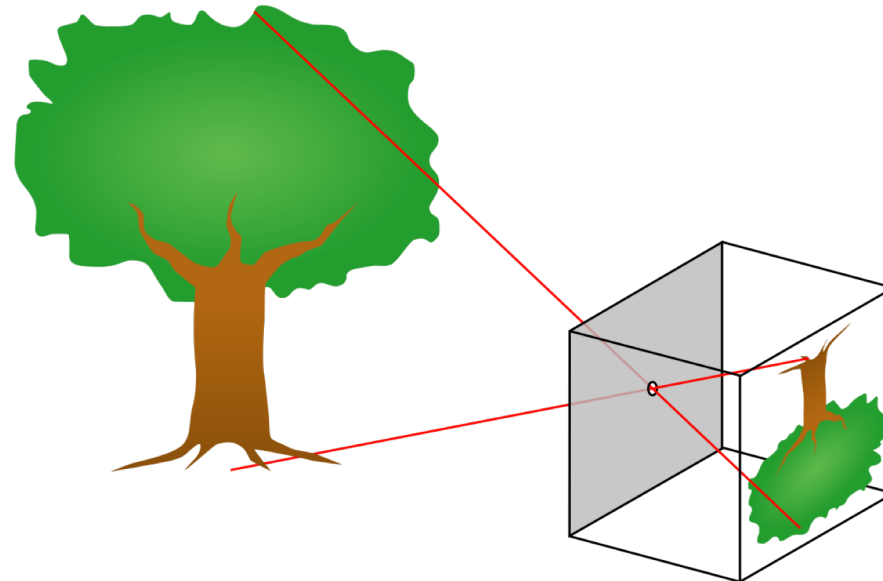


# Linear Algebra

Liangliang Nan

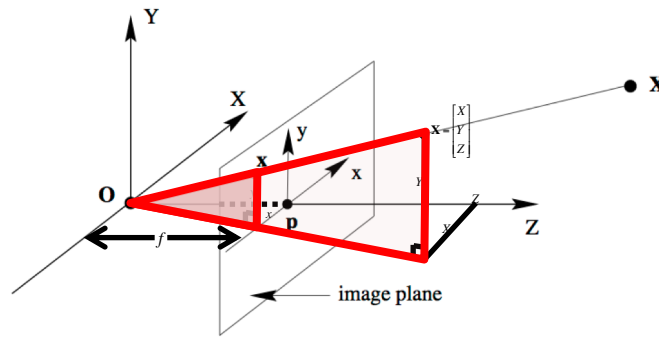
# What is linear algebra?



# What is linear algebra?



How to get the pixel location on an image of a 3D point?



$$\mathbf{x} = \mathbf{K} \begin{bmatrix} \mathbf{R} & \mathbf{t} \end{bmatrix} \mathbf{X}$$

# What is linear algebra?

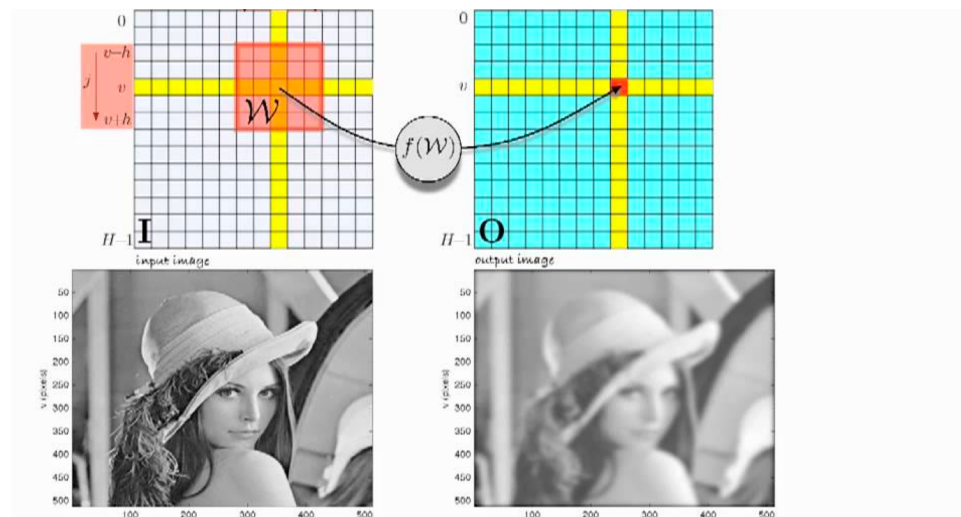
- Difficult problem can be handled effectively
  - Representation
    - 3D points, lines, planes in the 3D space
      - Distance?

# What is linear algebra?

- Difficult problem can be handled effectively
  - Representation
  - Coordinates will be used
    - Perform geometrical transformations/computation
      - The rotation of a 3D object
      - The intersection of 3 non-parallel planes
    - Associate 3D with 2D points
      - The projection of a 3D object onto a plane

# What is linear algebra?

- Difficult problem can be handled effectively
  - Representation
  - Coordinates will be used
  - Images are matrices of numbers
    - Find properties of these number



# What is linear algebra?

- Difficult problem can be handled effectively
  - Representation
  - Coordinates will be used
  - Images are matrices of numbers
- Linear algebra
  - Organize information in cases where certain **mathematical structures** are present
  - The study of those structures
    - Vectors and linear functions

# Scope of the lectures

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**Part 1:** Vectors, matrices, matrix/vector arithmetic, geometric transformations

**Part 2:** Linear systems, linear-least squares

**Part 3:** Eigen values/vectors, singular value decomposition

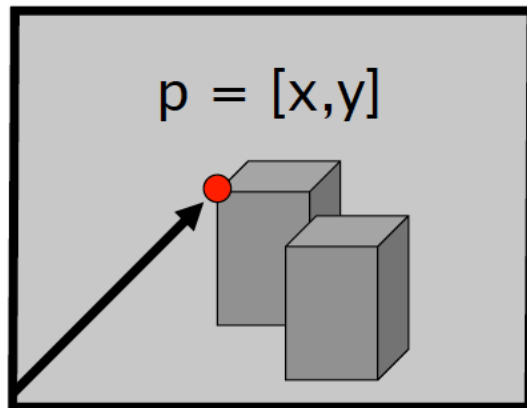


# Linear Algebra - Part 1

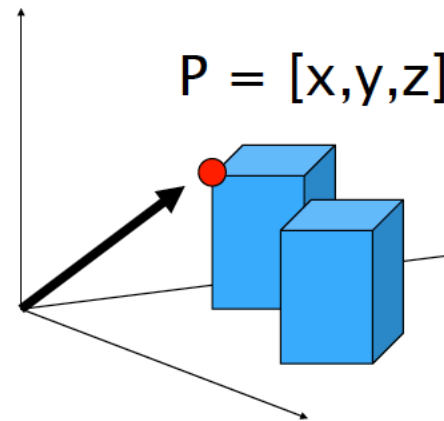
## **Vectors, matrices, matrix/vector arithmetic, geometric transformations**

