

# Unit 2J: Orthonormal Bases

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# Summary

- ① Orthogonal and orthonormal sets
- ② Orthogonal and orthonormal bases
- ③ Orthogonal Projection
- ④ Orthonormality and Matrices

# Orthogonal and orthonormal sets

## Orthogonal set

A set  $S$  of vectors in  $\mathbb{R}^n$  is called **orthogonal** if every pair of distinct vectors in  $S$  are orthogonal.

A set  $S$  of vectors in  $\mathbb{R}^n$  is called **orthonormal** if  $S$  is orthogonal and every vector in  $S$  is a unit vector.

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## Example.

It is easy to check that the standard basis  $E = \{e_1, e_2, \dots, e_n\}$  for  $\mathbb{R}^n$  is an orthonormal set.

### Example.

Let  $u_1 = (2, 0, 0)$ ,  $u_2 = (0, 1, 1)$  and  $u_3 = (0, 1, -1)$ . Find the orthonormal set associated with the set  $S = \{u_1, u_2, u_3\}$ .

**Solution:**

We have

$$u_1 \cdot u_2 = 2 \cdot 0 + 0 \cdot 1 + 0 \cdot 0 = 0$$

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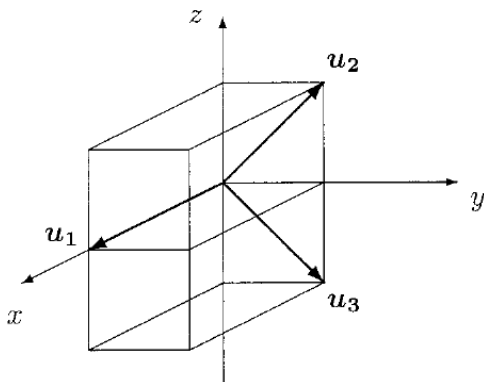
$$u_1 \cdot u_2 = 2 \cdot 0 + 0 \cdot 1 + 0 \cdot 0 = 0$$

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$$u_2 \cdot u_3 = 0 \cdot 0 + 1 \cdot 1 + 1 \cdot (-1) = 0.$$

Hence the set  $S$  is orthogonal.

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Then for  $i = 1, 2, 3$

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Therefore, the set  $\{v_1, v_2, v_3\}$  is orthonormal.  $\square$

## Note that

The process of converting an orthogonal set to an orthonormal set by multiplying each vector  $u$  by  $\frac{1}{\|u\|}$  is called **normalizing**.

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## Theorem

An orthogonal set of nonzero vectors in a vector space is linearly independent.

## Proof.

Let  $S = \{v_1, v_2, \dots, v_k\}$ .

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$$c_1v_1 + c_2v_2 + \dots + c_kv_k = 0.$$

Since  $S$  is orthogonal,  $v_i \cdot v_j = 0$  for  $i \neq j$ . For  $i = 1, 2, \dots, k$ ,

$$0 = (c_1v_1 + c_2v_2 + \dots + c_kv_k) \cdot v_i$$

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$$\begin{aligned} 0 &= (c_1v_1 + c_2v_2 + \dots + c_kv_k) \cdot v_i \\ &= (c_1v_1) \cdot v_i + (c_2v_2) \cdot v_i + \dots + (c_kv_k) \cdot v_i \end{aligned}$$



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$$\begin{aligned} 0 &= (c_1v_1 + c_2v_2 + \dots + c_kv_k) \cdot v_i \\ &= (c_1v_1) \cdot v_i + (c_2v_2) \cdot v_i + \dots + (c_kv_k) \cdot v_i \\ &= c_1(v_1 \cdot v_i) + c_2(v_2 \cdot v_i) + \dots + c_k(v_k \cdot v_i) \end{aligned}$$

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Given that  $v_i \neq 0$  for each  $i$ , we have  $v_i \cdot v_i \neq 0$ . This means,  $c_i = 0$  for each  $i$ . Therefore,  $S$  is linearly independent. □

# Orthogonal and orthonormal bases

## Orthogonal basis

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## Example

Let  $u_1 = (2, 0, 0)$ ,  $u_2 = (0, 1, 1)$  and  $u_3 = (0, 1, -1)$ .

