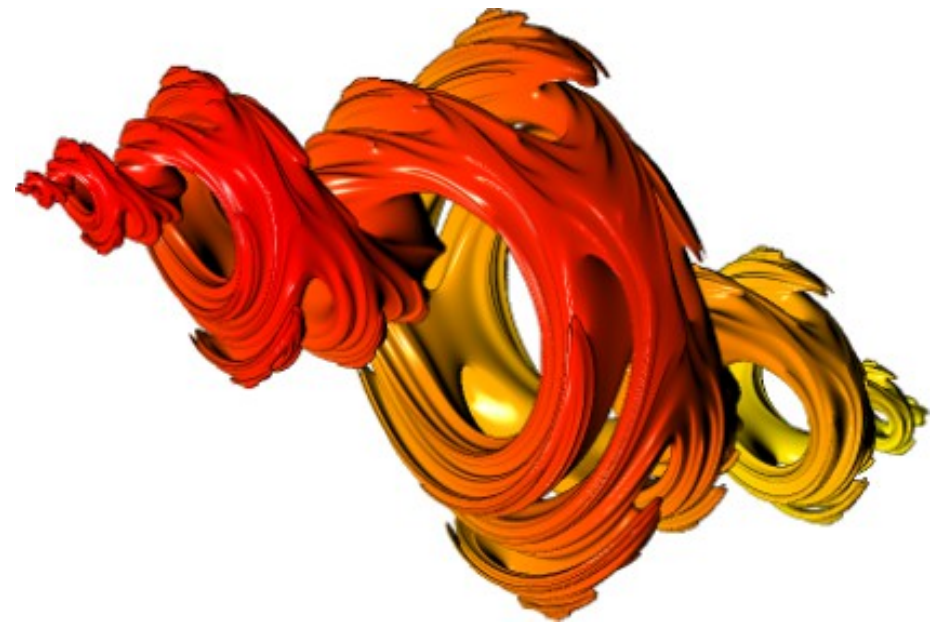
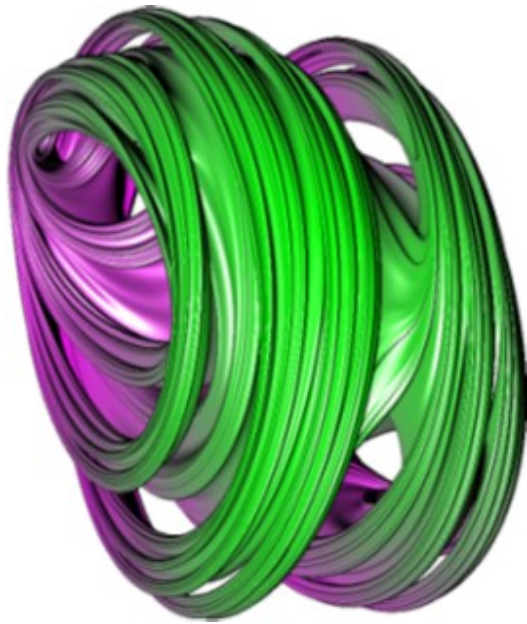
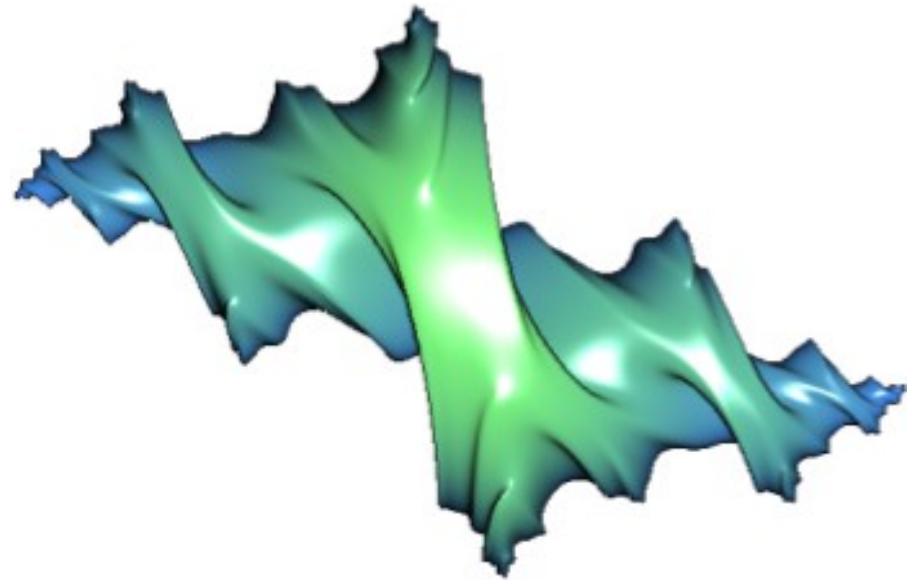
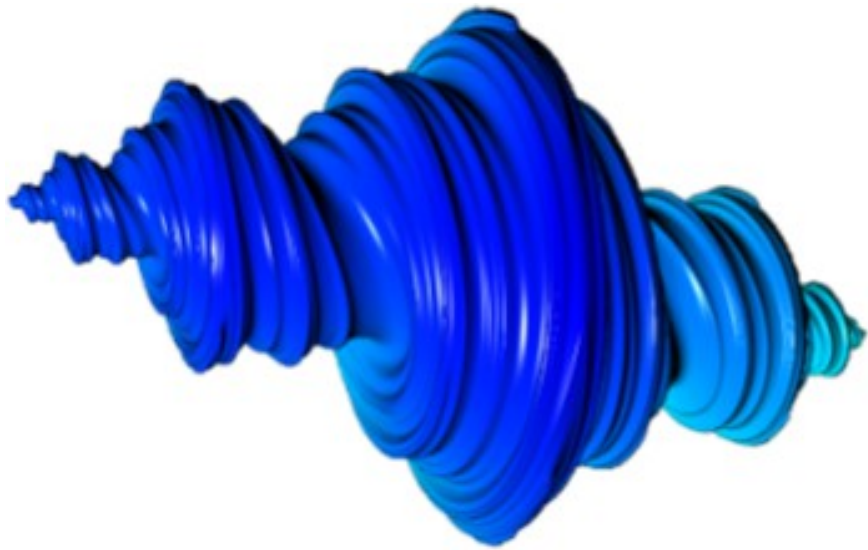


# Geometry

CS475 / 675, Fall 2016

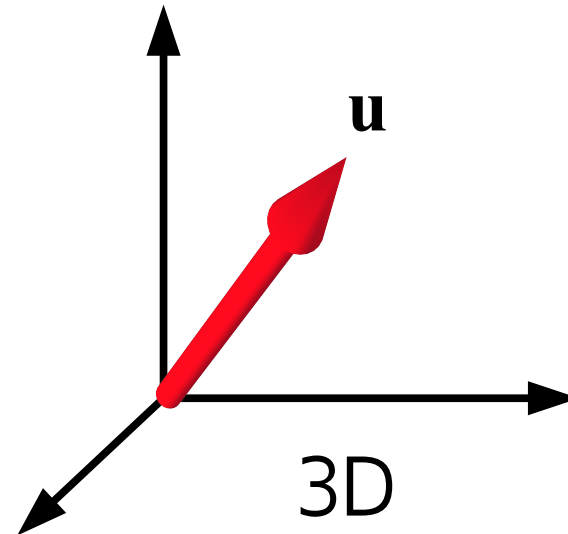
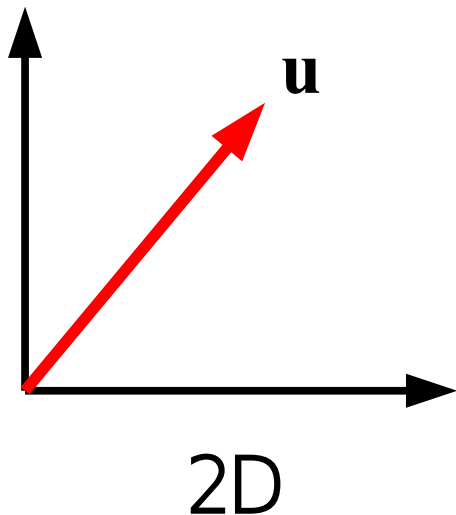
Siddhartha Chaudhuri



Ödegaard and Wennergren, 2D projections of 4D “Julia sets”

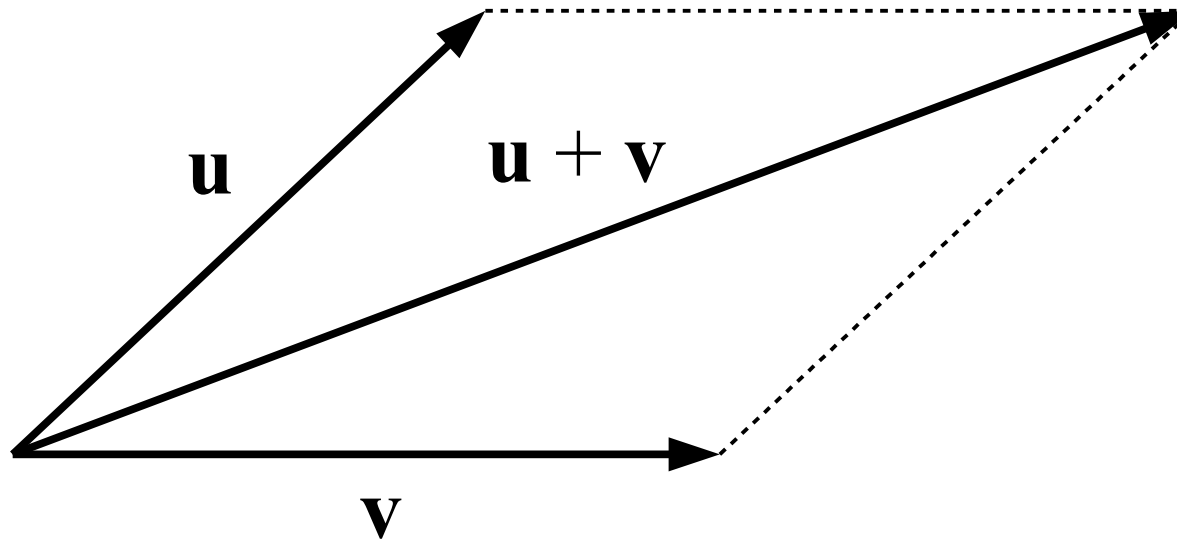
# Vectors: the “Physical” View

- Object  $\mathbf{u}$  with direction and length/magnitude
- $-\mathbf{u}$  is vector in opposite direction with same length
- $\|\mathbf{u}\|$  denotes length of  $\mathbf{u}$ 
  - *Unit vector* has length 1, is typically written with hat:  $\hat{\mathbf{u}}$
- Multiplying a vector by a scalar changes its length



# Addition of Vectors

- “Parallelogram Rule”