Liangliang Nan

# Vectors (i.e., 2D and 3D vectors)



Image



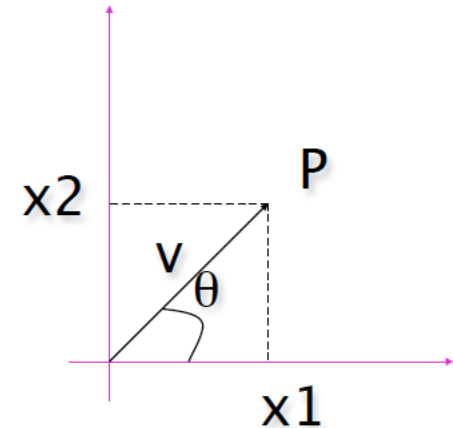
3D world

# Vectors (i.e., 2D vectors)

$$\mathbf{v} = (x_1, x_2)$$

Magnitude:  $\|\mathbf{v}\| = \sqrt{x_1^2 + x_2^2}$

If  $\|\mathbf{v}\| = 1$ ,  $\mathbf{v}$  is a UNIT vector

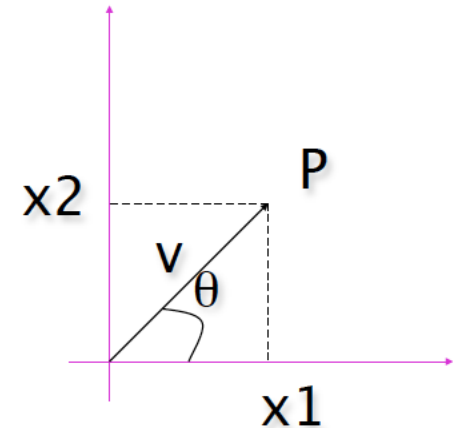


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If  $\|\mathbf{v}\| = 1$ ,  $\mathbf{v}$  is a UNIT vector



What is the unit vector that has the same direction as  $\mathbf{v}$ ?



# Vectors (i.e., 2D vectors)

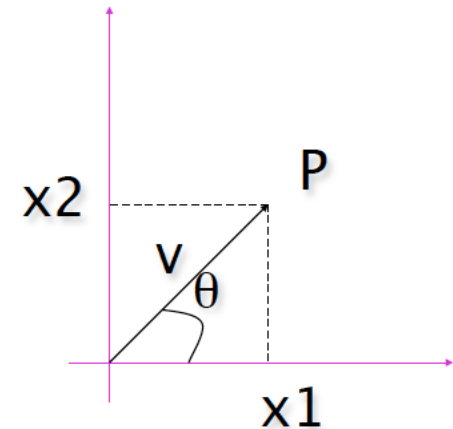
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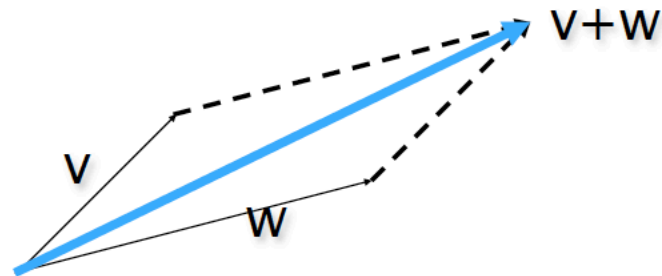
$$\frac{\mathbf{v}}{\|\mathbf{v}\|} = \left( \frac{x_1}{\|\mathbf{v}\|}, \frac{x_2}{\|\mathbf{v}\|} \right) \text{ is a unit vector}$$

Orientation:  $\theta = \tan^{-1}\left(\frac{x_2}{x_1}\right)$



# Vector Addition

$$\mathbf{v} + \mathbf{w} = (x_1, x_2) + (y_1, y_2) = (x_1 + y_1, x_2 + y_2)$$



Compute

$$(0, 1) + (1, 0) = ?$$

$$(1, 1) + (1, 1) = ?$$

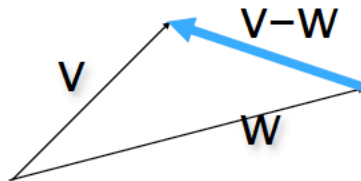
$$(1, 0) + (-1, 0) = ?$$

A parallelogram

# Vector Subtraction

- Inverse operation of addition

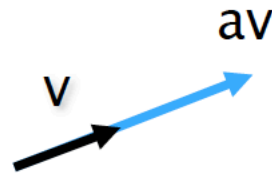
$$\mathbf{v} - \mathbf{w} = (x_1, x_2) - (y_1, y_2) = (x_1 - y_1, x_2 - y_2)$$



# Scalar Product

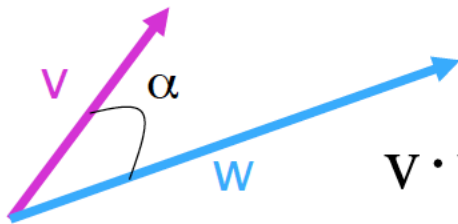
- Length is changed
- Direction is preserved

$$a\mathbf{v} = a(x_1, x_2) = (ax_1, ax_2)$$





# Inner (dot) Product

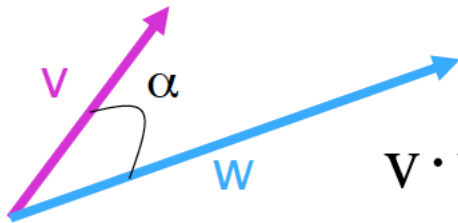


$$\mathbf{v} \cdot \mathbf{w} = (x_1, x_2) \cdot (y_1, y_2) = x_1 y_1 + x_2 y_2$$

