

5.4 Projection matrices

Review. To compute the orthogonal projection $\hat{\mathbf{b}}$ of \mathbf{b} onto W , we have two options:

- **(using least squares)** Write $W = \text{col}(A)$, where the columns of A are a basis of W . Then, $\hat{\mathbf{b}} = A\hat{\mathbf{x}}$ where $\hat{\mathbf{x}}$ is the least squares solution to $A\mathbf{x} = \mathbf{b}$ (i.e. $A^T A\hat{\mathbf{x}} = A^T \mathbf{b}$).
- **(using orthogonality)** If $\mathbf{v}_1, \dots, \mathbf{v}_m$ is an orthogonal basis of W , then

$$\hat{\mathbf{b}} = \left(\frac{\mathbf{b} \cdot \mathbf{v}_1}{\mathbf{v}_1 \cdot \mathbf{v}_1} \right) \mathbf{v}_1 + \dots + \left(\frac{\mathbf{b} \cdot \mathbf{v}_m}{\mathbf{v}_m \cdot \mathbf{v}_m} \right) \mathbf{v}_m.$$

In the least squares approach, assuming $A^T A$ is invertible (which, by Example 45, is automatically the case if the columns of A are independent), we have $\hat{\mathbf{x}} = (A^T A)^{-1} A^T \mathbf{b}$ and hence:

(projection matrix) The projection $\hat{\mathbf{b}}$ of \mathbf{b} onto $\text{col}(A)$ is

$$\hat{\mathbf{b}} = \underbrace{A(A^T A)^{-1} A^T}_{P} \mathbf{b}.$$

The matrix $P = A(A^T A)^{-1} A^T$ is the **projection matrix** for projecting onto $\text{col}(A)$.

Example 66.

- (a) Determine the projection matrix P for projecting onto $W = \text{span} \left\{ \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}$
- (b) Determine (once more!) the projection of $\begin{bmatrix} 3 \\ 7 \\ 4 \end{bmatrix}$ onto W using the projection matrix.

Solution.

(a) Choosing $A = \begin{bmatrix} 1 & 0 \\ -1 & 0 \\ 0 & 1 \end{bmatrix}$, the projection matrix P is $A(A^T A)^{-1} A^T = \begin{bmatrix} 1 & 0 \\ -1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix}^{-1} \begin{bmatrix} 1 & -1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$

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

(b) The projection of $\begin{bmatrix} 3 \\ 7 \\ 4 \end{bmatrix}$ onto W is $\begin{bmatrix} 1/2 & -1/2 & 0 \\ -1/2 & 1/2 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 3 \\ 7 \\ 4 \end{bmatrix} = \begin{bmatrix} -2 \\ 2 \\ 4 \end{bmatrix}.$

This is, of course, exactly the answer we found in Example 64.

Comment. You can choose A in any way such that its columns are a basis for W . The final projection matrix will always be the same.

Example 67. (homework) If P is a projection matrix, then what is P^2 ?

For instance. For P as in the previous example, $P^2 = \begin{bmatrix} 1/2 & -1/2 & 0 \\ -1/2 & 1/2 & 0 \\ 0 & 0 & 1 \end{bmatrix}^2 = \begin{bmatrix} 1/2 & -1/2 & 0 \\ -1/2 & 1/2 & 0 \\ 0 & 0 & 1 \end{bmatrix} = P.$

Solution. Can you see why it is always true that $P^2 = P$?

[Recall that P projects a vector onto a space W (actually, $W = \text{col}(P)$). Hence P^2 takes a vector \mathbf{b} , projects it onto W to get $\hat{\mathbf{b}}$, and then projects $\hat{\mathbf{b}}$ onto W again. But the projection of $\hat{\mathbf{b}}$ onto W is just $\hat{\mathbf{b}}$ (why?!), so that P^2 always has the exact same effect as P . Therefore, $P^2 = P$.]

Example 68. (homework)

(a) What is the matrix P for projecting onto $W = \text{span}\left\{\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}\right\}$?

(b) Using the projection matrix, project $\begin{bmatrix} 2 \\ 3 \\ 3 \end{bmatrix}$ onto $W = \text{span}\left\{\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}\right\}$.

Solution.

(a) Choosing $A = \begin{bmatrix} 1 & 1 \\ 1 & -1 \\ 1 & 1 \end{bmatrix}$, the projection matrix P is $A(A^T A)^{-1} A^T = \begin{bmatrix} 1 & 1 \\ 1 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 3 & 1 \\ 1 & 3 \end{bmatrix}^{-1} \begin{bmatrix} 1 & 1 & 1 \\ 1 & -1 & 1 \end{bmatrix}$

$$= \begin{bmatrix} 1 & 1 \\ 1 & -1 \\ 1 & 1 \end{bmatrix} \frac{1}{8} \begin{bmatrix} 3 & -1 \\ -1 & 3 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 \\ 1 & -1 & 1 \end{bmatrix} = \frac{1}{8} \begin{bmatrix} 1 & 1 \\ 1 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 2 & 4 & 2 \\ 2 & -4 & 2 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 1 & 0 & 1 \\ 0 & 2 & 0 \\ 1 & 0 & 1 \end{bmatrix}.$$

(b) The projection is $\frac{1}{2} \begin{bmatrix} 1 & 0 & 1 \\ 0 & 2 & 0 \\ 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 2 \\ 3 \\ 3 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 5 \\ 6 \\ 5 \end{bmatrix}$.

