

Orthogonal projection

Theorem: Let V be an inner product space, let \mathbf{v} be a vector in V , and let W be a finite dimensional linear subspace of V . Then there is a unique vector $\mathbf{w} := \pi_W(\mathbf{v}) \in W$ such that $\mathbf{v} - \mathbf{w} \in W^\perp$. Furthermore:

- $\pi_W(\mathbf{v})$ is the vector in W which is closest to \mathbf{v} .
- Let $\pi_W^\perp(\mathbf{v}) := \mathbf{v} - \pi_W(\mathbf{v})$. Then

$$\mathbf{v} = \pi_W(\mathbf{v}) + \pi_W^\perp(\mathbf{v}),$$

and $\pi_W(\mathbf{v}) \in W$ and $\pi_W^\perp(\mathbf{v}) \in W^\perp$.

- The maps

$$\pi_W: V \rightarrow W \quad \text{and} \quad \pi_W^\perp: V \rightarrow W^\perp$$

are linear transformations.

Orthogonality

Definition: A sequence of vectors $(\mathbf{w}_1, \dots, \mathbf{w}_m)$ in an inner product space V is *orthogonal* if $(\mathbf{w}_i | \mathbf{w}_j) = 0$ whenever $i \neq j$.

Theorem: Let $S := (\mathbf{w}_1, \dots, \mathbf{w}_m)$ be an orthogonal sequence of nonzero vectors in V .

- The sequence S is automatically linearly independent, hence is an ordered basis for $W := \text{span } S$.
- If \mathbf{v} in V , then

$$\pi_W(\mathbf{v}) = a_1 \mathbf{w}_1 + a_2 \mathbf{w}_2 + \dots + a_m \mathbf{w}_m, \quad \text{where}$$

$$a_i := \frac{(\mathbf{v} | \mathbf{w}_i)}{(\mathbf{w}_i | \mathbf{w}_i)}$$

Gram-Schmidt

Algorithm: Let V be an inner product space, let $(\mathbf{v}_1, \dots, \mathbf{v}_m)$ be a linearly independent sequence of vectors in V , and let $W_i := \text{span}(\mathbf{v}_1, \dots, \mathbf{v}_i)$ for i with $1 \leq i \leq m$. Then there is an *orthogonal* sequence $(\mathbf{w}_1, \dots, \mathbf{w}_m)$ such that for each i , $(\mathbf{w}_1, \dots, \mathbf{w}_i)$ is an orthogonal basis for W_i .

This is computed step by step, starting by taking $\mathbf{w}_1 := \mathbf{v}_1$. Now assume that $(\mathbf{w}_1, \dots, \mathbf{w}_i)$ is already computed. Then we can compute π_{W_i} and $\pi_{W_i}^\perp$. Hence we can compute

$$\mathbf{w}_{i+1} := \pi_{W_i}^\perp(\mathbf{v}_{i+1}) = \mathbf{v}_{i+1} - (a_1 \mathbf{w}_1 + a_2 \mathbf{w}_2 + \dots + a_i \mathbf{w}_i),$$

$$\text{where } a_i := \frac{(\mathbf{v}_{i+1} | \mathbf{w}_i)}{(\mathbf{w}_i | \mathbf{w}_i)}.$$

It is easy to check that this works.