

# Geometric Transformation

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COLLEGE OF COMPUTING

HANYANG ERICA CAMPUS

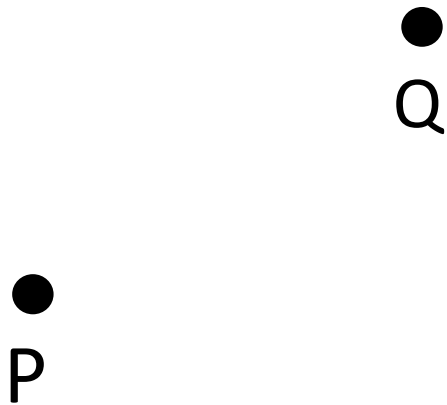
Q YOUN HONG (홍규연)

# Recap on Math

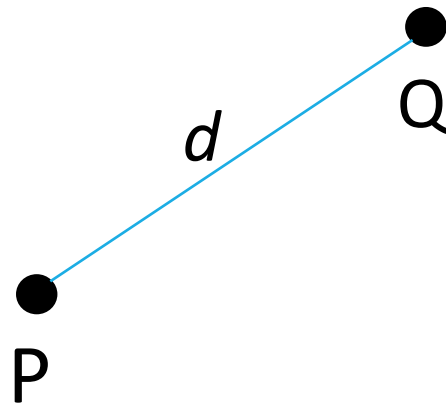
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# Points, Scalars and Vectors

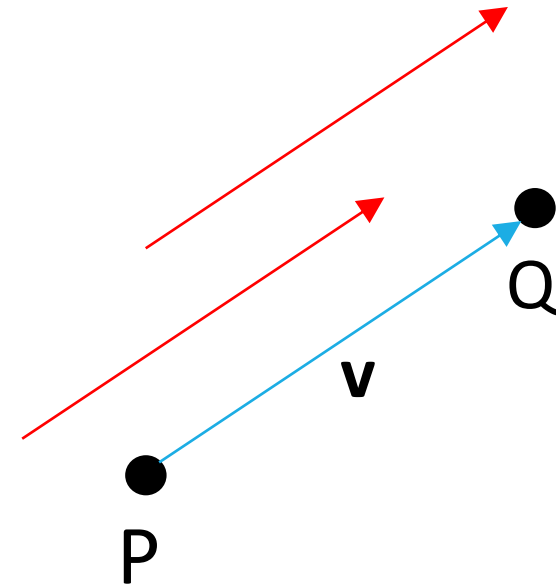
- Point: a location in space
- Scalar: real number, e.g. distance
- Vector: direction with magnitude



Points

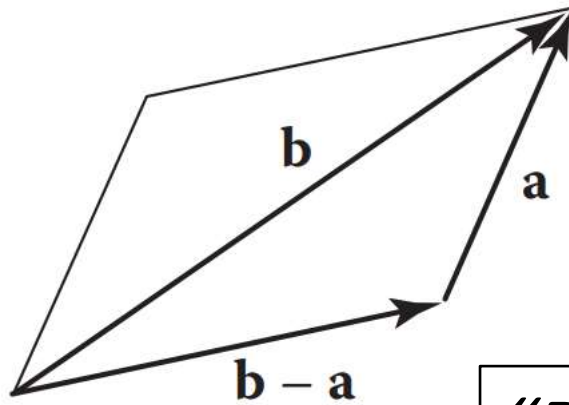
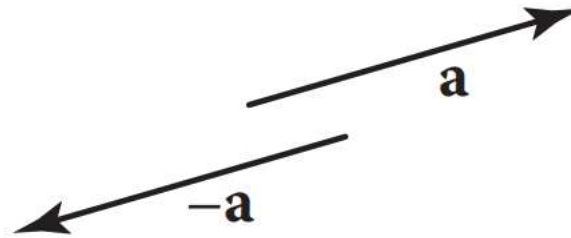
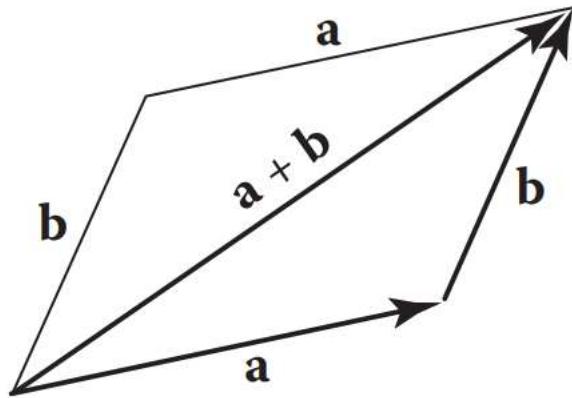


Scalar

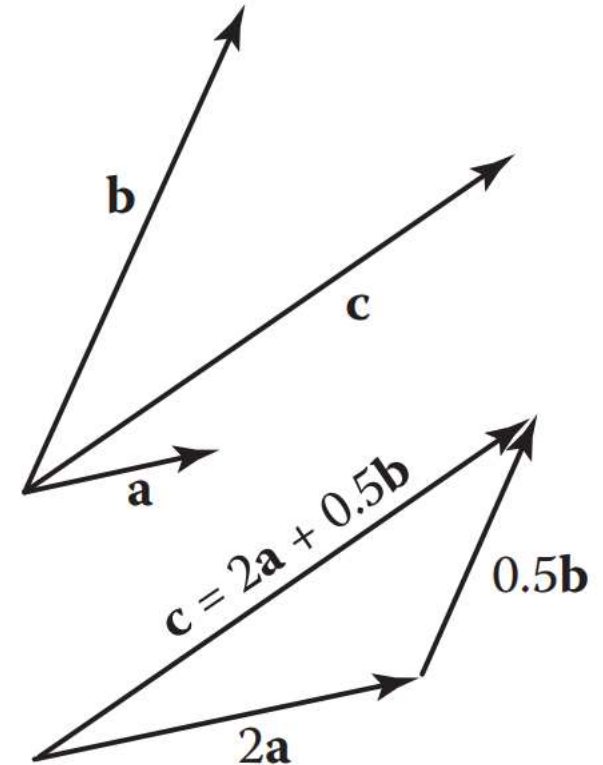


Vector

# Vector Operations

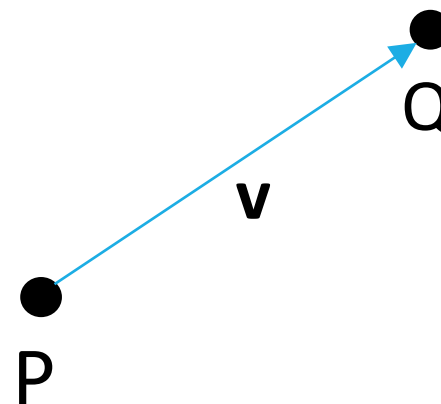


*"Parallelogram rule"*



# Point-Vector Operations

- Point + Vector = Point ( $Q = P + \mathbf{v}$ )  
(point-vector addition)
- Point - Point = Vector ( $\mathbf{v} = Q - P$ )  
(point-point subtraction)



ex)  $P + 3\mathbf{v} = ?$

ex)  $2P - Q + 3\mathbf{v} = ?$

ex)  $P + 3Q - \mathbf{v} = ?$

# Line



- Parametric form of a line:

$$P(\alpha) = P_0 + \alpha \mathbf{d},$$

$P_0$ : an arbitrary point (origin)

$\alpha$ : an arbitrary scalar

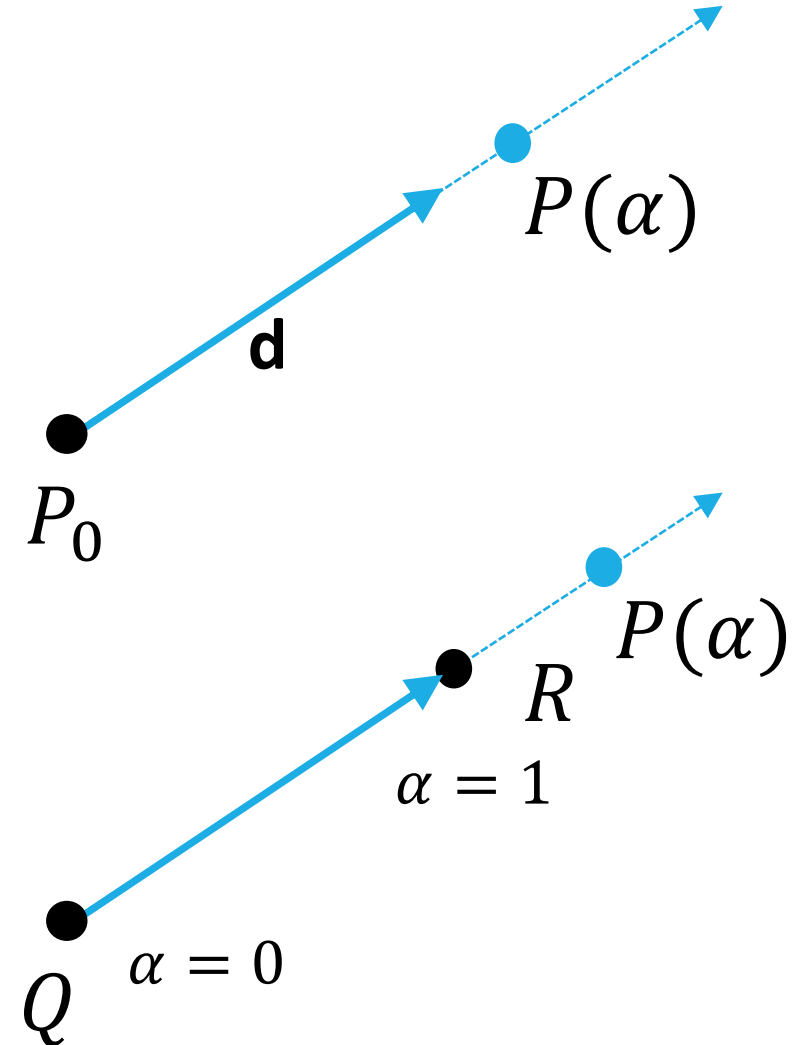
$\mathbf{d}$ : an arbitrary vector (direction)

- Affine sum:

$$P = Q + \alpha(R - Q) = (1 - \alpha)Q + \alpha R$$

$$P = \alpha_1 Q + \alpha_2 R, \alpha_1 + \alpha_2 = 1 \text{ (affine sum)}$$

Convex sum if  $\alpha_1 \geq 0, \alpha_2 \geq 0$ .



# Cartesian Coordinates of Vectors



- A 2D vector can be written as a combination of any two nonzero vectors that are not parallel (linearly independent):

$$\mathbf{c} = a_c \mathbf{a} + b_c \mathbf{b}$$

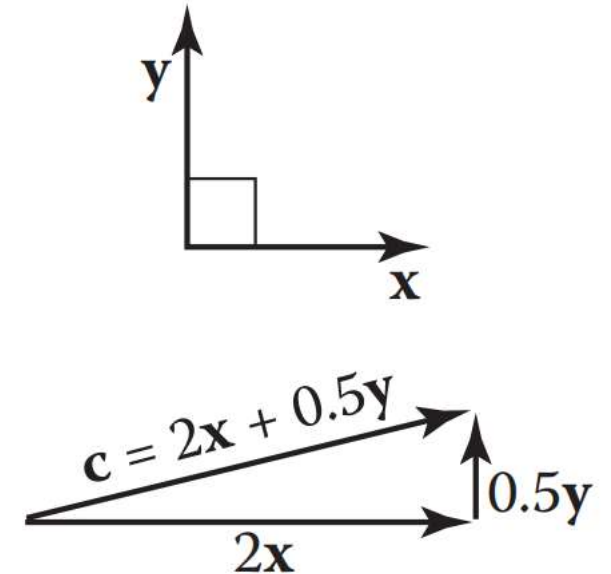
- If we use two orthonormal vectors  $\mathbf{x}, \mathbf{y}$ ,

$$\mathbf{a} = x_a \mathbf{x} + y_a \mathbf{y}$$

The coordinates of  $\mathbf{a} = (x_a, y_a)$ , or written as

$$\mathbf{a} = \begin{bmatrix} x_a \\ y_a \end{bmatrix}$$

- Also applies to 3D, 4D,...

