

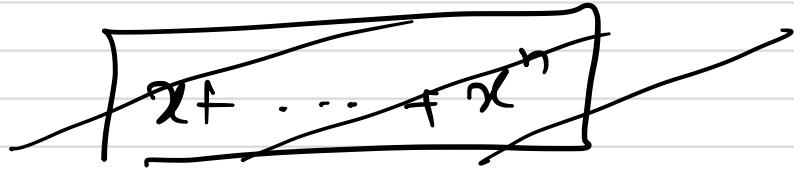
# Linear Algebra

## Lecture 16

Inner product spaces,  
orthogonality,



$$\alpha_0 + \alpha_1 x + x^2 + \cdots + \alpha_n x^n$$



$$1 + x + \dots + x^n$$

$$1 + x + \dots + x^n$$

$$\left\{ \frac{1}{x-\alpha} : \alpha \in \mathbb{C} \right\}$$

$$\frac{\beta_1}{x-\alpha_1} + \frac{\beta_2}{x-\alpha_2} + \cdots + \frac{\beta_n}{x-\alpha_n} = 0$$

$$\beta_1 + \beta_2 \frac{x-\alpha_1}{x-\alpha_2} + \cdots + \beta_n \frac{x-\alpha_n}{x-\alpha_n} = 0$$

Definition: Inner product.

An inner product or dot product is on a vector space  $V$  (over  $\mathbb{F}$ ) is a map  $\langle , \rangle : V \times V \rightarrow \mathbb{R}$  satisfying the following properties.

For  $x, y, z \in V$  and  $\alpha \in \mathbb{R}$

- i)  $\langle x, x \rangle \geq 0$  and  $\langle x, x \rangle = 0 \Leftrightarrow x = 0$
- ii)  $\langle x, y \rangle = \langle y, x \rangle$
- iii)  $\langle \alpha x + z, y \rangle = \langle x, y \rangle + \langle z, y \rangle$   
 $\langle x, y + z \rangle = \langle x, y \rangle + \langle x, z \rangle$
- iv)  $\langle \alpha x, y \rangle = \alpha \langle x, y \rangle$

$(V, \langle , \rangle)$  is called inner product space.

Example: Let  $V = \mathbb{R}^n$ , Define

$$\langle x, y \rangle = x^T y \quad \text{for every } x, y \in \mathbb{R}^n$$

Example: Let  $A \in \mathbb{R}^{n \times n}$

Define  $\langle \cdot, \cdot \rangle : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$  as

$$\langle x, y \rangle = x^T A y$$

$$x^T A y \stackrel{??}{=} y^T A x$$

$$\langle x, y \rangle \stackrel{??}{=} \langle y, x \rangle$$

$$\begin{aligned}\langle x, y+z \rangle &= \overbrace{x^T A(y+z)} \\ &= x^T A y + x^T A z \\ &= \langle x, y \rangle + \langle x, z \rangle\end{aligned}$$

$$\mathbb{R}^{2 \times 2}, \quad A = \begin{bmatrix} 1 & 0 \\ 0 & 3 \end{bmatrix}$$

$$A = \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix}$$

$$x = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

$$A = \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix}$$

$$A = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$$

$$x = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

$$\langle \alpha, \alpha \rangle$$

$$= \alpha^T A \alpha = (\alpha_1, \alpha_2) \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} \alpha_1 \\ \alpha_2 \end{pmatrix}$$

$$= 2\alpha_1^2 + 2\alpha_1\alpha_2 + \alpha_2^2$$

$$\left( \begin{smallmatrix} + \\ - \end{smallmatrix} \right), \left( \begin{smallmatrix} + \\ + \end{smallmatrix} \right), \left( \begin{smallmatrix} - \\ + \end{smallmatrix} \right), \left( \begin{smallmatrix} - \\ - \end{smallmatrix} \right)$$

$$A = \begin{bmatrix} a_{11} & a_{12} \\ a_{12} & a_{22} \end{bmatrix} = \begin{bmatrix} a & b \\ b & c \end{bmatrix} \begin{array}{l} f_{xx} \\ f_{xy} \\ f_{yx} \\ f_{yy} \end{array}$$

$$\boxed{\begin{aligned} a &> 0 \\ ac - b^2 &> 0 \end{aligned}}$$

Example: Let  $V = \mathbb{C}$  be a vector space over  $\mathbb{R}$ .

$$\langle z, w \rangle = \operatorname{Re}(z\bar{w})$$

Example: Let  $V = C[0, 1]$ , (the set of all continuous functions on  $[0, 1]$ ), over  $\mathbb{R}$  be the given vector space.

$$\langle f, g \rangle = \int_0^1 f(t)g(t)dt ; \quad \forall f, g \in C[0, 1]$$

$$\langle f, f \rangle = \int_0^1 [f(t)]^2 dt \geq 0 \quad \checkmark$$

$$\langle f, f \rangle = 0 \Leftrightarrow f = 0$$

$$(\text{Let } f^2(t) = g(t))$$

$$\int_0^{t_0+\delta} g(t)dt = 0$$

$$\int_0^{t_0-\delta} g(t)dt \geq \int_{t_0-\delta}^{t_0} g(t)dt \geq \int_{t_0-\delta}^{t_0} \frac{\epsilon}{2} dt = \epsilon \delta$$

