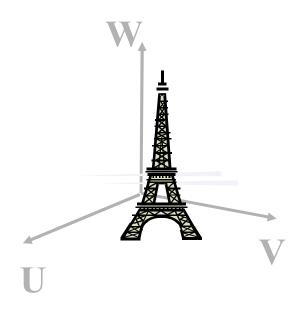
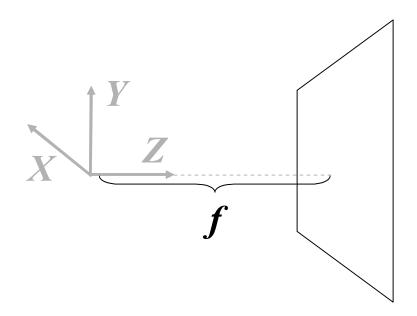
# Lecture 13: Camera Projection II

Reading: T&V Section 2.4

## **Recall: Imaging Geometry**

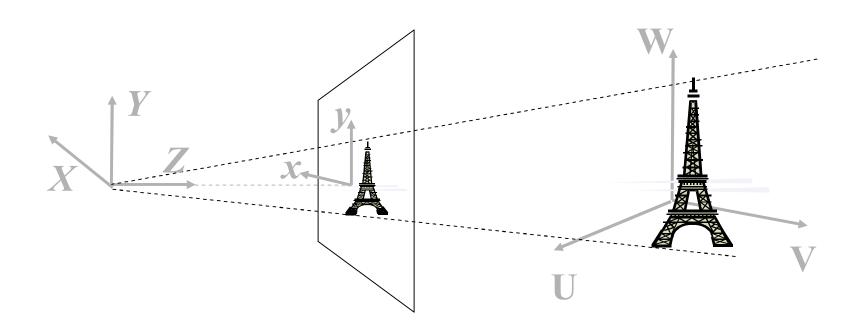
Object of Interest in World Coordinate System (U,V,W)



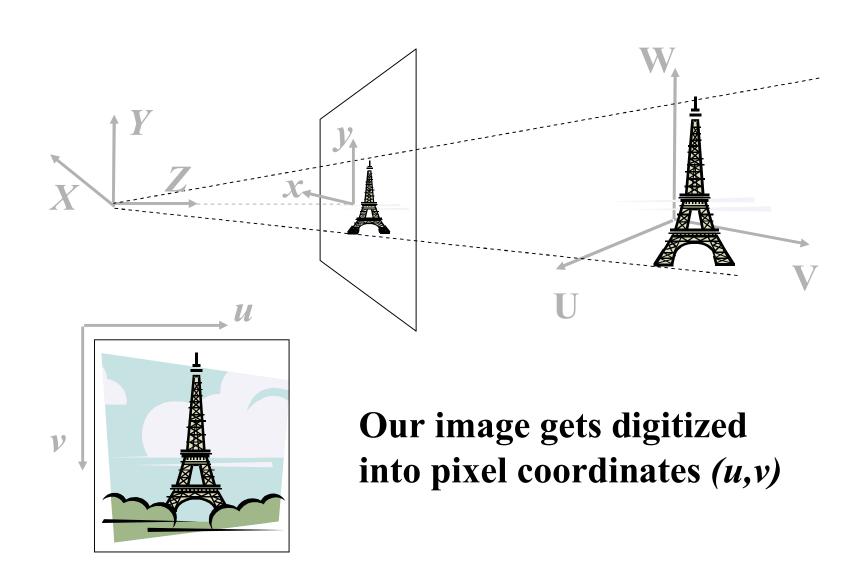


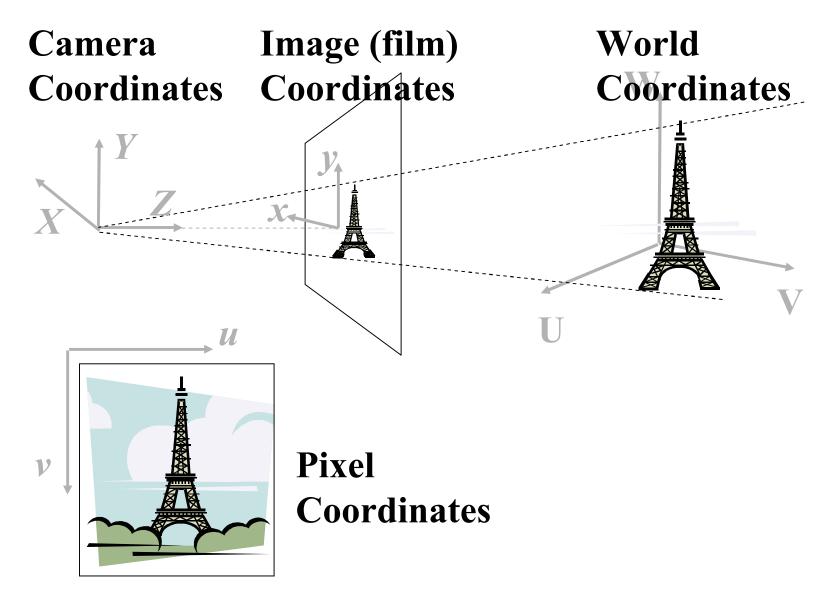
#### Camera Coordinate System (X,Y,Z).

