### Orthogonal Projections and Reflections (with exercises) by Dan Klain

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# **Orthogonal Projections**

Let  $X_1, \ldots, X_k$  be a family of linearly independent (column) vectors in  $\mathbb{R}^n$ , and let

$$W = \operatorname{Span}(X_1, \ldots, X_k).$$

In other words, the vectors  $X_1, \ldots, X_k$  form a basis for the *k*-dimensional subspace *W* of  $\mathbb{R}^n$ .

Suppose we are given another vector  $Y \in \mathbb{R}^n$ . How can we project Y onto W orthogonally? In other words, can we find a vector  $\hat{Y} \in W$  so that  $Y - \hat{Y}$  is orthogonal (perpendicular) to all of W? See Figure 1.

To begin, translate this question into the language of matrices and dot products. We need to find a vector  $\hat{Y} \in W$  such that

(1) 
$$(Y - \hat{Y}) \perp Z$$
, for all vectors  $Z \in W$ .

Actually, it's enough to know that  $Y - \hat{Y}$  is perpendicular to the vectors  $X_1, \ldots, X_k$  that span *W*. This would imply that (1) holds. (Why?)

Expressing this using dot products, we need to find  $\hat{Y} \in W$  so that

(2) 
$$X_i^T(Y - \hat{Y}) = 0$$
, for all  $i = 1, 2, ..., k$ .

This condition involves taking k dot products, one for each  $X_i$ . We can do them all at once by setting up a matrix A using the  $X_i$  as the columns of A, that is, let

$$A = \left[ \begin{array}{c|c} X_1 & X_2 & \cdots & X_k \end{array} \right].$$

Note that each vector  $X_i \in \mathbb{R}^n$  has *n* coordinates, so that *A* is an  $n \times k$  matrix. The set of conditions listed in (2) can now be re-written:

$$A^T(Y - \hat{Y}) = 0,$$

which is equivalent to

$$A^T Y = A^T \hat{Y}$$

Meanwhile, we need the projected vector  $\hat{Y}$  to be a vector in W, since we are projecting onto W. This means that  $\hat{Y}$  lies in the span of the vectors  $X_1, \ldots, X_k$ . In other



FIGURE 1. Projection of a vector onto a subspace.

words,

$$\hat{Y} = c_1 X_1 + c_2 X_2 + \dots + c_k X_k = A \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_k \end{bmatrix} = AC$$

where *C* is a *k*-dimensional column vector. On combining this with the matrix equation (3) we have

$$A^T Y = A^T A C.$$

If we knew what *C* was then we would also know  $\hat{Y}$ , since we were given the columns  $X_i$  of *A*, and  $\hat{Y} = AC$ . To solve for *C* just invert the  $k \times k$  matrix  $A^T A$  to get

$$(4) (ATA)-1ATY = C.$$

How do we know that  $(A^T A)^{-1}$  exists? Let's assume it does for now, and then address this question later on.

Now finally we can find our projected vector  $\hat{Y}$ . Since  $\hat{Y} = AC$ , multiply both sides of (4) to obtain

$$A(A^T A)^{-1} A^T Y = AC = \hat{Y}.$$

The matrix

$$Q = A(A^T A)^{-1} A^T$$

is called the *projection matrix for the subspace* W. According to our derivation above, the projection matrix Q maps a vector  $Y \in \mathbb{R}^n$  to its orthogonal projection (i.e. its shadow)  $QY = \hat{Y}$  in the subspace W.

It is easy to check that *Q* has the following nice properties:

(1) 
$$Q^T = Q$$
.  
(2)  $Q^2 = Q$ .

One can show that any matrix satisfying these two properties is in fact a projection matrix for its own column space. You can prove this using the hints given in the exercises.

There remains one problem. At a crucial step in the derivation above we took the inverse of the  $k \times k$  matrix  $A^T A$ . But how do we know this matrix is invertible? It is invertible, because the columns of A, the vectors  $X_1, \ldots, X_k$ , were assumed to be *linearly independent*. But this claim of invertibility needs a proof.

**Lemma 1.** Suppose A is an  $n \times k$  matrix, where  $k \leq n$ , such that the columns of A are linearly independent. Then the  $k \times k$  matrix  $A^T A$  is invertible.

*Proof of Lemma 1:* Suppose that  $A^T A$  is *not* invertible. In this case, there exists a vector  $X \neq 0$  such that  $A^T A X = 0$ . It then follows that

$$(AX) \cdot (AX) = (AX)^T A X = X^T A^T A X = X^T 0 = 0,$$

so that the length  $||AX|| = \sqrt{(AX) \cdot (AX)} = 0$ . In other words, the length of AX is zero, so that AX = 0. Since  $X \neq 0$ , this implies that the columns of A are linearly dependent. Therefore, if the columns of A are linearly independent, then  $A^TA$  must be invertible.

