

Lecture 4.

Camera Models: Intrinsic Parameters

ECEN 5283 Computer Vision

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Goals

- ▶ To review the geometric transforms for **coordinate changes**.
- ▶ To introduce the **analytical machinery** necessary to establish **quantitative constraints** between 2D image measurements and 3D real-world objects.
- ▶ To define physical parameters of camera, i.e., **intrinsic parameters** and **extrinsic parameters**, and to study the role of intrinsic parameters for geometric camera modeling.

Coordinate System Changes

- ▶ When several different coordinate systems are considered at the same time, we denote the coordinate vector of the point P in the frame F as

$${}^F P = {}^F \overrightarrow{OP} = \begin{pmatrix} x \\ y \\ z \end{pmatrix} \Leftrightarrow \overrightarrow{OP} = x\mathbf{i} + y\mathbf{j} + z\mathbf{k}.$$

- ▶ Let us consider two coordinate systems (two frames)

$$(A) = (O_A, \mathbf{i}_A, \mathbf{j}_A, \mathbf{k}_A)$$

$$(B) = (O_B, \mathbf{i}_B, \mathbf{j}_B, \mathbf{k}_B)$$

- ▶ **Question:** How to express ${}^B P$ as a function of ${}^A P$.

Four Cases of Coordinate Changes

► Case I: 3D Translation

$$O_A \neq O_B, \mathbf{i}_A = \mathbf{i}_B, \mathbf{j}_A = \mathbf{j}_B, \mathbf{k}_A = \mathbf{k}_B.$$

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

► Case II: 3D Rotation

$$O_A = O_B = O, \mathbf{i}_A \neq \mathbf{i}_B, \mathbf{j}_A \neq \mathbf{j}_B, \mathbf{k}_A \neq \mathbf{k}_B.$$

$${}^B P = {}^B R {}^A P$$

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

► Case III: 2D Rotation

$$O_A = O_B = O, \mathbf{k}_A = \mathbf{k}_B = \mathbf{k}.$$

$${}^B R = \begin{pmatrix} \cos \theta & \sin \theta & 0 \\ -\sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

► Case IV: 3D Rigid Transformation

$$O_A \neq O_B, \mathbf{i}_A \neq \mathbf{i}_B, \mathbf{j}_A \neq \mathbf{j}_B, \mathbf{k}_A \neq \mathbf{k}_B.$$