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- We have shown that the set  $S = \{u_1, u_2, u_3\}$  is orthogonal.



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- We have shown that the set  $S = \{u_1, u_2, u_3\}$  is orthogonal. It is clear that each vector in  $S$  is nonzero. Hence by the previous theorem,  $S$  is linearly independent.
- We also know that any linearly independent set of  $n$  elements of an  $n$ -dimensional vector space  $V$  must span  $V$  and thus is a basis.

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- We also know that any linearly independent set of  $n$  elements of an  $n$ -dimensional vector space  $V$  must span  $V$  and thus is a basis. Therefore,  $S$  is an orthogonal basis for  $\mathbb{R}^3$ .
- Moreover, the orthonormal set associated with  $S$  is an orthonormal basis for  $\mathbb{R}^3$ .

## Theorem

Let  $B = \{v_1, v_2, \dots, v_k\}$  be an orthogonal basis for a vector space  $V$ . Then for any vector  $w \in V$ ,

$$w = \frac{w \cdot v_1}{v_1 \cdot v_1} v_1 + \frac{w \cdot v_2}{v_2 \cdot v_2} v_2 + \dots + \frac{w \cdot v_k}{v_k \cdot v_k} v_k,$$

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that is,

$$w = \left( \frac{w \cdot v_1}{v_1 \cdot v_1}, \frac{w \cdot v_2}{v_2 \cdot v_2}, \dots, \frac{w \cdot v_k}{v_k \cdot v_k} \right)$$

with respect to Basis  $B$ .

## Theorem...

Moreover, if  $B = \{v_1, v_2, \dots, v_k\}$  is an orthonormal basis for a vector space  $V$ , then for any vector  $w \in V$ ,

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## Proof.

Let  $B = \{v_1, v_2, \dots, v_k\}$  be an orthogonal basis for a vector space  $V$  and  $w \in V$ .

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for some real numbers  $c_1, c_2, \dots, c_k$ .

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$$w \cdot v_i = (c_1v_1 + c_2v_2 + \dots + c_kv_k) \cdot v_i$$

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$$\begin{aligned} w \cdot v_i &= (c_1v_1 + c_2v_2 + \dots + c_kv_k) \cdot v_i \\ &= (c_1v_1) \cdot v_i + (c_2v_2) \cdot v_i + \dots + (c_kv_k) \cdot v_i \end{aligned}$$

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## Proof...

Given that  $v_i \neq 0$  for each  $i$ , we have  $v_i \cdot v_i \neq 0$ .

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Therefore,

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for each  $i$ .

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Therefore,

$$c_i = \frac{w \cdot v_i}{v_i \cdot v_i}$$

for each  $i$ . That means,

$$w = \frac{w \cdot v_1}{v_1 \cdot v_1} v_1 + \frac{w \cdot v_2}{v_2 \cdot v_2} v_2 + \dots + \frac{w \cdot v_k}{v_k \cdot v_k} v_k.$$

## Proof ...

That is,

$$w = \left( \frac{w \cdot v_1}{v_1 \cdot v_1}, \frac{w \cdot v_2}{v_2 \cdot v_2}, \dots, \frac{w \cdot v_k}{v_k \cdot v_k} \right)$$

with respect to Base  $B$ .

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with respect to Base  $B$ .

If  $B = \{v_1, v_2, \dots, v_k\}$  is an orthonormal basis for a vector space  $V$ , then the result follows from the previous result, because  $v_i \cdot v_i = \|v_i\|_2 = 1$  for all  $i$ . □

## Example

Let

$$v_1 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \quad v_2 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ -1 \end{pmatrix}.$$

Show that  $B = \{v_1, v_2\}$  is an orthonormal basis for  $\mathbb{R}^2$ . Find a vector  $x \in \mathbb{R}^2$  with respect to the basis  $B$ .

