# 19 – Orthogonal Projections

## Definition: Projection Onto a Line

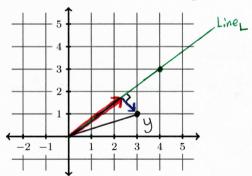
Let  $L = \text{Span}\{\mathbf{u}\}$  for some nonzero  $\mathbf{u}$  in  $\mathbb{R}^n$ . For any  $\mathbf{y}$  in  $\mathbb{R}^n$ , define the **orthogonal projection** of  $\mathbf{y}$  onto L to be

$$\operatorname{proj}_L(\mathbf{y}) = \left(\frac{\mathbf{y} \cdot \mathbf{u}}{\mathbf{u} \cdot \mathbf{u}}\right) \mathbf{u}.$$

Note: we sometimes write  $\operatorname{proj}_{\mathbf{u}}(\mathbf{y})$  in place of  $\operatorname{proj}_{L}(\mathbf{y})$ 

1. Let 
$$\mathbf{y} = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$$
 and  $L = \operatorname{Span}\left\{ \begin{bmatrix} 4 \\ 3 \end{bmatrix} \right\}$ .

(b) Let 
$$\hat{\mathbf{y}} = \operatorname{proj}_L(\mathbf{y})$$
 and  $\mathbf{b} = \mathbf{y} - \operatorname{proj}_L(\mathbf{y})$ . Graph  $\mathbf{L}, \mathbf{y}, \hat{\mathbf{y}}$ , and  $\mathbf{b}$ .



## Definition

A set of vectors  $\{\mathbf{v}_1, \dots, \mathbf{v}_k\}$  is said to be **orthogonal** if each pair of vectors in the set is orthogonal. If the set is orthogonal *and* every vector is a unit vector, then the set is said to be **orthonormal**.

**2.** Let 
$$\mathbf{v}_1 = \begin{bmatrix} 1 \\ 1/3 \\ 1/3 \end{bmatrix}$$
,  $\mathbf{v}_2 = \begin{bmatrix} -2 \\ 4 \\ 2 \end{bmatrix}$ ,  $\mathbf{v}_3 = \begin{bmatrix} -1 \\ -4 \\ 7 \end{bmatrix}$ . Verify that  $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$  is orthogonal but *not* orthonormal.

**3.** Give an example of an orthonormal set of three vectors in  $\mathbb{R}^3$ .

$$\left\{ \begin{bmatrix} 1\\ 0\\ 0 \end{bmatrix}, \begin{bmatrix} 9\\ 1\\ 0 \end{bmatrix}, \begin{bmatrix} 9\\ 1\\ 0 \end{bmatrix} \right\}$$

#### Theorem

If a set of nonzero vectors forms an orthogonal set, then the vectors are linearly independent.

## Definition: Projection Onto a Subspace

Let W be a subspace of  $\mathbb{R}^n$ , and let  $\{\mathbf{u}_1, \dots, \mathbf{u}_k\}$  be any *orthogonal* basis for W. For any  $\mathbf{y}$  in  $\mathbb{R}^n$ , define the **orthogonal projection of y onto** W to be

$$\operatorname{proj}_{W}(\mathbf{y}) = \operatorname{proj}_{\mathbf{u}_{1}}(\mathbf{y}) + \cdots + \operatorname{proj}_{\mathbf{u}_{k}}(\mathbf{y}) = \left(\frac{\mathbf{y} \cdot \mathbf{u}_{1}}{\mathbf{u}_{1} \cdot \mathbf{u}_{1}}\right) \mathbf{u}_{1} + \cdots + \left(\frac{\mathbf{y} \cdot \mathbf{u}_{k}}{\mathbf{u}_{k} \cdot \mathbf{u}_{k}}\right) \mathbf{u}_{k}.$$

Note:  $\operatorname{proj}_W(\mathbf{y})$  gives the same answer no matter which orthogonal basis you use.

- **4.** Let  $W = \text{Span}\{\mathbf{w}_1, \mathbf{w}_2\}$  for  $\mathbf{w}_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \mathbf{w}_2 = \begin{bmatrix} -1 \\ 3 \\ -2 \end{bmatrix}$ .
  - (a) Verify that  $\mathbf{w}_1, \mathbf{w}_2$  is an orthogonal basis for W.
    - $\overline{W}_1 \cdot \overline{W}_2 \ni -1 + 3 2 = 0 \rightarrow \overline{W}_1, \overline{W}_2 \text{ are orthogonal}$
    - · By theorem they are Linearly Independent and we are given that they span was so they form a basis for W.
  - (b) Compute  $\operatorname{proj}_{W}(\mathbf{y})$  for  $\mathbf{y} = \begin{bmatrix} -1 \\ 4 \\ 3 \end{bmatrix}$ .  $\operatorname{proj}_{W}(\overline{\mathbf{y}}) = \operatorname{proj}_{\overline{W}_{1}}(\overline{\mathbf{y}}) + \operatorname{proj}_{\overline{W}_{2}}(\overline{\mathbf{y}})$   $= \frac{\overline{\mathbf{y}} \cdot \overline{W}_{1}}{\overline{W}_{1}} \cdot \overline{W}_{1} + \frac{\overline{\mathbf{y}} \cdot \overline{W}_{2}}{\overline{W}_{2}} \cdot \overline{W}_{2} \longrightarrow \frac{-1 + 4 + 3}{1^{2} + 1^{2} + 1^{2}} \begin{bmatrix} \frac{1}{4} \end{bmatrix} + \frac{1 + 12 + (-6)}{(-1)^{2} + 3^{2} + (-2)^{2}} \begin{bmatrix} -1 \\ 3 \end{bmatrix}$   $\Rightarrow \frac{6}{3} \begin{bmatrix} \frac{1}{4} \end{bmatrix} + \frac{7}{14} \begin{bmatrix} -1 \\ \frac{3}{2} \end{bmatrix} = \begin{bmatrix} \frac{3}{2} \\ \frac{7}{4} \end{bmatrix}$
  - (c) Let  $\hat{\mathbf{y}} = \text{proj}_W(\mathbf{y})$  and  $\mathbf{b} = \mathbf{y} \text{proj}_W(\mathbf{y})$ . Use GeoGebra to graph  $W, \mathbf{y}, \hat{\mathbf{y}}, \hat{\mathbf{y}}$ , and  $\mathbf{b}$ .

# Theorem: Best Approximation Theorem

Let W be a subspace of  $\mathbb{R}^n$ , and let  $\hat{\mathbf{y}} = \operatorname{proj}_W(\mathbf{y})$ . Then  $\hat{\mathbf{y}}$  is the vector of W that is *closest* to  $\mathbf{y}$  in the sense that  $\operatorname{dist}(\mathbf{y}, \hat{\mathbf{y}}) \leq \operatorname{dist}(\mathbf{y}, \mathbf{w})$  for all  $\mathbf{w}$  in W.