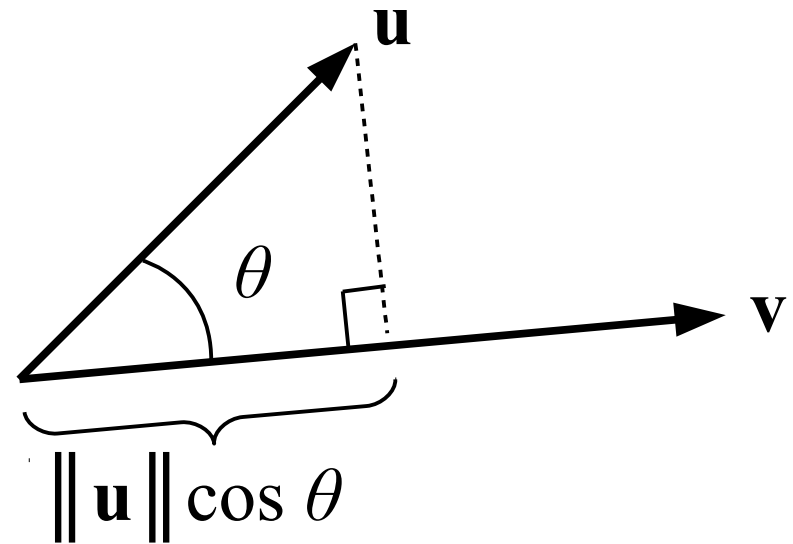


- $\mathbf{u} + (-\mathbf{v}) \equiv \mathbf{u} - \mathbf{v}$

# Dot Product

- If  $\theta$  is the angle between  $\mathbf{u}$  and  $\mathbf{v}$ , then

$$\mathbf{u} \cdot \mathbf{v} = \|\mathbf{u}\| \|\mathbf{v}\| \cos \theta$$



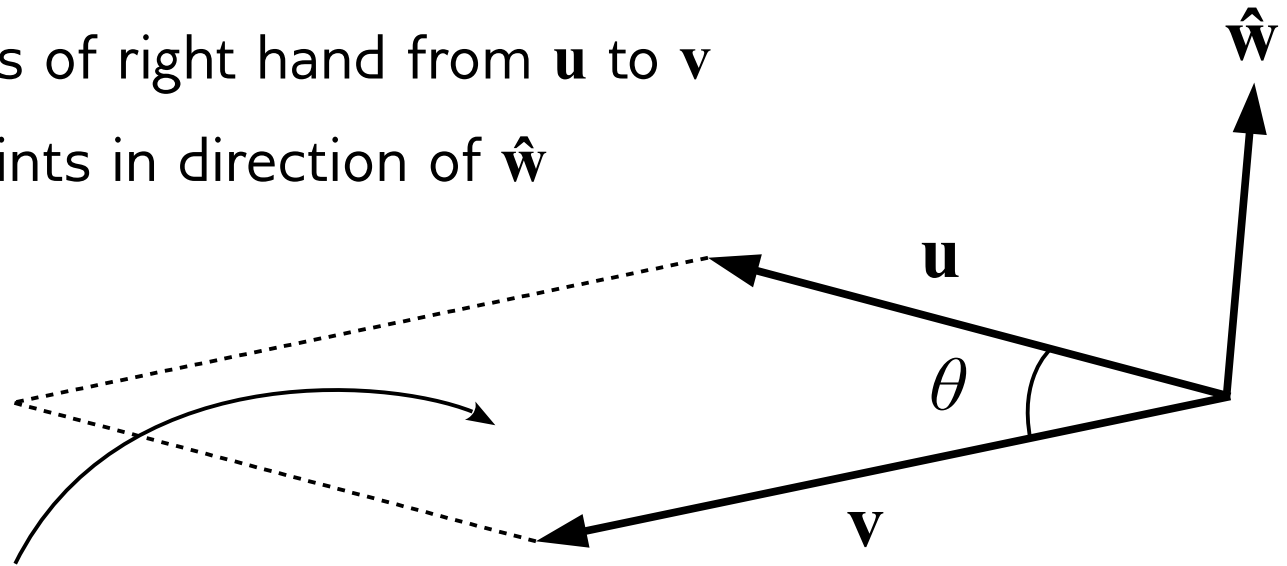
- This is also the length of the orthogonal projection of  $\mathbf{u}$  onto  $\mathbf{v}$ , times the length of  $\mathbf{v}$
- **Note:**  $\mathbf{u} \cdot \mathbf{u} = \|\mathbf{u}\|^2$

# Orthogonal and Orthonormal Vectors

- *Orthogonal* vectors are at right angles to each other
  - $\mathbf{u} \cdot \mathbf{v} = 0$
- *Orthonormal* vectors are at right angles to each other *and* each has unit length
- A set  $\{\mathbf{u}_i\}$  of orthonormal vectors has
  - $\mathbf{u}_i \cdot \mathbf{u}_j = 1$  if  $i = j$
  - $\mathbf{u}_i \cdot \mathbf{u}_j = 0$  if  $i \neq j$

# Cross Product (only 3D)

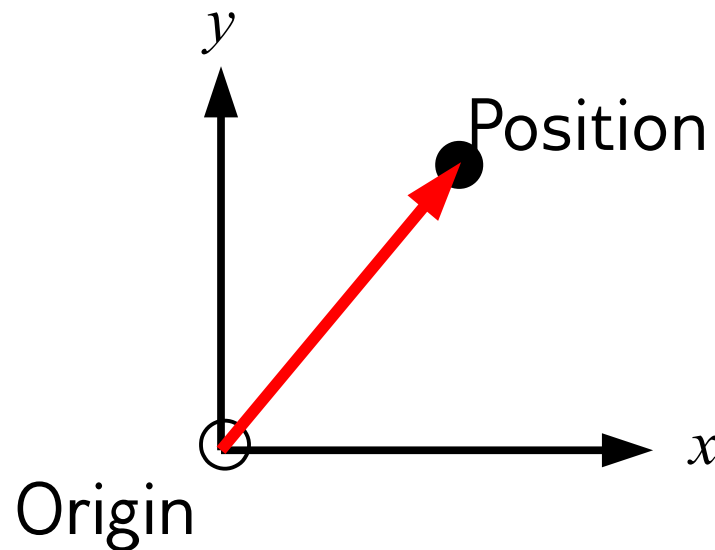
- If  $\theta$  is the angle between  $\mathbf{u}$  and  $\mathbf{v}$ , then 
$$\mathbf{u} \times \mathbf{v} = (\|\mathbf{u}\| \|\mathbf{v}\| \sin \theta) \hat{\mathbf{w}}$$
  - $\hat{\mathbf{w}}$  is a unit vector orthogonal to both  $\mathbf{u}$  and  $\mathbf{v}$
  - “Right-Hand Rule”:
    - Curl fingers of right hand from  $\mathbf{u}$  to  $\mathbf{v}$
    - Thumb points in direction of  $\hat{\mathbf{w}}$



Area of parallelogram =  $\|\mathbf{u} \times \mathbf{v}\|$

# Position Vector

- Identifies point in space
  - Interpreted as: tip of vector “arrow” when the other end is placed at a fixed point called the *origin* of the space





# Vectors: the “Physical” View

- Directions and positions are *different!*
- “Legal” Operations:
  - $\text{Direction} = \text{Scalar} * \text{Direction}$
  - $\text{Direction} = \text{Direction} + \text{Direction}$
  - $\text{Position} = \text{Position} + \text{Direction}$
  - $\text{Direction} = \text{Position} - \text{Position}$
- “Illegal” Operations:
  - $\text{Position} = \text{Scalar} * \text{Position}$
  - $\text{Position} = \text{Position} + \text{Position}$
  - $\text{Direction} = \text{Position} + \text{Direction}$
  - $\text{Position} = \text{Position} - \text{Position}$

# Cartesian/Euclidean $n$ -space $\mathbb{R}^n$

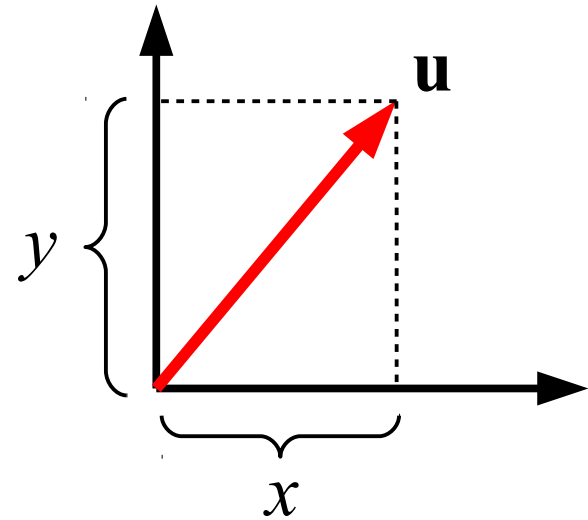
- Vectors represented by real  $n$ -tuples of *coordinates*  $(u_1, u_2, \dots, u_n)$

- 2D:  $(x, y)$

- 3D:  $(x, y, z)$

- Represents extent along orthogonal *coordinate axes*

- Right-handed system: Curl right hand fingers from  $x$  axis to  $y$  axis, thumb points along  $z$  axis



- Length/magnitude:  $\|\mathbf{u}\| = (u_1^2 + u_2^2 + \dots + u_n^2)^{1/2}$

- From Pythagoras' Theorem

# Cartesian/Euclidean $n$ -space $\mathbb{R}^n$

- Let  $\mathbf{u} = (u_1, u_2, \dots, u_n)$  and  $\mathbf{v} = (v_1, v_2, \dots, v_n)$
- Addition of vectors:

$$\mathbf{u} + \mathbf{v} \equiv (u_1 + v_1, u_2 + v_2, \dots, u_n + v_n)$$

- Multiplication of vector by scalar:

$$s\mathbf{u} \equiv (su_1, su_2, \dots, su_n)$$

# Cartesian/Euclidean $n$ -space $\mathbb{R}^n$

- Dot Product:

$$\mathbf{u} \cdot \mathbf{v} \equiv (u_1 v_1 + u_2 v_2 + \dots + u_n v_n)$$

- Cross Product (remember, 3D only!):

$$\mathbf{u} \times \mathbf{v} \equiv (u_2 v_3 - u_3 v_2, u_3 v_1 - u_1 v_3, u_1 v_2 - u_2 v_1)$$

# Another Way to Remember the Cross Product in $\mathbb{R}^3$

- Let  $\mathbf{u} = (x_1, y_1, z_1)$  and  $\mathbf{v} = (x_2, y_2, z_2)$
- $\mathbf{u} \times \mathbf{v} \equiv (y_1 z_2 - z_1 y_2, z_1 x_2 - x_1 z_2, \underbrace{x_1 y_2 - y_1 x_2})$

If  $\mathbf{u}$  and  $\mathbf{v}$  are 2D instead, this is the area of the parallelogram between them

# Vectors: the “Mathematical” View

- *Vector space*: Set of objects (*vectors*) closed under
  - Addition of two vectors
  - Multiplication of a vector by a scalar (from field  $F$ )
- Necessary properties of vector spaces
  - Commutative addition:  $\mathbf{x} + \mathbf{y} = \mathbf{y} + \mathbf{x}$
  - Associative addition:  $(\mathbf{x} + \mathbf{y}) + \mathbf{z} = \mathbf{x} + (\mathbf{y} + \mathbf{z})$
  - Existence of additive identity  $\mathbf{0}$ :  $\mathbf{x} + \mathbf{0} = \mathbf{0} + \mathbf{x}$ 
    - Can show:  $0 \cdot \mathbf{x} = \mathbf{0}$  (prove!)
  - Existence of additive inverse  $-\mathbf{x}$ :  $\mathbf{x} + (-\mathbf{x}) = \mathbf{0}$ 
    - Can show:  $-\mathbf{x} = -1 \cdot \mathbf{x}$  (prove!)

# Vectors: the “Mathematical” View

- Necessary properties of vector spaces (contd.)
  - Associative scalar multiplication:  $r(s\mathbf{x}) = (rs)\mathbf{x}$
  - Distributive scalar sum:  $(r + s)\mathbf{x} = r\mathbf{x} + s\mathbf{x}$
  - Distributive scalar multiplication:  $r(\mathbf{x} + \mathbf{y}) = r\mathbf{x} + r\mathbf{y}$
  - Scalar multiplication identity:  $1\cdot\mathbf{x} = \mathbf{x}$ 
    - **Note:** This says that the identity for multiplication of two scalars is also the identity for multiplication of a vector by a scalar

# Basis and Dimension

- *Linear combination* of vectors  $\{\mathbf{u}_i\}$ :  $\sum c_i \mathbf{u}_i$ 
  - $\{c_i\}$ : Scalar coefficients
- $\{\mathbf{u}_i\}$  is *linearly independent* if no  $\mathbf{u}_i$  can be expressed as a linear combination of the others
- *Span* of  $\{\mathbf{u}_i\}$ : Set of all linear combinations
- *Basis* of vector space: (Possibly infinite) set of linearly independent vectors whose span is the entire space
- *Dimension* of vector space: Cardinality of basis
  - **Note:** All bases of a given space have same (possibly infinite) cardinality



# Basis

- Let a vector space have basis  $B = [\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n]$
- Any vector  $\mathbf{u}$  can be written as  $\sum_{i=1 \dots n} u_i \mathbf{u}_i$ 
  - Or compactly,  $(u_1, u_2, \dots, u_n)$
  - These are the **coordinates of  $\mathbf{u}$  in basis  $B$**
- **Note:**
  - We've generalized our original notion of coordinates: they're now relative to the selected basis (axes)
  - A point has different coordinates in different bases

# Back to $\mathbb{R}^n$

- Has a basis  $[\hat{\mathbf{u}}_1, \hat{\mathbf{u}}_2, \dots, \hat{\mathbf{u}}_n]$  along coordinate axes

$$\hat{\mathbf{u}}_1 = (1, 0, \dots, 0)$$

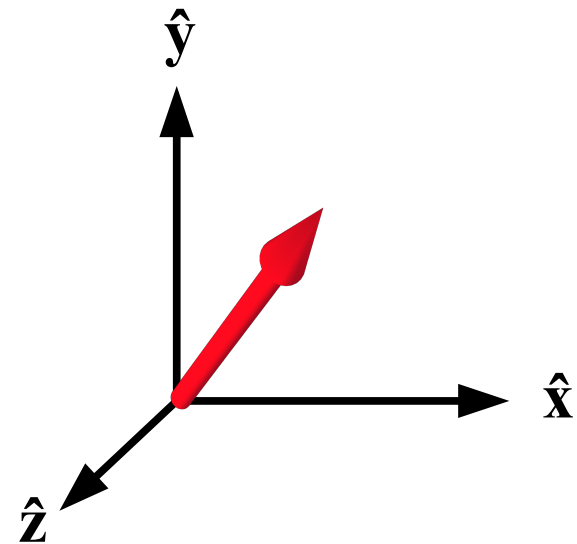
$$\hat{\mathbf{u}}_2 = (0, 1, \dots, 0)$$

$$\vdots$$

$$\hat{\mathbf{u}}_n = (0, 0, \dots, 1)$$

- 2D:  $\hat{\mathbf{x}}, \hat{\mathbf{y}}$

- 3D:  $\hat{\mathbf{x}}, \hat{\mathbf{y}}, \hat{\mathbf{z}}$



- Any  $n$  linearly independent vectors form a basis
  - Dot and cross products have same formula in right-handed orthonormal bases (but **not** in other bases!)

