

Orthogonal project and Gram-Schmidt orthogonalization

Zexi Sun

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1 Orthogonal projection

For a vector v its orthogonal projection $P_E v$ onto the subspace E is a vector w st

1. $w \in E$
2. $v - w \perp E$

We denote with $w = P_E v$ for the orthogonal projection.

- Existence
- Uniqueness (Thm3.2)
- Ways to find (Prep3.3, find $P_E v$ with an orthogonal basis)

2 Thm 3.2

The orthogonal projection $w = P_E v$ minimizes the distance from v to E . i.e,
 $\forall x \in E$,

$$\|v - w\| \leq \|v - x\|.$$

Moreover, if for some $x \in E$

$$\|v - w\| = \|v - x\|,$$

then $x = w$.

3 Prep 3.3

Let v_1, \dots, v_r be an orthogonal basis in E . Then the orthogonal projection $P_E v$ of a vector v is given by

$$P_E v = \sum_{k=1}^r \alpha_k v_k, \text{ where } \alpha_k = \frac{(v, v_k)}{\|v_k\|^2}.$$

In other words,

$$P_E v = \sum_{k=1}^r \frac{(v, v_k)}{\|v_k\|^2} v_k.$$

This formula applied to an orthogonal system (not a basis) gives us a projection onto its span.

It's also easy to see that P_E is a **linear transformation** by seeing the linearity of P_E from the def and uniqueness of orthogonal projection. (i.e, easy to check that for any x and y the vector $\alpha x + \beta y - (\alpha P_E x + \beta P_E y)$ is orthogonal any vector in E , so by def $P_E(\alpha x + \beta y) = \alpha P_E x + \beta P_E y$).

The matrix of P_E where E is in \mathbb{C}^n or \mathbb{R}^n is given by

$$P_E = \sum_{k=1}^r \frac{1}{\|v_k\|^2} v_k v_k^*$$

4 Gram-Schmidt orthogonalization algorithm

Gram-Schmidt constructs from a linearly independent system x_1, \dots, x_n an orthogonal system v_1, \dots, v_n st $\text{span}\{x_1, x_2, \dots, x_n\} = \text{span}\{v_1, v_2, \dots, v_n\}$.

Moreover, for all $r \leq n$, we have

$$\text{span}\{x_1, \dots, x_r\} = \text{span}\{v_1, \dots, v_r\}.$$

Step1: Define $v_1 = x_1$.

$$\text{Define } E_1 : \text{span}\{x_1\} = \text{span}\{v_1\}.$$

Step2: Define v_2 by

$$v_2 = x_2 - P_{E_1} x_2 = x_2 - \frac{(x_2, v_1)}{\|v_1\|^2} v_1.$$

Define $E_2 = \text{span}\{v_1, v_2\}$. (note that $\text{span}\{x_1, x_2\} = E_2$)

Step3: Define v_3 by

$$v_3 = x_3 - P_{E_2} x_3 = x_3 - \frac{(x_3, v_1)}{\|v_1\|^2} v_1 - \frac{(x_3, v_2)}{\|v_2\|^2} v_2.$$

Define $E_3 = \text{span}\{v_1, v_2, v_3\}$.

Step $r+1$: Define $v_{r+1} = x_{r+1} - P_{E_r} x_{r+1} = x_{r+1} - \sum_{k=1}^r \frac{(x_{r+1}, v_k)}{\|v_k\|^2} v_k$.

5 Orthogonal Complement

For a subspace E its orthogonal complement E^\perp is the set of all vectors orthogonal to E .

$$E^\perp = \{x : x \perp E\}$$

If $x, y \perp E$ then for any linear combination $\alpha x + \beta y \perp E$, therefore E^\perp is a subspace.

By def of orthogonal projection, any vector in an IPS V admits a unique representation $v = v_1 + v_2$, where $v_1 \in E, v_2 \in E^\perp$.

6 Prep 3.6

For a subspace E ,

$$(E^\perp)^\perp = E$$