$${}^B P = {}^B R {}^A P + {}^B O_A$$

Coordinate System Changes: Rigid Transform



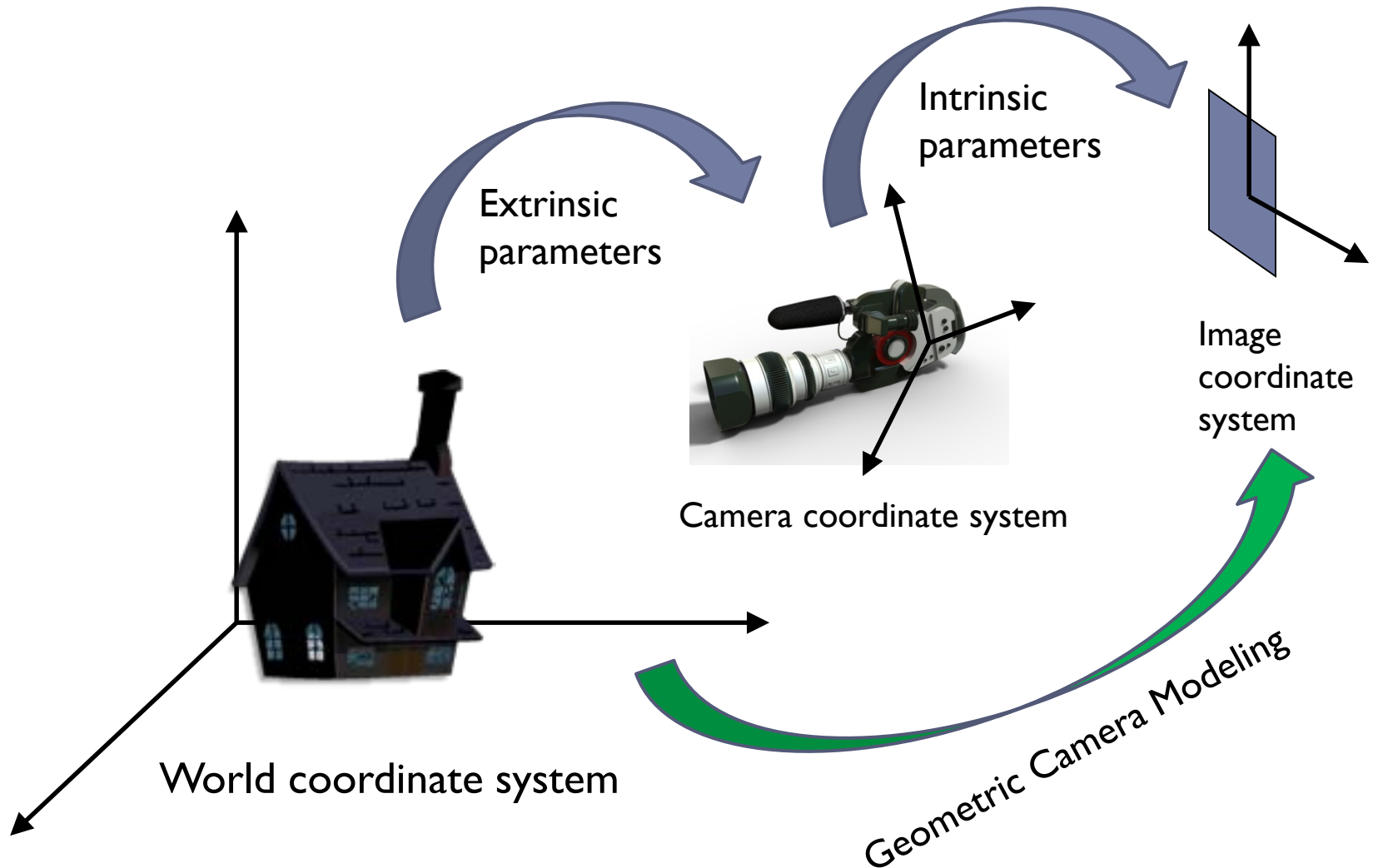
- ▶ Rigid transformation using homogeneous coordinates

$${}^B P = {}^B_A R {}^A P + {}^B O_A$$

$$\begin{pmatrix} {}^B P \\ 1 \end{pmatrix} = \begin{pmatrix} {}^B_A R & {}^B O_A \\ \mathbf{0}^T & 1 \end{pmatrix} \begin{pmatrix} {}^A P \\ 1 \end{pmatrix}$$

$$\begin{pmatrix} {}^B P \\ 1 \end{pmatrix} = {}^B_A T \begin{pmatrix} {}^A P \\ 1 \end{pmatrix}, \quad \text{where } {}^B_A T = \begin{pmatrix} {}^B_A R & {}^B O_A \\ \mathbf{0}^T & 1 \end{pmatrix}_{4 \times 4}$$

Geometric Camera Modeling: Intrinsic and Extrinsic Parameters



Camera Parameters and the Perspective Projection



- ▶ In reality, the perspective equation is only valid when all distances are measured in the **camera's reference frame**, and **image coordinates** have their own origin.
- ▶ In practice, the world and camera coordinate systems are related by a set of parameters.
 - ▶ Intrinsic parameters relates the **camera's coordinate system** to the **image pixel coordinate system**.
 - ▶ Extrinsic parameters relate the **camera's coordinate system** to a **fixed world coordinate system** and specify its position and orientation in space.
- ▶ A process to estimate intrinsic & extrinsic parameters is known as *geometric camera calibration*.