Check. The error $\begin{bmatrix} 2 \\ 3 \\ 3 \end{bmatrix} - \frac{1}{2} \begin{bmatrix} 5 \\ 6 \\ 5 \end{bmatrix} = \begin{bmatrix} -1/2 \\ 0 \\ 1/2 \end{bmatrix}$ is indeed orthogonal to W .

5.5 Gram–Schmidt

Example 69. Find an orthogonal basis for $W = \text{span}\left\{\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}\right\}$.

Solution. We already have the basis $\mathbf{w}_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$, $\mathbf{w}_2 = \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$ for W . However, that basis is not orthogonal.

We can construct an orthogonal basis $\mathbf{q}_1, \mathbf{q}_2$ for W as follows:

- $\mathbf{q}_1 = \mathbf{w}_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$

Since this is our first basis vector, we don't yet have other basis vectors it needs to be orthogonal to.

- $\mathbf{q}_2 = \mathbf{w}_2 - \left(\text{projection of } \mathbf{w}_2 \text{ onto } \mathbf{q}_1\right) = \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix} - \frac{1}{3} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 2/3 \\ -4/3 \\ 2/3 \end{bmatrix}$

Make sure our way to construct \mathbf{q}_2 makes sense to you!

\mathbf{q}_2 is the error of the projection of \mathbf{w}_2 onto \mathbf{q}_1 . This guarantees that it is orthogonal to \mathbf{q}_1 .

On the other hand, since \mathbf{q}_2 is a combination of \mathbf{w}_2 and \mathbf{q}_1 , we know that \mathbf{q}_2 actually is in W .

We have thus found the orthogonal basis $\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 2/3 \\ -4/3 \\ 2/3 \end{bmatrix}$ for W .

Important comment. Normalizing these, we get $\frac{1}{\sqrt{3}} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \frac{1}{\sqrt{6}} \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix}$, which is an orthonormal basis for W .

Comment. There are, of course, many orthogonal bases $\mathbf{q}_1, \mathbf{q}_2$ for W . Up to the length of the vectors, ours is the unique one with the property that $\text{span}\{\mathbf{q}_1\} = \text{span}\{\mathbf{w}_1\}$ and $\text{span}\{\mathbf{q}_1, \mathbf{q}_2\} = \text{span}\{\mathbf{w}_1, \mathbf{w}_2\}$.

Example 70. Extend the previous basis to an orthogonal basis of $\text{span}\left\{\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix}\right\}$.

[Of course, this span is all of \mathbb{R}^3 .]

Solution. Let us repeat the previous step so the entire procedure becomes more transparent.

We begin with the (not orthogonal) basis $w_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$, $w_2 = \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$, $w_3 = \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix}$.

We then construct an orthogonal basis q_1, q_2, q_3 as follows:

- $q_1 = w_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$
- $q_2 = w_2 - \left(\text{projection of } w_2 \text{ onto } q_1\right) = \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix} - \frac{1}{3} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 2/3 \\ -4/3 \\ 2/3 \end{bmatrix}$
- $q_3 = w_3 - \left(\text{projection of } w_3 \text{ onto } \text{span}\{q_1, q_2\}\right) = w_3 - \left(\text{projection of } w_3 \text{ onto } q_1\right) - \left(\text{projection of } w_3 \text{ onto } q_2\right)$
 $= \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix} - \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} - \frac{1}{4} \begin{bmatrix} 2 \\ -4 \\ 2 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$

Make sure you see why q_3 is orthogonal to both q_1 and q_2 !

Also note that breaking up the projection onto $\text{span}\{q_1, q_2\}$ into the projections onto q_1 and q_2 is only possible because q_1 and q_2 are orthogonal.

Indeed, $\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \frac{2}{3} \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix}, \frac{1}{2} \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$ is an orthogonal basis of \mathbb{R}^3 .

If we prefer, we can normalize to obtain an orthonormal basis: $\frac{1}{\sqrt{3}} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \frac{1}{\sqrt{6}} \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix}, \frac{1}{\sqrt{2}} \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$.

(Gram–Schmidt orthogonalization)

Given a basis w_1, w_2, \dots for W , produce an orthogonal basis q_1, q_2, \dots for W .

- $q_1 = w_1$
- $q_2 = w_2 - \left(\text{projection of } w_2 \text{ onto } q_1\right)$
- $q_3 = w_3 - \left(\text{projection of } w_3 \text{ onto } q_1\right) - \left(\text{projection of } w_3 \text{ onto } q_2\right)$
- $q_4 = \dots$

Important comment. When working numerically it actually saves time to compute an orthonormal basis q_1, q_2, \dots by the same approach but always normalizing each q_i along the way. The reason this saves time is that now the projections onto q_i only require a single dot product (instead of two). This is called **Gram–Schmidt orthonormalization**.

Note. When normalizing, the orthonormal basis q_1, q_2, \dots is the unique one with the property that $\text{span}\{q_1, q_2, \dots, q_k\} = \text{span}\{w_1, w_2, \dots, w_k\}$ for all $k = 1, 2, \dots$.