

# Inner Product Spaces

Sriram Vadlamani

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# 1 Introduction

## 1.1 A bilinear Form

let  $E$  be an  $R$ -vs and  $\theta : E \times E \rightarrow R$ , we say,  $\theta$  is a bilinear form if:  
 $\forall (u, v, w) \in E^2$  and  $\forall a \in R$ ,

- $\theta(u + v, w) = \theta(u, w) + \theta(v, w)$
- $\theta(au, v) = a\theta(u, v)$

### Proposition 1

let  $E$  be an  $R$ -vs of finite dimension 'n'. We have  $B = (e_1, e_2, \dots, e_n)$  the basis of  $E$ .

$\theta : E \times E \rightarrow R$  a bilinear form.

Then  $\forall (x, y) \in E^2, \theta(x, y) = X^t \cdot M \cdot Y$

Where  $M$  is the matrix of the bilinear form defined as:

$$M = (\theta(e_i, e_j))$$

And  $X$  and  $Y$  are the coordinates of 'x' and 'y'.

# 2 Inner product

A space  $(E, \theta)$  is said to be an inner product space iff

- The bilinear form is '*symmetric*'
- The form is positive definite.

## 2.1 Symmetry

A bilinear form is symmetric iff

$$\forall (x, y) \in E^2$$

$$\theta(x, y) = \theta(y, x)$$

## 2.2 Positive definite

A bilinear form is said to be positive definite iff

$$\forall x \in E, \theta(x, x) \geq 0, \text{ and}$$

$$\forall x \in E, \theta(x, x) = 0 \implies x = 0$$

# 3 Theorems

## 3.1 Cauchy-Schwartz and Minkowski Theorems

**Cauchy-Schwartz:** Let  $E$  be an  $R$ -vs and  $\theta : E \rightarrow R$  a positive definite and symmetric bilinear form, then

$$\forall (x, y) \in E^2 \mid \theta(x, y) \mid \leq \sqrt{\theta(x, x)} \times \sqrt{\theta(y, y)}$$

**Minkowski's:** Let  $(E, \theta)$  be an inner product space on  $\mathbb{R}$ , then:  
 $\forall (x, y) \in E^2, \sqrt{\theta(x + y, x + y)} \leq \sqrt{\theta(x, x)} + \sqrt{\theta(y, y)}$

## 4 Orthogonality

We call  $N : E \rightarrow \mathbb{R}$  a norm  $\forall (x, y) \in E^2$  and  $\forall \lambda \in \mathbb{R}$  we have:

- $N(x) \geq 0$
- $N(\lambda \cdot x) = \lambda \cdot N(x)$
- $N(x) = 0 \iff x = 0$
- $N(x + y) \leq N(x) + N(y)$

and we say that 'N' is a norm. In geometry, this is what we call a triangular inequality.

**Proposition 2** The norm for any vector  $x \in E$  is  $\sqrt{\theta(x, x)}$ .

### 4.1 Pythagorean Theorem

for  $(E, \theta)$  an inner product space, and  $\forall (x, y) \in E^2$  such that  $\langle x, y \rangle = 0$ , we have:

$$\|x + y\|^2 = \|x\|^2 + \|y\|^2$$

### 4.2 The orthogonal of a subspace

For  $(E, \theta)$  an inner product space,  $A^\perp$  is defined as:  
 $\{\forall x \in E, \forall y \in A, \langle x, y \rangle = 0\}$  where  $A \subset E$

**Some important remarks** are:

- $A \subset B \Rightarrow B^\perp \subset A^\perp$
- $A^\perp = \text{span}(A)^\perp$
- $A \cap A^\perp \subset \{0\}$

## 5 Orthonormal sets

### Definition

Let  $(E, \langle, \rangle)$  be an inner product space on  $R$  and  $X = \{x_1, x_2, \dots, x_n\} \subset E$ . And  $B = \{e_1, e_2, \dots, e_n\}$  a basis of  $E$ . we say that  $X$  is an orthonormal set in  $E$  if:

$$\langle x_i, x_j \rangle = \delta_{ij} \text{ and}$$

$$\langle e_i, e_j \rangle \geq \delta_{ij}$$

$$\delta_{ij} = 1 \text{ if } i = j$$

$$\delta_{ij} = 0 \text{ if } i \neq j$$

**Proposition 7:** Any orthogonal set / family of vectors of non-zero vectors from an inner product space is linearly independent.

## 6 Gram-Schmidt Theorem

Let  $(E, \langle, \rangle)$  be a euclidian space and  $B = \{e_1, e_2, \dots, e_n\}$  a basis of  $E$ . Then there exists an orthogonal basis  $O = \{f_1, f_2, \dots, f_n\}$  from  $E$ . such that  $\forall k \in [1, n]$ ,  $f_k \in \text{span}(B)$ .

## 7 Theorem of orthogonal supplementary

Let  $(E, \langle, \rangle)$  be a euclidian space and 'F' a sub-vector space of  $E$ , then:

$$E = F \oplus F^\perp$$

**Corollary:**  $F^{\perp\perp} = F$ , only in a euclidian space, i.e, Finite dimension.

## 8 Orthogonal Projections

### Definition

Let  $(E, \langle, \rangle)$  be a euclidian space and  $F$  a sub-vector space of  $E$ . We call orthogonal projector on 'F' and we denote  $P_F$  the projector on  $F$  parallel to  $F^\perp$  i.e.,  $P_F \in L(E)$  such that:

$$P_F^2 = P_F, \text{ and } \text{Im}(P_F) = F \text{ and } \text{ker}(P_F) = F^\perp$$

**Proposition:**  $\forall x \in E$

$$P_F(x) = \sum_{i=1}^n \langle x, e_i \rangle \cdot e_i$$

## 9 Distance to a subspace

**Proposition:** Let  $F$  be a sub-vector space of a euclidian space  $(E, <, >)$ . Then, the map:

$y \Rightarrow \|x - y\| \quad \forall x \in E$  reaches it's minimum value at  $P_F(x)$ .

We call distance of  $x$  to  $F$  and denote  $d(x, F)$  the real number:  $\|x - P_F(x)\|$ .

## 10 Extra remarks

- If  $F$  is of finite dimension, then  $E = F \oplus F^\perp$
- if  $E$  is finite,  $F^{\perp\perp} = F$ , but if it's not, then  $F \subset F^{\perp\perp}$