Intrinsic Parameters

- ▶ We want to associate with a camera with a normalized image plane parallel to its physical retina but located at a unit distance from the pinhole.

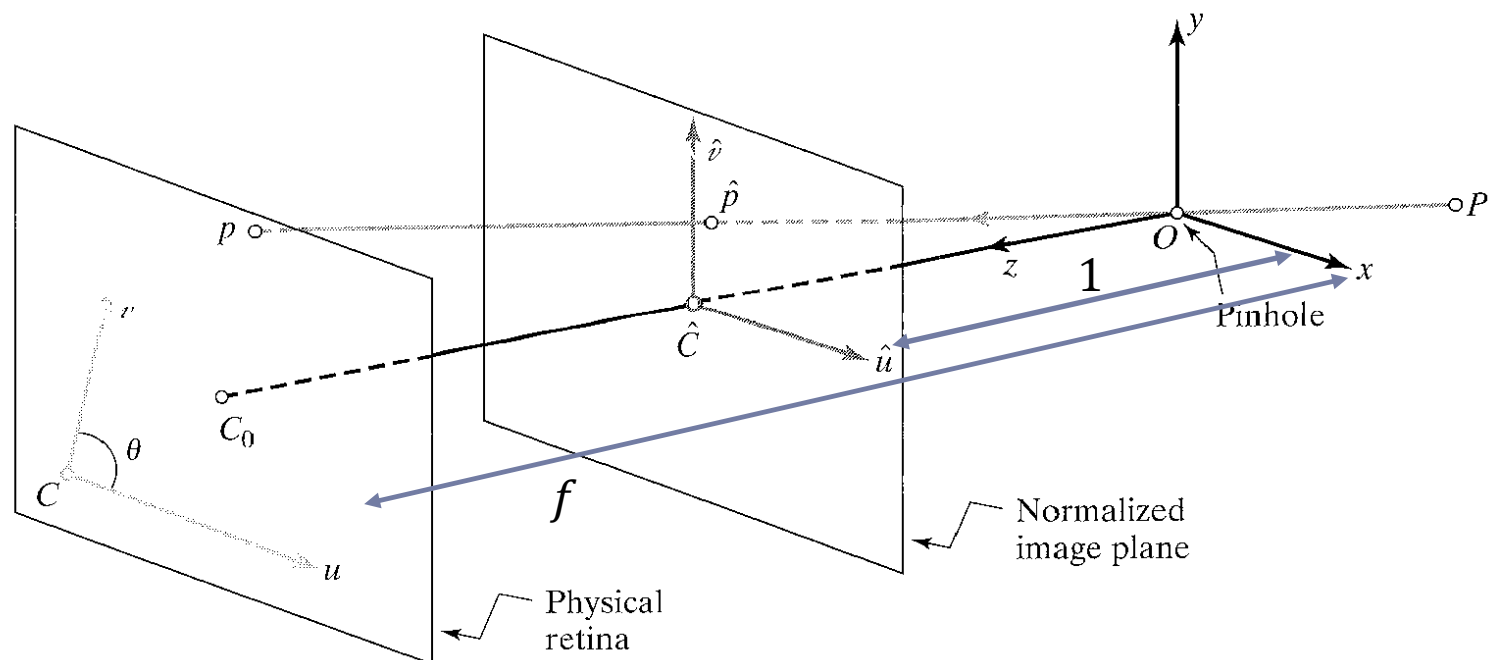


Figure 2.8 Physical and normalized image coordinate systems.

