

Section 6.2: Orthogonal Sets

Remark: We know that if $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_p\}$ is a basis for a subspace W of \mathbb{R}^n , then each vector \mathbf{x} in W can be realized (uniquely) as a sum

$$\mathbf{x} = c_1 \mathbf{b}_1 + \dots + c_p \mathbf{b}_p.$$

If n is very large, the computations needed to determine the coefficients c_1, \dots, c_p may require a lot of time (and machine memory).

Question: Can we seek a basis whose nature simplifies this task? And what properties should such a basis possess?

Orthogonal Sets

Definition:

An indexed set $\{\mathbf{u}_1, \dots, \mathbf{u}_p\}$ in \mathbb{R}^n is said to be an **orthogonal set** provided each pair of distinct vectors in the set is orthogonal. That is, provided

$$\mathbf{u}_i \cdot \mathbf{u}_j = 0 \quad \text{whenever } i \neq j.$$

Example: Show that the set $\left\{ \begin{bmatrix} 3 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ 2 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ -4 \\ 7 \end{bmatrix} \right\}$ is an orthogonal set.

Call these $\vec{u}_1, \vec{u}_2, \vec{u}_3$ in the given order.

we'll show that $\vec{u}_1 \cdot \vec{u}_2 = 0$, $\vec{u}_1 \cdot \vec{u}_3 = 0$ and $\vec{u}_2 \cdot \vec{u}_3 = 0$

$$\left\{ \left[\begin{array}{c} \vec{u}_1 \\ 3 \\ 1 \\ 1 \end{array} \right], \left[\begin{array}{c} \vec{u}_2 \\ -1 \\ 2 \\ 1 \end{array} \right], \left[\begin{array}{c} \vec{u}_3 \\ -1 \\ -4 \\ 7 \end{array} \right] \right\}$$

$$\vec{u}_1 \cdot \vec{u}_2 = 3(-1) + 1(2) + (1)(1) = -3 + 2 + 1 = 0$$

$$\vec{u}_1 \cdot \vec{u}_3 = 3(-1) + 1(-4) + 1(7) = -3 - 4 + 7 = 0$$

$$\vec{u}_2 \cdot \vec{u}_3 = (-1)(-1) + 2(-4) + 1(7) = 1 - 8 + 7 = 0$$

Orthogonal Basis

Definition:

An **orthogonal basis** for a subspace W of \mathbb{R}^n is a basis that is also an orthogonal set.

Theorem:

Let $\{\mathbf{u}_1, \dots, \mathbf{u}_p\}$ be an orthogonal basis for a subspace W of \mathbb{R}^n . Then each vector \mathbf{y} in W can be written as the linear combination

$$\mathbf{y} = c_1\mathbf{u}_1 + c_2\mathbf{u}_2 + \cdots + c_p\mathbf{u}_p, \quad \text{where the weights } c_j = \frac{\mathbf{y} \cdot \mathbf{u}_j}{\mathbf{u}_j \cdot \mathbf{u}_j}.$$

Remark: What's nice about this is how simple the formula for the c 's is.

Example

$\left\{ \begin{bmatrix} 3 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ 2 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ -4 \\ 7 \end{bmatrix} \right\}$ is an orthogonal basis of \mathbb{R}^3 . Express the vector $\mathbf{y} = \begin{bmatrix} -2 \\ 3 \\ 0 \end{bmatrix}$ as a linear combination of the basis vectors.

$$\vec{y} \cdot \vec{u}_1 = -2(3) + 3(1) + 0 = -3$$

$$\vec{y} \cdot \vec{u}_2 = -2(-1) + 3(2) + 0 = 8$$

$$\vec{y} \cdot \vec{u}_3 = -2(-1) + 3(-4) + 0 = -10$$

$$\vec{u}_1 \cdot \vec{u}_1 = 3^2 + 1^2 + 1^2 = 11$$

$$\vec{u}_2 \cdot \vec{u}_2 = (-1)^2 + 2^2 + 1^2 = 6$$

$$\vec{u}_3 \cdot \vec{u}_3 = (-1)^2 + (-4)^2 + 7^2 = 66$$

$$\vec{y} = \frac{-3}{11} \vec{u}_1 + \frac{8}{6} \vec{u}_2 - \frac{10}{66} \vec{u}_3$$

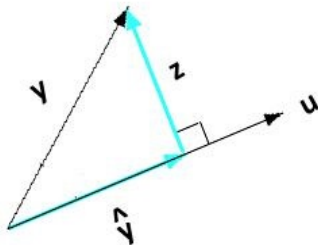
$$\vec{y} = \frac{-3}{11} \vec{u}_1 + \frac{4}{3} \vec{u}_2 - \frac{5}{33} \vec{u}_3$$

Projection

Given a nonzero vector \mathbf{u} , suppose we wish to decompose another nonzero vector \mathbf{y} into a sum of the form

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

in such a way that $\hat{\mathbf{y}}$ is parallel to \mathbf{u} and \mathbf{z} is perpendicular to \mathbf{u} .



Projection

Since $\hat{\mathbf{y}}$ is parallel to \mathbf{u} , there is a scalar α such that

$$\hat{\mathbf{y}} = \alpha \mathbf{u}.$$

Find a formula for this scalar.

