

§6.3 Orthogonal Projections

Last time we saw how to project a vector onto a line. What if we want a projection onto a plane? Or in general some span of vectors in \mathbb{R}^n ?

Definition

Let W be a subspace of \mathbb{R}^n . The orthogonal complement of W is the set of vectors orthogonal to every vector in W . We notate it as W^\perp . In other words

$$u \cdot v = 0$$

for any u in W and v in W^\perp

Example

Suppose $\{v_1, \dots, v_n\}$ is an orthogonal basis of \mathbb{R}^n and let $W = \text{span}\{v_1, \dots, v_k\}$ for $k \leq n$.

Then $W^\perp = \text{span}\{v_{k+1}, \dots, v_n\}$

Theorem (Orthogonal Decomposition Theorem)

Let W be a subspace of \mathbb{R}^n . Then any vector y in \mathbb{R}^n can be written uniquely as

$$y = \hat{y} + z$$

Where \hat{y} is in W and z is in W^\perp . Here \hat{y} is called the projection of y onto W and is sometimes written as $\hat{y} = \text{Proj}_W y$

How do we find \hat{y} and z ?

If $\{u_1, \dots, u_m\}$ is an orthogonal basis of W , then

$$\bullet \hat{y} = \left(\frac{y \cdot u_1}{u_1 \cdot u_1} \right) u_1 + \left(\frac{y \cdot u_2}{u_2 \cdot u_2} \right) u_2 + \dots + \left(\frac{y \cdot u_m}{u_m \cdot u_m} \right) u_m$$

$$\bullet z = y - \hat{y}$$

Notice this generalizes the case from last time when projecting onto $L = \text{span}\{u\}$.

Proof

Since \hat{y} is in W and $\{u_1, u_2, \dots, u_m\}$ is an orthogonal basis for W , the theorem from §6.2 tells us

$$\hat{y} = \left(\frac{\hat{y} \cdot u_1}{u_1 \cdot u_1} \right) u_1 + \left(\frac{\hat{y} \cdot u_2}{u_2 \cdot u_2} \right) u_2 + \dots + \left(\frac{\hat{y} \cdot u_m}{u_m \cdot u_m} \right) u_m$$

which is almost what we want. It suffices to show $\hat{y} \cdot u_i = y \cdot u_i$ for $i=1, \dots, m$ to obtain the result.

Since $z = y - \hat{y}$ is orthogonal to W , we have

$$0 = z \cdot u_i = (y - \hat{y}) \cdot u_i = y \cdot u_i - \hat{y} \cdot u_i$$

so indeed $y \cdot u_i = \hat{y} \cdot u_i$ for all $i=1, \dots, m$

Thus

$$\hat{y} = \left(\frac{y \cdot u_1}{u_1 \cdot u_1} \right) u_1 + \left(\frac{y \cdot u_2}{u_2 \cdot u_2} \right) u_2 + \dots + \left(\frac{y \cdot u_m}{u_m \cdot u_m} \right) u_m$$

Example

Let W be the subspace of \mathbb{R}^3 spanned by

$$u_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \quad u_2 = \begin{bmatrix} 3 \\ -2 \\ -1 \end{bmatrix}$$

a) Verify $\{u_1, u_2\}$ is an orthogonal basis of W

b) Write $y = \begin{bmatrix} -2 \\ 6 \\ 3 \end{bmatrix}$ as a sum of a vector in W and a vector in W^\perp .

Solution

a) This is clear.

b) $y = \hat{y} + z$ where

$$\hat{y} = \left(\frac{y \cdot u_1}{u_1 \cdot u_1} \right) u_1 + \left(\frac{y \cdot u_2}{u_2 \cdot u_2} \right) u_2$$
$$= \left(\frac{-2+6+3}{1+1+1} \right) u_1 + \left(\frac{-6-12-3}{9+4+1} \right) u_2$$

$$= \frac{7}{3} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} + \left(\frac{-21}{14} \right) \begin{bmatrix} 3 \\ -2 \\ -1 \end{bmatrix}$$

$= -3/2$

$$= \begin{bmatrix} -13/6 \\ 16/3 \\ 23/6 \end{bmatrix}$$

after getting common denominators

$$\text{Now } z = y - \hat{y} = \begin{bmatrix} -2 \\ 6 \\ 3 \end{bmatrix} - \begin{bmatrix} -13/6 \\ 16/3 \\ 23/6 \end{bmatrix} = \begin{bmatrix} 1/6 \\ 2/3 \\ -5/6 \end{bmatrix}$$

Theorem (Best Approximation Theorem)

With the notation as before if W is a subspace of \mathbb{R}^n , y any vector in \mathbb{R}^n , and $\hat{y} = \text{proj}_W y$ then \hat{y} is the ~~best~~ vector in W closest to y . In other words

$$\|y - \hat{y}\| < \|y - v\|$$

for all v in W with $v \neq \hat{y}$

Thus, just as in §6.2, $\|z\| = \|y - \hat{y}\|$ is the shortest distance from y to W .

Remark

Notice y is in W if and only if $y = \hat{y} = \text{proj}_W y$.

This of course implies $z = 0$ so the distance of y to W is zero.

The previous result requires only an orthogonal basis for W . The result is even nicer when an orthonormal basis is selected.

Theorem

Let $\{u_1, \dots, u_m\}$ be an orthonormal basis of subspace W of \mathbb{R}^n . For any vector y in \mathbb{R}^n ,

$$\text{proj}_W y = (y \cdot u_1)u_1 + (y \cdot u_2)u_2 + \dots + (y \cdot u_m)u_m$$

In particular, if $U = [u_1 | u_2 | \dots | u_m]$ is the $n \times m$ matrix with columns u_1, \dots, u_m , then also

$$\text{proj}_W y = UU^T y$$

Proof

The first statement follows from the previous formula as $u_i \cdot u_i = 1$ for $i = 1, \dots, m$.

The second statement follows as the entries of $U^T y$ are $u_1^T y, u_2^T y, \dots, u_m^T y$ which are exactly $u_1 \cdot y, u_2 \cdot y, \dots, u_m \cdot y$

Example

Suppose $\{u_1, u_2\}$ is an orthonormal basis for W
where $W = \text{span}\{u_1, u_2\}$

$$u_1 = \begin{bmatrix} 1/3 \\ 2/3 \\ 2/3 \end{bmatrix} \quad u_2 = \begin{bmatrix} -2/3 \\ -1/3 \\ 2/3 \end{bmatrix}$$

a) check this is an orthonormal basis

b) compute $\text{proj}_W y$ where $y = \begin{bmatrix} 3 \\ 7 \\ 1 \end{bmatrix}$

Solution

a) $\|u_1\| = \|u_2\| = 1$ and $u_1 \cdot u_2 = 0$

$$b) \text{proj}_W y = \begin{bmatrix} 1/3 & -2/3 \\ 2/3 & -1/3 \\ 2/3 & 2/3 \end{bmatrix} \begin{bmatrix} 1/3 & 2/3 & 2/3 \\ -2/3 & -1/3 & 2/3 \end{bmatrix} \begin{bmatrix} 3 \\ 7 \\ 1 \end{bmatrix}$$

$$= \begin{bmatrix} 5/9 & 4/9 & -2/9 \\ 4/9 & 5/9 & 2/9 \\ -2/9 & 2/9 & 8/9 \end{bmatrix} \begin{bmatrix} 3 \\ 7 \\ 1 \end{bmatrix}$$

$$= \begin{bmatrix} 41/9 \\ 49/9 \\ 16/9 \end{bmatrix}$$