Normalized Image Plane

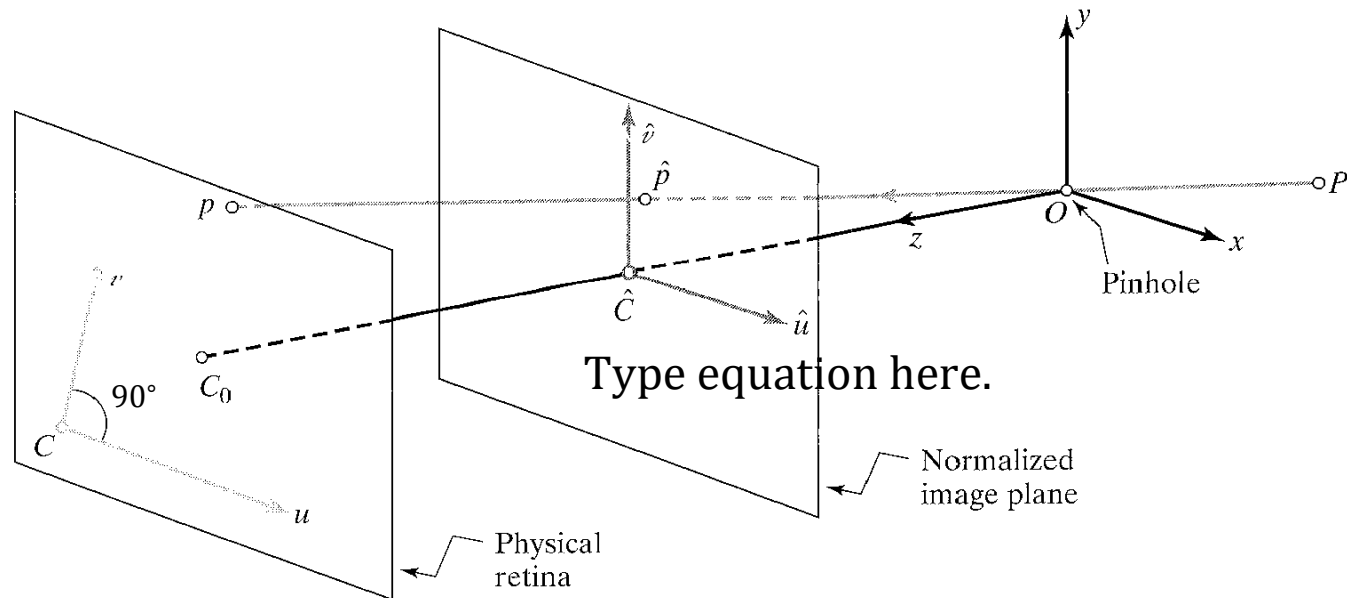


Figure 2.8 Physical and normalized image coordinate systems.

$$\begin{cases} \hat{u} = \frac{x}{z} \\ \hat{v} = \frac{y}{z} \end{cases} \Leftrightarrow \hat{\mathbf{p}} = \frac{1}{z}(\mathbf{I} \quad 0)\mathbf{P} \quad \text{where} \quad \begin{cases} \mathbf{P} = (x \quad y \quad z \quad 1)^T \\ \hat{\mathbf{p}} = (\hat{u} \quad \hat{v} \quad 1)^T \end{cases}$$