**Example:** Compute the projection matrix Q for the 2-dimensional subspace W of  $\mathbb{R}^4$  spanned by the vectors (1, 1, 0, 2) and (-1, 0, 0, 1). What is the orthogonal projection of the vector (0, 2, 5, -1) onto W?

Solution: Let

$$A = \begin{bmatrix} 1 & -1 \\ 1 & 0 \\ 0 & 0 \\ 2 & 1 \end{bmatrix}.$$

Then

$$A^{T}A = \begin{bmatrix} 6 & 1 \\ 1 & 2 \end{bmatrix}$$
 and  $(A^{T}A)^{-1} = \begin{bmatrix} \frac{2}{11} & \frac{-1}{11} \\ \frac{-1}{11} & \frac{6}{11} \end{bmatrix}$ ,

so that the projection matrix *Q* is given by

$$Q = A(A^{T}A)^{-1}A^{T} = \begin{bmatrix} 1 & -1\\ 1 & 0\\ 0 & 0\\ 2 & 1 \end{bmatrix} \begin{bmatrix} \frac{2}{11} & \frac{-1}{11}\\ \frac{-1}{11} & \frac{6}{11} \end{bmatrix} \begin{bmatrix} 1 & 1 & 0 & 2\\ -1 & 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} \frac{10}{11} & \frac{3}{11} & 0 & \frac{-1}{11}\\ \frac{3}{11} & \frac{2}{11} & 0 & \frac{3}{11}\\ 0 & 0 & 0 & 0\\ \frac{-1}{11} & \frac{3}{11} & 0 & \frac{10}{11} \end{bmatrix}$$

We can now compute the orthogonal projection of the vector (0, 2, 5, -1) onto *W*. This is

$$\begin{bmatrix} \frac{10}{11} & \frac{3}{11} & 0 & \frac{-1}{11} \\ \frac{3}{11} & \frac{2}{11} & 0 & \frac{3}{11} \\ 0 & 0 & 0 & 0 \\ \frac{-1}{11} & \frac{3}{11} & 0 & \frac{10}{11} \end{bmatrix} \begin{bmatrix} 0 \\ 2 \\ 5 \\ -1 \end{bmatrix} = \begin{bmatrix} \frac{7}{11} \\ \frac{1}{11} \\ 0 \\ \frac{-4}{11} \end{bmatrix}$$



FIGURE 2. Reflection of the vector v across the plane W.

### Reflections

We have seen earlier in the course that reflections of space across (i.e. through) a plane is linear transformation. Like rotations, a reflection preserves lengths and angles, although, unlike rotations, a reflection reverses orientation ("handedness").

Once we have projection matrices it is easy to compute the matrix of a reflection. Let W denote a plane passing through the origin, and suppose we want to reflect a vector v across this plane, as in Figure 2.

Let *u* denote a unit vector along  $W^{\perp}$ , that is, let *u* be a normal to the plane *W*. We will think of *u* and *v* as column vectors. The projection of *v* along the line through *u* is then given by:

$$\hat{v} = Proj_u(v) = u(u^T u)^{-1} u^T v.$$

But since we chose *u* to be a *unit* vector,  $u^T u = u \cdot u = 1$ , so that

$$\hat{v} = Proj_u(v) = uu^T v.$$

Let  $Q_u$  denote the matrix  $uu^T$ , so that  $\hat{v} = Q_u v$ .

What is the reflection of v across W? It is the vector  $\operatorname{Refl}_W(v)$  that lies on the other side of W from v, exactly the same distance from W as is v, and having the same projection into W as v. See Figure 2. The distance between v and its reflection is exactly twice the distance of v to W, and the difference between v and its reflection is perpendicular to W. That is, the difference between v and its reflection is exactly twice the projection of v along the unit normal u to W. This observation yields the equation:

$$v - \operatorname{Refl}_W(v) = 2Q_u v$$

so that

$$\operatorname{Refl}_W(v) = v - 2Q_uv = Iv - 2Q_uv = (I - 2uu^T)v$$

The matrix  $H_W = I - 2uu^T$  is called the *reflection matrix for the plane W*, and is also sometimes called a *Householder matrix*.

