

Linear Algebra Primer

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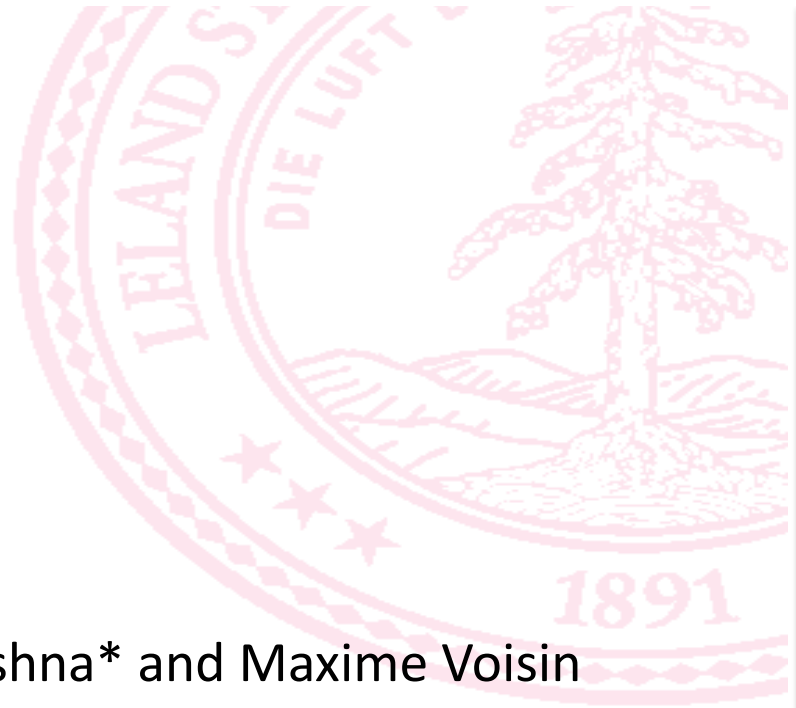
*Stanford Vision and Learning Lab

Another, very in-depth linear algebra review from CS229 is available here:

<http://cs229.stanford.edu/section/cs229-linalg.pdf>

And a video discussion of linear algebra from EE263 is here (lectures 3 and 4):

<https://see.stanford.edu/Course/EE263>



Outline

- [Vectors and matrices](#)
 - Basic Matrix Operations
 - Determinants, norms, trace
 - Special Matrices
- [Transformation Matrices](#)
 - Homogeneous coordinates
 - Translation
- [Matrix inverse](#)
- [Matrix rank](#)
- Eigenvalues and Eigenvectors
- Matrix Calculus





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Vectors and matrices are just collections of ordered numbers that represent something: location in space, speed, pixel brightness, etc. We'll define some common uses and standard operations on them.

$$\mathbf{v} = \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix} \quad \mathbf{v}^T = \begin{bmatrix} v_1 & v_2 & \dots & v_n \end{bmatrix}$$
$$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} \\ \vdots & & & & \vdots \\ a_{m1} & a_{m2} & a_{m3} & \dots & a_{mn} \end{bmatrix}$$



Vector

- A column vector $\mathbf{v} \in \mathbb{R}^{n \times 1}$ where

$$\mathbf{v} = \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix}$$

- A row vector $\mathbf{v}^T \in \mathbb{R}^{1 \times n}$ where

$$\mathbf{v}^T = [v_1 \quad v_2 \quad \dots \quad v_n]$$

T denotes the transpose operation

Vector

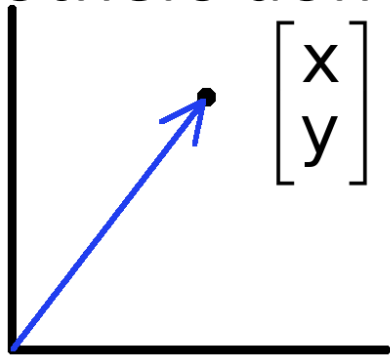
- We'll default to column vectors in this class

$$\mathbf{v} = \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix}$$

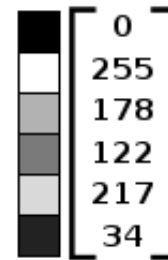
- You'll want to keep track of the orientation of your vectors when programming in python



Some vectors have a geometric interpretation, others don't...



- Some vectors have a geometric interpretation:
 - Points are just vectors from the origin.
 - We can make calculations like “distance” between 2 vectors



- Other vectors don't have a geometric interpretation:
 - Vectors can represent any kind of data (pixels, gradients at an image keypoint, etc)
 - Such vectors don't have a geometric interpretation
 - We can still make calculations like “distance” between 2 vectors





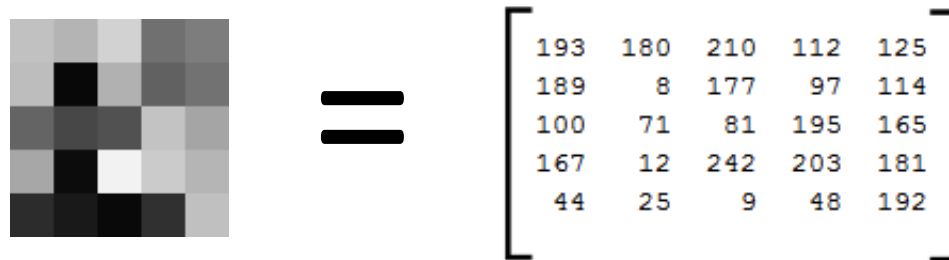
Matrix

- A matrix $\mathbf{A} \in \mathbb{R}^{m \times n}$ is an array of numbers with size m by n , i.e. m rows and n columns.

$$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} \\ \vdots & & & & \vdots \\ a_{m1} & a_{m2} & a_{m3} & \dots & a_{mn} \end{bmatrix}$$

- If $m = n$, we say that \mathbf{A} is square.

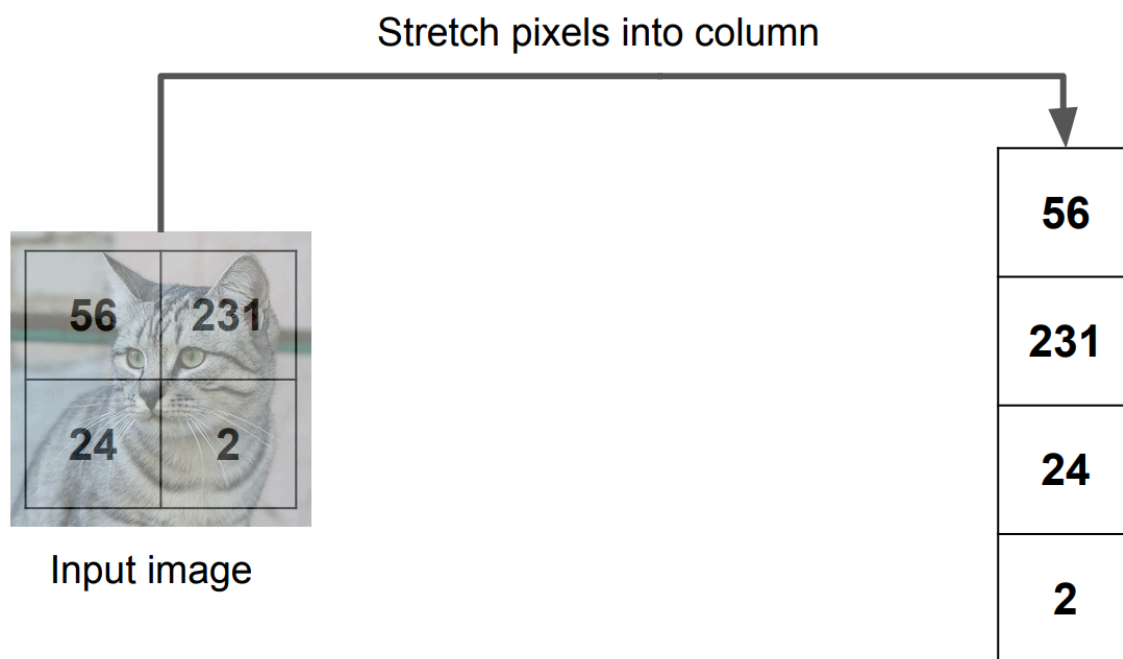
Images



- Python represents an image as a matrix of pixel brightnesses
- Note that the upper left corner is $[x, y] = (0,0)$

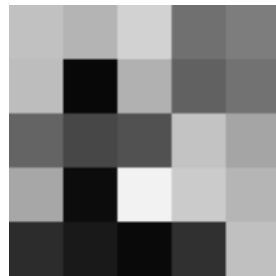
row column Python indices start at 0

Images can be represented as a **matrix** of pixels.
Images can also be represented as a **vector** of pixels

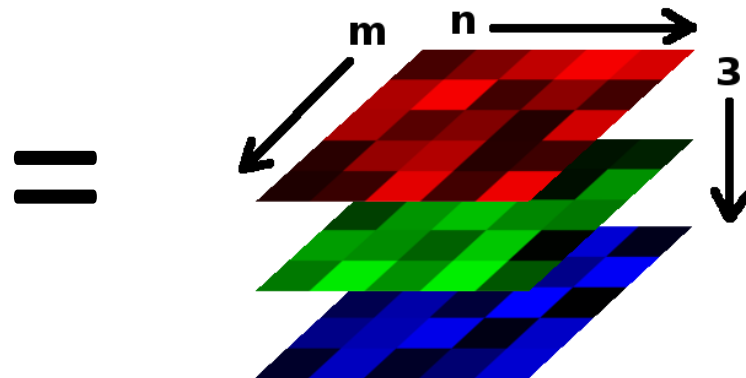


Color Images

- **Grayscale images** have **1** number per pixel, and are stored as an **$m \times n$ matrix**.



- **Color images** have **3** numbers per pixel – red, green, and blue brightnesses (RGB) - and are stored as an **$m \times n \times 3$ matrix**



Basic Matrix Operations

- We will discuss:
 - Addition
 - Scaling
 - Dot product
 - Multiplication
 - Transpose
 - Inverse / pseudoinverse
 - Determinant / trace





Matrix Operations

- Addition

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} + \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} = \begin{bmatrix} a+1 & b+2 \\ c+3 & d+4 \end{bmatrix}$$

– We can only add a matrix with matching dimensions, or a scalar.

