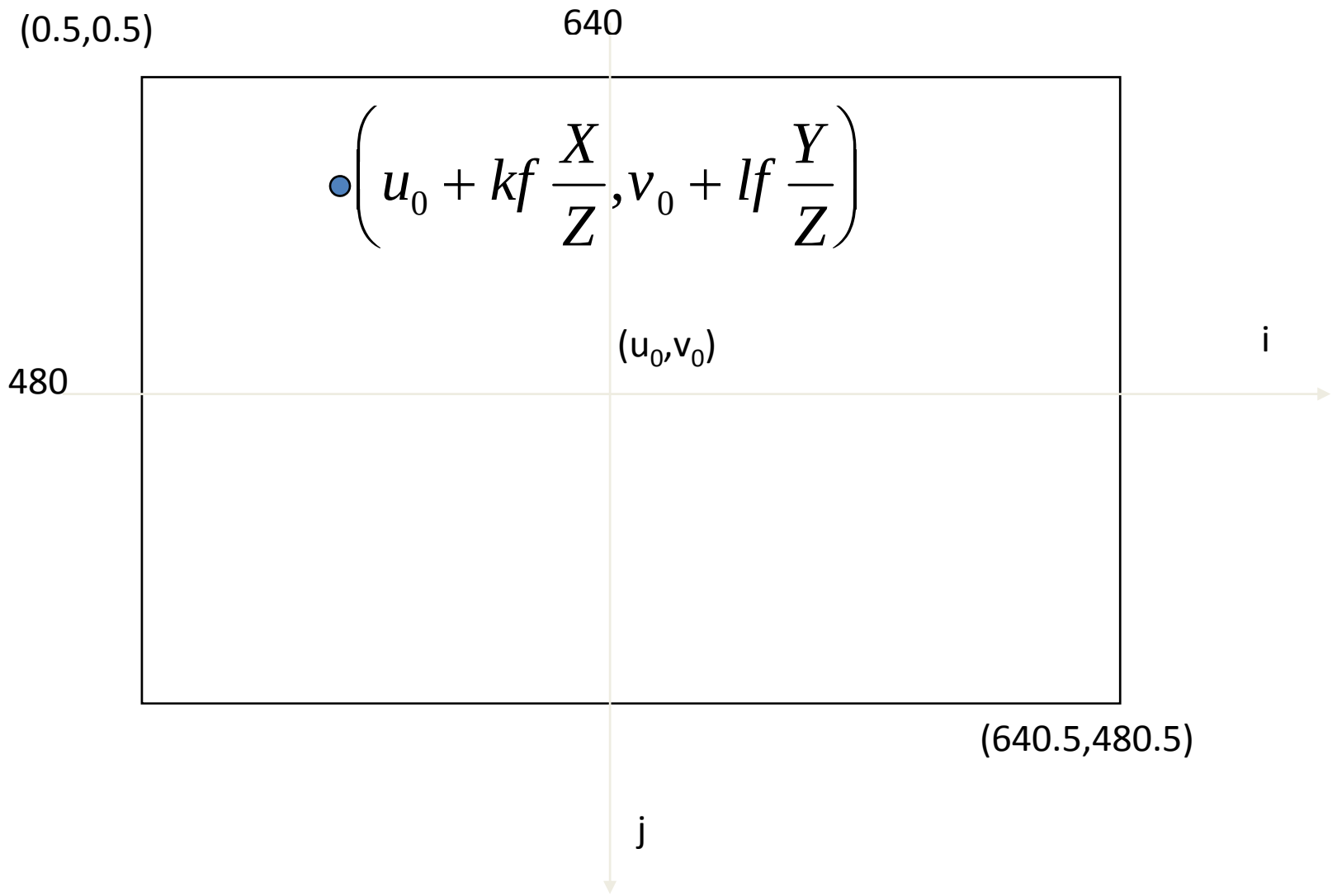


Pixel coordinates

We put the pixel coordinate origin on topleft



Pixel coordinates in 2D



Intrinsic parameters (in references)

3×3 Calibration Matrix K

$$p=\begin{bmatrix} u \\ v \\ w \end{bmatrix} = K[I \quad 0]P = \begin{bmatrix} \alpha & s & u_0 \\ & \beta & v_0 \\ & & 1 \end{bmatrix}\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}\begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

skew

Recover image (Euclidean) coordinates by normalizing :

$$\hat{u}=\frac{u}{w}=\frac{\alpha X + sY}{Z}+u_0$$

$$\hat{v}=\frac{v}{w}=\frac{\beta Y}{Z}+v_0$$

5 Degrees of Freedom !

Intrinsic parameters (in the textbook)

3×3 Calibration Matrix K

$$p=\begin{bmatrix} u \\ v \\ 1 \end{bmatrix}=\frac{1}{z}K[I \quad 0]P=\begin{bmatrix} \alpha & -\alpha\cot\theta & u_0 \\ & \frac{\beta}{\sin\theta} & v_0 \\ & & 1 \end{bmatrix}\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}\begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$$

Recover image (Euclidean) coordinates by normalizing :

$$u=\frac{\alpha x-\alpha\cot\theta}{z}+u_0$$

$$v=\frac{\beta y}{z\sin\theta}+v_0$$

Combining intrinsic and extrinsic param.

- Perspective projection mapping (including intrinsic and extrinsic parameters).

$$p = \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \frac{1}{z} K [I \quad 0] TP = \begin{bmatrix} \alpha & -\alpha \cot\theta & u_0 \\ & \frac{\beta}{\sin\theta} & v_0 \\ & & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} {}^{c}_{w}R & {}^{c}_{w}O \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$$

$$= \frac{1}{z} K \begin{bmatrix} {}^{c}_{w}R & {}^{c}_{w}O \end{bmatrix} P = \frac{1}{z} MP$$

5+6 DOF = 11 !