## Solution:

We have

$$v_1 \cdot v_2 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix} \cdot \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

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We have

$$v_1 \cdot v_2 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix} \cdot \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ -1 \end{pmatrix} = \frac{1}{2} (1 \cdot 1 + 1 \cdot (-1)) = 0.$$



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Also, we have

$$v_1 \cdot v_1 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix} \cdot \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

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Also, we have

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Similarly,

$$v_2 \cdot v_2 = 1.$$

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To find a vector  $x \in \mathbb{R}^2$  with respect to the basis  $B$ , assume that  $x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$  with respect to the standard basis for  $\mathbb{R}^2$ .

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$$v_1 \cdot x = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix} \cdot \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

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$$v_1 \cdot x = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix} \cdot \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \frac{1}{\sqrt{2}}(x_1 + x_2)$$

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and

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Therefore,  $B = \{v_1, v_2\}$  is an orthonormal basis for  $\mathbb{R}^2$ .

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and

$$v_2 \cdot x = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ -1 \end{pmatrix} \cdot \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \frac{1}{\sqrt{2}}(x_1 - x_2).$$

## Solution...

Therefore,

$$x = \left( \frac{1}{\sqrt{2}}(x_1 + x_2), \frac{1}{\sqrt{2}}(x_1 - x_2) \right)$$

with respect to the basis  $B$ .

## Example

Let  $V$  be a plane in  $\mathbb{R}^3$  defined by the equation

$ax + by + cz = 0$ . Let  $n = \begin{pmatrix} a \\ b \\ c \end{pmatrix}$ . For any vector

$u = \begin{pmatrix} x \\ y \\ z \end{pmatrix} \in V$ , we have

$$n \cdot u = ax + by + cz = 0.$$

Thus  $n$  is orthogonal to  $V$ .

## Example ...

In fact,

$$\begin{aligned} V &= \{(x, y, z) \in \mathbb{R}^3 : ax + by + cz = 0\} \\ &= \{u \in \mathbb{R}^3 \mid n \cdot u = 0\}. \end{aligned}$$

The vector  $n$  is a normal vector of  $V$ .

## Example

Let  $V = \text{span}\{u_1, u_2\}$  be a subspace of  $\mathbb{R}^4$ , where

$$u_1 = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 0 \end{pmatrix}, u_2 = \begin{pmatrix} 0 \\ -1 \\ -1 \\ 1 \end{pmatrix}. \text{ Find all vectors that}$$

are orthogonal to  $V$ .

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$$\Leftrightarrow \begin{cases} w + x + y = 0 \\ -x - y + z = 0 \end{cases}$$

$$\Leftrightarrow (w, x, y, z) = (-t, -s + t, s, t) \text{ for some } s, t \in \mathbb{R}.$$

## Solution ...

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Therefore,  $\text{span}\{(0, -1, 1, 0), (-1, 1, 0, 1)\}$  is orthogonal to  $V$ . □

# Orthogonal projection

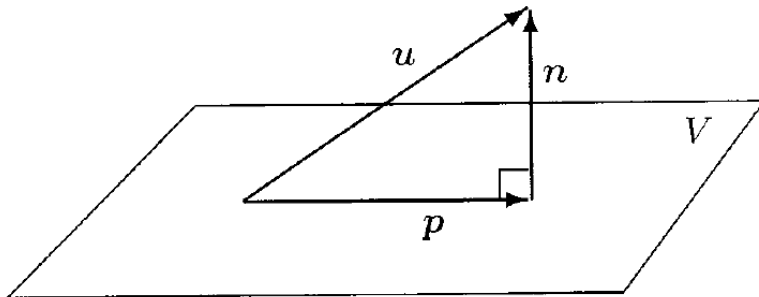
## Orthogonal projection

Let  $V$  be a subspace of  $\mathbb{R}^n$ . Every vector  $u \in \mathbb{R}^n$  can be written uniquely as

$$u = n + p$$

such that  $n$  is a vector orthogonal to  $V$  and  $p$  is a vector in  $V$ . The vector  $p$  is called the **(orthogonal) projection** of  $u$  onto  $V$ .





