

Spatial Transformations

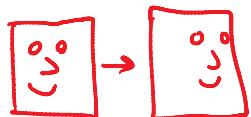
Goal: To learn how to represent affine transformations in an efficient way.

Images (and volumes) can be thought of as fields (or functions) of space. We will talk a lot about transformations of these spatial domains.

Some examples are: translation, rotation, scaling, etc.

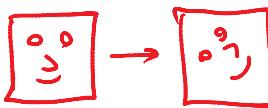
Translation

$$\begin{cases} u = x + a \\ v = y + b \end{cases}$$



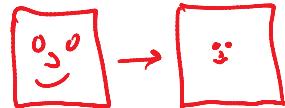
Rotation

$$\begin{cases} u = x \cos \theta + y \sin \theta \\ v = -x \sin \theta + y \cos \theta \end{cases}$$



Scaling

$$\begin{cases} u = s_1 x \\ v = s_2 y \end{cases}$$



Homogeneous Coordinates

Notice above that rotation & scaling can each be written as matrix-vector products.

$$\vec{u} = \begin{bmatrix} \cos \theta & \sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \vec{x}$$

Rotation

$$\vec{u} = \begin{bmatrix} s_1 & 0 \\ 0 & s_2 \end{bmatrix} \vec{x}$$

Scaling

So called
"affine" transforms.

However, translation cannot. But we can embed translations into a matrix-vector product if we bump up the dimension by 1.

so Γ

$\Gamma \vec{x}$

Γ_{trans}

we bump up the dimension by 1.

i.e. $\begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} \Leftrightarrow \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} \Leftrightarrow \begin{bmatrix} rx \\ ry \\ rz \\ r \end{bmatrix}$ for $r \neq 0$

We call this augmented coordinate system
"Homogeneous coordinates".

Consider the transform

$$\begin{bmatrix} u \\ v \\ r \end{bmatrix} = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} + \begin{bmatrix} e \\ f \\ 1 \end{bmatrix} \quad \text{in non-homogeneous coords}$$
$$\Rightarrow \begin{bmatrix} u \\ v \\ r \end{bmatrix} = \begin{bmatrix} a & b & e \\ c & d & f \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} ax + by + e \\ cx + dy + f \\ 1 \end{bmatrix}$$

Example:

Rotate by 30° , then translate by $\begin{bmatrix} 2 \\ 1 \end{bmatrix}$

Rotation matrix is

$$R_{30^\circ} = \begin{bmatrix} \frac{\sqrt{3}}{2} & -\frac{1}{2} \\ \frac{1}{2} & \frac{\sqrt{3}}{2} \end{bmatrix}$$

Thus, $P = \begin{bmatrix} \frac{\sqrt{3}}{2} & -\frac{1}{2} & 2 \\ \frac{1}{2} & \frac{\sqrt{3}}{2} & 1 \\ 0 & 0 & 1 \end{bmatrix}$

Composition of Transformations

Why do we care about representing translation as a matrix-vector product?

Because we can do this

a affine transform:

'Cuz we can do this...

→ Two (or more) affine transforms can be combined into one affine transform by simple matrix multiplication.

Consider the composite transform:

1) $P_1 \equiv$ rotate by 30°

2) $P_2 \equiv$ translate by $\begin{bmatrix} 2 \\ 1 \end{bmatrix}$

$$P_1 = \begin{bmatrix} \cos 30 & \sin 30 & 0 \\ -\sin 30 & \cos 30 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad P_2 = \begin{bmatrix} 1 & 0 & 2 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix}$$

Composite transform is

$$\begin{aligned} P_2 P_1 &= \begin{bmatrix} 1 & 0 & 2 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \cos 30 & \sin 30 & 0 \\ -\sin 30 & \cos 30 & 0 \\ 0 & 0 & 1 \end{bmatrix} \\ &= \begin{bmatrix} \cos 30 & \sin 30 & 2 \\ \sin 30 & \cos 30 & 1 \\ 0 & 0 & 1 \end{bmatrix} \end{aligned}$$

Now suppose we add a 3rd transform to the sequence:

3) $P_3 \equiv$ rotate by 45° , then translate by $\begin{bmatrix} -1 \\ 1 \end{bmatrix}$

$$P_3 = \begin{bmatrix} \cos 45 & \sin 45 & -1 \\ -\sin 45 & \cos 45 & -1 \\ 0 & 0 & 1 \end{bmatrix}$$

The transform resulting from the sequential composition of all 3 can be written

The transform resulting from the sequential composition of all 3 can be written

$$P = P_3 P_2 P_1$$

[Pause... Matlab exercise L03_transforms.m]

→ your end result should be

`P_all_three =`

0.2588	0.9659	1.1213
-0.9659	0.2588	0.2929
0	0	1.0000

Decomposing Affine Transforms

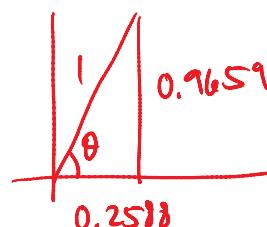
We can decompose a transformation matrix into a single rotation & translation (& scale, skew, etc.)