# Change of Basis (example in 3D)

- Let  $[\mathbf{u}, \mathbf{v}, \mathbf{n}]$  be a basis  $B$  for a vector space
- Let coordinates of  $\mathbf{u}, \mathbf{v}, \mathbf{n}$  in another basis  $B_0$  be
$$\mathbf{u} = (u_1, u_2, u_3)$$
$$\mathbf{v} = (v_1, v_2, v_3)$$
$$\mathbf{n} = (n_1, n_2, n_3)$$
- Let coordinates of  $\mathbf{a}$  in  $B$  be  $(a_1, a_2, a_3)$
- Let coordinates of  $\mathbf{a}$  in  $B_0$  be  $(a^0_1, a^0_2, a^0_3)$

# Change of Basis (example in 3D)

$$\begin{bmatrix} a_1^0 \\ a_2^0 \\ a_3^0 \end{bmatrix} = \begin{bmatrix} u_1 & v_1 & n_1 \\ u_2 & v_2 & n_2 \\ u_3 & v_3 & n_3 \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix}$$

# Change of Basis (example in 3D)

- If  $B$  and  $B_0$  are orthonormal, inverse of matrix is its transpose (*orthogonal matrix*)

$$\begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix} = \begin{bmatrix} u_1 & u_2 & u_3 \\ v_1 & v_2 & v_3 \\ n_1 & n_2 & n_3 \end{bmatrix} \begin{bmatrix} a_1^0 \\ a_2^0 \\ a_3^0 \end{bmatrix}$$

(only if bases are orthonormal)

# Metric

- Assign non-negative real number  $d(\mathbf{u}, \mathbf{v})$  to every pair of vectors  $\mathbf{u}, \mathbf{v}$ 
  - Interpret as distance between  $\mathbf{u}$  and  $\mathbf{v}$
- Necessary properties of metric
  - $d(\mathbf{u}, \mathbf{v}) \geq 0$
  - $d(\mathbf{u}, \mathbf{v}) = 0$  if and only if  $\mathbf{u} = \mathbf{v}$
  - $d(\mathbf{u}, \mathbf{v}) = d(\mathbf{v}, \mathbf{u})$
  - $d(\mathbf{u}, \mathbf{v}) \leq d(\mathbf{u}, \mathbf{w}) + d(\mathbf{w}, \mathbf{v})$  (*Triangle Inequality*)

# $\mathbb{R}^n$ as Metric Space

- Magnitude:  $L_2$  norm:  $\|\mathbf{u}\| = (u_1^2 + u_2^2 + \dots + u_n^2)^{1/2}$

- Metric  $d(\mathbf{u}, \mathbf{v})$  defined as  $\|\mathbf{u} - \mathbf{v}\|$

- Other common magnitude functions:  $L_p$  norms

$$\|\mathbf{u}\| = (|u_1|^p + |u_2|^p + \dots + |u_n|^p)^{1/p}$$

- $L_1$  is “Manhattan” norm
- $L_\infty$  is “max” norm

# Inner Product

- Assign scalar  $\langle \mathbf{u}, \mathbf{v} \rangle \in \text{field } F$  to pair of vectors  $\mathbf{u}, \mathbf{v}$ 
  - $F$  is assumed to be set of real or complex numbers
- Necessary properties of inner product
  - $\langle \mathbf{u}, \mathbf{v} \rangle = \overline{\langle \mathbf{v}, \mathbf{u} \rangle}$  (overline denotes complex conjugate)
  - $\langle s\mathbf{u}, \mathbf{v} \rangle = s \langle \mathbf{u}, \mathbf{v} \rangle$
  - $\langle \mathbf{u} + \mathbf{v}, \mathbf{w} \rangle = \langle \mathbf{u}, \mathbf{w} \rangle + \langle \mathbf{v}, \mathbf{w} \rangle$
- The dot product is a valid inner product for  $\mathbb{R}^n$

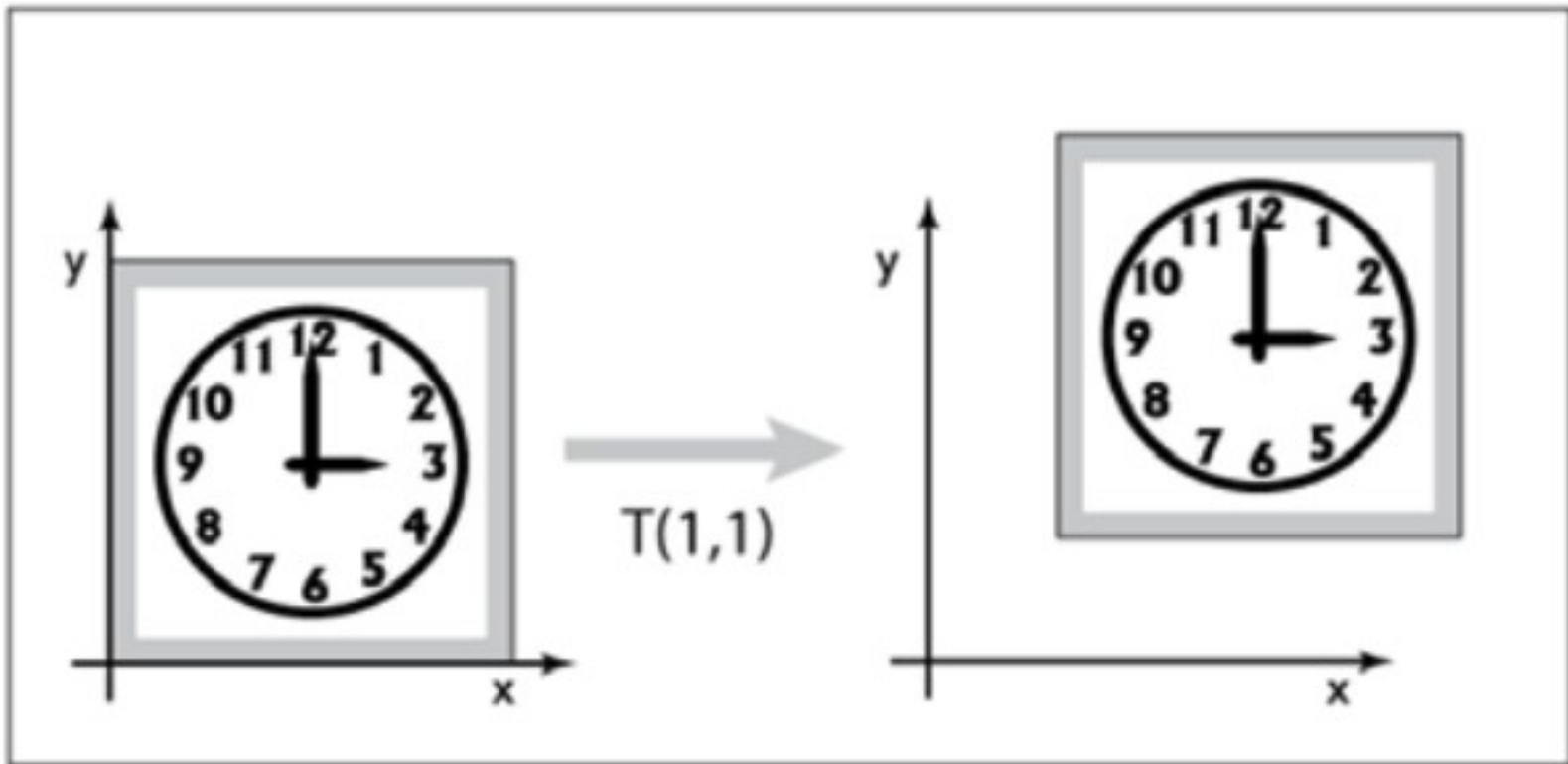


# Transformations of Vectors

(Thanks to Pat Hanrahan for this section)

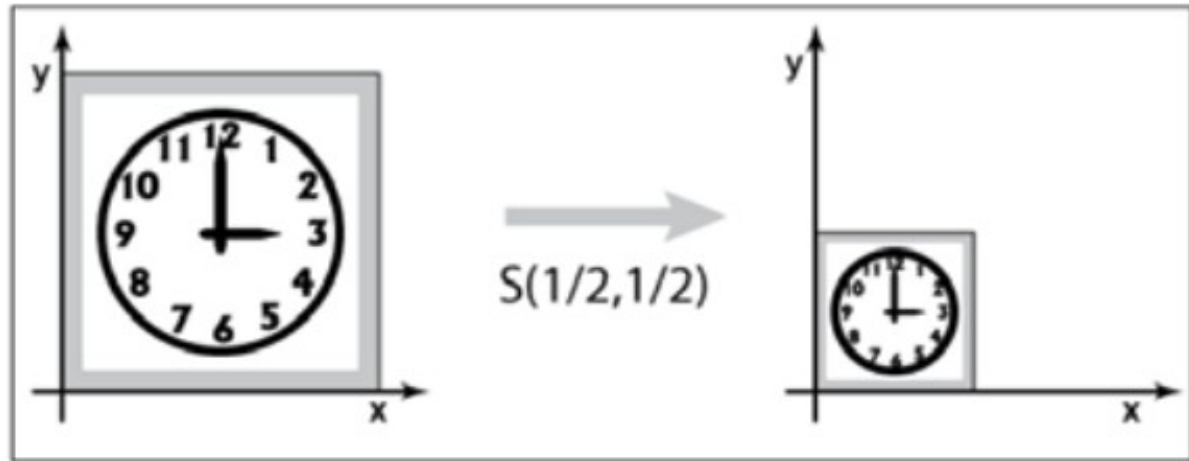
- $\mathbf{u}' = T(\mathbf{u})$
- Why?
  - Modeling:
    - Create objects in convenient coordinates
    - Multiple instances of prototype shape
    - Kinematics of linked structures
  - Viewing:
    - Map between window and device coordinates
    - Virtual camera projections: parallel/perspective
- We'll stick to  $\mathbb{R}^2$  for now