- Z is optic axis
- Image plane located f units out along optic axis
- f is called focal length

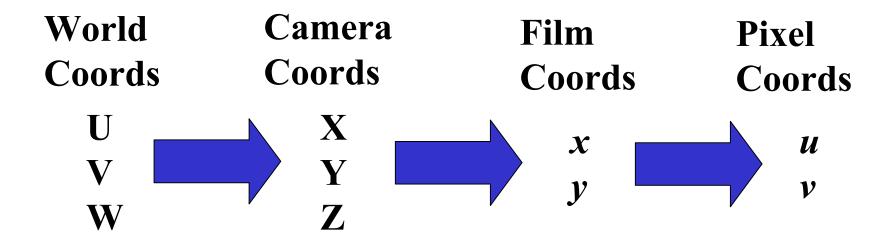


Forward Projection onto image plane. 3D (X,Y,Z) projected to 2D (x,y)





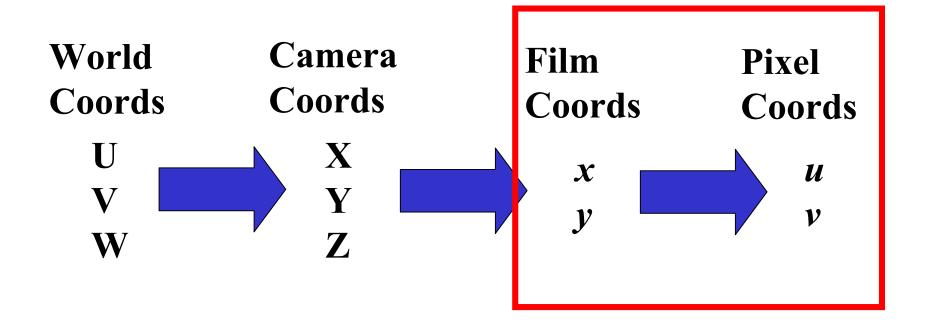
### **Forward Projection**



We want a mathematical model to describe how 3D World points get projected into 2D Pixel coordinates.

Our goal: describe this sequence of transformations by a big matrix equation!

#### **Intrinsic Camera Parameters**



**Affine Transformation** 

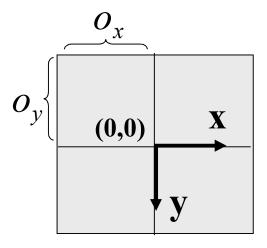
### Intrinsic parameters

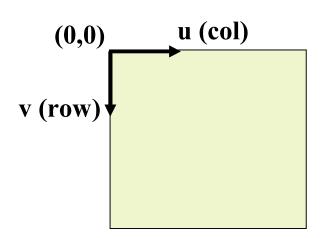
- Describes coordinate transformation between film coordinates (projected image) and pixel array
- Film cameras: scanning/digitization
- CCD cameras: grid of photosensors

### Intrinsic parameters (offsets)

film plane (projected image)

pixel array





$$u = f \frac{X}{Z} + o_x \qquad v = f \frac{Y}{Z} + o_y$$

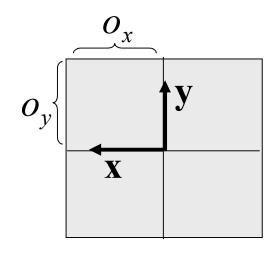
o<sub>x</sub> and o<sub>y</sub> called image center or principle point

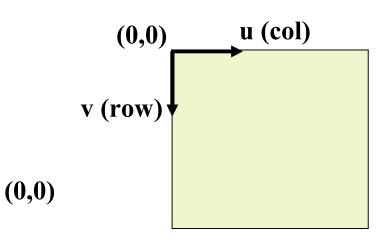
### Intrinsic parameters

sometimes one or more coordinate axes are flipped (e.g. T&V section 2.4)



