# MTHSC 3110 Section 6.2 – Orthogonal Sets

**Kevin James** 

# DEFINITION

A set  $\{\vec{u_1},\vec{u_2},\ldots,\vec{u_k}\}$  of non-zero vectors is said to be orthogonal if for every  $i\neq j$ ,

$$\vec{u}_i \cdot \vec{u}_j = 0,$$

that is, if every pair of vectors is orthogonal.

Show that 
$$\left\{ \begin{pmatrix} 3\\1\\1 \end{pmatrix}, \begin{pmatrix} -1\\2\\1 \end{pmatrix}, \begin{pmatrix} 1\\4\\-7 \end{pmatrix} \right\}$$
 is orthogonal.

Show that 
$$\left\{ \begin{pmatrix} 3\\1\\1 \end{pmatrix}, \begin{pmatrix} -1\\2\\1 \end{pmatrix}, \begin{pmatrix} 1\\4\\-7 \end{pmatrix} \right\}$$
 . is orthogonal.

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If  $\{\vec{u}_1, \vec{u}_2, \dots, \vec{u}_k\}$  is an orthogonal set of nonzero vectors, then it is linearly independent. Hence it is a basis for the space that it spans.

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## Proof.

Suppose that  $\sum_{i=1}^k c_i \vec{u_i} = \vec{0}$ .

If  $\{\vec{u}_1, \vec{u}_2, \dots, \vec{u}_k\}$  is an orthogonal set of nonzero vectors, then it is linearly independent. Hence it is a basis for the space that it spans.

# Proof.

$$0 = \vec{u_j} \cdot \vec{0} =$$

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# $\overline{\text{P}_{\text{ROOF}}}$ .

$$0 = \vec{u_j} \cdot \vec{0} = = \vec{u_j} \cdot \sum_{i=1}^k c_i \vec{u_i}$$

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## Proof.

$$0 = \vec{u_j} \cdot \vec{0} = = \vec{u_j} \cdot \sum_{i=1}^k c_i \vec{u_i}$$
$$= \sum_{i=1}^k c_i (\vec{u_j} \cdot \vec{u_i}) =$$

If  $\{\vec{u}_1, \vec{u}_2, \dots, \vec{u}_k\}$  is an orthogonal set of nonzero vectors, then it is linearly independent. Hence it is a basis for the space that it spans.

#### Proof.

$$0 = \vec{u_j} \cdot \vec{0} = = \vec{u_j} \cdot \sum_{i=1}^{\kappa} c_i \vec{u_i}$$
$$= \sum_{i=1}^{k} c_i (\vec{u_j} \cdot \vec{u_i}) = c_j (\vec{u_j} \cdot \vec{u_j}).$$

If  $\{\vec{u}_1, \vec{u}_2, \dots, \vec{u}_k\}$  is an orthogonal set of nonzero vectors, then it is linearly independent. Hence it is a basis for the space that it spans.

#### Proof.

Suppose that  $\sum_{i=1}^{k} c_i \vec{u_i} = \vec{0}$ . Then for 1 < i < k, we have

$$0 = \vec{u_j} \cdot \vec{0} = = \vec{u_j} \cdot \sum_{i=1}^k c_i \vec{u_i}$$

$$= \sum_{i=1}^k c_i (\vec{u_j} \cdot \vec{u_i}) = c_j (\vec{u_j} \cdot \vec{u_j}).$$

$$\Rightarrow c_j = 0 \text{ (because, } \vec{u_j} \neq \vec{0}\text{)}.$$

Thus  $c_1 = c_2 = \cdots = c_k = 0$ , and our set is indeed independent.



# **DEFINITION**

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#### THEOREM

Let  $W < \mathbb{R}^n$ , and let  $S = \{\vec{u}_1, \vec{u}_2, \dots, \vec{u}_k\}$  be an orthogonal basis for W. Then if the vector  $\vec{y}$  in W is given in terms of the basis S by

$$\vec{y} = c_1 \vec{u}_1 + c_2 \vec{u}_2 + \cdots + c_k \vec{u}_k$$

then

$$c_j = \frac{\vec{y} \cdot \vec{u_j}}{\vec{u_j} \cdot \vec{u_j}}$$



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$$\vec{u}_{j} \cdot \vec{y} = \vec{u}_{j} \cdot \sum_{i=1}^{k} c_{i} \vec{u}_{i} = \sum_{i=1}^{k} c_{i} \left( \vec{u}_{j} \cdot \vec{u}_{i} \right)$$

$$= c_{j} \left( \vec{u}_{j} \cdot \vec{u}_{j} \right).$$

$$\Rightarrow c_{j} = \frac{\vec{u}_{j} \cdot \vec{y}}{\vec{u}_{j} \cdot \vec{u}_{j}}.$$