# Translation

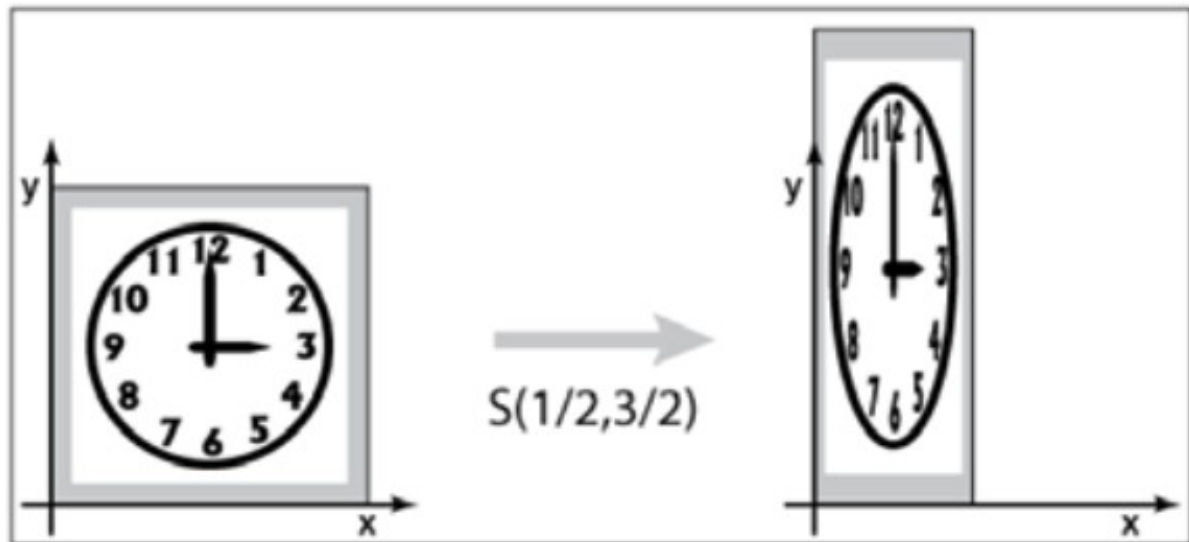


# Scaling

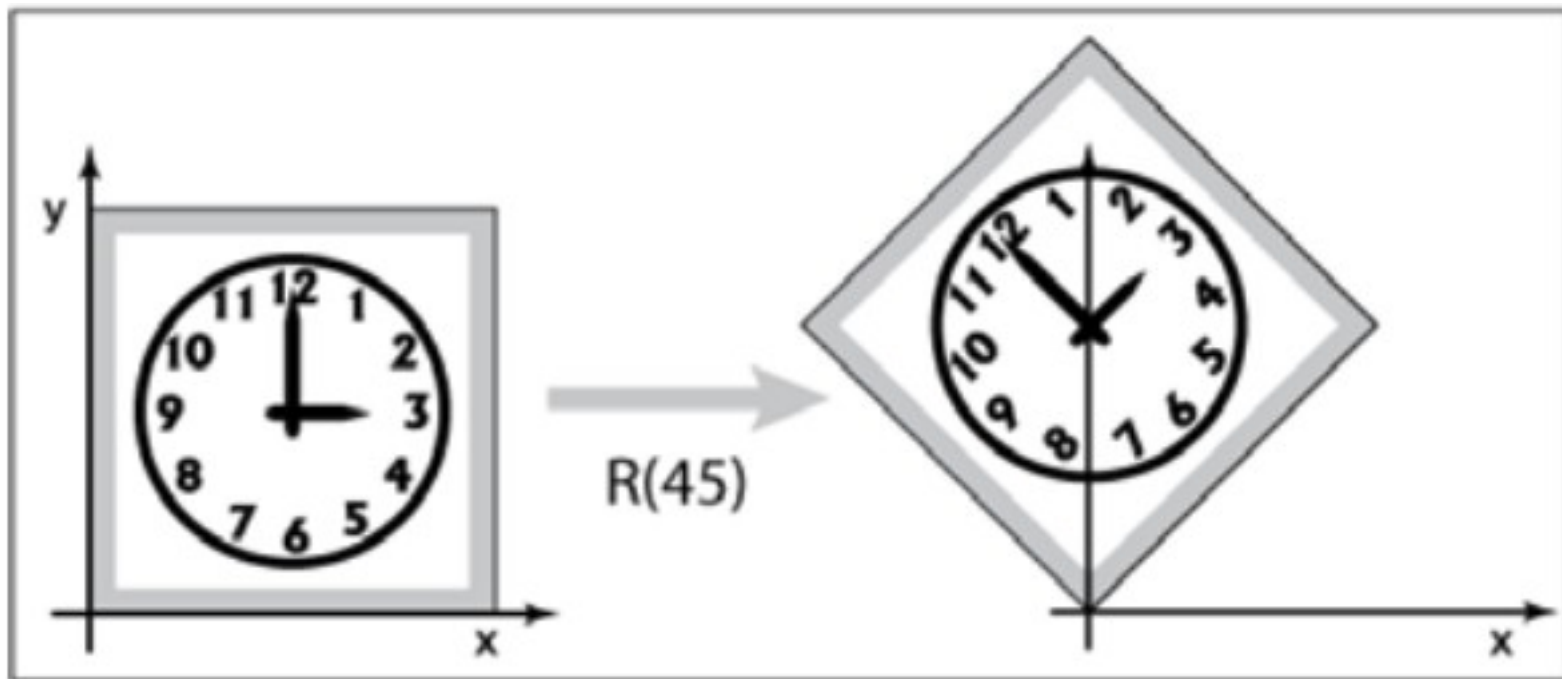
Uniform



Non-uniform

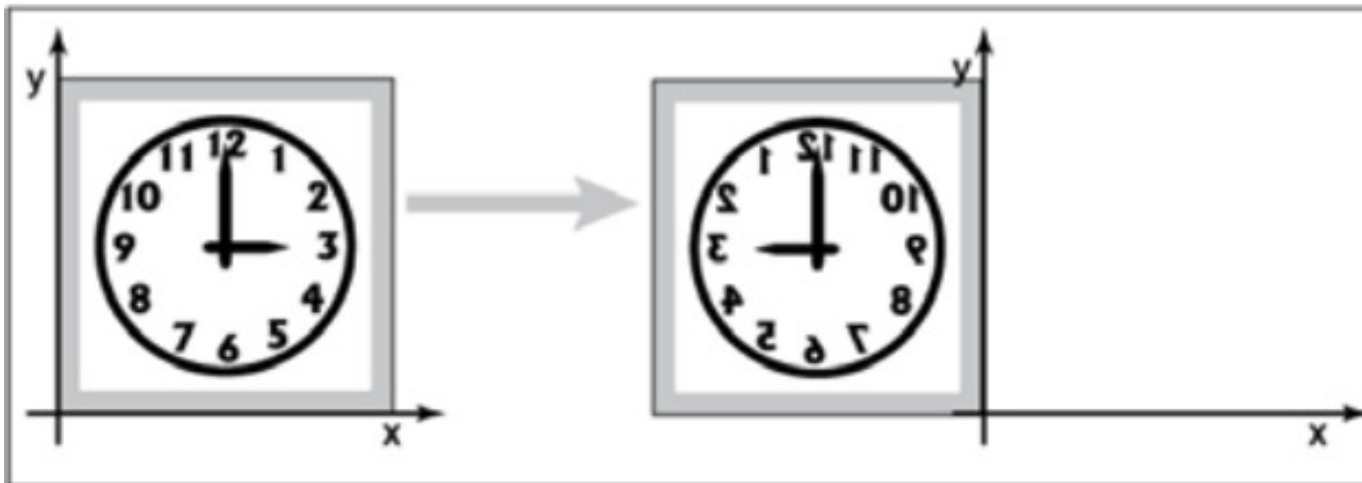
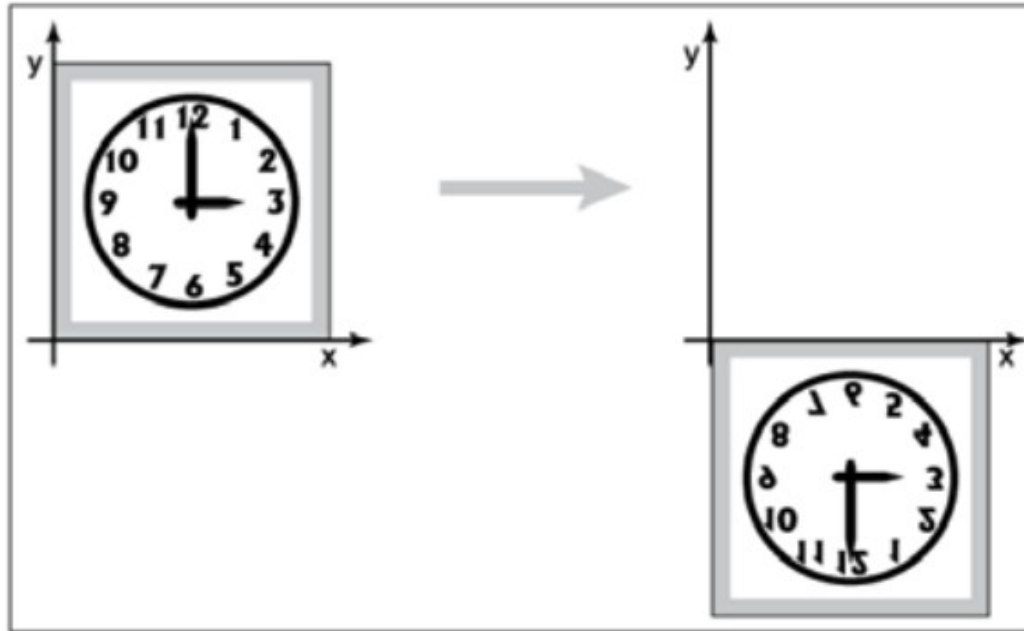


# Rotation

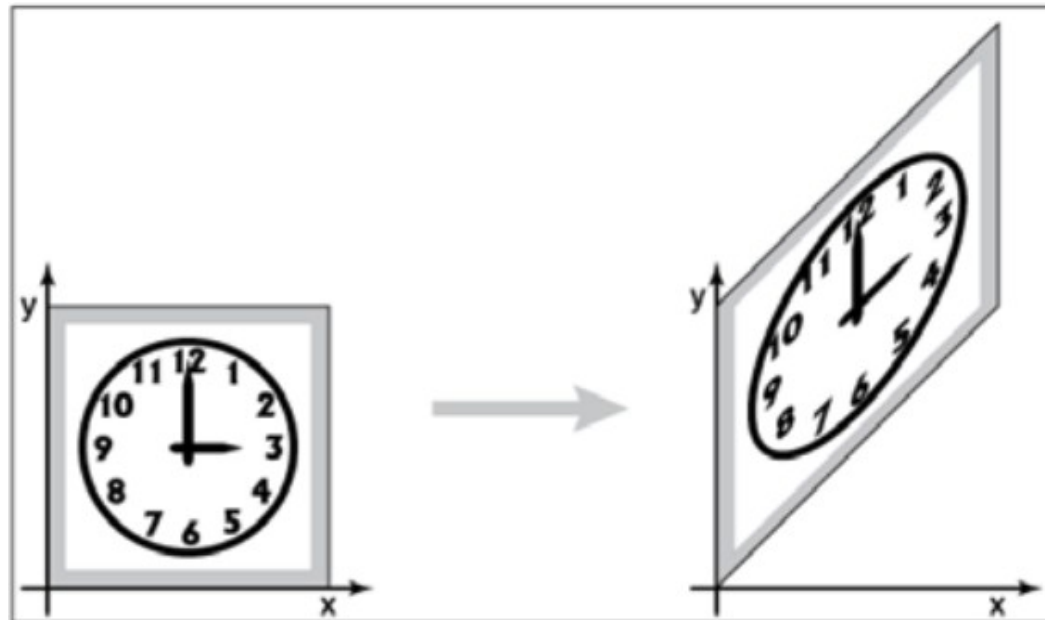
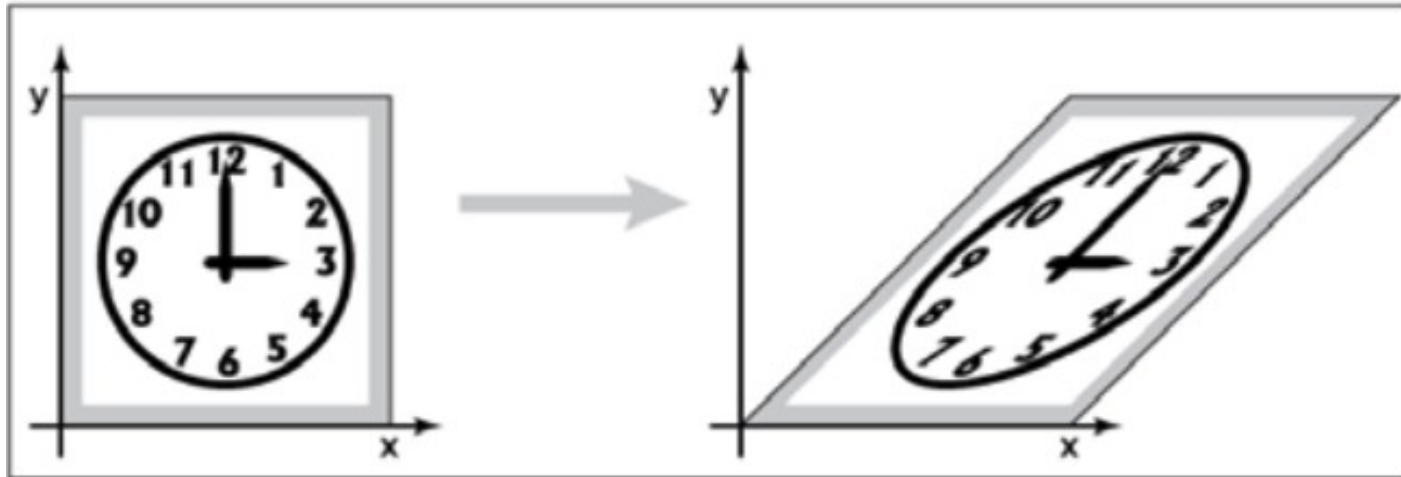


(counter-clockwise about origin)

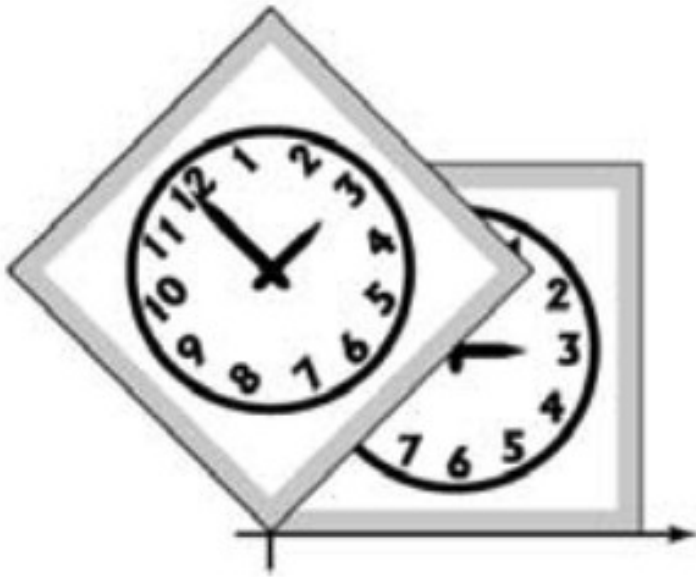
# Reflection



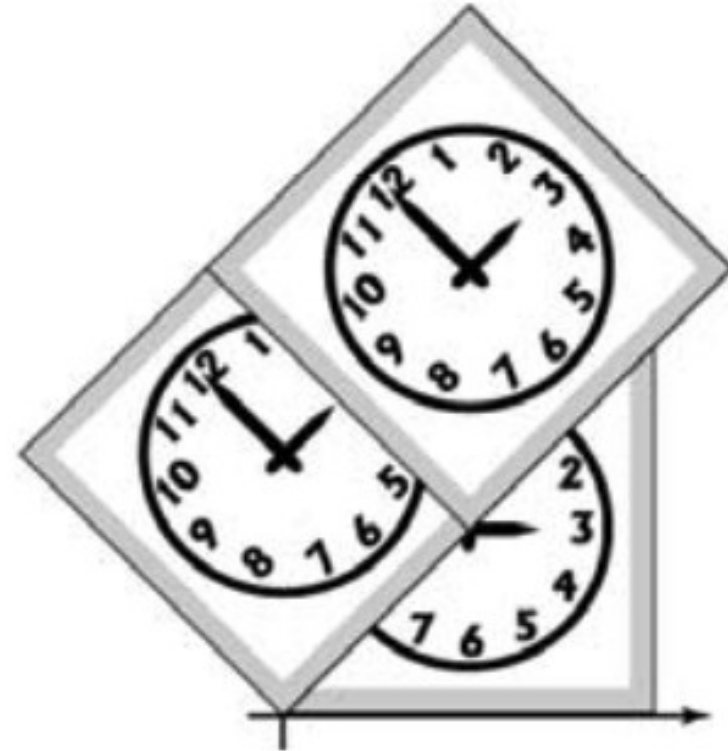
# Shear



# Composing Transformations



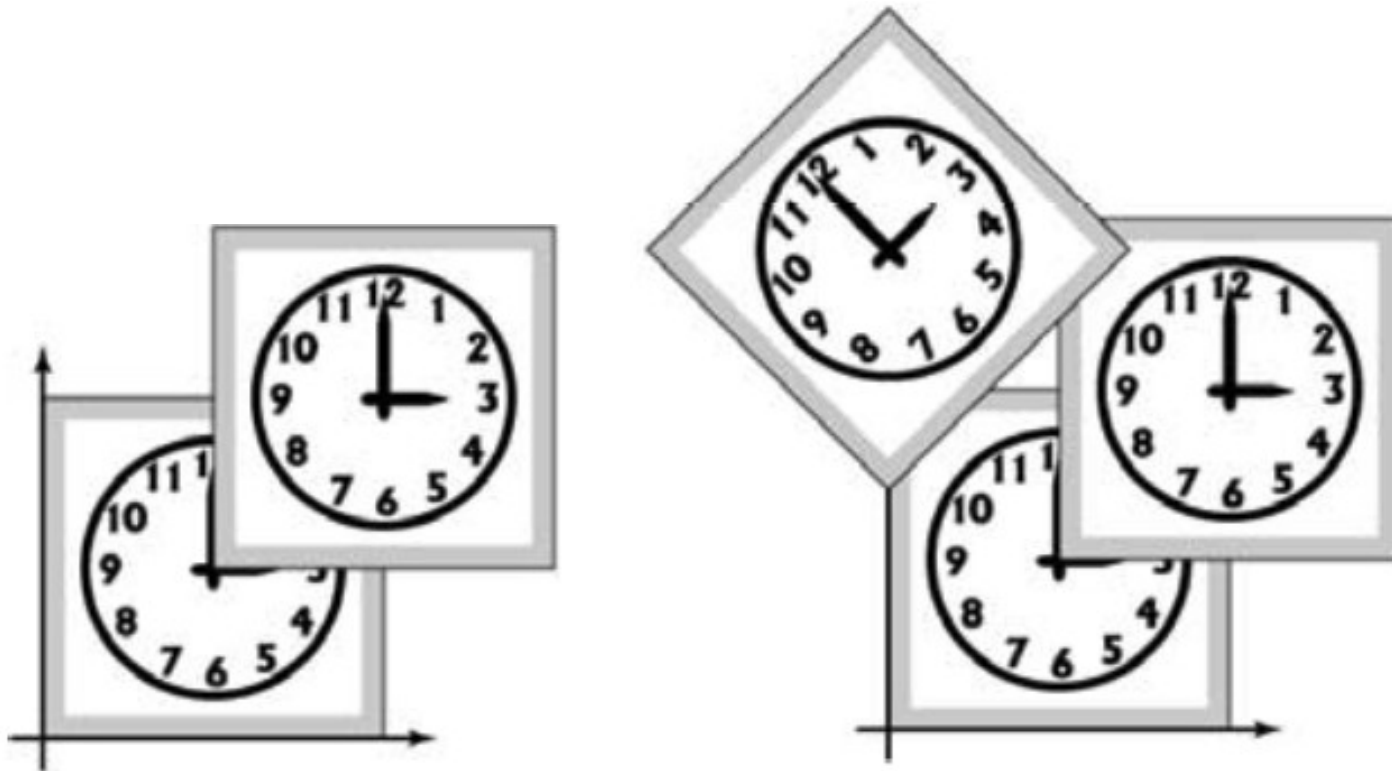
$R(45)$



$T(1, 1) R(45)$

Rotate, then Translate

# Composing Transformations



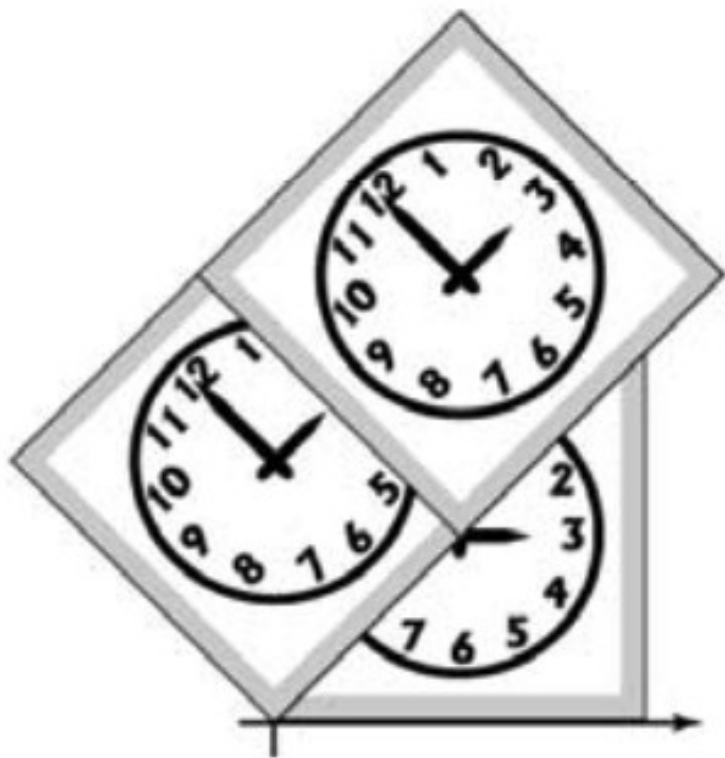
$T(1, 1)$

$R(45) T(1, 1)$

Translate, then Rotate

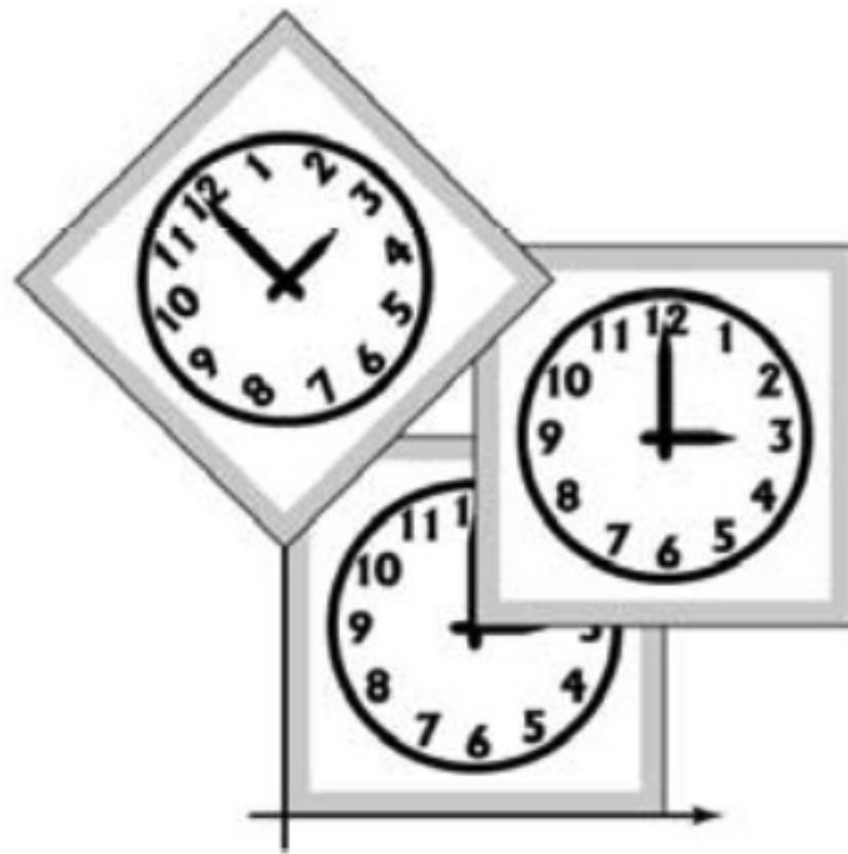


# Order Matters!



$T(1, 1) R(45)$

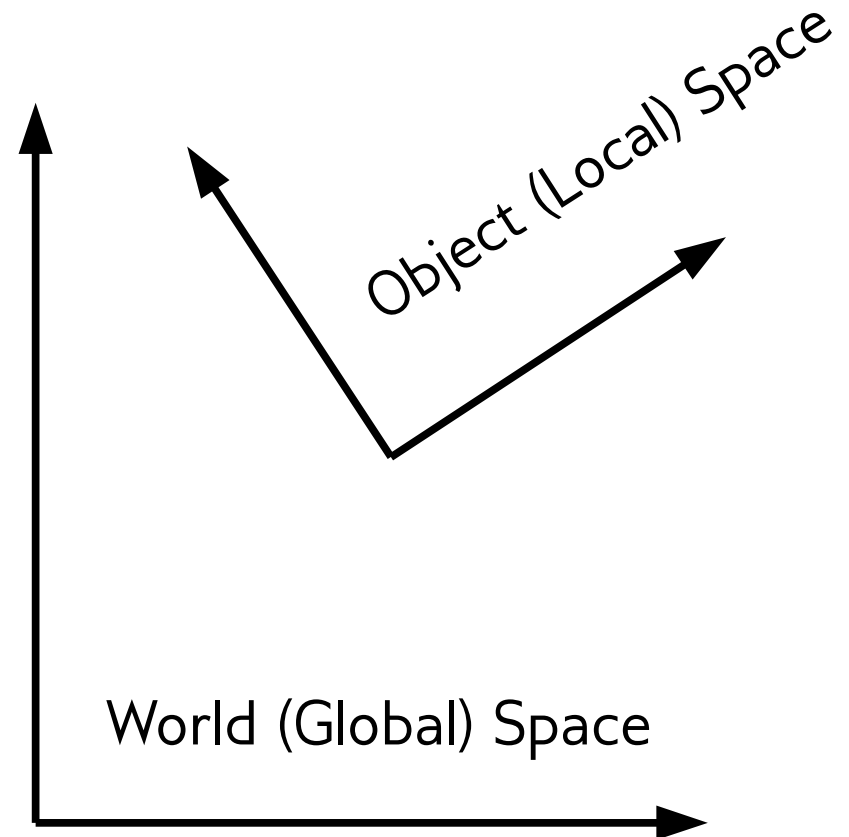
$\neq$



$R(45) T(1, 1)$

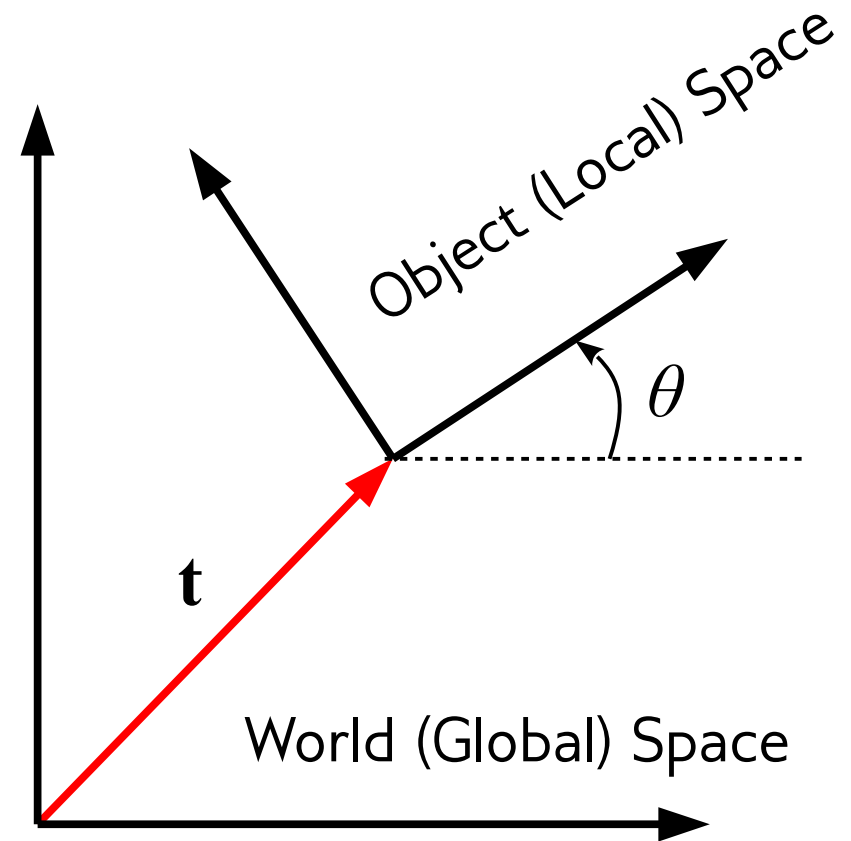
# World Space and Object Space

- Transformation maps from one to the other
- Construct by composing sequence of basic transforms
  - **Remember:** Transforms apply *Right-to-Left* in our notation!



