

Orthogonal Sets

Linear Algebra

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Orthogonal Sets

Definition

A set of vectors $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p\}$ in \mathbb{R}^n is said to be an **orthogonal set** if $\mathbf{v}_i \cdot \mathbf{v}_j = 0$ when $i \neq j$ (ie, the vectors are pairwise orthogonal).

Example

Is $\left\{ \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}$ an orthogonal set?

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Since

$$\mathbf{v}_1 \cdot \mathbf{v}_2 = 0, \quad \mathbf{v}_1 \cdot \mathbf{v}_3 = 0, \quad \text{and} \quad \mathbf{v}_2 \cdot \mathbf{v}_3 = 0,$$

the set is indeed an orthogonal set.

Linear Independence of Orthogonal Sets

Theorem

If $S = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p\}$ is an **orthogonal set** of **nonzero** vectors in \mathbb{R}^n , then

S is a **linearly independent** set, and is therefore a **basis** for the subspace spanned by S .

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Proof.

Suppose S is an orthogonal set and that the vectors in S are linearly **dependent**. This implies there exist scalars c_1, c_2, \dots, c_p , with at least one $c_i \neq 0$, such that

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

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$$\begin{aligned} 0 &= \mathbf{0} \cdot \mathbf{v}_1 = (c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \dots + c_p\mathbf{v}_p) \cdot \mathbf{v}_1 \\ &= c_1(\mathbf{v}_1 \cdot \mathbf{v}_1) + c_2(\mathbf{v}_2 \cdot \mathbf{v}_1) + \dots + c_p(\mathbf{v}_p \cdot \mathbf{v}_1) \\ &= c_1(\mathbf{v}_1 \cdot \mathbf{v}_1). \end{aligned}$$

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Since \mathbf{v}_1 is nonzero, we have $c_1 = 0$. Similarly, we can show that $c_1 = c_2 = \dots = c_p = 0$, which contradicts our original assumption. Thus S is **linearly independent**.



Orthogonal Bases

Definition

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

Example

► The set $S_1 = \left\{ \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}$ is an orthogonal basis for \mathbb{R}^3 .

► The set $S_2 = \left\{ \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}$ is a basis for \mathbb{R}^3 that is **not** orthogonal.

Finding Coordinates With an Orthogonal Basis

Suppose that $\mathcal{B} = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_p\}$ is an **orthogonal basis** for a subspace W of \mathbb{R}^n .

Find \mathcal{B} -coordinates of \mathbf{y} in W .

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Theorem

Let $S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_p\}$ be an orthogonal basis for a subspace W of \mathbb{R}^n . For each \mathbf{y} in W , the weights in the linear combination

$\mathbf{y} = c_1\mathbf{u}_1 + c_2\mathbf{u}_2 + \dots + c_p\mathbf{u}_p$ are given by

$$c_j = \frac{\mathbf{y} \cdot \mathbf{u}_j}{\mathbf{u}_j \cdot \mathbf{u}_j} \quad \text{for } j = 1, 2, \dots, p.$$

Finding Coordinates with an Orthogonal Basis

Example

Express the vector $\mathbf{y} = \begin{bmatrix} 3 \\ 7 \\ 4 \end{bmatrix}$ as a linear combination of the orthogonal basis $\left\{ \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}$.

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Thus we have

$$\mathbf{y} = \begin{bmatrix} 3 \\ 7 \\ 4 \end{bmatrix} = -2 \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix} + 5 \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} + 4 \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}.$$

$$c_1 = \frac{\mathbf{y} \cdot \mathbf{u}_1}{\mathbf{u}_1 \cdot \mathbf{u}_1} = \frac{3 - 7}{2} = -2$$

$$c_2 = \frac{\mathbf{y} \cdot \mathbf{u}_2}{\mathbf{u}_2 \cdot \mathbf{u}_2} = \frac{3 + 7}{2} = 5$$

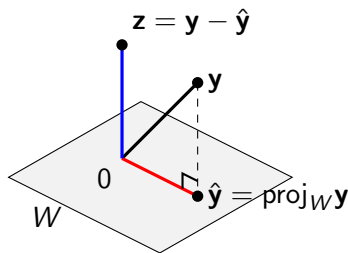
$$c_3 = \frac{\mathbf{y} \cdot \mathbf{u}_3}{\mathbf{u}_3 \cdot \mathbf{u}_3} = \frac{4}{1} = 4$$

Orthogonal Projections

- ▶ In physics, forces are generally expressed as vectors, \mathbf{F} .
- ▶ An object is moving with velocity \mathbf{v} .
- ▶ We often want to decompose the force \mathbf{F} into two components $\mathbf{F} = \mathbf{F}_{\text{perp}} + \mathbf{F}_{\text{parallel}}$

Given a nonzero vector \mathbf{u} in \mathbb{R}^n , we wish to decompose a vector \mathbf{y} in \mathbb{R}^n into the sum of two vectors, one that is **parallel** to \mathbf{u} and the other is **orthogonal** to \mathbf{u} :

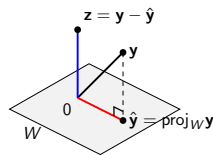
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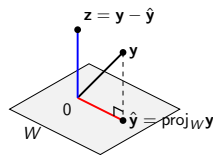


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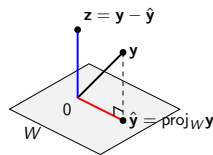
We want to find $\hat{\mathbf{y}} = \text{proj}_{\mathbf{u}}\mathbf{y} = c\mathbf{u}$. The other component $\mathbf{z} = \mathbf{y} - \hat{\mathbf{y}}$ is **orthogonal** to \mathbf{u} , so

$$0 = (\mathbf{y} - c\mathbf{u}) \cdot \mathbf{u} = (\mathbf{y} \cdot \mathbf{u}) - c(\mathbf{u} \cdot \mathbf{u}).$$



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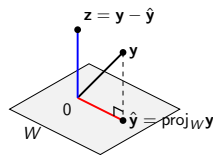
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Definition

The vector $\hat{\mathbf{y}}$ is called the **orthogonal projection of \mathbf{y} onto \mathbf{u}** , and the vector \mathbf{z} is called the **component of \mathbf{y} orthogonal to \mathbf{u}** . We have

$$\hat{\mathbf{y}} = \text{proj}_{\mathbf{u}}\mathbf{y} = \left(\frac{\mathbf{y} \cdot \mathbf{u}}{\mathbf{u} \cdot \mathbf{u}} \right) \mathbf{u}, \quad \mathbf{z} = \mathbf{y} - \hat{\mathbf{y}}.$$

Application of Projections

Example

Suppose a sailboat has constant velocity given by $\mathbf{v} = \begin{bmatrix} 3 \\ 4 \end{bmatrix}$. The wind has a constant velocity of $\mathbf{w} = \begin{bmatrix} 10 \\ -10 \end{bmatrix}$. Find the component of the force that in the direction of the sailboat's motion.

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We have

$$\hat{\mathbf{w}} = \text{proj}_{\mathbf{v}} \mathbf{w} = \left(\frac{\mathbf{w} \cdot \mathbf{v}}{\mathbf{v} \cdot \mathbf{v}} \right) \mathbf{v} = \frac{-10}{25} \mathbf{v} = \begin{bmatrix} -6/5 \\ -8/5 \end{bmatrix}$$

$$\mathbf{z} = \mathbf{w} - \hat{\mathbf{w}} = \begin{bmatrix} 56/5 \\ -42/5 \end{bmatrix}$$

And we can verify that

$$\hat{\mathbf{w}} \cdot \mathbf{z} = \begin{bmatrix} -6/5 \\ -8/5 \end{bmatrix} \cdot \begin{bmatrix} 56/5 \\ -42/5 \end{bmatrix} = 0.$$

Orthonormal Sets

Definition

A set $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_p\}$ is an **orthonormal set** if it is an orthogonal set of **unit** vectors.

Example

The set $S = \{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n\}$ is an orthonormal set that forms a basis for \mathbb{R}^n .

- ▶ Orthonormal sets (bases) are particularly nice to work with as many formulas simplify considerably.
- ▶ For example, $\hat{\mathbf{y}} = \text{proj}_{\mathbf{u}} \mathbf{y} = \left(\frac{\mathbf{y} \cdot \mathbf{u}}{\mathbf{u} \cdot \mathbf{u}} \right) \mathbf{u} = (\mathbf{y} \cdot \mathbf{u}) \mathbf{u}$.

Matrices Constructed from Orthonormal Sets

Let $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_p\}$ be an **orthonormal set** in \mathbb{R}^n and let $U = [\mathbf{u}_1 \ \mathbf{u}_2 \ \dots \ \mathbf{u}_p]_{n \times p}$. Then

$$U^T U = \begin{bmatrix} \mathbf{u}_1^T \\ \mathbf{u}_2^T \\ \vdots \\ \mathbf{u}_p^T \end{bmatrix}_{p \times n} [\mathbf{u}_1 \ \mathbf{u}_2 \ \dots \ \mathbf{u}_p]_{n \times p}$$

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Theorem

The columns of an $m \times n$ matrix U form an **orthonormal set** if and only if $U^T U = I_{n \times n}$.

Linear Mapping from Matrix whose Columns form an Orthonormal Set

Theorem

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

- (a) $(U\mathbf{x}) \cdot (U\mathbf{y}) = \mathbf{x} \cdot \mathbf{y}$.
- (b) $\|U\mathbf{x}\| = \|\mathbf{x}\|$.
- (c) $(U\mathbf{x}) \cdot (U\mathbf{y}) = 0$ if and only if $\mathbf{x} \cdot \mathbf{y} = 0$.

In other words, the linear mapping $\mathbf{x} \mapsto U\mathbf{x}$ preserves **length** and **orthogonality**.

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

(a) $(U\mathbf{x}) \cdot (U\mathbf{y}) = \mathbf{x} \cdot \mathbf{y}$.

(b) $\|U\mathbf{x}\| = \|\mathbf{x}\|$.

(c) $(U\mathbf{x}) \cdot (U\mathbf{y}) = 0$ if and only if $\mathbf{x} \cdot \mathbf{y} = 0$.

In other words, the linear mapping $\mathbf{x} \mapsto U\mathbf{x}$ preserves **length** and **orthogonality**.

Proof.

Statements (b) and (c) follow from statement (a). Note that

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