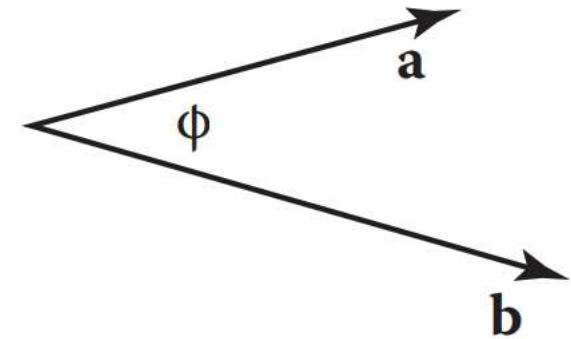


# Dot Product (내적)



- Dot product (= scalar product = inner product):

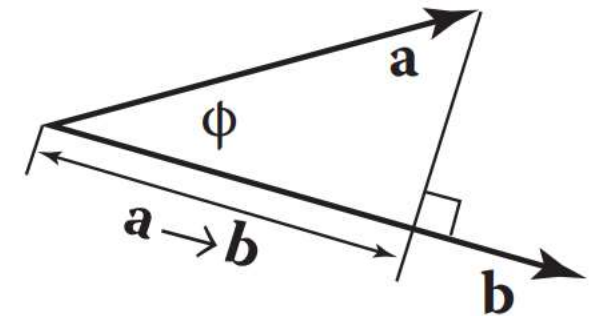
$$\mathbf{a} \cdot \mathbf{b} = \|\mathbf{a}\| \|\mathbf{b}\| \cos \Phi$$



$$\mathbf{a} \cdot \mathbf{b} = \|\mathbf{a}\| \|\mathbf{b}\| \cos \phi$$

- The projection of vector **a** onto vector **b**:

$$\mathbf{a} \rightarrow \mathbf{b} = \|\mathbf{a}\| \cos \Phi = \frac{\mathbf{a} \cdot \mathbf{b}}{\|\mathbf{b}\|}$$





# Dot Product (cont'd)

- Some dot product rules:

$$\mathbf{a} \cdot \mathbf{b} = \mathbf{b} \cdot \mathbf{a}$$

$$\mathbf{a} \cdot (\mathbf{b} + \mathbf{c}) = \mathbf{a} \cdot \mathbf{b} + \mathbf{a} \cdot \mathbf{c}$$

$$(k\mathbf{a}) \cdot \mathbf{b} = \mathbf{a} \cdot (k\mathbf{b}) = k\mathbf{a} \cdot \mathbf{b}$$

- $\mathbf{x} \cdot \mathbf{y} = 0$ ?
- If  $\mathbf{a} = (x_a, y_a)$ ,  $\mathbf{b} = (x_b, y_b)$ , then

$$\mathbf{a} \cdot \mathbf{b} = x_a x_b + y_a y_b$$

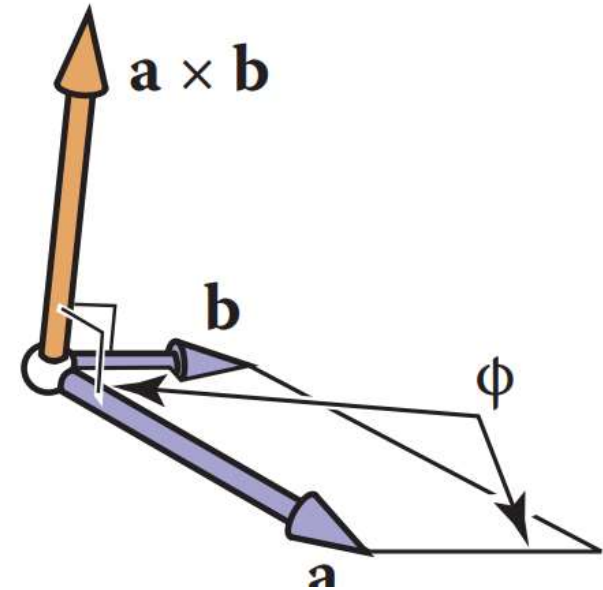
# Cross Product (외적)



- Cross Product (= vector product = exterior product):

$$\|\mathbf{a} \times \mathbf{b}\| = \|\mathbf{a}\| \|\mathbf{b}\| \sin \Phi$$

- $\mathbf{a} \times \mathbf{b}$  is perpendicular to both  $\mathbf{a}$  and  $\mathbf{b}$
- $\|\mathbf{a} \times \mathbf{b}\|$  = area of parallelogram made by  $\mathbf{a}$  and  $\mathbf{b}$



# Cross Product (cont'd)

- Right-hand rule applies
- If  $x = (1,0,0)$ ,  $y = (0,1,0)$ ,  $z = (0,0,1)$ ,

$$x \times y = +z$$

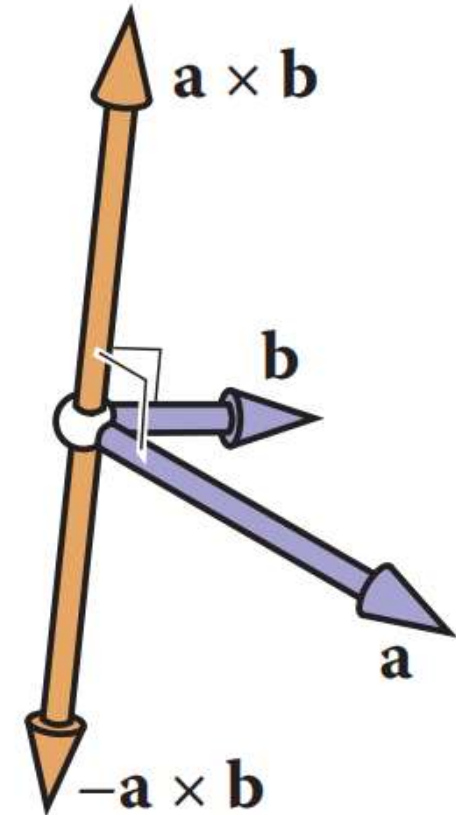
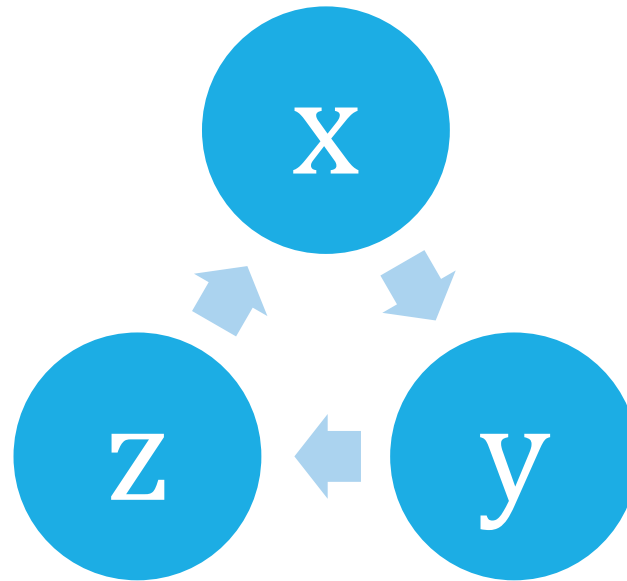
$$y \times x = -z$$

$$y \times z = +x$$

$$z \times y = -x$$

$$z \times x = +y$$

$$x \times z = -y$$



# Cross Product (cont'd)



- Some cross product rules:

$$\mathbf{a} \times \mathbf{b} = -\mathbf{b} \times \mathbf{a}$$

$$\mathbf{a} \times (\mathbf{b} + \mathbf{c}) = \mathbf{a} \times \mathbf{b} + \mathbf{a} \times \mathbf{c}$$

$$(k\mathbf{a}) \times \mathbf{b} = \mathbf{a} \times (k\mathbf{b}) = k(\mathbf{a} \times \mathbf{b})$$

- If  $\mathbf{a} = (x_a, y_a, z_a)$ ,  $\mathbf{b} = (x_b, y_b, z_b)$ , then

$$\mathbf{a} \times \mathbf{b} = (y_a z_b - z_a y_b, z_a x_b - x_a z_b, x_a y_b - y_a x_b)$$

# Coordinate Frames



- Orthonormal bases

- In 2D, use  $\mathbf{u}$ ,  $\mathbf{v}$  as bases such that  $\|\mathbf{u}\| = \|\mathbf{v}\| = 1$  and  $\mathbf{u} \cdot \mathbf{v} = 0$
- In 3D, use  $\mathbf{u}$ ,  $\mathbf{v}$ ,  $\mathbf{w}$  as bases such that

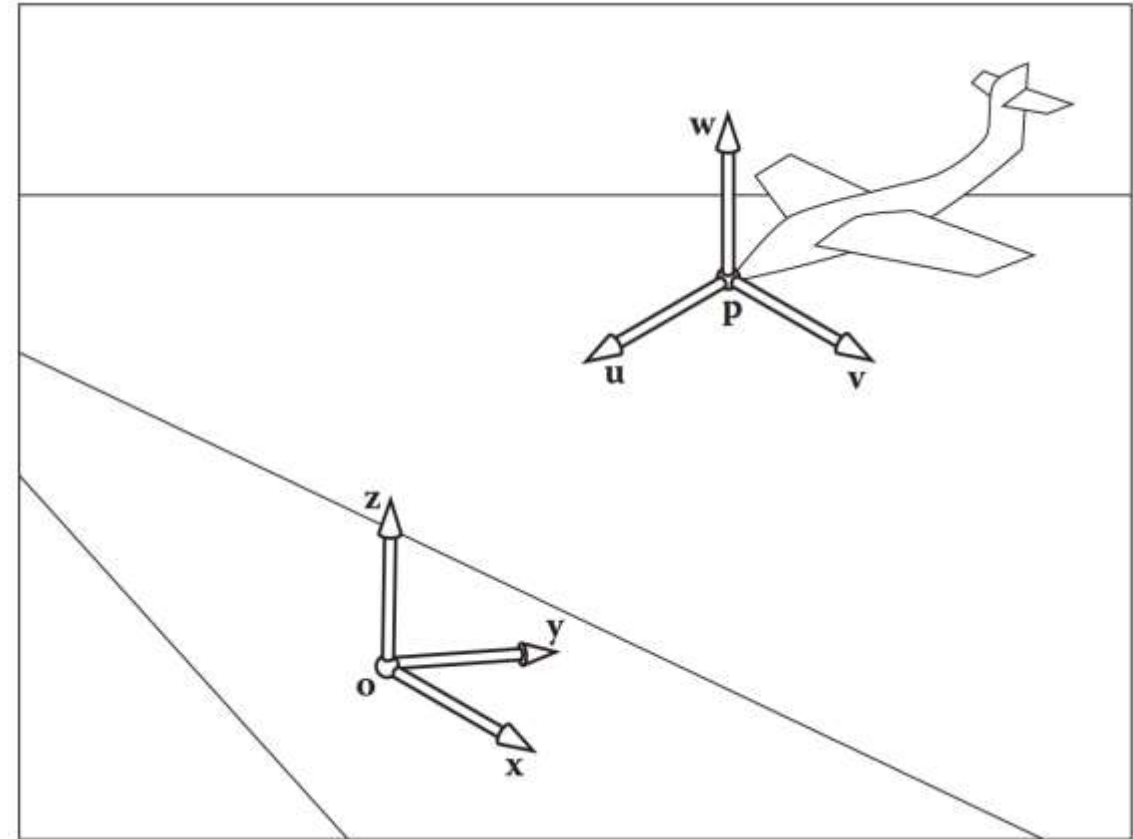
$$\begin{aligned}\|\mathbf{u}\| &= \|\mathbf{v}\| = \|\mathbf{w}\| = 1, \\ \mathbf{u} \cdot \mathbf{v} &= \mathbf{v} \cdot \mathbf{w} = \mathbf{w} \cdot \mathbf{u} = 0.\end{aligned}$$

- The Cartesian canonical orthonormal basis is just one of infinitely many orthonormal bases (bases are not explicitly stored)
- The Cartesian canonical coordinate system: x,y,z-axes + o

