## Lecture 17 Inner Product Space

Let  $V = \mathbb{R}^2$  and  $P = (x_1, x_2)$  and  $Q = (y_1, y_2)$  be two vectors in V. The dot product of P and Q is defined as  $(x_1, x_2) \cdot (y_1, y_2) = x_1y_1 + x_2y_2$ . Then the length of P,  $||P|| = \sqrt{(x_1, x_2) \cdot (x_1, x_2)}$ , distance between P and Q is  $d(p,q) = \sqrt{(x_1 - y_1)^2 + (x_2 - y_2)^2} = \sqrt{(x_1 - y_1, x_2 - y_2) \cdot (x_1 - y_1, x_2 - y_2)}$  and the angle  $(\theta)$  between P and Q is defined as  $\cos\theta = \frac{P \cdot Q}{||P|||Q||}$ .

Observe that the above dot product satisfies the following properties:

- 1.  $(x \cdot x) \ge 0$  and  $(x \cdot x) = 0$  if and only if x = 0;
- 2.  $(x \cdot y) = (y \cdot x), \forall x, y \in \mathbb{R}^n;$
- 3.  $((\alpha x) \cdot y) = \alpha(x \cdot y), \forall \alpha \in \mathbb{R};$

4. 
$$((x+y) \cdot z) = (x \cdot z) + (y \cdot z)$$

In an arbitrary vector space, we define a function which satisfies the above four conditions, we call this function **inner product**, with the help of this function we can define the geometric concepts such as length of a vector, distance between two vectors and angle between the vectors.

**Definition 1.** Let V be a vector space over  $\mathbb{F}$ . A function  $\langle , \rangle : V \times V \longrightarrow \mathbb{F}$  is called an inner product on V if it satisfies the following properties.

1.  $\langle x, x \rangle \ge 0 \ \forall x \in V \text{ and } \langle x, x \rangle = 0 \text{ if and only if } x = 0;$ 2.  $\langle x, y \rangle = \overline{\langle y, x \rangle}, \ \forall x, y \in V;$ 3.  $\langle \alpha x + \beta y, z \rangle = \alpha \langle x, z \rangle + \beta \langle y, z \rangle, \ \forall \alpha \in \mathbb{F} \text{ and } \forall x, y, z \in V.$ 

A vector space  $V(\mathbb{F})$  together with an inner product  $\langle , \rangle$  is called an inner product space and denoted by  $(V, \langle , \rangle)$ .

Example 2. 1. Let  $V = \mathbb{R}^n$  over  $\mathbb{R}$  with  $\langle x, y \rangle = x \cdot y$ , that is,  $\langle (x_1, x_2, \dots, x_n), (y_1, y_2, \dots, y_n) = x_1y_1 + x_2y_2 + \dots + x_ny_n$ . 2. Let  $V = \mathbb{C}^n$  over  $\mathbb{C}$ . Define  $\langle (x_1, x_2, \dots, x_n), (y_1, y_2, \dots, y_n) = x_1\overline{y_1} + x_2\overline{y_2} + \dots + x_n\overline{y_n}$ . 3. Let  $V = \mathbb{R}^2$ ,  $\mathbb{F} = \mathbb{R}$  and  $A = \begin{bmatrix} a & b \\ b & c \end{bmatrix}$  such that a, c > 0 and  $ac - b^2 > 0$ . Define  $\langle x, y \rangle = y^T A x$ . 4. Let V = C[a, b],  $\mathbb{F} = \mathbb{R}$ . Define  $\langle f(x), g(x) \rangle = \int_a^b f(x)\overline{g(x)}dx$ . 5. Let  $V = M_n(\mathbb{R})$ ,  $\mathbb{F} = \mathbb{R}$ . Then for  $A, B \in V$ , define  $\langle A, B \rangle = trace(AB^T)$ . **Proposition 3.** Every finite dimensional vector space is an inner product space.

*Proof.* Let  $B = \{v_1, \ldots, v_n\}$  be an ordered basis of  $V(\mathbb{F})$ . Then for  $u, v \in V$ , define  $\langle u, v \rangle = \alpha_1 \overline{\beta_1} + \ldots + \alpha_n \overline{\beta_n}$ , where  $(\alpha_1, \ldots, \alpha_n)^T = [u]_B$  and  $(\beta_1, \ldots, \beta_n)^T = [v]_B$ .

Note that  $\langle v, v \rangle > 0$  for non-zero  $v \in V$ . This leads us to define the concept of length of a vector in an inner product space.

**Definition 4.** The length of a vector v (norm of a vector v) is defined as  $||v|| = \sqrt{\langle v, v \rangle}$ .

**Theorem 5** (Cauchy-Schwartz Inequality). Let V be an inner product space. Then  $|\langle v, u \rangle| \leq ||v|| ||u||, \forall u, v \in V$ . The equality holds if and only if the set  $\{u, v\}$  is linearly dependent.

**Proof:** Clearly, the result is true for u = 0. Suppose  $u \neq 0$ . Let  $w = v - \frac{\langle v, u \rangle}{||u||^2} u$ . Then  $w \in V$ . By the property  $\langle w, w \rangle \ge 0$ , we get  $||v||^2 - \frac{|\langle v, u \rangle|^2}{||u||^2} \ge 0$ . Therefore,  $|\langle v, u \rangle| \le ||v|| ||u||$ .

For equality, if u = 0 then the set  $\{0, v\}$  is L.D.. If  $u \neq 0$  then from the above we have  $v = \frac{\langle v, u \rangle}{||u||^2} u$ . Conversely, let u, v are L.D. then  $u = \alpha v$  for some  $\alpha \in \mathbb{F}$ . Then  $|\langle u, v \rangle| = |\langle \alpha v, v \rangle| = |\alpha|||v||^2 = ||u|| ||v||$ .

