Image Formation

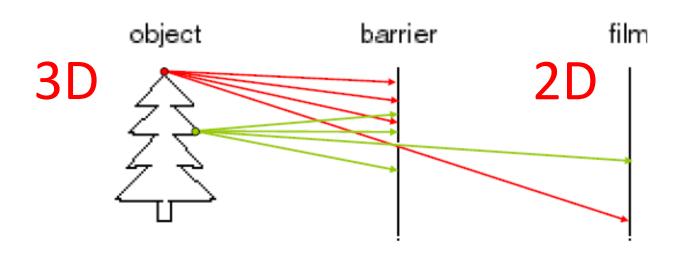
Lecture 02

Saket Anand

Outline

- Pinhole camera model
 - Need for Geometric Transformations
- Geometric Transformations 2D and 3D
 - -Scaling, Rotation and Translation
 - Homogeneous Coordinates
 - Perspective Projection
- Image Formation Process

Pinhole camera



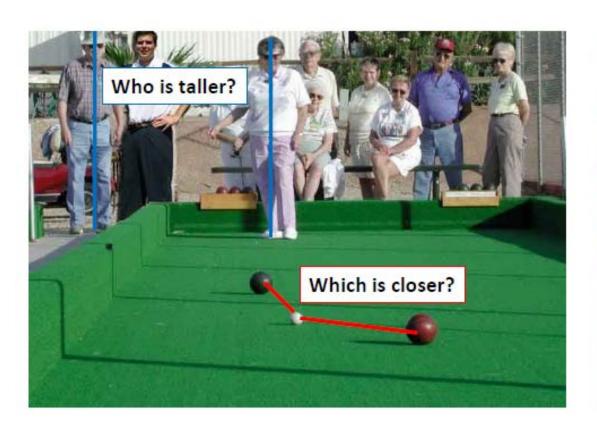
Pinhole model:

- Captures pencil of rays all rays through a single point
- The point is called Center of Projection (focal point)
- The image is formed on the Image Plane
- Add a barrier to block off most of the rays
 - This reduces blurring
 - The opening known as the aperture

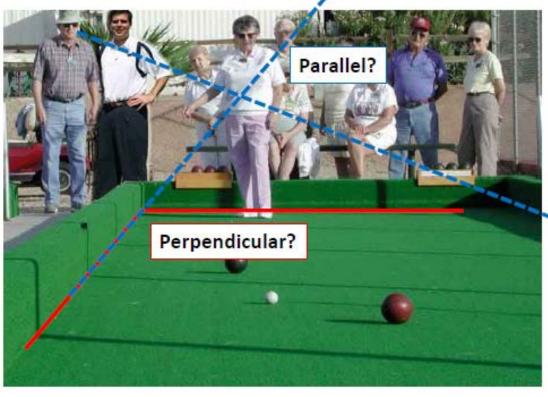
Object to Image (3D to 2D) transformation occurs via a **perspective projection**