Good to know for Python assignments 😊

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} + 7 = \begin{bmatrix} a+7 & b+7 \\ c+7 & d+7 \end{bmatrix}$$

- Scaling

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} \times 3 = \begin{bmatrix} 3a & 3b \\ 3c & 3d \end{bmatrix}$$



Vector Norms

- **Examples of vector norms**

$$\|x\|_1 = \sum_{i=1}^n |x_i| \quad \left| \quad \|x\|_2 = \sqrt{\sum_{i=1}^n x_i^2} \quad \left| \quad \|x\|_\infty = \max_i |x_i| \right.$$



Vector Norms

- More formally, a norm is any function $f : \mathbb{R}^n \rightarrow \mathbb{R}$ that satisfies the following 4 properties:
- **Non-negativity:** For all $x \in \mathbb{R}^n$, $f(x) \geq 0$
- **Definiteness:** $f(x) = 0$ if and only if $x = [0, 0 \dots 0]$.
- **Homogeneity:** For all $x \in \mathbb{R}^n$, $t \in \mathbb{R}$, $f(tx) = |t|f(x)$
- **Triangle inequality:** For all $x, y \in \mathbb{R}^n$, $f(x + y) \leq f(x) + f(y)$



Vector Operation: inner product

- **Inner product (dot product) of two vectors**
 - Multiply corresponding entries of two vectors and add up the result
 - $\mathbf{x} \cdot \mathbf{y}$ is also $|\mathbf{x}| |\mathbf{y}| \cos(\text{the angle between } \mathbf{x} \text{ and } \mathbf{y})$

$$\mathbf{x}^T \mathbf{y} = \begin{bmatrix} x_1 & \dots & x_n \end{bmatrix} \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} = \sum_{i=1}^n x_i y_i \quad (\text{scalar})$$

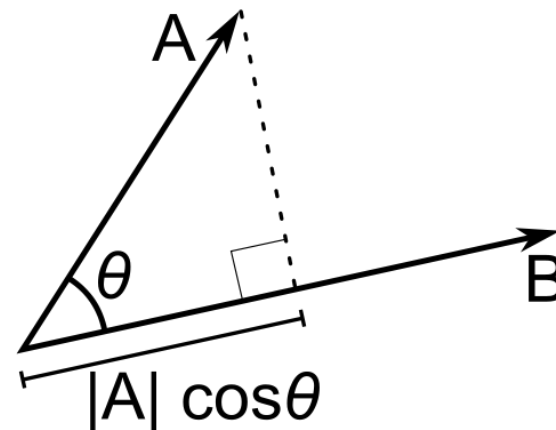
Vector Operation: inner product

- **Inner product** (dot product) **of two vectors**

- If B is a **unit** vector:

- Then $A \cdot B = |A| |B| \cos(\Theta) = |A| \times 1 \times \cos(\Theta) = |A| \cos(\Theta)$

- $A \cdot B$ gives the length of A which lies in the direction of B





Matrix Operations

- The **product** of two matrices

$$A \in \mathbb{R}^{m \times n}$$

$$B \in \mathbb{R}^{n \times p}$$

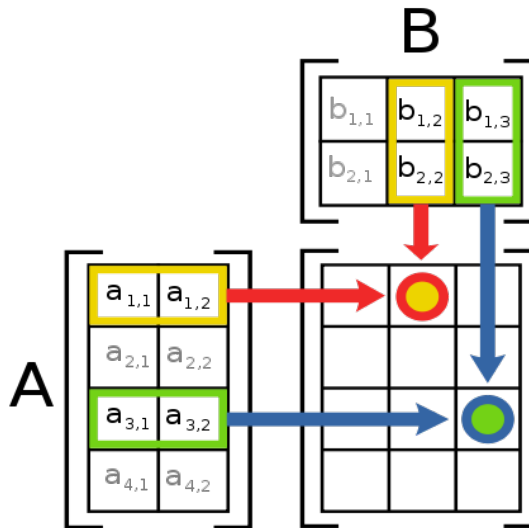
$$C = AB \in \mathbb{R}^{m \times p}$$

$$C_{ij} = \sum_{k=1}^n A_{ik} B_{kj}$$

$$C = AB = \begin{bmatrix} - & a_1^T & - \\ - & a_2^T & - \\ & \vdots & \\ - & a_m^T & - \end{bmatrix} \begin{bmatrix} | & | & \cdots & | \\ b_1 & b_2 & \cdots & b_p \\ | & | & & | \end{bmatrix} = \begin{bmatrix} a_1^T b_1 & a_1^T b_2 & \cdots & a_1^T b_p \\ a_2^T b_1 & a_2^T b_2 & \cdots & a_2^T b_p \\ \vdots & \vdots & \ddots & \vdots \\ a_m^T b_1 & a_m^T b_2 & \cdots & a_m^T b_p \end{bmatrix}.$$

Matrix Operations

- Multiplication
- The product AB is:



- Each entry in the result is:
(that row of A) dot product with (that column of B)
- Many uses, which will be covered later



Matrix Operations

- Multiplication example:

$$\begin{array}{ccc} A & \times & B \\ \downarrow & & \nearrow \\ \begin{bmatrix} 0 & 2 \\ 4 & 6 \end{bmatrix} & & \begin{bmatrix} 1 & 3 \\ 5 & 7 \end{bmatrix} \\ & & \begin{bmatrix} 10 & 14 \\ 34 & 54 \end{bmatrix} \end{array}$$

$$0 \cdot 3 + 2 \cdot 7 = 14$$

- Each entry of the matrix product is made by taking: the dot product of the corresponding row in the left matrix, with the corresponding column in the right one.





Matrix Operations

- The product of two matrices

Matrix multiplication is associative: $(AB)C = A(BC)$.

Matrix multiplication is distributive: $A(B + C) = AB + AC$.

Matrix multiplication is, in general, *not* commutative; that is, it can be the case that $AB \neq BA$. (For example, if $A \in \mathbb{R}^{m \times n}$ and $B \in \mathbb{R}^{n \times q}$, the matrix product BA does not even exist if m and q are not equal!)

Matrix Operations

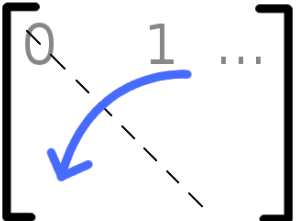
- Powers
 - By convention, we can refer to the matrix product AA as A^2 , and AAA as A^3 , etc.
 - **Important: only square matrices can be multiplied that way!**
(make sure you understand why)





Matrix Operations

- Transpose a matrix: flip the matrix, so row 1 becomes column 1


$$\begin{bmatrix} 0 & 1 & \dots \\ 2 & 3 & \\ 4 & 5 & \end{bmatrix}^T = \begin{bmatrix} 0 & 2 & 4 \\ 1 & 3 & 5 \end{bmatrix}$$

- A useful identity:

$$(ABC)^T = C^T B^T A^T$$

Matrix Operations

- **Determinant**

- $\det(\mathbf{A})$ returns a scalar
- Represents area (or volume) of the parallelogram described by the vectors in the rows of the matrix

- For $\mathbf{A} = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$, $\det(\mathbf{A}) = ad - bc$

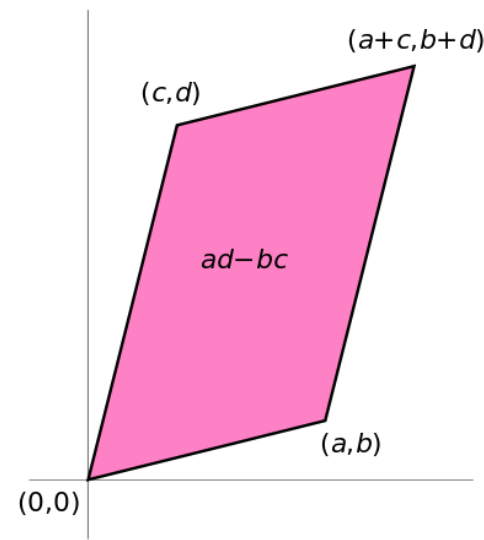
- Properties:

$$\det(\mathbf{AB}) = \det(\mathbf{BA})$$

$$\det(\mathbf{A}^{-1}) = \frac{1}{\det(\mathbf{A})}$$

$$\det(\mathbf{A}^T) = \det(\mathbf{A})$$

$$\det(\mathbf{A}) = 0 \Leftrightarrow \mathbf{A} \text{ is singular}$$





Matrix Operations

- **Trace**

$\text{tr}(\mathbf{A}) = \text{sum of diagonal elements}$

$$\text{tr}\left(\begin{bmatrix} 1 & 3 \\ 5 & 7 \end{bmatrix}\right) = 1 + 7 = 8$$

- Invariant to a lot of transformations, so it's used sometimes in proofs. (Rarely in this class though.)
- Properties:

$$\text{tr}(\mathbf{AB}) = \text{tr}(\mathbf{BA})$$

$$\text{tr}(\mathbf{A} + \mathbf{B}) = \text{tr}(\mathbf{A}) + \text{tr}(\mathbf{B})$$



Matrix Operations

- **Vector Norms** (we've talked about them earlier)

$$\|x\|_1 = \sum_{i=1}^n |x_i|$$

$$\|x\|_\infty = \max_i |x_i|$$

$$\|x\|_2 = \sqrt{\sum_{i=1}^n x_i^2}$$

$$\|x\|_p = \left(\sum_{i=1}^n |x_i|^p \right)^{1/p}$$

- **Matrix norms:** Norms can also be defined for matrices, such as

$$\|A\|_F = \sqrt{\sum_{i=1}^m \sum_{j=1}^n A_{ij}^2} = \sqrt{\text{tr}(A^T A)}.$$



Special Matrices

- **Identity matrix \mathbf{I}**

Square matrix, 1's along diagonal, 0's elsewhere

$\mathbf{I} \cdot [\text{a matrix A}] = [\text{that matrix A}]$
 $[\text{a matrix A}] \cdot \mathbf{I} = [\text{that matrix A}]$

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

- **Diagonal matrix**

Square matrix with numbers along diagonal, 0's elsewhere

$[\text{diagonal matrix A}] \cdot [\text{another matrix B}]$ scales the rows of matrix B

$$\begin{bmatrix} 3 & 0 & 0 \\ 0 & 7 & 0 \\ 0 & 0 & 2.5 \end{bmatrix}$$

Special Matrices

- Symmetric matrix

$$\mathbf{A}^T = \mathbf{A}$$

$$\begin{bmatrix} 1 & 2 & 5 \\ 2 & 1 & 7 \\ 5 & 7 & 1 \end{bmatrix}$$


- Skew-symmetric matrix

$$\mathbf{A}^T = -\mathbf{A}$$

$$\begin{bmatrix} 0 & -2 & -5 \\ 2 & 0 & -7 \\ 5 & 7 & 0 \end{bmatrix}$$



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Matrix multiplication can be used to transform vectors. A matrix used in this way is called a **transformation matrix**.





