#### Inner Productt

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#### **Dot Product**

A **dot product** is the numerical product of the lengths of two vectors, multiplied by the cosine of the angle between them, that is  $\vec{a} \cdot \vec{b} = |\vec{a}| |\vec{b}| \cos \theta$ , where  $\theta$  represents the angle between the two vectors.

A simply way to calculate a dot product is by multiplying the components of each vector separately and then adding these products together.

Example: 
$$\vec{a} = [4, 3], \vec{b} = [1, 2]$$
  
 $\vec{a} \cdot \vec{b} = (4 \times 3) + (3 \times 2) = 11$ 

# Inner product

The generalization of the dot product to an arbitrary vector space is called an *inner product*.

Let V be a vector space. An inner product on V is a rule that assigns to each pair v,  $w \in V$  a real number  $\langle v, w \rangle$  such that, for all u, v,  $w \in V$  and  $\alpha \in \mathbb{R}$ ,

- (1)  $\langle v, v \rangle \geq 0$ , with equality if and only if v = 0,
- (2) < v, w > = < w, v >,
- (3) < u + v, w > = < u, w > + < v, w >
- $(4) < \alpha v, w > = \alpha < v, w > .$

## Inner product: Matrix space

We get an inner product on  $M_{m\times n}$  by defining, for  $A, B \in M_{m\times n}$ ,

$$\langle \mathbf{A}, \mathbf{B} \rangle = \sum_{i=1}^{m} \sum_{j=1}^{n} a_{ij} b_{ij}$$

This inner product is identical to the dot product on  $R_{mn}$  if an  $m \times n$  matrix is viewed as an  $mn \times 1$  matrix by stacking its columns.

#### Example:

$$\langle \begin{bmatrix} 2 & -1 & 3 \\ 5 & 0 & 4 \end{bmatrix}, \begin{bmatrix} 1 & 3 & 8 \\ 0 & 1 & -2 \end{bmatrix} \rangle = (2)(1) + (-1)(3) + (3)(8) + (5)(0) + (0)(1) + (4)(-2)$$
$$= 15.$$

## Inner product: Polynomial space

The idea of inner product is applicable to all vector space, such as polynomial space (What about function space?).

For example, let  $x_1, x_2, ..., x_n$  be fixed numbers. We get an inner product on  $P_n$  by defining, for  $p, q \in P_n$ ,

$$\langle p, q \rangle = \sum_{i=1}^{n} p(x_i)q(x_i)$$
  
=  $p(x_1)q(x_1) + p(x_2)q(x_2) + \cdots + p(x_n)q(x_n)$ .

Example: If  $x_1 = -1$ ,  $x_2 = 0$ , and  $x_3 = 1$ , then for  $p = x^2$  and q = x + 1, we have

$$< p, q > = p(-1)q(-1) + p(0)q(0) + p(1)q(1)$$
  
= 2

#### Norm

Let V be an inner product space and let  $v \in V$ . The **norm** (or **length**) of v is denoted ||v|| and is defined by

$$||v|| = \sqrt{\langle v, v \rangle}$$

The *distance* between two vectors in V is the norm of their difference

$$dist(\mathbf{v}, \mathbf{w}) = \|\mathbf{v} - \mathbf{w}\|$$

# Cauchy-Schwarz Theorem

**Cauchy-Schwarz Theorem**: For all  $v, w \in V$ , w have

$$|< V, W>| \le ||V|| ||W||.$$

The theorem implies that

$$\theta = \cos^{-1} \frac{\langle \mathbf{v}, \mathbf{w} \rangle}{\|\mathbf{v}\| \|\mathbf{w}\|}$$

and also  $\langle v, w \rangle = ||v|| ||w|| \cos \theta$ .

We say that v is **orthogonal** (or **perpendicular**) to w if and only if  $\langle v, w \rangle = 0$ :

## Cauchy-Schwarz Theorem

The idea of Cauchy-Schwarz theorem can similarly be applied to polynomial space and function space.

Example: If  $x_1 = -1$ ,  $x_2 = 0$ , and  $x_3 = 1$ , then for  $p = x^2$  and q = x + 1. The angle  $\theta$  between the two functions is as follows.

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< p, q > = 2 (as shown before)

||p|| = \sqrt{< p, p > } = \sqrt{p(-1)^2 + p(0)^2 + p(1)^2 } = \sqrt{2}

||q|| = \sqrt{< q, q > } = \sqrt{q(-1)^2 + q(0)^2 + q(1)^2 } = \sqrt{5}

\theta = \cos^{-1}(< p, q > /(||p|| ||q||)) = \cos^{-1}(2/\sqrt{10})
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## Pythagorean Theorem

**Pythagorean theorem**: Let v,  $w \in V$ . If  $v \perp w$ , then  $||v + w||^2 = ||v||^2 + ||w||^2$ .

The idea of Pythagorean theorem can similarly be applied to polynomial space and function space.