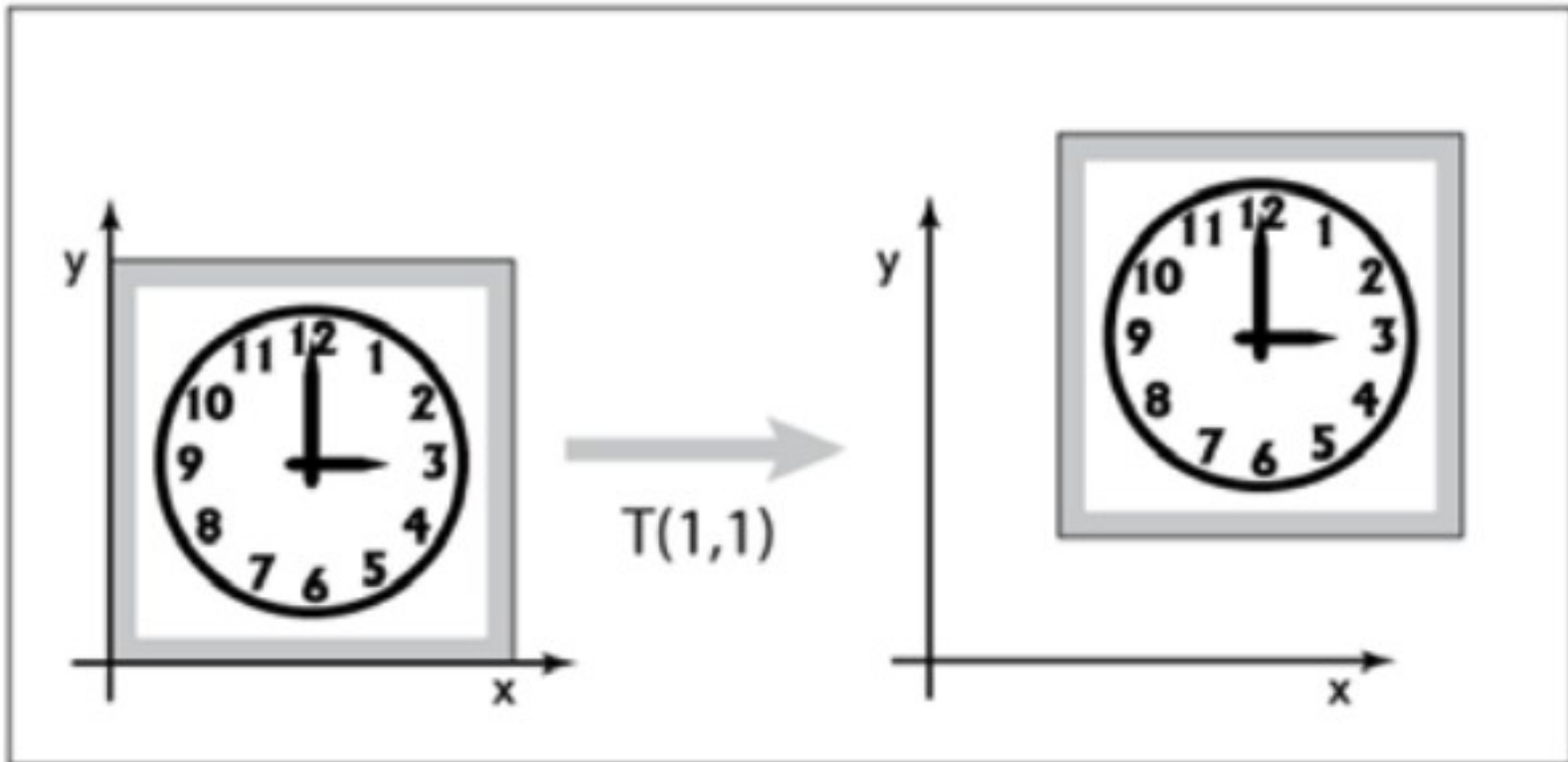
# World Space and Object Space

- Let's look at the example on the right
- Object  $\rightarrow$  World:
  - Rotate by  $\theta$  (ccw), then translate by  $\mathbf{t}$
- World  $\rightarrow$  Object:
  - Translate by  $-\mathbf{t}$ , then rotate by  $-\theta$



Make sure you understand this!

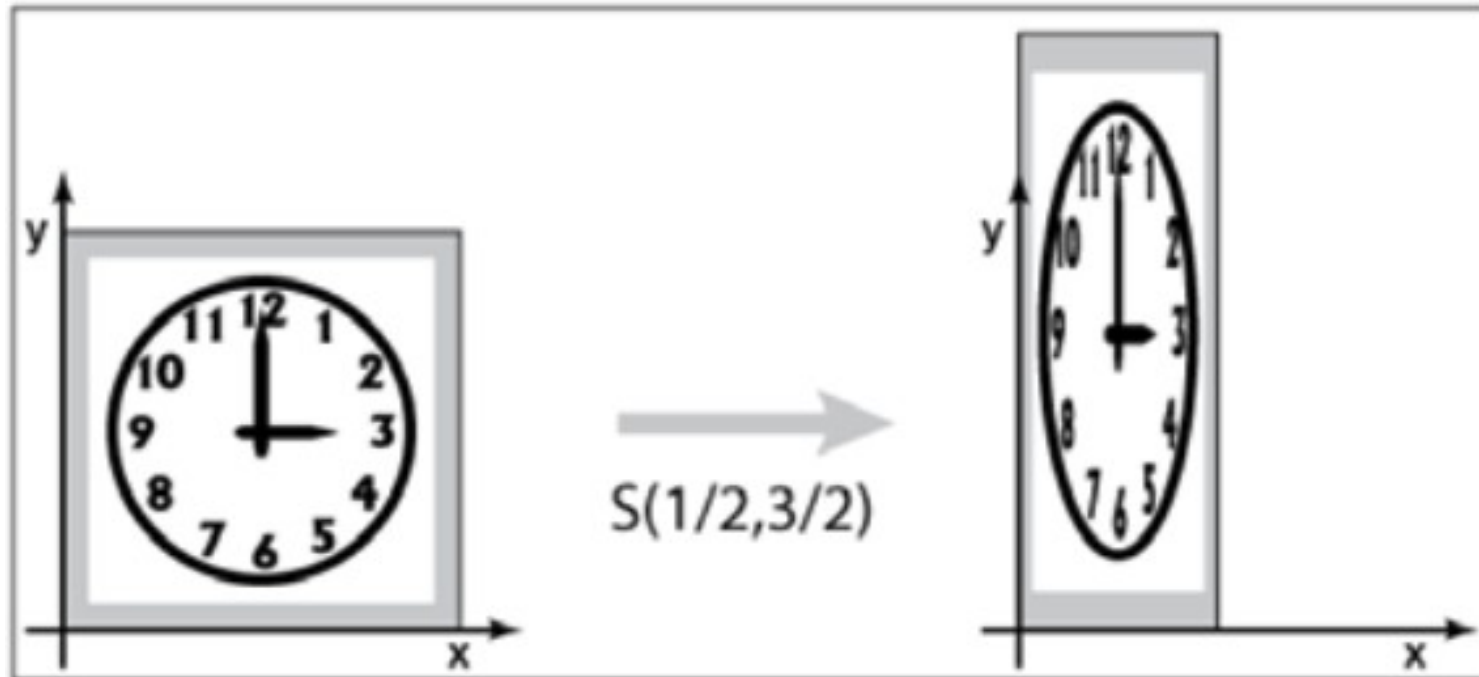
# Translation



$$x' = x + t_x$$

$$y' = y + t_y$$

# Scaling (and Reflection)

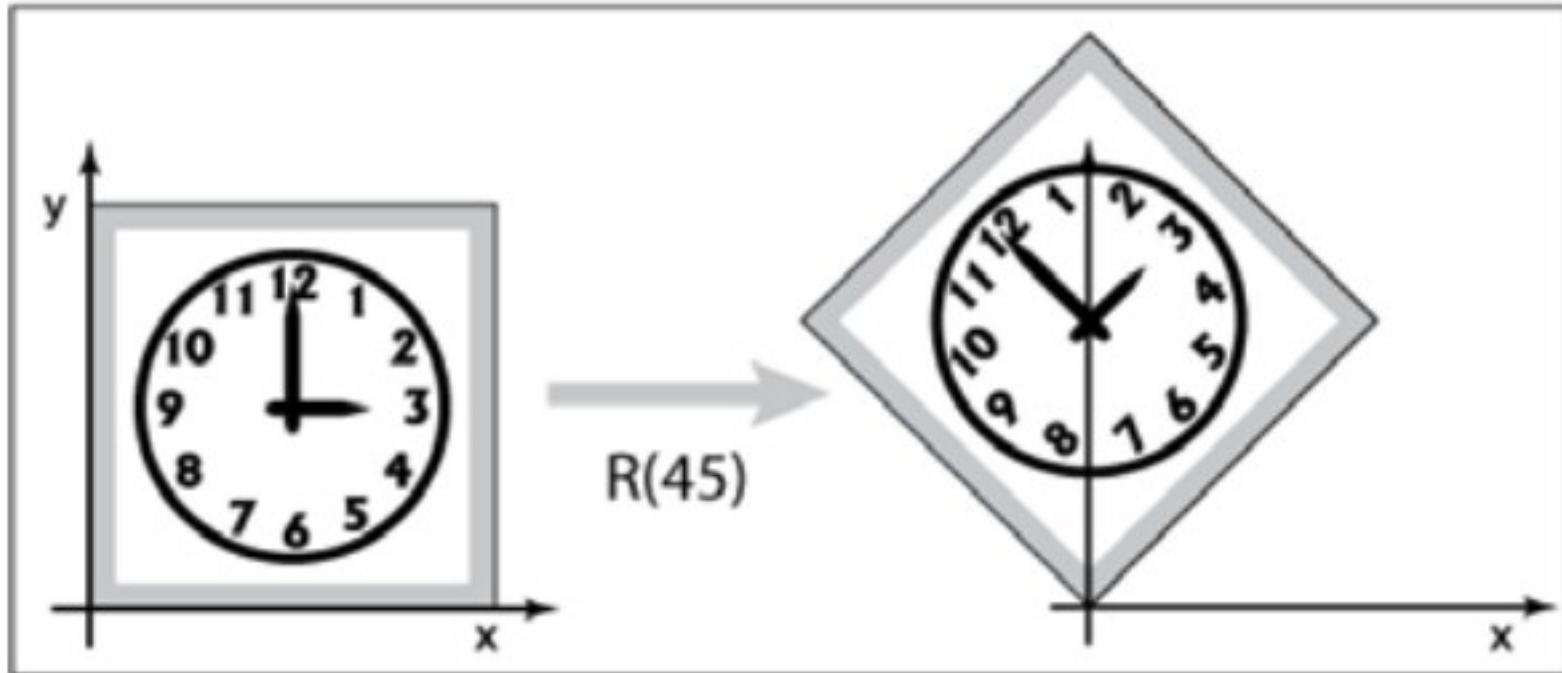


$$x' = s_x x$$

$$y' = s_y y$$

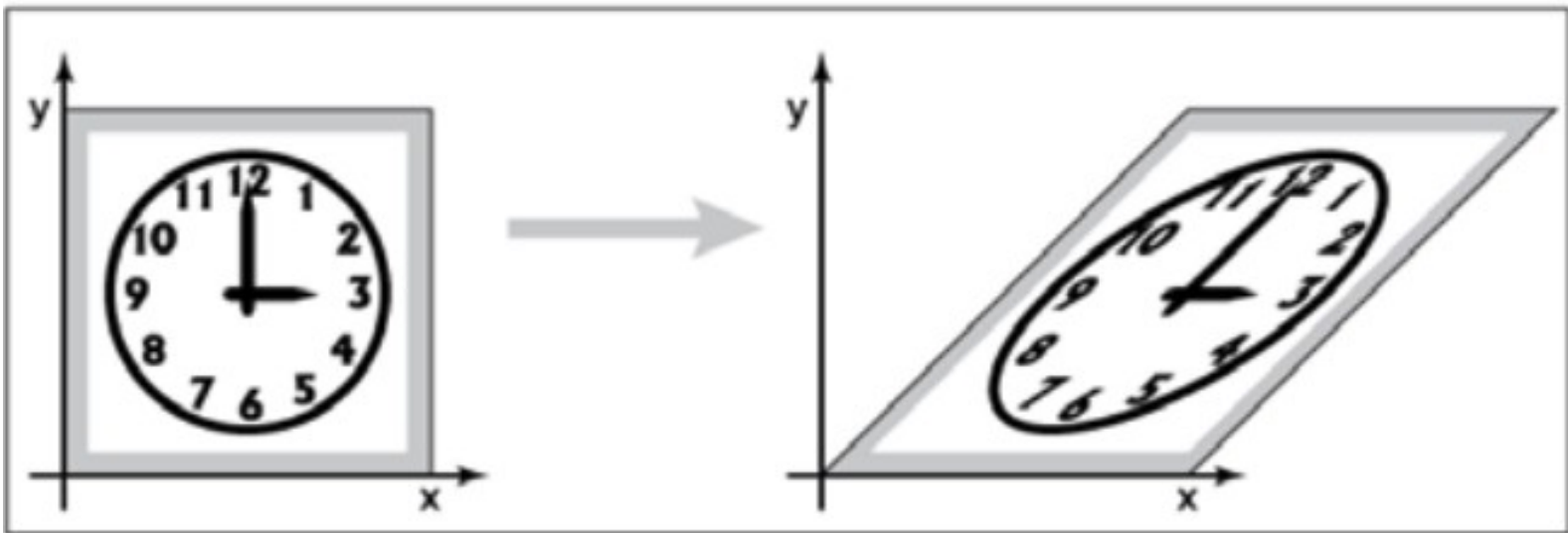
(Negative scaling coefficients give reflection)