Transformation: scaling

- **Matrices can be used to transform vectors** in useful ways, through multiplication: $Ax = x'$
- Simplest transformation is **scaling**:

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

(Verify to yourself that the matrix multiplication works out this way)

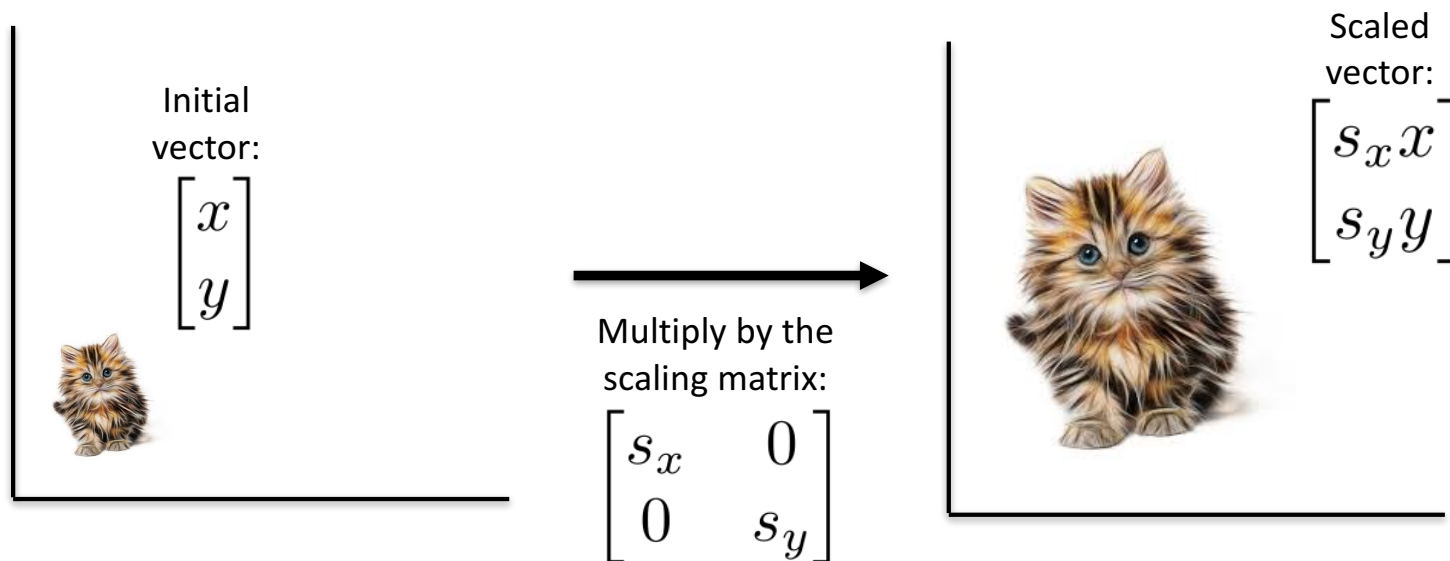
Scaling
matrix

Initial
vector

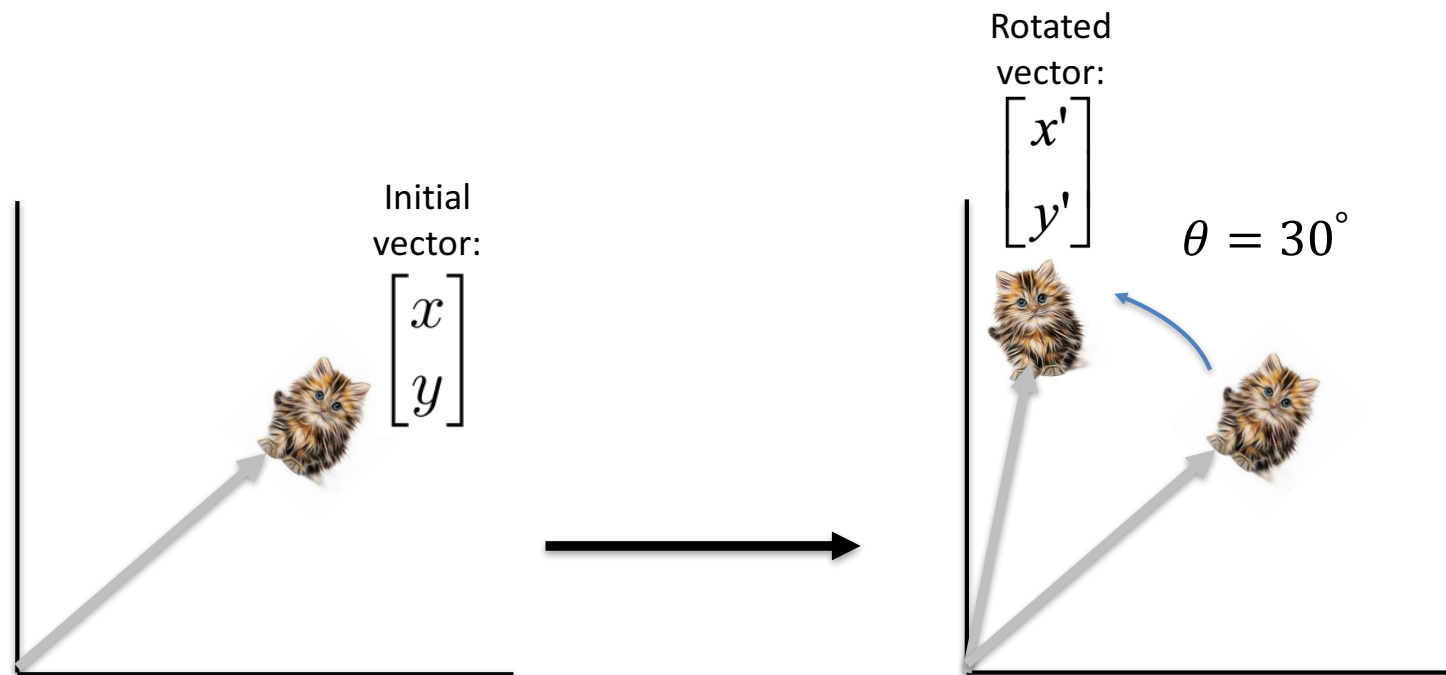
Scaled
vector

Transformation: scaling

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

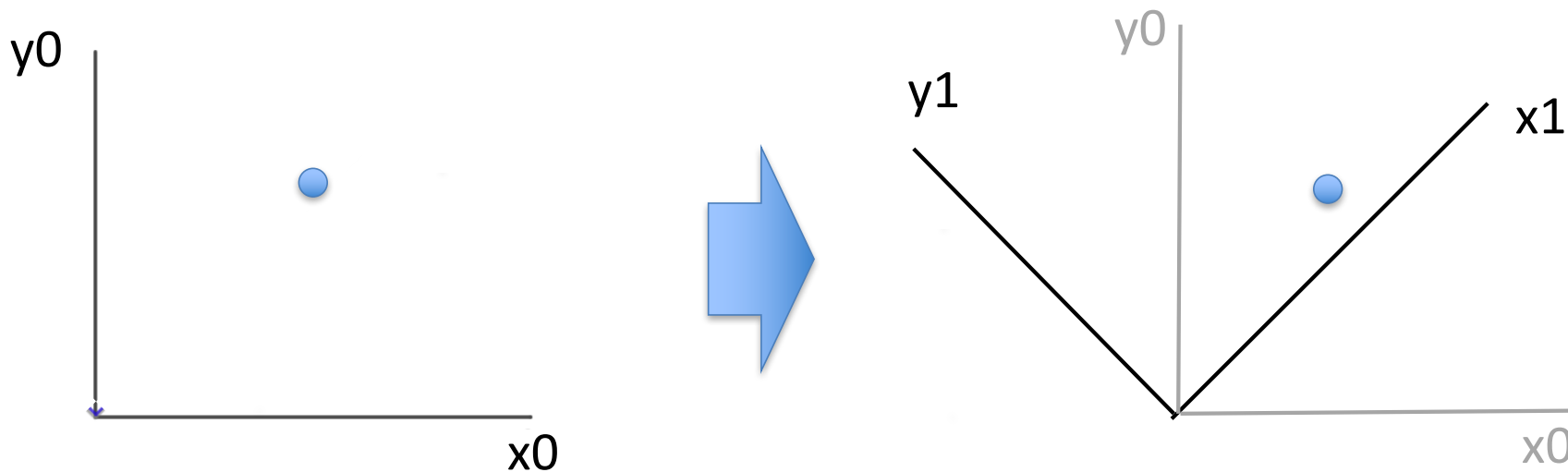


Transformation: rotation



Before learning how to rotate a vector, let's understand how to do something slightly different

- How can you convert a vector represented in frame “0” to a new, rotated coordinate frame “1”?