pixel array

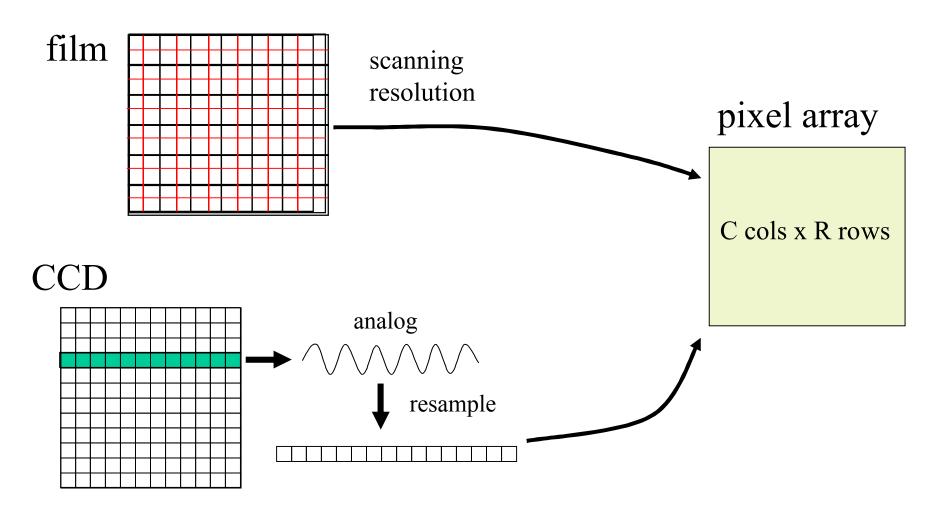




$$u = -f\frac{X}{Z} + o_x \qquad v = -f\frac{Y}{Z} + o_y$$

## Intrinsic parameters (scales)

sampling determines how many rows/cols in the image



# Effective Scales: $s_x$ and $s_y$

$$u = \frac{1}{S_x} f \frac{X}{Z} + o_x \qquad v = \frac{1}{S_y} f \frac{Y}{Z} + o_y$$

Note, since we have different scale factors in x and y, we don't necessarily have square pixels!

Aspect ratio is  $s_y / s_x$ 

### Perspective projection matrix

Adding the intrinsic parameters into the perspective projection matrix:

$$\begin{bmatrix} x' \\ y' \\ z' \end{bmatrix} = \begin{bmatrix} f/s_x & 0 & o_x & 0 \\ 0 & f/s_y & o_y & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

To verify:

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

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

$$u = \frac{1}{S_x} f \frac{X}{Z} + o_x$$

$$v = \frac{1}{S_y} f \frac{Y}{Z} + o_y$$

#### Note:

Sometimes, the image and the camera coordinate systems have opposite orientations: [the book does it this way]

$$f\frac{X}{Z} = -(u - o_x)s_x$$

$$f\frac{Y}{Z} = -(v - o_y)s_y$$

$$\begin{bmatrix} x' \\ y' \\ z' \end{bmatrix} = \begin{bmatrix} -f/s_x & 0 & +o_x & 0 \\ 0 & -f/s_y & +o_y & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

#### Note 2

In general, I like to think of the conversion as a separate 2D affine transformation from film coords (x,y) to pixel coordinates (u,v):

$$\begin{pmatrix} u' \\ v' \\ w' \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ 0 & 0 & 1 \end{pmatrix} \begin{bmatrix} f & 0 & 0 & 0 \\ 0 & f & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

$$\mathbf{M}_{aff}$$

$$u = M_{int} P_C = M_{aff} M_{proj} P_C$$

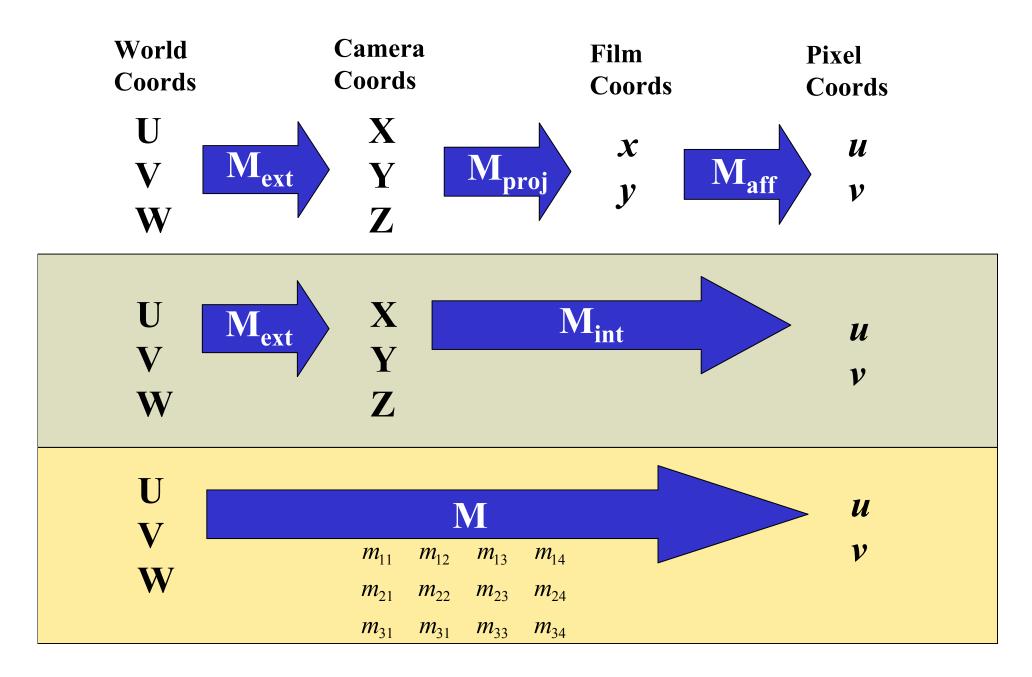
#### Huh?