Information Loss in Perspective Projection

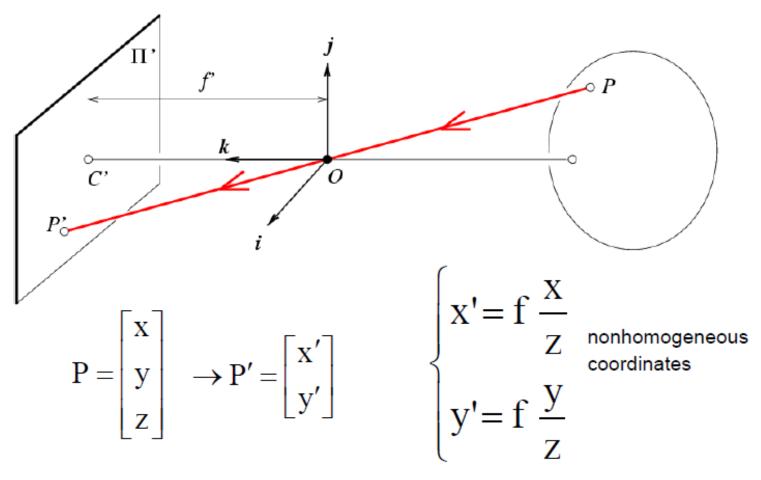
Length and Distances



Angles

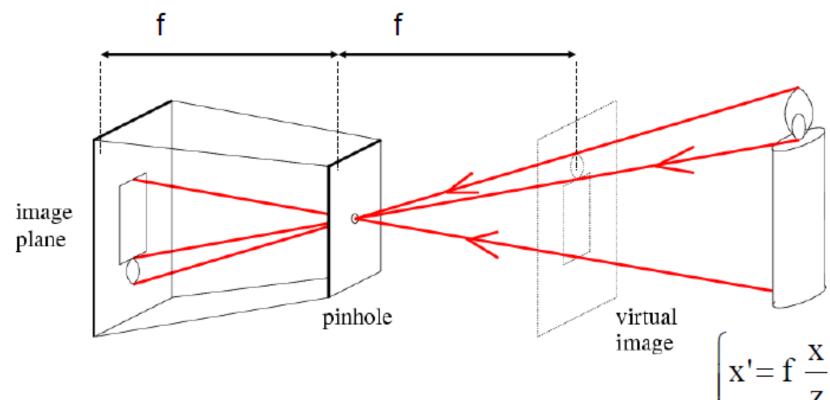


Pinhole Camera Model



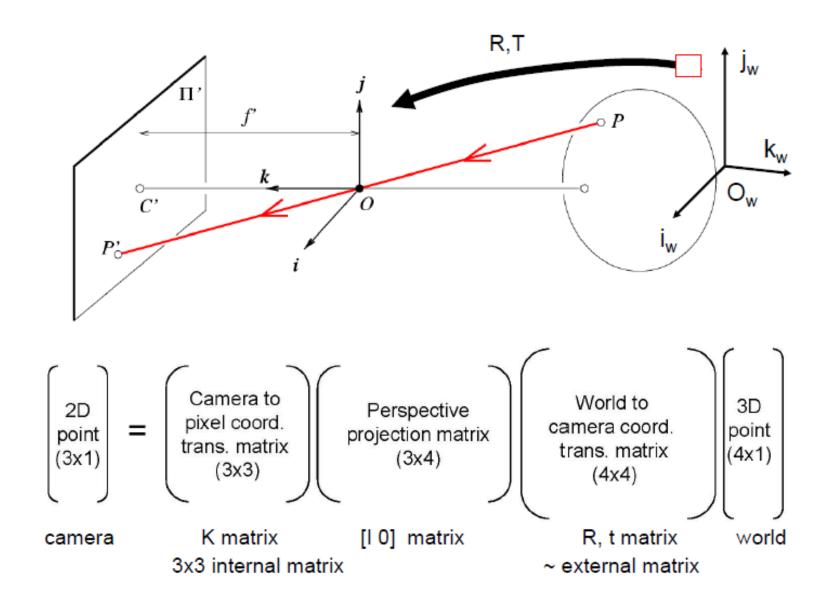
Derived using similar triangles

Pinhole Camera Model



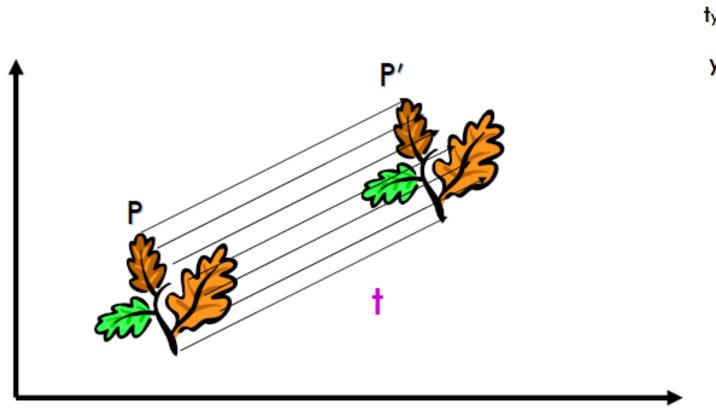
Common to draw image plane *in front* of the focal point. Moving the image plane merely scales the image.

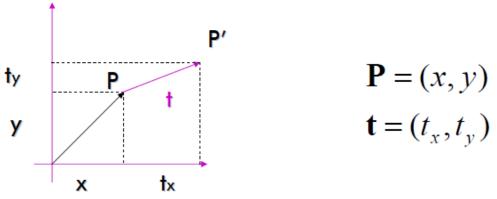
World to Camera – Camera to World



Geometric Transformations

2D Translation





$$\mathbf{P'} = \mathbf{P} + \mathbf{t} = (\mathbf{x} + \mathbf{t}_{\mathbf{x}}, \mathbf{y} + \mathbf{t}_{\mathbf{y}})$$

$$\mathbf{P'} \to \begin{bmatrix} x + t_x \\ y + t_y \end{bmatrix} = \begin{bmatrix} 1 & 0 & t_x \\ 0 & 1 & t_y \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

Homogeneous Coordinates

 Multiply the coordinates by a non-zero scalar and add an extra coordinate equal to that scalar. For example,

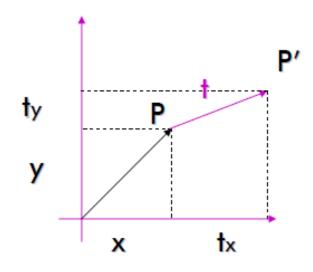
$$(x, y) \rightarrow (x \cdot z, y \cdot z, z) \quad z \neq 0$$

 $(x, y, z) \rightarrow (x \cdot w, y \cdot w, z \cdot w, w) \quad w \neq 0$

Back to Cartesian coordinates

$$(x, y, z) \quad z \neq 0 \rightarrow (x/z, y/z)$$
$$(x, y, z, w) \quad w \neq 0 \rightarrow (x/w, y/w, z/w)$$

2D Translation using Homogeneous Coordinates

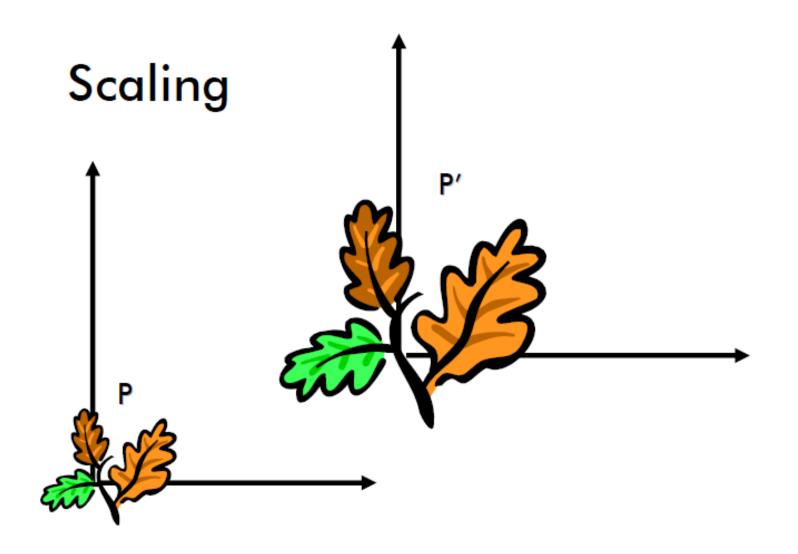


$$\mathbf{P}' \qquad \mathbf{P} = (x, y) \to (x, y, 1)$$

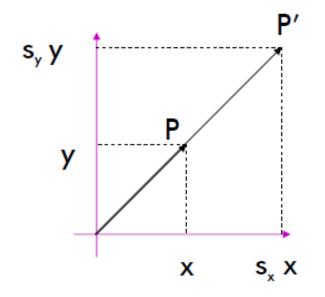
$$\mathbf{t} = (t_x, t_y) \to (t_x, t_y, 1)$$

$$\mathbf{P}' \to \begin{bmatrix} x + t_x \\ y + t_y \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & t_x \\ 0 & 1 & t_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \overline{x} \\ y \\ 1 \end{bmatrix}$$

$$= \begin{bmatrix} \mathbf{I} & \mathbf{t} \\ 0 & 1 \end{bmatrix} \cdot \mathbf{P} = \mathbf{T} \cdot \mathbf{P}$$



