

# Dot product

**Definition: Dot Product.** Let  $v, w \in \mathbb{R}^n$  with  $v = \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix}$  and  $w = \begin{bmatrix} w_1 \\ \vdots \\ w_n \end{bmatrix}$ , with  $v_n, w_n \in \mathbb{R}$ . The dot product of  $v$  and  $w$  is denoted as  $v \cdot w$  and is defined by

$$\begin{bmatrix} v_1 & \dots & v_n \end{bmatrix} \begin{bmatrix} w_1 \\ \vdots \\ w_n \end{bmatrix} = v_1 w_1 + \dots + v_n w_n$$

So, dot product in  $\mathbb{R}^n$  is a map  $: \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$

- ◆ The geometric interpretation of the dot product is that it shows **orthogonality** between vectors in  $\mathbb{R}^n$ 
  - ◆ When two vectors are orthogonal, the dot product will be zero
  - ◆ When they are aligned, it will be equal to the product of their magnitudes
- ◆ The computation of the dot product is very simple
  - ◆ simply line up the components, multiply, and sum
- ◆ Length relates to dot product in the following way (note that  $\langle, \rangle$  is another notation for the dot product)
  - ◆  $v \cdot v = \langle v, v \rangle = ||v||^2$ . We know that  $||v||$  is the length of  $v$

## Bilinearity

**Definition: Bilinear map.** A map of  $f : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$  is called bilinear if  $f$  is **linear map** in one component when the other component is fixed.

$$f : (v, w) \mapsto f(v, w)$$

$$f : (v_0, \cdot) \mapsto f(v_0, \cdot)$$

$$f(v_1, \lambda w_1 + w_2) = \lambda f(v_1, w_1) + f(v_1, w_2)$$

- ◆ Essentially, this is saying that the map acts linear for one component at a time
  - ◆ Meaning that the map is consistent under scalar multiplication and addition when one of the values is held constant

Example: Bilinearity for the dot product in

$$\langle, \rangle : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R} \text{ and } v_0 \in \mathbb{R}^n$$

Then,  $\langle v_0, \cdot \rangle : \mathbb{R}^n \rightarrow \mathbb{R}$  is linear (as the first element is fixed to  $v_0$ )  
 $\langle \cdot, v_0 \rangle : \mathbb{R}^n \rightarrow \mathbb{R}$  is also linear.

## Inner product


Example:  $v, w \in \mathbb{R}^n$

$$\begin{bmatrix} v_1 & v_2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} w_1 \\ w_2 \end{bmatrix} \\ = v_1 w_1 + v_2 w_2$$

Say instead of  $I$ , we have

$$\begin{bmatrix} v_1 & v_2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} w_1 \\ w_2 \end{bmatrix}$$

This would be similar to dot product, but a bit scaled in one of the axes

- ◆ This introduces the idea of the *inner product*, a concept that is introduced in class but not fully developed
  - ◆ [From Wikipedia](#) : Inner products allow formal definitions of intuitive geometric notions, such as lengths, angles, and orthogonality (zero inner product) of vectors

Remark: Inner product in  $\mathbb{R}^n$ . All dot products are a class of inner products. All inner products are bilinear, but not all bilinear maps are inner products.

Remark: There are a class of matrices called positive definite matrices. If  $A \in M_2(\mathbb{R})$  is positive definite, then  $\begin{bmatrix} v_1 & v_2 \end{bmatrix} A \begin{bmatrix} w_1 \\ w_2 \end{bmatrix}$  defines an inner product and after a choice of basis, all inner products are of this above form.

Positive definite means that for  $v^T A v \geq 0$  and  $v^T A v = 0$  only if  $v = \vec{0} \in \mathbb{R}^n$ .