

# Geometric transformations

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# Notation

There are a number of important and useful geometric transformations of the plane that can be applied to images. Image transformations are implemented in the same way as subsampling: by scanning the pixels of the target and modifying them using interpolates of pixels from the source. This means it is important that transformations are invertible. Adopt the convention that a point  $\mathbf{x} = (x, y)$  is mapped by a transformation to the point  $\mathbf{u} = (u, v) = (u(x, y), v(x, y))$ , and  $\mathbf{u} = (u, v)$  is mapped to  $\mathbf{x} = (x, y)$  by the inverse. In vector notation,  $\mathbf{x}$  is mapped to  $\mathbf{u}$ , and so on. Write  $\mathbf{A}$  for a  $2 \times 2$  matrix, whose  $i, j$ 'th component is  $a_{ij}$ .

# Translation

## **Definition: 4.1** *Translation*

**Translation** maps the point  $(x, y)$  to the point  $(u, v) = (x + t_x, y + t_y)$  for two constants  $t_x$  and  $t_y$ . Here  $(x, y) = (u - t_x, v - t_y)$ . In vector notation,

$$\mathbf{u} = \mathbf{x} + \mathbf{t} \text{ and } \mathbf{x} = \mathbf{u} - \mathbf{t}.$$

**Useful Fact:** *Translation preserves lengths and angles. Choose two points  $\mathbf{x}_1$  and  $\mathbf{x}_2$ . The squared distance from  $\mathbf{x}_1$  to  $\mathbf{x}_2$  is  $(\mathbf{x}_1 - \mathbf{x}_2)^T (\mathbf{x}_1 - \mathbf{x}_2)$ ; but for a translation  $(\mathbf{u}_1 - \mathbf{u}_2) = (\mathbf{x}_1 - \mathbf{x}_2)$ . A similar argument shows that angles are preserved (**exercises**).*

**Definition: 4.2** *Rotation*

**Rotation** takes the point  $(x, y)$  to the point

$$(u, v) = x \cos \theta - y \sin \theta, x \sin \theta + y \cos \theta.$$

Here  $\theta$  is the angle of rotation, rotation is anti-clockwise, and

$$(x, y) = u \cos \theta + v \sin \theta, -u \sin \theta + v \cos \theta.$$

Write  $\mathcal{R}$  for a  $2 \times 2$  rotation matrix (a matrix where  $\mathcal{R}^T \mathcal{R} = \mathcal{I}$  and  $\det(\mathcal{R}) = 1$ ); then

$$\mathbf{u} = \mathcal{R}\mathbf{x} \text{ and } \mathbf{x} = \mathcal{R}^{-1}\mathbf{u} = \mathcal{R}^T \mathbf{u}.$$

**Useful Fact:** *Rotation preserves lengths and angles. Choose two points  $\mathbf{x}_1$  and  $\mathbf{x}_2$ . The squared distance from  $\mathbf{x}_1$  to  $\mathbf{x}_2$  is  $(\mathbf{x}_1 - \mathbf{x}_2)^T (\mathbf{x}_1 - \mathbf{x}_2)$ ; but for a rotation  $(\mathbf{u}_1 - \mathbf{u}_2) = \mathcal{R}(\mathbf{x}_1 - \mathbf{x}_2)$  and  $\mathcal{R}^T \mathcal{R} = \mathcal{I}$ . A similar argument shows that angles are preserved (**exercises** ).*

**Definition: 4.3** *Euclidean transformations*

**A Euclidean transformation** is a rotation and translation, so  $(u(x, y), v(x, y)) = (x \cos \theta - y \sin \theta + t_x, x \sin \theta + y \cos \theta + t_y)$ . Euclidean transformations preserve lengths and angles (and so areas) and are sometimes referred to as rigid body transformations. Here  $(x, y) = ((u - t_x) \cos \theta + (v - t_y) \sin \theta, -(u - t_x) \sin \theta + (v - t_y) \cos \theta)$ . In vector notation, for  $\mathcal{R}$  a rotation,

$$\mathbf{u} = \mathcal{R}\mathbf{x} + \mathbf{t} \text{ and } \mathbf{x} = \mathcal{R}^T(\mathbf{u} - \mathbf{t}).$$

**Useful Fact:** *Euclidean transformations preserve lengths and angles (you can think of a Euclidean transformation as a rotation followed by a translation).*

**Definition: 4.4** *Uniform scaling*

For **uniform scaling**,  $(u, v) = (sx, sy)$  for  $s > 0$ . Here  $(x, y) = (1/su, 1/sv)$ . In vector notation,

$$\mathbf{u} = s\mathbf{x} \text{ and } \mathbf{x} = (1/s)\mathbf{u}.$$

**Useful Fact:** *Uniform scaling preserves angles, but not lengths (**exercises** ). Uniform scaling preserves ratios of lengths (**exercises** )*