Scaling Equation



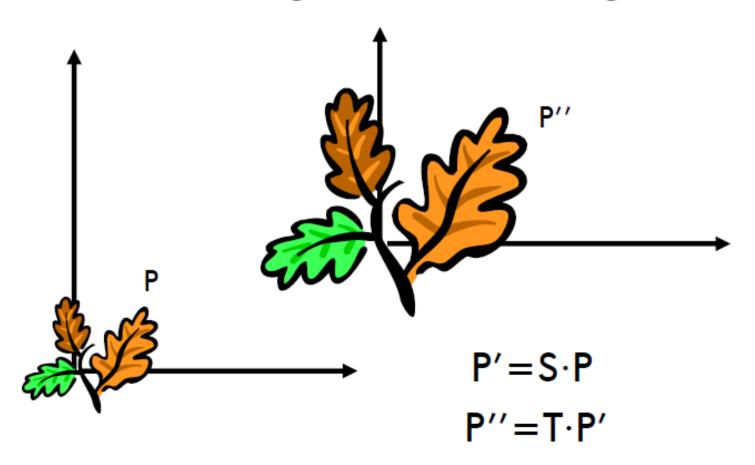
$$\mathbf{P} = (x, y) \rightarrow \mathbf{P'} = (s_x x, s_y y)$$

$$\mathbf{P} = (x, y) \to (x, y, 1)$$

$$\mathbf{P'} = (s_x x, s_y y) \to (s_x x, s_y y, 1)$$

$$\mathbf{P'} \rightarrow \begin{bmatrix} s_x x \\ s_y y \\ 1 \end{bmatrix} = \begin{bmatrix} s_x & 0 & 0 \\ 0 & s_y & 0 \\ 0 & 0 & 1 \end{bmatrix} \cdot \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} \mathbf{S'} & \mathbf{0} \\ \mathbf{0} & \mathbf{1} \end{bmatrix} \cdot \mathbf{P} = \mathbf{S} \cdot \mathbf{P}$$

Scaling & Translating



$$P''=T \cdot P'=T \cdot (S \cdot P)=(T \cdot S) \cdot P=A \cdot P$$

Scaling & Translating

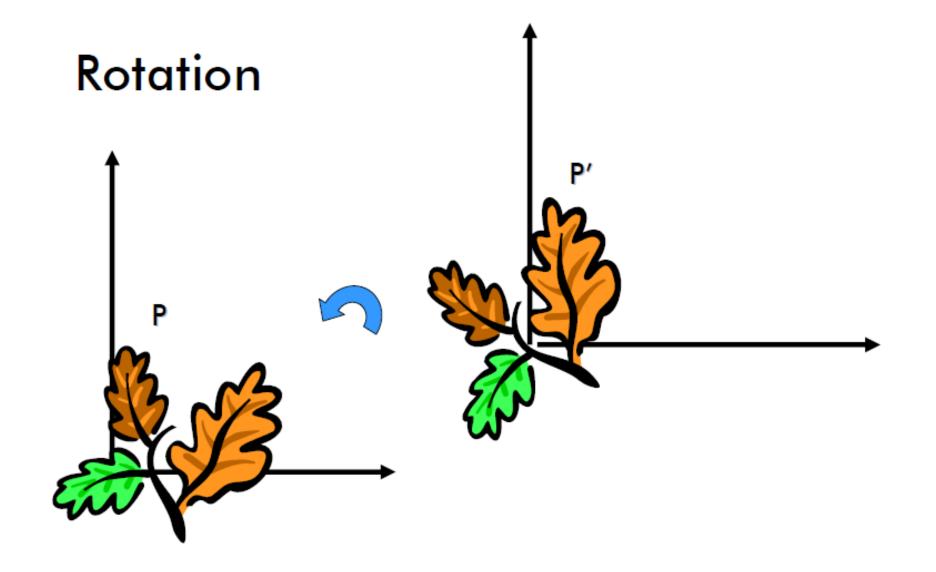
$$\mathbf{P''} = \mathbf{T} \cdot \mathbf{S} \cdot \mathbf{P} = \begin{bmatrix} 1 & 0 & t_{x} \\ 0 & 1 & t_{y} \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} s_{x} & 0 & 0 \\ 0 & s_{y} & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} s_{x} & 0 & t_{x} \\ 0 & s_{y} & t_{y} \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} s_{x}x + t_{x} \\ s_{y}y + t_{y} \\ 1 \end{bmatrix}$$

Translating & Scaling = Scaling & Translating ?

$$\mathbf{P}''' = \mathbf{T} \cdot \mathbf{S} \cdot \mathbf{P} = \begin{bmatrix} 1 & 0 & t_x \\ 0 & 1 & t_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} s_x & 0 & 0 \\ 0 & s_y & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} s_x & 0 & t_x \\ 0 & s_y & t_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} s_x x + t_x \\ s_y y + t_y \\ 1 \end{bmatrix}$$

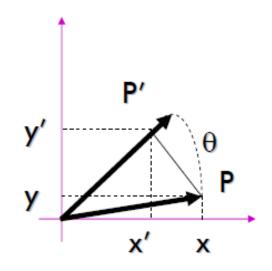
$$\mathbf{P'''} = \mathbf{S} \cdot \mathbf{T} \cdot \mathbf{P} = \begin{bmatrix} \mathbf{s}_{\mathbf{x}} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{s}_{\mathbf{y}} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & 1 \end{bmatrix} \begin{bmatrix} 1 & \mathbf{0} & \mathbf{t}_{\mathbf{x}} \\ \mathbf{0} & 1 & \mathbf{t}_{\mathbf{y}} \\ \mathbf{0} & \mathbf{0} & 1 \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \mathbf{y} \\ 1 \end{bmatrix} =$$

$$= \begin{bmatrix} \mathbf{s}_{\mathbf{x}} & \mathbf{0} & \mathbf{s}_{\mathbf{x}} \mathbf{t}_{\mathbf{x}} \\ \mathbf{0} & \mathbf{s}_{\mathbf{y}} & \mathbf{s}_{\mathbf{y}} \mathbf{t}_{\mathbf{y}} \\ \mathbf{0} & \mathbf{0} & \mathbf{1} \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \mathbf{y} \\ \mathbf{1} \end{bmatrix} = \begin{bmatrix} \mathbf{s}_{\mathbf{x}} \mathbf{x} + \mathbf{s}_{\mathbf{x}} \mathbf{t}_{\mathbf{x}} \\ \mathbf{s}_{\mathbf{y}} \mathbf{y} + \mathbf{s}_{\mathbf{y}} \mathbf{t}_{\mathbf{y}} \\ \mathbf{1} \end{bmatrix}$$