# Coordinate Systems



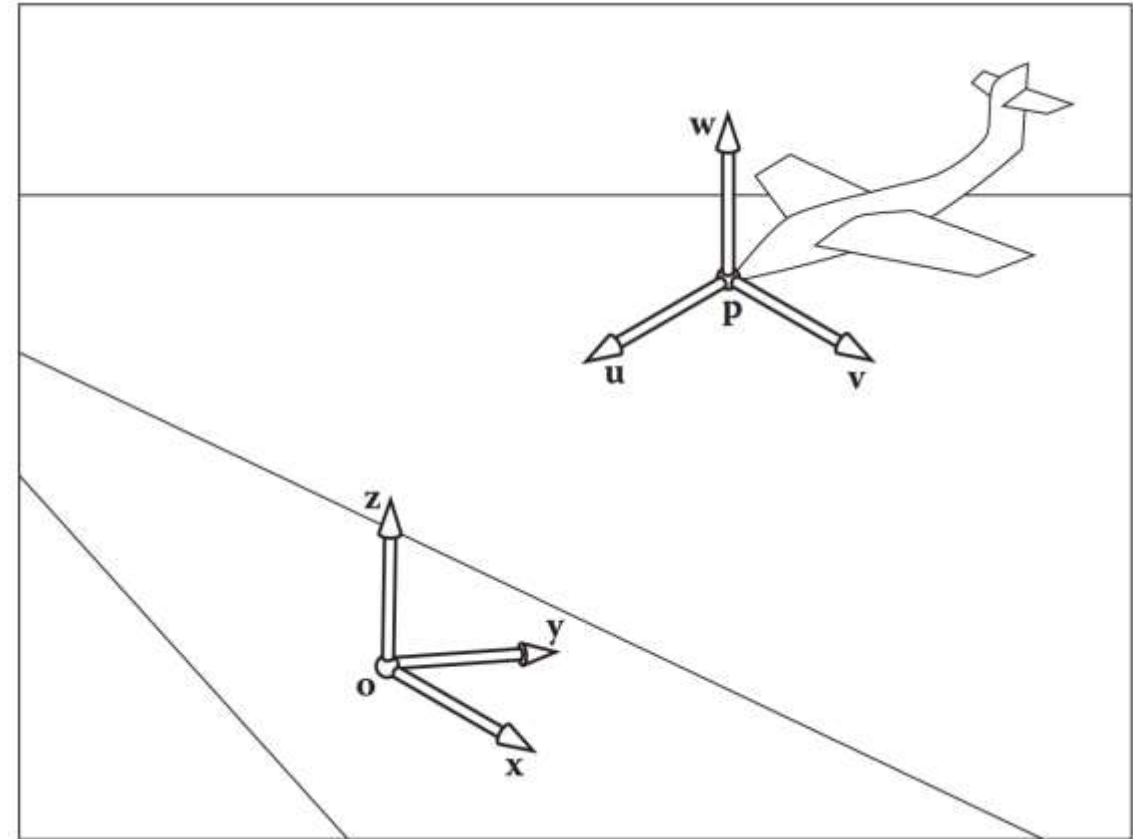
- The global model is typically stored in the canonical coordinate system (global/world coordinate system)
- We can define the model in another coordinate system (a frame of reference/coordinate frame)



# Coordinate Systems



- $\mathbf{u} = x_u \mathbf{x} + y_u \mathbf{y} + z_u \mathbf{z}$   
 $\mathbf{p} = \mathbf{o} + x_p \mathbf{x} + y_p \mathbf{y} + z_p \mathbf{z}$
- Express a vector  $\mathbf{a}$  in the airplane's coordinate frame?  
$$\mathbf{a} = u_a \mathbf{u} + v_a \mathbf{v} + w_a \mathbf{w}$$
  
⇒ Get  $(u_a, v_a, w_a)$  by  
⇒  $u_a = \mathbf{a} \cdot \mathbf{u}, v_a = \mathbf{a} \cdot \mathbf{v},$   
 $w_a = \mathbf{a} \cdot \mathbf{w}$



# 2D Geometric Transformation

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# 2D Linear Transformation

- Linear transformation

⇒ Use 2 x 2 matrix to transform a 2D vector

$$\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} a_{11}x + a_{12}y \\ a_{21}x + a_{22}y \end{bmatrix}$$

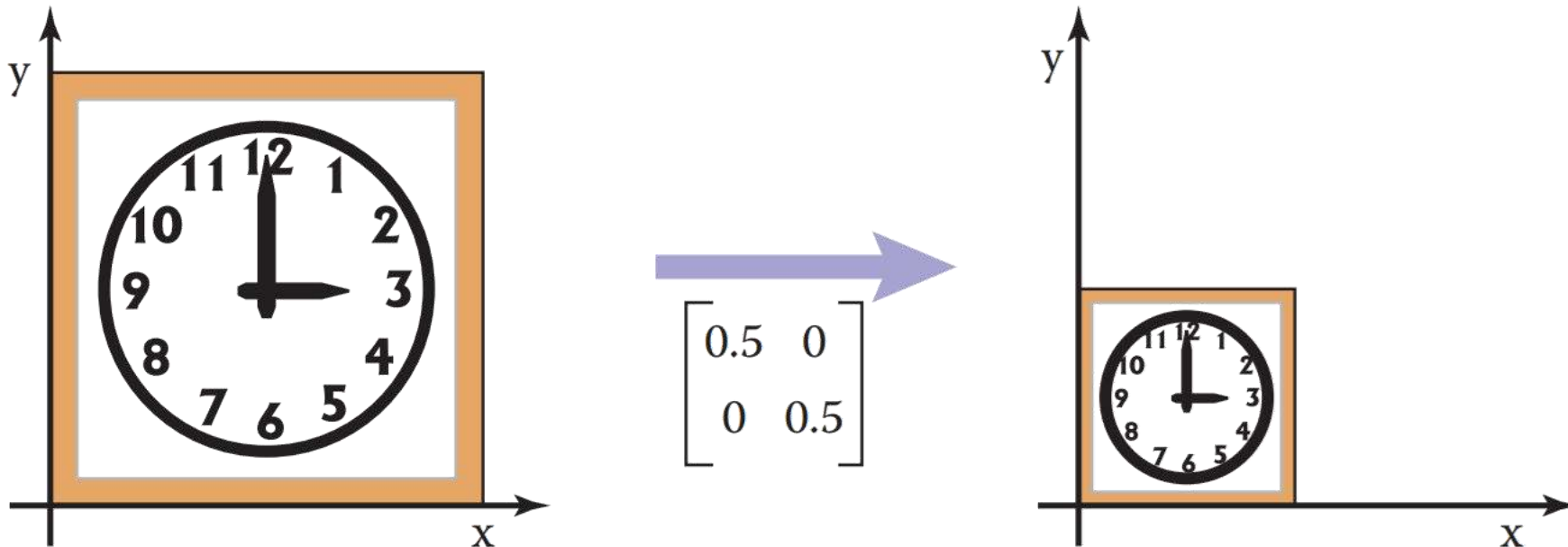
# 2D Scaling



- Scaling: changes length (and direction)

$$\text{scale}(s_x, s_y) = \begin{bmatrix} s_x & 0 \\ 0 & s_y \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} s_x x \\ s_y y \end{bmatrix}$$

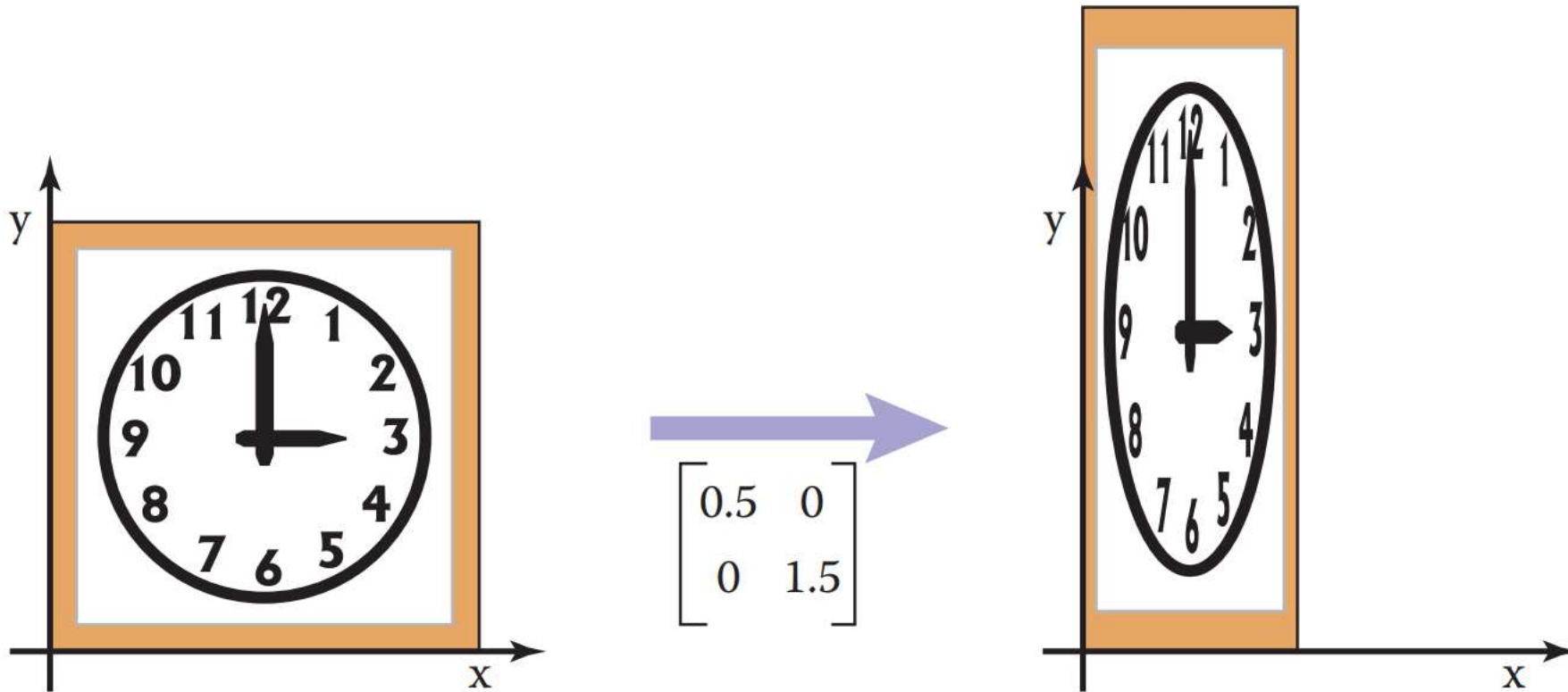
- If  $s_x = s_y$ , uniform scaling



# 2D Scaling



- If  $s_x \neq s_y$ , nonuniform scaling



# 2D Rotation

- A vector  $\mathbf{a} = (x_a, y_a)$  can be written as a polar form

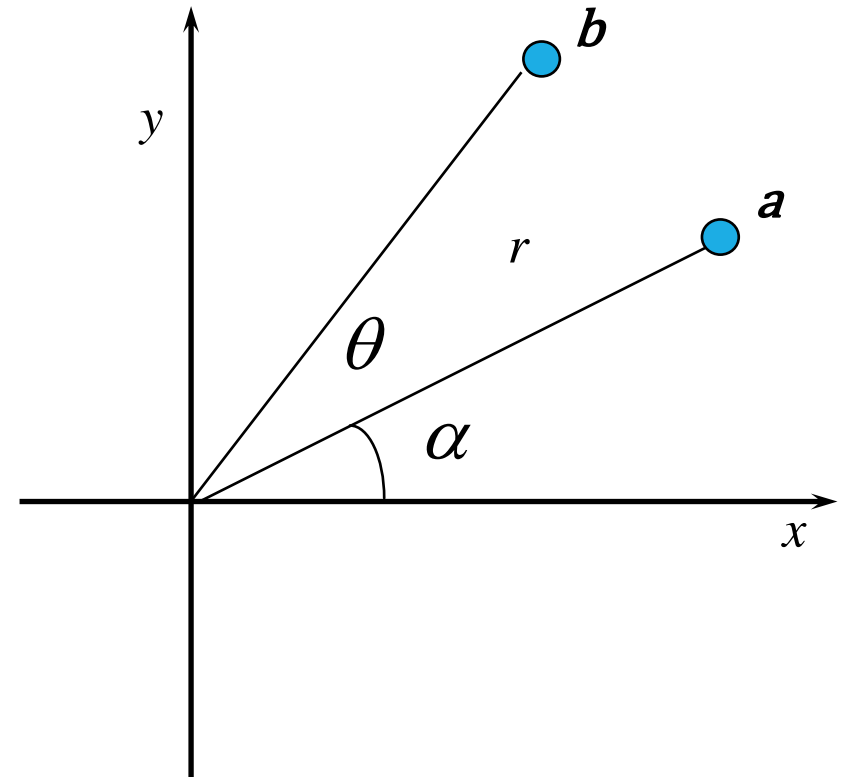
$$\mathbf{a} = (x_a, y_a) = (r \cos \alpha, r \sin \alpha)$$

where  $r = \sqrt{x_a^2 + y_a^2}$

- Rotating  $\mathbf{a}$  counter-clockwise by  $\theta$  to  $\mathbf{b}$ :

$$x_b = r \cos(\alpha + \theta) = r \cos \alpha \cos \theta - r \sin \alpha \sin \theta$$

$$y_b = r \sin(\alpha + \theta) = r \sin \alpha \cos \theta + r \cos \alpha \sin \theta$$

