

## ORTHOGONAL MATRICES

Informally, an orthogonal  $n \times n$  matrix is the  $n$ -dimensional analogue of the rotation matrices  $R_\theta$  in  $\mathbb{R}^2$ . When does a linear transformation of  $\mathbb{R}^3$  (or  $\mathbb{R}^n$ ) deserve to be called a rotation? Rotations are ‘rigid motions’, in the geometric sense of preserving the lengths of vectors and the angle between vectors. The cosine of the angle formed by two vectors is given by their inner product (or ‘dot product’) divided by the products of their lengths. Thus if our linear transformation preserves lengths of vectors and also the inner product of two vectors, it will automatically be a ‘rigid motion’. In fact, preservation of inner products already implies preservation of lengths, since  $|v|^2 = v \cdot v$ , for any  $v$ .

**Definition.** An  $n \times n$  matrix is *orthogonal* if  $A^t A = \mathbb{I}_n$ .

Recall the basic property of the transpose (for *any*  $A$ ):

$$Av \cdot w = v \cdot A^t w, \quad \forall v, w \in \mathbb{R}^n.$$

It implies that requiring  $A$  to have the property:

$$Av \cdot Aw = v \cdot w, \quad \forall v, w \in \mathbb{R}^n.$$

is the same as requiring:

$$v \cdot A^t A w = v \cdot w, \quad \forall v, w \in \mathbb{R}^n.$$

This is certainly true for orthogonal matrices; thus the action of an orthogonal matrix on vectors in  $\mathbb{R}^n$  preserves lengths and angles.

*Basic properties.* (1) A matrix is orthogonal exactly when its column vectors have length one, and are pairwise orthogonal; likewise for the row vectors. In short, the columns (or the rows) of an orthogonal matrix are an *orthonormal basis* of  $\mathbb{R}^n$ , and any orthonormal basis gives rise to a number of orthogonal matrices.

(2) Any orthogonal matrix is invertible, with  $A^{-1} = A^t$ . If  $A$  is orthogonal, so are  $A^T$  and  $A^{-1}$ .

(3) The product of orthogonal matrices is orthogonal: if  $A^t A = I_n$  and  $B^t B = I_n$ ,

$$(AB)^t(AB) = (B^t A^t)AB = B^t(A^t A)B = B^t B = \mathbb{I}_n.$$

(2) and (3) (plus the fact that the identity is orthogonal) can be summarized by saying the  $n \times n$  orthogonal matrices form a matrix group, the *orthogonal group*  $\mathbb{O}_n$ .

(4) The  $2 \times 2$  rotation matrices  $R_\theta$  are orthogonal. Recall:

$$R_\theta = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}.$$

( $R_\theta$  rotates vectors by  $\theta$  radians, counterclockwise.)

(5) The determinant of an orthogonal matrix is equal to 1 or -1. The reason is that, since  $\det(A) = \det(A^t)$  for any  $A$ , and the determinant of the product is the product of the determinants, we have, for  $A$  orthogonal:

$$1 = \det(I_n) = \det(A^t A) = \det(A^t) \det(A) = (\det A)^2.$$

(6) Any real eigenvalue of an orthogonal matrix has absolute value 1. To see this, consider that  $|Rv| = |v|$  for any  $v$ , if  $R$  is orthogonal. But if  $v \neq 0$  is an eigenvector with eigenvalue  $\lambda$ :

$$Rv = \lambda v \quad \Rightarrow \quad |v| = |Rv| = |\lambda||v|,$$

hence  $|\lambda| = 1$ . (Actually, it is also true that each complex eigenvalue must have modulus 1, and the argument is similar).

### Reflections.

A linear transformation  $T$  of  $\mathbb{R}^n$  is a *reflection* if there is a one-dimensional subspace  $L$  (a line through 0) so that  $Tv = -v$  for  $v \in L$  and  $Tv = v$  for  $v$  in the orthogonal complement  $L^\perp$ . Letting  $n$  be a unit vector spanning  $L$ , we find the expression for  $T$ :

$$Tv = v - 2(v \cdot n)n, \quad v \in \mathbb{R}^n.$$

If  $\{v_1, \dots, v_{n-1}\}$  is a basis of  $L$ , in the basis  $\mathcal{B} = \{v_1, \dots, v_{n-1}, n\}$  of  $\mathbb{R}^n$  the matrix of the reflection  $T$  takes the form (say, for  $n = 3$ ):

$$[T]_{\mathcal{B}} = \begin{bmatrix} -1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}.$$