Did he just say it was "a fine" transformation?

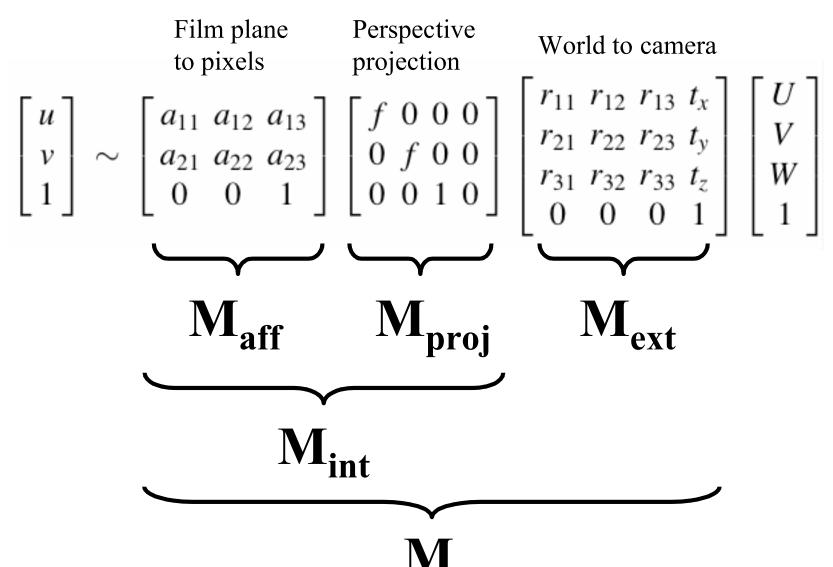
No, it was "affine" transformation, a type of 2D to 2D mapping defined by 6 parameters.

More on this in a moment...

# CSE486, Penn State Summary: Forward Projection



# CSE486, Penn State Summary: Projection Equation



# Lecture 13/14: Intro to Image Mappings

# Image Mappings Overview

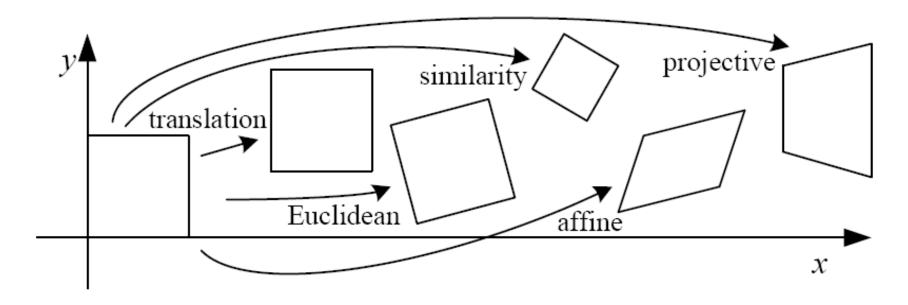
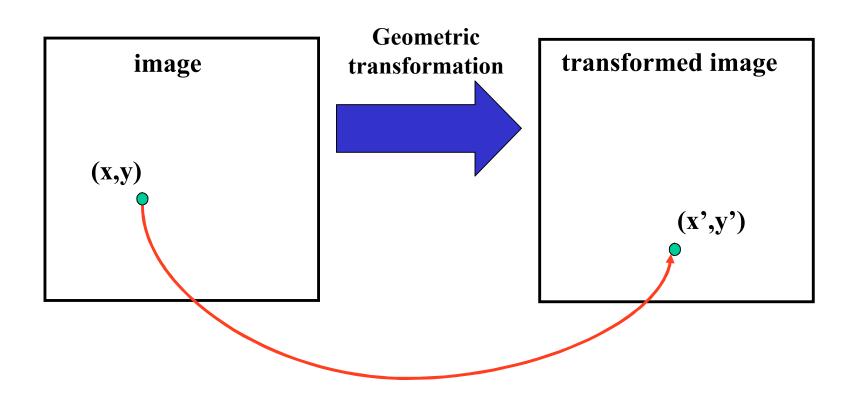