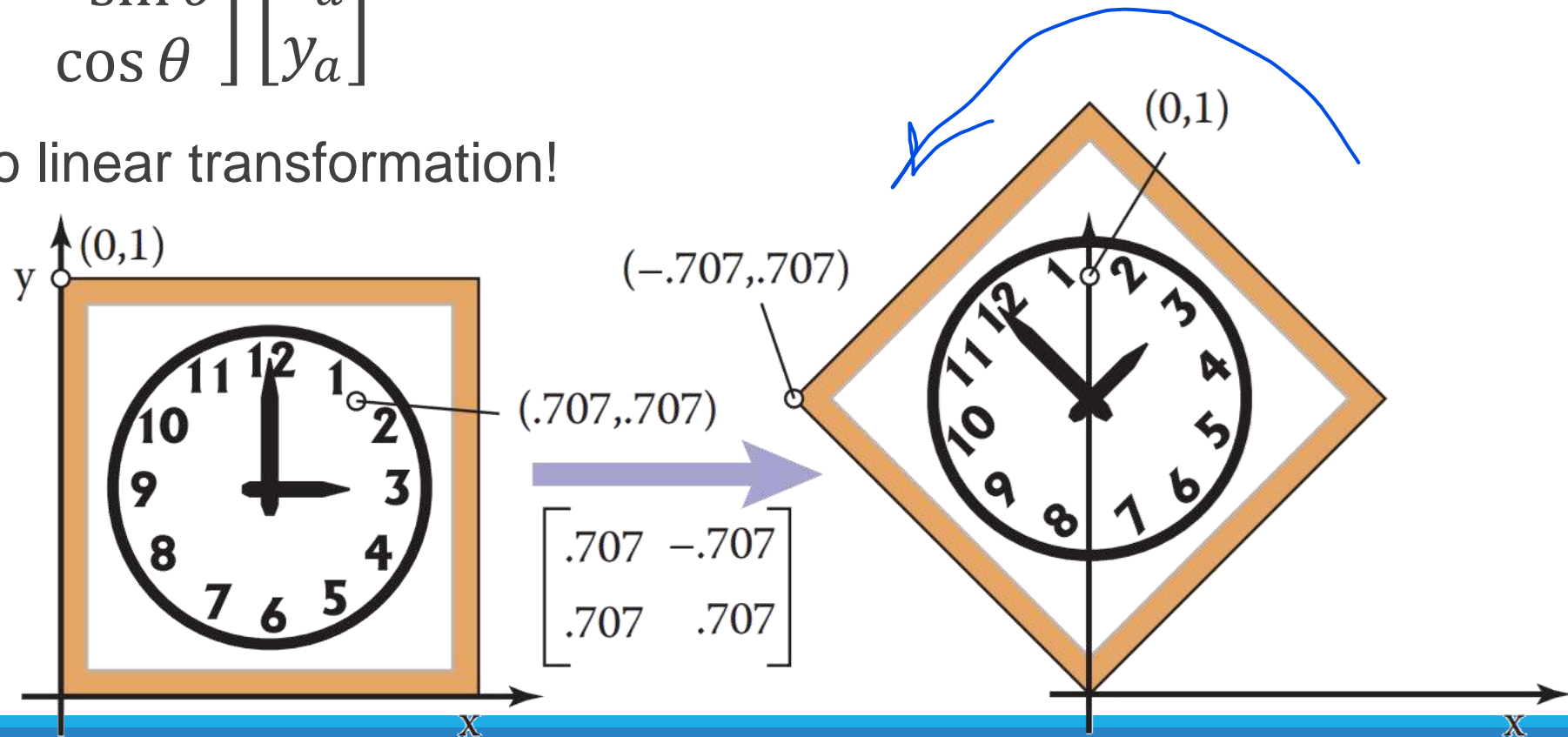


# 2D Rotation (cont'd)

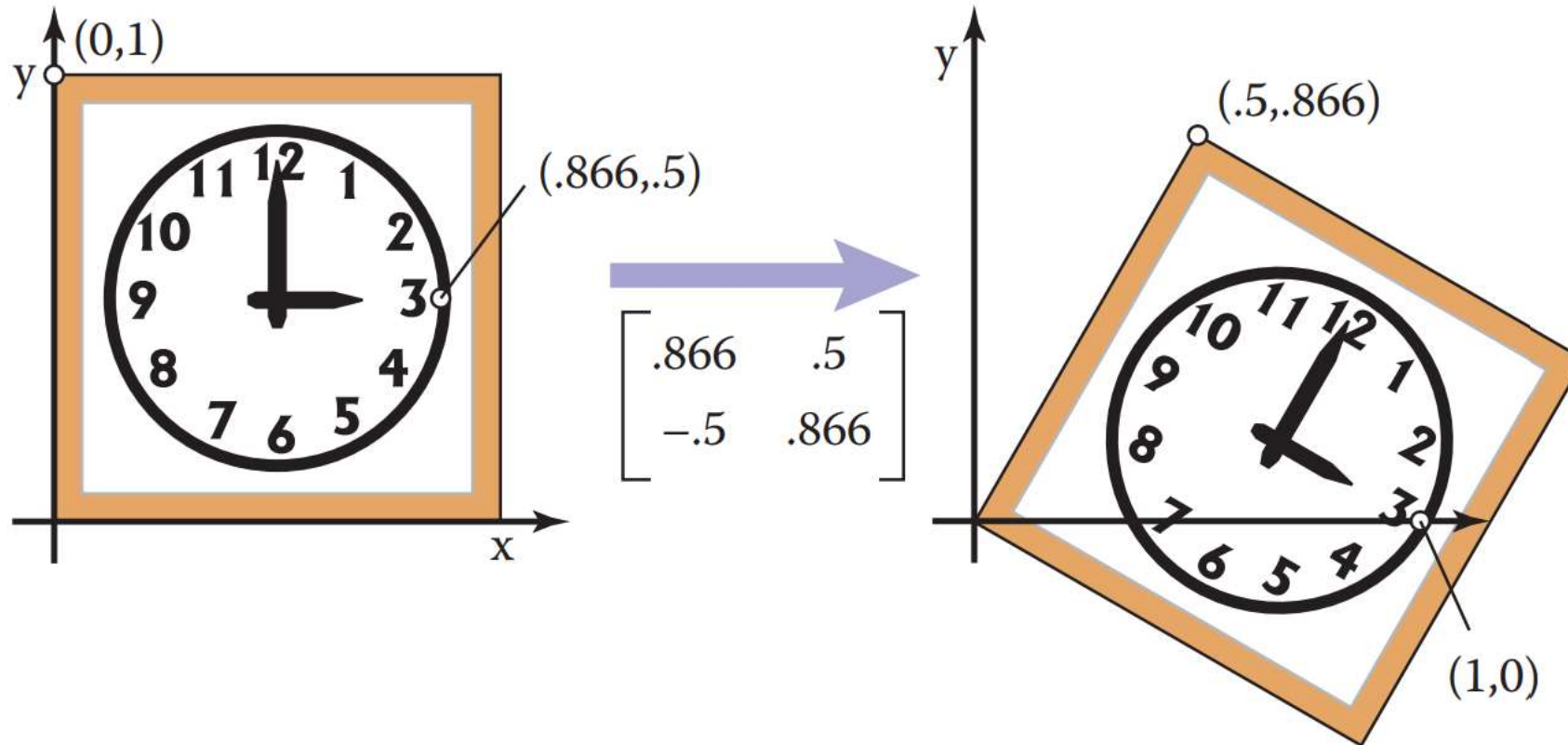
$$\begin{aligned}x_b &= r \cos \alpha \cos \theta - r \sin \alpha \sin \theta \\y_b &= r \sin \alpha \cos \theta + r \cos \alpha \sin \theta\end{aligned}$$

$$\begin{aligned}\begin{bmatrix} x_b \\ y_b \end{bmatrix} &= \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} r \cos \alpha \\ r \sin \alpha \end{bmatrix} \\ &= \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} x_a \\ y_a \end{bmatrix}\end{aligned}$$

$\therefore$  Rotation is also linear transformation!



# 2D Rotation (cont'd)



# Rotation Properties



$$R = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$

- Orthogonal matrix: Two columns (rows) are orthogonal  
 $(\cos \theta, \sin \theta) \cdot (-\sin \theta, \cos \theta) = 0$
- What is the inverse matrix  $R^{-1}$ , of  $R$  ? ( $R^{-1}R = RR^{-1} = I$ )

# Rotation Properties

- Matrix for rotating by  $-\theta$  angle is as follows:

$$R_{-\theta} = \begin{bmatrix} \cos(-\theta) & -\sin(-\theta) \\ \sin(-\theta) & \cos(-\theta) \end{bmatrix} = \begin{bmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{bmatrix} = R^T$$

- Geometrically,  $RR_{-\theta} = R_{-\theta}R = I$
- Therefore,  $RR^T = R^TR = I$

$$\therefore R^{-1} = R^T$$



# Other 2D Linear Transformations



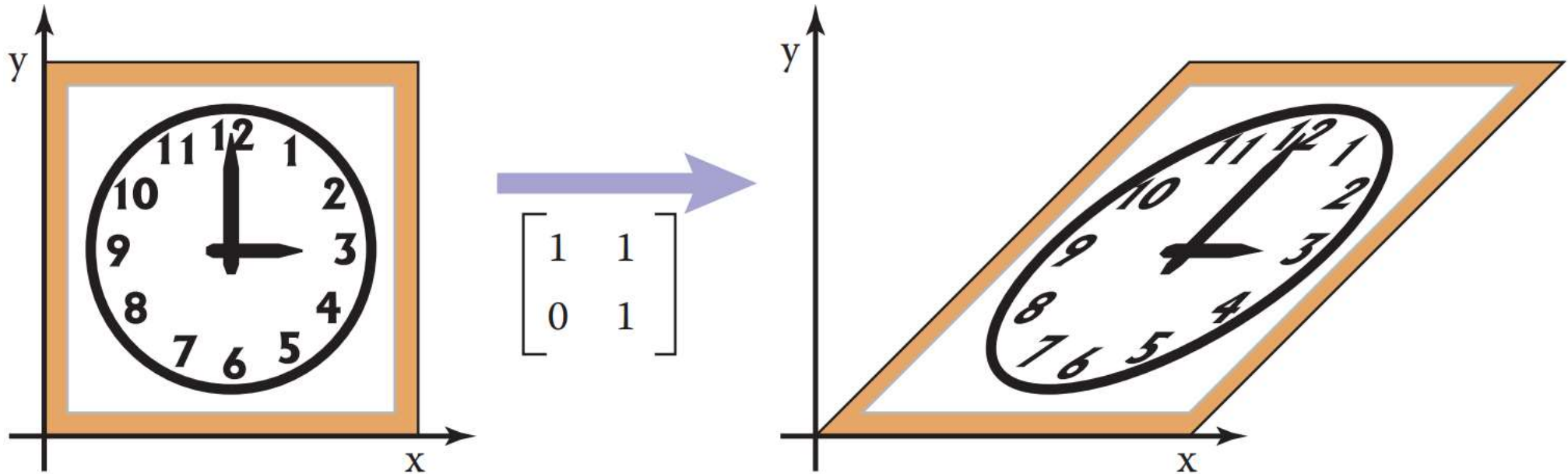
Q) What do these transformations do?

$$\begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & a \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ a & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} \begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix}$$

# 2D Shear



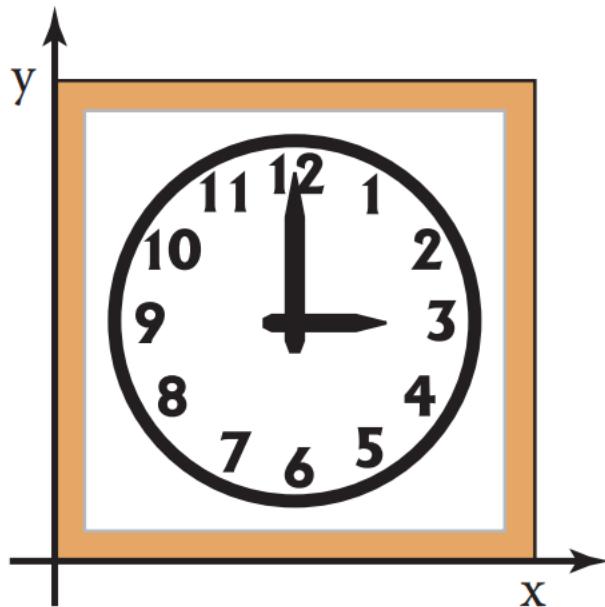
$$\text{shear}_x(s) = \begin{bmatrix} 1 & s \\ 0 & 1 \end{bmatrix}, \text{shear}_y(s) = \begin{bmatrix} 1 & 0 \\ s & 1 \end{bmatrix}$$