In the above case

$$\begin{bmatrix} \cos\theta & \sin\theta & t_1 \\ -\sin\theta & \cos\theta & t_2 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 0.2588 & 0.9659 & 1.1213 \\ -0.9659 & 0.2588 & 0.2929 \\ 0 & 0 & 1.0000 \end{bmatrix}$$

$$\Rightarrow \cos\theta = 0.2588$$

$$\sin\theta = 0.9659$$



$$\Rightarrow \theta = 75^\circ$$

$$t_1 = 1.1213$$

$$t_2 = -1.7071$$

Thus, $P_3 P_2 P_1$ is equivalent to

- Rotation by 75° , followed by
- Translation of $\begin{bmatrix} 1.1213 \\ -1.7071 \end{bmatrix}$

It is often helpful to think of these transforms

It is often helpful to think of these transforms in terms of their affine part & translation part.

$$P = \begin{bmatrix} A & T \\ \hline 0 & 1 \end{bmatrix} \quad A \in \mathbb{R}^{d \times d} \quad d=2 \text{ for 2D}$$

$$T \in \mathbb{R}^{d \times 1} \quad d=3 \text{ for 3D}$$

etc.

Combining Simple Transforms

Translations:

$$\begin{bmatrix} I & T_1 \\ \hline 0 & 1 \end{bmatrix} \begin{bmatrix} I & T_2 \\ \hline 0 & 1 \end{bmatrix} = \begin{bmatrix} I & T_1 + T_2 \\ \hline 0 & 1 \end{bmatrix}$$

Thus, combining translations simply adds the translations.

Inverse of transl. T_i is $-T_i$

Rotations:

$$\begin{bmatrix} R_0 & 0 \\ \hline 0 & 1 \end{bmatrix} \begin{bmatrix} R_{02} & 0 \\ \hline 0 & 1 \end{bmatrix} = \begin{bmatrix} R_0 R_{02} & 0 \\ \hline 0 & 1 \end{bmatrix} = \begin{bmatrix} R_{0+02} & 0 \\ \hline 0 & 1 \end{bmatrix}$$

Thus, combining rotations simply adds their angles.

Inverse of rotation by θ is rotation by $-\theta$.

Scaling:

$$\begin{bmatrix} S_1 & 0 \\ \hline 0 & 1 \end{bmatrix} \begin{bmatrix} S_2 & 0 \\ \hline 0 & 1 \end{bmatrix} = \begin{bmatrix} S_1 S_2 & 0 \\ \hline 0 & 1 \end{bmatrix}$$

Thus, combining scales simply multiplies them.

Inverse of scaling by S is scaling by $\frac{1}{S}$

Inverse of scaling by S is scaling by $\frac{1}{S}$.

CAUTION: Be aware of the complexities that arise when you are combining different kinds of simple transforms.

Example: transl. by $\begin{bmatrix} 2 \\ 1 \end{bmatrix}$, rotate by 30° , transl. by $\begin{bmatrix} -2 \\ -1 \end{bmatrix}$.

$$P = \begin{bmatrix} I & -T \\ 0 & 1 \end{bmatrix} \begin{bmatrix} R & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} I & T \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} R & RT-T \\ 0 & 1 \end{bmatrix}$$

Notice that the translations do not undo each other because there is a rotation between them.

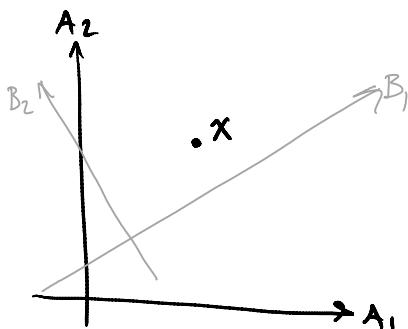
Converting Between Coordinate Systems

Suppose we have a point x and two different coordinate systems (CS) labeled A and B.

x_A ≡ coords of x w.r.t. CS A

x_B ≡ coords of x w.r.t. CS B

e.g.