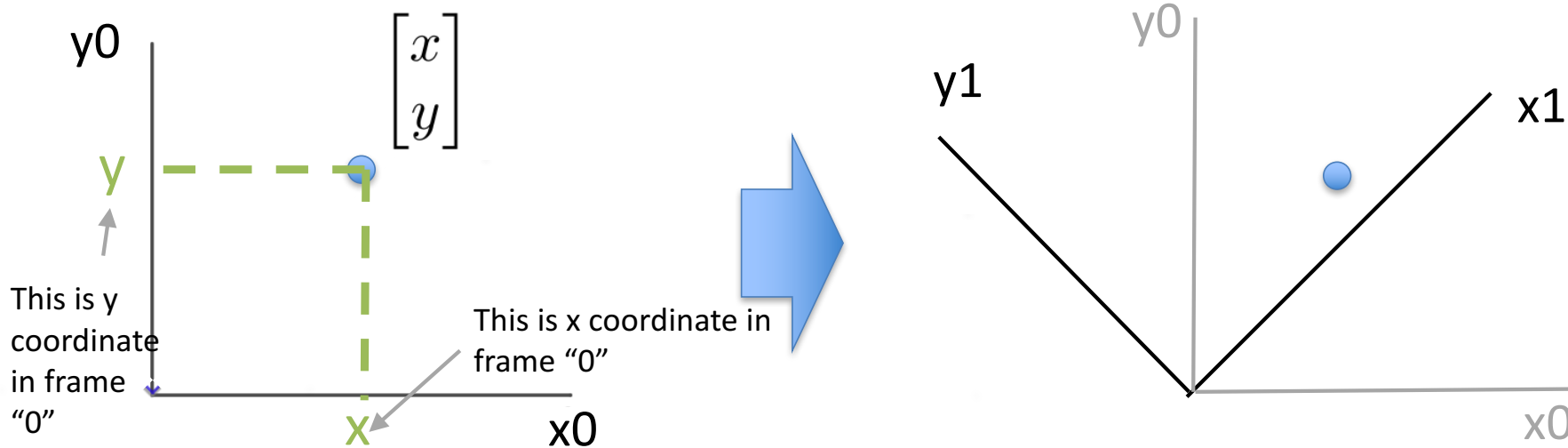


It is the **same** data point, but it is represented in a new, rotated frame.



Before learning how to rotate a vector, let's understand how to do something slightly different

- How can you convert a vector represented in frame "0" to a new, rotated coordinate frame "1"?

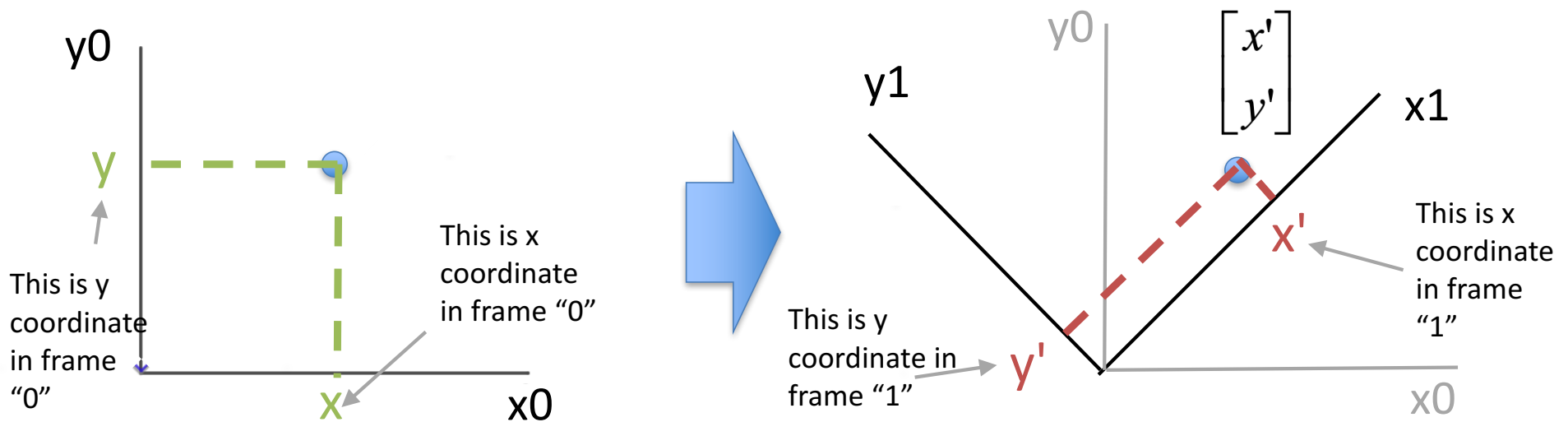


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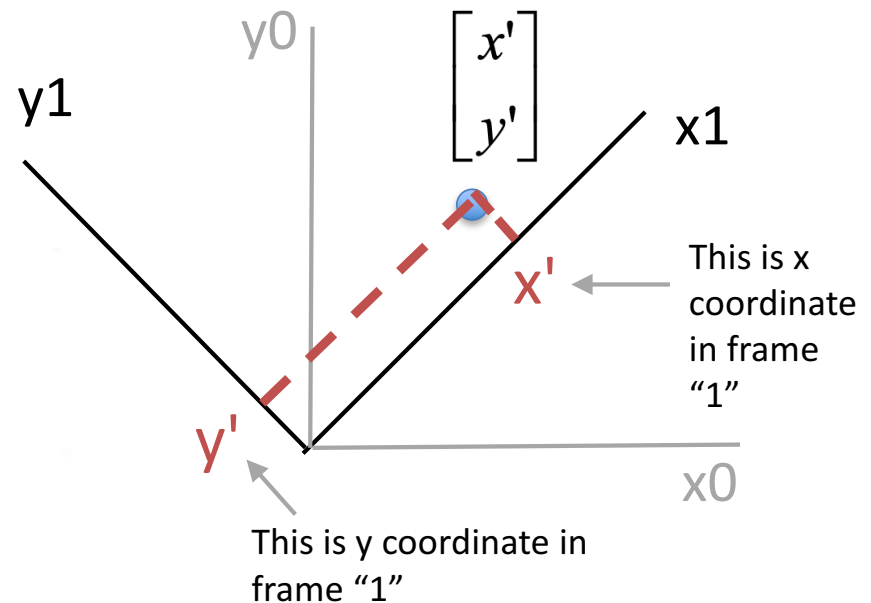
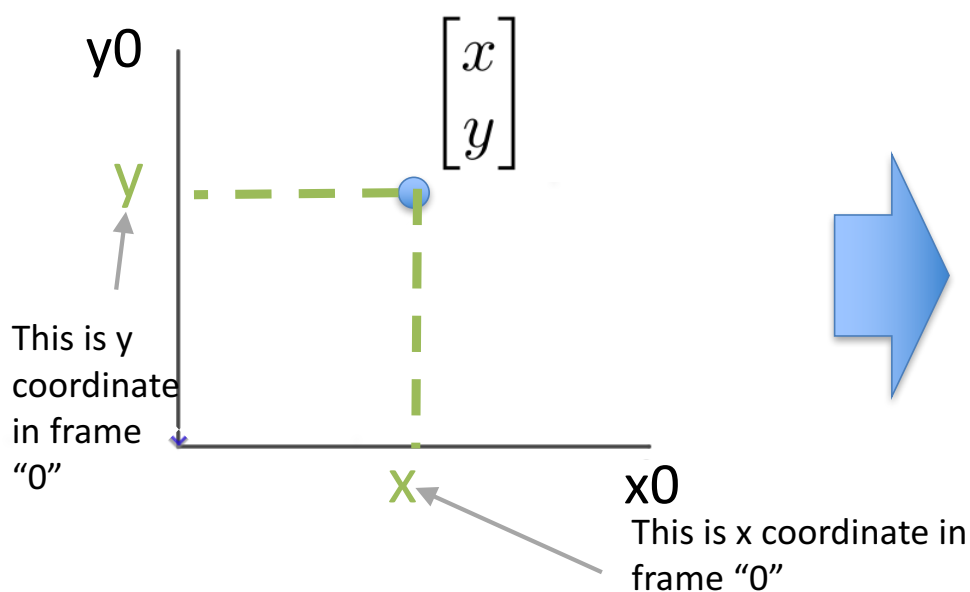
- How can you convert a vector represented in frame "0" to a new, rotated coordinate frame "1"?



It is the **same** data point, but it is represented in a new, rotated frame.

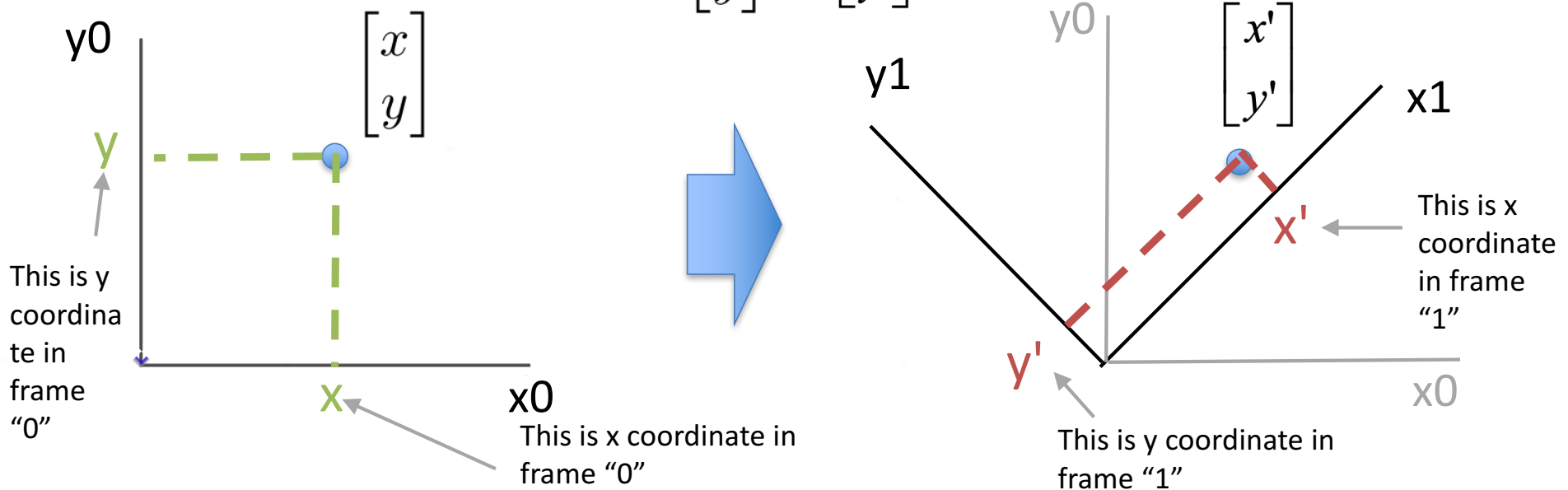
Before learning how to rotate a vector, let's understand how to do something slightly different

- How can you convert a vector represented in frame "0" to a new, rotated coordinate frame "1"?
- Remember what a vector is:
[component in direction of the frame's x axis, component in direction of y axis]



Before learning how to rotate a vector, let's understand how to do something slightly different

- How can you convert a vector represented in frame "0" to a new, rotated coordinate frame "1"?
- Goal: How do we go from $\begin{bmatrix} x \\ y \end{bmatrix}$ to $\begin{bmatrix} x' \\ y' \end{bmatrix}$?