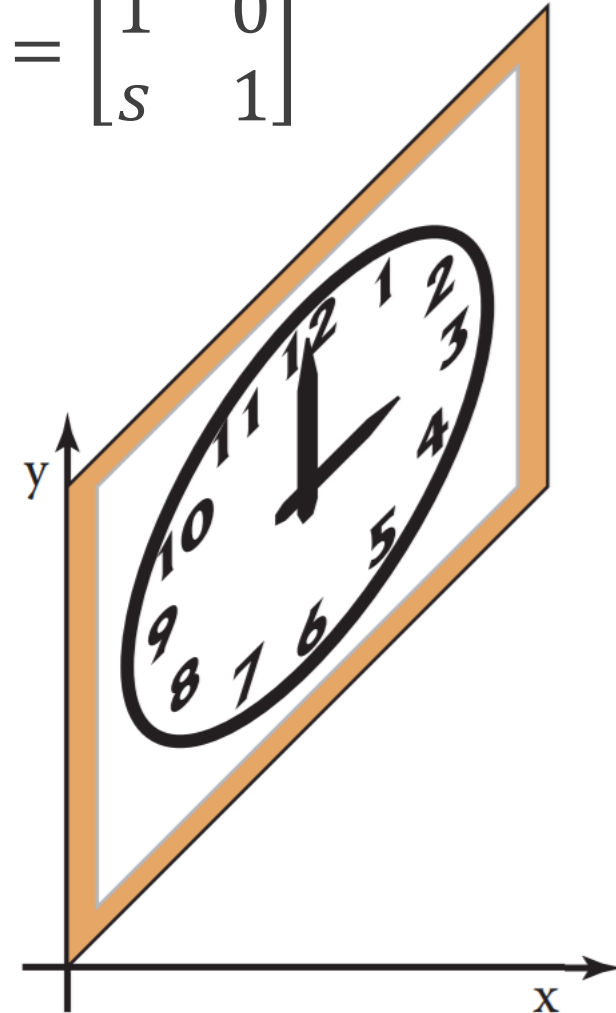
# 2D Shear



$$\text{shear}_x(s) = \begin{bmatrix} 1 & s \\ 0 & 1 \end{bmatrix}, \text{shear}_y(s) = \begin{bmatrix} 1 & 0 \\ s & 1 \end{bmatrix}$$




$$\begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix}$$

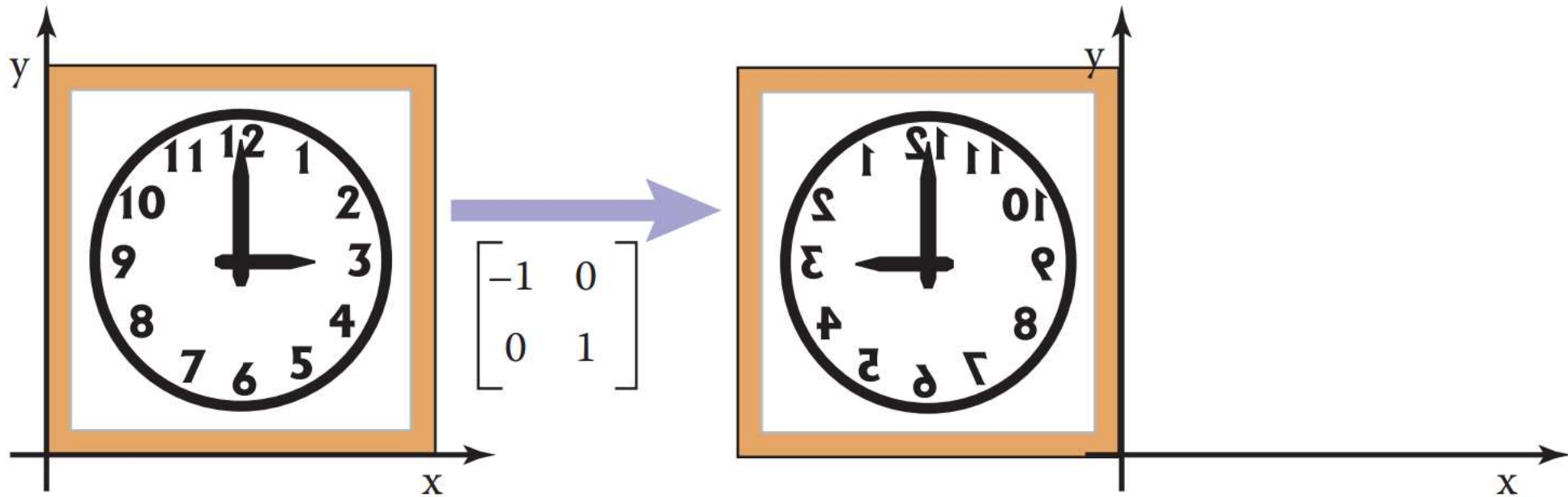


# 2D Reflection

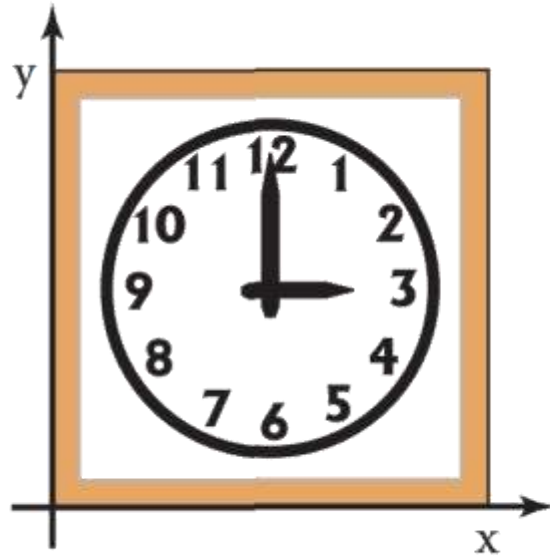


- Reflection: mirror a vector across either of the coordinate axes

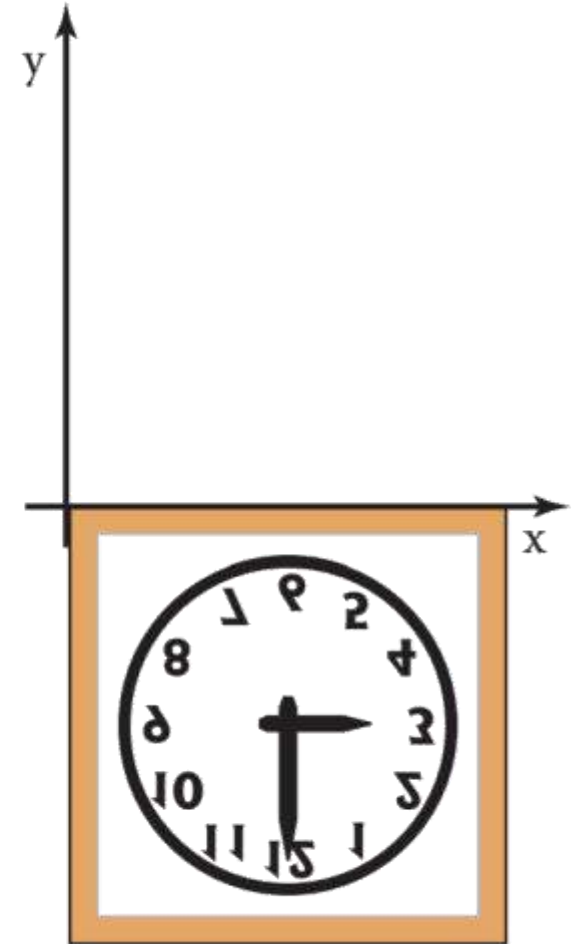
$$\text{reflect}_x = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}, \text{reflect}_y = \begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix}$$



# 2D Reflection



$$\begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$$



$$\text{reflect}_x = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix},$$
$$\text{reflect}_y = \begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix}$$