Suppose $\vec{y} = \hat{\mathbf{y}} + \vec{z}$ where $\hat{\mathbf{y}} = \alpha \vec{u}$ and $\vec{u} \perp \vec{z}$

$$\vec{y} = \hat{\mathbf{y}} + \vec{z} = \alpha \vec{u} + \vec{z} \quad \text{dot w/ } \vec{u}.$$

$$\vec{u} \cdot \vec{y} = \vec{u} \cdot (\alpha \vec{u} + \vec{z}) = \alpha \vec{u} \cdot \vec{u} + \vec{u} \cdot \vec{z}$$

$$\alpha \vec{u} \cdot \vec{u} = \vec{u} \cdot \vec{y} \Rightarrow \alpha = \frac{\vec{u} \cdot \vec{y}}{\vec{u} \cdot \vec{u}} = \frac{\vec{u} \cdot \vec{y}}{\|\vec{u}\|^2}$$

$$\text{so } \hat{\mathbf{y}} = \left(\frac{\vec{u} \cdot \vec{y}}{\|\vec{u}\|^2} \right) \vec{u} \quad \text{and} \quad \vec{z} = \vec{y} - \hat{\mathbf{y}}$$

Projection onto the subspace $L = \text{Span}\{\mathbf{u}\}$

Projection Notation

We'll use the following notation for the project of a vector \mathbf{y} onto the line $L = \text{Span}\{\mathbf{u}\}$ for nonzero vector \mathbf{u} .

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

This may also be written as $\text{proj}_{\mathbf{u}} \mathbf{y}$.

This is read as “the projection of \mathbf{y} onto \mathbf{u} (or onto L).”

Example

Let $\mathbf{y} = \begin{bmatrix} 7 \\ 6 \end{bmatrix}$ and $\mathbf{u} = \begin{bmatrix} 4 \\ 2 \end{bmatrix}$. Write $\mathbf{y} = \hat{\mathbf{y}} + \mathbf{z}$ where $\hat{\mathbf{y}}$ is in $\text{Span}\{\mathbf{u}\}$ and \mathbf{z} is orthogonal to \mathbf{u} .

$$\hat{\mathbf{y}} = \left(\frac{\vec{u} \cdot \vec{y}}{\vec{u} \cdot \vec{u}} \right) \vec{u}$$

$$\vec{u} \cdot \vec{y} = 4(7) + 2(6) = 40$$

$$\vec{u} \cdot \vec{u} = 4^2 + 2^2 = 20$$

$$\hat{\mathbf{y}} = \frac{40}{20} \vec{u} = 2 \begin{bmatrix} 4 \\ 2 \end{bmatrix} = \begin{bmatrix} 8 \\ 4 \end{bmatrix}$$

$$\vec{y} = \hat{\mathbf{y}} + \vec{z} \Rightarrow \vec{z} = \vec{y} - \hat{\mathbf{y}} = \begin{bmatrix} 7 \\ 6 \end{bmatrix} - \begin{bmatrix} 8 \\ 4 \end{bmatrix} = \begin{bmatrix} -1 \\ 2 \end{bmatrix}$$

$$\therefore \vec{y} = \begin{bmatrix} 7 \\ 6 \end{bmatrix} = \begin{bmatrix} 8 \\ 4 \end{bmatrix} + \begin{bmatrix} -1 \\ 2 \end{bmatrix}$$

$\hat{\mathbf{y}}$

\vec{z}

Note that

$$\vec{u} \cdot \vec{z} = \begin{bmatrix} 4 \\ 2 \end{bmatrix} \cdot \begin{bmatrix} -1 \\ 2 \end{bmatrix} = -4 + 4 = 0$$
$$= \vec{z}^T \vec{u}$$

Example Continued...

Determine the distance between the point $(7, 6)$ and the line $\text{Span}\{\mathbf{u}\}$.

Call this L.

$$\text{dist}(L, \vec{y}) = \|\vec{z}\|$$

$$= \sqrt{(-1)^2 + 2^2} = \sqrt{5}$$

Distance between point and line

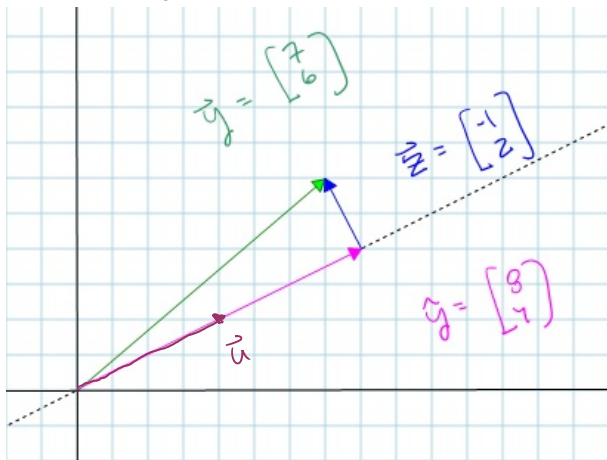


Figure: The distance between the point $(7, 6)$ and the line $\text{Span}\{u\}$ is the norm of z .

Orthonormal Sets

Definition:

A set $\{\mathbf{u}_1, \dots, \mathbf{u}_p\}$ is called an **orthonormal set** if it is an orthogonal set of **unit vectors**.

Definition:

An **orthonormal basis** of a subspace W of \mathbb{R}^n is a basis that is also an orthonormal set.

Remark: So an **orthonormal** set (or basis) is an orthogonal set (or basis) with the extra condition that each vector has norm $\sqrt{\mathbf{u}_j \cdot \mathbf{u}_j} = 1$.

Remark: Any orthogonal set can be normalized to obtain an orthonormal one.