Before learning how to rotate a vector, let's understand how to do something slightly different

- How can you convert a vector represented in frame “0” to a new, rotated coordinate frame “1”?
- Goal: How do we go from $\begin{bmatrix} x \\ y \end{bmatrix}$ to $\begin{bmatrix} x' \\ y' \end{bmatrix}$?
- **Answer: we use dot products!!**
 - x' (the new x coordinate) is the length of the original vector which lies in the direction of the new x axis
 - y' (the new y coordinate) is the length of the original vector which lies in the direction of the new y axis
- So:
 - x' is [original vector] **dot** [the new x axis]
 - y' is [original vector] **dot** [the new y axis]



Before learning how to rotate a vector, let's understand how to do something slightly different

- How can you convert a vector represented in frame “0” to a new, rotated coordinate frame “1”?
- Goal: How do we go from $\begin{bmatrix} x \\ y \end{bmatrix}$ to $\begin{bmatrix} x' \\ y' \end{bmatrix}$?
 - x' (the new x coordinate) is [original vector] **dot** [the new x axis]
 - y' (the new y coordinate) is [original vector] **dot** [the new y axis]
- So:
 - $x' = [x, y] \text{ dot [the new x axis]}$
 - $y' = [x, y] \text{ dot [the new y axis]}$



Before learning how to rotate a vector, let's understand how to do something slightly different

- How can you convert a vector represented in frame “0” to a new, rotated coordinate frame “1”?

- Goal: How do we go from $\begin{bmatrix} x \\ y \end{bmatrix}$ to $\begin{bmatrix} x' \\ y' \end{bmatrix}$?

- $x' = [x, y] \text{ dot [the new x axis]}$

- $y' = [x, y] \text{ dot [the new y axis]}$

- So:

- $\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} \text{[the new x axis] dot [x, y]} \\ \text{[the new y axis] dot [x, y]} \end{bmatrix}$



Before learning how to rotate a vector, let's understand how to do something slightly different

- How can you convert a vector represented in frame “0” to a new, rotated coordinate frame “1”?

- Goal: How do we go from $\begin{bmatrix} x \\ y \end{bmatrix}$ to $\begin{bmatrix} x' \\ y' \end{bmatrix}$?

- $\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} \text{[the new x axis] dot [x, y]} \\ \text{[the new y axis] dot [x, y]} \end{bmatrix}$

- So:

- $\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} \text{[the new x axis]} \\ \text{[the new y axis]} \end{bmatrix} \times \begin{bmatrix} x \\ y \end{bmatrix}$

- $\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} \text{2x2 Rotation Matrix} \end{bmatrix} \times \begin{bmatrix} x \\ y \end{bmatrix}$

→ it is a matrix-vector multiplication!!



Before learning how to rotate a vector, let's understand how to do something slightly different

- How can you convert a vector represented in frame “0” to a new, rotated coordinate frame “1”?
- Goal: How do we go from $\begin{bmatrix} x \\ y \end{bmatrix}$ to $\begin{bmatrix} x' \\ y' \end{bmatrix}$?

→ answer: it is a matrix-vector multiplication!!

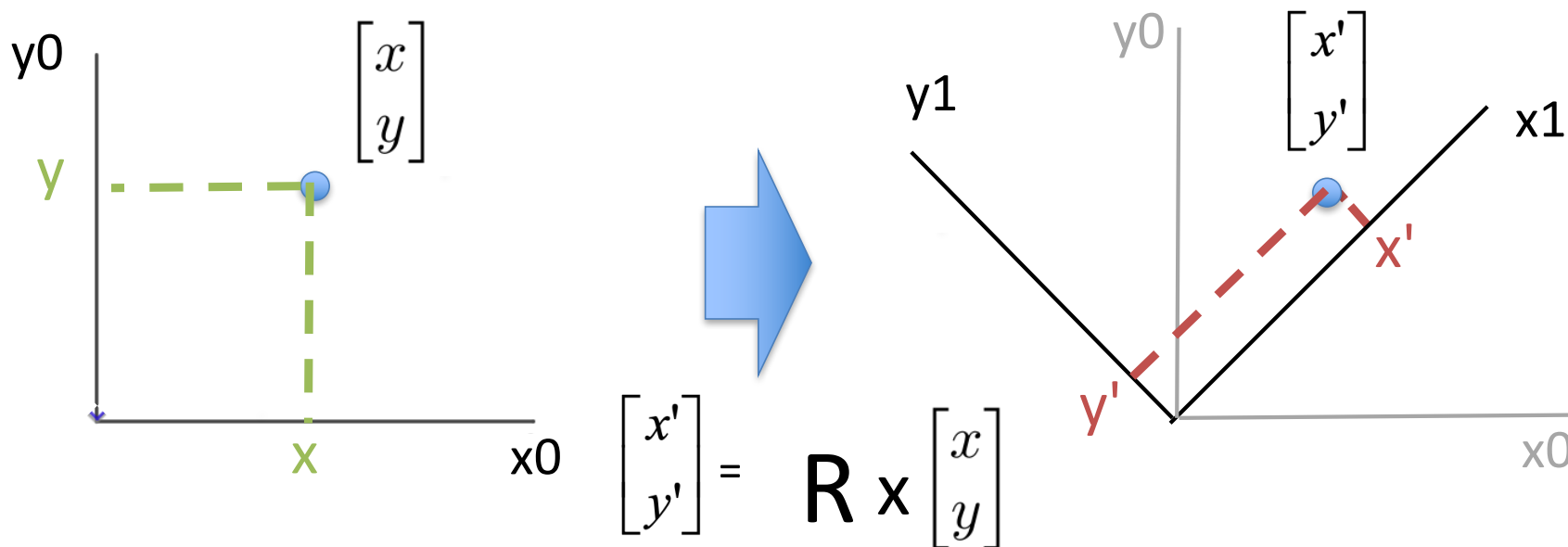
$$\circ \begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} \text{2x2 Rotation Matrix} \end{bmatrix} \times \begin{bmatrix} x \\ y \end{bmatrix}$$

$$\circ \begin{bmatrix} x' \\ y' \end{bmatrix} = \mathbf{R} \times \begin{bmatrix} x \\ y \end{bmatrix}$$

$$\mathbf{P}' = \mathbf{R} \mathbf{P}$$

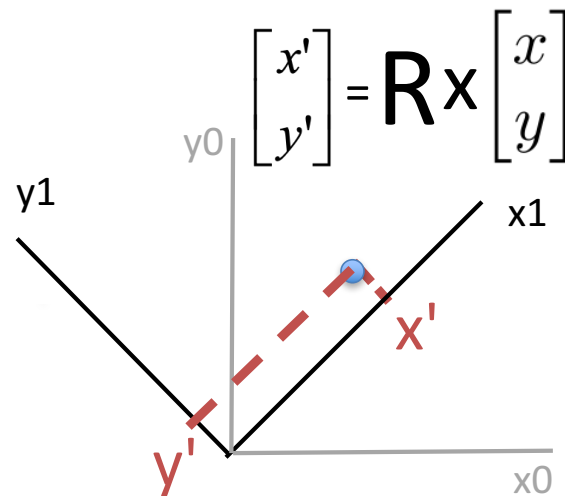
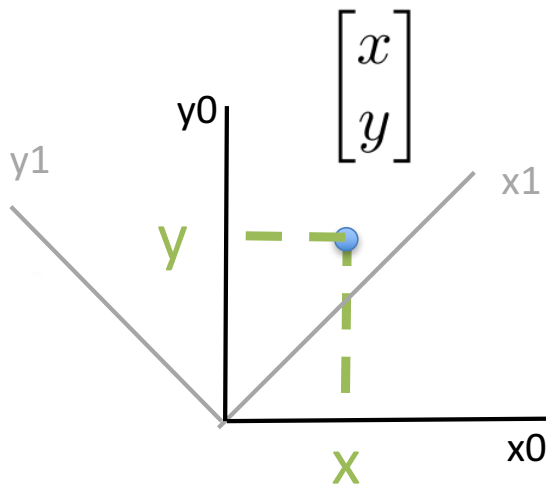
Before learning how to rotate a vector, let's understand how to do something slightly different

- How can you convert a vector represented in frame “0” to a new, rotated coordinate frame “1”?
- Answer: using a matrix-vector multiplication!!



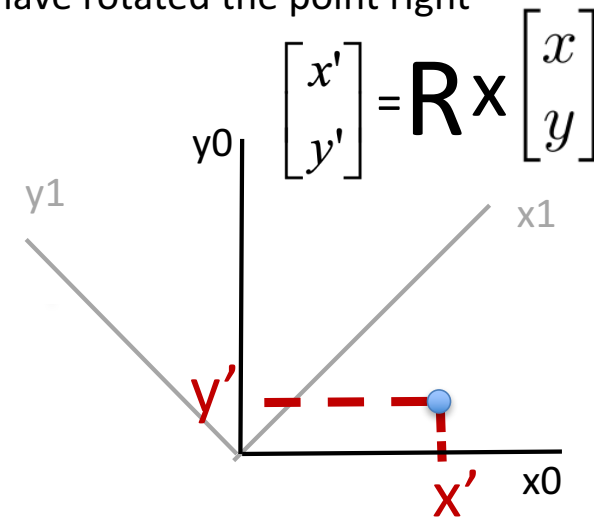
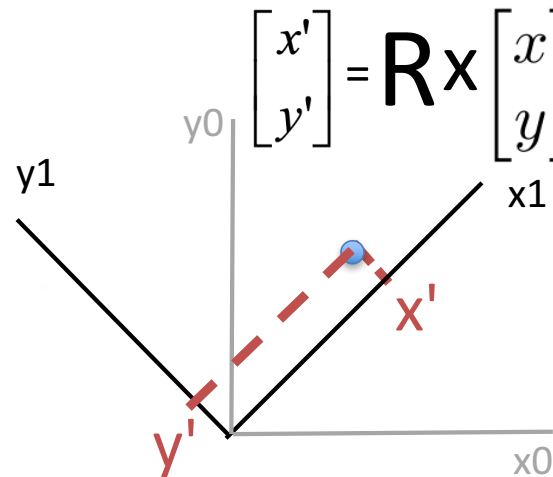
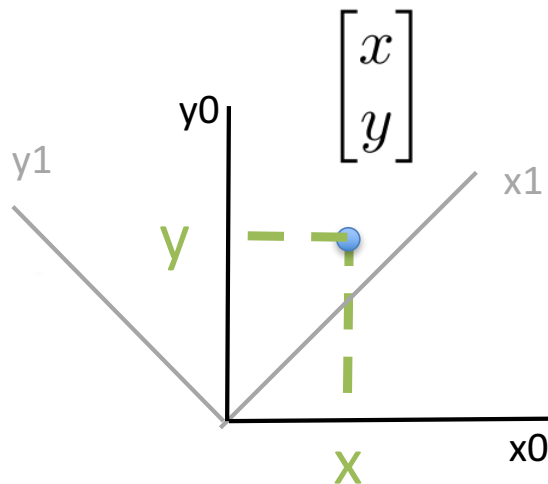
Now, here's how we do a rotation!!