# Composition of Linear Transformations



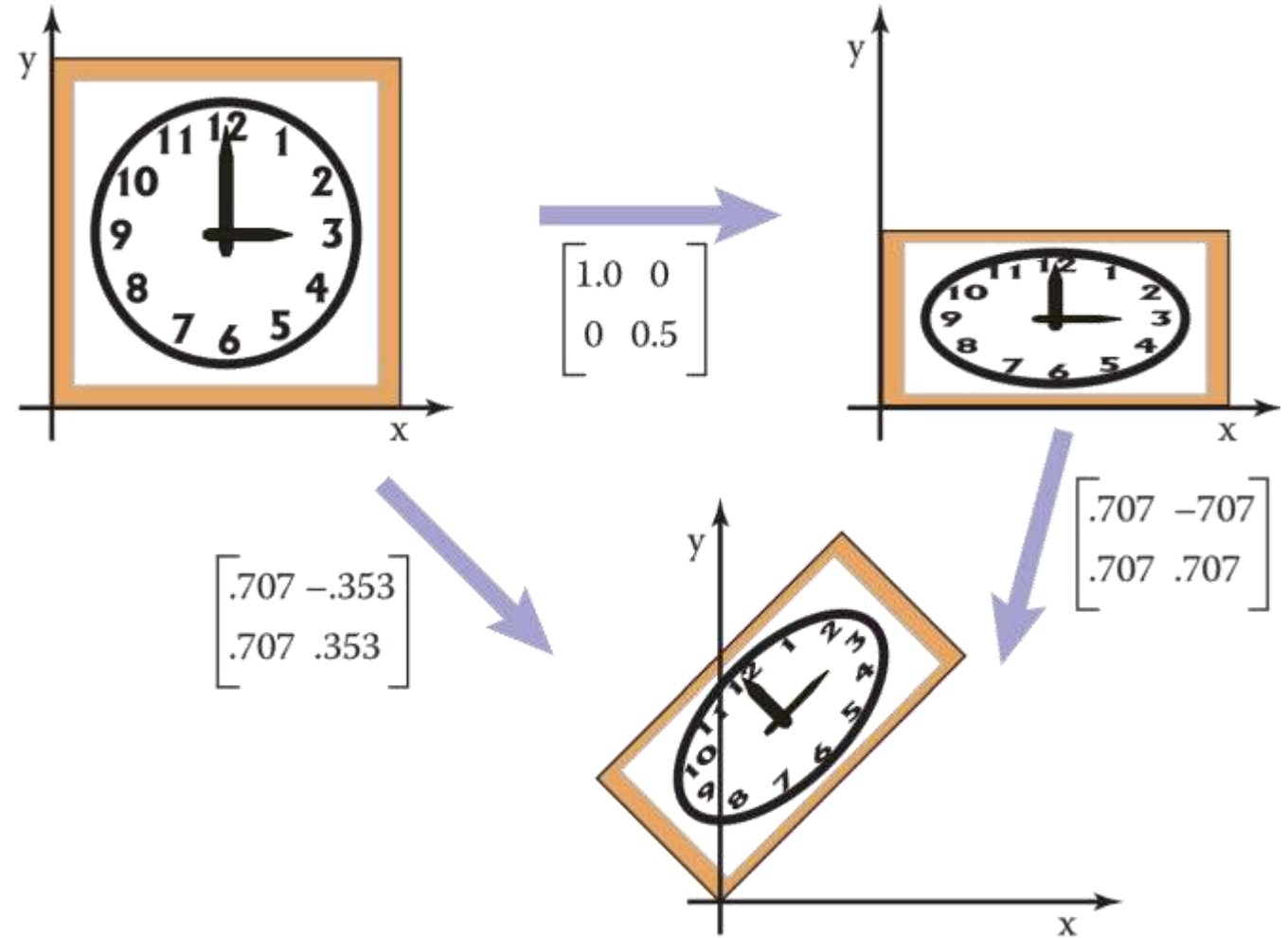
- Apply more than one transformation

⇒ Multiply transformation matrices

ex.  $v_2 = Sv_1$ , then  $v_3 = Rv_2$

⇒  $v_3 = Rv_2 = R(Sv_1) = (RS)v_1$

⇒  $M = RS$

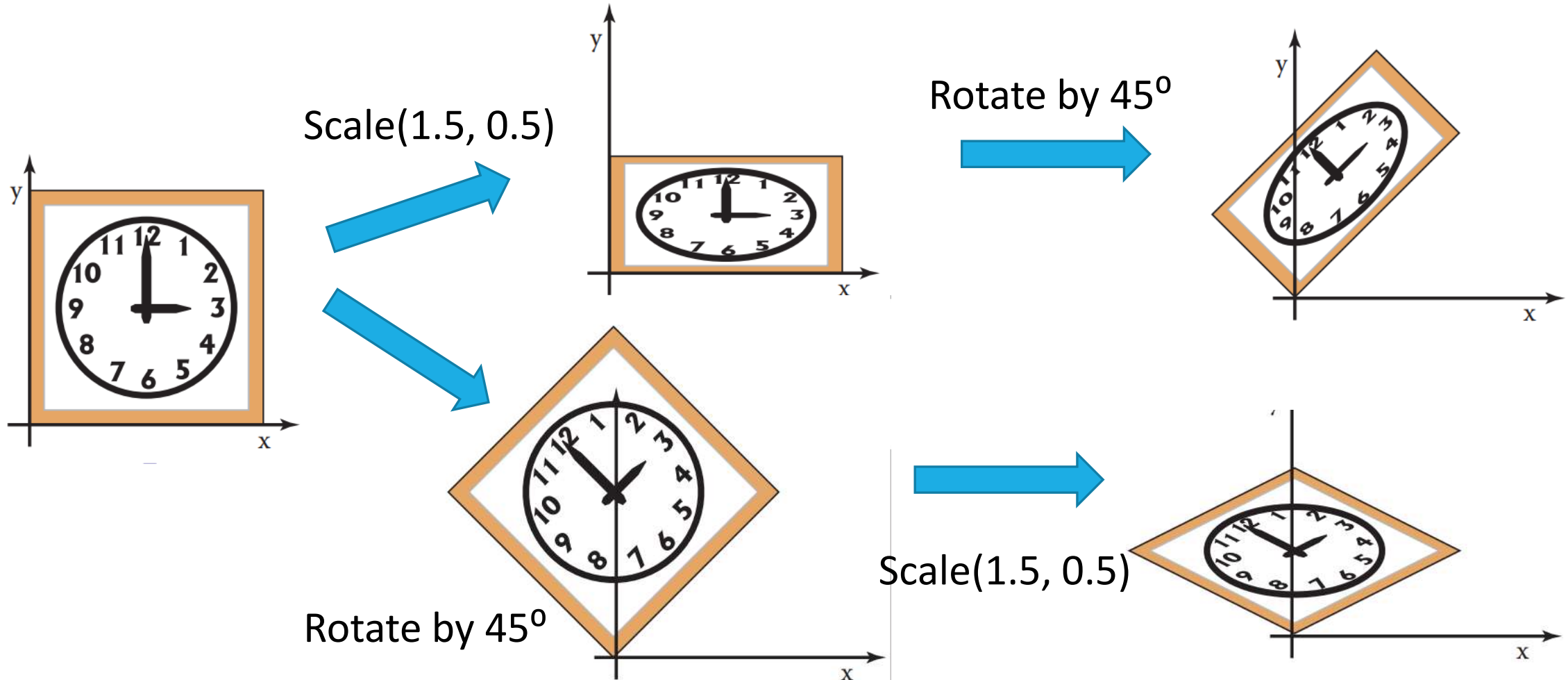


First apply non-uniform scaling, and then rotate by 45 degrees

# Composition of Transformations



- What about changing the order of transformation?

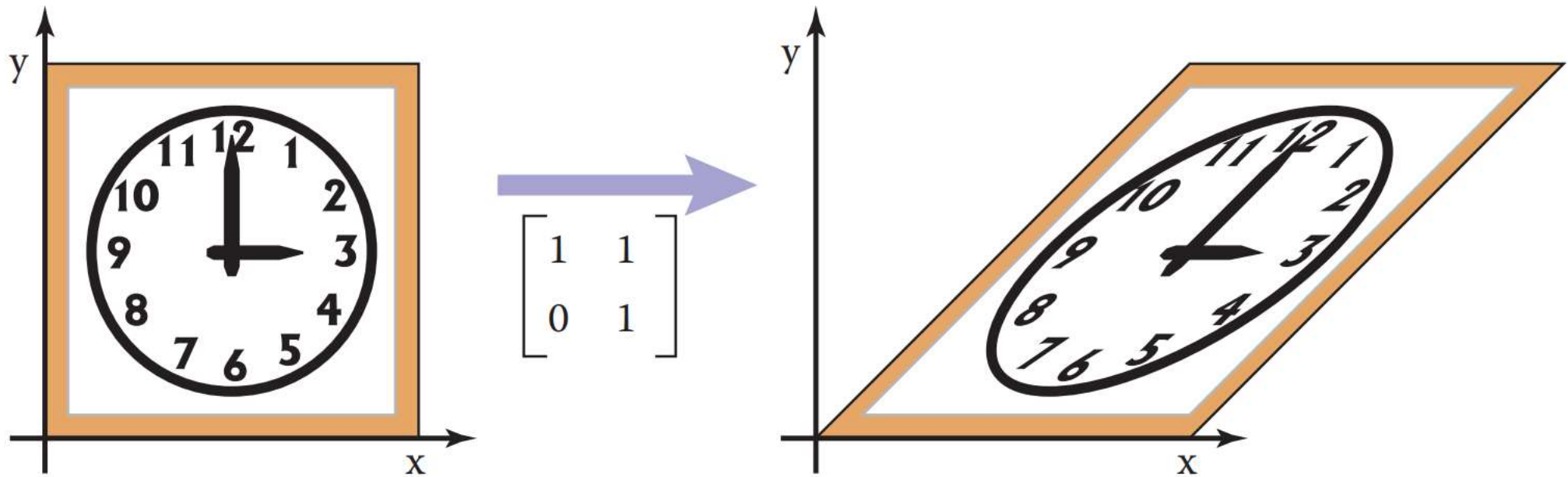


# Decomposition of Transformations



Q) Can we express shear transformation by the product of scaling and rotation transformation matrices?

Q) What about arbitrary linear transformation?





# Decomposition of Transformation



- For symmetric transformation, use eigen value decomposition

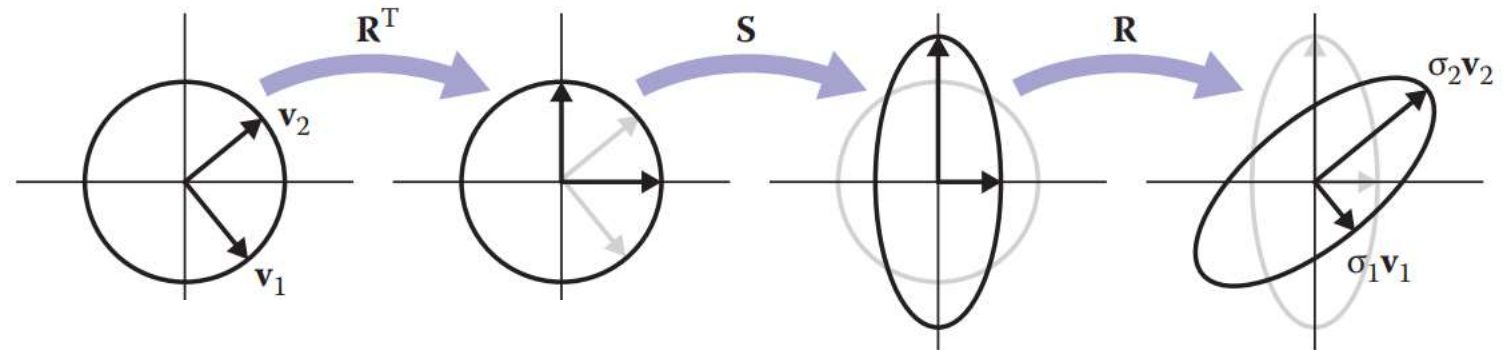
$$A = RSR^T = R \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix} R^T$$

ex)

$$\begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix} = \mathbf{R} \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix} \mathbf{R}^T$$

$$= \begin{bmatrix} 0.8507 & -0.5257 \\ 0.5257 & 0.8507 \end{bmatrix} \begin{bmatrix} 2.618 & 0 \\ 0 & 0.382 \end{bmatrix} \begin{bmatrix} 0.8507 & 0.5257 \\ -0.5257 & 0.8507 \end{bmatrix}$$

$$= \text{rotate } (31.7^\circ) \text{ scale } (2.618, 0.382) \text{ rotate } (-31.7^\circ).$$



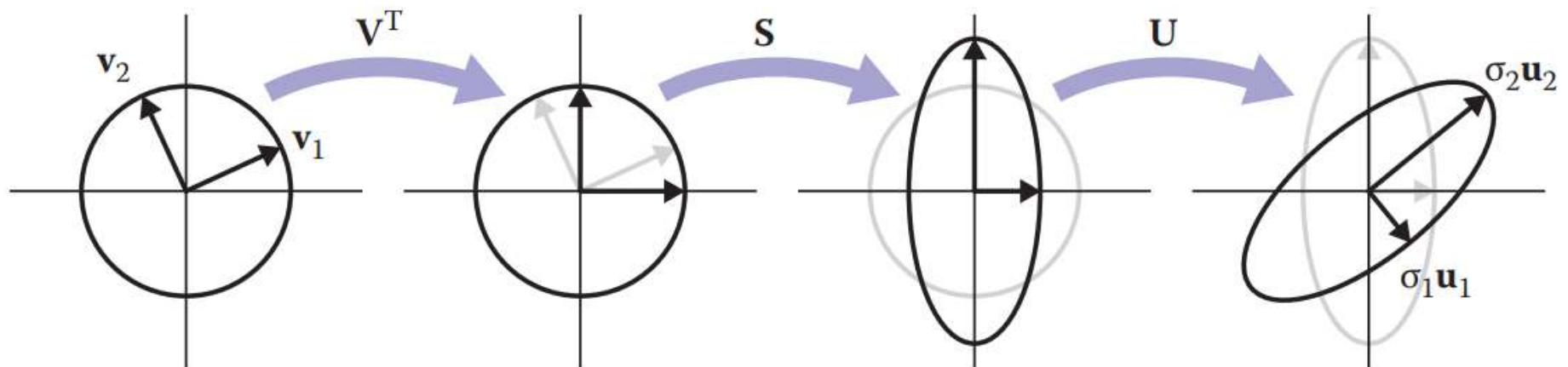
# Decomposition of Transformation



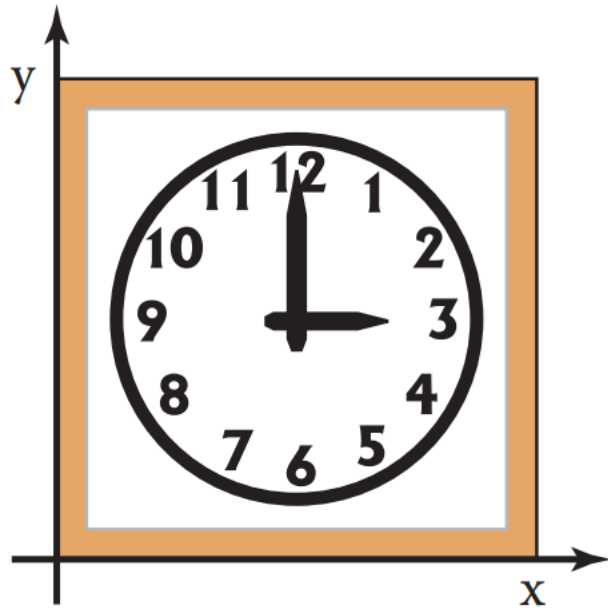
- For arbitrary transformation, use singular value decomposition (SVD)

$$A = USV^T$$

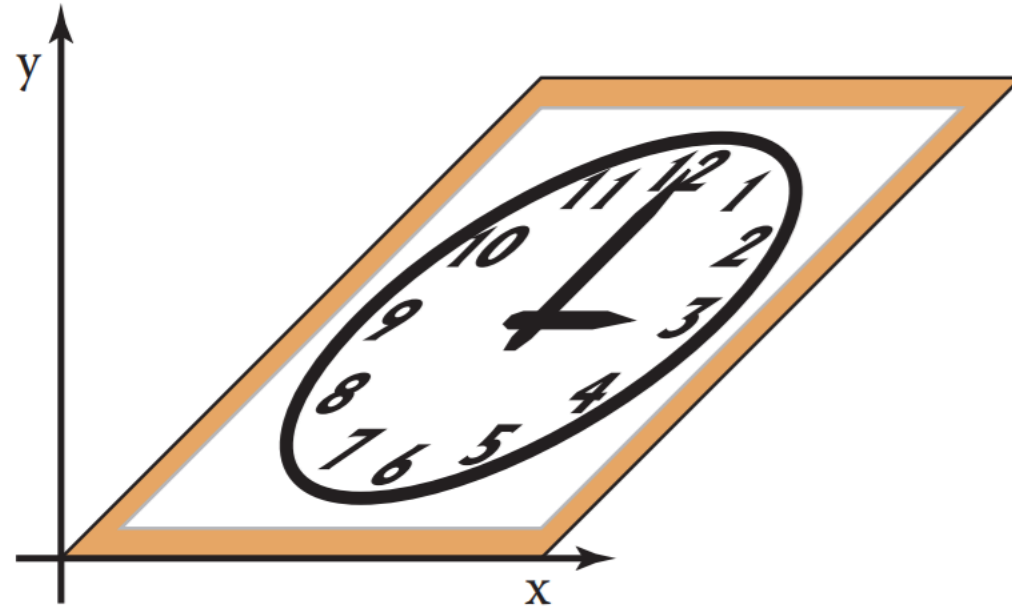
where  $U$ ,  $V$  are orthogonal (rotation) matrices, and  $S$  (eigen values) is a (non-uniform) scale matrix