We saw above that

$$S = \left\{ \vec{v_1} = \begin{pmatrix} 3 \\ 1 \\ 1 \end{pmatrix}, \vec{v_2} = \begin{pmatrix} -1 \\ 2 \\ 1 \end{pmatrix}, \vec{v_3} = \begin{pmatrix} 1 \\ 4 \\ -7 \end{pmatrix} \right\} \text{ is an}$$

orthogonal set in  $\mathbb{R}^3$ . Since it is linearly independent and has three vectors in it, it must be a basis for  $\mathbb{R}^3$ . Express the vector

$$\vec{y} = \begin{pmatrix} 2 \\ 4 \\ 6 \end{pmatrix}$$
 in terms of the vectors in *S*.

$$\vec{y} = c_1 \vec{v_1} + c_2 \vec{v_2} + c_3 \vec{v_3}$$
, where

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, where

$$c_1 = \frac{v_1 \cdot y}{\vec{v_1} \cdot \vec{v_1}}$$

$$\vec{y} = c_1 \vec{v_1} + c_2 \vec{v_2} + c_3 \vec{v_3}$$
, where

$$c_1 = \frac{\vec{v_1} \cdot \vec{y}}{\vec{v_1} \cdot \vec{v_1}} = \frac{(3)(2) + (1)(4) + (1)(6)}{3^2 + 1^2 + 1^2} = \frac{16}{11}$$

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$$c_2 = \frac{\vec{v_2} \cdot \vec{y}}{\vec{v_2} \cdot \vec{v_2}}$$

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$$c_2 = \frac{\vec{v_2} \cdot \vec{y}}{\vec{v_2} \cdot \vec{v_2}} = \frac{(-1)(2) + (2)(4) + (1)(6)}{(-1)^2 + 2^2 + 1^2} = \frac{12}{6} = 2$$

$$\vec{y} = c_1 \vec{v_1} + c_2 \vec{v_2} + c_3 \vec{v_3}$$
, where

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$$c_{3} = \frac{\vec{v_{3}} \cdot \vec{y}}{\vec{v_{3}} \cdot \vec{v_{3}}}$$

$$\vec{y} = c_1 \vec{v_1} + c_2 \vec{v_2} + c_3 \vec{v_3}$$
, where

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$$c_{3} = \frac{\vec{v_{3}} \cdot \vec{y}}{\vec{v_{3}} \cdot \vec{v_{3}}} = \frac{(1)(2) + (4)(4) + (-7)(6)}{1^{2} + 4^{2} + (-7)^{2}} = \frac{-24}{66} = \frac{-4}{11}$$

# **Orthogonal Projections**

Earlier we saw how to find the component of  $\vec{v}$  in the direction of  $\vec{u}$  and the component orthogonal to  $\vec{u}$ . We revisit this idea to introduce some notation, and to extend it to projecting onto a subspace.

Given  $\vec{y} \in \mathbb{R}^n$  and  $\vec{u} \in \mathbb{R}^n$ , find  $\hat{y}, \vec{z} \in \mathbb{R}^n$  so that

- $\mathbf{2} \ \hat{\mathbf{y}} = \alpha \vec{\mathbf{u}} \ (\alpha \in \mathbb{R}).$
- $\vec{\mathbf{3}} \ \vec{u} \cdot \vec{z} = 0.$

As before, we see that

$$\vec{y} \cdot \vec{u} = \hat{y} \cdot \vec{u} = (\alpha \vec{u}) \cdot \vec{u} = \alpha ||\vec{u}||^2$$

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so

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

and

$$\vec{z} = \vec{y} - \frac{\vec{y} \cdot \vec{u}}{\|\vec{u}\|^2} \vec{u}$$

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# NOTATION

$$\hat{y} = \mathsf{Proj}_L(\vec{y}) = \frac{\vec{y} \cdot \vec{u}}{\|\vec{u}\|^2} \vec{u}$$

where  $L = \operatorname{Span}(\vec{u})$ .

# TERMINOLOGY

 $\vec{y} = \hat{y} + \vec{z}$ :

 $\hat{y}$  is the orthogonal projection of  $\vec{y}$  onto L

 $\vec{z}$  is the component of  $\vec{y}$  orthogonal to L.

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# EXAMPLE

Let  $\vec{y} = \begin{pmatrix} 7 \\ 6 \end{pmatrix}$ ,  $\vec{u} = \begin{pmatrix} 4 \\ 2 \end{pmatrix}$  and  $L = \operatorname{Span}(\vec{u})$ . Compute the orthogonal projection of  $\vec{y}$  onto L and the component of  $\vec{y}$  orthogonal to L. Plot  $\vec{y}, \vec{u}, \hat{y}$  and  $\vec{z}$ . Compute the distance from  $\vec{y}$  to L. (Note: the subspace L here is the line through  $\vec{0}$  and  $\vec{u}$ .)