The projection of

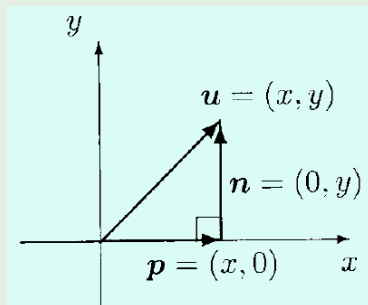
$$u = \begin{pmatrix} x \\ y \end{pmatrix}$$

onto the  $x$ -axis is

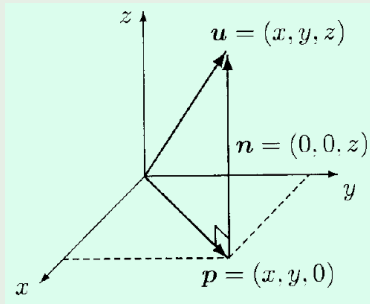
$$p = \begin{pmatrix} x \\ 0 \end{pmatrix}.$$

Here,

$$n = \begin{pmatrix} 0 \\ y \end{pmatrix}.$$



The projection of  $u = \begin{pmatrix} x \\ y \\ z \end{pmatrix}$  onto the  $xy$ -plane  
is  $p = \begin{pmatrix} x \\ y \\ 0 \end{pmatrix}$ . Here,  $n = \begin{pmatrix} 0 \\ 0 \\ z \end{pmatrix}$ .



## Theorem

Let  $V$  be a subspace of  $\mathbb{R}^n$  and  $w$  a vector in  $\mathbb{R}^n$ .

(a) If  $\{v_1, v_2, \dots, v_k\}$  be an orthogonal basis for  $V$ , then

$$\frac{w \cdot v_1}{v_1 \cdot v_1} v_1 + \frac{w \cdot v_2}{v_2 \cdot v_2} v_2 + \dots + \frac{w \cdot v_k}{v_k \cdot v_k} v_k,$$

is the projection of  $w$  onto  $V$ .

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is the projection of  $w$  onto  $V$ .

- (b) If  $\{v_1, v_2, \dots, v_k\}$  is an orthonormal basis for  $V$ , then

$$(w \cdot v_1)v_1 + (w \cdot v_2)v_2 + \dots + (w \cdot v_k)v_k,$$

is the projection of  $w$  onto  $V$ .

Proof.

(a) Let

$$p = \frac{w \cdot v_1}{v_1 \cdot v_1} v_1 + \frac{w \cdot v_2}{v_2 \cdot v_2} v_2 + \dots + \frac{w \cdot v_k}{v_k \cdot v_k} v_k, \quad n = w - p.$$

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Then for  $i = 1, 2, \dots, k$ ,

$$n \cdot v_i = w \cdot v_i - p \cdot v_i$$

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Then for  $i = 1, 2, \dots, k$ ,

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This means,  $n$  is orthogonal to  $V$ . Since  $w = n + p$  where  $n \perp V$  and  $p \in V$ ,  $p$  is the projection of  $w$  onto  $V$ .

**Proof...**

(b) This is a consequence of part (a), because

$$v_i \cdot v_i = \|v_i\|_2^2 = 1.$$



## Example

Let  $V$  be a subspace of  $\mathbb{R}^3$  spanned by the orthogonal vectors  $v_1 = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$  and  $v_2 = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$ . Find the projection of  $w = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$  onto  $V$ .

### Solution:

The projection of  $w = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$  onto  $V$  is equal to

$$\frac{w \cdot v_1}{v_1 \cdot v_1} v_1 + \frac{w \cdot v_2}{v_2 \cdot v_2} v_2 = \frac{1}{2} \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} + \frac{1}{2} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}.$$

# Gram-Schmidt Process

## Discussion

- (a) Let  $\{u_1, u_2\}$  be a basis for a vector space  $V$  where  $V$  is either  $\mathbb{R}^2$  or a plane in  $\mathbb{R}^3$  containing the origin.

