

Today §6.3: Orthogonal Projections

Next: §6.4: Orthogonalization

Reminders: Please fill out your CAPEs.

MyMathLab Homework #8: Due March 15 by 11:59pm

MATLAB Homework #5: Due **TONIGHT** by 11:59pm

MATLAB QUIZ: Tuesday, March 13 @ your usual section time  
↑ Conflict quiz times on Monday.

A collection  $\underline{u}_1, \underline{u}_2, \dots, \underline{u}_p$  of vectors in  $\mathbb{R}^n$  is orthogonal if  $\underline{u}_i \cdot \underline{u}_j = 0$ .  
They are orthonormal if, in addition,  $\|\underline{u}_i\| = 1$  for all  $i, j$ .

Set  $U = \begin{bmatrix} \underline{u}_1 & \underline{u}_2 & \dots & \underline{u}_p \end{bmatrix}$  ( $n \times p$  matrix)

Then the  $(i, j)$ -entry of  $U^T U$  is  $\langle \underline{u}_i, \underline{u}_j \rangle$ . Therefore  
 $\{\underline{u}_1, \underline{u}_2, \dots, \underline{u}_p\}$  are orthonormal if and only if

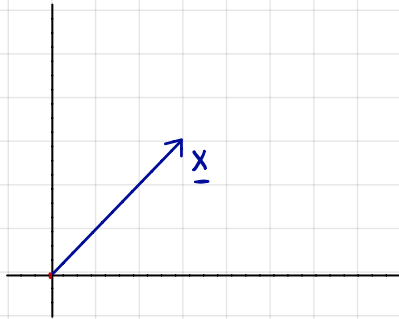
$$U^T U =$$

Note: if  $n \neq p$ , this does not mean that  $U U^T =$

Definition: If the columns of an  $n \times n$  matrix  $U$  form an orthonormal basis for  $\mathbb{R}^n$ , we call  $U$  an **orthogonal matrix**. This is equivalent to

Better yet: thinking of such a  $U$  as a linear transformation,

# Orthogonal Projections



E.g.  $\underline{u} = \begin{bmatrix} 1 \\ 2 \\ -2 \end{bmatrix}$ ,  $\underline{v} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$ ,  $\underline{w} = \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}$

What's really going on:

If  $V \subseteq \mathbb{R}^n$  is a subspace, it has an orthogonal complement  $V^\perp \subseteq \mathbb{R}^n$ . The two are complementary:  $\dim V + \dim V^\perp = n$ .

Vectors in  $V$  are linearly independent from vectors in  $V^\perp$ .

Thus, any vector  $y \in \mathbb{R}^n$  has a unique decomposition:

How do we find  $\text{Proj}_V$  for a given subspace  $V$ ?

We already saw that if  $V = \text{span}\{\underline{u}\}$  (1-dimensional)

$$\text{Proj}_V(\underline{y}) = \frac{\underline{y} \cdot \underline{u}}{\underline{u} \cdot \underline{u}} \underline{u}$$

Theorem: Let  $\{\underline{u}_1, \underline{u}_2, \dots, \underline{u}_p\}$  be an orthogonal basis for  $V$ . Then

$$\text{Proj}_V(\underline{y}) = \frac{\underline{y} \cdot \underline{u}_1}{\|\underline{u}_1\|^2} \underline{u}_1 + \frac{\underline{y} \cdot \underline{u}_2}{\|\underline{u}_2\|^2} \underline{u}_2 + \dots + \frac{\underline{y} \cdot \underline{u}_p}{\|\underline{u}_p\|^2} \underline{u}_p$$

This allows us to compute the matrix of  $\text{Proj}_V$ .

Start with an orthonormal basis for  $V$ ; then

$$\text{Proj}_V(y) = (y \cdot \underline{u}_1) \underline{u}_1 + (y \cdot \underline{u}_2) \underline{u}_2 + \dots + (y \cdot \underline{u}_p) \underline{u}_p$$

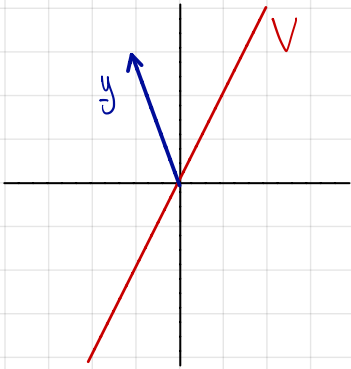
Theorem: Let  $V \subseteq \mathbb{R}^n$  be a subspace, and fix an orthonormal basis  $\{\underline{u}_1, \dots, \underline{u}_p\}$  for  $V$ . Let  $U = [\underline{u}_1 \ \underline{u}_2 \ \dots \ \underline{u}_p]$ .

$$\therefore \text{Proj}_V(y) =$$

E.g. Compute the matrix of the orthogonal projection  
in  $\mathbb{R}^3$  onto the subspace  $\text{span}\left\{\begin{bmatrix} 1 \\ -2 \\ 2 \end{bmatrix}\right\}$



What is an orthogonal projection, really?



Theorem:  $\text{Proj}_V(y)$  is the point in  $V$  that is closest to  $y$ .  
I.e. it is the best approximation of  $y$  in  $V$ .

E.g.  $\left\{ \underline{u}_1 = \begin{bmatrix} 2 \\ 5 \\ -1 \end{bmatrix}, \underline{u}_2 = \begin{bmatrix} -2 \\ 1 \\ 1 \end{bmatrix} \right\}$  is an orthogonal basis for  $V$ .  
Find the closest point in  $V$  to  $y = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$ .