Rotation Equations

Counter-clockwise rotation by an angle θ



$$x' = \cos \theta x - \sin \theta y$$
$$y' = \cos \theta y + \sin \theta x$$

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

$$P' = R P$$

Degrees of Freedom

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

Note: R belongs to the category of *normal* matrices and satisfies many interesting properties:

$$\mathbf{R} \cdot \mathbf{R}^{\mathsf{T}} = \mathbf{R}^{\mathsf{T}} \cdot \mathbf{R} = \mathbf{I}$$
$$\det(\mathbf{R}) = 1$$

Rotation + Scaling +Translation

$$P' = (T R S) P$$

$$\mathbf{P'} = \mathbf{T} \cdot \mathbf{R} \cdot \mathbf{S} \cdot \mathbf{P} = \begin{bmatrix} 1 & 0 & t_{x} \\ 0 & 1 & t_{y} \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \cos \theta & -\sin \theta & 0 \\ \sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} s_{x} & 0 & 0 \\ 0 & s_{y} & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} =$$

$$= \begin{bmatrix} \cos \theta & -\sin \theta & t_x \\ \sin \theta & \cos \theta & t_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} s_x & 0 & 0 \\ 0 & s_y & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} =$$

$$= \begin{bmatrix} R' & t \\ 0 & 1 \end{bmatrix} \begin{bmatrix} S & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} X \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} R'S & t \\ 0 & 1 \end{bmatrix} \begin{bmatrix} X \\ y \\ 1 \end{bmatrix}$$

If $s_x = s_y$, this is a similarity transformation!

Geometric Transformations

Isometries or Euclidean Transformation

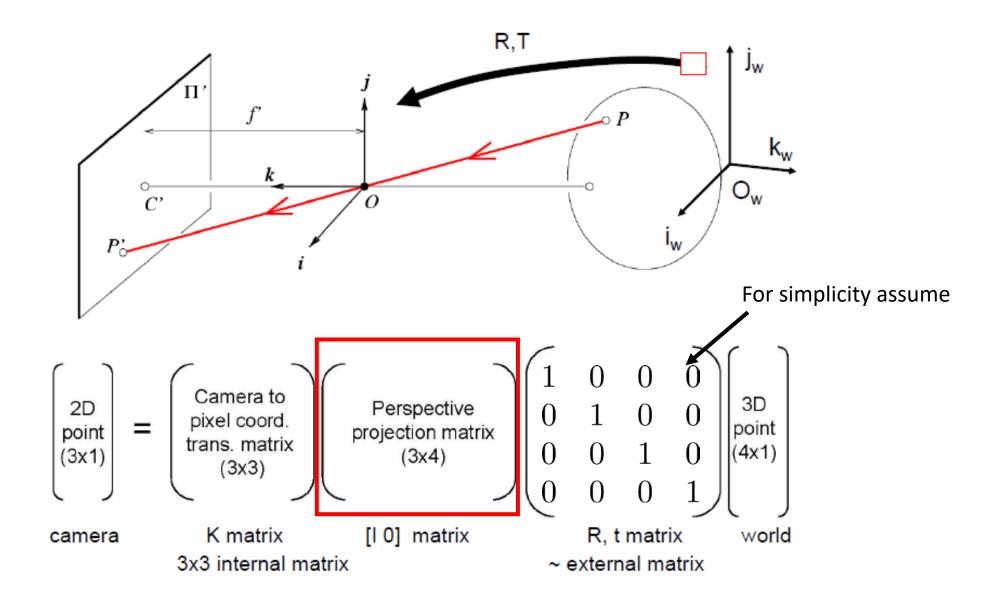
Similarity Transformation

Affine Transformation

Projective Transformations

Camera Coordinates and Image Formation

World to Camera – Camera to World



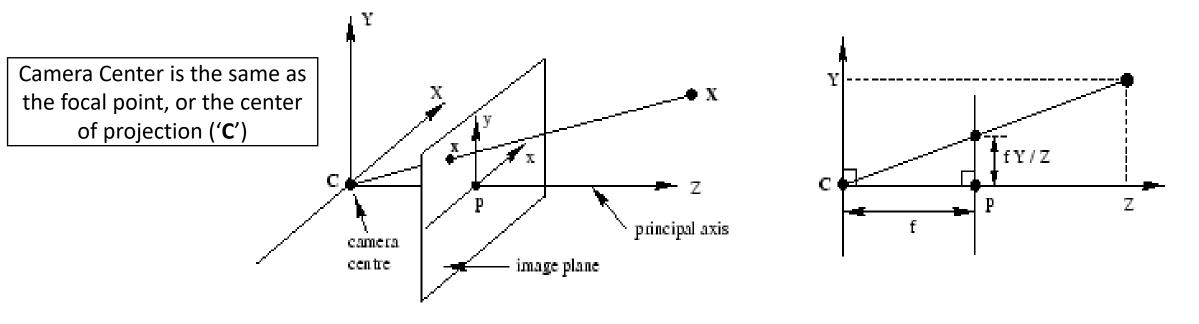
Perspective Projection Transformation

$$\mathbf{X}' = \begin{bmatrix} \mathbf{x} \\ \mathbf{y} \\ \mathbf{z} \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \mathbf{y} \\ \mathbf{z} \\ 1 \end{bmatrix} \qquad \mathbf{X}' = \mathbf{M} \mathbf{X}$$

$$\begin{array}{ccc} & & & \frac{X}{Z} \\ & & & \\ & & \frac{y}{Z} \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ \end{array}$$

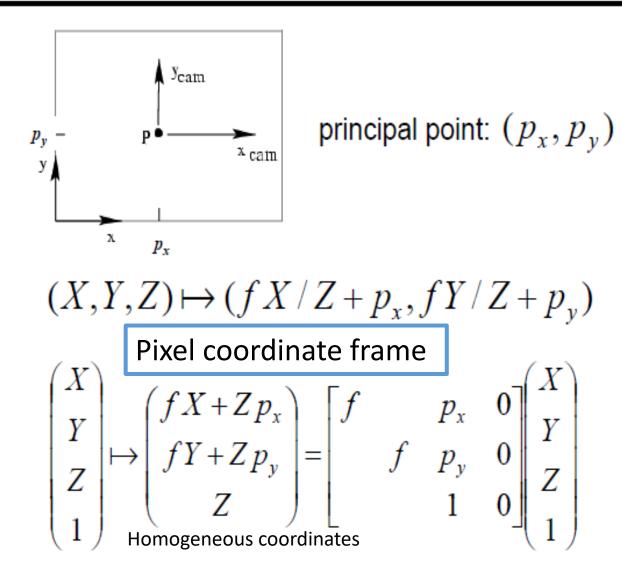
X point in 3D homogeneous c. X' point in 2D homogeneous c.