Perspective projection

Physical Retina of the Camera

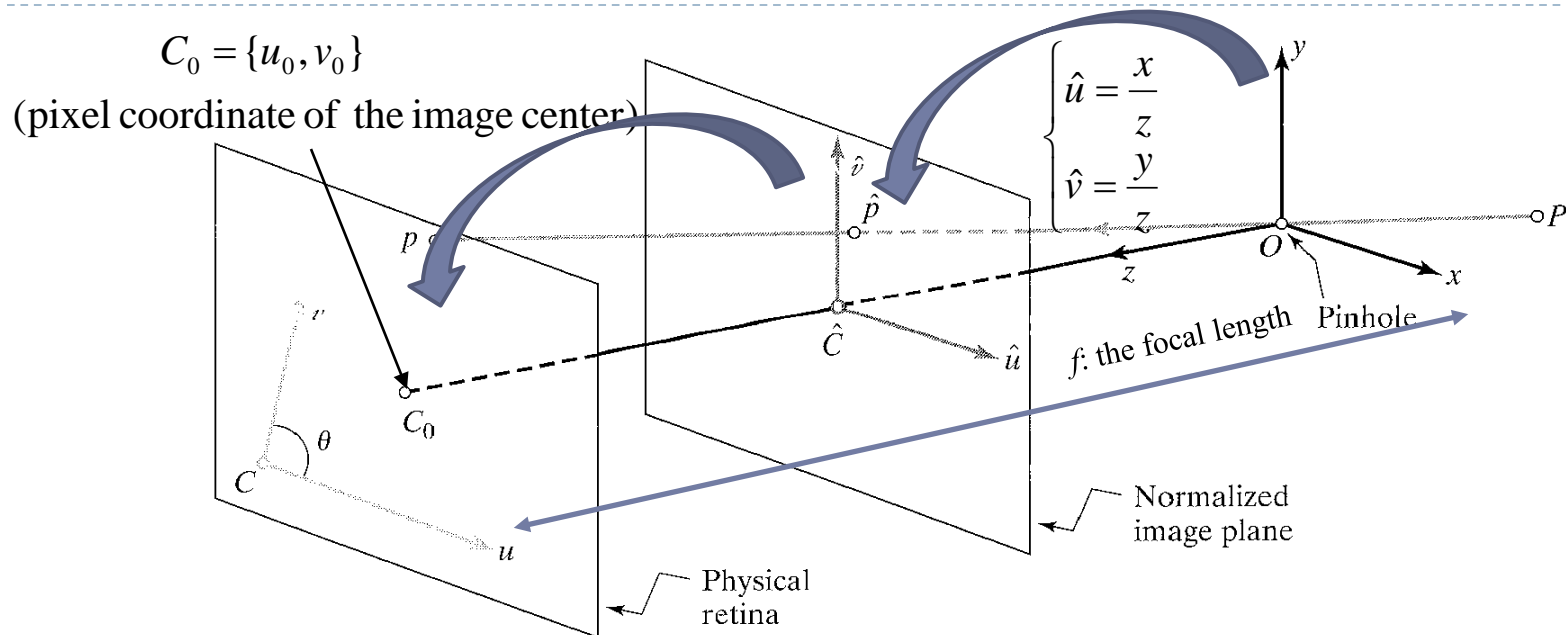


Figure 2.8 Physical and normalized image coordinate systems.

$$\begin{cases} u = kf \frac{x}{z} \\ v = lf \frac{y}{z} \end{cases} \quad \text{where a pixel has dimension } \frac{1}{k} \times \frac{1}{l}, f \text{ is the focal length.} \Rightarrow \text{Pixel coordinates} \quad \begin{cases} u = \alpha \frac{x}{z} + u_0 \\ v = \beta \frac{y}{z} + v_0 \end{cases}$$

(assuming $\theta = \frac{\pi}{2}$)

$\frac{1}{l}$ Pixel

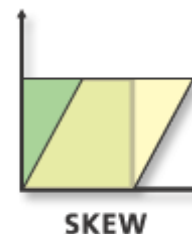
$\alpha = kf$ and $\beta = lf$

Affine Transformation Review

- ▶ An affine transformation can differentially scale the data, skew it, rotate it, and translate it.

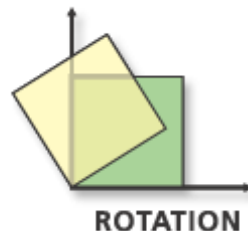
$$\begin{pmatrix} u \\ v \\ 1 \end{pmatrix} = \begin{pmatrix} a & b & c \\ d & e & f \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} \hat{u} \\ \hat{v} \\ 1 \end{pmatrix}$$

$$\begin{pmatrix} s_u & 0 & 0 \\ 0 & s_v & 0 \\ 0 & 0 & 1 \end{pmatrix}$$



$$\begin{pmatrix} 1 & h_x & 0 \\ h_y & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

$$\begin{pmatrix} \cos \theta & -\sin \theta & 0 \\ \sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{pmatrix}$$