# Geometric Interpretation of Theorem 5

Let  $\{\vec{u}_1, \vec{u}_2\}$  be an orthogonal basis for  $\mathbb{R}^2$ . Put

$$\hat{y_1} = rac{ec{y} \cdot ec{u_1}}{\|ec{u_1}\|^2} ec{u_1} = \mathsf{Proj}_{ec{u_1}}(ec{y})$$

$$\hat{y_2} = \frac{\vec{y} \cdot \vec{u_2}}{\|\vec{u_2}\|^2} \vec{u_2} = \mathsf{Proj}_{\vec{u_2}}(\vec{y})$$

Then

$$\vec{y} = \hat{y_1} + \hat{y_2}.$$

# DEFINITION

Orthonormal Sets A set  $\vec{u}_1, \vec{u}_2, \ldots, \vec{u}_p$  is called *orthonormal* if it is orthogonal and  $\|\vec{u}_i\| = 1$  for  $1 \leq i \leq p$ . In this case, if  $W = \operatorname{Span}(\vec{u}_1, \ldots, \vec{u}_p)$ , then the set is called an *orthonormal basis* for W.

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The set  $\vec{e_1}, \vec{e_2}, \dots, \vec{e_n}$  is an orthonormal basis for  $\mathbb{R}^n$ .

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# EXAMPLE

Show that the set  $\{\vec{v}_1, \vec{v}_2, \vec{v}_3\}$  is an orthornormal basis for  $\mathbb{R}^3$ , where

$$\vec{v}_1 = \left( \begin{array}{c} 3/\sqrt{11} \\ 1/\sqrt{11} \\ 1/\sqrt{11} \end{array} \right), \quad \vec{v}_2 = \left( \begin{array}{c} -1/\sqrt{6} \\ 2/\sqrt{6} \\ 1/\sqrt{6} \end{array} \right), \quad \vec{v}_3 = \left( \begin{array}{c} 1/\sqrt{66} \\ 4/\sqrt{66} \\ -7/\sqrt{66} \end{array} \right).$$



An  $m \times n$  matrix U has orthonormal columns if and only if  $U^T U = I$ .

**Proof:** proof strategy: interpret the entries of  $U^TU$  in terms of inner products of the columns of U.

Let U be an  $m \times n$  matrix with orthonormal columns, and  $\vec{x}, \vec{y} \in \mathbb{R}^n$ . Then

- $(U\vec{x}) \cdot (U\vec{y}) = \vec{x} \cdot \vec{y}.$
- $(U\vec{x}) \cdot (U\vec{y}) = 0 \text{ if and only if } \vec{x} \cdot \vec{y} = 0.$

Let U be an  $m \times n$  matrix with orthonormal columns, and  $\vec{x}, \vec{y} \in \mathbb{R}^n$ . Then

- $(U\vec{x}) \cdot (U\vec{y}) = \vec{x} \cdot \vec{y}.$
- 3  $(U\vec{x}) \cdot (U\vec{y}) = 0$  if and only if  $\vec{x} \cdot \vec{y} = 0$ .

#### Note

- 1 U preserves length
- U preserves orthonormality.

Let U be an  $m \times n$  matrix with orthonormal columns, and  $\vec{x}, \vec{y} \in \mathbb{R}^n$ . Then

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#### Note

- 1 U preserves length
- 2 U preserves orthonormality.

# Proof.

$$(U\vec{x}) \cdot (U\vec{y}) = (U\vec{x})^T (U\vec{y}) = (\vec{x}^T U^T)(U\vec{y}) = \vec{x}^T (U^T U)\vec{y} = \vec{x} \cdot \vec{y}$$

This proves part 2. 1 & 3 follow from 2.



# EXAMPLE

Let

$$U = \begin{pmatrix} 1/\sqrt{2} & 2/3 \\ 1/\sqrt{2} & -2/3 \\ 0 & 1/3 \end{pmatrix} \quad \text{and} \quad \vec{x} = \begin{pmatrix} \sqrt{2} \\ 3 \end{pmatrix}.$$

Check that U has orthonormal columns and that  $||U\vec{x}|| = ||\vec{x}||$ .

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#### Note

In the case where U is a an  $n \times n$  matrix with orthonormal columns, we see that  $U^T U = I$  so  $U^T = U^{-1}$ , so  $UU^T = I$  as well: that is, U has orthonormal rows too!