**Example:** Compute the reflection of the vector v = (-1, 3, -4) across the plane 2x - y + 7z = 0.

**Solution:** The vector w = (2, -1, 7) is normal to the plane, and  $w^T w = 2^2 + (-1)^2 +$  $7^2 = 54$ , so a unit normal will be

$$u = \frac{w}{|w|} = \frac{1}{\sqrt{54}}(2, -1, 7).$$

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The reflection matrix is then given by

$$H = I - 2uu^{T} = I - \frac{2}{54}ww^{T} = I - \frac{1}{27}\begin{bmatrix}2\\-1\\7\end{bmatrix}\begin{bmatrix}2&-1&7\end{bmatrix} = \cdots$$
$$\cdots = \begin{bmatrix}1&0&0\\0&1&0\\0&0&1\end{bmatrix} - \frac{1}{27}\begin{bmatrix}4&-2&14\\-2&1&-7\\14&-7&49\end{bmatrix},$$

so that

$$H = \begin{bmatrix} 23/27 & 2/27 & -14/27 \\ 2/27 & 26/27 & 7/27 \\ -14/27 & 7/27 & -22/27 \end{bmatrix}$$

The reflection of v across W is then given by

$$Hv = \begin{bmatrix} 23/27 & 2/27 & -14/27 \\ 2/27 & 26/27 & 7/27 \\ -14/27 & 7/27 & -22/27 \end{bmatrix} \begin{bmatrix} -1 \\ 3 \\ -4 \end{bmatrix} = \begin{bmatrix} 39/27 \\ 48/27 \\ 123/27 \end{bmatrix}$$

## **Reflections and Projections**

Notice in Figure 2 that the *projection* of *v* into *W* is the *midpoint* of the vector *v* and its reflection  $Hv = \operatorname{Refl}_W(v)$ ; that is,

$$Qv = \frac{1}{2}(v + Hv)$$
 or, equivalently  $Hv = 2Qv - v$ ,

where  $Q = Q_W$  denotes the projection onto W. (This is not the same as  $Q_u$  in the previous section, which was projection onto the *normal* of *W*.)

Let *I* denote the identity matrix, so that v = Iv for all vectors  $v \in \mathbb{R}^n$ . The identities above can now be expressed as matrix identities:

$$Q = \frac{1}{2}(I+H)$$
 and  $H = 2Q - I$ ,

So once you have computed the either the projection or reflection matrix for a subspace of  $\mathbb{R}^n$ , the other is quite easy to obtain.

## **Exercises**

**1.** Suppose that *M* is an  $n \times n$  matrix such that  $M^T = M = M^2$ . Let *W* denote the column space of *M*.

(a) Suppose that  $Y \in W$ . (This means that Y = MX for some *X*.) Prove that MY = Y.

**(b)** Suppose that v is a vector in  $\mathbb{R}^n$ . Why is  $Mv \in W$ ?

(c) If  $Y \in W$ , why is  $v - Mv \perp Y$ ?

(d) Conclude that Mv is the projection of v into W.

**2.** Compute the projection of the vector v = (1, 1, 0) onto the plane x + y - z = 0.

**3.** Compute the projection matrix *Q* for the subspace *W* of  $\mathbb{R}^4$  spanned by the vectors (1,2,0,0) and (1,0,1,1).

**4.** Compute the orthogonal projection of the vector z = (1, -2, 2, 2) onto the subspace *W* of Problem **3.** above. What does your answer tell you about the relationship between the vector *z* and the subspace *W*?

**5.** Recall that a square matrix *P* is said to be an *orthogonal matrix* if  $P^T P = I$ . Show that Householder matrices are always orthogonal matrices; that is, show that  $H^T H = I$ .

**6.** Compute the Householder matrix for reflection across the plane x + y - z = 0.

**7.** Compute the reflection of the vector v = (1, 1, 0) across the plane x + y - z = 0. What happens when you add v to its reflection? How does this sum compare to your answer from Exercise 2? Draw a sketch to explain this phenomenon.

**8.** Compute the reflection of the vector v = (1, 1) across the line  $\ell$  in  $\mathbb{R}^2$  spanned by the vector (2,3). Sketch the vector v, the line  $\ell$  and the reflection of v across  $\ell$ . (Do not confuse the spanning vector for  $\ell$  with the normal vector to  $\ell$ .)

**9.** Compute the Householder matrix *H* for reflection across the hyperplane  $x_1 + 2x_2 - x_3 - 3x_4 = 0$  in  $\mathbb{R}^4$ . Then compute the projection matrix *Q* for this hyperplane.

**10.** Compute the Householder matrix for reflection across the plane z = 0 in  $\mathbb{R}^3$ . Sketch the reflection involved. Your answer should not be too surprising!

