

Inner Product Spaces: Orthogonality

Recall: If \mathbf{v} and \mathbf{w} are vectors in \mathbb{R}^n and θ is the angle between these two vectors, then

$$\cos \theta = \frac{\mathbf{v} \cdot \mathbf{w}}{\|\mathbf{v}\| \|\mathbf{w}\|}.$$

Thus, $\mathbf{v} \cdot \mathbf{w} = 0$ implies that \mathbf{v} and \mathbf{w} are perpendicular.

Definition: Let V be an inner product space. We say that \mathbf{v} and \mathbf{w} are **orthogonal**, denoted

Example: Consider $M_{2 \times 2}(\mathbb{R})$ with the inner product defined by

$$\langle A, B \rangle = \text{tr}(A^T B).$$

Let

$$A = \begin{bmatrix} 1/2 & 1/2 \\ 1/2 & 1/2 \end{bmatrix}, \quad B = \begin{bmatrix} 1/2 & -1/2 \\ -1/2 & 1/2 \end{bmatrix}, \quad \text{and} \quad C = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}.$$

(Pythagorean) Theorem: Let V be an inner product space. If \mathbf{v} and \mathbf{w} are orthogonal, then

$$\|\mathbf{v} + \mathbf{w}\|^2 = \|\mathbf{v}\|^2 + \|\mathbf{w}\|^2.$$

Lemma: Let V be an inner product space and $\mathbf{v}, \mathbf{w} \in V$ such that $\mathbf{w} \neq \mathbf{0}$. Then \mathbf{w} is orthogonal to $\mathbf{v} - \frac{\langle \mathbf{v}, \mathbf{w} \rangle}{\langle \mathbf{w}, \mathbf{w} \rangle} \mathbf{w}$.

Cauchy-Schwarz Inequality: Let V be an inner product space. Then

$$|\langle \mathbf{v}, \mathbf{w} \rangle| \leq \|\mathbf{v}\| \|\mathbf{w}\|.$$

The Cauchy-Schwarz Inequality is what allows us to make the following formal definition for the angle between vectors of inner product spaces over the real numbers.

Definition: Let V be a *real* inner product space. We define the **angle θ between \mathbf{v} and \mathbf{w} in V** by

We now list some properties of the norm of a vector in an inner product space.

Proposition: let V be an inner product space. The following properties hold for all $\mathbf{v}, \mathbf{w} \in V$ and scalars α :

1. $\|\alpha\mathbf{v}\| = |\alpha| \|\mathbf{v}\|$;
2. (**Triangle Inequality**) $\|\mathbf{v} + \mathbf{w}\| \leq \|\mathbf{v}\| + \|\mathbf{w}\|$;
3. $\|\mathbf{v}\| = 0 \implies \mathbf{v} = \mathbf{0}$.

Orthonormal Bases

Goal: To find bases that behave like the standard basis for \mathbb{R}^n with respect to the dot product.

Definition: Let V be an inner product space.

1. $\{\mathbf{v}_1, \dots, \mathbf{v}_k\} \subseteq V$ is called **orthogonal** if $\langle \mathbf{v}_i, \mathbf{v}_j \rangle = 0$ whenever $i \neq j$.
2. \mathbf{v} in V is called a **unit vector** if $\|\mathbf{v}\| = 1$.
3. $\{\mathbf{v}_1, \dots, \mathbf{v}_k\} \subseteq V$ is called **orthonormal** if

Examples:

1. Consider $M_{2 \times 2}(\mathbb{R})$ with the inner product defined by

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Let

$$A = \begin{bmatrix} 1/2 & 1/2 \\ 1/2 & 1/2 \end{bmatrix}, \quad B = \begin{bmatrix} 1/2 & -1/2 \\ -1/2 & 1/2 \end{bmatrix}, \quad \text{and} \quad C = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}.$$

$$2. \left\{ \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}, \begin{bmatrix} -1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} \right\} \subseteq \mathbb{R}^2$$

Proposition: If $\{\mathbf{v}_1, \dots, \mathbf{v}_k\}$ is an orthogonal set and $\mathbf{v}_i \neq \mathbf{0}$ for all i , then $\{\mathbf{v}_1, \dots, \mathbf{v}_k\}$ is linearly independent.

Definition: $\{\mathbf{v}_1, \dots, \mathbf{v}_k\}$ is called an **orthonormal basis** for an inner product space V if

Examples:

1. The standard basis for \mathbb{C}^n with respect to the standard Hermitian inner product is an orthonormal basis.

$$2. \left\{ \begin{bmatrix} 1/\sqrt{2} \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\}$$

3.

$$\left\{ \frac{1}{\sqrt{2}}, \sqrt{\frac{3}{2}}x, \sqrt{\frac{5}{2}} \left(\frac{3}{2}x^2 - \frac{1}{2} \right), \sqrt{\frac{7}{2}} \left(\frac{5}{2}x^3 - \frac{3}{2}x \right) \right\}$$

Question: Can we always find orthonormal bases? How? For this we will need something called projections!

Definition: Let V be an inner product space and $\mathbf{w}, \mathbf{v} \in V$ with $\mathbf{w} \neq \mathbf{0}$.

1. The **projection of \mathbf{v} onto \mathbf{w}** is

2. The **perpendicular vector of \mathbf{v} with respect to \mathbf{w}** is