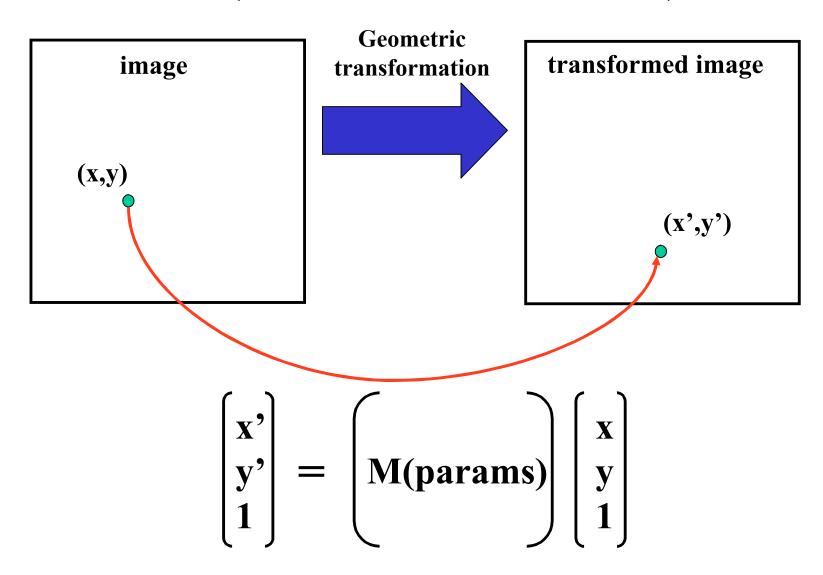
FIGURE 1. Basic set of 2D planar transformations

### Geometric Image Mappings

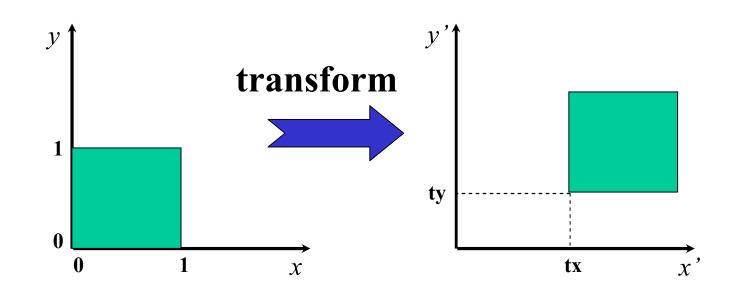


#### **Linear Transformations**

(Can be written as matrices)



#### **Translation**



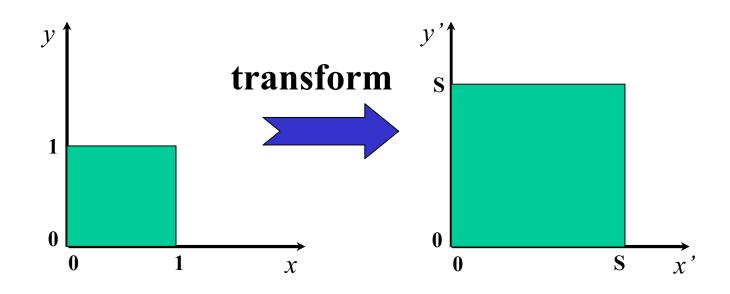
$$x' = x + t_x$$
$$y' = y + t_y$$

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

equations

matrix form

#### Scale

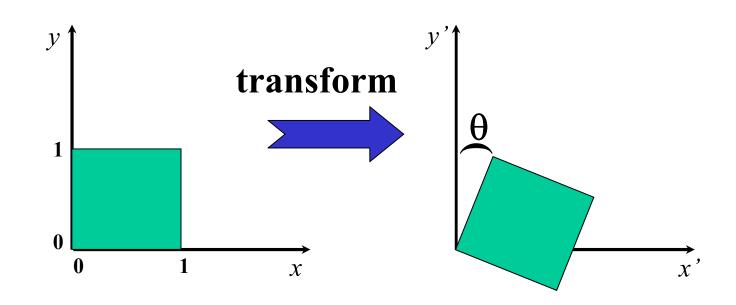


$$x' = s x_i$$
$$y' = s y_i$$

$$\begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix} = \begin{bmatrix} s & 0 & 0 \\ 0 & s & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

matrix form

#### **Rotation**



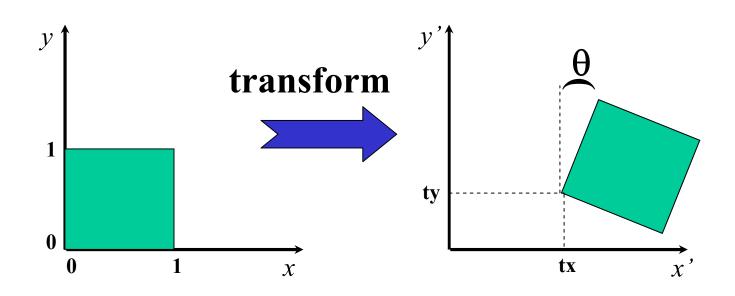
$$x' = x_i \cos \theta - y_i \sin \theta$$
  