Pinhole Camera Model

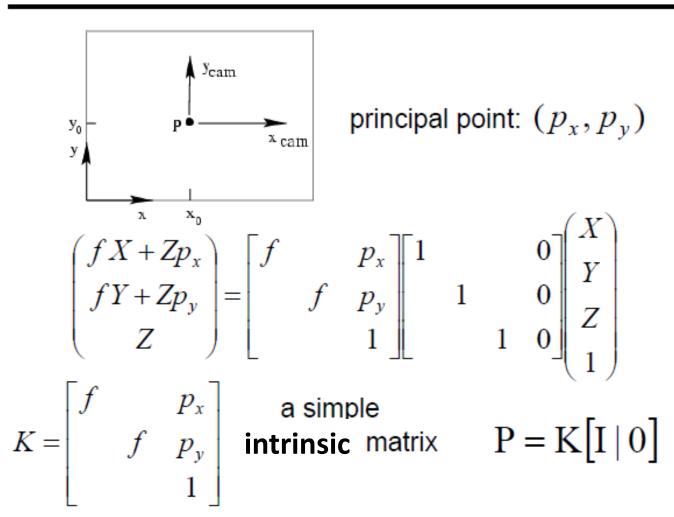


- **Principal Axis** Line from the camera centre perpendicular to the image plane.
- Normalized Camera Coordinate Camera centre at the origin (C), x and y axes are aligned with the image axes and image plane ($P_z = f$); units in m/cm/ft etc.
- **Principal point** point where the principal axis intersects the image plane (z=f)
- Pixel coordinate frame Origin (0,0) is in the corner of an image; units are in pixels.

Principal point offset

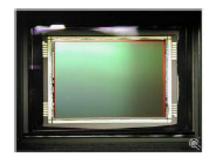


Principal point offset



Pixel coordinates





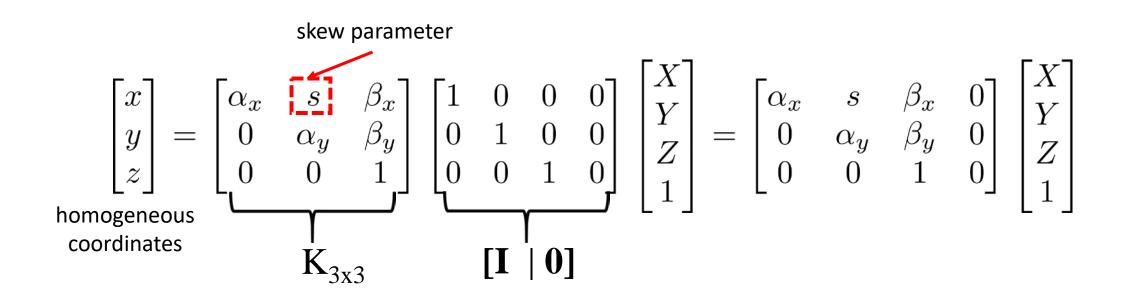
Pixel size:
$$\frac{1}{m_x} \times \frac{1}{m_y}$$

 m_{x} pixels per meter in horizontal direction m_{y} pixels per meter in vertical direction

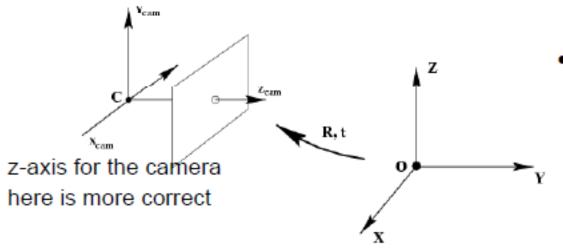
$$K = \begin{bmatrix} m_x & & \\ & m_y & \\ & 1 \end{bmatrix} \begin{bmatrix} f & & p_x \\ & f & p_y \\ & & 1 \end{bmatrix} = \begin{bmatrix} \alpha_x & & \beta_x \\ & \alpha_y & \beta_y \\ & & 1 \end{bmatrix}$$
pixels/m m pixels

Intrinsic Camera Matrix

- Skew parameter *s* is non-zero, only if x and y axes are non-orthogonal, i.e., pixels are not rectangular
- Recent cameras rarely have <u>non-square</u> pixels

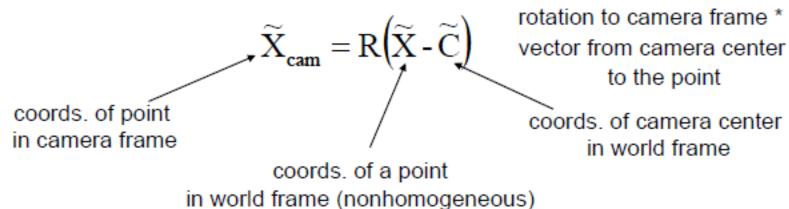


Camera rotation and translation



In general, the camera coordinate frame will be related to the world coordinate frame by a rotation and a translation

3D nonhomogeneous



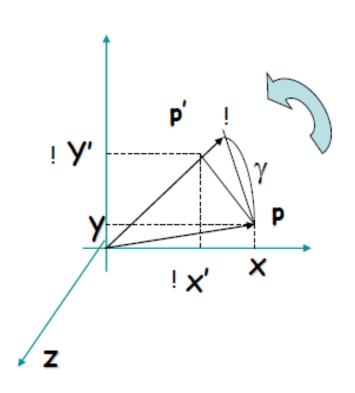
3D Rotation Matrices

- Representations
 - Euler angles
 - -Axis-angle
 - -Rodrigues' formula
 - Unit quaternion

- Important to note:
 - -Points along the axis of rotation are invariant to rotation

Euler Angles

3D Rotation around the coordinate axes counter-clockwise.



$$R_{x}(\alpha) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos \alpha & -\sin \alpha \\ 0 & \sin \alpha & \cos \alpha \end{bmatrix}$$

$$R_{y}(\beta) = \begin{bmatrix} \cos \beta & 0 & \sin \beta \\ 0 & 1 & 0 \\ -\sin \beta & 0 & \cos \beta \end{bmatrix}$$

$$R_{z}(\gamma) = \begin{bmatrix} \cos \gamma & -\sin \gamma & 0 \\ \sin \gamma & \cos \gamma & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Axis Angle

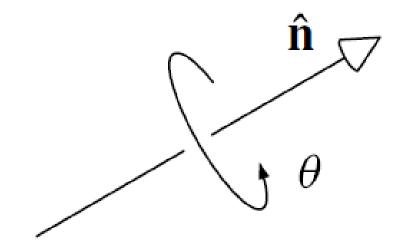
Let $rot(\hat{\mathbf{n}}, \theta)$ be the corresponding rotation.

