

Lecture Three

Section 3.1 – Inner Products

Definition

An **inner product** on a real vector space V is a function that associates a real number $\langle \mathbf{u}, \mathbf{v} \rangle$ with each pair of vectors in V in such a way that the following axioms are satisfied for all vectors \mathbf{u}, \mathbf{v} , and \mathbf{w} in V and all scalars k .

1. $\langle \mathbf{u}, \mathbf{v} \rangle = \langle \mathbf{v}, \mathbf{u} \rangle$ *Symmetry axiom*
2. $\langle \mathbf{u} + \mathbf{v}, \mathbf{w} \rangle = \langle \mathbf{u}, \mathbf{w} \rangle + \langle \mathbf{v}, \mathbf{w} \rangle$ *Additivity axiom*
3. $\langle k\mathbf{u}, \mathbf{v} \rangle = k \langle \mathbf{u}, \mathbf{v} \rangle$ *Homogeneity axiom*
4. $\langle \mathbf{v}, \mathbf{v} \rangle \geq 0$ and $\langle \mathbf{v}, \mathbf{v} \rangle = 0$ iff $\mathbf{v} = \mathbf{0}$ *Positivity axiom*

A real vector space with an inner product is called a **real inner product space**.

$$\langle \mathbf{u}, \mathbf{v} \rangle = \mathbf{u} \cdot \mathbf{v} = u_1 v_1 + u_2 v_2 + \cdots + u_n v_n$$

This is called the **Euclidean inner product** (or the **standard inner product**)

Definition

If V is a real inner product space, then the norm (or length) of a vector \mathbf{v} in V is denoted by $\|\mathbf{v}\|$ and is defined by

$$\|\mathbf{v}\| = \sqrt{\langle \mathbf{v}, \mathbf{v} \rangle}$$

And the **distance** between two vectors is denoted by $d(\mathbf{u}, \mathbf{v})$ and is defined by

$$d(\mathbf{u}, \mathbf{v}) = \|\mathbf{u} - \mathbf{v}\| = \sqrt{\langle \mathbf{u} - \mathbf{v}, \mathbf{u} - \mathbf{v} \rangle}$$

A vector of norm 1 is called a **unit vector**.

Theorem

If \mathbf{u} and \mathbf{v} are vectors in a real inner product space V , and if k is a scalar, then:

- a) $\|\mathbf{v}\| \geq 0$ with equality iff $\mathbf{v} = \mathbf{0}$
- b) $\|k\mathbf{v}\| = |k| \|\mathbf{v}\|$
- c) $d(\mathbf{u}, \mathbf{v}) = d(\mathbf{v}, \mathbf{u})$
- d) $d(\mathbf{u}, \mathbf{v}) \geq 0$ with equality iff $\mathbf{u} = \mathbf{v}$

Although the Euclidean inner product is the most important inner product on R^n , there are various applications in which is desirable to modify it by weighing each term differently. More precisely, if w_1, w_2, \dots, w_n are positive real numbers, which we will call weighs, and if $\mathbf{u} = (u_1, u_2, \dots, u_n)$ and \mathbf{v} are vectors in R^n , then it can be shown that the formula

$$\langle \mathbf{u}, \mathbf{v} \rangle = w_1 u_1 v_1 + w_2 u_2 v_2 + \dots + w_n u_n v_n$$

Defines an inner product on R^n that we call the **weighted Euclidean inner product** with weights w_1, w_2, \dots, w_n

Example

Let $\mathbf{u} = (u_1, u_2)$ and $\mathbf{v} = (v_1, v_2)$ be vectors in R^2 , verify that the weighted Euclidean inner product $\langle \mathbf{u}, \mathbf{v} \rangle = 3u_1 v_1 + 2u_2 v_2$ satisfies the four inner product axioms.

Solution

$$\text{Axiom 1: } \langle \mathbf{u}, \mathbf{v} \rangle = 3u_1 v_1 + 2u_2 v_2 = 3v_1 u_1 + 2v_2 u_2 = \langle \mathbf{v}, \mathbf{u} \rangle$$

$$\begin{aligned} \text{Axiom 2: } \langle \mathbf{u} + \mathbf{v}, \mathbf{w} \rangle &= 3(u_1 + v_1)w_1 + 2(u_2 + v_2)w_2 \\ &= 3(u_1 w_1 + v_1 w_1) + 2(u_2 w_2 + v_2 w_2) \\ &= 3u_1 w_1 + 3v_1 w_1 + 2u_2 w_2 + 2v_2 w_2 \\ &= (3u_1 w_1 + 2u_2 w_2) + (3v_1 w_1 + 2v_2 w_2) \\ &= \langle \mathbf{u}, \mathbf{w} \rangle + \langle \mathbf{v}, \mathbf{w} \rangle \end{aligned}$$

$$\begin{aligned} \text{Axiom 3: } \langle k\mathbf{u}, \mathbf{v} \rangle &= 3(ku_1)v_1 + 2(ku_2)v_2 \\ &= k(3u_1 v_1 + 2u_2 v_2) \\ &= k\langle \mathbf{u}, \mathbf{v} \rangle \end{aligned}$$

$$\begin{aligned} \text{Axiom 3: } \langle \mathbf{v}, \mathbf{v} \rangle &= 3v_1 v_1 + 2v_2 v_2 \\ &= 3v_1^2 + 2v_2^2 \geq 0 \\ v_1 = v_2 = 0 &\text{ iff } \mathbf{v} = \mathbf{0} \end{aligned}$$

Exercises Section 3.1 – Inner Products

1. Let $\langle \mathbf{u}, \mathbf{v} \rangle$ be the Euclidean inner product on R^2 , and let $\mathbf{u} = (1, 1)$, $\mathbf{v} = (3, 2)$, $\mathbf{w} = (0, -1)$, and $k = 3$. Compute the following.

a) $\langle \mathbf{u}, \mathbf{v} \rangle$	c) $\langle \mathbf{u} + \mathbf{v}, \mathbf{w} \rangle$	e) $d(\mathbf{u}, \mathbf{v})$
b) $\langle k\mathbf{v}, \mathbf{w} \rangle$	d) $\ \mathbf{v}\ $	f) $\ \mathbf{u} - k\mathbf{v}\ $

2. Let $\langle \mathbf{u}, \mathbf{v} \rangle$ be the Euclidean inner product on R^2 , and let $\mathbf{u} = (1, 1)$, $\mathbf{v} = (3, 2)$, $\mathbf{w} = (0, -1)$ and $k = 3$. Compute the following for the weighted Euclidean inner product $\langle \mathbf{u}, \mathbf{v} \rangle = 2u_1v_1 + 3u_2v_2$.

a) $\langle \mathbf{u}, \mathbf{v} \rangle$	c) $\langle \mathbf{u} + \mathbf{v}, \mathbf{w} \rangle$	e) $d(\mathbf{u}, \mathbf{v})$
b) $\langle k\mathbf{v}, \mathbf{w} \rangle$	d) $\ \mathbf{v}\ $	f) $\ \mathbf{u} - k\mathbf{v}\ $

3. Let $\langle \mathbf{u}, \mathbf{v} \rangle$ be the Euclidean inner product on R^2 , and let $\mathbf{u} = (3, -2)$, $\mathbf{v} = (4, 5)$, $\mathbf{w} = (-1, 6)$, and $k = -4$. Verify the following.

a) $\langle \mathbf{u}, \mathbf{v} \rangle = \langle \mathbf{v}, \mathbf{u} \rangle$	d) $\langle k\mathbf{u}, \mathbf{v} \rangle = k\langle \mathbf{u}, \mathbf{v} \rangle$
b) $\langle \mathbf{u} + \mathbf{v}, \mathbf{w} \rangle = \langle \mathbf{u}, \mathbf{w} \rangle + \langle \mathbf{v}, \mathbf{w} \rangle$	e) $\langle \mathbf{0}, \mathbf{v} \rangle = \langle \mathbf{v}, \mathbf{0} \rangle = 0$
c) $\langle \mathbf{u}, \mathbf{v} + \mathbf{w} \rangle = \langle \mathbf{u}, \mathbf{v} \rangle + \langle \mathbf{u}, \mathbf{w} \rangle$	

4. Let $\langle \mathbf{u}, \mathbf{v} \rangle$ be the Euclidean inner product on R^2 , and let $\mathbf{u} = (3, -2)$, $\mathbf{v} = (4, 5)$, $\mathbf{w} = (-1, 6)$, and $k = -4$. Verify the following for the weighted Euclidean inner product $\langle \mathbf{u}, \mathbf{v} \rangle = 4u_1v_1 + 5u_2v_2$

a) $\langle \mathbf{u}, \mathbf{v} \rangle = \langle \mathbf{v}, \mathbf{u} \rangle$	d) $\langle k\mathbf{u}, \mathbf{v} \rangle = k\langle \mathbf{u}, \mathbf{v} \rangle$
b) $\langle \mathbf{u} + \mathbf{v}, \mathbf{w} \rangle = \langle \mathbf{u}, \mathbf{w} \rangle + \langle \mathbf{v}, \mathbf{w} \rangle$	e) $\langle \mathbf{0}, \mathbf{v} \rangle = \langle \mathbf{v}, \mathbf{0} \rangle = 0$
c) $\langle \mathbf{u}, \mathbf{v} + \mathbf{w} \rangle = \langle \mathbf{u}, \mathbf{v} \rangle + \langle \mathbf{u}, \mathbf{w} \rangle$	

5. Let $\mathbf{u} = (u_1, u_2)$ and $\mathbf{v} = (v_1, v_2)$. Show that the following are inner product on R^3 by verifying that the inner product axioms hold. $\langle \mathbf{u}, \mathbf{v} \rangle = 3u_1v_1 + 5u_2v_2$

6. Show that the following identity holds for the vectors in any inner product space

$$\|\mathbf{u} + \mathbf{v}\|^2 + \|\mathbf{u} - \mathbf{v}\|^2 = 2\|\mathbf{u}\|^2 + 2\|\mathbf{v}\|^2$$

Section 3.2 – Angle and Orthogonality in Inner Product Spaces

Cosine Formula

If u and v are nonzero vectors that implies $\Rightarrow \cos \theta = \frac{u \cdot v}{\|u\| \cdot \|v\|} \rightarrow \boxed{\theta = \cos^{-1} \left(\frac{\langle u, v \rangle}{\|u\| \cdot \|v\|} \right)}$

$$-1 \leq \frac{u \cdot v}{\|u\| \cdot \|v\|} \leq 1$$

Example

Let R^4 have the Euclidean inner product. Find the cosine angle θ between the vectors $u = (4, 3, 1, -2)$ and $v = (-2, 1, 2, 3)$.

Solution

$$\|u\| = \sqrt{4^2 + 3^2 + 1^2 + (-2)^2} = \sqrt{30}$$

$$\|v\| = \sqrt{(-2)^2 + 1^2 + 2^2 + 3^2} = \sqrt{18} = 3\sqrt{2}$$

$$\langle u, v \rangle = 4(-2) + 3(1) + 1(2) - 2(3) = -9$$

$$\begin{aligned} \cos \theta &= \frac{\langle u, v \rangle}{\|u\| \cdot \|v\|} \\ &= -\frac{9}{3\sqrt{30}\sqrt{2}} \\ &= -\frac{3}{\sqrt{60}} \\ &= -\frac{3}{2\sqrt{15}} \end{aligned}$$

Theorem – Cauchy-Schwarz Inequality

If v and w are vectors in a real inner product space V , then

$$\|\langle u, v \rangle\| \leq \|u\| \cdot \|v\|$$

The following two alternative forms of the Cauchy-Schwarz inequality are useful to know:

$$\langle u, v \rangle^2 \leq \langle u, u \rangle \langle v, v \rangle$$

$$\langle u, v \rangle^2 \leq \|u\|^2 \cdot \|v\|^2$$

Theorem

If \mathbf{u} , \mathbf{v} and \mathbf{w} are vectors in a real inner product space V , and if k is any scalar, then

$$a) \quad \|\mathbf{u} + \mathbf{v}\| \leq \|\mathbf{u}\| + \|\mathbf{v}\| \quad (\text{Triangle inequality for vectors})$$

$$b) \quad d(\mathbf{u}, \mathbf{v}) \leq d(\mathbf{u}, \mathbf{w}) + d(\mathbf{w}, \mathbf{v}) \quad (\text{Triangle inequality for distances})$$

Proof (a)

$$\begin{aligned} \|\mathbf{u} + \mathbf{v}\|^2 &= \langle \mathbf{u} + \mathbf{v}, \mathbf{u} + \mathbf{v} \rangle \\ &= \langle \mathbf{u}, \mathbf{u} \rangle + 2\langle \mathbf{u}, \mathbf{v} \rangle + \langle \mathbf{v}, \mathbf{v} \rangle \\ &\leq \langle \mathbf{u}, \mathbf{u} \rangle + 2|\langle \mathbf{u}, \mathbf{v} \rangle| + \langle \mathbf{v}, \mathbf{v} \rangle \\ &\leq \langle \mathbf{u}, \mathbf{u} \rangle + 2\|\mathbf{u}\|\|\mathbf{v}\| + \langle \mathbf{v}, \mathbf{v} \rangle \\ &= \|\mathbf{u}\|^2 + 2\|\mathbf{u}\|\|\mathbf{v}\| + \|\mathbf{v}\|^2 \\ &= (\|\mathbf{u}\| + \|\mathbf{v}\|)^2 \end{aligned}$$

$$\|\mathbf{u} + \mathbf{v}\|^2 \leq (\|\mathbf{u}\| + \|\mathbf{v}\|)^2$$

$$\|\mathbf{u} + \mathbf{v}\| \leq \|\mathbf{u}\| + \|\mathbf{v}\|$$

Definition

Two vectors \mathbf{u} and \mathbf{v} in an inner product space are called orthogonal if $\langle \mathbf{u}, \mathbf{v} \rangle = 0$

Example

The vectors $\mathbf{u} = (1, 1)$ and $\mathbf{v} = (1, -1)$ are orthogonal with respect to the Euclidean inner product on \mathbb{R}^2 , since

$$\mathbf{u} \cdot \mathbf{v} = 1(1) + 1(-1) = 0$$

They are not orthogonal with the respect to the weighted Euclidean inner product

$\langle \mathbf{u}, \mathbf{v} \rangle = 3u_1v_1 + 2u_2v_2$, since

$$\langle \mathbf{u}, \mathbf{v} \rangle = 3(1)(1) + 2(1)(-1) = 1 \neq 0$$

Example

$U = \begin{pmatrix} 1 & 0 \\ 1 & 1 \end{pmatrix}$ and $V = \begin{pmatrix} 0 & 2 \\ 0 & 0 \end{pmatrix}$ are orthogonal, since

$$U \cdot V = 1(0) + 0(2) + 1(0) + 1(0) = 0$$

Definition

If W is a subspace of an inner product space V , then the set of all vectors are orthogonal to every vector in W is called the **orthogonal complement** of W and is denoted by the symbol W^\perp

Theorem

If W is a subspace of an inner product space V , then:

- a) W^\perp is a subspace of V .
- b) $W \cap W^\perp = \{0\}$

Proof

- a) Let set W^\perp contains at least the zero vector, since $\langle 0, w \rangle = 0$ for every vector w in W . We need to show that W^\perp is closed under addition and scalar multiplication.

Suppose that u and v are vectors in W^\perp , so every vector w in W we have $\langle u, w \rangle = 0$ and $\langle v, w \rangle = 0$

$$\langle u + v, w \rangle = \langle u, w \rangle + \langle v, w \rangle = 0 + 0 = 0 \quad \text{Closed under addition}$$

$$\langle ku, w \rangle = k \langle u, w \rangle = k(0) = 0 \quad \text{Closed under scalar multiplication}$$

Which proves that $u + w$ and ku are in W^\perp

- b) If v is any vector in both W and W^\perp , then v is orthogonal to itself; that is, $\langle v, v \rangle = 0$. It follows from the positivity axiom for inner products that $v = 0$

Theorem

If W is a subspace of a finite-dimensional inner product space V , then the orthogonal complement of W^\perp is W ; that is

$$(W^\perp)^\perp = W$$

Example

Let W be the subspace of \mathbb{R}^6 spanned by the vectors

$$\begin{aligned} \mathbf{w}_1 &= (1, 3, -2, 0, 2, 0), & \mathbf{w}_2 &= (2, 6, -5, -2, 4, -3) \\ \mathbf{w}_3 &= (0, 0, 5, 10, 0, 15), & \mathbf{w}_4 &= (2, 6, 0, 8, 4, 18) \end{aligned}$$

Find a basis for the orthogonal complement of W .

Solution

The Space W is the same as the row space of the matrix

$$A = \begin{bmatrix} 1 & 3 & -2 & 0 & 2 & 0 \\ 2 & 6 & -5 & -2 & 4 & -3 \\ 0 & 0 & 5 & 10 & 0 & 15 \\ 2 & 6 & 0 & 8 & 4 & 18 \end{bmatrix} \xrightarrow{\text{rref}} \begin{bmatrix} 1 & 3 & 0 & 4 & 2 & 0 \\ 0 & 0 & 1 & 2 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

The solution

$$\begin{aligned} (x_1, x_2, x_3, x_4, x_5, x_6) &= (-3x_2 - 4x_4 - 2x_5, x_2, -2x_4, x_4, x_5, 0) \\ &= x_2(-3, 1, 0, 0, 0, 0) + x_4(-4, 0, -2, 1, 0, 0) + x_5(-2, 0, 0, 0, 1, 0) \end{aligned}$$

$$\mathbf{v}_1 = (-3, 1, 0, 0, 0, 0), \quad \mathbf{v}_2 = (-4, 0, -2, 1, 0, 0), \quad \mathbf{v}_3 = (-2, 0, 0, 0, 1, 0)$$

Definition

A collection of vectors in \mathbb{R}^n (or inner space) is called orthogonal if any 2 are perpendicular.

$$\mathbf{v}_i \cdot \mathbf{v}_j = \mathbf{v}_i^T \mathbf{v}_j = \begin{cases} 0 & \text{for } i \neq j \quad (\text{orthogonal vectors}) \\ 1 & \text{for } i = j \quad (\text{unit vectors}) \end{cases}$$

Theorem

If $\mathbf{v}_1, \dots, \mathbf{v}_m$ are nonzero orthogonal vectors, then they are linearly independent.

Definition

A vector \mathbf{v} is called normal if $\|\mathbf{v}\| = 1$

A collection of vectors $\mathbf{v}_1, \dots, \mathbf{v}_m$ is called orthonormal if they are orthogonal and each $\|\mathbf{v}_i\| = 1$.

An orthonormal basis is a basis made up of orthonormal vectors.

Example

Q rotates every vector in the plane through the angle θ .

$$Q = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$

$$Q^{-1} = \frac{1}{\cos^2 \theta + \sin^2 \theta} \begin{bmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{bmatrix} = \begin{bmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{bmatrix} = \underline{Q^T} \quad \cos^2 \theta + \sin^2 \theta = 1$$

The dot product $(\cos \theta \sin \theta - \sin \theta \cos \theta = 0)$, the columns are orthogonal.

They are unit vectors because $\cos^2 \theta + \sin^2 \theta = 1$. Those columns give an orthonormal basis for the plane \mathbf{R}^2 .

We have: $QQ^T = I = Q^T Q$ (This type is called **rotation**)

Exercises **Section 3.2 – Angle and Orthogonality in Inner Product Spaces**

1. Which of the following form orthonormal sets?

a) $(1, 0), (0, 2)$ in \mathbf{R}^2

b) $\left(\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}}\right), \left(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}\right)$ in \mathbf{R}^2

c) $\left(-\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}}\right), \left(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}\right)$ in \mathbf{R}^2

d) $\left(\frac{1}{\sqrt{2}}, 0, \frac{1}{\sqrt{2}}\right), \left(\frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}, -\frac{1}{\sqrt{3}}\right), \left(-\frac{1}{\sqrt{2}}, 0, \frac{1}{\sqrt{2}}\right)$ in \mathbf{R}^3

e) $\left(\frac{2}{3}, -\frac{2}{3}, \frac{1}{3}\right), \left(\frac{2}{3}, \frac{1}{3}, -\frac{2}{3}\right), \left(\frac{1}{3}, \frac{2}{3}, \frac{2}{3}\right)$ in \mathbf{R}^3

f) $\left(\frac{1}{\sqrt{6}}, \frac{1}{\sqrt{6}}, -\frac{2}{\sqrt{6}}\right), \left(\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}}, 0\right)$ in \mathbf{R}^3

2. Find the cosine of the angle between \mathbf{u} and \mathbf{v} .

a) $\mathbf{u} = (1, -3), \mathbf{v} = (2, 4)$

d) $\mathbf{u} = (4, 1, 8), \mathbf{v} = (1, 0, -3)$

b) $\mathbf{u} = (-1, 0), \mathbf{v} = (3, 8)$

e) $\mathbf{u} = (1, 0, 1, 0), \mathbf{v} = (-3, -3, -3, -3)$

c) $\mathbf{u} = (-1, 5, 2), \mathbf{v} = (2, 4, -9)$

f) $\mathbf{u} = (2, 1, 7, -1), \mathbf{v} = (4, 0, 0, 0)$

3. Find the cosine of the angle between \mathbf{A} and \mathbf{B} .

a) $\mathbf{A} = \begin{pmatrix} 2 & 6 \\ 1 & -3 \end{pmatrix} \quad \mathbf{B} = \begin{pmatrix} 3 & 2 \\ 1 & 0 \end{pmatrix}$

b) $\mathbf{A} = \begin{pmatrix} 2 & 4 \\ -1 & 3 \end{pmatrix} \quad \mathbf{B} = \begin{pmatrix} -3 & 1 \\ 4 & 2 \end{pmatrix}$

4. Determine whether the given vectors are orthogonal with respect to the Euclidean inner product.

a) $\mathbf{u} = (-1, 3, 2), \mathbf{v} = (4, 2, -1)$

d) $\mathbf{u} = (-4, 6, -10, 1), \mathbf{v} = (2, 1, -2, 9)$

b) $\mathbf{u} = (a, b), \mathbf{v} = (-b, a)$

e) $\mathbf{u} = (-4, 6, -10, 1), \mathbf{v} = (2, 1, -2, 9)$

c) $\mathbf{u} = (-2, -2, -2), \mathbf{v} = (1, 1, 1)$

5. Do there exist scalars k and l such that the vectors $\mathbf{u} = (2, k, 6)$, $\mathbf{v} = (l, 5, 3)$, and $\mathbf{w} = (1, 2, 3)$ are mutually orthogonal with respect to the Euclidean inner product?

6. Let \mathbf{R}^3 have the Euclidean inner product. For which values of k are \mathbf{u} and \mathbf{v} orthogonal?

a) $\mathbf{u} = (2, 1, 3), \mathbf{v} = (1, 7, k)$

b) $\mathbf{u} = (k, k, 1), \mathbf{v} = (k, 5, 6)$

7. Let V be an inner product space. Show that if \mathbf{u} and \mathbf{v} are orthogonal unit vectors in V , then $\|\mathbf{u} - \mathbf{v}\| = \sqrt{2}$

8. Let \mathbf{S} be a subspace of \mathbb{R}^n . Explain what $(\mathbf{S}^\perp)^\perp = \mathbf{S}$ means and why it is true.
9. The methane molecule CH_4 is arranged as if the carbon atom were at the center of a regular tetrahedron with four hydrogen atoms at the vertices. If vertices are placed at $(0, 0, 0)$, $(1, 1, 0)$, $(1, 0, 1)$ and $(0, 1, 1)$ – (*note* that all six edges have length $\sqrt{2}$, so the tetrahedron is regular). What is the cosine of the angle between the rays going from the center $(\frac{1}{2}, \frac{1}{2}, \frac{1}{2})$ to the vertices?
10. Determine if the given vectors are orthogonal.
 $\mathbf{x}_1 = (1, 0, 1, 0)$, $\mathbf{x}_2 = (0, 1, 0, 1)$, $\mathbf{x}_3 = (1, 0, -1, 0)$, $\mathbf{x}_4 = (1, 1, -1, -1)$
11. Which of the following sets of vectors are orthogonal with respect to the Euclidean inner
- a) $\left(\frac{1}{\sqrt{2}}, 0, \frac{1}{\sqrt{2}}\right)$ $\left(\frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}, -\frac{1}{\sqrt{3}}\right)$ $\left(-\frac{1}{\sqrt{2}}, 0, \frac{1}{\sqrt{2}}\right)$
- b) $\left(\frac{2}{3}, -\frac{2}{3}, \frac{1}{3}\right)$ $\left(\frac{2}{3}, \frac{1}{3}, -\frac{2}{3}\right)$ $\left(\frac{1}{3}, \frac{2}{3}, \frac{2}{3}\right)$

Section 3.3 – Gram-Schmidt Process

Definition

A set of two or more vectors in a real inner product space is said to be **orthogonal** if all pairs of distinct vectors in the set are orthogonal. An orthogonal set in which each vector has norm 1 is said to be **orthonormal**.

Theorem

1. If $S = \{v_1, v_2, \dots, v_n\}$ is an orthogonal basis for an inner product space V , and if u is any vector in V , then

$$u = \frac{\langle u, v_1 \rangle}{\|v_1\|^2} v_1 + \frac{\langle u, v_2 \rangle}{\|v_2\|^2} v_2 + \dots + \frac{\langle u, v_n \rangle}{\|v_n\|^2} v_n$$

2. If $S = \{v_1, v_2, \dots, v_n\}$ is an orthonormal basis for an inner product space V , and if u is any vector in V , then

$$u = \langle u, v_1 \rangle v_1 + \langle u, v_2 \rangle v_2 + \dots + \langle u, v_n \rangle v_n$$

Proof

1. Since $S = \{v_1, v_2, \dots, v_n\}$ is a basis for V , every vector u in V can be expressed in the form

$$u = c_1 v_1 + c_2 v_2 + \dots + c_n v_n$$

Let show that $c_i = \frac{\langle u, v_i \rangle}{\|v_i\|^2}$ for $i = 1, 2, \dots, n$

$$\begin{aligned} \langle u, v_i \rangle &= \langle c_1 v_1 + c_2 v_2 + \dots + c_n v_n, v_i \rangle \\ &= c_1 \langle v_1, v_i \rangle + c_2 \langle v_2, v_i \rangle + \dots + c_n \langle v_n, v_i \rangle \end{aligned}$$

Since S is an orthogonal set, all of the inner products in the last equality are zero except the i^{th} , so we have

$$\langle u, v_i \rangle = c_i \langle v_i, v_i \rangle = c_i \|v_i\|^2$$

The Gram-Schmidt Process

To convert a basis $\{u_1, u_2, \dots, u_r\}$ into an orthogonal basis $\{v_1, v_2, \dots, v_r\}$, perform the following computations:

$$\text{Step 1: } v_1 = u_1$$

$$\text{Step 2: } v_2 = u_2 - \frac{\langle u_2, v_1 \rangle}{\|v_1\|^2} v_1$$

$$\text{Step 3: } v_3 = u_3 - \frac{\langle u_3, v_1 \rangle}{\|v_1\|^2} v_1 - \frac{\langle u_3, v_2 \rangle}{\|v_2\|^2} v_2$$

$$\text{Step 4: } v_4 = u_4 - \frac{\langle u_4, v_1 \rangle}{\|v_1\|^2} v_1 - \frac{\langle u_4, v_2 \rangle}{\|v_2\|^2} v_2 - \frac{\langle u_4, v_3 \rangle}{\|v_3\|^2} v_3$$

To convert the orthogonal basis into an orthonormal basis $\{q_1, q_2, q_3\}$, normalize the orthogonal basis

vectors.
$$q_i = \frac{v_i}{\|v_i\|}$$

Example

Assume that the vector space R^3 has the Euclidean inner product. Apply the Gram-Schmidt process to transform the basis vectors

$$u_1 = (1, 1, 1) \quad u_2 = (0, 1, 1) \quad u_3 = (0, 0, 1)$$

Into the orthogonal basis $\{v_1, v_2, v_3\}$, and then normalize the orthogonal basis vectors to obtain an orthonormal basis $\{q_1, q_2, q_3\}$

Solution

$$v_1 = u_1 = (1, 1, 1)$$

$$\begin{aligned} v_2 &= u_2 - \frac{\langle u_2, v_1 \rangle}{\|v_1\|^2} v_1 \\ &= (0, 1, 1) - \frac{0+1+1}{1^2+1^2+1^2} (1, 1, 1) \\ &= (0, 1, 1) - \frac{2}{3} (1, 1, 1) \end{aligned}$$

$$= \left(-\frac{2}{3}, \frac{1}{3}, \frac{1}{3} \right)$$

$$\begin{aligned}
\mathbf{v}_3 &= \mathbf{u}_3 - \text{proj}_{\mathbf{v}_2} \mathbf{u}_3 \\
&= \mathbf{u}_3 - \frac{\langle \mathbf{u}_3, \mathbf{v}_1 \rangle}{\|\mathbf{v}_1\|^2} \mathbf{v}_1 - \frac{\langle \mathbf{u}_3, \mathbf{v}_2 \rangle}{\|\mathbf{v}_2\|^2} \mathbf{v}_2 \\
&= (0, 0, 1) - \frac{0+0+1}{1^2+1^2+1^2} (1, 1, 1) - \frac{0+0+\frac{1}{3}}{\left(\frac{2}{3}\right)^2 + \left(\frac{1}{3}\right)^2 + \left(\frac{1}{3}\right)^2} \left(-\frac{2}{3}, \frac{1}{3}, \frac{1}{3} \right) \\
&= (0, 0, 1) - \frac{1}{3} (1, 1, 1) - \frac{\frac{1}{3}}{\frac{2}{3}} \left(-\frac{2}{3}, \frac{1}{3}, \frac{1}{3} \right) \\
&= (0, 0, 1) - \frac{1}{3} (1, 1, 1) - \frac{1}{2} \left(-\frac{2}{3}, \frac{1}{3}, \frac{1}{3} \right) \\
&= \left(0, -\frac{1}{2}, \frac{1}{2} \right)
\end{aligned}$$

$$\mathbf{q}_1 = \frac{\mathbf{v}_1}{\|\mathbf{v}_1\|} = \frac{(1, 1, 1)}{\sqrt{3}} = \left(\frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}} \right)$$

$$\begin{aligned}
\mathbf{q}_2 &= \frac{\mathbf{v}_2}{\|\mathbf{v}_2\|} = \frac{\left(-\frac{2}{3}, \frac{1}{3}, \frac{1}{3} \right)}{\sqrt{\left(\frac{2}{3}\right)^2 + \left(\frac{1}{3}\right)^2 + \left(\frac{1}{3}\right)^2}} \\
&= \frac{\left(-\frac{2}{3}, \frac{1}{3}, \frac{1}{3} \right)}{\frac{\sqrt{6}}{3}} \\
&= \left(-\frac{2}{\sqrt{6}}, \frac{1}{\sqrt{6}}, \frac{1}{\sqrt{6}} \right)
\end{aligned}$$

$$\begin{aligned}
\mathbf{q}_3 &= \frac{\mathbf{v}_3}{\|\mathbf{v}_3\|} = \frac{\left(0, -\frac{1}{2}, \frac{1}{2} \right)}{\sqrt{0^2 + \left(-\frac{1}{2}\right)^2 + \left(\frac{1}{2}\right)^2}} \\
&= \frac{\left(0, -\frac{1}{2}, \frac{1}{2} \right)}{\frac{\sqrt{2}}{2}} \\
&= \left(0, -\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}} \right)
\end{aligned}$$

Gram-Schmidt Process (Orthonormal)

Suppose v_1, \dots, v_n linearly independent in \mathbb{R}^n , construct n **orthonormal** u_1, \dots, u_n that span the same space: $\text{span} \{u_1, \dots, u_k\} = \text{span} \{v_1, \dots, v_k\}$

Step 1: Since v_i are linearly independent ($\neq 0$), so $\|v_1\| \neq 0$ (to create a normal vector)

Let $u_1 = \frac{v_1}{\|v_1\|} = q_1$, then $\|u_1\| = 1$ since u_1 is orthonormal and $\text{span} \{u_1\} = \text{span} \{v_1\}$

$$w_1 = v_1 \Rightarrow v_1 = \|w_1\| u_1$$

Step 2: $w_2 = v_2 - (v_2 \cdot q_1) q_1$

$$\Rightarrow w_2 = v_2 - \frac{v_2 \cdot u_1}{\|v_1\|} v_1 \quad (w_2 \perp u_1)$$

$$v_2 = \|w_2\| u_2 + (v_2 \cdot u_1) u_1 \quad w_2 = \|w_2\| u_2$$

$$q_2 = \frac{w_2}{\|w_2\|}$$

Step 3: $w_3 = v_3 - (v_3 \cdot q_1) q_1 - (v_3 \cdot q_2) q_2$

$$q_3 = \frac{w_3}{\|w_3\|}$$

	$u_1 = \frac{v_1}{\ v_1\ }$
$w_2 = v_2 - (v_2 \cdot u_1) u_1$	$u_2 = \frac{w_2}{\ w_2\ }$
$w_3 = v_3 - (v_3 \cdot u_1) u_1 - (v_3 \cdot u_2) u_2$	$u_3 = \frac{w_3}{\ w_3\ }$
$w_n = v_n - (v_n \cdot u_1) u_1 - (v_n \cdot u_2) u_2 - \dots - (v_n \cdot u_{n-1}) u_{n-1}$	$u_n = \frac{w_n}{\ w_n\ }$

Example

Use the Gram-Schmidt process to find an orthonormal basis for the subspaces of

$$v_1 = (1, 1, 0, 0) \quad v_2 = (0, 1, 1, 0) \quad v_3 = (1, 0, 1, 1)$$

Solution

$$\begin{aligned} \text{Step 1: } u_1 &= \frac{v_1}{\|v_1\|} = \frac{(1, 1, 0, 0)}{\sqrt{1^2+1^2+0+0}} \\ &= \frac{(1, 1, 0, 0)}{\sqrt{2}} \\ &= \left(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}, 0, 0 \right) \end{aligned}$$

$$\begin{aligned} \text{Step 2: } w_2 &= v_2 - (v_2 \cdot u_1) u_1 \\ &= (0, 1, 1, 0) - \left[(0, 1, 1, 0) \cdot \left(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}, 0, 0 \right) \right] \left(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}, 0, 0 \right) \\ &= (0, 1, 1, 0) - \frac{1}{\sqrt{2}} \left(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}, 0, 0 \right) \\ &= (0, 1, 1, 0) - \left(\frac{1}{2}, \frac{1}{2}, 0, 0 \right) \\ &= \left(-\frac{1}{2}, \frac{1}{2}, 1, 0 \right) \end{aligned}$$

$$\|w_2\| = \sqrt{\left(-\frac{1}{2}\right)^2 + \left(\frac{1}{2}\right)^2 + 1} = \sqrt{\frac{3}{2}} = \frac{\sqrt{3}}{\sqrt{2}} \frac{\sqrt{2}}{\sqrt{2}} = \frac{\sqrt{6}}{2}$$

$$\begin{aligned} u_2 &= \frac{w_2}{\|w_2\|} = \frac{\left(-\frac{1}{2}, \frac{1}{2}, 1, 0\right)}{\frac{\sqrt{6}}{2}} \\ &= \frac{2}{\sqrt{6}} \left(-\frac{1}{2}, \frac{1}{2}, 1, 0\right) \\ &= \left(-\frac{1}{\sqrt{6}}, \frac{1}{\sqrt{6}}, \frac{2}{\sqrt{6}}, 0\right) \end{aligned}$$

$$\text{Step 3: } v_3 \cdot u_1 = (1, 0, 1, 1) \cdot \left(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}, 0, 0 \right) = \frac{1}{\sqrt{2}}$$

$$v_3 \cdot u_2 = (1, 0, 1, 1) \cdot \left(-\frac{1}{\sqrt{6}}, \frac{1}{\sqrt{6}}, \frac{2}{\sqrt{6}}, 0 \right) = -\frac{1}{\sqrt{6}} + \frac{2}{\sqrt{6}} = \frac{1}{\sqrt{6}}$$

$$\begin{aligned} w_3 &= v_3 - (v_3 \cdot u_1) u_1 - (v_3 \cdot u_2) u_2 \\ &= (1, 0, 1, 1) - \frac{1}{\sqrt{2}} \left(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}, 0, 0 \right) - \frac{1}{\sqrt{6}} \left(-\frac{1}{\sqrt{6}}, \frac{1}{\sqrt{6}}, \frac{2}{\sqrt{6}}, 0 \right) \end{aligned}$$

$$\begin{aligned}
&= (1, 0, 1, 1) - \left(\frac{1}{2}, \frac{1}{2}, 0, 0\right) - \left(-\frac{1}{6}, \frac{1}{6}, \frac{2}{6}, 0\right) \\
&= \left(\frac{2}{3}, -\frac{2}{3}, \frac{2}{3}, 1\right) \\
u_3 &= \frac{w_3}{\|w_3\|} = \frac{1}{\sqrt{\left(\frac{2}{3}\right)^2 + \left(-\frac{2}{3}\right)^2 + \left(\frac{2}{3}\right)^2 + 1^2}} \left(\frac{2}{3}, -\frac{2}{3}, \frac{2}{3}, 1\right) \\
&= \frac{1}{\sqrt{\frac{21}{9}}} \left(\frac{2}{3}, -\frac{2}{3}, \frac{2}{3}, 1\right) \\
&= \frac{3}{\sqrt{21}} \left(\frac{2}{3}, -\frac{2}{3}, \frac{2}{3}, 1\right) \\
&= \left(\frac{2}{\sqrt{21}}, -\frac{2}{\sqrt{21}}, \frac{2}{\sqrt{21}}, \frac{3}{\sqrt{21}}\right)
\end{aligned}$$

QR-Decomposition

Problem

If A is an $m \times n$ matrix with linearly independent column vectors, and if Q is the matrix that results by applying the Gram-Schmidt process to the column vectors of A , what relationship, if any, exists between A and Q ?

To solve this problem, suppose that the column vectors of A are u_1, u_2, \dots, u_n and the orthonormal column vectors of Q are q_1, q_2, \dots, q_n .

$$\begin{aligned} u_1 &= \langle u_1, q_1 \rangle q_1 + \langle u_1, q_2 \rangle q_2 + \dots + \langle u_1, q_n \rangle q_n \\ u_2 &= \langle u_2, q_1 \rangle q_1 + \langle u_2, q_2 \rangle q_2 + \dots + \langle u_2, q_n \rangle q_n \\ &\vdots \\ u_n &= \langle u_n, q_1 \rangle q_1 + \langle u_n, q_2 \rangle q_2 + \dots + \langle u_n, q_n \rangle q_n \end{aligned}$$

$$R = \begin{bmatrix} \langle u_1, q_1 \rangle & \langle u_2, q_1 \rangle & \dots & \langle u_n, q_1 \rangle \\ 0 & \langle u_2, q_2 \rangle & \dots & \langle u_n, q_2 \rangle \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \langle u_n, q_n \rangle \end{bmatrix}$$

The equation $A = QR$ is a factorization of A into the product of a matrix Q with orthonormal column vectors and an invertible upper triangular matrix R . We call it the **QR-decomposition of A** .

Theorem

If A is an $m \times n$ matrix with linearly independent column vectors, then A can be factored as

$$A = QR$$

Where Q is an $m \times n$ matrix with orthonormal column vectors, and R is an $n \times n$ invertible upper triangular matrix.

Example

Find the QR -decomposition of

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

Solution

The column vectors of are

$$\mathbf{u}_1 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}, \quad \mathbf{u}_2 = \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix}, \quad \mathbf{u}_3 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

From the previous example

$$\mathbf{q}_1 = \left(\frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}} \right) \quad \mathbf{q}_2 = \left(-\frac{2}{\sqrt{6}}, \frac{1}{\sqrt{6}}, \frac{1}{\sqrt{6}} \right) \quad \mathbf{q}_3 = \left(0, -\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}} \right)$$

$$\begin{aligned} R &= \begin{bmatrix} \langle \mathbf{u}_1, \mathbf{q}_1 \rangle & \langle \mathbf{u}_2, \mathbf{q}_1 \rangle & \langle \mathbf{u}_3, \mathbf{q}_1 \rangle \\ 0 & \langle \mathbf{u}_2, \mathbf{q}_2 \rangle & \langle \mathbf{u}_3, \mathbf{q}_2 \rangle \\ 0 & 0 & \langle \mathbf{u}_3, \mathbf{q}_3 \rangle \end{bmatrix} \\ &= \begin{bmatrix} (1)\frac{1}{\sqrt{3}} + (1)\frac{1}{\sqrt{3}} + (1)\frac{1}{\sqrt{3}} & 0 + (1)\frac{1}{\sqrt{3}} + (1)\frac{1}{\sqrt{3}} & 0 + 0 + (1)\frac{1}{\sqrt{3}} \\ 0 & 0\left(\frac{-2}{\sqrt{6}}\right) + (1)\frac{1}{\sqrt{6}} + (1)\frac{1}{\sqrt{6}} & 0\left(\frac{-2}{\sqrt{6}}\right) + 0\frac{1}{\sqrt{6}} + (1)\frac{1}{\sqrt{6}} \\ 0 & 0 & 0 + (0)\frac{-1}{\sqrt{2}} + (1)\frac{1}{\sqrt{2}} \end{bmatrix} \\ &= \begin{bmatrix} \frac{3}{\sqrt{3}} & \frac{2}{\sqrt{3}} & \frac{1}{\sqrt{3}} \\ 0 & \frac{2}{\sqrt{6}} & \frac{1}{\sqrt{6}} \\ 0 & 0 & \frac{1}{\sqrt{2}} \end{bmatrix} \end{aligned}$$

$$\begin{pmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 1 \end{pmatrix} = \begin{pmatrix} \frac{1}{\sqrt{3}} & -\frac{2}{\sqrt{6}} & 0 \\ \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{6}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{2}} \end{pmatrix} \cdot \begin{pmatrix} \frac{3}{\sqrt{3}} & \frac{2}{\sqrt{3}} & \frac{1}{\sqrt{3}} \\ 0 & \frac{2}{\sqrt{6}} & \frac{1}{\sqrt{6}} \\ 0 & 0 & \frac{1}{\sqrt{2}} \end{pmatrix}$$

$\mathbf{A} \quad = \quad \mathbf{Q} \quad \mathbf{R}$

Exercises **Section 3.3 – Gram-Schmidt Process**

1. Use the Gram-Schmidt process to find an orthonormal basis for the subspaces of \mathbf{R}^m .

a) $\mathbf{u}_1 = (1, -3), \mathbf{u}_2 = (2, 2)$

b) $\mathbf{u}_1 = (1, 0), \mathbf{u}_2 = (3, -5)$

c) $\{(1, 1, 1), (-1, 1, 0), (1, 2, 1)\}$

d) $\{(1, 1, 1), (0, 1, 1), (0, 0, 1)\}$

e) $\{(1, 1, 1, 1), (1, 2, 1, 0), (1, 3, 0, 0)\}$

f) $\{(0, 2, -1, 1), (0, 0, 1, 1), (-2, 1, 1, -1)\}$

g) $\mathbf{u}_1 = (1, 0, 0), \mathbf{u}_2 = (3, 7, -2), \mathbf{u}_3 = (0, 4, 1)$

h) $\mathbf{u}_1 = (0, 2, 1, 0), \mathbf{u}_2 = (1, -1, 0, 0), \mathbf{u}_3 = (1, 2, 0, -1), \mathbf{u}_4 = (1, 0, 0, 1)$

2. Find the **QR**-decomposition of

a) $\begin{bmatrix} 1 & -1 \\ 2 & 3 \end{bmatrix}$

b) $\begin{bmatrix} 3 & 5 \\ -4 & 0 \end{bmatrix}$

c) $\begin{bmatrix} 1 & 2 \\ 0 & 1 \\ 1 & 4 \end{bmatrix}$

d) $\begin{bmatrix} 1 & 2 & 1 \\ 1 & 1 & 1 \\ 0 & 3 & 1 \end{bmatrix}$

e) $\begin{bmatrix} 1 & 0 & 1 \\ -1 & 1 & 1 \\ 1 & 0 & 1 \\ -1 & 1 & 1 \end{bmatrix}$

3. Verify that the Cauchy-Schwarz inequality holds for the given vectors using the Euclidean inner product

$\mathbf{u} = (0, -2, 2, 1), \mathbf{v} = (-1, -1, 1, 1)$

Section 3.4 – Orthogonal Matrices

Definition

A square matrix A is said to be orthogonal if its transpose is the same as its inverse, that is, if

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

or, equivalently, if

$$AA^T = A^T A = I$$

Example

The matrix $A = \begin{pmatrix} \frac{3}{7} & \frac{2}{7} & \frac{6}{7} \\ -\frac{6}{7} & \frac{3}{7} & \frac{2}{7} \\ \frac{2}{7} & \frac{6}{7} & -\frac{3}{7} \end{pmatrix}$

Solution

$$A^T A = \begin{pmatrix} \frac{3}{7} & -\frac{6}{7} & \frac{2}{7} \\ \frac{2}{7} & \frac{3}{7} & \frac{6}{7} \\ \frac{6}{7} & \frac{2}{7} & -\frac{3}{7} \end{pmatrix} \begin{pmatrix} \frac{3}{7} & \frac{2}{7} & \frac{6}{7} \\ -\frac{6}{7} & \frac{3}{7} & \frac{2}{7} \\ \frac{2}{7} & \frac{6}{7} & -\frac{3}{7} \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

Example

The matrix $A = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$

Solution

$$A^T A = \begin{bmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

Theorem

The following are equivalent for $n \times n$ matrix A .

- a) A is orthogonal.
- b) The row vectors of A form an orthonormal set in R^n with the Euclidean inner product.
- c) The column vectors of A form an orthonormal set in R^n with the Euclidean inner product.

Theorem

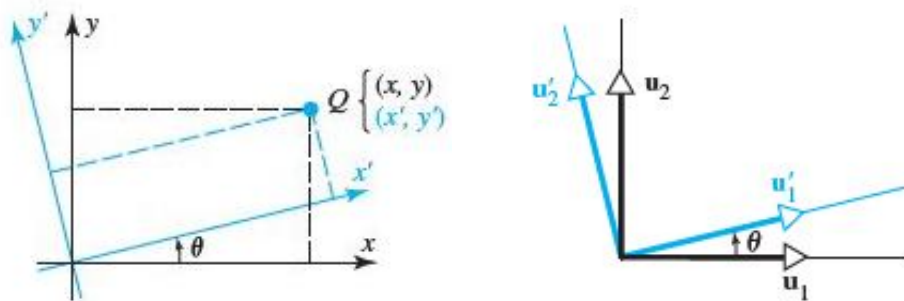
- a) The inverse of an orthogonal matrix is orthogonal
- b) A product of orthogonal matrices is orthogonal
- c) If A is orthogonal, then $\det(A) = 1$ or $\det(A) = -1$

Theorem

If A is an $n \times n$ matrix, then the following are equivalent

- a) A is orthogonal.
- b) $\|Ax\| = \|x\|$ for all x in R^n .
- c) $Ax \cdot Ay = x \cdot y$ for all x and y in R^n .

Let u_1 and u_2 be the unit vectors along the x - and y -axes and unit vectors u'_1 and u'_2 along the x' - and y' -axes.



The new coordinates (x', y') and the old coordinates (x, y) of a point Q will be related by

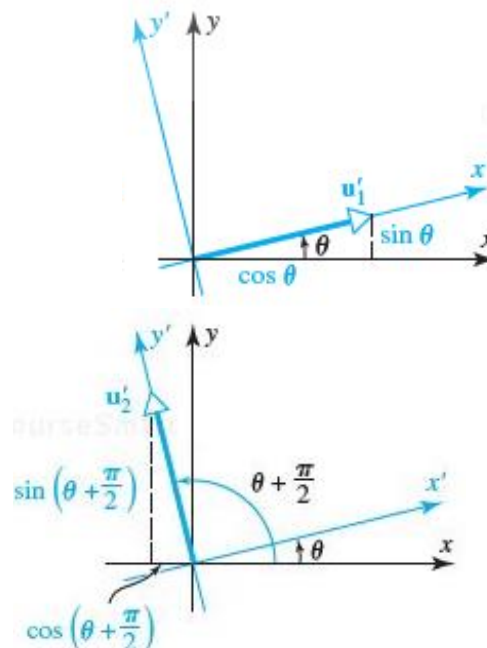
$$\begin{bmatrix} x' \\ y' \end{bmatrix} = P^{-1} \begin{bmatrix} x \\ y \end{bmatrix} \quad P = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$

$$P^{-1} = P^T = \begin{bmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{bmatrix}$$

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

$$\rightarrow \begin{cases} x' = x \cos \theta + y \sin \theta \\ y' = -x \sin \theta + y \cos \theta \end{cases}$$

These are sometimes called the **rotation equations**.



Example

Use the form $\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$ to find the new coordinates of the point $Q(2,1)$ if the coordinate axes of a rectangular coordinate system are rotated through an angle of $\theta = \frac{\pi}{4}$.

Solution

$$\begin{aligned} \begin{bmatrix} x' \\ y' \end{bmatrix} &= \begin{bmatrix} \cos \frac{\pi}{4} & \sin \frac{\pi}{4} \\ -\sin \frac{\pi}{4} & \cos \frac{\pi}{4} \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} \\ &= \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix} \begin{bmatrix} 2 \\ 1 \end{bmatrix} \\ &= \begin{bmatrix} \frac{3}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{bmatrix} \end{aligned}$$

The new coordinates of Q are $(x', y') = \left(\frac{3}{\sqrt{2}}, -\frac{1}{\sqrt{2}} \right)$

Exercises Section 3.4 – Orthogonal Matrices

1. Show that the matrix is orthogonal

$$a) \quad A = \begin{bmatrix} \frac{4}{5} & 0 & -\frac{3}{5} \\ -\frac{9}{25} & \frac{4}{5} & -\frac{12}{25} \\ \frac{12}{25} & \frac{3}{5} & \frac{16}{25} \end{bmatrix}$$

$$b) \quad A = \begin{bmatrix} \frac{1}{3} & \frac{2}{3} & \frac{2}{3} \\ \frac{2}{3} & -\frac{2}{3} & \frac{1}{3} \\ -\frac{2}{3} & -\frac{1}{3} & \frac{2}{3} \end{bmatrix}$$

2. Determine if the matrix is orthogonal. For those that is orthogonal find the inverse.

$$a) \quad \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$b) \quad \begin{bmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix}$$

$$c) \quad \begin{bmatrix} 0 & 1 & \frac{1}{\sqrt{2}} \\ 1 & 0 & 0 \\ 0 & 0 & \frac{1}{\sqrt{2}} \end{bmatrix}$$

$$d) \quad \begin{bmatrix} -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{3}} \\ 0 & -\frac{2}{\sqrt{6}} & \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{3}} \end{bmatrix}$$

$$e) \quad \begin{bmatrix} \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & -\frac{5}{6} & \frac{1}{6} & \frac{1}{6} \\ \frac{1}{2} & \frac{1}{6} & \frac{1}{6} & -\frac{5}{6} \\ \frac{1}{2} & \frac{1}{6} & -\frac{5}{6} & \frac{1}{6} \end{bmatrix}$$

$$f) \quad \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & \frac{1}{\sqrt{3}} & -\frac{1}{2} & 0 \\ 0 & \frac{1}{\sqrt{3}} & 0 & 1 \\ 0 & \frac{1}{\sqrt{3}} & \frac{1}{2} & 0 \end{bmatrix}$$

3. Prove that if A is orthogonal, then A^T is orthogonal.
4. Find a last column so that the resulting matrix is orthogonal

$$\begin{bmatrix} \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{6}} & \cdots \\ \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{6}} & \cdots \\ -\frac{1}{\sqrt{3}} & \frac{2}{\sqrt{6}} & \cdots \end{bmatrix}$$

5. Determine if the given matrix is orthogonal. If it is, find its inverse

$$\begin{bmatrix} \frac{1}{9} & \frac{4}{5} & \frac{3}{7} \\ \frac{4}{9} & \frac{3}{5} & -\frac{2}{7} \\ \frac{8}{9} & -\frac{2}{5} & \frac{3}{7} \end{bmatrix}$$

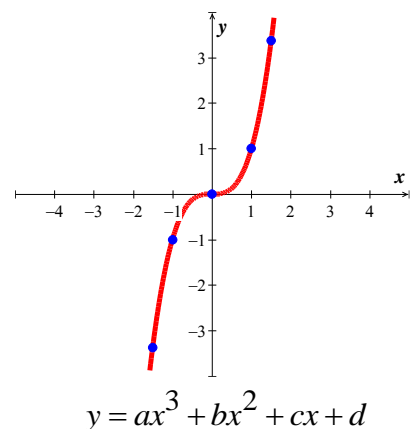
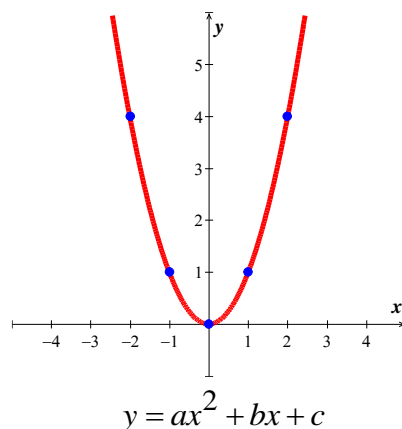
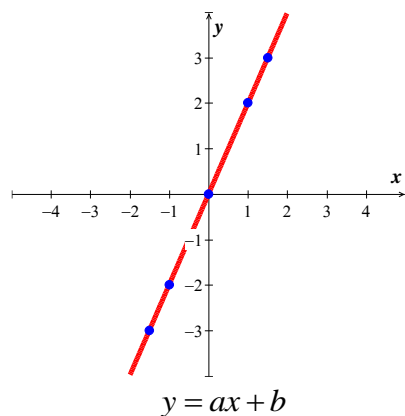
Section 3.5 – Least Squares Analysis

The use to **best** fit data, we will use results about orthogonal projections in inner product spaces to obtain a technique for fitting a line or other polynomial.

Fitting a Curve to Data

The common problem is to obtain a mathematical relationship between 2 variables x and y by **fitting** a curve to points in the xy -plane.

Some possibility of fitting the data



Least Squares Fit of a Straight Line

Recall that a system of equations $A\mathbf{x} = \mathbf{y}$ is called inconsistent if it does not have a solution. Suppose we want to fit a straight line $y = mx + b$ to the determined points $(x_1, y_1), \dots, (x_n, y_n)$

If the data points were collinear, the line would pass through all n points and the unknown coefficients m and b would satisfy the equations

$$\begin{array}{rcl} y_1 = mx_1 + b \\ y_2 = mx_2 + b \\ \vdots \quad \quad \quad \vdots \\ y_n = mx_n + b \end{array} \Rightarrow \begin{bmatrix} x_1 & 1 \\ x_2 & 1 \\ \vdots & \vdots \\ x_n & 1 \end{bmatrix} \begin{bmatrix} m \\ b \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}$$

$$A \quad \mathbf{x} = \mathbf{y}$$

The problem is to find m and b that minimize the errors in some sense.

Least Square Problem

Given a linear system $A\mathbf{x} = \mathbf{y}$ of m equations in n unknowns, find a vector \mathbf{x} that minimizes $\|\mathbf{y} - A\mathbf{x}\|$ with respect to the Euclidean inner product on \mathbf{R}^m . We call such as \mathbf{x} a least squares solution of the system, we call $\mathbf{y} - A\mathbf{x}$ the least squares error vectors, and we call $\|\mathbf{y} - A\mathbf{x}\|$ the least squares error.

$$A\mathbf{x} = \begin{pmatrix} e_1 \\ e_2 \\ \vdots \\ e_m \end{pmatrix}$$

The term “*least square solution*” results from the fact the minimizing $\|\mathbf{y} - A\mathbf{x}\| = e_1^2 + e_2^2 + \dots + e_m^2$

Example

Find the sums of squares of the errors of (2, 4), (4, 8), (6, 6)

Solution

$$4 = 2m + b \Rightarrow 4 - 2m - b = e_1$$

$$8 = 4m + b \Rightarrow 8 - 4m - b = e_2$$

$$6 = 6m + b \Rightarrow 6 - 6m - b = e_3$$

$$e_1^2 + e_2^2 + \dots + e_m^2 = (4 - 2m - b)^2 + (8 - 4m - b)^2 + (6 - 6m - b)^2$$

The least squares problem for this example to find the values m and b for which $e_1^2 + e_2^2 + \dots + e_m^2$ is a minimum.

Theorem

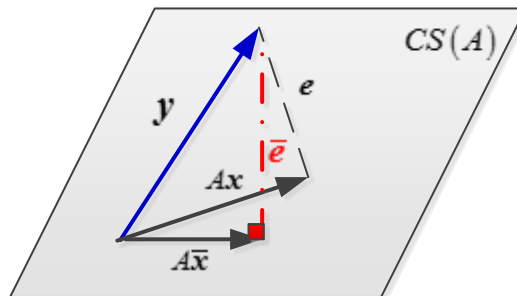
If A is an $m \times n$ matrix, the equation $A\mathbf{x} = \mathbf{y}$ has a solution if and only if \mathbf{y} is in the column space of A .

$$\mathbf{y} - A\mathbf{x} = \mathbf{e}$$

$A\mathbf{x}$ is a vector that is in the column space of A . For this A the column space is a plane is \mathbf{R}^m

\mathbf{y} is a vector, not in the column space of A (otherwise $A\mathbf{x} = \mathbf{y}$ has an exact solution)

\mathbf{e} is the error vector, the difference between \mathbf{y} and $A\mathbf{x}$



The length $\|\mathbf{e}\|$ is a minimum exactly when $\mathbf{e} \perp CS(A)$

Best Approximation *Theorem*

If $CS(A)$ is a finite dimensional subspace of an inner product space, and if \mathbf{y} is a vector in V , then $proj_{CS(A)} \mathbf{y}$ is the best approximation to \mathbf{y} from $CS(A)$ in the sense that

$$\left\| \mathbf{y} - proj_{CS(A)} \mathbf{y} \right\| < \left\| \mathbf{y} - \mathbf{v} \right\|$$

For every vector \mathbf{w} in $CS(A)$ that is different from $proj_{CS(A)} \mathbf{y}$

Theorem

For every linear system $A\mathbf{x} = \mathbf{y}$, the associated normal system

$$A^T A \mathbf{x} = A^T \mathbf{y}$$

is consistent, and all solutions are least squares solutions of $A\mathbf{x} = \mathbf{y}$

If the columns of A are linearly independent, then $A^T A$ is invertible so has a unique solution $\bar{\mathbf{x}}$. This solution is often expressed theoretically as

$$\begin{aligned} (A^T A)^{-1} A^T A \bar{\mathbf{x}} &= (A^T A)^{-1} A^T \mathbf{y} \\ \bar{\mathbf{x}} &= (A^T A)^{-1} A^T \mathbf{y} \end{aligned}$$

Proof

Let the vector $\bar{\mathbf{x}}$ is a least squares solution to $A\mathbf{x} = \mathbf{y} \Leftrightarrow (\mathbf{y} - A\bar{\mathbf{x}}) \perp CS(A)$

$$(\mathbf{y} - A\bar{\mathbf{x}}) \cdot \mathbf{z} = 0 \quad \mathbf{z} \text{ in } CS(A) \quad \& \quad \mathbf{z} = A\mathbf{w}$$

$$(\mathbf{y} - A\bar{\mathbf{x}}) \cdot A\mathbf{w} = 0 \quad \mathbf{w} \text{ in } \mathbf{R}^n$$

$$A^T (\mathbf{y} - A\bar{\mathbf{x}}) \cdot \mathbf{w} = 0$$

$$A^T (\mathbf{y} - A\bar{\mathbf{x}}) = 0$$

$$A^T \mathbf{y} - A^T A \bar{\mathbf{x}} = 0$$

$$A^T \mathbf{y} = A^T A \bar{\mathbf{x}}$$

Theorem

If A is an $m \times n$ matrix, then the following are equivalent

- a) A has linearly independent column vectors.
- b) $A^T A$ is invertible.

Example

Find the equation of the line that best fits the given points in the least-squares sense.

(40, 482), (45, 467), (50, 452), (55, 432), (60, 421)

Solution

Let $y = mx + b$ be the equation of the line that best fits the given points. Then

$$\begin{pmatrix} 40 & 1 \\ 45 & 1 \\ 50 & 1 \\ 55 & 1 \\ 60 & 1 \end{pmatrix} \begin{pmatrix} m \\ b \end{pmatrix} = \begin{pmatrix} 482 \\ 467 \\ 452 \\ 432 \\ 421 \end{pmatrix}$$

$$\text{Where } A = \begin{pmatrix} 40 & 1 \\ 45 & 1 \\ 50 & 1 \\ 55 & 1 \\ 60 & 1 \end{pmatrix} \quad x = \begin{pmatrix} m \\ b \end{pmatrix} \quad y = \begin{pmatrix} 482 \\ 467 \\ 452 \\ 432 \\ 421 \end{pmatrix}$$

Using the normal equation formula: $A^T A x = A^T y$

$$\begin{pmatrix} 40 & 45 & 50 & 55 & 60 \\ 1 & 1 & 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} 40 & 1 \\ 45 & 1 \\ 50 & 1 \\ 55 & 1 \\ 60 & 1 \end{pmatrix} \begin{pmatrix} m \\ b \end{pmatrix} = \begin{pmatrix} 40 & 45 & 50 & 55 & 60 \\ 1 & 1 & 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} 482 \\ 467 \\ 452 \\ 432 \\ 421 \end{pmatrix}$$

$$\begin{pmatrix} 12,750 & 250 \\ 250 & 5 \end{pmatrix} \begin{pmatrix} m \\ b \end{pmatrix} = \begin{pmatrix} 111,970 \\ 2,255 \end{pmatrix}$$

$$X = A^{-1}B$$

$$\begin{pmatrix} m \\ b \end{pmatrix} = \frac{1}{1250} \begin{pmatrix} 5 & -250 \\ -250 & 12,750 \end{pmatrix} \begin{pmatrix} 111,970 \\ 2,255 \end{pmatrix} = \begin{pmatrix} -3.12 \\ 607 \end{pmatrix}$$

Thus $y = -3.12x + 607$

Or

$$\begin{pmatrix} 12,750 & 250 & 111,970 \\ 250 & 5 & 2,225 \end{pmatrix} \xrightarrow{rref} \begin{pmatrix} 1 & 0 & -3.12 \\ 0 & 1 & 607 \end{pmatrix}$$

Example

Given the system equation:
$$\begin{cases} x_1 - x_2 = 4 \\ 3x_1 + 2x_2 = 1 \\ -2x_1 + 4x_2 = 3 \end{cases}$$

- a) Find the least-squares solution of the linear system $A\mathbf{x} = \mathbf{y}$
- b) Find the orthogonal projection of \mathbf{y} on the column space of A
- c) Find the error vector and the error

Solution

$$a) \quad A = \begin{pmatrix} 1 & -1 \\ 3 & 2 \\ -2 & 4 \end{pmatrix} \quad \mathbf{x} = \begin{pmatrix} m \\ b \end{pmatrix} \quad \mathbf{y} = \begin{pmatrix} 4 \\ 1 \\ 3 \end{pmatrix}$$

$$A^T A \mathbf{x} = A^T \mathbf{y}$$

$$\begin{pmatrix} 1 & 3 & -2 \\ -1 & 2 & 4 \end{pmatrix} \begin{pmatrix} 1 & -1 \\ 3 & 2 \\ -2 & 4 \end{pmatrix} \begin{pmatrix} m \\ b \end{pmatrix} = \begin{pmatrix} 1 & 3 & -2 \\ -1 & 2 & 4 \end{pmatrix} \begin{pmatrix} 4 \\ 1 \\ 3 \end{pmatrix}$$

$$\begin{pmatrix} 14 & -3 \\ -3 & 21 \end{pmatrix} \begin{pmatrix} m \\ b \end{pmatrix} = \begin{pmatrix} 1 \\ 10 \end{pmatrix}$$

$$\begin{pmatrix} m \\ b \end{pmatrix} = \frac{1}{285} \begin{pmatrix} 21 & 3 \\ 3 & 14 \end{pmatrix} \begin{pmatrix} 1 \\ 10 \end{pmatrix} = \begin{pmatrix} \frac{51}{285} \\ \frac{143}{285} \end{pmatrix} \quad X = A^{-1}B$$

$$= \begin{pmatrix} \frac{17}{95} \\ \frac{143}{285} \end{pmatrix}$$

Thus $y = 0.1789x + 0.5018$

- b) The orthogonal projection of \mathbf{y} on the column space of A

$$A\mathbf{x} = \begin{pmatrix} 1 & -1 \\ 3 & 2 \\ -2 & 4 \end{pmatrix} \begin{pmatrix} \frac{17}{95} \\ \frac{143}{285} \end{pmatrix} = \begin{pmatrix} -\frac{92}{285} \\ \frac{439}{285} \\ \frac{94}{57} \end{pmatrix}$$

$$c) \quad \mathbf{y} - A\mathbf{x} = \begin{pmatrix} 4 \\ 1 \\ 3 \end{pmatrix} - \begin{pmatrix} -\frac{92}{285} \\ \frac{439}{285} \\ \frac{94}{57} \end{pmatrix} = \begin{pmatrix} \frac{1232}{285} \\ -\frac{154}{285} \\ \frac{4}{3} \end{pmatrix}$$

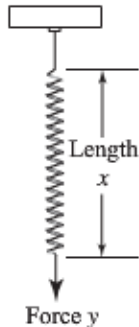
$$\text{The error: } \|\mathbf{y} - A\mathbf{x}\| = \sqrt{\left(\frac{1232}{285}\right)^2 + \left(-\frac{154}{285}\right)^2 + \left(\frac{4}{3}\right)^2} \approx 4.556$$

Exercises Section 3.5 – Least Squares Analysis

1. Find the equation of the line that best fits the given points in the least-squares sense.
 - a) $\{(0, 2), (1, 2), (2, 0)\}$
 - b) $\{(1, 5), (2, 4), (3, 1), (4, 1), (5, -1)\}$
 - c) $\{(0, 1), (1, 3), (2, 4), (3, 4)\}$
 - d) $\{(-2, 0), (-1, 0), (0, 1), (1, 3), (2, 5)\}$

2. Find the orthogonal projection of the vector \mathbf{u} on the subspace of \mathbf{R}^4 spanned by the vectors
 - a) $\mathbf{u} = (-3, -3, 8, 9); \quad \mathbf{v}_1 = (3, 1, 0, 1), \quad \mathbf{v}_2 = (1, 2, 1, 1), \quad \mathbf{v}_3 = (-1, 0, 2, -1)$
 - b) $\mathbf{u} = (6, 3, 9, 6); \quad \mathbf{v}_1 = (2, 1, 1, 1), \quad \mathbf{v}_2 = (1, 0, 1, 1), \quad \mathbf{v}_3 = (-2, -1, 0, -1)$
 - c) $\mathbf{u} = (-2, 0, 2, 4); \quad \mathbf{v}_1 = (1, 1, 3, 0), \quad \mathbf{v}_2 = (-2, -1, -2, 1), \quad \mathbf{v}_3 = (-3, -1, 1, 3)$

3. Find the standard matrix for the orthogonal projection P of \mathbf{R}^2 on the line passes through the origin and makes an angle θ with the positive x -axis.

4. Hooke's law in physics states that the length x of a uniform spring is a linear function of the force y applied to it. If we express the relationship as $y = mx + b$, then the coefficient m is called the spring constant. Suppose a particular unstretched spring has a measured length of 6.1 inches. (i.e., $x = 6.1$ when $y = 0$). Forces of 2 pounds, 4 pounds, and 6 pounds are then applied to the spring, and the corresponding lengths are found to be 7.6 inches, 8.7 inches, and 10.4 inches. Find the spring constant.
 

5. Prove: If A has a linearly independent column vectors, and if \mathbf{b} is orthogonal to the column space of A , then the least squares solution of $A\mathbf{x} = \mathbf{b}$ is $\mathbf{x} = \mathbf{0}$.

6. Let A be an $m \times n$ matrix with linearly independent row vectors. Find a standard matrix for the orthogonal projection of \mathbf{R}^n onto the row space of A .

7. Let W be the line with parametric equations $x = 2t, \quad t = -t, \quad z = 4t$
 - a) Find a basis for W .
 - b) Find the standard matrix for the orthogonal projection on W .
 - c) Use the matrix in part (b) to find the orthogonal projection of a point $P_0(x_0, y_0, z_0)$ on W .
 - d) Find the distance between the point $P_0(2, 1, -3)$ and the line W .

8. In \mathbf{R}^3 , consider the line l given by the equations $x = t, \quad t = t, \quad z = t$
 And the line m given by the equations $x = s, \quad t = 2s - 1, \quad z = 1$
 Let P be the point on l , and let Q be a point on m . Find the values of t and s that minimize the distance between the lines by minimizing the squared distance $\|P - Q\|^2$

9. Determine whether the statement is true or false,
- a) If A is an $m \times n$ matrix, then $A^T A$ is a square matrix.
 - b) If $A^T A$ is invertible, then A is invertible.
 - c) If A is invertible, then $A^T A$ is invertible.
 - d) If $A\mathbf{x} = \mathbf{b}$ is a consistent linear system, then $A^T A\mathbf{x} = A^T \mathbf{b}$ is also consistent.
 - e) If $A\mathbf{x} = \mathbf{b}$ is an inconsistent linear system, then $A^T A\mathbf{x} = A^T \mathbf{b}$ is also inconsistent.
 - f) Every linear system has a least squares solution.
 - g) Every linear system has a unique least squares solution.
 - h) If A is an $m \times n$ matrix with linearly independent columns and \mathbf{b} is in R^m , then $A\mathbf{x} = \mathbf{b}$ has a unique least squares solution.