- Until now, we've seen how to convert a data point represented in frame "0" to a new, rotated coordinate frame "1". We rotated the frame to the left.



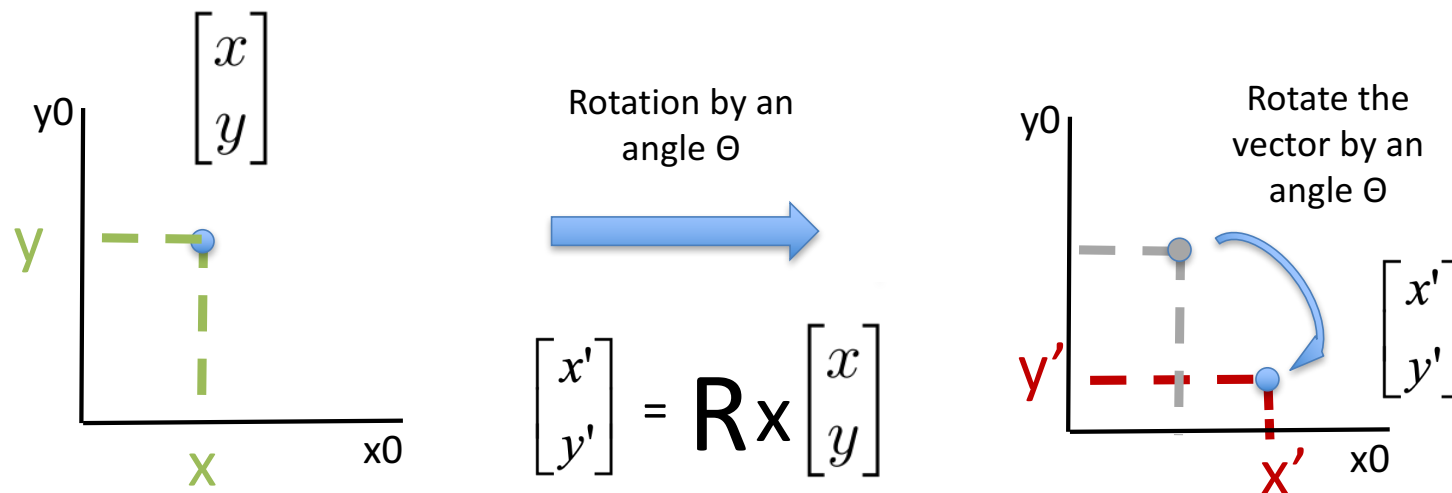
Now, here's how we do a rotation!!

- Until now, we've seen how to convert a data point represented in frame "0" to a new, rotated coordinate frame "1". We rotated the frame to the left.
- Insight: rotating the frame to the left == rotating a data point to the right
 - Suppose we express a point in the new coordinate system "1"
 - If we plot the result in the **original** coordinate system, we have rotated the point right



Now, here's how we do a rotation!!

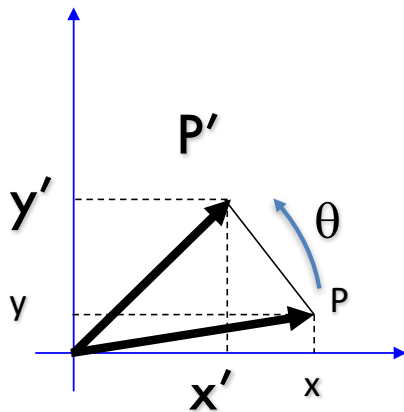
- Until now, we've seen how to convert a data point represented in frame "0" to a new, rotated coordinate frame "1".
- Insight: rotating the frame to the left == rotating a data point to the right



- Thus, rotation matrices can be used to rotate vectors. We'll usually think of them in that sense-- as operators to rotate vectors

2D Rotation Matrix Formula: what is R?

Counter-clockwise rotation by an angle θ



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

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

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

$$\mathbf{P}' = \mathbf{R} \, \mathbf{P}$$





Transformation Matrices

- Multiple transformation matrices can be used to transform a point:
 $p' = R_2 R_1 S p$
- The effect of this is to apply their transformations one after the other, from **right to left**.
- In the example above, the result is
 $(R_2 (R_1 (S p)))$
- The result is exactly the same if we multiply the matrices first, to form a single transformation matrix:
 $p' = (R_2 R_1 S) p$



Homogeneous system

- In general, a matrix multiplication lets us linearly combine components of a vector

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} \times \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} ax + by \\ cx + dy \end{bmatrix}$$

- This is sufficient for scale, rotate, skew transformations.
- But notice, we can't add a constant! ☹️



Homogeneous system

- The (somewhat hacky) solution? Stick a “1” at the end of every vector:

$$\begin{bmatrix} a & b & c \\ d & e & f \\ 0 & 0 & 1 \end{bmatrix} \times \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} ax + by + c \\ dx + ey + f \\ 1 \end{bmatrix}$$

- Now we can rotate, scale, and skew like before, **AND translate** (note how the multiplication works out, above)
- This is called “homogeneous coordinates”



Homogeneous system

- In homogeneous coordinates, the multiplication works out so the rightmost column of the matrix is a vector that gets added.

$$\begin{bmatrix} a & b & c \\ d & e & f \\ 0 & 0 & 1 \end{bmatrix} \times \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} ax + by + c \\ dx + ey + f \\ 1 \end{bmatrix}$$

- Generally, a homogeneous transformation matrix will have a bottom row of $[0 \ 0 \ 1]$, so that the result has a “1” at the bottom too.

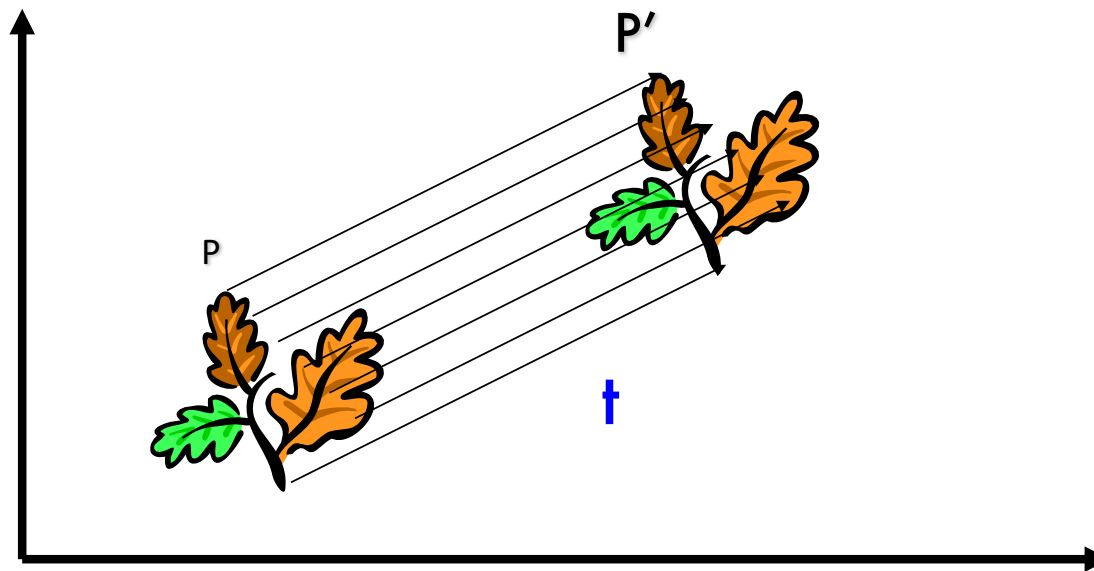


Homogeneous system

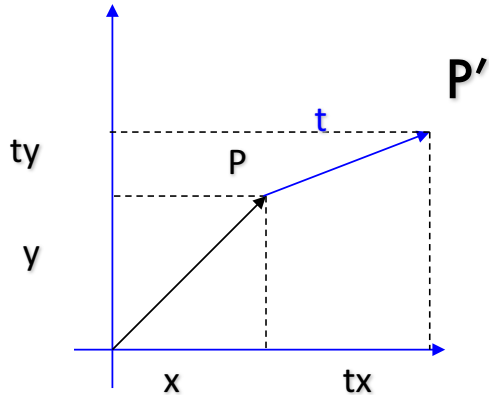
- One more thing we might want: to divide the result by something
 - For example, we may want to divide by a coordinate, to make things scale down as they get farther away in a camera image
 - Matrix multiplication can't actually divide
 - So, **by convention**, in homogeneous coordinates, we'll divide the result by its last coordinate after doing a matrix multiplication

$$\begin{bmatrix} x \\ y \\ 7 \end{bmatrix} \Rightarrow \begin{bmatrix} x/7 \\ y/7 \\ 1 \end{bmatrix}$$

2D Translation using Homogeneous Coordinates



2D Translation using Homogeneous Coordinates



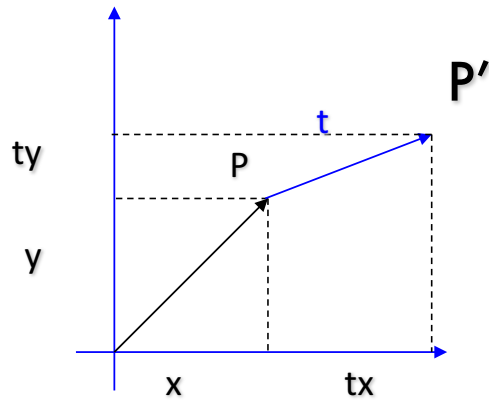
$$\mathbf{P} = (x, y) \rightarrow (x, y, 1)$$