Many to one:

$$rot(-\hat{\mathbf{n}}, -\theta) = rot(\hat{\mathbf{n}}, \theta)$$

 $rot(\hat{\mathbf{n}}, \theta + 2k\pi) = rot(\hat{\mathbf{n}}, \theta)$, for any integer k.

When $\theta = 0$, the rotation axis is indeterminate, giving an infinity-to-one mapping.

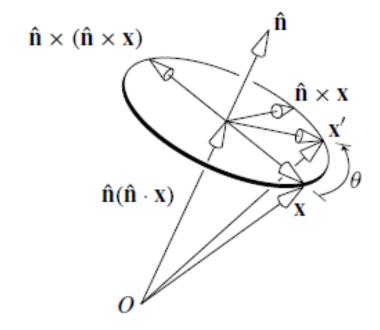


Rodrigues's Formula

Given point x, decompose into components parallel and perpendicular to the rotation axis

$$\mathbf{x} = \hat{\mathbf{n}}(\hat{\mathbf{n}} \cdot \mathbf{x}) - \hat{\mathbf{n}} \times (\hat{\mathbf{n}} \times \mathbf{x})$$

Only \mathbf{x}_{\perp} is affected by the rotation, yielding *Rodrigues's* formula:



Rodrigues's Formula

Axis-angle to R

$$\mathbf{x}' = \mathbf{x} + (\sin \theta) \ \hat{\mathbf{n}} \times \mathbf{x} + (1 - \cos \theta) \ \hat{\mathbf{n}} \times (\hat{\mathbf{n}} \times \mathbf{x})$$

$$N = \left(\begin{array}{ccc} 0 & -n_3 & n_2 \\ n_3 & 0 & -n_1 \\ -n_2 & n_1 & 0 \end{array} \right)$$

$$N\mathbf{x} = \hat{\mathbf{n}} \times \mathbf{x}$$

$$\mathbf{x}' = \mathbf{x} + (\sin \theta) N \mathbf{x} + (1 - \cos \theta) N^2 \mathbf{x}$$

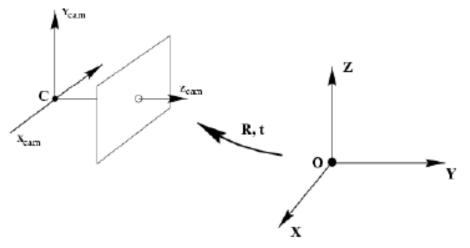
$$R = I + (\sin \theta)N + (1 - \cos \theta)N^2$$

R to Axis-angle

$$\theta = \cos^{-1}\left(\frac{trace(\mathbf{R}) - 1}{2}\right)$$

$$n = rac{1}{2\sin(heta)} egin{bmatrix} m{R}_{32} - m{R}_{23} \ m{R}_{13} - m{R}_{31} \ m{R}_{21} - m{R}_{12} \end{bmatrix}$$

Camera rotation and translation



In nonhomogeneous coordinates:

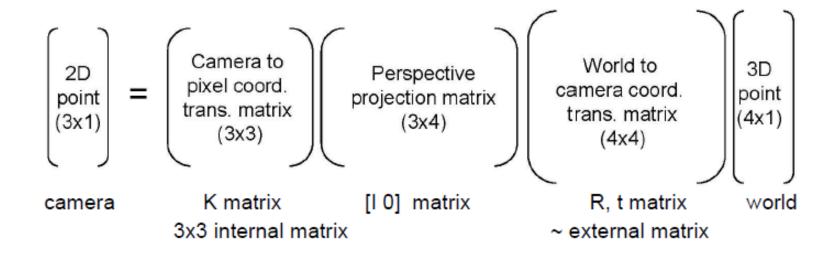
$$\widetilde{X}_{cam} = R(\widetilde{X} - \widetilde{C})$$

$$\begin{array}{ll} \textit{homogeneous coor.} \\ \textit{in 3D} & X_{cam} = \begin{bmatrix} R & -R\widetilde{C} \\ 0 & 1 \end{bmatrix} \begin{pmatrix} \widetilde{X} \\ 1 \end{pmatrix} = \begin{bmatrix} R & -R\widetilde{C} \\ 0 & 1 \end{bmatrix} X \\ \end{array}$$

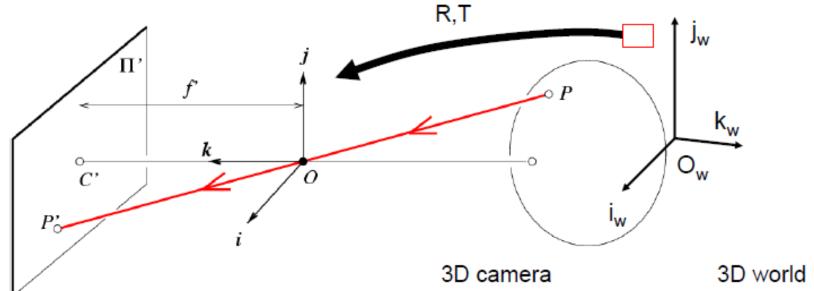
in 2D plane
$$x = K \Big[I \, | \, 0 \Big] X_{cam} = K \Big[R \, | \, -R \widetilde{C} \Big] X \qquad \qquad P = K \Big[R \, | \, t \Big], \qquad t = -R \widetilde{C}$$

Note: C is the null space of the camera projection matrix (PC=0) $K[R \mid -R\tilde{C}] [\tilde{C} \mid 1]^T = 0$

Image Formation Summary



- Rotation and Translation (world to camera)
- Perspective Projection (camera 3D to image plane 2D)
- Scaling and Shifting (image plane 2D to pixel 2D)
- Use of homogeneous coordinates



In homogeneous coordinates

$$X = \begin{bmatrix} R & T \\ 0 & 1 \end{bmatrix}_{4 \times 4} X_{w}$$

We need 5 + 3 + 3 = 11 degrees of freedom (DOF) maximum.