# Decomposition of Transformation



$$\begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$$



$$\begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} = \mathbf{R}_2 \begin{bmatrix} \sigma_1 & 0 \\ 0 & \sigma_2 \end{bmatrix} \mathbf{R}_1$$

$$= \begin{bmatrix} 0.8507 & -0.5257 \\ 0.5257 & 0.8507 \end{bmatrix} \begin{bmatrix} 1.618 & 0 \\ 0 & 0.618 \end{bmatrix} \begin{bmatrix} 0.5257 & 0.8507 \\ -0.8507 & 0.5257 \end{bmatrix}$$

$$= \text{rotate } (31.7^\circ) \text{ scale } (1.618, 0.618) \text{ rotate } (-58.3^\circ).$$

# 2D Translation

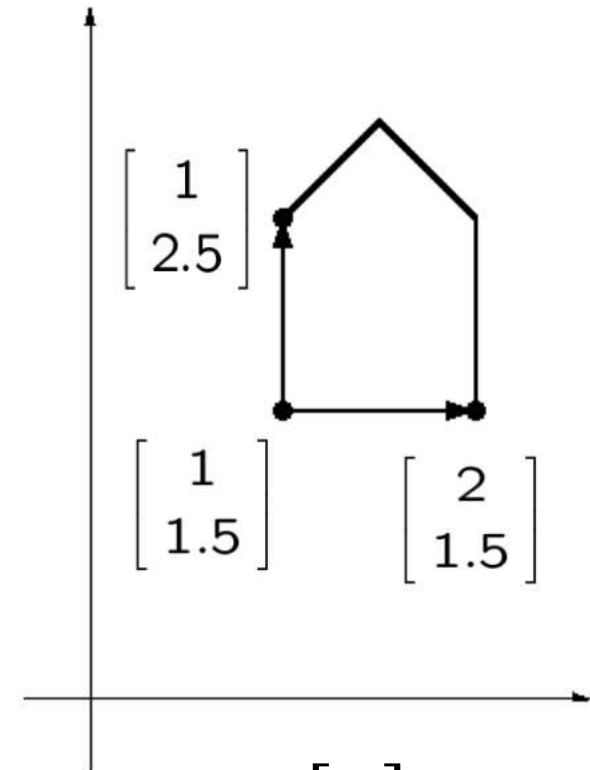


- Translation: move all points of an object by the same amounts

$$x' = x + x_t$$

$$y' = y + y_t$$

- Translation cannot be expressed by 2 x 2 linear transformation matrix
- How to solve this problem?



$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} x \\ y \end{bmatrix} + \begin{bmatrix} 1 \\ 1.5 \end{bmatrix}$$

# Homogeneous Coordinate

- A point (x,y) in 2D is represented by

$$\begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

- Homogeneous coordinates can express both linear transformation and translation at the same time

⇒ Affine transformation

- Rigid-body transformation: translation + rotation

# Homogeneous Coordinate



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

$$\begin{bmatrix} \hat{x} \\ \hat{y} \end{bmatrix} = \begin{bmatrix} \alpha & 0 \\ 0 & \beta \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

$$\begin{bmatrix} \hat{x} \\ \hat{y} \end{bmatrix} = \begin{bmatrix} t_x \\ t_y \end{bmatrix} + \begin{bmatrix} x \\ y \end{bmatrix}$$



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

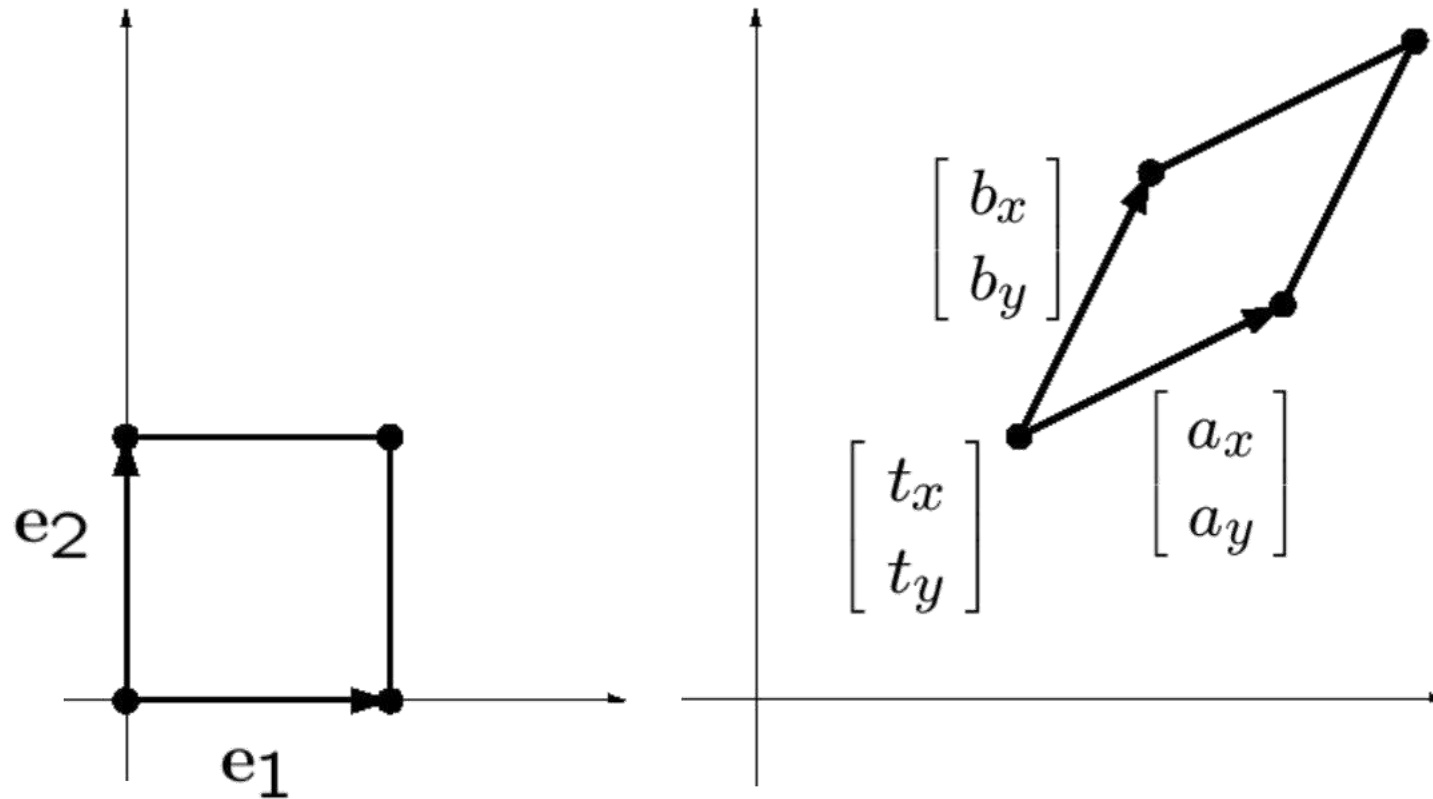
$$\begin{bmatrix} \hat{x} \\ \hat{y} \\ 1 \end{bmatrix} = \begin{bmatrix} \alpha & 0 & 0 \\ 0 & \beta & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

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

# Homogeneous Coordinates



Q) Write a transformation matrix for the following transformation

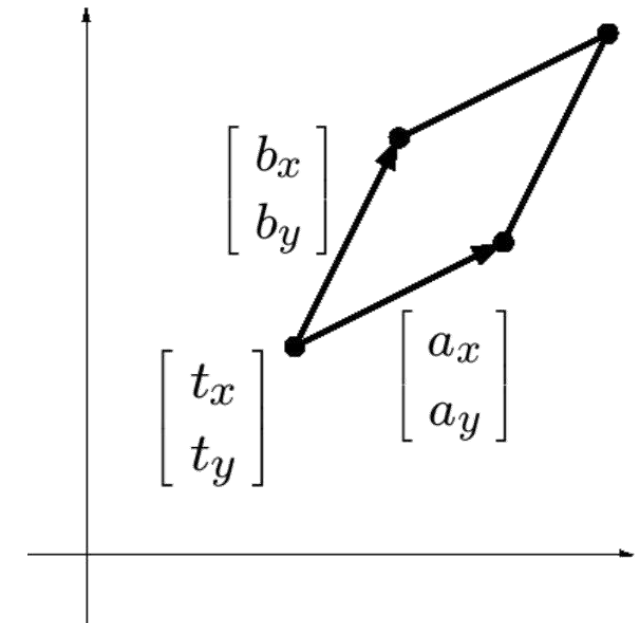
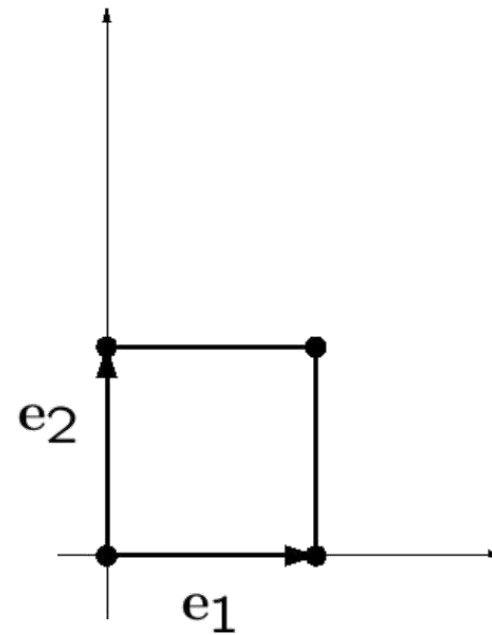


# Homogeneous Coordinates



$$\begin{bmatrix} \hat{x} \\ \hat{y} \\ 1 \end{bmatrix} = \begin{bmatrix} a_x & b_x & t_x \\ a_y & b_y & t_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} \hat{x} \\ \hat{y} \end{bmatrix} = \begin{bmatrix} a_x & b_x \\ a_y & b_y \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} + \begin{bmatrix} t_x \\ t_y \end{bmatrix}$$





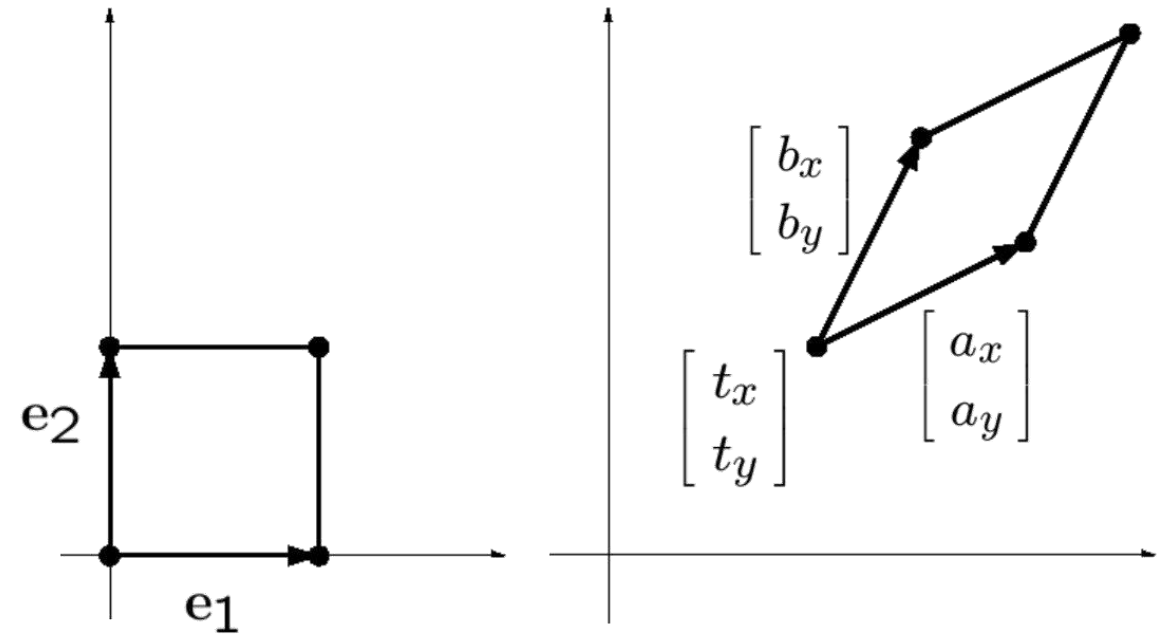
# Homogeneous Coordinates



$$\begin{bmatrix} t_x \\ t_y \\ 1 \end{bmatrix} = \begin{bmatrix} a_x & b_x & t_x \\ a_y & b_y & t_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} a_x \\ a_y \\ 0 \end{bmatrix} = \begin{bmatrix} a_x & b_x & t_x \\ a_y & b_y & t_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$