*Remark.* Like rotations, reflections are rigid motions of space. *How are they different?* (Not at all a trivial question!) The following distinction

is easy to work with: a rotation  $R$  is really the end result of a ‘continuous motion’, where we start by ‘doing nothing’ (the identity map), and gradually move all vectors by rotations  $R(t)$  (where  $t$  denotes ‘time’), until we arrive at the desired rotation  $R = R(1)$ . Since  $\det(I_n) = 1$  and  $\det(R(t))$  depends continuously on  $t$  (and is always 1 or  $-1$ ), we must have  $\det R = 1$  if  $R$  is a rotation. On the other hand, from the above matrix expression we see that a reflection (as defined above) clearly has determinant  $-1$ .

*Ex 1. Reflection on a given plane in  $\mathbb{R}^3$ .* Find the orthogonal matrix (in the standard basis) that implements reflection on the plane with equation:

$$2x_1 + 3x_2 + x_3 = 0$$

*Solution.* The orthogonal line  $L$  is spanned by the unit vector:

$$n = \frac{1}{\sqrt{14}}(2, 3, 1).$$

From the expression given above:

$$Tv = v - 2\frac{1}{14}\langle v, (2, 1, 3) \rangle(2, 1, 3),$$

in particular:

$$\begin{aligned} Te_1 &= e_1 - (4/14)(2, 1, 3) = (3/7, -2/7, -6/7) \\ Te_2 &= e_2 - (2/14)(2, 1, 3) = (-2/7, 6/7, -3/7) \\ Te_3 &= e_3 - (6/14)(2, 1, 3) = (-6/7, -3/7, -2/7) \end{aligned} .$$

These are the columns of the matrix of  $T$  in the standard basis:

$$[T]_{\mathcal{B}_0} = \frac{1}{7} \begin{bmatrix} 3 & -2 & -6 \\ -2 & 6 & -3 \\ -6 & -3 & -2 \end{bmatrix} .$$

### Rotations.

*Ex. 2. Rotation in  $\mathbb{R}^3$  with given axis and given angle.* Find the  $3 \times 3$  orthogonal matrix that implements the rotation  $R$  in  $\mathbb{R}^3$  with axis the subspace spanned by  $(1, 2, 3)$  (a line  $L$ ), by an angle of  $\pi/6$  radians (counterclockwise when looking down the axis).

*Solution.* The first step is to find a positive orthonormal basis of  $\mathbb{R}^3$ ,  $\mathcal{B} = \{u, v_1, v_2\}$ , where  $u$  is a unit vector spanning  $L$  and  $v_1, v_2$  are an orthonormal

basis for the orthogonal plane  $L^\perp$ . ‘Positive’ means  $\det[u|v_1|v_2] = 1$ . We take  $u = (1/\sqrt{14})(1, 2, 3)$ . Start with any basis of  $L^\perp$ , say (by ‘trial and error’): take any two l.i. vectors satisfying the equation  $x_1 + 2x_2 + 3x_3 = 0$  of  $L^\perp$ :

$$w_1 = (3, 0, -1), \quad w_2 = (0, 3, -2).$$

Since:

$$\det \begin{bmatrix} 1 & 3 & 0 \\ 2 & 0 & 3 \\ 3 & -1 & -2 \end{bmatrix} = 33 > 0,$$

the basis  $\{u, w_1, w_2\}$  of  $\mathbb{R}^3$  is positive. Applying Gram-Schmidt to  $w_1, w_2$ , we obtain an orthonormal basis of the plane  $L^\perp$ :

$$v_1 = (1/\sqrt{10})(3, 0, -1), \quad v_2 = (1/\sqrt{35})(-1, 5, -3).$$

(The basis  $\{u, v_1, v_2\}$  is still positive.)

*Second step:* computing the action of  $R$  on the basis vectors. Since  $u$  is a vector on the axis, it is fixed by  $R$ :  $Ru = u$ . In the plane  $L^\perp$ ,  $R$  acts exactly like the counterclockwise rotation by  $\pi/6$  radians in  $\mathbb{R}^2$ , so we know what  $R$  does to an orthonormal basis. We obtain:

$$\begin{aligned} Ru &= (1/\sqrt{14})(1, 2, 3) \\ Rv_1 &= \cos(\pi/6)v_1 + \sin(\pi/6)v_2 = (\sqrt{3}/2)v_1 + (1/2)v_2 \\ Rv_2 &= \sin(\pi/6)v_1 + \cos(\pi/6)v_2 = (-1/2)v_1 + (\sqrt{3}/2)v_2. \end{aligned}$$

This means that the matrix of  $R$  in the ‘adapted’ basis  $\mathcal{B}$  is:

$$[R]_{\mathcal{B}} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \sqrt{3}/2 & -1/2 \\ 0 & 1/2 & \sqrt{3}/2 \end{bmatrix}.$$

This is orthogonal; in fact, in block form, it has a 1 (corresponding to the axis  $u$ , an eigenvector of  $R$  with eigenvalue 1) and a  $2 \times 2$  block given by the rotation  $R_{\pi/6}$  in the plane.