$$y' = x_i \sin \theta + y_i \cos \theta$$

$$\begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix} = \begin{bmatrix} \cos \theta & -\sin \theta & 0 \\ \sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

equations

matrix form

### **Euclidean (Rigid)**



$$x' = x_i \cos \theta - y_i \sin \theta + t_x$$
  
$$y' = x_i \sin \theta + y_i \cos \theta + t_y$$

$$x' = x_i \cos \theta - y_i \sin \theta + t_x y' = x_i \sin \theta + y_i \cos \theta + t_y$$
$$\begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix} = \begin{bmatrix} \cos \theta & -\sin \theta & t_x \\ \sin \theta & \cos \theta & t_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

equations

matrix form

#### **Partitioned Matrices**

A partitioned matrix, or a block matrix, is a matrix M that has been constructed from other smaller matrices. These smaller matrices are called blocks or sub-matrices of M.

For instance, if we partition the below  $5 \times 5$  matrix as follows

$$L = \begin{pmatrix} 1 & 0 & 1 & 2 & 3 \\ 0 & 1 & 1 & 2 & 3 \\ \hline 2 & 3 & 9 & 9 & 9 \\ 2 & 3 & 9 & 9 & 9 \\ 2 & 3 & 9 & 9 & 9 \end{pmatrix},$$

then we can define the matrices

$$A = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}, B = \begin{pmatrix} 1 & 2 & 3 \\ 1 & 2 & 3 \end{pmatrix}, C = \begin{pmatrix} 2 & 3 \\ 2 & 3 \\ 2 & 3 \end{pmatrix}, D = \begin{pmatrix} 9 & 9 & 9 \\ 9 & 9 & 9 \\ 9 & 9 & 9 \end{pmatrix}$$

and write  $\,L\,$  as

$$L = \begin{pmatrix} A & B \\ C & D \end{pmatrix}, \text{ or } L = \begin{pmatrix} A & B \\ \hline C & D \end{pmatrix}.$$

#### **Partitioned Matrices**

$$\begin{bmatrix} x' \\ \underline{y'} \\ 1 \end{bmatrix} = \begin{bmatrix} \cos \theta & -\sin \theta & t_x \\ \sin \theta & \cos \theta & t_y \\ \hline 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ \underline{y} \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} p' \\ \mathbf{1}_{x1} \\ 1 \end{bmatrix} = \begin{bmatrix} \mathbf{R} & t \\ \mathbf{1}_{x2} & \mathbf{1}_{x1} \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \mathbf{p} \\ \mathbf{1}_{x1} \\ 1 \end{bmatrix} \quad \mathbf{matrix form}$$

$$p' = Rp + t$$

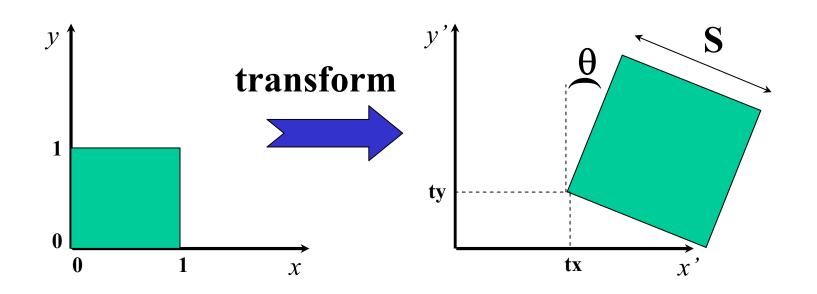
equation form

# CSE486, Penn State Another Example (from last time)

$$\begin{pmatrix}
\mathbf{P_{C}} \\
\mathbf{1x1} \\
1
\end{pmatrix} = \begin{pmatrix}
\mathbf{3x3} & \mathbf{3x1} \\
\mathbf{R} & \mathbf{T} \\
\mathbf{1x3} & \mathbf{1x1} \\
0 & 1
\end{pmatrix} \begin{pmatrix}
\mathbf{3x1} \\
\mathbf{P_{W}} \\
\mathbf{1x1} \\
1
\end{pmatrix}$$