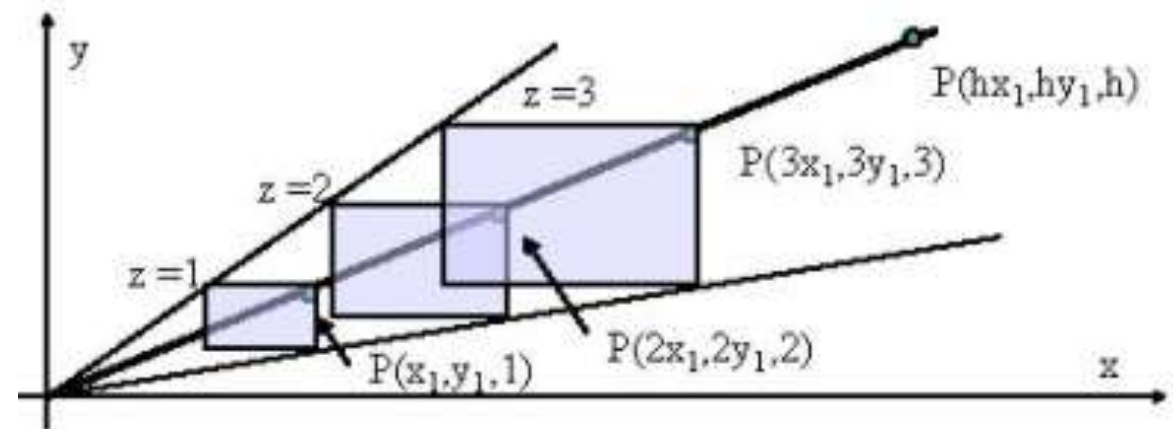
$$\begin{bmatrix} b_x \\ b_y \\ 0 \end{bmatrix} = \begin{bmatrix} a_x & b_x & t_x \\ a_y & b_y & t_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$



# Homogeneous Coordinates

- A point  $p$  is expressed as  $p = \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$
- A vector  $v$  is expressed as  $v = \begin{bmatrix} x \\ y \\ 0 \end{bmatrix}$
- In homogeneous coordinates,

$$p = \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} hx \\ hy \\ h \end{bmatrix}$$

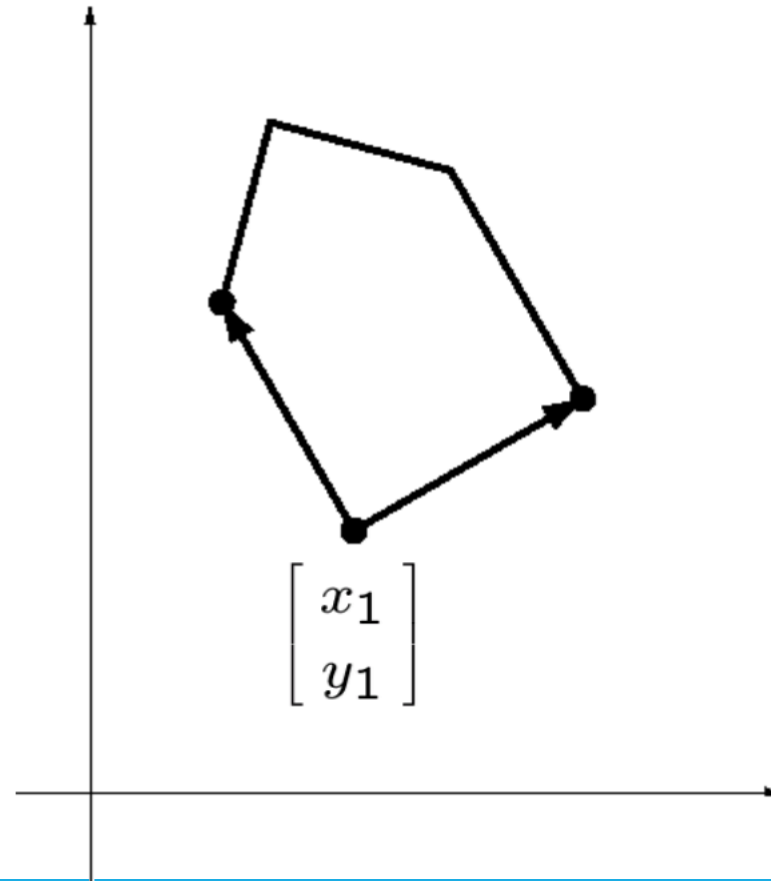
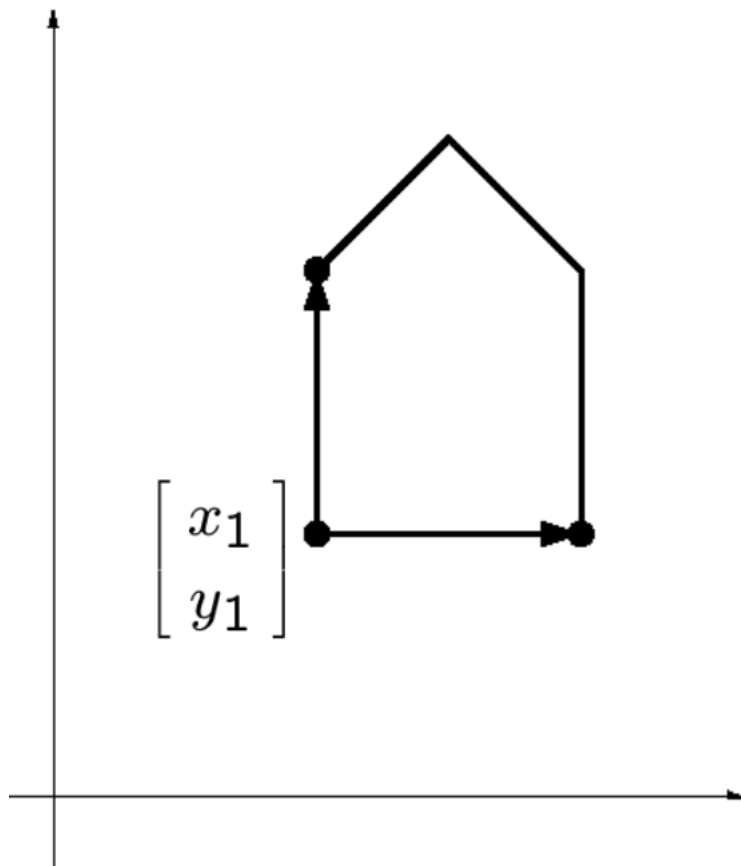


3D Representation of homogeneous space

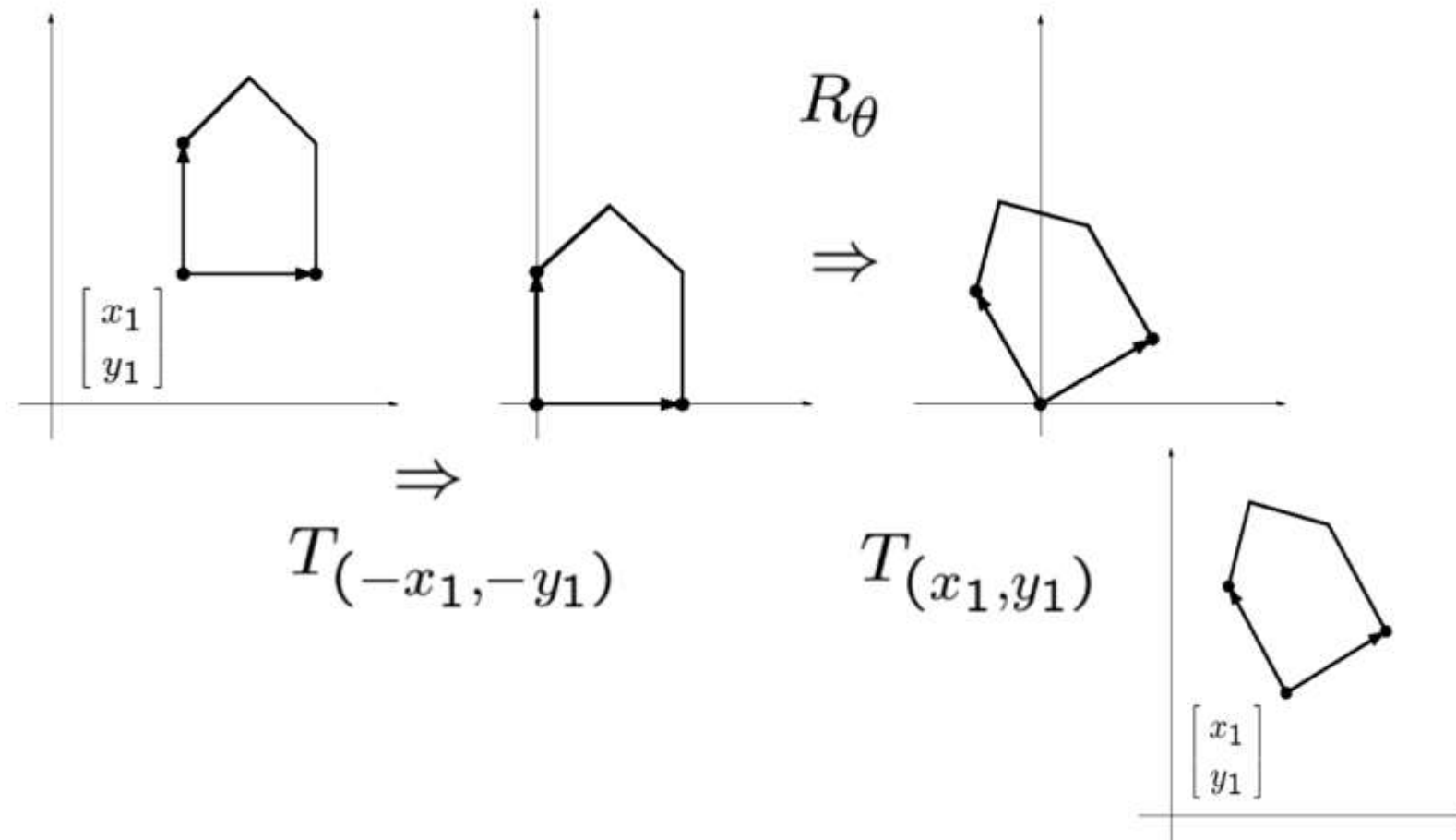
# 일반적인 2D Transformation



Q) Write a transformation matrix for the following transformation



# 일반적인 2D Transformation



# 일반적인 2D Transformation



$$\begin{aligned} & \begin{bmatrix} 1 & 0 & x_1 \\ 0 & 1 & y_1 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \cos \theta & -\sin \theta & 0 \\ \sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & -x_1 \\ 0 & 1 & -y_1 \\ 0 & 0 & 1 \end{bmatrix} \\ &= \begin{bmatrix} 1 & 0 & x_1 \\ 0 & 1 & y_1 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \cos \theta & -\sin \theta & -x_1 \cos \theta + y_1 \sin \theta \\ \sin \theta & \cos \theta & -x_1 \sin \theta - y_1 \cos \theta \\ 0 & 0 & 1 \end{bmatrix} \\ &= \begin{bmatrix} \cos \theta & -\sin \theta & x_1(1 - \cos \theta) + y_1 \sin \theta \\ \sin \theta & \cos \theta & y_1(1 - \cos \theta) - x_1 \sin \theta \\ 0 & 0 & 1 \end{bmatrix} \end{aligned}$$

# Summary

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- Points, scalars, vectors
- Coordinates, coordinate frames
- 2D transformation
- Homogeneous coordinates