

Orthogonal Projection

Jeremy Teitelbaum

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

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