The inner product is a **SCALAR!**

$$\mathbf{v} \cdot \mathbf{w} = (x_1, x_2) \cdot (y_1, y_2) = \|\mathbf{v}\| \cdot \|\mathbf{w}\| \cos \alpha$$


# Inner (dot) Product



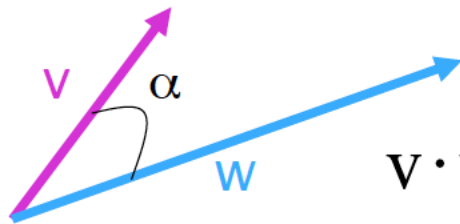
$$\mathbf{v} \cdot \mathbf{w} = (x_1, x_2) \cdot (y_1, y_2) = x_1 y_1 + x_2 y_2$$

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$$\mathbf{v} \cdot \mathbf{w} = (x_1, x_2) \cdot (y_1, y_2) = \|\mathbf{v}\| \cdot \|\mathbf{w}\| \cos \alpha$$

if  $\mathbf{v} \perp \mathbf{w}$ ,  $\mathbf{v} \cdot \mathbf{w} =$  

# Inner (dot) Product



$$\mathbf{v} \cdot \mathbf{w} = (x_1, x_2) \cdot (y_1, y_2) = x_1 y_1 + x_2 y_2$$

The inner product is a **SCALAR!**

$$\mathbf{v} \cdot \mathbf{w} = (x_1, x_2) \cdot (y_1, y_2) = \|\mathbf{v}\| \cdot \|\mathbf{w}\| \cos \alpha$$

Compute the inner product of

$$(0, 1) \cdot (1, 0) = ?$$

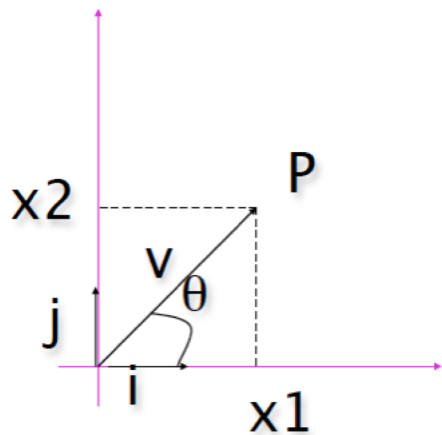
$$(1, 1) \cdot (1, 1) = ?$$

$$(1, 0) \cdot (-1, 0) = ?$$



# Orthonormal Basis

- Unit vectors; orthogonal to each other
- Any vector  $\rightarrow$  linear combination of the basis



$$\mathbf{i} = (1,0) \quad \|\mathbf{i}\| = 1 \quad \mathbf{i} \cdot \mathbf{j} = 0$$

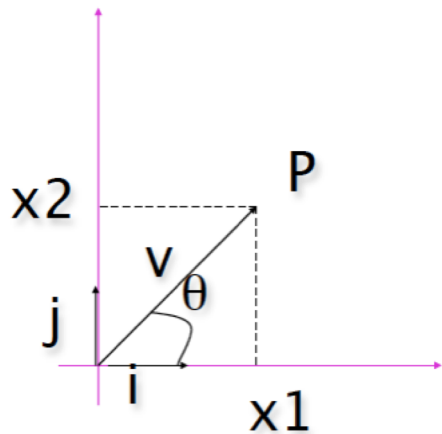
$$\mathbf{j} = (0,1) \quad \|\mathbf{j}\| = 1$$

$$\mathbf{v} = (x_1, x_2)$$

$$\mathbf{v} = x_1 \mathbf{i} + x_2 \mathbf{j}$$

# Orthonormal Basis

- Unit vectors; orthogonal to each other
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$$\mathbf{i} = (1,0) \quad \|\mathbf{i}\| = 1 \quad \mathbf{i} \cdot \mathbf{j} = 0$$

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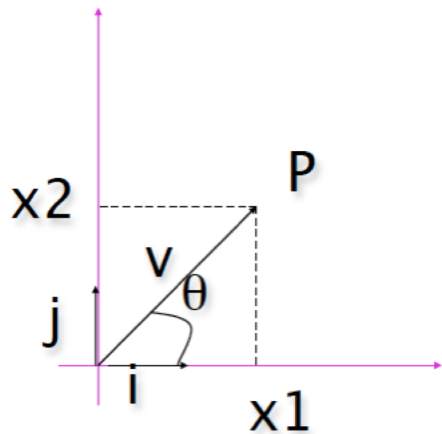
$$\mathbf{v} = x_1 \mathbf{i} + x_2 \mathbf{j}$$

$$\mathbf{v} \cdot \mathbf{i} =$$

$$\mathbf{v} \cdot \mathbf{j} =$$


# Orthonormal Basis

- Unit vectors; orthogonal to each other
- Any vector  $\rightarrow$  linear combination of the basis



$$\mathbf{i} = (1,0) \quad \|\mathbf{i}\| = 1 \quad \mathbf{i} \cdot \mathbf{j} = 0$$

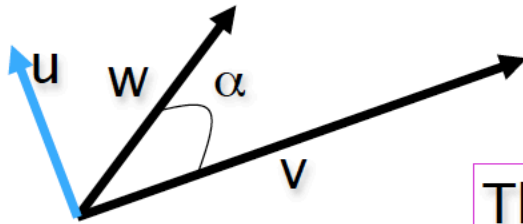
$$\mathbf{j} = (0,1) \quad \|\mathbf{j}\| = 1$$

$$\mathbf{v} = (x_1, x_2) \quad \mathbf{v} = x_1 \mathbf{i} + x_2 \mathbf{j}$$

$$\mathbf{v} \cdot \mathbf{i} = ? = (x_1 \mathbf{i} + x_2 \mathbf{j}) \cdot \mathbf{i} = x_1 \cdot 1 + x_2 \cdot 0 = x_1$$

$$\mathbf{v} \cdot \mathbf{j} = (x_1 \mathbf{i} + x_2 \mathbf{j}) \cdot \mathbf{j} = x_1 \cdot 0 + x_2 \cdot 1 = x_2$$