# Gram-Schmidt Process

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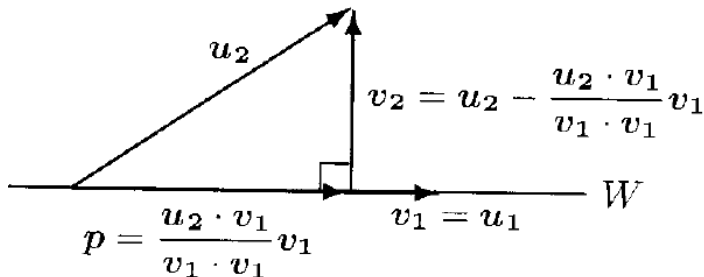
# Gram-Schmidt Process

## Discussion

- (a) Let  $\{u_1, u_2\}$  be a basis for a vector space  $V$  where  $V$  is either  $\mathbb{R}^2$  or a plane in  $\mathbb{R}^3$  containing the origin. Let  $W$  be the subspace of  $V$  spanned by  $u_1$ . ( $W$  is a line through the origin.) Then the projection of  $u_2$  onto  $W$  is

$$p = \frac{u_2 \cdot u_1}{u_1 \cdot u_1} u_1$$

Define  $v_1 = u_1$  and  $v_2 = u_2 - p = u_2 - \frac{u_2 \cdot v_1}{v_1 \cdot v_1} v_1$ .



Then  $\{v_1, v_2\}$  is an orthogonal basis for  $V$ .

- (b) Let  $\{u_1, u_2, u_3\}$  be a basis for  $\mathbb{R}^3$  and let  $V$  be the subspace of  $\mathbb{R}^3$  spanned by  $u_1, u_2$ . Let  $W$  be the subspace of  $V$  spanned by  $u_1$ . ( $V$  is a plane containing the origin.) Define

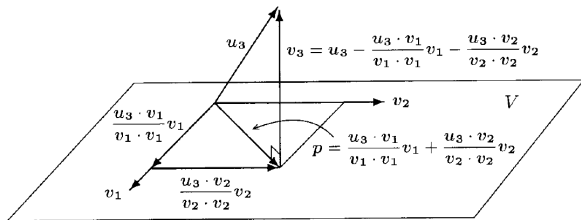
$$v_1 = u_1, \quad v_2 = u_2 - \frac{u_2 \cdot v_1}{v_1 \cdot v_1} v_1.$$

By Part (a),  $\{v_1, v_2\}$  is an orthogonal basis for  $V$ .  
Then the projection of  $u_3$  onto  $V$  is

$$p = \frac{u_3 \cdot v_1}{v_1 \cdot v_1} v_1 + \frac{u_3 \cdot v_2}{v_2 \cdot v_2} v_2.$$

Define

$$v_3 = u_3 - p = u_3 - \frac{u_3 \cdot v_1}{v_1 \cdot v_1} v_1 - \frac{u_3 \cdot v_2}{v_2 \cdot v_2} v_2.$$



Then  $\{v_1, v_2, v_3\}$  is an orthogonal basis for  $\mathbb{R}^3$ .

In general, we have the following process, known as **Gram-Schmidt Process**:

### Gram-Schmidt Process:

Let  $\{u_1, u_2, \dots, u_k\}$  be a basis for a vector space  $V$ . Define

$$v_1 = u_1,$$

$$v_2 = u_2 - \frac{u_2 \cdot v_1}{v_1 \cdot v_1} v_1,$$

$$v_3 = u_3 - \frac{u_3 \cdot v_1}{v_1 \cdot v_1} v_1 - \frac{u_3 \cdot v_2}{v_2 \cdot v_2} v_2,$$

$$\vdots$$

$$v_k = u_k - \frac{u_k \cdot v_1}{v_1 \cdot v_1} v_1 - \frac{u_k \cdot v_2}{v_2 \cdot v_2} v_2 - \dots - \frac{u_k \cdot v_{k-1}}{v_{k-1} \cdot v_{k-1}} v_{k-1}$$

Then  $\{v_1, v_2, v_3\}$  is an orthogonal basis for  $V$ .  
Furthermore, let

$$w_1 = \frac{1}{\|v_1\|}v_1, w_2 = \frac{1}{\|v_2\|}v_2, \dots, w_k = \frac{1}{\|v_k\|}v_k.$$

Then  $\{w_1, w_2, w_3\}$  is an orthonormal basis for  $V$ .

