

Orthogonal Complements

DEFINITION: If a vector \mathbf{z} is orthogonal to every vector in a subspace W of \mathbb{R}^n , then \mathbf{z} is said to be **orthogonal to W** .

EXAMPLE: Let $\mathbf{z} = \begin{bmatrix} 2 \\ 3 \end{bmatrix}$ and

$$W = \left\{ \begin{bmatrix} 6 \\ -4 \end{bmatrix} t : t \in \mathbb{R} \right\}$$

be a subspace of \mathbb{R}^2 . Then \mathbf{z} is orthogonal to every vector in W , since

$$\mathbf{z} \cdot \left(\begin{bmatrix} 6 \\ -4 \end{bmatrix} t \right) = \left(\begin{bmatrix} 2 \\ 3 \end{bmatrix} \cdot \begin{bmatrix} 6 \\ -4 \end{bmatrix} \right) t = 0 \cdot t = 0$$

EXAMPLE: Let $\mathbf{z} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$ and

$$W = \left\{ \begin{bmatrix} -1 \\ 5 \\ -3 \end{bmatrix} t_1 + \begin{bmatrix} 7 \\ 4 \\ -5 \end{bmatrix} t_2 : t_1, t_2 \in \mathbb{R} \right\}$$

be a subspace of \mathbb{R}^3 . Then \mathbf{z} is orthogonal to every vector in W , since

$$\begin{aligned} \mathbf{z} \cdot \left(\begin{bmatrix} -1 \\ 5 \\ -3 \end{bmatrix} t_1 + \begin{bmatrix} 7 \\ 4 \\ -5 \end{bmatrix} t_2 \right) \\ = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \cdot \begin{bmatrix} -1 \\ 5 \\ -3 \end{bmatrix} t_1 + \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \cdot \begin{bmatrix} 7 \\ 4 \\ -5 \end{bmatrix} t_2 = 0 \cdot t_1 + 0 \cdot t_2 = 0 \end{aligned}$$

DEFINITION: The set of all vectors \mathbf{z} that are orthogonal to W is called the **orthogonal complement** of W and is denoted by W^\perp .

EXAMPLE: Let

$$H = \left\{ \begin{bmatrix} 2 \\ 3 \end{bmatrix} s : s \in \mathbb{R} \right\}$$

and

$$W = \left\{ \begin{bmatrix} 6 \\ -4 \end{bmatrix} t : t \in \mathbb{R} \right\}$$

be subspaces of \mathbb{R}^2 . Then every vector in H is orthogonal to every vector in W , since

$$\left(\begin{bmatrix} 2 \\ 3 \end{bmatrix} s \right) \cdot \left(\begin{bmatrix} 6 \\ -4 \end{bmatrix} t \right) = \left(\begin{bmatrix} 2 \\ 3 \end{bmatrix} \cdot \begin{bmatrix} 6 \\ -4 \end{bmatrix} \right) st = 0.$$

Moreover, one can show that there are no other vectors in \mathbb{R}^2 which are orthogonal to every vector in W . Therefore $H = W^\perp$.

EXAMPLE: Let L_1 be a line through the origin in \mathbb{R}^2 , and let L_2 be the line through the origin and perpendicular to L_1 . Then each vector on L_1 is orthogonal to every vector in L_2 . Moreover, one can show that there are no other vectors in \mathbb{R}^2 which are orthogonal to every vector in L_1 . Therefore

$$L_1 = L_2^\perp$$

Also, for the same reason we have

$$L_2 = L_1^\perp$$

EXAMPLE: Let $H = \left\{ \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} s : s \in \mathbb{R} \right\}$ and

$$W = \left\{ \begin{bmatrix} -1 \\ 5 \\ -3 \end{bmatrix} t_1 + \begin{bmatrix} 7 \\ 4 \\ -5 \end{bmatrix} t_2 : t_1, t_2 \in \mathbb{R} \right\}$$

be subspaces of \mathbb{R}^3 . Then every vector in H is orthogonal to every vector in W , since

$$\begin{aligned} & \left(\begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} s \right) \cdot \left(\begin{bmatrix} -1 \\ 5 \\ -3 \end{bmatrix} t_1 + \begin{bmatrix} 7 \\ 4 \\ -5 \end{bmatrix} t_2 \right) \\ &= \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \cdot \begin{bmatrix} -1 \\ 5 \\ -3 \end{bmatrix} st_1 + \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \cdot \begin{bmatrix} 7 \\ 4 \\ -5 \end{bmatrix} st_2 = 0 \cdot st_1 + 0 \cdot st_2 = 0 \end{aligned}$$

Moreover, one can show that there are no other vectors in \mathbb{R}^3 which are orthogonal to every vector in W . Therefore $H = W^\perp$.