# Vector (cross) Product

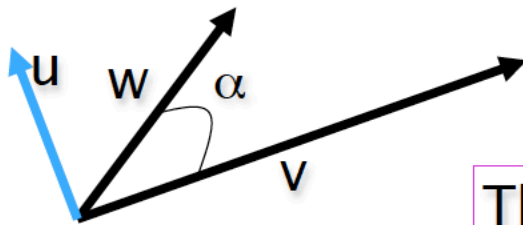


$$u = v \times w$$

The cross product is a **VECTOR!**

$$\text{Magnitude: } \|u\| = \|v \times w\| = \|v\| \|w\| \sin \alpha$$

# Vector (cross) Product



$$u = v \times w$$

The cross product is a **VECTOR!**

Magnitude:  $\|u\| = \|v \times w\| = \|v\| \|w\| \sin \alpha$

Orientation:

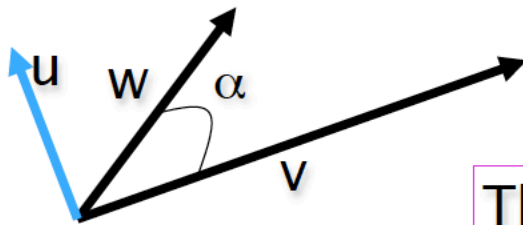
$$u \perp v \Rightarrow u \cdot v =$$

$$u \perp w \Rightarrow u \cdot w =$$





# Vector (cross) Product




$$u = v \times w$$

The cross product is a **VECTOR!**

Magnitude:  $\|u\| = \|v \times w\| = \|v\| \|w\| \sin \alpha$

Orientation:  $u \perp v \Rightarrow u \cdot v = (v \times w) \cdot v = 0$

$$u \perp w \Rightarrow u \cdot w = (v \times w) \cdot w = 0$$

if  $v \parallel w$  ?  $\rightarrow u =$  

# Vector Product Computation

- Orthonormal basis

$$\mathbf{i} = (1,0,0) \quad \|\mathbf{i}\| = 1 \quad \mathbf{i} = \mathbf{j} \times \mathbf{k}$$

$$\mathbf{j} = (0,1,0) \quad \|\mathbf{j}\| = 1 \quad \mathbf{j} = \mathbf{k} \times \mathbf{i}$$

$$\mathbf{k} = (0,0,1) \quad \|\mathbf{k}\| = 1 \quad \mathbf{k} = \mathbf{i} \times \mathbf{j}$$

$$\mathbf{u} = \mathbf{v} \times \mathbf{w} = (x_1, x_2, x_3) \times (y_1, y_2, y_3)$$



# Vector Product Computation

- Orthonormal basis

$$\mathbf{i} = (1,0,0) \quad \|\mathbf{i}\| = 1 \quad \mathbf{i} = \mathbf{j} \times \mathbf{k}$$

$$\mathbf{j} = (0,1,0) \quad \|\mathbf{j}\| = 1 \quad \mathbf{j} = \mathbf{k} \times \mathbf{i}$$

$$\mathbf{k} = (0,0,1) \quad \|\mathbf{k}\| = 1 \quad \mathbf{k} = \mathbf{i} \times \mathbf{j}$$

$$\mathbf{u} = \mathbf{v} \times \mathbf{w} = (x_1, x_2, x_3) \times (y_1, y_2, y_3)$$

$$= (x_2 y_3 - x_3 y_2) \mathbf{i} + (x_3 y_1 - x_1 y_3) \mathbf{j} + (x_1 y_2 - x_2 y_1) \mathbf{k}$$

# Matrices

$$A_{n \times m} = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1m} \\ a_{21} & a_{22} & \dots & a_{2m} \\ \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nm} \end{bmatrix}$$



Pixel's intensity value

# Matrix addition

$$A_{n \times m} = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1m} \\ a_{21} & a_{22} & \dots & a_{2m} \\ \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nm} \end{bmatrix}$$



Pixel's intensity value

Sum:  $C_{n \times m} = A_{n \times m} + B_{n \times m} \quad c_{ij} = a_{ij} + b_{ij}$

A and B must have the same dimensions!

Example:  $\begin{bmatrix} 2 & 5 \\ 3 & 1 \end{bmatrix} + \begin{bmatrix} 6 & 2 \\ 1 & 5 \end{bmatrix} = \begin{bmatrix} 8 & 7 \\ 4 & 6 \end{bmatrix}$

# Matrix product

$$A_{n \times m} = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1m} \\ a_{21} & a_{22} & \dots & a_{2m} \\ \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nm} \end{bmatrix} \mathbf{a}_i$$

$$B_{m \times p} = \begin{bmatrix} b_{11} & b_{12} & \dots & b_{1p} \\ b_{21} & b_{22} & \dots & b_{2p} \\ \vdots & \vdots & \vdots & \vdots \\ b_{m1} & b_{m2} & \dots & b_{mp} \end{bmatrix} \mathbf{b}_j$$

Product:

$$C_{n \times p} = A_{n \times m} B_{m \times p}$$

$$c_{ij} = \mathbf{a}_i \cdot \mathbf{b}_j = \sum_{k=1}^m a_{ik} b_{kj}$$

A and B must have compatible dimensions!

$$A_{n \times n} B_{n \times n} \begin{matrix} = \\ ? \end{matrix} B_{n \times n} A_{n \times n}$$

# Matrix transpose

Transpose:

$$C_{m \times n} = A^T_{n \times m}$$

$$c_{ij} = a_{ji}$$

$$(A + B)^T = A^T + B^T$$

$$(AB)^T = B^T A^T$$

If  $A^T = A$       A is symmetric

Examples:

$$\begin{bmatrix} 6 & 2 \\ 1 & 5 \end{bmatrix}^T = \begin{bmatrix} 6 & 1 \\ 2 & 5 \end{bmatrix}$$

$$\begin{bmatrix} 6 & 2 \\ 1 & 5 \\ 3 & 8 \end{bmatrix}^T = \begin{bmatrix} 6 & 1 & 3 \\ 2 & 5 & 8 \end{bmatrix}$$

$$\begin{bmatrix} 5 & 2 \\ 1 & 5 \end{bmatrix}$$

$$\begin{bmatrix} 3 & 2 \\ 2 & 7 \end{bmatrix}$$



# Matrix determinant

$$\det \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} = \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = a_{11}a_{22} - a_{21}a_{12}$$


$$\det \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} = a_{11} \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} - a_{12} \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix} + a_{13} \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix}$$

(alternate signs)

