

Linear Algebra and SVD

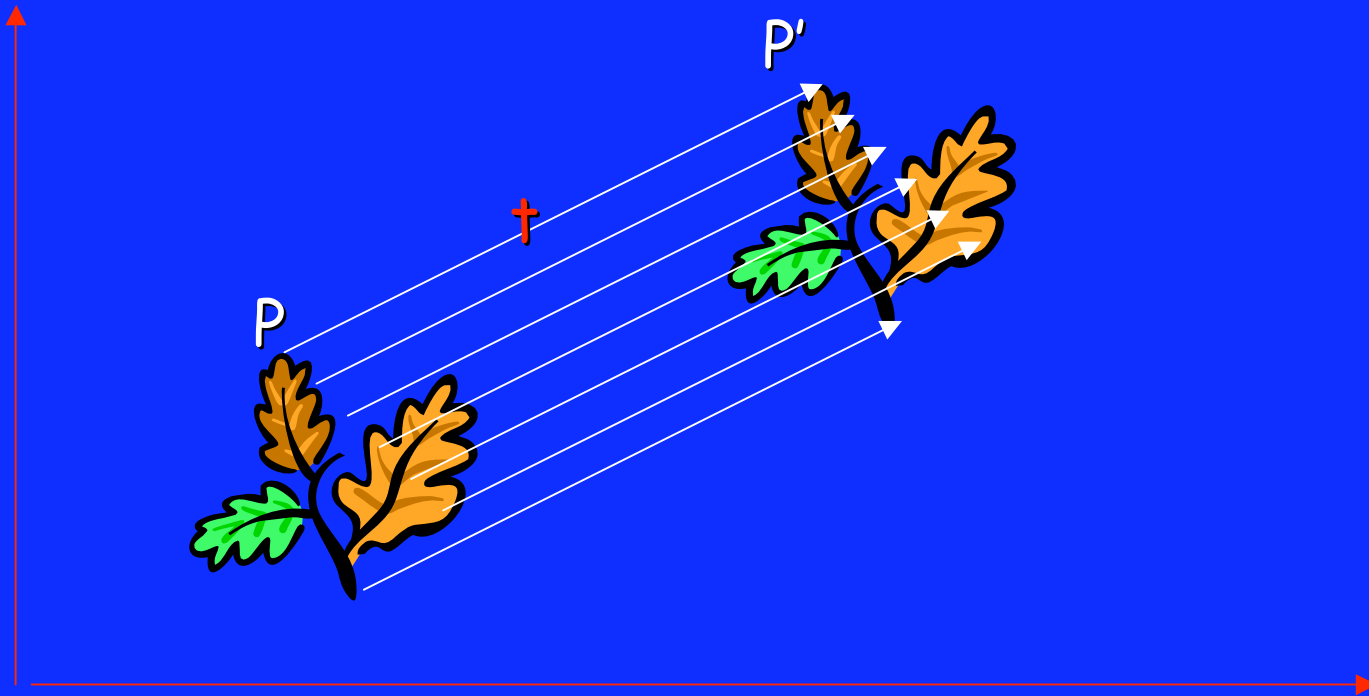
(Some slides adapted from Octavia Camps)

Goals

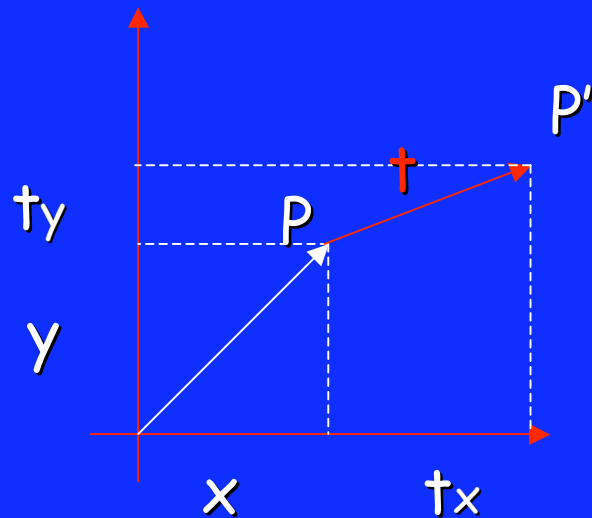
- Represent points as column vectors.
- Represent motion as matrices.
- Move geometric objects with matrix multiplication.
- Introduce SVD