# CCW Rotation By $\theta$ About Origin



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

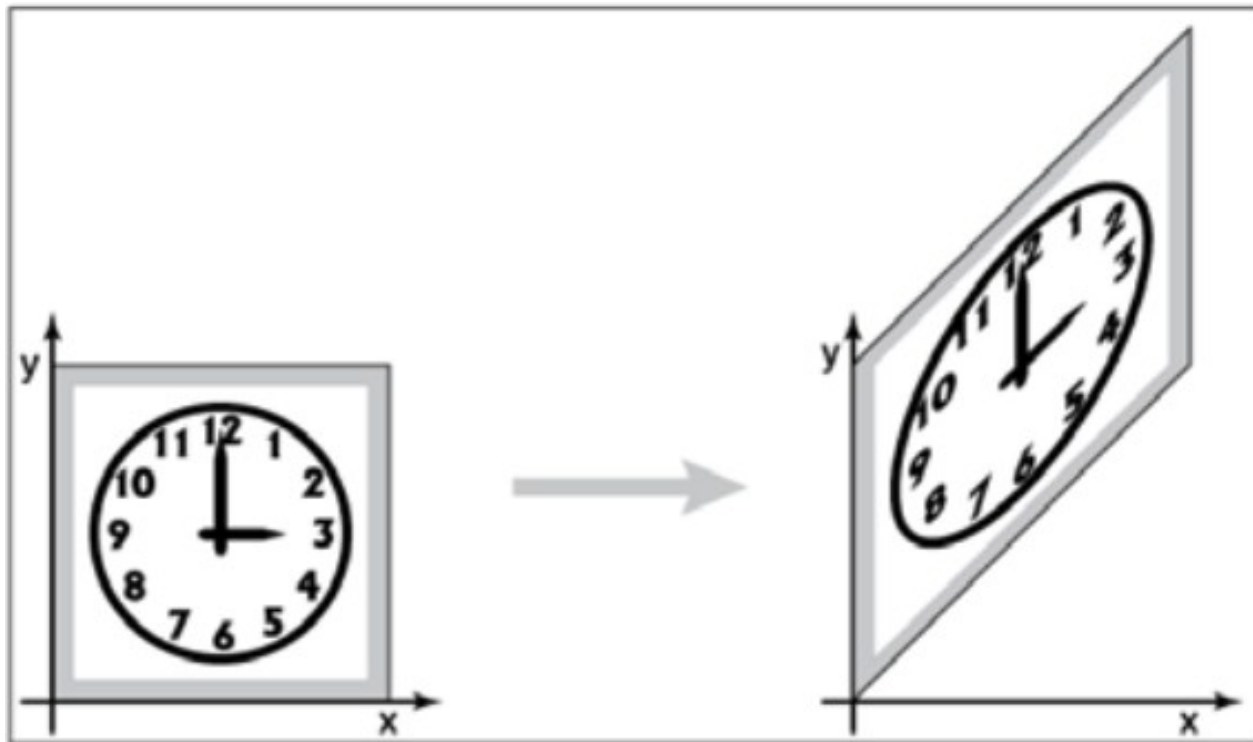
# Horizontal Shear



$$x' = x + sy$$

$$y' = y$$

# Vertical Shear



$$x' = x$$

$$y' = sx + y$$



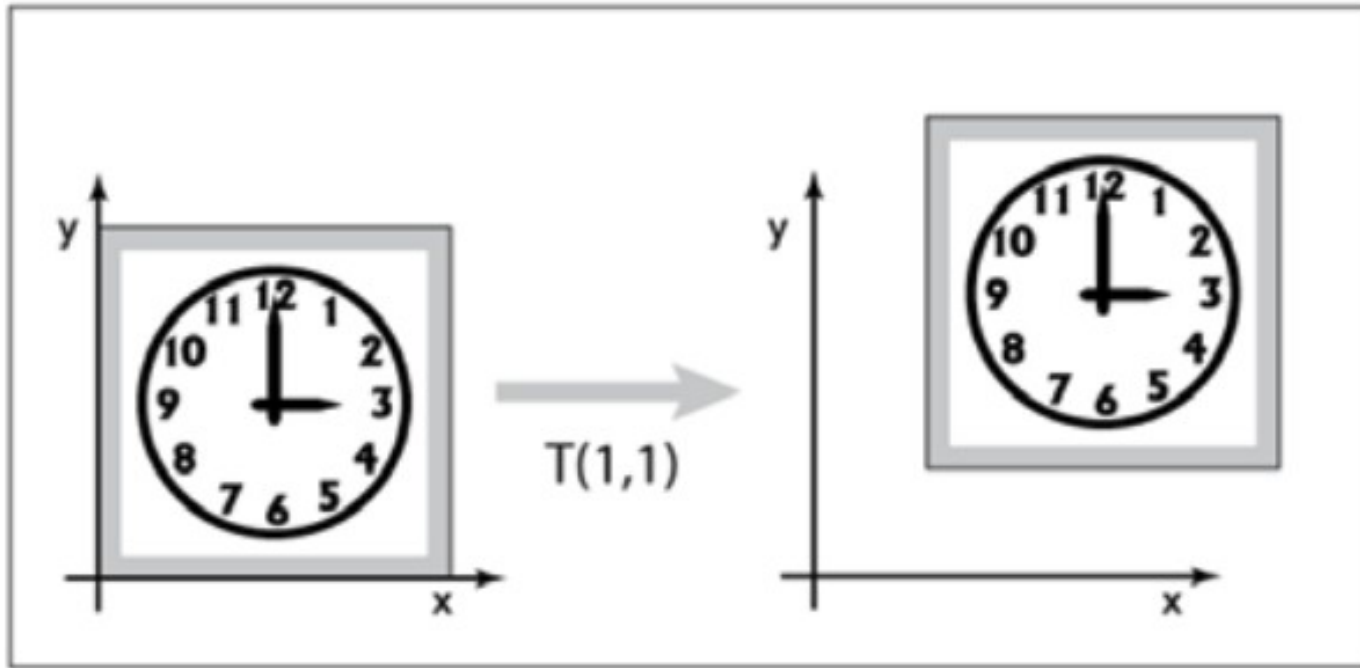
# Types of 2D Transformations

- *Linear Transforms*:  $T(\mathbf{u} + \mathbf{v}) = T(\mathbf{u}) + T(\mathbf{v})$ 
  - Scaling
  - Rotation
  - Shear
  - Reflection
- *Affine Transforms*:  $T(\mathbf{u}) = L(\mathbf{u}) + \mathbf{a}$ , where  $L$  is linear and  $\mathbf{a}$  is a fixed vector
  - Translation
- Other, e.g. perspective projection
- How do we represent these in a common format?

# Homogenous Coordinates (2D)

- Point  $(x, y) \rightarrow (x, y, 1)$
- Direction  $(x, y) \rightarrow (x, y, 0)$
- For any scalar  $c$ ,  $(cx, cy, ch) \equiv (x, y, h)$
- To convert back:
  - If  $h$  is 0:  $(x, y, 0) \rightarrow (x, y)$
  - If  $h$  is non-zero:  $(x, y, h) \rightarrow (x / h, y / h)$
- **Note:**
  - **Not** 3D vector space, just a new representation for 2D
  - Legal/illegal operations for directions & positions automatically distinguished!