A must be square

Example:  $\det \begin{bmatrix} 2 & 5 \\ 3 & 1 \end{bmatrix} =$



$$\begin{vmatrix} 1 & 2 & 1 \\ 3 & 0 & 4 \\ 8 & 4 & 10 \end{vmatrix} =$$




# Matrix inverse

- Reciprocal



$$8 \times (1/8) = \mathbf{1}$$

- Inverse of a matrix



$$A \times A^{-1} = \mathbf{I}$$

$$\mathbf{I} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

When the inverse comes first:

$$(1/8) \times 8 = \mathbf{1}$$

$$A^{-1} \times A = \mathbf{I}$$

Identity Matrix

# Matrix inverse: compute

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix}^{-1} = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

↑  
determinant

1. **Swap** the positions of a and d
2. Put **negatives** in front of b and c
3. **Divide** everything by the determinant

# Matrix inverse: always exist?

- A must be square

$$\begin{bmatrix} 6 & 1 & 3 \\ 2 & 5 & 8 \end{bmatrix}$$

# Matrix inverse: always exist?

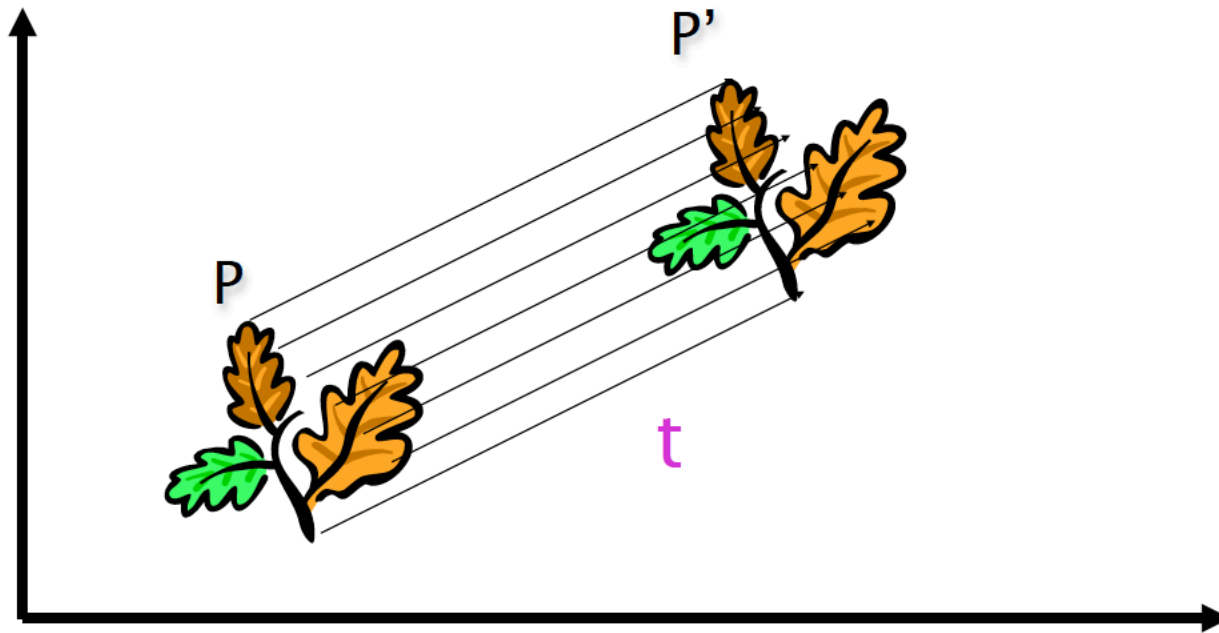
- A must be square
- A must be non-singular

$$\begin{bmatrix} 3 & 4 \\ 6 & 8 \end{bmatrix}^{-1}$$

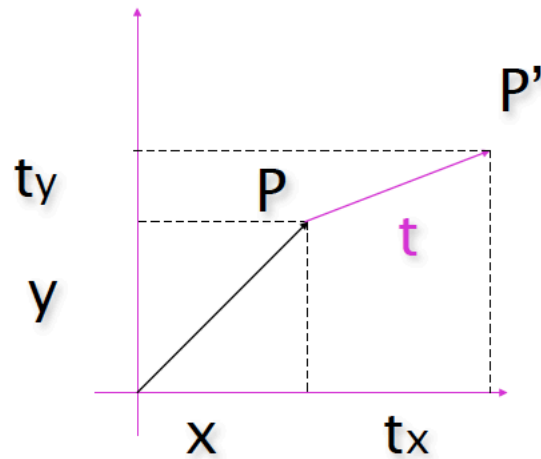


# 2D Geometric Transformations

# 2D Translation



# 2D Translation Equation

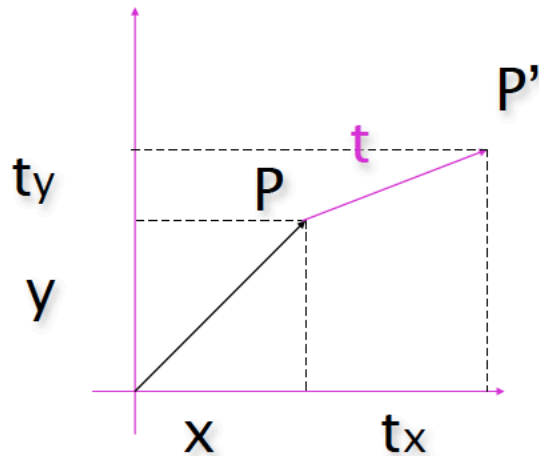


$$\mathbf{P} = (x, y)$$

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

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

# 2D Translation using Matrices



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

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

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

Not square

What is the inverse transformation?