Euclidean transformations

2D Translation



2D Translation Equation

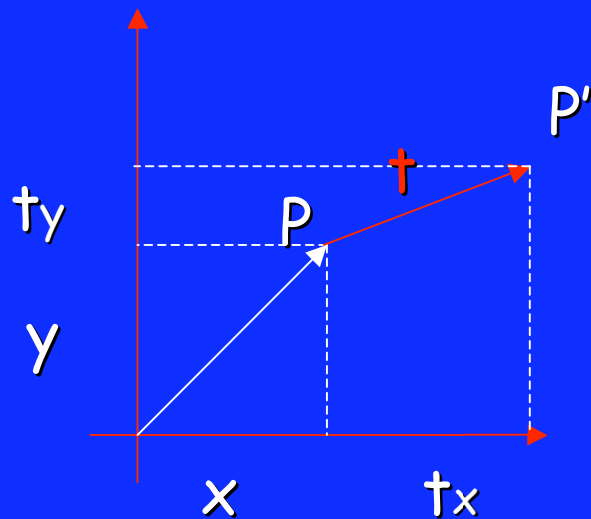


$$\mathbf{P} = (x, y)$$

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

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

2D Translation using Matrices



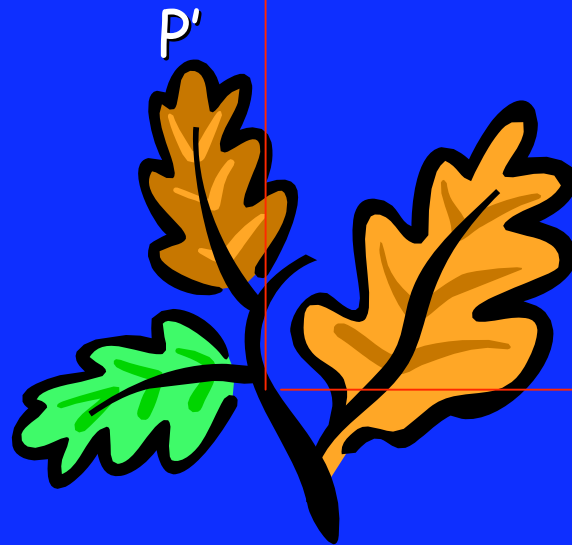
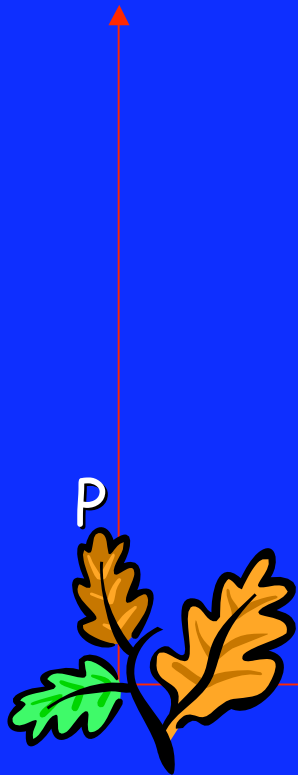
$$\mathbf{P} = (x, y)$$

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

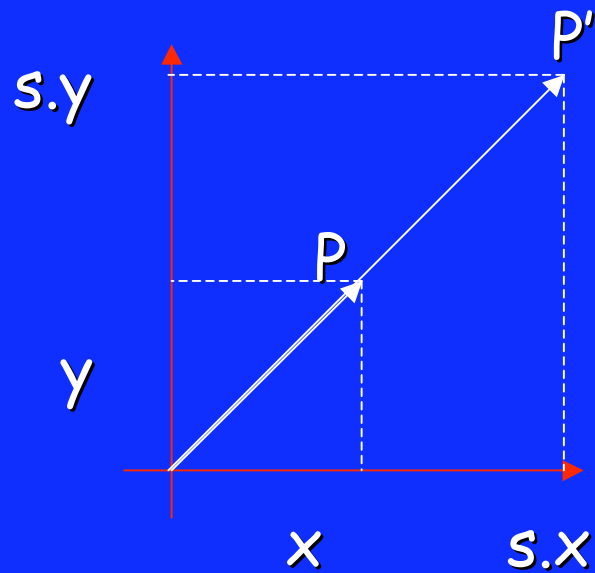
$$\mathbf{P}' \rightarrow \begin{bmatrix} x + t_x \\ y + t_y \end{bmatrix} = \begin{bmatrix} 1 & 0 & t_x \\ 0 & 1 & t_y \end{bmatrix} \cdot \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

The diagram shows the matrix multiplication for 2D translation. The translation vector \mathbf{t} is highlighted in a red box, and the point \mathbf{P} is highlighted in a yellow box. The homogeneous coordinate 1 is highlighted in a red circle with a red arrow pointing to it.