$$\begin{pmatrix} 1 & 0 & u_0 \\ 0 & 1 & v_0 \\ 0 & 0 & 1 \end{pmatrix}$$

Normalized Image Plan and Physical Retina: Revisited

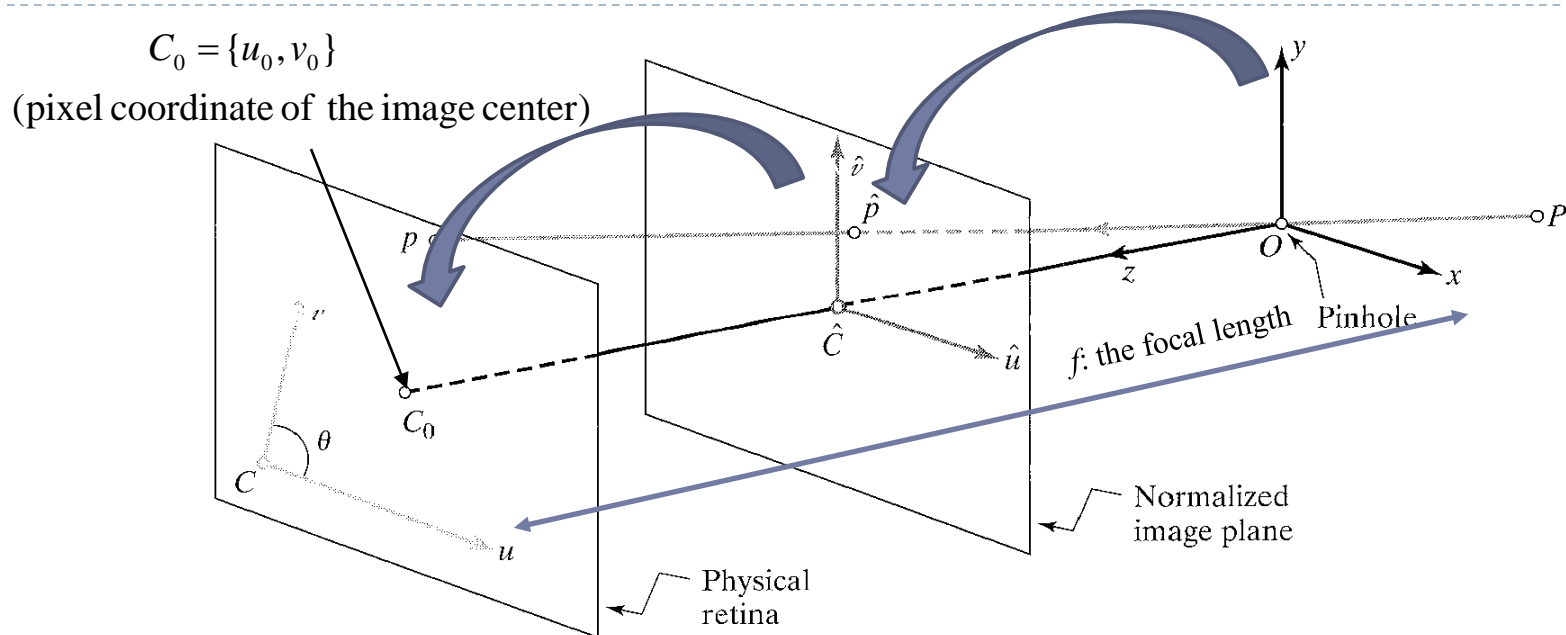


Figure 2.8 Physical and normalized image coordinate systems.

$$\begin{cases} \hat{u} = \frac{x}{z} \\ \hat{v} = \frac{y}{z} \end{cases} \xrightarrow{\text{Affine Transformation}} \begin{cases} u = \alpha \hat{u} - \alpha \cot \theta \hat{v} + u_0 \\ v = \frac{\beta}{\sin \theta} \hat{v} + v_0 \end{cases} \quad \left(\cot \theta = \frac{\cos \theta}{\sin \theta} \right)$$

$$\alpha = kf, \beta = lf$$

Planar Affine Transformation