# 2D Translation using Matrices

- Homogeneous coordinates
  - Multiply the coordinates by a non-zero scalar and add an extra coordinate equal to that scalar

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

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

# 2D Translation using Matrices

- Homogeneous coordinates
  - Multiply the coordinates by a non-zero scalar and add an extra coordinate equal to that scalar

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

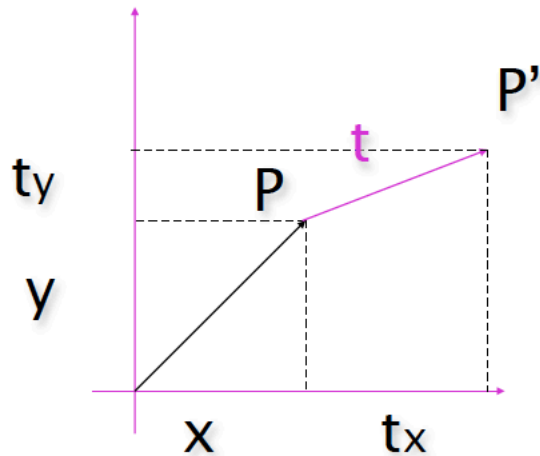
$$(x, y, z) \rightarrow (x \cdot w, y \cdot w, z \cdot w, w) \quad w \neq 0$$

- Back to Cartesian coordinates
  - Divide by the last coordinate and eliminate it

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

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

# 2D Translation using Homogeneous Coordinates



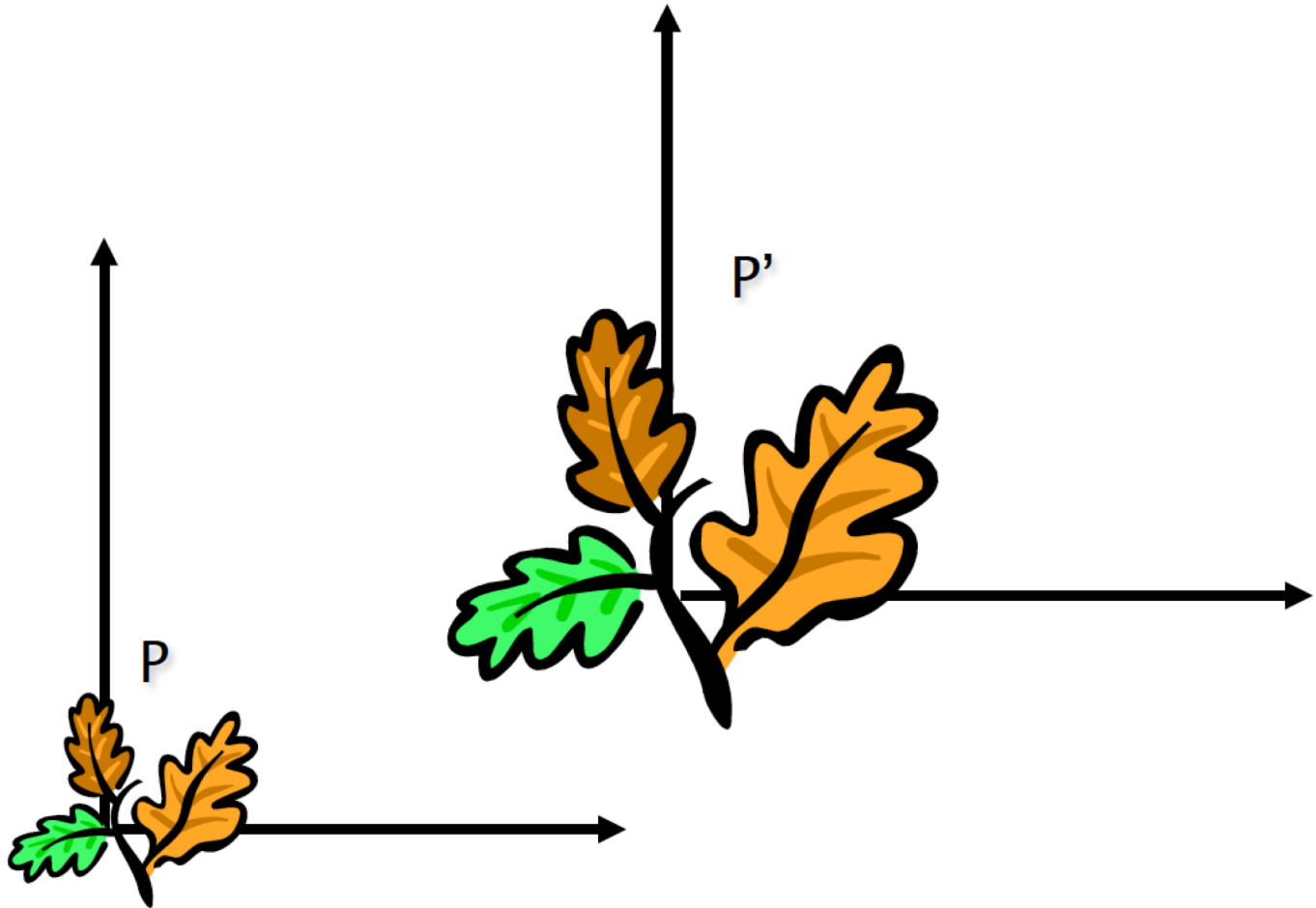
$$\mathbf{P} = (x, y) \rightarrow (x, y, 1)$$

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

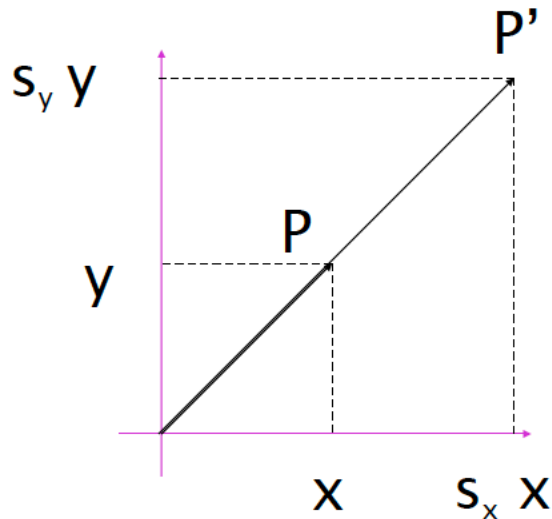
$$\mathbf{P}' \rightarrow \begin{bmatrix} x + t_x \\ y + t_y \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & t_x \\ 0 & 1 & t_y \\ 0 & 0 & 1 \end{bmatrix} \cdot \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