Camera Parameters

- Intrinsic
 - Focal length (2-dof) f_x and f_y
 - Principal point (2-dof) c_{x} , c_{y}
 - Skew factor (1-dof) s
- Extrinsic
 - Rotation (3-dof) R
 - Translation (3-dof) \mathbf{t}
- Total degrees of freedom: 3 + 3 + 2 + 2 + 1 = 11
 - Need 11 equations to estimate these parameters
 - Direct Linear Transform
 - Calibration toolboxes

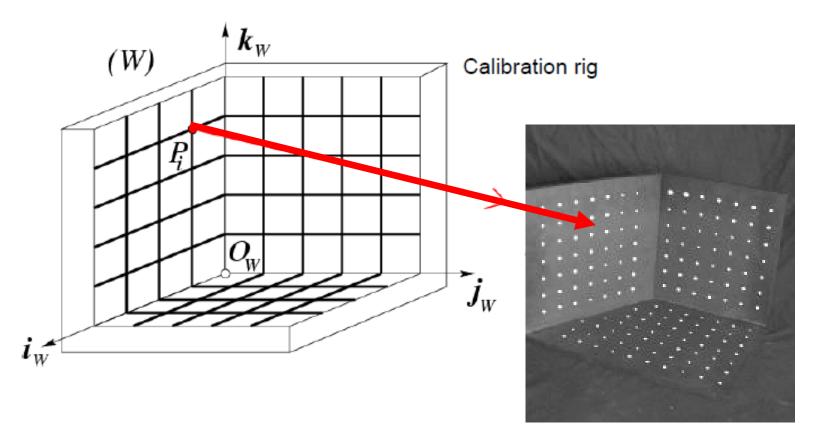
Camera Calibration

Projective Camera

$$P' = M P_{W} = K R T P_{W}$$
Internal parameters
$$External parameters$$

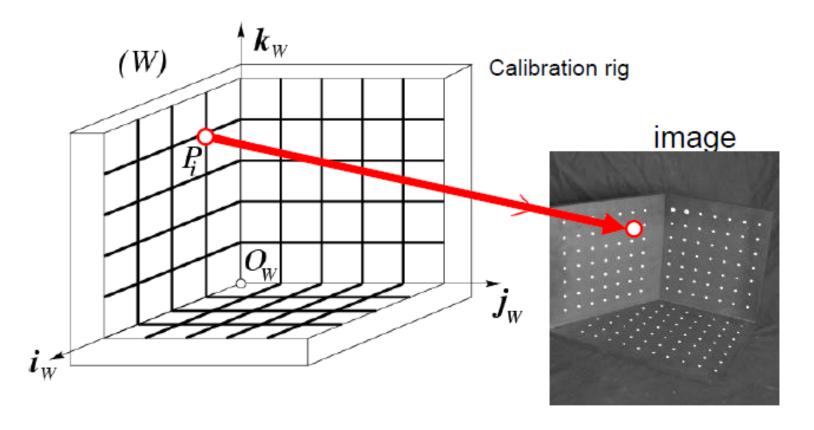
$$K = \begin{bmatrix} \alpha & -\alpha \cot \theta & u_{o} \\ 0 & \frac{\beta}{\sin \theta} & v_{o} \\ 0 & 0 & 1 \end{bmatrix} \qquad R = \begin{bmatrix} \mathbf{r}_{1}^{T} \\ \mathbf{r}_{2}^{T} \\ \mathbf{r}_{3}^{T} \end{bmatrix} \qquad T = \begin{bmatrix} t_{x} \\ t_{y} \\ t_{z} \end{bmatrix}$$

Goal: To estimate internal and external parameters from one or multiple images

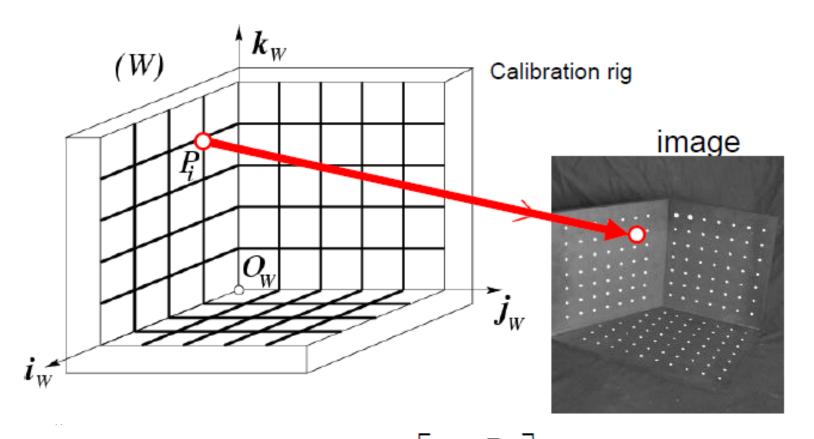


- •P₁... P_n with known positions in $[O_w, i_w, j_w, k_w]$
- •p₁, ... p_n known positions in the image

Goal: compute intrinsic and extrinsic parameters



In practice, using more than 6 correspondences enables more robust results



$$P_{i} \rightarrow M \ P_{i} \rightarrow p_{i} = \begin{bmatrix} u_{i} \\ v_{i} \end{bmatrix} = \begin{bmatrix} \frac{\mathbf{m}_{1} P_{i}}{\mathbf{m}_{3} P_{i}} \\ \frac{\mathbf{m}_{2} P_{i}}{\mathbf{m}_{3} P_{i}} \end{bmatrix} \qquad M = \begin{bmatrix} \mathbf{m}_{1} P_{i} \\ \mathbf{m}_{2} P_{i} \\ \mathbf{m}_{3} P_{i} \end{bmatrix}$$

$$\begin{cases} -u_{1}(\mathbf{m}_{3} P_{1}) + \mathbf{m}_{1} P_{1} = 0 \\ -v_{1}(\mathbf{m}_{3} P_{1}) + \mathbf{m}_{2} P_{1} = 0 \\ \vdots \\ -u_{n}(\mathbf{m}_{3} P_{n}) + \mathbf{m}_{1} P_{n} = 0 \\ -v_{n}(\mathbf{m}_{3} P_{n}) + \mathbf{m}_{2} P_{n} = 0 \end{cases}$$
Homogenous