*Third step.* Now use the change of basis formula. We have:

$$[R]_{\mathcal{B}_0} = P[R]_{\mathcal{B}}P^{-1} = P[R]_{\mathcal{B}}P^t,$$

since  $P$  is orthogonal. The change of basis matrix  $P$  is given by the calculation in the first step:

$$P = \begin{bmatrix} 1/\sqrt{14} & 3/\sqrt{10} & -1/\sqrt{35} \\ 2/\sqrt{14} & 0 & 5/\sqrt{35} \\ 3/\sqrt{14} & -1/\sqrt{10} & -3/\sqrt{35} \end{bmatrix}.$$

And that's it. Anyone so inclined can go ahead and multiply out the matrices, but this is best left to a computer.

*Remark on eigenvalues.*

(i) Reflections: the eigenvalues are  $-1$  (with eigenspace  $L$ ) and  $1$  (with eigenspace  $L^\perp$ .) Reflections are *diagonalizable*.

(ii) Rotations. As remarked above, any vector on the axis of a rotation in  $\mathbb{R}^3$  is an eigenvector for the eigenvalue  $\lambda = 1$ . In fact, the eigenspace  $E(1)$  is the axis. Is it possible to see algebraically that a rotation  $A$  in  $\mathbb{R}^3$  (as defined above: an orthogonal matrix with determinant 1) has 1 as an eigenvalue? Well, since the determinant is 1, the characteristic equation has the form:

$$\lambda^3 + a_2\lambda^2 + a_1\lambda + 1 = 0.$$

A polynomial of degree three has at least one real root, and each real root is either 1 or  $-1$  (since  $A$  is orthogonal- property (6) above). If all the roots are real, at least one must be 1 (since their product is the determinant, namely 1). If there are two complex conjugate roots  $(a \pm ib)$  and the real root is  $-1$ , the product of the three roots is  $(a + ib)(a - ib)(-1) = -(a^2 + b^2)$ , which can't possibly be equal to 1. This shows that at least one of the roots is 1, pictorially: *every rotation in  $\mathbb{R}^3$  has an axis*. From the *geometric* construction of a rotation, we see that, in fact, the other two eigenvalues must be complex (since in the plane orthogonal to the axis there are no fixed directions, hence no real eigenvectors.)

**Orthogonal projections.** Let  $V \subset \mathbb{R}^n$  be a subspace,  $\dim(V) = r$ . The orthogonal projection  $P : \mathbb{R}^n \rightarrow \mathbb{R}^n$  is defined by:

$$Pv = v, \text{ if } v \in V; \quad Pv = 0, \text{ if } v \in V^\perp.$$

*Note:  $P$  is not given by an orthogonal matrix!*

Let  $\{v_1, \dots, v_r\}$  be an *orthonormal* basis of  $V$ ; form the corresponding matrix  $A = [v_1|v_2|\dots|v_r] \in \mathbb{M}_{n \times r}$ . The matrix of  $P$  in the standard basis  $\{e_1, \dots, e_n\}$  of  $\mathbb{R}^n$  is given by:  $[P]_{std} = [Pe_1|\dots|Pe_n]$ . From:

$$Pe_i = (e_i \cdot v_1)v_1 + \dots + (e_i \cdot v_r)v_r,$$

we see that the columns of  $[P]_{std}$  are linear combinations of the columns of  $A$ . Thus  $[P]_{std} = AB$  for an  $r \times n$  matrix  $B$ , with entries in the  $i^{th}$  column:

$$B_{ji} = e_i \cdot v_j, \quad j = 1, \dots, r.$$

That is,  $B_{ji}$  is the  $i^{\text{th}}$  component of  $v_j$  (in the standard basis), which is the entry  $A_{ij}$  of  $A$ . So  $B = A^t$  and:

$$[P]_{std} = AA^t.$$

From this it is clear that  $[P]_{std}$  is *symmetric*. (Contrast with  $A^tA = \mathbb{I}_r$ .)