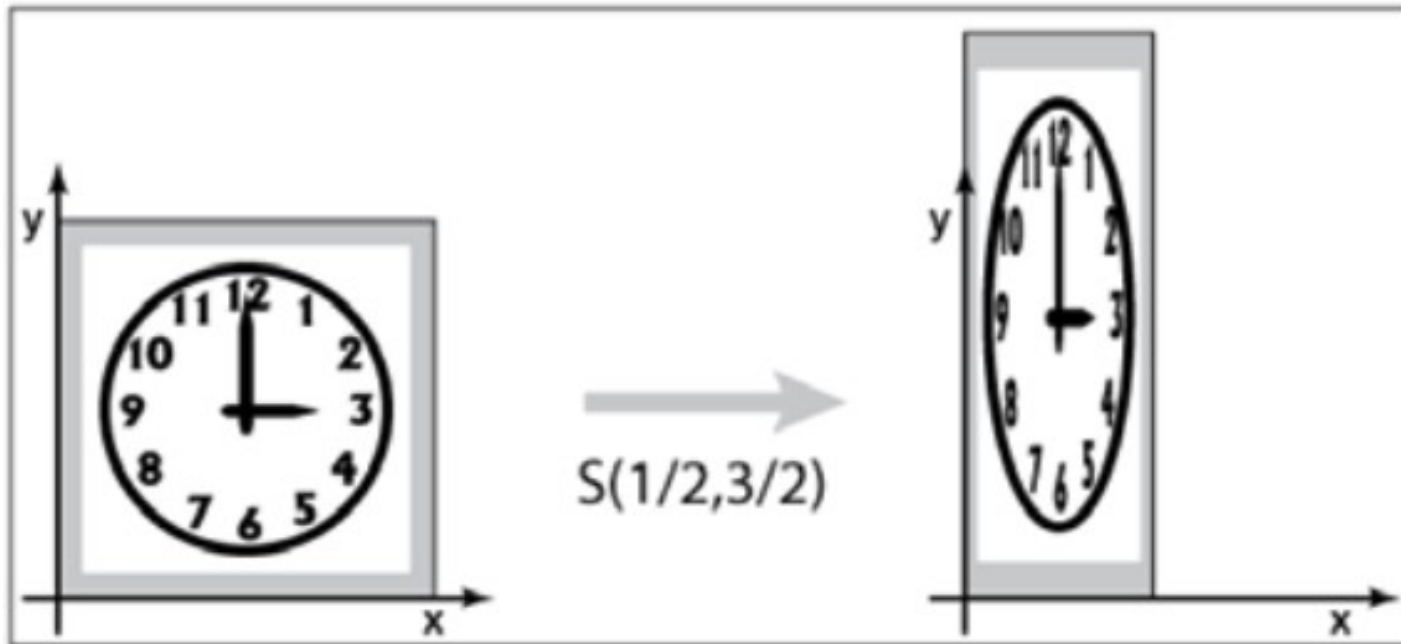
# Translation



$$\begin{aligned}x' &= x + t_x \\y' &= y + t_y\end{aligned}$$

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

# Scaling (and Reflection)

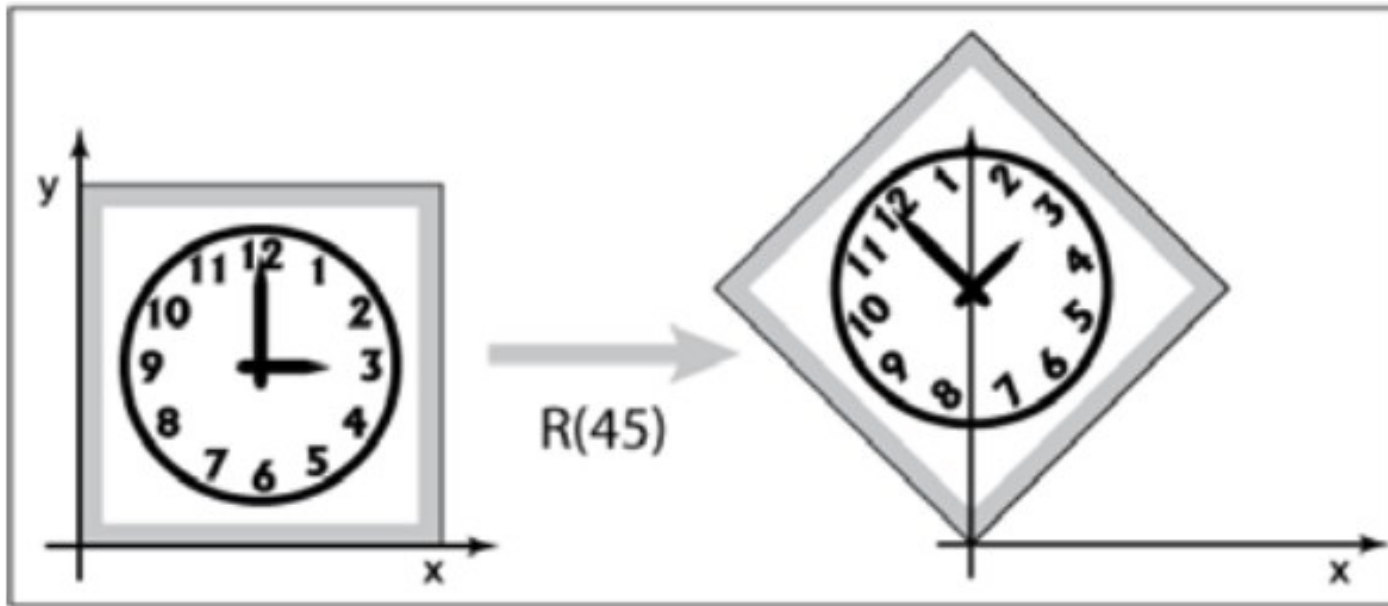


$$x' = s_x x$$

$$y' = s_y y$$

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

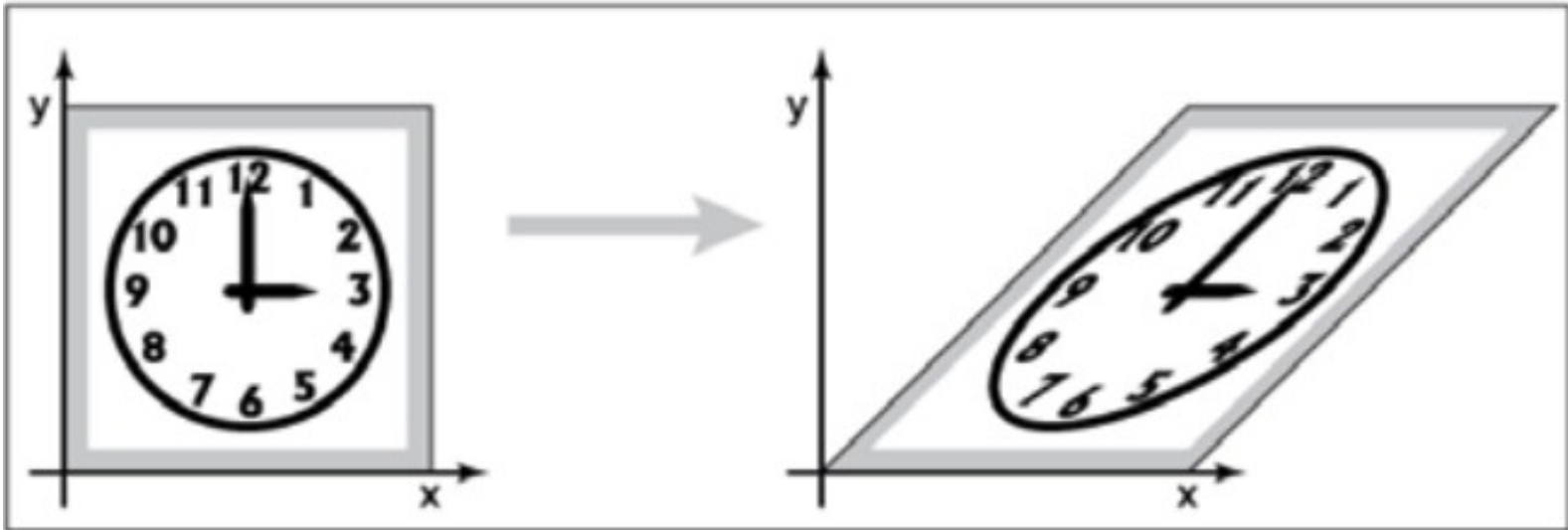
# CCW Rotation By $\theta$ About Origin



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

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

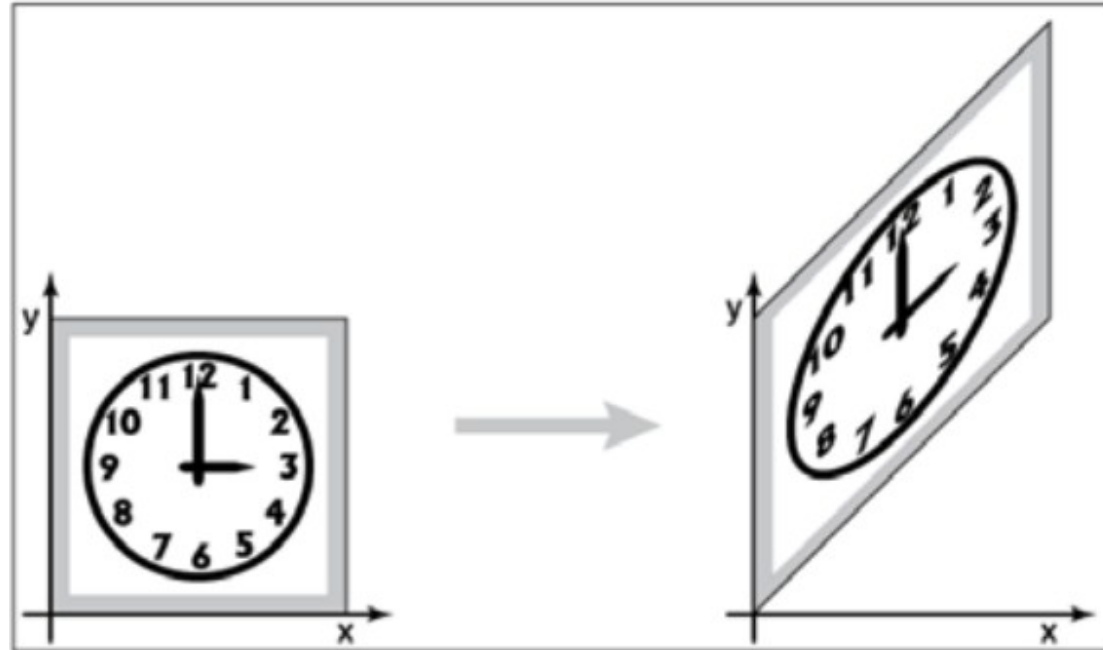
# Horizontal Shear



$$\begin{aligned}x' &= x + sy \\y' &= y\end{aligned}$$

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

# Vertical Shear



$$\begin{aligned}x' &= x \\y' &= sx + y\end{aligned}$$

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

# What about 3D?

- Very similar:  $(x, y, z) \rightarrow (x, y, z, h)$
- Look up the formulæ!
- Rotation is a mess
  - Common method:
    - Map axis of rotation to a coordinate axis (similar to change of basis)
    - Rotate around the coordinate axis
    - Map back
  - Other approaches based on Euler angles and quaternions



# Why Use Matrices?

- Compute the matrix once

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

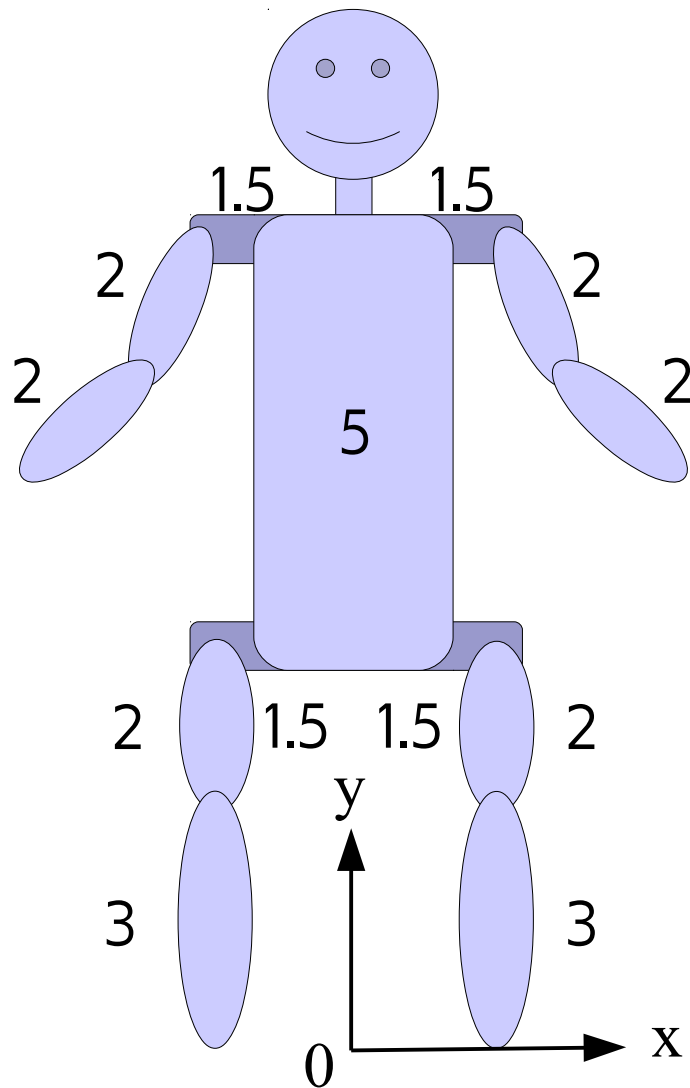
$$y' = x \sin \theta + y \cos \theta$$

- Don't repeatedly evaluate sines and cosines
- Combine sequence of transforms into a single transform
  - Store  $M = ABCD$ , apply  $M(\mathbf{u})$  instead of  $A(B(C(D(\mathbf{u}))))$
- The inverse of a sequence of transforms is just the matrix inverse
  - $(ABCD)^{-1} = D^{-1}C^{-1}B^{-1}A^{-1} = M^{-1}$

# Hierarchical Modeling

- Graphics systems maintain a *current transformation matrix* (CTM)
  - All geometry is transformed by the CTM
  - CTM defines object space in which geometry is specified
  - Transformation commands are concatenated onto the CTM. The last one added is applied first:
    - $CTM = CTM * T$
- Graphics systems also maintain a *transformation stack*
  - The CTM can be pushed onto the stack
  - The CTM can be restored from the stack

# Example: Articulated Robot



body

torso

head

shoulder

leftArm

upperArm

lowerArm

hand

rightArm

upperArm

lowerArm

hand

hips

leftLeg

upperLeg

lowerLeg

foot

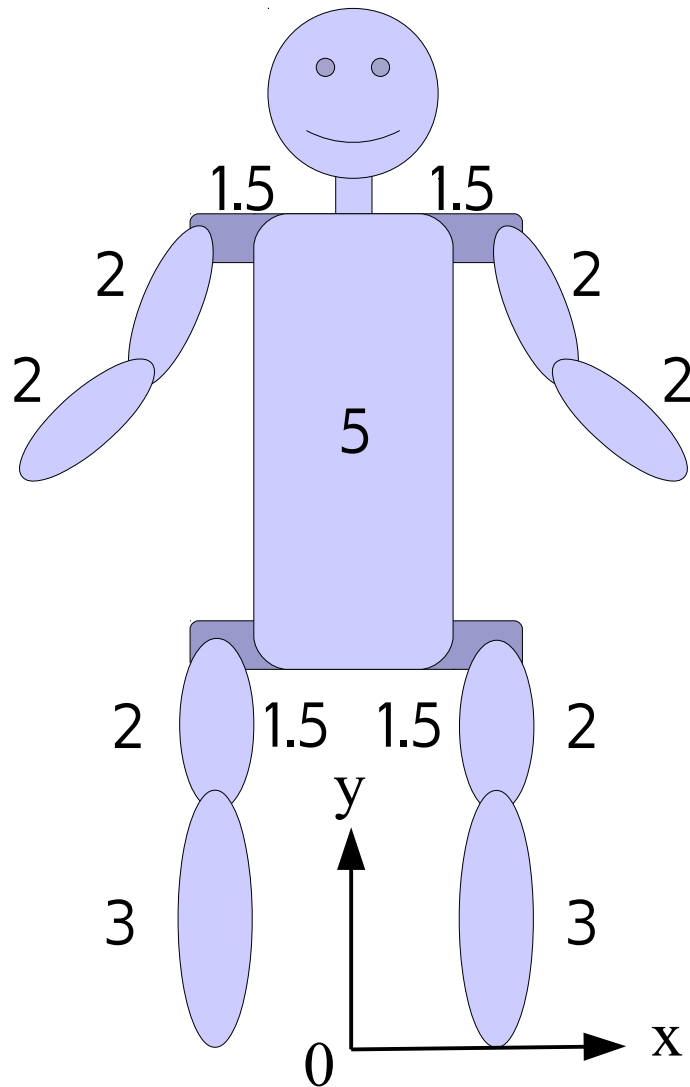
rightLeg

upperLeg

lowerLeg

foot

# Example: Articulated Robot



```
translate(0, 5, 0);
torso();
pushMatrix();
translate(0, 5, 0);
  shoulder();
  pushMatrix();
    rotateY(neck_y);
    rotateX(neck_x);
    head();
  popMatrix();
pushMatrix();
  translate(1.5, 0, 0);
  rotateX(l_shoulder_x);
  upperArm();
  pushMatrix();
    translate(0,-2,0);
    rotateX(l_elbow_x);
    lowerArm();
    ...
  popMatrix();
popMatrix();
...
```