EXAMPLE: Let W be a plane through the origin in \mathbb{R}^3 , and let L be the line through the origin and perpendicular to W . Then each vector on L is orthogonal to every vector \mathbf{z} in W . Moreover, one can show that there are no other vectors in \mathbb{R}^3 which are orthogonal to every vector in W . Therefore

$$L = W^\perp$$

Also, for the same reason we have

$$W = L^\perp$$

THEOREM 6.10: If W is a subspace of \mathbb{R}^n , then $\mathbf{v} \in W^\perp$ if and only if \mathbf{v} is orthogonal to every vector in a spanning set for W .

THEOREM 6.11: Let W be a subspace of \mathbb{R}^n . Then W^\perp is a subspace of \mathbb{R}^n , and $W \cap W^\perp = \{\mathbf{0}\}$.

THEOREM 6.12: Let W be a subspace of \mathbb{R}^n . Let $\{\mathbf{v}_1, \dots, \mathbf{v}_k\}$ be an orthogonal basis for W contained in an orthogonal basis $\{\mathbf{v}_1, \dots, \mathbf{v}_k, \mathbf{v}_{k+1}, \dots, \mathbf{v}_n\}$ for \mathbb{R}^n . Then $\{\mathbf{v}_{k+1}, \dots, \mathbf{v}_n\}$ is an orthogonal basis for W^\perp .

COROLLARY 6.13: Let W be a subspace of \mathbb{R}^n . Then

$$\dim(W) + \dim(W^\perp) = n = \dim(\mathbb{R}^n)$$

An Orthogonal Projection

THEOREM 6.15 (Projection Theorem): Let W be a subspace of \mathbb{R}^n . Then each \mathbf{y} in \mathbb{R}^n can be written uniquely in the form

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

where $\hat{\mathbf{y}}$ is in W and \mathbf{z} is in W^\perp . In fact, if $\{\mathbf{u}_1, \dots, \mathbf{u}_p\}$ is any orthogonal basis of W , then

$$\hat{\mathbf{y}} = \frac{\mathbf{y} \cdot \mathbf{u}_1}{\mathbf{u}_1 \cdot \mathbf{u}_1} \mathbf{u}_1 + \dots + \frac{\mathbf{y} \cdot \mathbf{u}_p}{\mathbf{u}_p \cdot \mathbf{u}_p} \mathbf{u}_p$$

and $\mathbf{z} = \mathbf{y} - \hat{\mathbf{y}}$

DEFINITION: The vector $\hat{\mathbf{y}}$ is called the **orthogonal projection** of \mathbf{y} onto W and written as

$$\text{proj}_W \mathbf{y}$$

EXAMPLE: Let

$$\mathbf{u}_1 = \begin{bmatrix} 2 \\ 5 \\ -1 \end{bmatrix}, \quad \mathbf{u}_2 = \begin{bmatrix} -2 \\ 1 \\ 1 \end{bmatrix}, \quad \text{and} \quad \mathbf{y} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$$

- (i) Find the orthogonal projection of \mathbf{y} onto $W = \text{Span}\{\mathbf{u}_1, \mathbf{u}_2\}$;
- (ii) Write \mathbf{y} as the sum of a vector in W and a vector orthogonal to W .

Solution:

- (i) By the Theorem above, the orthogonal projection of \mathbf{y} onto W is

$$\begin{aligned} \hat{\mathbf{y}} &= \frac{\mathbf{y} \cdot \mathbf{u}_1}{\mathbf{u}_1 \cdot \mathbf{u}_1} \mathbf{u}_1 + \frac{\mathbf{y} \cdot \mathbf{u}_2}{\mathbf{u}_2 \cdot \mathbf{u}_2} \mathbf{u}_2 \\ &= \frac{9}{30} \begin{bmatrix} 2 \\ 5 \\ -1 \end{bmatrix} + \frac{3}{6} \begin{bmatrix} -2 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} -2/5 \\ 2 \\ 1/5 \end{bmatrix} \end{aligned}$$

- (ii) By the Theorem above we have $\mathbf{z} = \mathbf{y} - \hat{\mathbf{y}}$, therefore

$$\mathbf{z} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} - \begin{bmatrix} -2/5 \\ 2 \\ 1/5 \end{bmatrix} = \begin{bmatrix} 7/5 \\ 0 \\ 14/5 \end{bmatrix}$$

So,

$$\mathbf{y} = \begin{bmatrix} -2/5 \\ 2 \\ 1/5 \end{bmatrix} + \begin{bmatrix} 7/5 \\ 0 \\ 14/5 \end{bmatrix}$$

Application: Distance from a Point to a Subspace

DEFINITION: Let W be a subspace of \mathbb{R}^n , and assume all vectors in W have initial point at the origin. Let P be any point in n -dimensional space. Then the **minimum distance** from P to W is the shortest distance between P and the terminal point of any vector in W .

THEOREM 6.17: Let W be a subspace of \mathbb{R}^n , and let P be a point in n -dimensional space. If \mathbf{v} is the vector from the origin to P , then the minimum distance from P to W is $\|\mathbf{v} - \text{proj}_W \mathbf{v}\|$.