$$P_C = R P_W + T$$

### Similarity (scaled Euclidean)

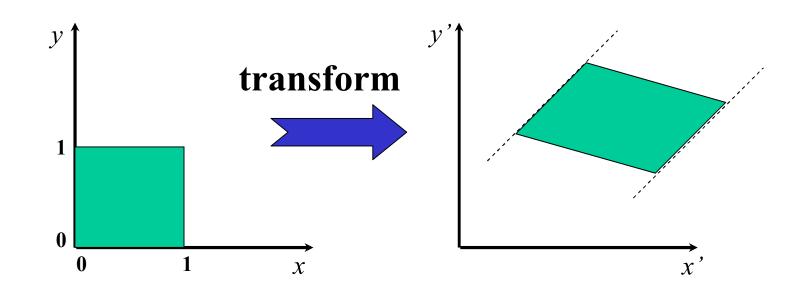


$$p' = sRp + t$$

$$\begin{bmatrix} p' \\ 1 \end{bmatrix} = \begin{bmatrix} sR & t \\ 0 & 1 \end{bmatrix} \begin{bmatrix} p \\ 1 \end{bmatrix}$$

matrix form

#### **Affine**



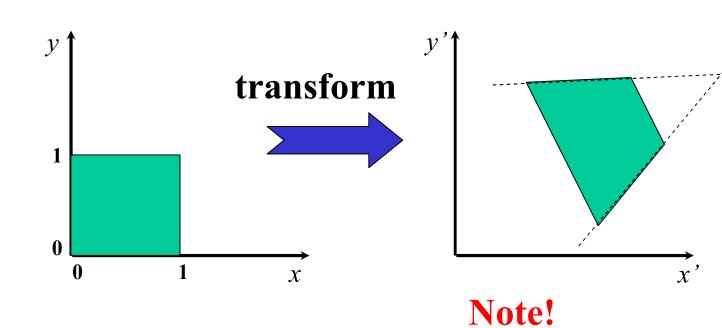
$$p' = Ap + b$$

$$\begin{bmatrix} p' \\ 1 \end{bmatrix} = \begin{bmatrix} A & b \\ 0 & 1 \end{bmatrix} \begin{bmatrix} p \\ 1 \end{bmatrix}$$