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

# Scaling



# Scaling Equation



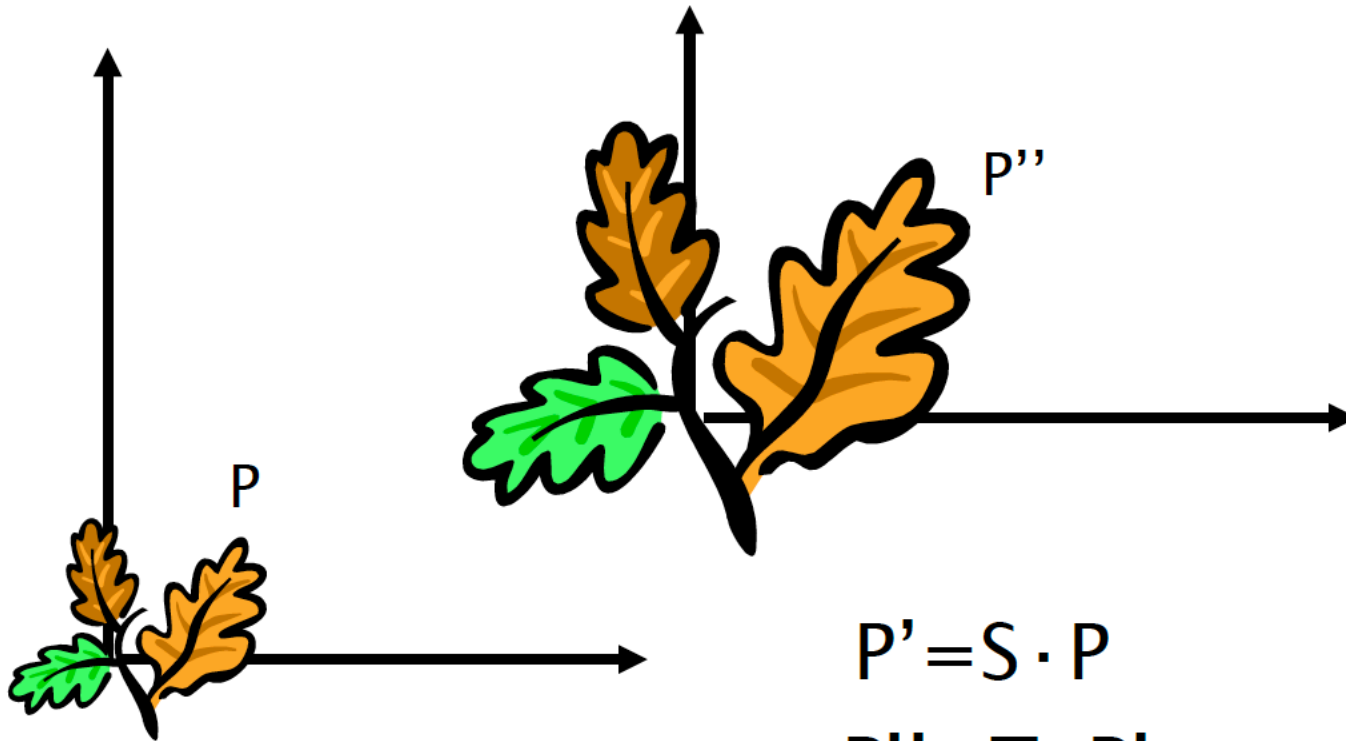
$$\mathbf{P} = (x, y) \rightarrow \mathbf{P}' = (s_x x, s_y y)$$

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

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

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

# Scaling & Translation



$$P' = S \cdot P$$

$$P''' = T \cdot P'$$

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

# Scaling & Translation

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

$$= \underbrace{\begin{bmatrix} \text{?} \\ \text{?} \\ \text{?} \end{bmatrix}}_A \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