**Proposition 6.** Let  $(V(\mathbb{F}), \langle , \rangle)$  be an inner product space. Then 1.  $||u+v|| \le ||v|| + ||u||, \forall u, v \in V.$  (Triangle inequality) 2.  $||u+v||^2 + ||u-v||^2 = 2(||v|| + ||u||)^2 \forall u, v \in V.$  (Parallelogram law)

**Proof:** By definition,  $||u+v||^2 = \langle u+v, u+v \rangle = ||v||^2 + \langle u, v \rangle + \overline{\langle u, v \rangle} + ||u||^2 = ||v||^2 + 2Re(\langle u, v \rangle) + ||u||^2 \leq ||v||^2 + 2|\langle u, v \rangle| + ||u||^2 = (||u|| + ||v||)^2$ . Prove the second statement yourself.

**Definition 7.** Let u and v be vectors in an inner product space  $(V, \langle , \rangle)$ . Then u and v are **orthogonal** if  $\langle u, v \rangle = 0$ . A set S of an inner product space is called an **orthogonal set** of vectors if  $\langle u, v \rangle = 0$  for all  $u, v \in S$  and  $u \neq v$ . An **orthonormal set** is an orthogonal set S with the additional property that ||u|| = 1 for every  $u \in S$ .

**Proposition 8.** An orthogonal set of non-zero vectors is linearly independent.

**Proof:** Let S be an orthogonal set (finite or infinite) of non-zero vectors in a given inner product space. Suppose  $v_I, v_2, \ldots, v_m$  are distinct vectors in S and take  $w = \alpha_1 v_1 + \cdots + \alpha_m v_m$ . Then  $\langle w, v_i \rangle = \langle \alpha_1 v_1 + \cdots + \alpha_m v_m, v_i \rangle = \alpha_1 \langle v_1, v_i \rangle + \alpha_2 \langle v_2, v_i \rangle + \cdots + \alpha_m \langle v_m, v_i \rangle = \alpha_i \langle v_i, v_i \rangle$ . Note that  $v_i \neq 0$  so that  $\langle v_i, v_i \rangle \neq 0$ . If w = 0, then  $\alpha_i = 0$  for each *i*. Therefore, S is linearly independent.

## Gram-Schmidt orthogonalization process

**Theorem 9.** Let  $(V, \langle , \rangle)$  be an inner product space and  $S = \{v_1, v_2, \ldots, v_n\}$  be a linearly independent set of vectors in V. Then we get an orthogonal set  $\{w_1, w_2, \ldots, w_n\}$  in V such that

$$L(\{v_1, v_2, \dots, v_n\}) = L(\{w_1, w_2, \dots, w_n\}).$$

Proof.  $w_1 = v_1$ , then  $L(\{w_1\}) = L(\{v_1\})$ ;  $w_2 = v_2 - \frac{\angle v_2, w_1}{\langle w_1, w_1 \rangle} w_1$ , then  $\langle w_2, w_1 \rangle = 0$  with  $L(\{w_1, w_2\}) = L(\{v_1, v_2\})$ ;  $w_3 = v_3 - \frac{\langle v_3, w_2 \rangle}{\langle w_2, w_2 \rangle} w_2 - \frac{\langle v_3, w_1 \rangle}{\langle w_1, w_1 \rangle} w_1$ , then  $\langle w_3, w_1 \rangle = 0$ , and  $\langle w_3, w_2 \rangle = 0$  with  $L(\{w_1, w_2, w_3\}) = L(\{v_1, v_2, v_3\})$ ; Inductively,

 $w_n = v_n - \frac{\langle v_n, w_{n-1} \rangle}{\langle w_{n-1}, w_{n-1} \rangle} w_{n-1} - \frac{\langle v_n, w_{n-2} \rangle}{\langle w_{n-2}, w_{n-2} \rangle} w_{n-2} - \dots - \frac{\langle v_n, w_1 \rangle}{\langle w_1, w_1 \rangle} w_1, \text{ then } \langle w_n, w_i \rangle = 0 \text{ for } i \neq n \text{ with } L(\{v_1, v_2, \dots, v_n\}) = L(\{w_1, w_2, \dots, w_n\}).$ 

**Remark 10.** 1. The method by means of which orthogonal vectors  $w_1, \ldots, w_n$  are obtained is known as the **Gram-Schmidt orthogonalization process**.

2. Every finite-dimensional inner product space has an orthonormal basis.

3.Let  $\{v_1, \ldots, v_n\}$  be an orthonormal basis for an inner product space V. Then for any  $w \in V$ ,  $w = \langle w, v_1 \rangle v_1 + \cdots + \langle w, v_n \rangle v_n$ .

**Example 11.** Find an orthogonal basis of  $\mathbb{R}^2$  with the inner product given by  $\langle (x_1, y_1), (x_2, y_2) \rangle = x_1 x_2 + 2x_1 y_2 + 2x_2 y_1 + 5y_1 y_2$ .

**Solution:** We know that  $\{e_1, e_2\}$  is a basis of  $\mathbb{R}^2$ . Since  $\langle e_1, e_2 \rangle = 2 \neq 0$ , the standard basis is not an orthogonal basis under the defined inner product. To get an orthogonal basis we use Gram-Schmidt process:  $w_1 = e_1$  and  $w_2 = e_2 - \langle e_2, e_1 \rangle \frac{e_1}{||e_1||^2}$  and  $||e_1||^2 = \langle e_1, e_1 \rangle = 1$  so that  $w_2 = e_2 - 2e_1$ . Thus  $\{e_1, e_2 - 2e_1\}$  is an orthogonal basis.