Scaling



Scaling Equation



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

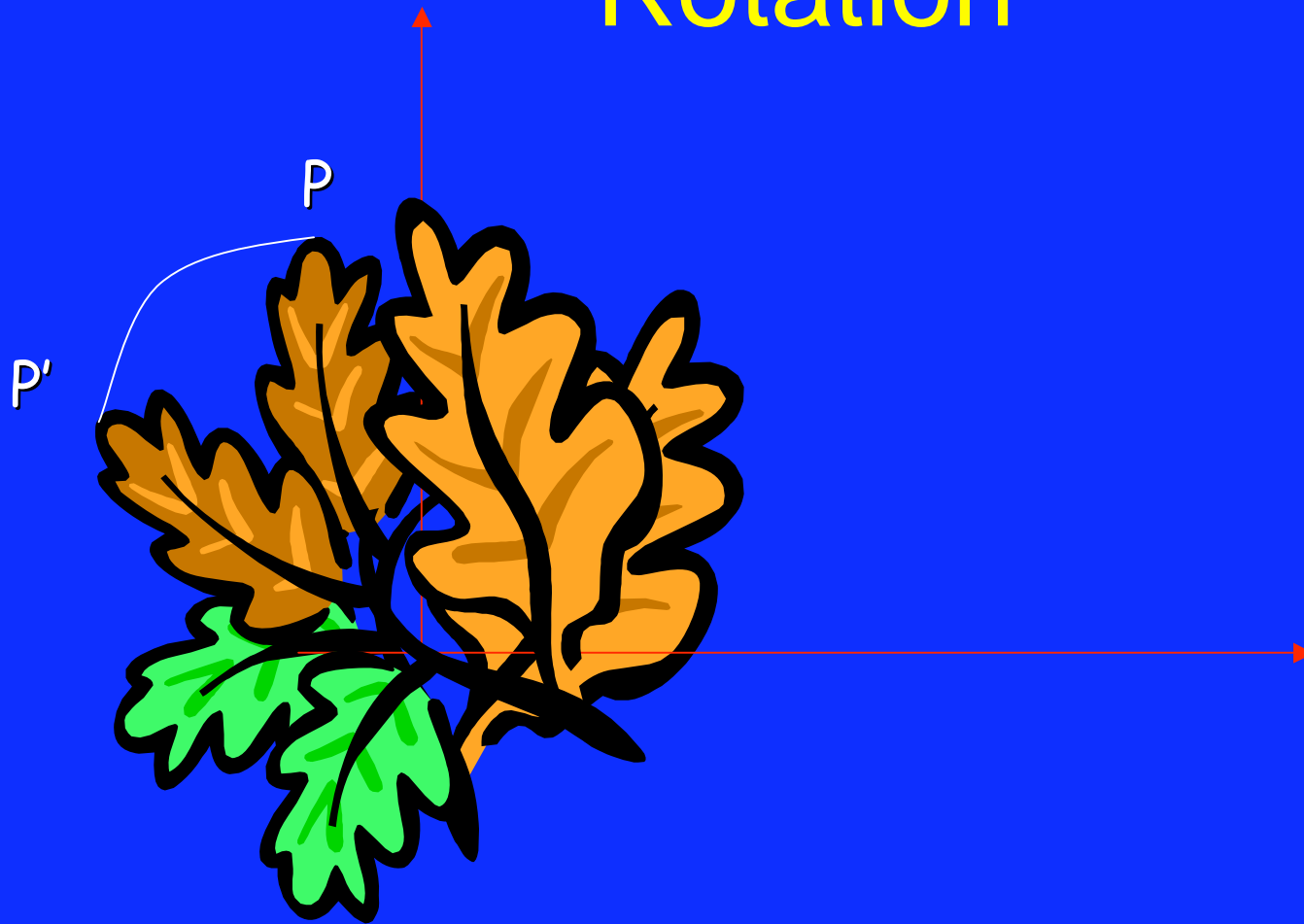
$$\mathbf{P}' = (sx, sy)$$

$$\mathbf{P}' = s \cdot \mathbf{P}$$

$$\mathbf{P}' \rightarrow \begin{bmatrix} sx \\ sy \end{bmatrix} = \underbrace{\begin{bmatrix} s & 0 \\ 0 & s \end{bmatrix}}_{\mathbf{S}} \cdot \begin{bmatrix} x \\ y \end{bmatrix}$$

