

## Section 6.2: 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

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

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

# Orthogonal Basis

**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

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

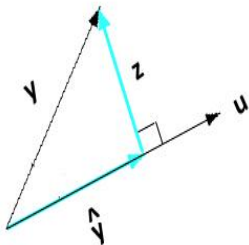
$$c_j = \frac{\mathbf{y} \cdot \mathbf{u}_j}{\mathbf{u}_j \cdot \mathbf{u}_j}.$$

## 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  $\alpha$  such that

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

Start w/  $\vec{y} = \hat{y} + \vec{z}$

Use  $\vec{z} \cdot \vec{u} = 0$  and  $\hat{y} = \alpha \vec{u}$

$$\vec{u} \cdot \vec{y} = \vec{u} \cdot (\hat{y} + \vec{z}) = \vec{u} \cdot \hat{y} + \underbrace{\vec{u} \cdot \vec{z}}_0$$

$$\Rightarrow \vec{u} \cdot \vec{y} = \vec{u} \cdot (\alpha \vec{u}) = \alpha \vec{u} \cdot \vec{u} \quad \text{since } \hat{y} = \alpha \vec{u}$$

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

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

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

**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  $\mathbf{u}$ .

$$\vec{y} \cdot \vec{u} = 7 \cdot 4 + 6 \cdot 2 = 40 \qquad \vec{u} \cdot \vec{u} = 4^2 + 2^2 = 20$$

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

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

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

## Example Continued...

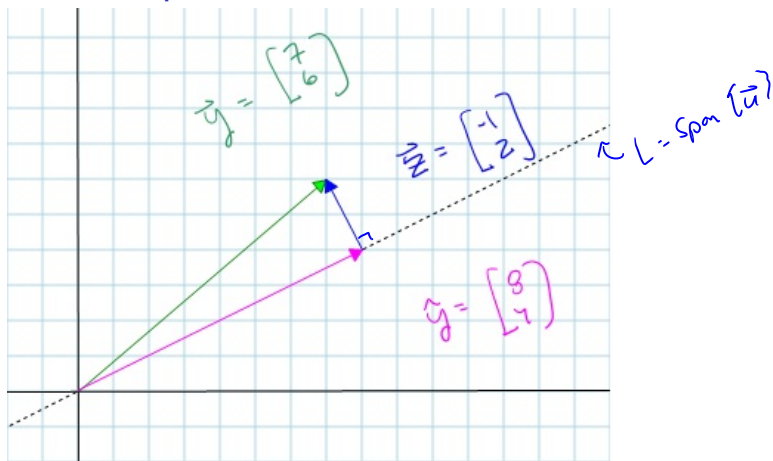
Determine the distance between the point  $(7, 6)$  and the line  $\text{Span}\{\mathbf{u}\}$ .

This distance is just

$$\text{dist.}(\hat{y}, \tilde{y}) = \|\vec{z}\| \quad \vec{z} = \begin{bmatrix} -1 \\ 2 \end{bmatrix}$$

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

## 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.



## Example

The set  $\left\{ \begin{bmatrix} \frac{3}{5} \\ \frac{4}{5} \end{bmatrix}, \begin{bmatrix} -\frac{4}{5} \\ \frac{3}{5} \end{bmatrix} \right\}$  is an orthonormal basis for  $\mathbb{R}^2$ .

Note that if  $\mathbf{u}_1 = \begin{bmatrix} \frac{3}{5} \\ \frac{4}{5} \end{bmatrix}$  and  $\mathbf{u}_2 = \begin{bmatrix} -\frac{4}{5} \\ \frac{3}{5} \end{bmatrix}$ , then

$$\mathbf{u}_1 \cdot \mathbf{u}_1 = \left(\frac{3}{5}\right)^2 + \left(\frac{4}{5}\right)^2 = 1$$

$$\mathbf{u}_1 \cdot \mathbf{u}_2 = \left(\frac{3}{5}\right)\left(-\frac{4}{5}\right) + \left(\frac{4}{5}\right)\left(\frac{3}{5}\right) = 0$$

$$\mathbf{u}_2 \cdot \mathbf{u}_2 = \left(-\frac{4}{5}\right)^2 + \left(\frac{3}{5}\right)^2 = 1$$

## Orthogonal Matrix

Consider the matrix  $U = \begin{bmatrix} \frac{3}{5} & -\frac{4}{5} \\ \frac{4}{5} & \frac{3}{5} \end{bmatrix}$  whose columns are the vectors in the last example. Compute the product

$$U^T U = \begin{bmatrix} \frac{3}{5} & \frac{4}{5} \\ -\frac{4}{5} & \frac{3}{5} \end{bmatrix} \begin{bmatrix} \frac{3}{5} & -\frac{4}{5} \\ \frac{4}{5} & \frac{3}{5} \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$U^{-1} = U^T$$

What does this say about  $U^{-1}$ ?

# Orthogonal Matrix

**Definition:** A square matrix  $U$  is called an **orthogonal matrix** if  $U^T = U^{-1}$ .

**Theorem:** An  $n \times n$  matrix  $U$  is orthogonal if and only if its columns form an orthonormal basis of  $\mathbb{R}^n$ .

The linear transformation associated to an orthogonal matrix preserves *lengths* and *angles* in the following sense: