Linear Algebra and Application

ANLT 201

Overview

- Orthogonal Projections
- Reading 6.2 and 6.3 from Lay

Show that $\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$ is an orthogonal set, where

$$\mathbf{u}_1 = \begin{bmatrix} 3 \\ 1 \\ 1 \end{bmatrix}, \quad \mathbf{u}_2 = \begin{bmatrix} -1 \\ 2 \\ 1 \end{bmatrix}, \quad \mathbf{u}_3 = \begin{bmatrix} -1/2 \\ -2 \\ 7/2 \end{bmatrix}$$

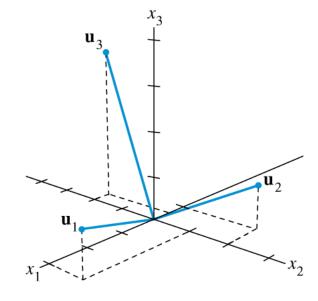


FIGURE 1

Solution

$$\mathbf{u}_1 = \begin{bmatrix} 3 \\ 1 \\ 1 \end{bmatrix}, \quad \mathbf{u}_2 = \begin{bmatrix} -1 \\ 2 \\ 1 \end{bmatrix}, \quad \mathbf{u}_3 = \begin{bmatrix} -1/2 \\ -2 \\ 7/2 \end{bmatrix}$$

$$\mathbf{u}_{1} \cdot \mathbf{u}_{2} = 3(-1) + 1(2) + 1(1) = 0$$

$$\mathbf{u}_{1} \cdot \mathbf{u}_{3} = 3\left(-\frac{1}{2}\right) + 1(-2) + 1\left(\frac{7}{2}\right) = 0$$

$$\mathbf{u}_{2} \cdot \mathbf{u}_{3} = -1\left(-\frac{1}{2}\right) + 2(-2) + 1\left(\frac{7}{2}\right) = 0$$

Each pair of distinct vectors is orthogonal, and so $\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$ is an orthogonal set. See Fig. 1; the three line segments there are mutually perpendicular.

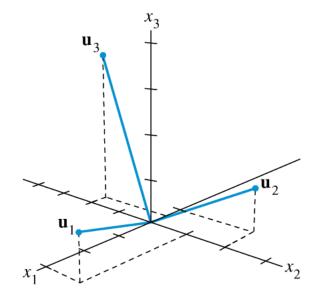
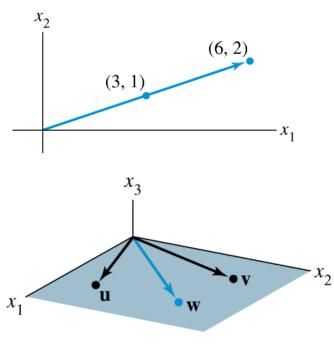


FIGURE 1

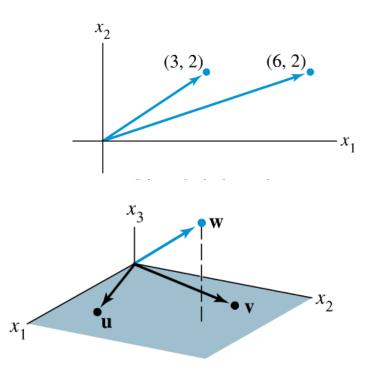
Definition: Orthogonal Set

A set of vectors $\{\mathbf{u}_1, \dots, \mathbf{u}_p\}$ in \mathbb{R}^n is said to be an **orthogonal set** if each pair of distinct vectors from the set is orthogonal, that is, if $\mathbf{u}_i \cdot \mathbf{u}_j = 0$ whenever $i \neq j$.

Linear Independence



Linearly dependent, w in Span{u, v}



Linearly independent, w not in Span{u, v}

Definition: Linear Independence

An indexed set of vectors $\{\mathbf{v}_1, \dots, \mathbf{v}_p\}$ in \mathbb{R}^n is said to be **linearly independent** if the vector equation

$$x_1\mathbf{v}_1 + x_2\mathbf{v}_2 + \dots + x_p\mathbf{v}_p = \mathbf{0}$$

has only the trivial solution. The set $\{\mathbf{v}_1, \dots, \mathbf{v}_p\}$ is said to be **linearly dependent** if there exist weights c_1, \dots, c_p , not all zero, such that

$$c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2 + \dots + c_p \mathbf{v}_p = \mathbf{0} \tag{2}$$

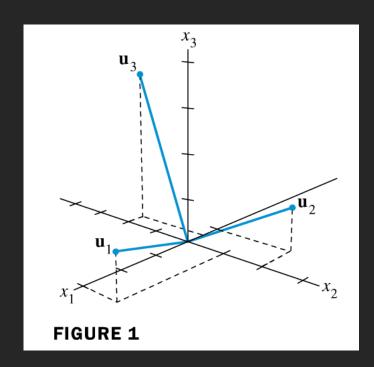
Theorem

If $S = \{ u_1, ..., u_p \}$ is an orthogonal set of nonzero vectors in R^n , then S is linearly independent.

Proof

PROOF If
$$\mathbf{0} = c_1 \mathbf{u}_1 + \cdots + c_p \mathbf{u}_p$$
 for some scalars c_1, \dots, c_p , then $0 = \mathbf{0} \cdot \mathbf{u}_1 = (c_1 \mathbf{u}_1 + c_2 \mathbf{u}_2 + \cdots + c_p \mathbf{u}_p) \cdot \mathbf{u}_1$ $= (c_1 \mathbf{u}_1) \cdot \mathbf{u}_1 + (c_2 \mathbf{u}_2) \cdot \mathbf{u}_1 + \cdots + (c_p \mathbf{u}_p) \cdot \mathbf{u}_1$ $= c_1(\mathbf{u}_1 \cdot \mathbf{u}_1) + c_2(\mathbf{u}_2 \cdot \mathbf{u}_1) + \cdots + c_p(\mathbf{u}_p \cdot \mathbf{u}_1)$ $= c_1(\mathbf{u}_1 \cdot \mathbf{u}_1)$

because \mathbf{u}_1 is orthogonal to $\mathbf{u}_2, \dots, \mathbf{u}_p$. Since \mathbf{u}_1 is nonzero, $\mathbf{u}_1 \cdot \mathbf{u}_1$ is not zero and so $c_1 = 0$. Similarly, c_2, \dots, c_p must be zero. Thus S is linearly independent.



Express the vector
$$\mathbf{y} = \begin{bmatrix} 6 \\ 1 \\ -8 \end{bmatrix}$$

as a linear combination of the vectors

$$\mathbf{u}_1 = \begin{bmatrix} 3 \\ 1 \\ 1 \end{bmatrix}, \quad \mathbf{u}_2 = \begin{bmatrix} -1 \\ 2 \\ 1 \end{bmatrix}, \quad \mathbf{u}_3 = \begin{bmatrix} -1/2 \\ -2 \\ 7/2 \end{bmatrix}$$

Orthogonal as seen in previous example

Theorem

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

$$\mathbf{y} = c_1 \mathbf{u}_1 + \dots + c_p \mathbf{u}_p$$

are given by

$$c_j = \frac{\mathbf{y} \cdot \mathbf{u}_j}{\mathbf{u}_j \cdot \mathbf{u}_j} \qquad (j = 1, \dots, p)$$

Proof

PROOF As in the preceding proof, the orthogonality of $\{\mathbf{u}_1, \dots, \mathbf{u}_p\}$ shows that

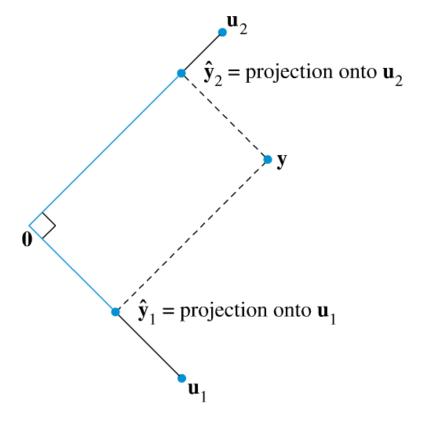
$$\mathbf{y} \cdot \mathbf{u}_1 = (c_1 \mathbf{u}_1 + c_2 \mathbf{u}_2 + \dots + c_p \mathbf{u}_p) \cdot \mathbf{u}_1 = c_1 (\mathbf{u}_1 \cdot \mathbf{u}_1)$$

Since $\mathbf{u}_1 \cdot \mathbf{u}_1$ is not zero, the equation above can be solved for c_1 . To find c_j for j = 2, ..., p, compute $\mathbf{y} \cdot \mathbf{u}_j$ and solve for c_j .

Geometrical Interpretation of the Theorem

- The first term in the expression is the projection of y onto the subspace spanned by u₁ (the line through u₁ and the origin),
- The second term is the projection of y onto the subspace spanned by u₂.
- Thus this expression expresses y as the sum of its projections onto the (orthogonal) axes determined by u₁ and u₂.

$$\mathbf{y} = \frac{\mathbf{y} \cdot \mathbf{u}_1}{\mathbf{u}_1 \cdot \mathbf{u}_1} \mathbf{u}_1 + \frac{\mathbf{y} \cdot \mathbf{u}_2}{\mathbf{u}_2 \cdot \mathbf{u}_2} \mathbf{u}_2$$



Geometrical Interpretation in Higher Dimension

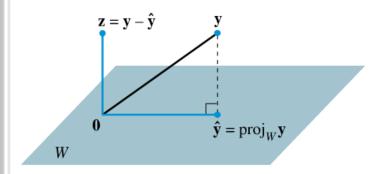
Let W be a subspace of \mathbb{R}^n . Then each y in \mathbb{R}^n can be written uniquely in the form

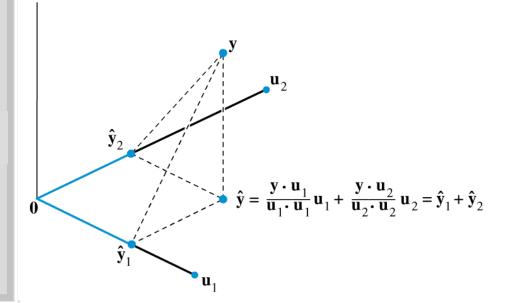
$$\mathbf{y} = \hat{\mathbf{y}} + \mathbf{z} \tag{1}$$

The vector $\hat{\mathbf{y}}$ in (1) is called the **orthogonal projection of y onto** W

Here $\hat{\mathbf{y}}_1$ and $\hat{\mathbf{y}}_2$ denote

the projections of \mathbf{y} onto the lines spanned by \mathbf{u}_1 and \mathbf{u}_2 , respectively. The orthogonal projection $\hat{\mathbf{y}}$ of \mathbf{y} onto W is the sum of the projections of \mathbf{y} onto one-dimensional subspaces that are orthogonal to each other.





$$\mathbf{y} \cdot \mathbf{u}_1 = 11, \qquad \mathbf{y} \cdot \mathbf{u}_2 = -12, \qquad \mathbf{y} \cdot \mathbf{u}_3 = -33$$
 $\mathbf{u}_1 \cdot \mathbf{u}_1 = 11, \qquad \mathbf{u}_2 \cdot \mathbf{u}_2 = 6, \qquad \mathbf{u}_3 \cdot \mathbf{u}_3 = 33/2$

Solution (Example Revisited)

Express the vector
$$\mathbf{y} = \begin{bmatrix} 6 \\ 1 \\ -8 \end{bmatrix}$$

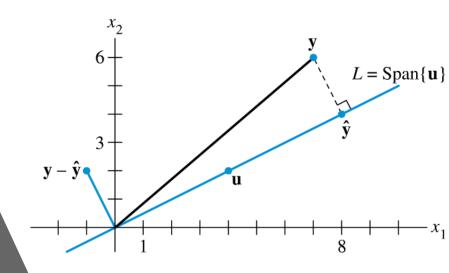
as a linear combination of the vectors

$$\mathbf{u}_1 = \begin{bmatrix} 3 \\ 1 \\ 1 \end{bmatrix}, \quad \mathbf{u}_2 = \begin{bmatrix} -1 \\ 2 \\ 1 \end{bmatrix}, \quad \mathbf{u}_3 = \begin{bmatrix} -1/2 \\ -2 \\ 7/2 \end{bmatrix}$$

$$\mathbf{y} = \frac{\mathbf{y} \cdot \mathbf{u}_1}{\mathbf{u}_1 \cdot \mathbf{u}_1} \mathbf{u}_1 + \frac{\mathbf{y} \cdot \mathbf{u}_2}{\mathbf{u}_2 \cdot \mathbf{u}_2} \mathbf{u}_2 + \frac{\mathbf{y} \cdot \mathbf{u}_3}{\mathbf{u}_3 \cdot \mathbf{u}_3} \mathbf{u}_3$$
$$= \frac{11}{11} \mathbf{u}_1 + \frac{-12}{6} \mathbf{u}_2 + \frac{-33}{33/2} \mathbf{u}_3$$
$$= \mathbf{u}_1 - 2\mathbf{u}_2 - 2\mathbf{u}_3$$

Let y = [76] and u = [42]. Find the orthogonal projection of \mathbf{y} onto \mathbf{u} . Then write \mathbf{y} as the sum of two orthogonal vectors, one in Span $\{u\}$ and one orthogonal to \mathbf{u} .

Find the distance from **y** to *L* as shown in the figure.



Solution

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

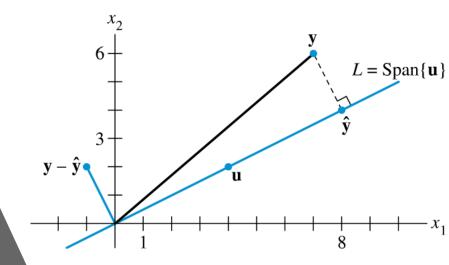
$$\mathbf{y} - \hat{\mathbf{y}} = \begin{bmatrix} 7 \\ 6 \end{bmatrix} - \begin{bmatrix} 8 \\ 4 \end{bmatrix} = \begin{bmatrix} -1 \\ 2 \end{bmatrix}$$

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

$$\uparrow \qquad \uparrow \qquad \uparrow$$

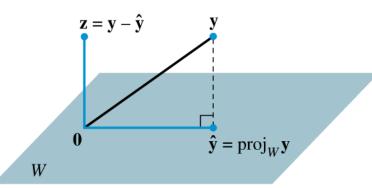
$$\mathbf{\hat{y}} \qquad (\mathbf{y} - \hat{\mathbf{y}})$$

$$\|\mathbf{y} - \hat{\mathbf{y}}\| = \sqrt{(-1)^2 + 2^2} = \sqrt{5}$$



The distance from a point \mathbf{y} in \mathbb{R}^n to a subspace W is defined as the distance from \mathbf{y} to the nearest point in W. Find the distance from \mathbf{y} to $\mathbb{W} = \mathbb{S}_{n} \{ \mathbf{u}_{1}, \mathbf{u}_{2} \}$, where

$$\mathbf{y} = \begin{bmatrix} -1 \\ -5 \\ 10 \end{bmatrix}, \quad \mathbf{u}_1 = \begin{bmatrix} 5 \\ -2 \\ 1 \end{bmatrix}, \quad \mathbf{u}_2 = \begin{bmatrix} 1 \\ 2 \\ -1 \end{bmatrix}$$



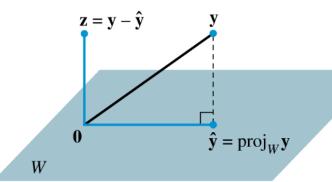
Solution

$$\hat{\mathbf{y}} = \frac{\mathbf{y} \cdot \mathbf{u}_1}{\mathbf{u}_1 \cdot \mathbf{u}_1} \mathbf{u}_1 + \frac{\mathbf{y} \cdot \mathbf{u}_2}{\mathbf{u}_2 \cdot \mathbf{u}_2} \mathbf{u}_2$$

$$\hat{\mathbf{y}} = \frac{15}{30}\mathbf{u}_1 + \frac{-21}{6}\mathbf{u}_2 = \frac{1}{2} \begin{bmatrix} 5 \\ -2 \\ 1 \end{bmatrix} - \frac{7}{2} \begin{bmatrix} 1 \\ 2 \\ -1 \end{bmatrix} = \begin{bmatrix} -1 \\ -8 \\ 4 \end{bmatrix}$$

$$\mathbf{y} - \hat{\mathbf{y}} = \begin{bmatrix} -1 \\ -5 \\ 10 \end{bmatrix} - \begin{bmatrix} -1 \\ -8 \\ 4 \end{bmatrix} = \begin{bmatrix} 0 \\ 3 \\ 6 \end{bmatrix}$$
$$\|\mathbf{y} - \hat{\mathbf{y}}\|^2 = 3^2 + 6^2 = 45$$

The distance from **y** to W is $\sqrt{45} = 3\sqrt{5}$.



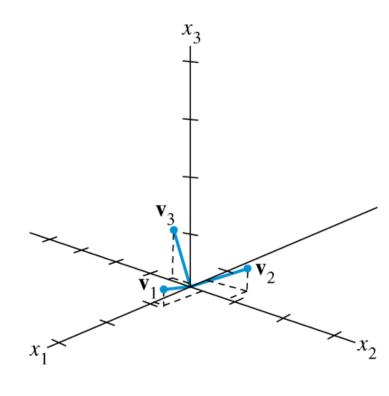
Definition: Orthonormal Set

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

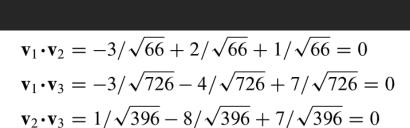
The simplest example of an orthonormal set is the standard basis $\{\mathbf{e}_1, \dots, \mathbf{e}_n\}$ for \mathbb{R}^n . Any nonempty subset of $\{\mathbf{e}_1, \dots, \mathbf{e}_n\}$ is orthonormal, too. Here is a more complicated example.