**Definition: 4.5** *Non-uniform scaling*

For **non-uniform scaling**,  $(u, v) = (sx, ty)$  for  $s$  and  $t$  both positive, and so  $(x, y) = (1/su, 1/tv)$ . Write  $\text{diag}((s, t))$  for the matrix with  $s$  and  $t$  on the diagonal. In vector notation,

$$\mathbf{u} = \text{diag}((s, t))\mathbf{x} \text{ and } \mathbf{x} = \text{diag}((1/s, 1/t))\mathbf{u}.$$

**Useful Fact:** *Non-uniform scaling will usually change both lengths and angles.*

**Definition: 4.6** *Affine transformations*

**Affine transformations** are better written in vector notation. Write  $\mathcal{A}$  for an invertible  $2 \times 2$  matrix, and  $\mathbf{t}$  for some constant vector. Then

$$\mathbf{u} = \mathcal{A}\mathbf{x} + \mathbf{t} \text{ and } \mathbf{x} = \mathcal{A}^{-1}(\mathbf{u} - \mathbf{t}).$$

**Useful Fact:** *Affine transformations will usually change both lengths and angles.*



**Definition: 4.7** *Projective transformations*

**Projective transformations** involve quite inefficient notation if one does not know homogenous coordinates (Section ??), and writing them in vector form is clumsy. Write  $p_{ij}$  for the  $i, j$ 'th component of a  $3 \times 3$  matrix  $\mathcal{P}$  that is invertible. Then

$$\begin{bmatrix} u \\ v \end{bmatrix} = \begin{bmatrix} \frac{p_{11}x + p_{12}y + p_{13}}{p_{31}x + p_{32}y + p_{33}} \\ \frac{p_{21}x + p_{22}y + p_{23}}{p_{31}x + p_{32}y + p_{33}} \end{bmatrix}.$$

The inverse transformation is obtained by applying the inverse of  $\mathcal{P}$  to  $\mathbf{u}$  according to the recipe above. For a vector representation, write

$$\mathcal{P} = \begin{bmatrix} \mathbf{p}_1^T & p_{13} \\ \mathbf{p}_2^T & p_{23} \\ \mathbf{p}_3^T & p_{33} \end{bmatrix}$$

for a  $3 \times 3$  array with inverse  $\mathcal{Q}$ . Then

$$\mathbf{u} = \begin{bmatrix} \frac{\mathbf{p}_1^T \mathbf{x} + p_{13}}{\mathbf{p}_3^T \mathbf{x} + p_{33}} \\ \frac{\mathbf{p}_2^T \mathbf{x} + p_{23}}{\mathbf{p}_3^T \mathbf{x} + p_{33}} \end{bmatrix} \quad \text{and} \quad \mathbf{x} = \begin{bmatrix} \frac{\mathbf{q}_1^T \mathbf{u} + q_{13}}{\mathbf{q}_3^T \mathbf{u} + q_{33}} \\ \frac{\mathbf{q}_2^T \mathbf{u} + q_{23}}{\mathbf{q}_3^T \mathbf{u} + q_{33}} \end{bmatrix}$$

This definition means that, if  $\mathcal{P} = \lambda \mathcal{Q}$  for some  $\lambda \neq 0$ , then  $\mathcal{P}$  and  $\mathcal{Q}$  implement the same projective transformation.

- All the transformations are special cases of projective transformations

# Things to think about

- 4.1. Which transformations preserve angles?
- 4.2. Which transformations preserve lengths?
- 4.3. Could there be a family of transformations that preserves lengths, but not angles? Why?
- 4.4. Assume that the  $2 \times 2$  matrix  $\mathcal{N}$  has the property  $\mathcal{N}^T \mathcal{N} = \mathcal{I}$  and  $\det(\mathcal{N}) = -1$ . Check that there is some rotation  $\mathcal{R}$  such that  $\mathcal{N} = \mathcal{R} \text{diag}((1, -1))$ .
- 4.5. What happens if you apply  $\text{diag}((1, -1))$  to an image?
- 4.6. Figure 4.1 shows two image coordinate systems. What transformation takes the coordinates of a point in the left-hand coordinate system to the coordinates of the same point in the right-hand coordinate system?
- 4.7. Write  $\mathcal{R}$  for a rotation matrix. Show that the transformation that takes  $\mathbf{x}$  to  $\mathcal{R}(\mathbf{x} - \mathbf{t}) + \mathbf{t}$  is a rotation about the point  $\mathbf{t}$ .