

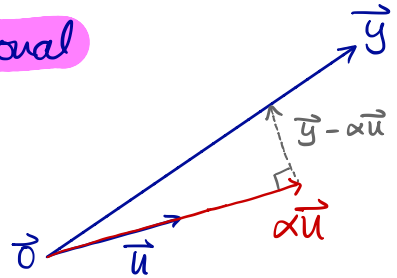
25 APR 2017

Orthogonal Projection

Given two vectors \vec{u} and \vec{y} in \mathbb{R}^n , how do we find the vector parallel to \vec{u} , let's call it $\alpha\vec{u}$, so that $\vec{u} \perp (\vec{y} - \alpha\vec{u})$?

this vector is called the **orthogonal projection** of \vec{y} onto \vec{u} .

To solve for α , we can write down the perpendicularity requirement in equation form:



$$(\vec{y} - \alpha\vec{u}) \cdot \vec{u} = 0 \quad \Leftrightarrow$$

$$\vec{y} \cdot \vec{u} - \alpha\vec{u} \cdot \vec{u} = 0 \quad \Leftrightarrow$$

$$\alpha = \frac{\vec{y} \cdot \vec{u}}{\vec{u} \cdot \vec{u}}$$

[we need to assume $\vec{u} \neq 0$ for orth. proj. to make sense]

So, we have

$$\boxed{\text{proj}_{\vec{u}} \vec{y} = \frac{\vec{y} \cdot \vec{u}}{\vec{u} \cdot \vec{u}} \vec{u}}$$

Example Find $\text{proj}_{\vec{u}} \vec{y}$ where $\vec{y} = \begin{bmatrix} 7 \\ 6 \end{bmatrix}$, $\vec{u} = \begin{bmatrix} 4 \\ 2 \end{bmatrix}$.

Solution

$$\text{we calculate } \text{proj}_{\vec{u}} \vec{y} = \frac{\begin{bmatrix} 7 \\ 6 \end{bmatrix} \cdot \begin{bmatrix} 4 \\ 2 \end{bmatrix}}{\begin{bmatrix} 4 \\ 2 \end{bmatrix} \cdot \begin{bmatrix} 4 \\ 2 \end{bmatrix}} \begin{bmatrix} 4 \\ 2 \end{bmatrix} = \boxed{\begin{bmatrix} 8 \\ 4 \end{bmatrix}}.$$

Orthonormal lists

An orthogonal list for which each vector has unit length is called **orthonormal**.

Example Normalize $\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} \right\}$.

Solution

$$\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ -1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}, \begin{bmatrix} 0 \\ 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} \right\}$$

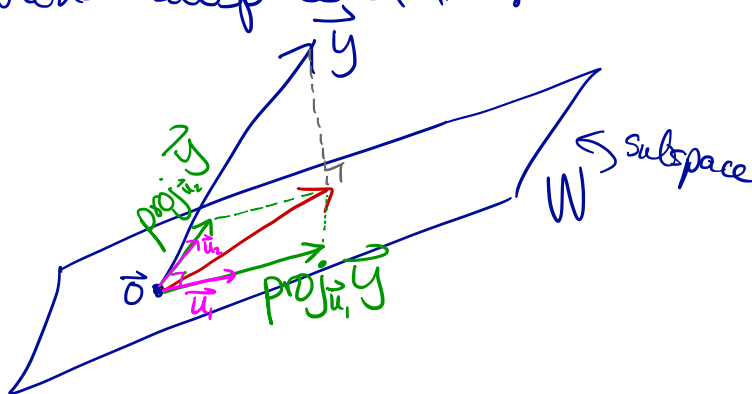
(Note: Green arrows in the original image point from the original vectors to the normalized ones with labels: $\div 1$, $\div \sqrt{2}$, and $\div \sqrt{2}$.)

Fact: A matrix has orthonormal columns iff $U^T U = I$

Proof The $(i, j)^{\text{th}}$ entry of $U^T U$ is equal to the i^{th} row of U^T (i.e., the i^{th} column of U) dotted with the j^{th} column of U . The $(i, j)^{\text{th}}$ entry of I is $\begin{cases} 1 & i=j \\ 0 & i \neq j \end{cases}$

Orthogonal Projection, higher dimension

How to project \vec{y} onto a plane or higher dimensional subspace of \mathbb{R}^n ?



Geometrically, we might be tempted to project \vec{y} onto two vectors that span W and add the results, since that makes sense in the picture (the green vectors add to give the red one). This is indeed the case:

Theorem If $\{\vec{u}_1, \dots, \vec{u}_p\}$ is an ^{orthogonal} basis for W , then $\hat{\vec{y}} = \underbrace{\frac{\vec{y} \cdot \vec{u}_1}{\vec{u}_1 \cdot \vec{u}_1} \vec{u}_1 + \dots + \frac{\vec{y} \cdot \vec{u}_p}{\vec{u}_p \cdot \vec{u}_p} \vec{u}_p}_{\text{sum of projections}}$

is in W , and $\vec{y} - \hat{\vec{y}}$ is orthogonal to W .

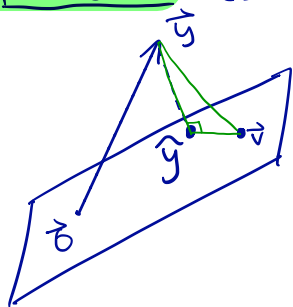
~~Proof~~ we calculate

$$\begin{aligned}(\vec{y} - \hat{y}) \cdot \vec{u}_1 &= \vec{y} \cdot \vec{u}_1 - \frac{\vec{y} \cdot \vec{u}_1}{\vec{u}_1 \cdot \vec{u}_1} \vec{u}_1 \cdot \vec{u}_1 - \underbrace{0 \dots 0}_{\text{rest of terms are zero (by orthogonality)}} \\&= \vec{y} \cdot \vec{u}_1 - \vec{y} \cdot \vec{u}_1 = 0.\end{aligned}$$

Similarly $(\vec{y} - \hat{y}) \cdot \vec{u}_2 = \dots = (\vec{y} - \hat{y}) \cdot \vec{u}_p = 0$.

So $\vec{y} - \hat{y}$ is orthogonal to each vector in a basis of W , & therefore to every vector in W .

Theorem (Best Approximation)



If $W \subset \mathbb{R}^n$ is a subspace and

$\hat{y} = \text{proj}_W \vec{y}$, then

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

for any $\vec{v} \in W$ other than \hat{y} .

~~Proof~~ $|\vec{y} - \hat{y}|^2 + |\hat{y} - \vec{v}|^2 = |\vec{y} - \vec{v}|^2$ Pythagoras, green triangle

$$\Rightarrow |\vec{y} - \hat{y}| < |\vec{y} - \vec{v}|.$$