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

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



2D Translation using Homogeneous Coordinates



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

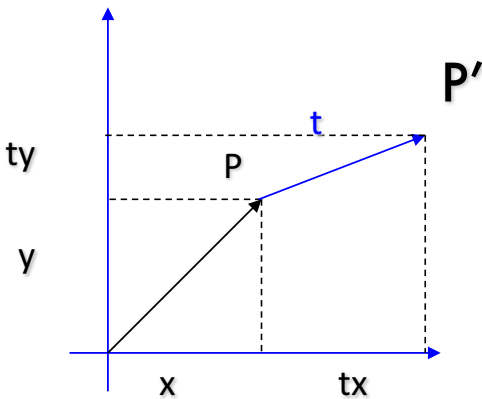
$$\mathbf{t} = (t_x, t_y) \rightarrow (t_x, t_y, 1)$$

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

\mathbf{P}



2D Translation using Homogeneous Coordinates



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

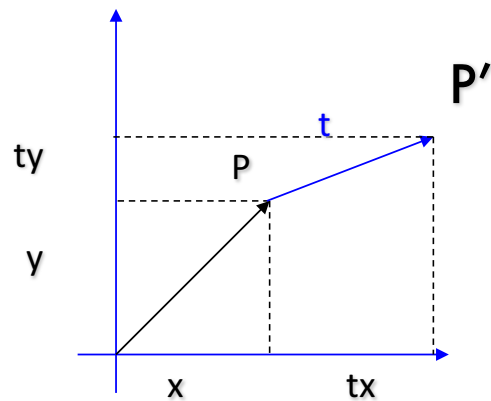
$$\mathbf{t} = (t_x, t_y) \rightarrow (t_x, t_y, 1)$$

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

\mathbf{P}



2D Translation using Homogeneous Coordinates



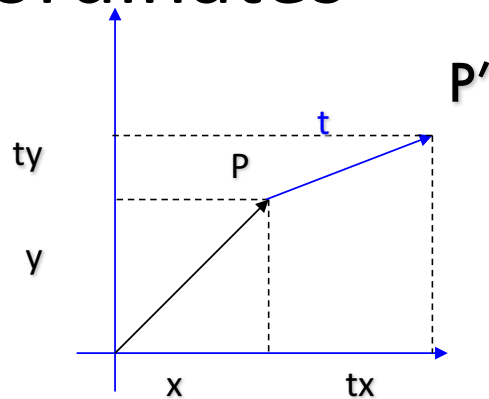
$$\mathbf{P} = (x, y) \rightarrow (x, y, 1)$$

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

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



2D Translation using Homogeneous Coordinates



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

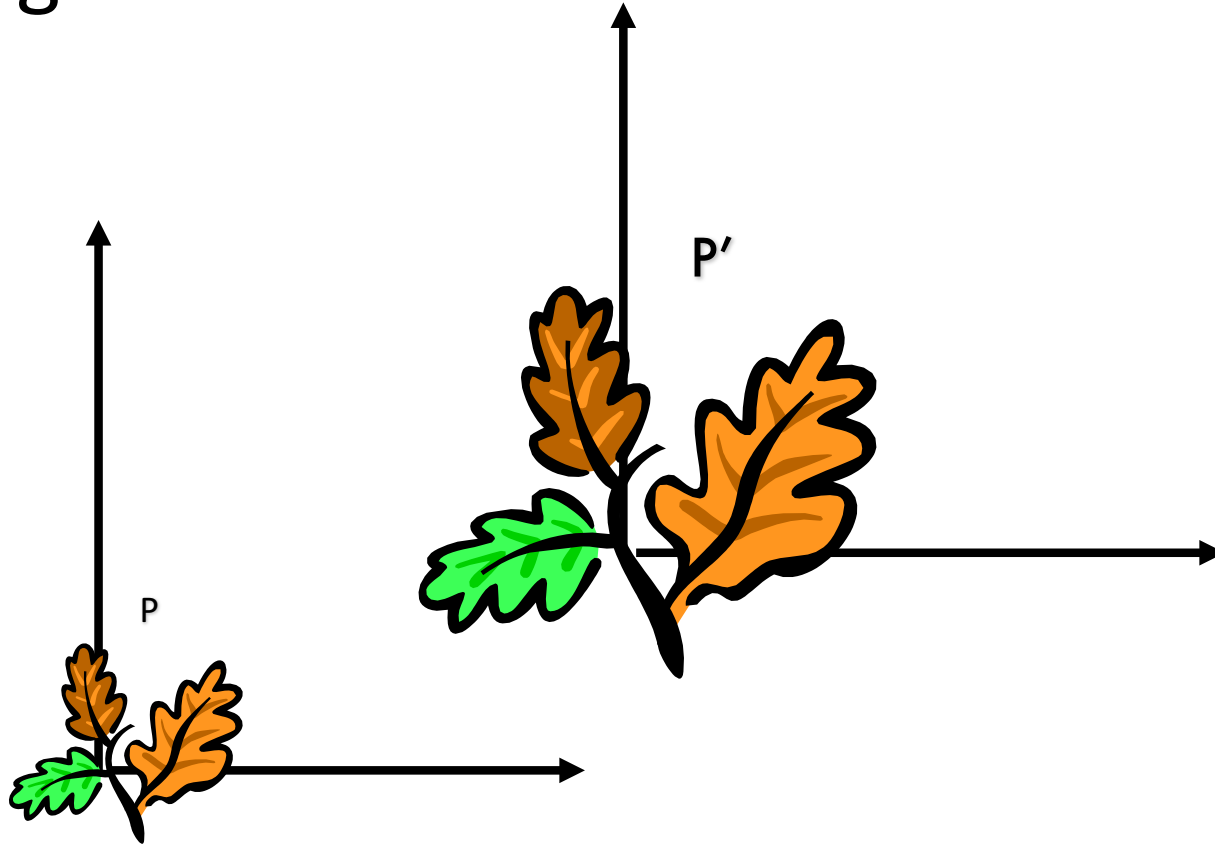
$$\mathbf{t} = (t_x, t_y) \rightarrow (t_x, t_y, 1)$$

$$\begin{aligned} \mathbf{P}' &\rightarrow \begin{bmatrix} x + t_x \\ y + t_y \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & t_x \\ 0 & 1 & t_y \\ 0 & 0 & 1 \end{bmatrix} \cdot \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} \\ &= \begin{bmatrix} \mathbf{I} & \mathbf{t} \\ 0 & 1 \end{bmatrix} \cdot \mathbf{P} = \mathbf{T} \cdot \mathbf{P} \end{aligned}$$

Diagram illustrating the matrix multiplication for 2D translation using homogeneous coordinates. The translation vector $\mathbf{t} = (t_x, t_y, 1)$ is shown as a blue dashed box, and the point $\mathbf{P} = (x, y, 1)$ is shown as a blue dashed box. Arrows point from the labels \mathbf{t} and \mathbf{P} to their respective boxes in the matrix equation.

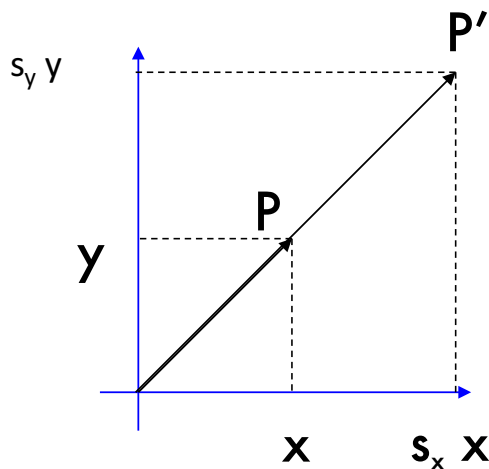


Scaling





Scaling Equation

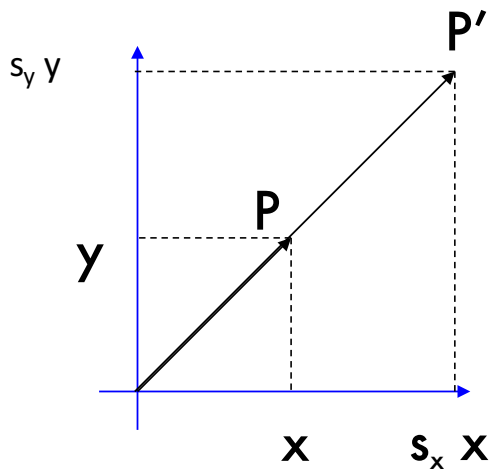


$$\mathbf{P} = (x, y) \rightarrow \mathbf{P}' = (s_x x, s_y y)$$

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

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

Scaling Equation



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

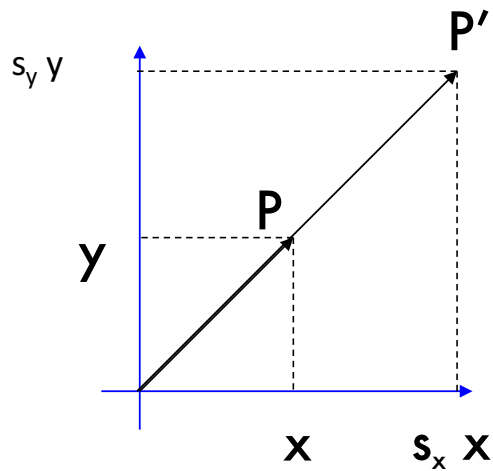
$$\mathbf{P} = (x, y) \rightarrow (x, y, 1)$$

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

$$\mathbf{P}' \rightarrow \begin{bmatrix} s_x x \\ s_y y \\ 1 \end{bmatrix} = \begin{bmatrix} \cdot \\ \cdot \\ \cdot \end{bmatrix} \cdot \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$