Show that $\{v_1, v_2, v_3\}$ is an orthonormal basis of \mathbb{R}^3 , where

$$\mathbf{v}_{1} = \begin{bmatrix} 3/\sqrt{11} \\ 1/\sqrt{11} \\ 1/\sqrt{11} \end{bmatrix}, \quad \mathbf{v}_{2} = \begin{bmatrix} -1/\sqrt{6} \\ 2/\sqrt{6} \\ 1/\sqrt{6} \end{bmatrix}, \quad \mathbf{v}_{3} = \begin{bmatrix} -1/\sqrt{66} \\ -4/\sqrt{66} \\ 7/\sqrt{66} \end{bmatrix}$$



Solution

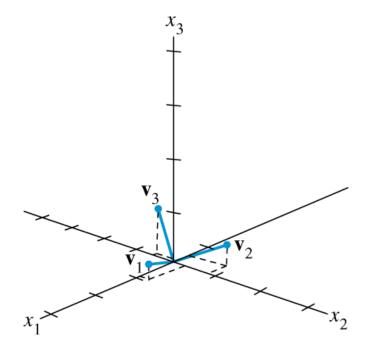


Thus $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ is an orthogonal set. Also,

$$\mathbf{v}_1 \cdot \mathbf{v}_1 = 9/11 + 1/11 + 1/11 = 1$$

 $\mathbf{v}_2 \cdot \mathbf{v}_2 = 1/6 + 4/6 + 1/6 = 1$
 $\mathbf{v}_3 \cdot \mathbf{v}_3 = 1/66 + 16/66 + 49/66 = 1$

which shows that v_1 , v_2 , and v_3 are unit vectors. Thus $\{v_1, v_2, v_3\}$ is an orthonormal set.



Find U^TU.

$$U = \begin{bmatrix} 1/\sqrt{2} & 2/3 \\ 1/\sqrt{2} & -2/3 \\ 0 & 1/3 \end{bmatrix}$$

$$U^{T}U = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} & 0\\ 2/3 & -2/3 & 1/3 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} & 2/3\\ 1/\sqrt{2} & -2/3\\ 0 & 1/3 \end{bmatrix} = \begin{bmatrix} 1 & 0\\ 0 & 1 \end{bmatrix}$$

Theorem

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

Proof

$$U^T U = \begin{bmatrix} \mathbf{u}_1^T \\ \mathbf{u}_2^T \\ \mathbf{u}_3^T \end{bmatrix} \begin{bmatrix} \mathbf{u}_1 & \mathbf{u}_2 & \mathbf{u}_3 \end{bmatrix} = \begin{bmatrix} \mathbf{u}_1^T \mathbf{u}_1 & \mathbf{u}_1^T \mathbf{u}_2 & \mathbf{u}_1^T \mathbf{u}_3 \\ \mathbf{u}_2^T \mathbf{u}_1 & \mathbf{u}_2^T \mathbf{u}_2 & \mathbf{u}_2^T \mathbf{u}_3 \\ \mathbf{u}_3^T \mathbf{u}_1 & \mathbf{u}_3^T \mathbf{u}_2 & \mathbf{u}_3^T \mathbf{u}_3 \end{bmatrix}$$

$$\mathbf{u}_{1}^{T}\mathbf{u}_{2} = \mathbf{u}_{2}^{T}\mathbf{u}_{1} = 0, \quad \mathbf{u}_{1}^{T}\mathbf{u}_{3} = \mathbf{u}_{3}^{T}\mathbf{u}_{1} = 0, \quad \mathbf{u}_{2}^{T}\mathbf{u}_{3} = \mathbf{u}_{3}^{T}\mathbf{u}_{2} = 0$$

The columns of U all have unit length if and only if

$$\mathbf{u}_1^T \mathbf{u}_1 = 1, \quad \mathbf{u}_2^T \mathbf{u}_2 = 1, \quad \mathbf{u}_3^T \mathbf{u}_3 = 1$$

Definition: Orthogonal Matrix

- An **orthogonal matrix** is a square invertible matrix U such that $U^{-1} = U^{T}$.
- Such a matrix has orthonormal columns and rows.