Homogenous linear system

known

unknown

$$\mathbf{P} \stackrel{\text{def}}{=} \begin{pmatrix} \mathbf{P}_{1}^{T} & \mathbf{0}^{T} & -u_{1} \mathbf{P}_{1}^{T} \\ \mathbf{0}^{T} & \mathbf{P}_{1}^{T} & -v_{1} \mathbf{P}_{1}^{T} \\ \vdots & & & \\ \mathbf{P}_{n}^{T} & \mathbf{0}^{T} & -u_{n} \mathbf{P}_{n}^{T} \\ \mathbf{0}^{T} & \mathbf{P}_{n}^{T} & -u_{n} \mathbf{P}_{n}^{T} \end{pmatrix}_{2n \times 12}$$

$$\boldsymbol{m} \stackrel{\text{def}}{=} \begin{pmatrix} \mathbf{m}_{1}^{\text{T}} \\ \mathbf{m}_{2}^{\text{T}} \\ \mathbf{m}_{3}^{\text{T}} \end{pmatrix}_{12x1}$$

Decomposition of the Camera Matrix

- Camera Centre: MC=0
 - -One dimensional null space of camera matrix M

- Extrinsic Parameters (R,t) and Intrinsic Parameters (K)
 - Use RQ decomposition where R is upper triangular and Q is orthogonal

RQ Decomposition

A 3-dimensional Givens rotation is a rotation about one of the three coordinate axes. The three Givens rotations are

$$\mathbf{Q}_x = \begin{bmatrix} 1 & & \\ & c & -s \\ & s & c \end{bmatrix} \quad \mathbf{Q}_y = \begin{bmatrix} c & s \\ & 1 \\ -s & c \end{bmatrix} \quad \mathbf{Q}_z = \begin{bmatrix} c & -s \\ s & c \\ & 1 \end{bmatrix} \tag{A4.1}$$

Objective

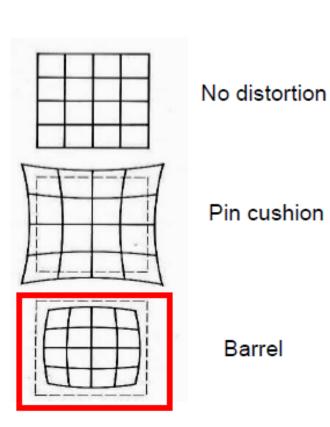
Carry out the RQ decomposition of a 3 × 3 matrix A using Givens rotations.

Algorithm

- (i) Multiply by Q_x so as to set A_{32} to zero.
- (ii) Multiply by Q_y so as to set A_{31} to zero. This multiplication does not change the second column of A, hence A_{32} remains zero.
- (iii) Multiply by Q_z so as to set A_{21} to zero. The first two columns are replaced by linear combinations of themselves. Thus, A_{31} and A_{32} remain zero.

Radial Distortion

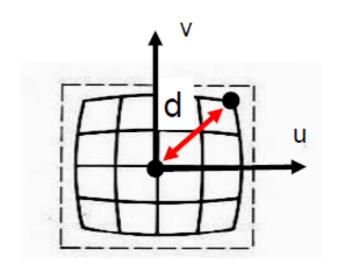
- Caused by imperfect lenses
- Deviations are most noticeable for rays that pass through the edge of the lens





Radial Distortion

Image magnification in(de)creases with distance from the optical center



$$\begin{bmatrix} \frac{1}{\lambda} & 0 & 0 \\ 0 & \frac{1}{\lambda} & 0 \\ 0 & 0 & 1 \end{bmatrix} \mathbf{M} \mathbf{P}_{i} \rightarrow \begin{bmatrix} \mathbf{u}_{i} \\ \mathbf{v}_{i} \end{bmatrix} = \mathbf{p}_{i}$$

$$d^2 = a u^2 + b v^2 + c u v$$

To model radial behavior

$$d^2 = a \ u^2 + b \ v^2 + c \ u \ v$$

$$\lambda = 1 \pm \sum_{p=1}^{3} \kappa_p d^{2p}$$
To model radial behavior

Polynomial function

Radial Distortion

$$\begin{bmatrix} \frac{1}{\lambda} & 0 & 0 \\ 0 & \frac{1}{\lambda} & 0 \\ 0 & 0 & 1 \end{bmatrix} \mathbf{M} \mathbf{P}_{i} \rightarrow \begin{bmatrix} \mathbf{u}_{i} \\ \mathbf{v}_{i} \end{bmatrix} = \mathbf{p}_{i} \qquad \mathbf{Q} = \begin{bmatrix} \mathbf{q}_{1} \\ \mathbf{q}_{2} \\ \mathbf{q}_{3} \end{bmatrix}$$

Is this a linear system of equations?

$$p_{i} = \begin{bmatrix} u_{i} \\ v_{i} \end{bmatrix} = \begin{bmatrix} \frac{\mathbf{q}_{1} P_{i}}{\mathbf{q}_{3} P_{i}} \\ \frac{\mathbf{q}_{2} P_{i}}{\mathbf{q}_{3} P_{i}} \end{bmatrix} \longrightarrow \begin{cases} u_{i} \mathbf{q}_{3} P_{i} = \mathbf{q}_{1} P_{i} \\ v_{i} \mathbf{q}_{3} P_{i} = \mathbf{q}_{2} P \end{cases}$$

$$v_{i} \mathbf{q}_{3} P_{i} = \mathbf{q}_{2} P$$
No! why?

General Calibration Problem

$$\begin{bmatrix} u_i \\ v_i \end{bmatrix} = \begin{bmatrix} \frac{\mathbf{q}_1 P_i}{\mathbf{q}_3 P_i} \\ \frac{\mathbf{q}_2 P_i}{\mathbf{q}_3 P_i} \end{bmatrix} \xrightarrow{X} X = f(P)$$
measurement parameter
$$f() \text{ is nonlinear}$$

Typical assumptions:

- zero-skew, square pixel
- u_0 , v_0 = known center of the image
- no distortion

→ Just estimate f → and R, T

Camera Calibration Toolbox for Matlab J. Bouguet - [1998-2000]

http://www.vision.caltech.edu/bouguetj/calib_doc/index.html#examples