# Scaling & Translation

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

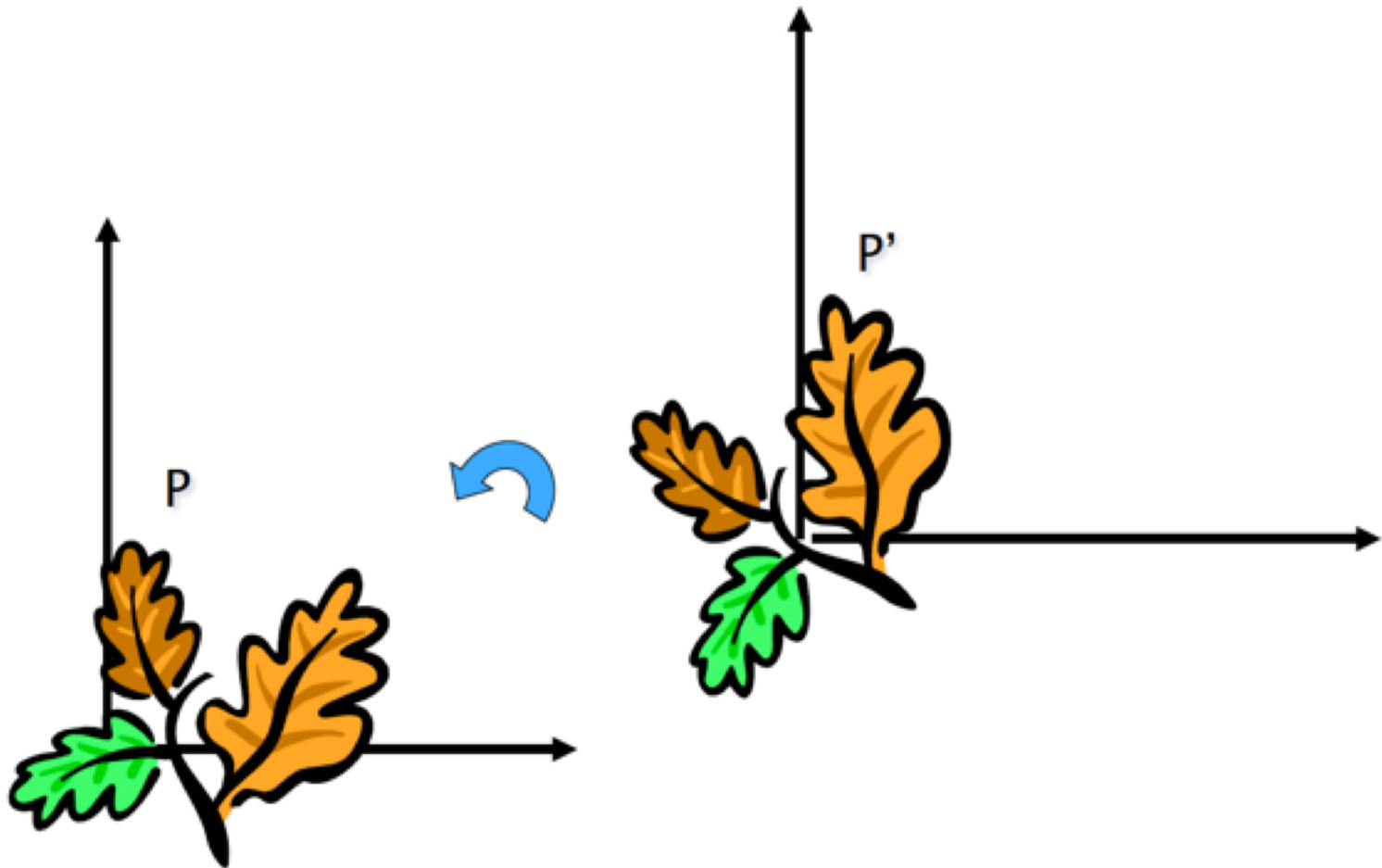


# Translation & Scaling = Scaling & Translation ?

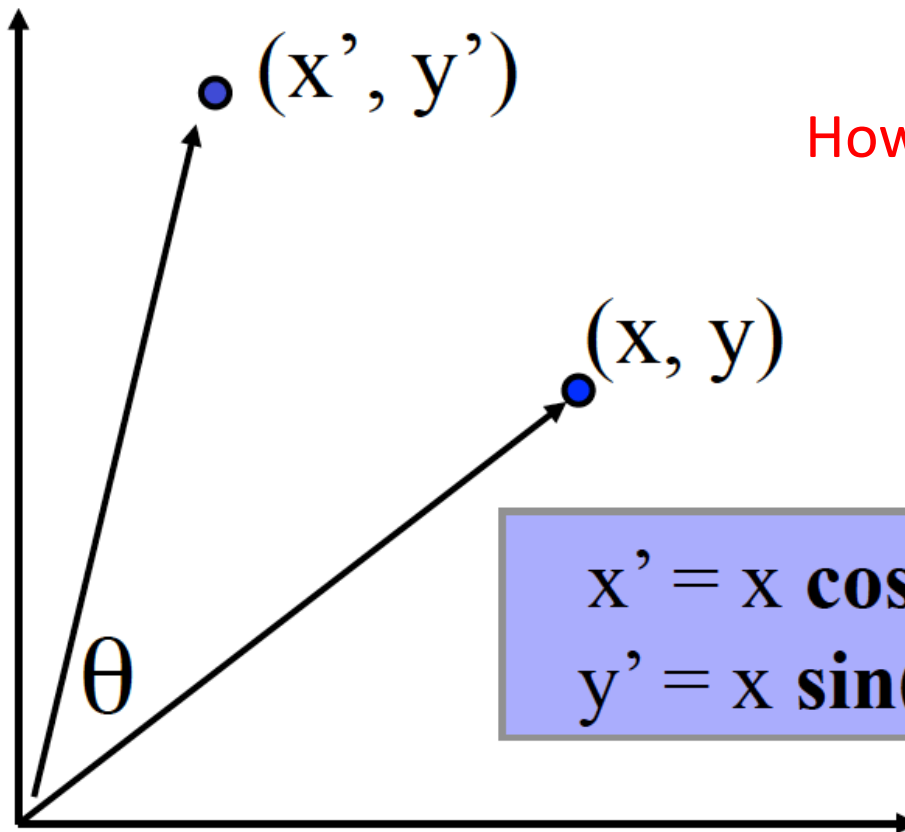
$$\mathbf{P}''' = \mathbf{T} \cdot \mathbf{S} \cdot \mathbf{P} = \begin{bmatrix} 1 & 0 & t_x \\ 0 & 1 & t_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} s_x & 0 & 0 \\ 0 & s_y & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} s_x & 0 & t_x \\ 0 & s_y & t_y \\ 0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

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

# Rotation



# Rotation Equations

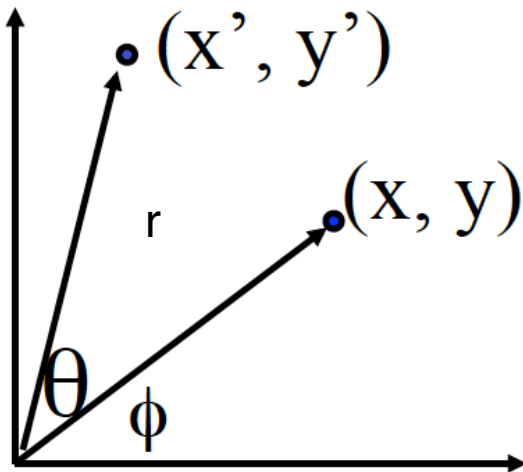


How to derive this equation?



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

# Rotation Equations



$$x = r \cos(\phi)$$

$$y = r \sin(\phi)$$

$$x' = r \cos(\phi + \theta)$$

$$y' = r \sin(\phi + \theta)$$

↓ Expand

$$x' = r \cos(\phi) \cos(\theta) - r \sin(\phi) \sin(\theta)$$

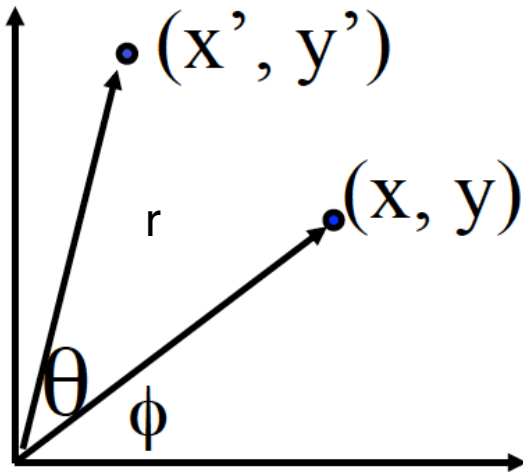
$$y' = r \sin(\phi) \cos(\theta) + r \cos(\phi) \sin(\theta)$$

↓ Substitute

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

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

# Rotation Equations



This is easy to capture in matrix form:

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \underbrace{\begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix}}_{\mathbf{R}} \begin{bmatrix} x \\ y \end{bmatrix}$$

What is the inverse transformation



- Rotation by  $-\theta$

$\mathbf{R}$  has many interesting properties:

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

# Translation + Rotation + Scaling

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

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

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

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

# Assignment 1

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## Matrix/vector arithmetic

# Next Lecture

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Part 2: Linear systems, linear least squares