#### **Selected Solutions to Exercises:**

2. We describe two ways to solve this problem.

*Solution 1:* Pick a basis for the plane. Since the plane is 2-dimensional, any two independent vectors in the plane will do, say, (1, -1, 0) and (0, 1, 1). Set

$$A = \begin{bmatrix} 1 & 0 \\ -1 & 1 \\ 0 & 1 \end{bmatrix} \text{ and } A^T A = \begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix}$$

The projection matrix Q for the plane is

$$Q = A(A^{T}A)^{-1}A^{T} = \begin{bmatrix} 1 & 0 \\ -1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 2/3 & 1/3 \\ 1/3 & 2/3 \end{bmatrix} \begin{bmatrix} 1 & -1 & 0 \\ 0 & 1 & 1 \end{bmatrix} = \begin{bmatrix} 2/3 & -1/3 & 1/3 \\ -1/3 & 2/3 & 1/3 \\ 1/3 & 1/3 & 2/3 \end{bmatrix}$$

We can now project any vector onto the plane by multiplying by *Q*:

$$Projection(v) = Qv = \begin{bmatrix} 2/3 & -1/3 & 1/3 \\ -1/3 & 2/3 & 1/3 \\ 1/3 & 1/3 & 2/3 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 1/3 \\ 1/3 \\ 2/3 \end{bmatrix}$$

Solution 2: First, project v onto the normal vector n = (1, 1, -1) to the plane:

$$y = Proj_n(v) = \frac{v \cdot n}{n \cdot n}n = (2/3, 2/3, -2/3).$$

Since *y* is the component of *v* orthogonal to the plane, the vector v - y is the orthogonal projection of *v* onto the plane. The solution (given in row vector notation) is

$$v - y = (1, 1, 0) - (2/3, 2/3, -2/3) = (1/3, 1/3, 2/3),$$

as in the previous solution.

**Note:** Which method is better? The second way is shorter for hyperplanes (subspaces of  $\mathbb{R}^n$  having dimension n - 1), but finding the projection matrix Q is needed if you are projecting from  $\mathbb{R}^n$  to some intermediate dimension k, where you no longer have a single normal vector to work with. For example, a 2-subspace in  $\mathbb{R}^4$  has a 2-dimensional orthogonal complement as well, so one must compute a projection matrix in order to project to either component of  $\mathbb{R}^4$ , as in the next problem.

**3.** Set

$$A = \begin{bmatrix} 1 & 1 \\ 2 & 0 \\ 0 & 1 \\ 0 & 1 \end{bmatrix} \text{ so that } A^{T}A = \begin{bmatrix} 5 & 1 \\ 1 & 3 \end{bmatrix}$$

The projection matrix Q for the plane is

$$Q = A(A^{T}A)^{-1}A^{T} = \begin{bmatrix} 1 & 1\\ 2 & 0\\ 0 & 1\\ 0 & 1 \end{bmatrix} \begin{bmatrix} \frac{3}{14} & -\frac{1}{14}\\ -\frac{1}{14} & \frac{5}{14} \end{bmatrix} \begin{bmatrix} 1 & 2 & 0 & 0\\ 1 & 0 & 1 & 1 \end{bmatrix} = \begin{bmatrix} \frac{6}{14} & \frac{4}{14} & \frac{4}{14} & \frac{4}{14}\\ \frac{4}{14} & \frac{12}{14} & -\frac{2}{14} & -\frac{2}{14}\\ \frac{4}{14} & -\frac{2}{14} & \frac{5}{14} & \frac{5}{14}\\ \frac{4}{14} & -\frac{2}{14} & \frac{5}{14} & \frac{5}{14} \end{bmatrix}$$

**5.** Here is a hint: Use the fact that  $H = I - 2uu^T$ , where *I* is the identity matrix and *u* is a unit column vector. What is  $H^T =$ ? What is  $H^T H =$ ?

**6.** The vector v = (1, 1, -1) is normal to the plane x + y - z = 0, so the vector  $u = \frac{1}{\sqrt{3}}(1, 1, -1) = \frac{1}{\sqrt{3}}v$  is a unit normal. Expressing *u* and *v* as column vectors we find that

$$I - 2uu^{T} = I - (2/3)vv^{T} = I - \frac{2}{3} \begin{bmatrix} 1\\1\\-1 \end{bmatrix} \begin{bmatrix} 1 & 1 & -1 \end{bmatrix}$$
$$= \begin{bmatrix} 1 & 0 & 0\\0 & 1 & 0\\0 & 0 & 1 \end{bmatrix} - \frac{2}{3} \begin{bmatrix} 1 & 1 & -1\\1 & 1 & -1\\-1 & -1 & 1 \end{bmatrix} = \begin{bmatrix} \frac{1}{3} & -\frac{2}{3} & \frac{2}{3}\\-\frac{2}{3} & \frac{1}{3} & \frac{2}{3}\\\frac{2}{3} & \frac{2}{3} & \frac{1}{3} \end{bmatrix}$$