Example:

Let  $V = M_{n \times n}(\mathbb{R})$  over  $\mathbb{R}$ .

$$\langle A, B \rangle = \text{trace}(AB^T)$$

Exercise: Let  $V$  be an inner product space.

$$\langle ax+by, cv+dw \rangle = ac\langle x, v \rangle$$

$$+ ad\langle x, w \rangle + bc\langle y, v \rangle + bd\langle y, w \rangle$$

In particular,

$$\langle x+y, x-y \rangle = \langle x, x \rangle - \langle y, y \rangle$$

$$\langle x+y, x+y \rangle = \langle x, x \rangle + 2\langle x, y \rangle + \langle y, y \rangle$$

## Definition (Norm)

Let  $V$  be an inner product space. The length or norm of a vector  $v \in V$  is denoted as  $\|v\|$  and is defined as the positive square root of  $\langle v, v \rangle$ .

$$\|v\| = \langle v, v \rangle^{1/2} = \sqrt{\langle v, v \rangle}$$

Lemma: Let  $V$  be an inner product space. The norm function  $\|\cdot\|: V \rightarrow \mathbb{R}$ , has following properties.

(1)  $\|x\| > 0$  and  $\|x\| = 0 \Leftrightarrow x = 0$

(2)  $\|ax\| = |a| \|x\| \quad \forall x \in V, a \in \mathbb{R}$ .

Furthermore, given a non-zero vector  $v \in V$ , there is a vector  $u \in V$  such that  $\|u\| = 1$  and  $v = \|v\| u$ .  $u$  is called as a unit vector in the direction of  $v$ .

## Cauchy-Schwarz Inequality.

Let  $V$  be an inner product space.

If  $x, y \in V$ , then

$$|\langle x, y \rangle| \leq \|x\| \|y\|$$

Further, the equality holds if and only if one is a multiple of another vector.

Proof: Fix  $x, y \in V$  and let  $y \neq 0$ .

$$f(t) = \langle x + ty, x + ty \rangle$$

$$= \langle x, x \rangle + (2t) \langle x, y \rangle + (t^2) \langle y, y \rangle$$

$$f'(t) = 2 \langle x, y \rangle + 2t \langle y, y \rangle$$

$$f'(t) = 0 \Rightarrow t = -\frac{\langle x, y \rangle}{\langle y, y \rangle} = t_0$$

is a point of extremum.

$$\Rightarrow 0 \leq f(t_0) \leq f(t)$$

$$\Rightarrow \langle x, x \rangle - 2 \frac{(\langle x, y \rangle)^2}{\langle y, y \rangle} + \frac{(\langle x, y \rangle)^2}{\langle y, y \rangle}$$

$$\geq 0$$

$$\Rightarrow \langle x, x \rangle \langle y, y \rangle \geq (\langle x, y \rangle)^2$$

$$\Rightarrow |\langle x, y \rangle| \leq \langle x, x \rangle^{\frac{1}{2}} \langle y, y \rangle^{\frac{1}{2}}$$

$$\Rightarrow |\langle x, y \rangle| \leq \|x\| \|y\|$$

□

Corollary

$$\frac{|\langle x, y \rangle|}{\|x\| \|y\|} \leq 1$$

$$|\gamma| \leq \frac{|\langle x, y \rangle|}{\|x\| \|y\|} \leq 1$$

$$\theta = \cos^{-1} \left( \frac{\langle x, y \rangle}{\|x\| \|y\|} \right) : \text{angle between } x \text{ & } y$$

Particular case of C-S inequality.

Let  $V = \mathbb{R}^n$

$$x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}, \quad y = \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}$$

$$\left| \sum_{i=1}^n x_i y_i \right| \leq \left( \sum_{i=1}^n x_i^2 \right)^{1/2} \left( \sum_{i=1}^n y_i^2 \right)^{1/2}$$

$$\Rightarrow \left( \sum_{i=1}^n x_i y_i \right)^2 \leq \left( \sum_{i=1}^n x_i^2 \right) \left( \sum_{i=1}^n y_i^2 \right)$$

Corollary: Let  $V$  be an inner product space. Then the norm function  $\|\cdot\| : V \rightarrow \mathbb{R}$  has the following properties.

i)  $\|x\| = 0 \iff x = 0 \quad \forall x \in V$

ii)  $\|ax\| = |a| \|x\| \quad \forall x \in V, a \in \mathbb{R}$

iii)  $\|x+y\| \leq \|x\| + \|y\| \quad \forall x, y \in V$

triangle inequality

Proof:

$$\begin{aligned}\|x+y\|^2 &= \langle x+y, x+y \rangle \\&= \langle x, x \rangle + 2 \langle x, y \rangle + \langle y, y \rangle \\&= \|x\|^2 + 2 \langle x, y \rangle + \|y\|^2 \\&\leq \|x\|^2 + 2 \|x\| \cdot \|y\| + \|y\|^2 \\&= (\|x\| + \|y\|)^2\end{aligned}$$

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Rmk: Norm is a "metric/distance" on  $V$ .