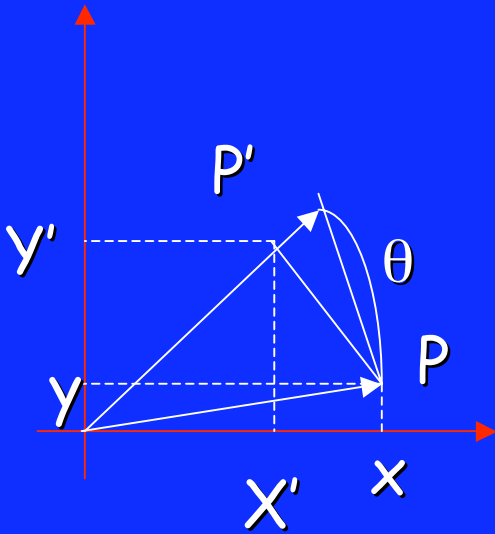
$$\mathbf{P}' = \mathbf{S} \cdot \mathbf{P}$$

Rotation



Rotation Equations

Counter-clockwise rotation by an angle θ



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

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

Why does multiplying points by R rotate them?

- Think of the rows of R as a new coordinate system. Taking inner products of each point with these expresses that point in that coordinate system.
 - This means rows of R must be orthonormal vectors (orthogonal unit vectors).
- Think of what happens to the points $(1,0)$ and $(0,1)$. They go to $(\cos \theta, -\sin \theta)$, and $(\sin \theta, \cos \theta)$. They remain orthonormal, and rotate clockwise by θ .
 - Any other point, (a,b) can be thought of as $a(1,0) + b(0,1)$. $R(a(1,0)+b(0,1)) = Ra(1,0) + Ra(0,1) = aR(1,0) + bR(0,1)$. So it's in the same position relative to the rotated coordinates that it was in before rotation relative to the x, y coordinates. That is, it's rotated.

Degrees of Freedom

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

\mathbf{R} is 2x2 \longrightarrow 4 elements

BUT! There is only 1 degree of freedom: θ

The 4 elements must satisfy the following constraints:

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

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

Transformations can be composed

- Matrix multiplication is associative.
- Combine series of transformations into one matrix. (*example, whiteboard*).
- In general, the order matters. (*example, whiteboard*).
- 2D Rotations can be interchanged.
Why?

Rotation and Translation

$$\begin{pmatrix} \cos q & -\sin q & tx \\ \sin q & \cos q & ty \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \\ 1 \end{pmatrix}$$