equations

matrix form

# **Projective**

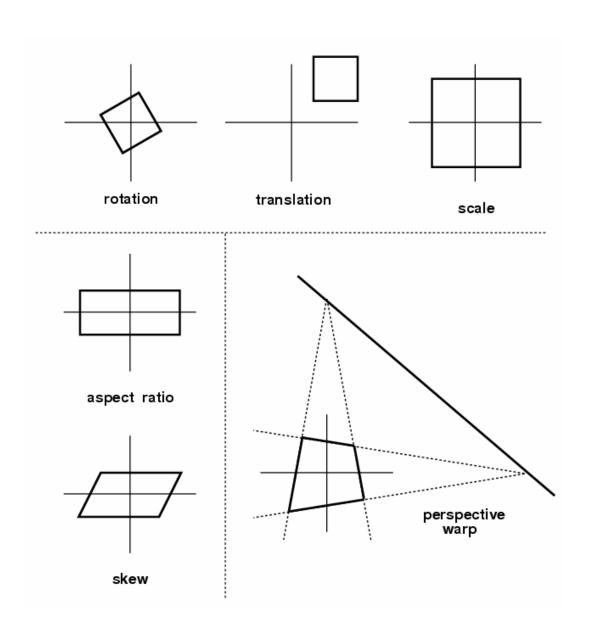


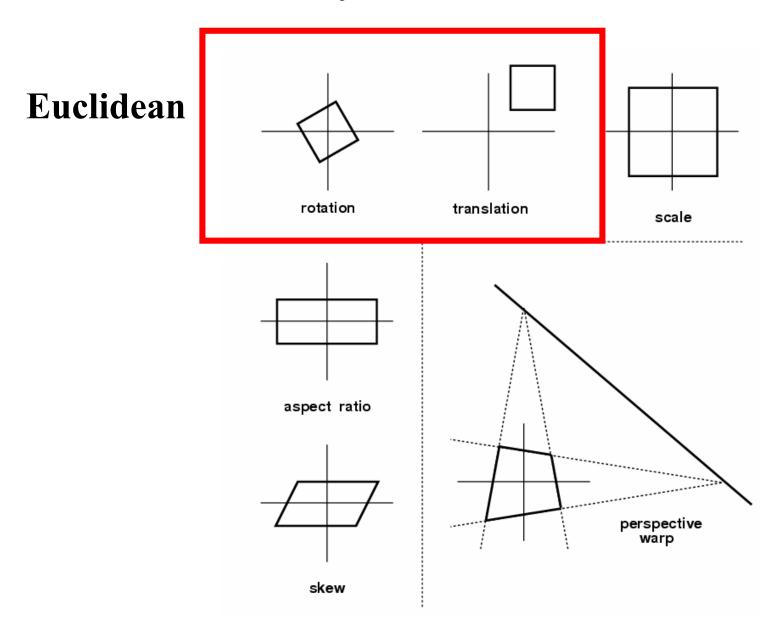
$$p' = \frac{Ap + b}{c^T p + 1}$$

equations

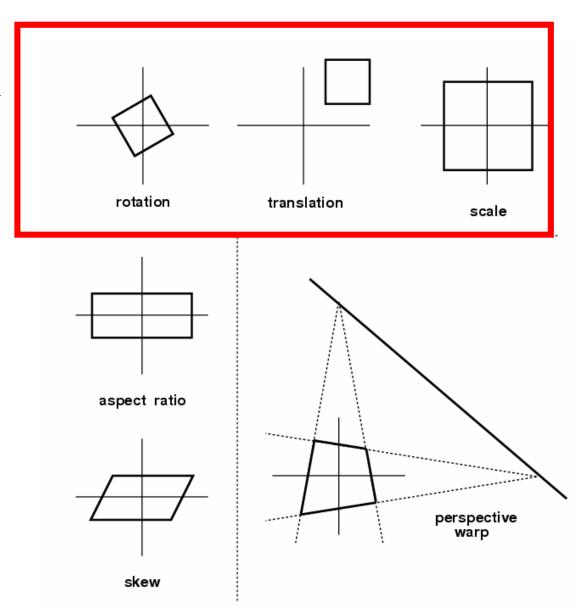
$$\begin{bmatrix} p' \\ 1 \end{bmatrix} \sim \begin{bmatrix} A & b \\ c^T & 1 \end{bmatrix} \begin{bmatrix} p \\ 1 \end{bmatrix}$$

matrix form

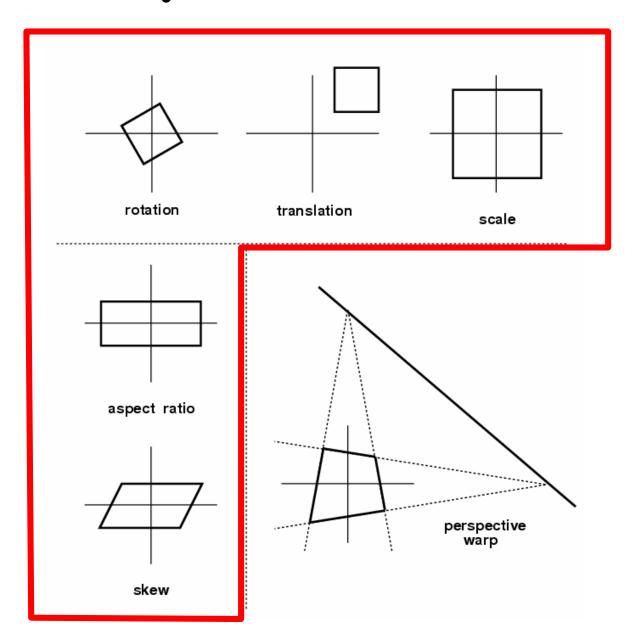




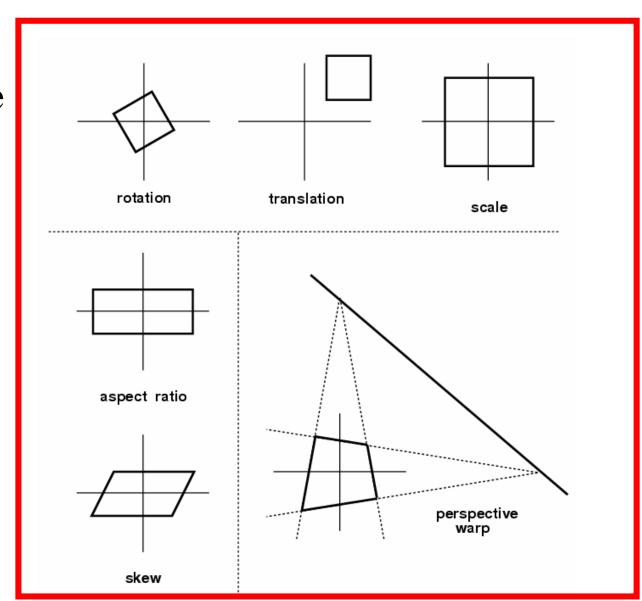
#### **Similarity**



#### **Affine**



#### **Projective**



# Robert Collins CSE486, Penn Stat Summary of 2D Transformations

Name	Matrix	# D.O.F.	Preserves:	Icon
translation	$\left[egin{array}{c c}I & t\end{array} ight]_{2 imes 3}$	2	orientation $+\cdots$	
rigid (Euclidean)	$\left[\begin{array}{c c} R & t\end{array}\right]_{2 imes 3}$	3	lengths $+\cdots$	$\Diamond$
similarity	$\left[\begin{array}{c c} sR \mid t\end{array}\right]_{2 imes 3}$	4	$angles + \cdots$	$\Diamond$
affine	$\left[\begin{array}{c} m{A} \end{array} ight]_{2 imes 3}$	6	parallelism $+\cdots$	
projective	$\left[\begin{array}{c} H \end{array} ight]_{3 imes 3}$	8	straight lines	