

# Orthogonal Projection

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# Orthogonal Decomposition

Let  $W$  be a subspace of  $\mathbf{R}^n$ . Then every vector  $y \in \mathbf{R}^n$  can be written

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

where  $\hat{y} \in W$  and  $z \in W^\perp$ .

## Orthogonal decomposition

To compute the decomposition, let  $\{u_1, u_2, \dots, u_k\}$  be an orthogonal basis of  $W$ . Let

$$\hat{y} = \sum_{i=1}^k \frac{y \cdot u_i}{u_i \cdot u_i} u_i.$$

Let  $z = y - \hat{y}$ .

Notice that, for any  $i = 1, \dots, n$ ,

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

so  $z \in W^\perp$ .

The vector  $\hat{y}$  is called the orthogonal projection of  $y$  onto  $W$ .

## Example

Let

$$y = \begin{bmatrix} -1 \\ 4 \\ 3 \end{bmatrix}$$

and

$$u_1 = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, u_2 = \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}$$

Check that  $u_1 \cdot u_2 = 0$  and then find the orthogonal projection of  $y$  into the span of  $\{u_1, u_2\}$ .

$$\hat{y} = \frac{y \cdot u_1}{u_1 \cdot u_1} u_1 + \frac{y \cdot u_2}{u_2 \cdot u_2} u_2 = \frac{3}{2} u_1 + \frac{5}{2} u_2$$

so

$$\hat{y} = \begin{bmatrix} -1 \\ 4 \\ 0 \end{bmatrix}$$

## Best approximation

Let  $W$  be a subspace of  $\mathbf{R}^n$ . Then  $\hat{y} = \text{proj}_W(y)$  is the point in  $W$  that is closest to  $y$  among all points in  $W$ .

If  $v \in W$ , then

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

for all  $v \in W$ .

$$\|y - v\|^2 = \|(y - \hat{y}) + (\hat{y} - v)\|^2 = \|(y - \hat{y})\|^2 + \|(\hat{y} - v)\|^2 + 2(y - \hat{y}) \cdot (\hat{y} - v)$$

The dot product is zero since  $y - \hat{y}$  is in  $W^\perp$  and  $\hat{y} - v$  is in  $W$ .

Therefore the minimum value occurs when  $\hat{y} - v = 0$ .

## Distance to a subspace

The distance from a point  $y$  to a subspace  $W$  is by definition the length of  $y - \hat{y}$  where  $\hat{y}$  is the orthogonal projection onto  $W$ .

## Orthonormal bases

If  $\{u_1, \dots, u_p\}$  is an orthonormal basis (meaning orthogonal, but all vectors have length one), then we can let

$$U = [u_1 \quad \cdots \quad u_p]$$

be the matrix whose columns are the  $u_i$ .

So  $U$  has  $n$  rows and  $p$  columns.

Then

$$\hat{y} = UU^T y$$

for any  $y \in \mathbf{R}^n$ . This is because  $U^T y$  is the vector whose entries are the  $u_i \cdot y$ .

$U^T y$  is a vector with  $p$  entries, and  $UU^T y$  is the sum of the columns of  $U$  – the  $u_i$  weighted by the elements of  $U^T y$ .

## Examples

$$\text{Let } u_1 = \begin{bmatrix} 1 \\ 3 \\ -2 \end{bmatrix} \text{ and } u_2 = \begin{bmatrix} 5 \\ 1 \\ 4 \end{bmatrix}.$$

These are orthogonal vectors. Let  $W$  be their span. ( $W$  is a plane in  $\mathbf{R}^3$ .)

$$\text{Let } y = \begin{bmatrix} 1 \\ 3 \\ 5 \end{bmatrix}.$$

These are orthogonal but not orthonormal.



## Examples (continued)

We can make them orthogonal by dividing them by their lengths:

$$v_1 = \frac{1}{\sqrt{14}}u_1$$

and

$$v_2 = \frac{1}{\sqrt{42}}u_2$$

Then we can make the matrix  $U$ :

$$U = \begin{bmatrix} 1/\sqrt{14} & 5/\sqrt{42} \\ 3/\sqrt{14} & 1/\sqrt{42} \\ -2/\sqrt{14} & 4/\sqrt{42} \end{bmatrix}$$

## Examples (continued)

The projection of  $y$  into  $W$  is  $UU^T y$ .

$$U^T y = \begin{bmatrix} 1/\sqrt{14} & 3/\sqrt{14} & -2/\sqrt{14} \\ 5/\sqrt{42} & 1/\sqrt{42} & 4/\sqrt{42} \end{bmatrix} \begin{bmatrix} 1 \\ 3 \\ 5 \end{bmatrix} = \begin{bmatrix} 0 \\ 28/\sqrt{42} \end{bmatrix}$$

so

$$UU^T y = \begin{bmatrix} 10/3 \\ 2/3 \\ 8/3 \end{bmatrix}$$

## Examples (continued)

The matrix  $UU^T$  computed directly is

$$\begin{bmatrix} 2/3 & 1/3 & 1/3 \\ 1/3 & 2/3 & -1/3 \\ 1/3 & -1/3 & 2/3 \end{bmatrix}$$

This is a *rank 2* matrix.

That is because its column space is  $W$  (which is two dimensional) and its null space is the one dimensional perpendicular to  $W$ .

## Examples (continued)

Find the distance from  $y = \begin{bmatrix} 5 \\ -9 \\ 5 \end{bmatrix}$  to the plane spanned by

$$u_1 = \begin{bmatrix} -3 \\ -5 \\ 1 \end{bmatrix}, u_2 = \begin{bmatrix} -3 \\ 2 \\ 1 \end{bmatrix}$$

The desired distance is

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

where  $\hat{y}$  is the projection of  $y$  into the plane spanned by  $u$ .

## Examples (continued)

Since

$$\hat{y} = \frac{y \cdot u_1}{u_1 \cdot u_1} u_1 + \frac{y \cdot u_2}{u_2 \cdot u_2} u_2$$

we get

$$\hat{y} = \frac{35}{35} u_1 + \frac{(-28)}{14} u_2 = u_1 - 2u_2 = \begin{bmatrix} 3 \\ -9 \\ -1 \end{bmatrix}$$

so

$$\|y - \hat{y}\| = \sqrt{(2)^2 + 0^2 + (6)^2} = \sqrt{40}$$