Scaling Equation



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

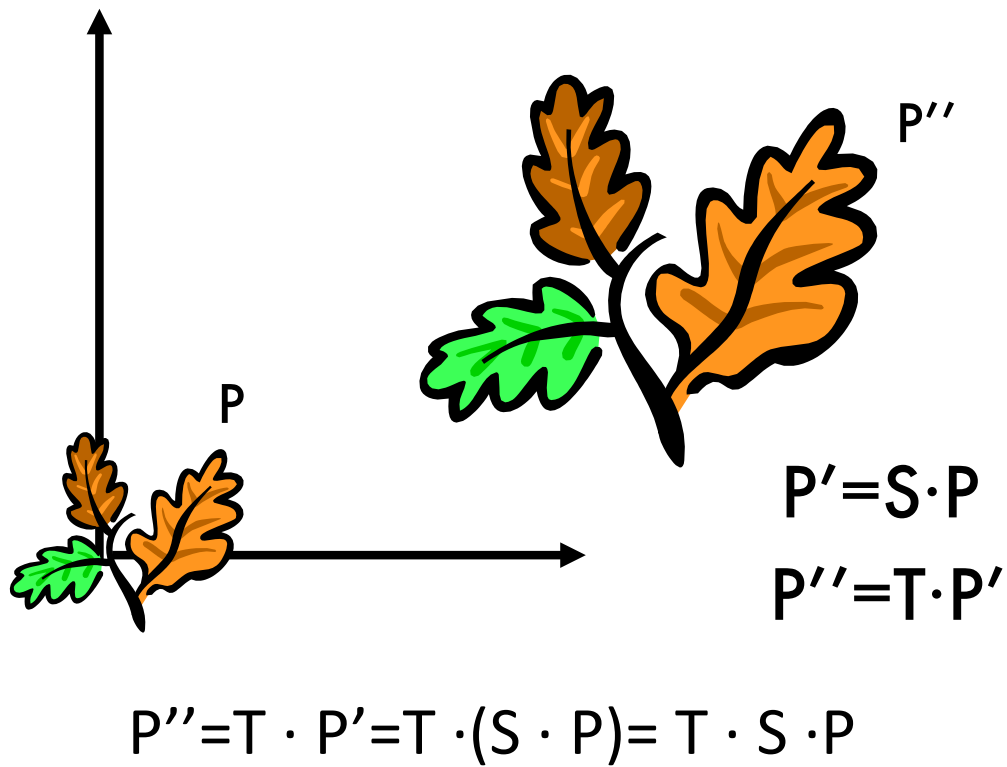
$$\mathbf{P} = (x, y) \rightarrow (x, y, 1)$$

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

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



Scaling & Translating



Scaling & Translating

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





Scaling & Translating

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

Translating & Scaling != Scaling & Translating

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



Translating & Scaling != Scaling & Translating

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

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



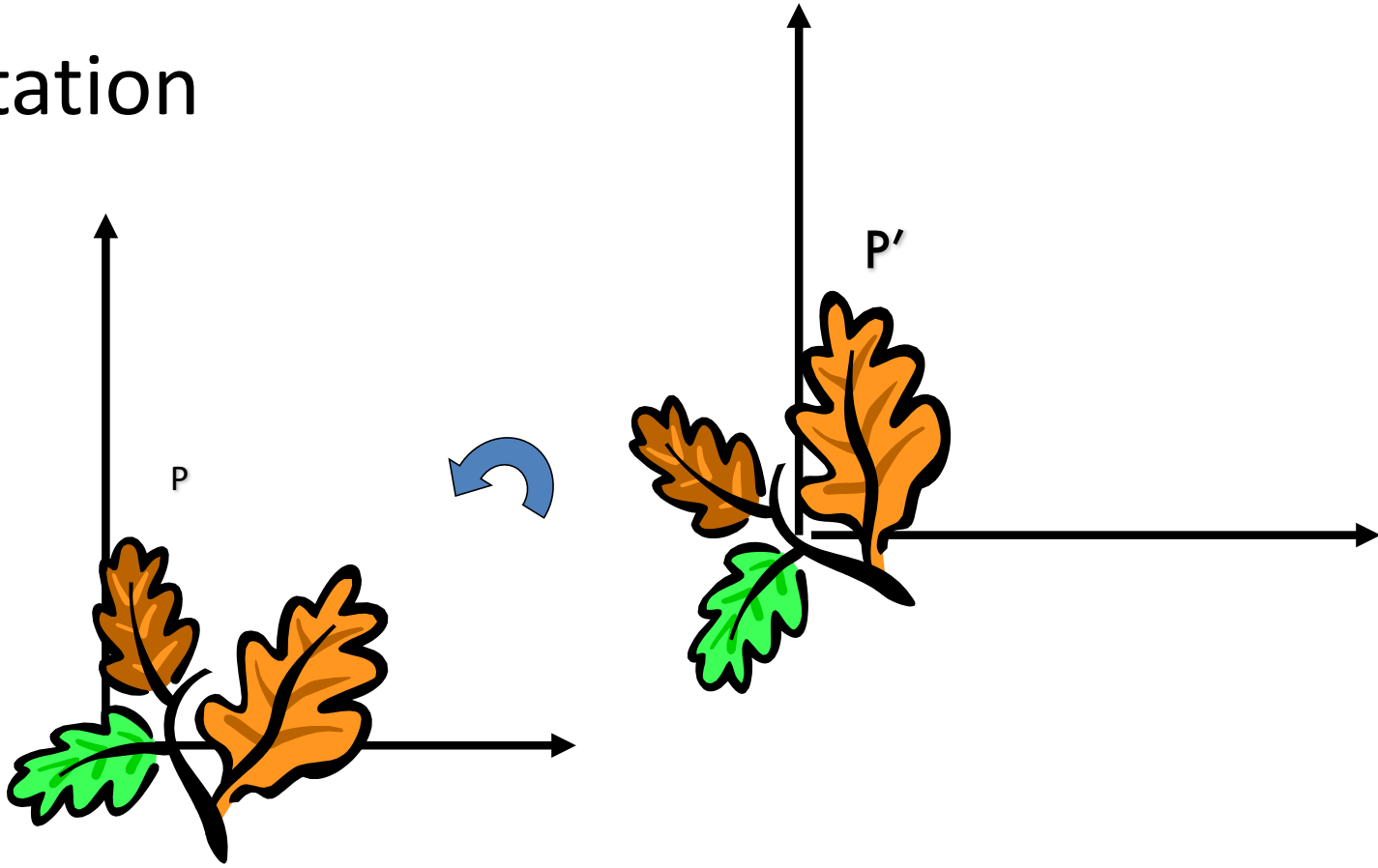


Translating & Scaling != Scaling & Translating

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

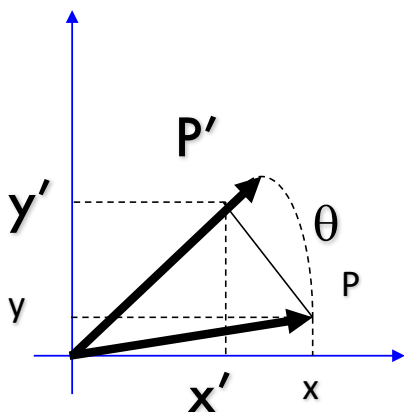
$$\begin{aligned} \mathbf{P}''' = \mathbf{S} \cdot \mathbf{T} \cdot \mathbf{P} &= \begin{bmatrix} s_x & 0 & 0 \\ 0 & s_y & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & t_x \\ 0 & 1 & t_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \\ &= \begin{bmatrix} s_x & 0 & s_x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

Rotation



Rotation Equations

Counter-clockwise rotation by an angle θ



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

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

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

$$\mathbf{P}' = \mathbf{R} \, \mathbf{P}$$





Rotation Matrix Properties

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

A 2D rotation matrix is 2x2

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

Note: \mathbf{R} belongs to the category of *normal* matrices and satisfies many interesting properties:

$$\mathbf{R} \cdot \mathbf{R}^T = \mathbf{R}^T \cdot \mathbf{R} = \mathbf{I}$$

$$\det(\mathbf{R}) = 1$$



Rotation Matrix Properties

- Transpose of a rotation matrix produces a rotation in the opposite direction

$$\mathbf{R} \cdot \mathbf{R}^T = \mathbf{R}^T \cdot \mathbf{R} = \mathbf{I}$$

$$\det(\mathbf{R}) = 1$$

- The rows of a rotation matrix are always mutually perpendicular (a.k.a. orthogonal) unit vectors
 - (and so are its columns)



Rotation Equation

$$\mathbf{P} = (x, y) \rightarrow \mathbf{P}' = (\cos\theta \cdot x - \sin\theta \cdot y, \quad \cos\theta \cdot y + \sin\theta \cdot x)$$

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

$$\begin{aligned} \mathbf{P}' &= (\cos\theta \cdot x - \sin\theta \cdot y, \cos\theta \cdot y + \sin\theta \cdot x) \\ &\rightarrow (\cos\theta \cdot x - \sin\theta \cdot y, \cos\theta \cdot y + \sin\theta \cdot x, 1) \end{aligned}$$



Rotation Equation

$$\mathbf{P} = (x, y) \rightarrow \mathbf{P}' = (\cos(\theta)x - \sin(\theta)y, \cos(\theta)y + \sin(\theta)x)$$

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

$$\begin{aligned}\mathbf{P}' &= (\cos(\theta)x - \sin(\theta)y, \cos(\theta)y + \sin(\theta)x) \\ &\rightarrow (\cos(\theta)x - \sin(\theta)y, \cos(\theta)y + \sin(\theta)x, 1)\end{aligned}$$

$$\begin{aligned}\mathbf{P}' \rightarrow \begin{bmatrix} \cos(\theta)x - \sin(\theta)y \\ \cos(\theta)y + \sin(\theta)x \\ 1 \end{bmatrix} &= \begin{bmatrix} \cos(\theta) & -\sin(\theta) & 0 \\ \sin(\theta) & \cos(\theta) & 0 \\ 0 & 0 & 1 \end{bmatrix} \cdot \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} \\ &= \begin{bmatrix} \mathbf{R}' & \mathbf{0} \\ \mathbf{0} & \mathbf{1} \end{bmatrix} \cdot \mathbf{P} = \mathbf{R} \cdot \mathbf{P}\end{aligned}$$



Scaling + Rotation + Translation

$$\mathbf{P}' = (\mathbf{T} \mathbf{R} \mathbf{S}) \mathbf{P}$$

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

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

This is the form of the
general-purpose
transformation matrix

$$= \begin{bmatrix} R & t \\ 0 & 1 \end{bmatrix} \begin{bmatrix} S & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \boxed{\begin{bmatrix} R & S & t \\ 0 & 1 \end{bmatrix}} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$