# Orthonormality and Matrices

Let  $\{v_1, v_2, \dots, v_k\}$  be an orthonormal basis for vector space  $V$ . Then

$$v_i \cdot v_j = v_i^T v_j = 0 \text{ if } i \neq j$$

$$\|v_i\| = \sqrt{v_i^T v_i} = 1 \text{ for all } i.$$



If we use the relatively common notation that

$$\delta_{ij} = \begin{cases} 1 & \text{if } i \neq j \\ 0 & \text{if } i = j. \end{cases}$$

then the two normality conditions can be written compactly as

$$v_i \cdot v_j = \delta_{ij} \quad \text{for all } i \in \{1, 2, \dots, k\}.$$

To express the two orthonormality conditions in terms of conditions on matrices, let's define an  $n \times k$  matrix  $A$  to be

$$A = (v_1 \quad v_2 \quad \cdots \quad v_k),$$

where  $v_i = A_{:i}$  is the  $i$ th column of  $A$ .

Next, let's consider the matrix  $A^T A$ . Observe that  $A^T A$  is a  $k \times k$  matrix. Let's ask what information is contained in the  $(ij)$  element of this matrix, i.e., in the element  $(A^T A)_{ij}$  ? That is,

$$(A^T A)_{ij} = v_i^T v_j,$$

which, in this case, equals 1 or 0, depending on whether or not  $i = j$ .

That is, in this case,

$$(A^T A)_{ij} = \delta_{ij}$$

and we can write the matrix product as

$$A^T A = I_k,$$

where  $I_k$  is the identity matrix of dimension  $k$ .  
This is the matrix way to express that a matrix  
has columns that form an orthonormal basis.

Given an  $n \times n$  matrix  $A$ , recall that the inverse matrix  $A^{-1}$  is the matrix such that

$$A^{-1}A = AA^{-1} = I_n.$$

So, we have the following definition:

### Orthogonal matrix

A square matrix  $A$  is called **orthogonal** if

$$A^{-1} = A^T.$$

## Example

Let

$$A = \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix}.$$

Determine if  $A$  is an orthogonal matrix.

## Example

Let

$$B = \begin{pmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{pmatrix}.$$

Determine if  $B$  is an orthogonal matrix.

## Practice Problems

- 1 Determine if the matrix

$$\begin{bmatrix} 2 & 3 \\ 1 & 2 \end{bmatrix}$$

is an orthogonal matrix.

- 2 Prove that the product of two orthogonal matrices is orthogonal.

- 3 If  $A = \frac{1}{3} \begin{pmatrix} 1 & 2 & 2 \\ 2 & 1 & -2 \\ x & 2 & y \end{pmatrix}$  is an orthogonal matrix, then find the values of  $x$  and  $y$ .





From the above discussion, we have the following result:

Let  $A$  be a square matrix of order  $n$ . The following statements are equivalent:

- ❶  $A$  is orthogonal.
- ❷ The columns of  $A$  form an orthonormal basis for  $\mathbb{R}^n$ .
- ❸ The rows of  $A$  form an orthonormal basis for  $\mathbb{R}^n$ .

Suppose that we have a set of vectors  $v_1, v_2, \dots, v_k \in \mathbb{R}^n$ , such that any two different vectors  $v_i$  and  $v_j$  in that set are orthogonal to each other. To organize this information into a matrix, let's define an  $n \times k$  matrix  $A$  to be

$$A = (v_1 \quad v_2 \quad \dots \quad v_k).$$

The requirement that the basis vectors be orthogonal to each other means that

$$A^T A = D,$$

where  $D$  is a  $k \times k$  diagonal matrix, all the diagonal entries of which are positive.

The matrix  $D$  is diagonal since the off-diagonal elements are the dot products between different basis vectors, which equal zero, since they are orthogonal; and the diagonal entries are all non-zero since each vector in the basis is non-zero and thus has a non-zero non-negative norm.