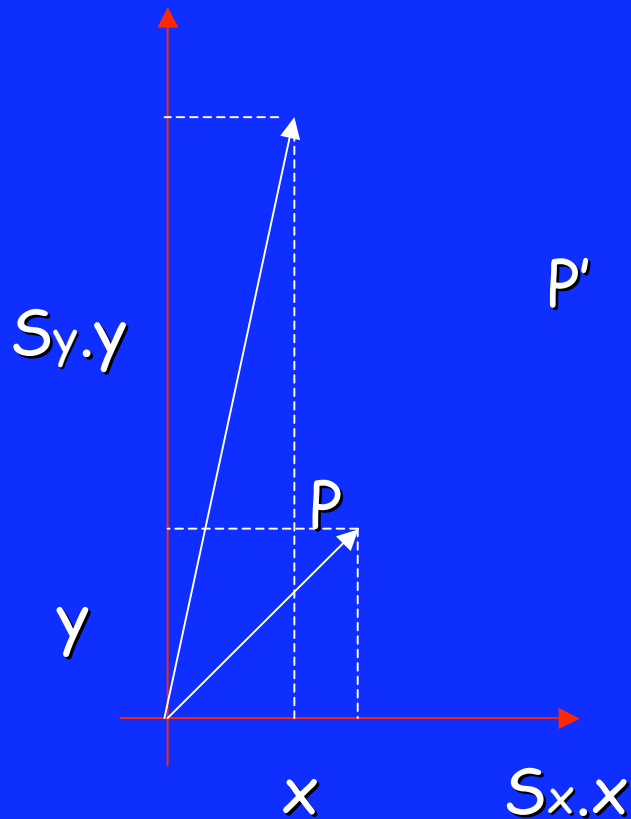
Rotation, Scaling and Translation

$$\begin{pmatrix} a & -b & tx \\ b & a & ty \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \\ 1 \end{pmatrix}$$

Rotation about an arbitrary point

- Can translate to origin, rotate, translate back. (*example, whiteboard*).
- This is also rotation with one translation.
 - Intuitively, amount of rotation is same either way.
 - But a translation is added.

Stretching Equation



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

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

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

$$\mathbf{P}' = \mathbf{S} \cdot \mathbf{P}$$

Stretching = tilting and projecting (with weak perspective)

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

SVD

- For any matrix, $M = R_1DR_2$
 - R_1, R_2 are rotation matrices
 - D is a diagonal matrix.
 - This decomposition is unique.
 - Efficient algorithms can compute this (in matlab, `svd`).

Linear Transformation

$$\mathbf{P}' \rightarrow \begin{bmatrix} a & b \\ c & d \end{bmatrix} \cdot \begin{bmatrix} x \\ y \end{bmatrix} \quad \text{SVD}$$

$$= \begin{bmatrix} \sin \theta & \cos \theta \\ -\cos \theta & \sin \theta \end{bmatrix} \begin{bmatrix} s_x & 0 \\ 0 & s_y \end{bmatrix} \begin{bmatrix} \sin \varphi & \cos \varphi \\ -\cos \varphi & \sin \varphi \end{bmatrix} \cdot \begin{bmatrix} x \\ y \end{bmatrix}$$

$$= s_y \begin{bmatrix} \sin \theta & \cos \theta \\ -\cos \theta & \sin \theta \end{bmatrix} \begin{bmatrix} \frac{s_x}{s_y} & 0 \\ 0 & 1 \end{bmatrix} \cdot \begin{bmatrix} \sin \varphi & \cos \varphi \\ -\cos \varphi & \sin \varphi \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

Affine Transformation

$$\mathbf{P}' \rightarrow \begin{bmatrix} a & b & tx \\ c & d & ty \end{bmatrix} \cdot \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

Simple 3D Rotation

$$\begin{pmatrix} \cos \theta & \sin \theta & 0 \\ -\sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 & x_2 & \cdot & \cdot & \cdot & x_n \\ y_1 & y_2 & & & & y_n \\ z_1 & z_2 & & & & z_n \end{pmatrix}$$

Rotation about z axis.

Rotates x,y coordinates. Leaves z coordinates fixed.

Full 3D Rotation

$$R = \begin{pmatrix} \cos \theta & \sin \theta & 0 \\ -\sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} \cos \beta & 0 & \sin \beta \\ 0 & 1 & 0 \\ -\sin \beta & 0 & \cos \beta \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & \cos \alpha & \sin \alpha \\ 0 & -\sin \alpha & \cos \alpha \end{pmatrix}$$

- Any rotation can be expressed as combination of three rotations about three axes.

$$RR^T = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

- Rows (and columns) of R are orthonormal vectors.
- R has determinant 1 (not -1).

- Intuitively, it makes sense that 3D rotations can be expressed as 3 separate rotations about fixed axes. Rotations have 3 degrees of freedom; two describe an axis of rotation, and one the amount.
- Rotations preserve the length of a vector, and the angle between two vectors. Therefore, $(1,0,0)$, $(0,1,0)$, $(0,0,1)$ must be orthonormal after rotation. After rotation, they are the three columns of R . So these columns must be orthonormal vectors for R to be a rotation. Similarly, if they are orthonormal vectors (with determinant 1) R will have the effect of rotating $(1,0,0)$, $(0,1,0)$, $(0,0,1)$. Same reasoning as 2D tells us all other points rotate too.
 - Note if R has determinant -1 , then R is a rotation plus a reflection.

3D Rotation + Translation

- Just like 2D case