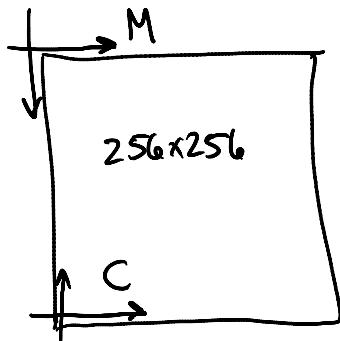
We want to be able to convert x_A to x_B (and vice versa).

$$P_{A \rightarrow B} x_A = x_B$$

Magic Rule: To get the transform $P_{A \rightarrow B}$, we can simply use the same transform that moves the B axes onto the A axes, using A's coord. system.

That is, suppose the B axes is an object in A's coordinate space. Compose the transform that will overlay the B axes on the A axes.

Example: Cartesian to Matlab index coords.



To convert from x_c (Cartesian) to x_m (Matlab index), we move M axes onto C.

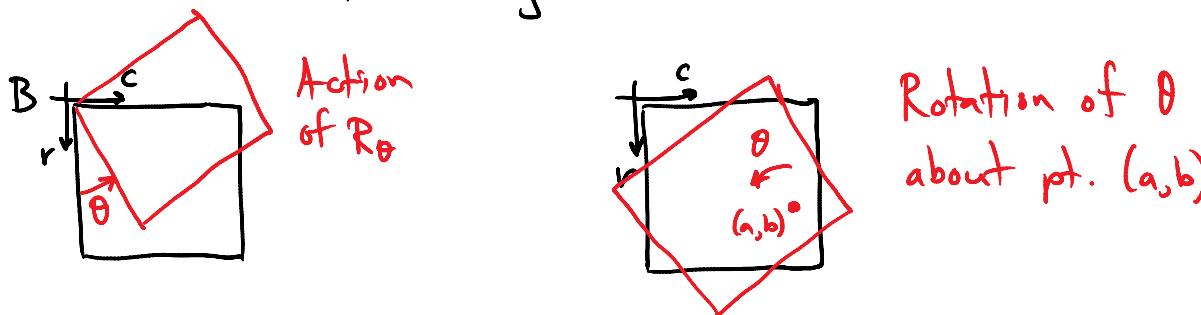
- 1) shift M by $\begin{bmatrix} 1 \\ -256 \end{bmatrix}$ $\rightarrow P_1 = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & -256 \\ 0 & 0 & 1 \end{bmatrix}$
- 2) reverse M's vertical axis $\rightarrow P_2 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$
- 3) swap M's axes $\rightarrow P_3 = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$

$$P_{C \rightarrow M} = P_3 P_2 P_1 = \begin{bmatrix} 0 & -1 & 256 \\ 1 & 0 & 1 \\ 0 & 0 & 1 \end{bmatrix}$$

Check: $\begin{bmatrix} 0 \\ 0 \end{bmatrix}_c \rightarrow \begin{bmatrix} 256 \\ 1 \end{bmatrix}_m \quad \begin{bmatrix} 255 \\ 255 \end{bmatrix}_c \rightarrow \begin{bmatrix} 1 \\ 256 \end{bmatrix}$

Converting Transforms Between Coordinate Systems

Suppose you want to rotate an image around a specified point with coords (a, b) . You already know how to build a rotation matrix, but it rotates about the origin.



The Theory

Given two coord systems A and B, and a transform matrix M_A , expressed w.r.t. CS A, find the equivalent transform matrix w.r.t. CS B, M_B .

Let P_{AB} be the change-of-basis matrix that converts from A coords to B coords.

$$\text{i.e. } X_B = P_{AB} X_A \quad \text{eg. } \begin{array}{c} a_1 \\ a_2 \\ b_1 \\ b_2 \end{array} \quad X_A = [1 \ 2]^T \quad X_B = [0 \ \sqrt{2}]^T$$

$$\text{Note that } P_{AB}^{-1} = P_{BA}$$

$$\text{Thus, } X_B = P_{AB} X_A \text{ and } X_A = P_{BA} X_B .$$

The transform M_A is applied to vectors in CS A.

i.e.

$$y_A = M_A X_A$$

$$P_{BA} y_B = M_A P_{BA} x_B \Rightarrow y_B = \underbrace{P_{AB}}_{M_B} M_A P_{BA} x_B$$

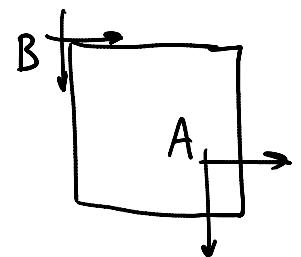
$$y_B = M_B x_B$$

Hence, $M_B = P_{AB} M_A P_{BA} (= P_{BA}^{-1} M_A P_{BA})$

Back to our rotation example :

$$M_A = R_\theta \text{ (rotation about A's origin)}$$

$$P_{AB} = T(a, b) \text{ (translation by } (a, b))$$



Hence, $M_B = T(a, b) R_\theta T(-a, -b)$

