

# Inner Product Space

## Linear Algebra

Department of Computer Engineering

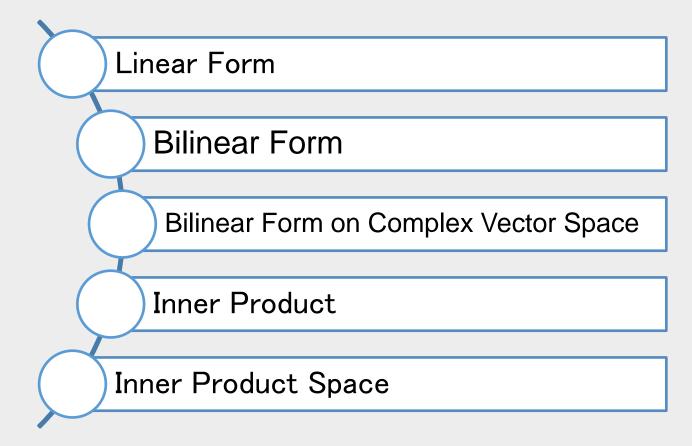
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# Overview





# Linear Form

#### What are Linear Functions?



- $\Box$   $f: \mathbb{R}^n \to \mathbb{R}$  means that f is a function that maps real n-vectors to real numbers
- $\Box$  f(x) is the value of function f at x (x is referred to as the argument of the function).

#### **Definition**

A function  $f: \mathbb{R}^n \to \mathbb{R}$  is linear if it satisfies the following two properties:

- $\square$  Additivity: For any n-vector x and y, f(x+y)=f(x)+f(y)
- $\square$  Homogeneity: For any n-vector x and any scalar  $\alpha \in R$ :  $f(\alpha x) = \alpha f(x)$

# Superposition property:



#### Definition

Superposition property:

$$f(\alpha x + \beta y) = \alpha f(x) + \beta f(y)$$

#### Note

☐ A function that satisfies the superposition property is called linear

# Homogeneity and Additivity



#### **Definition**

☐ Additivity:

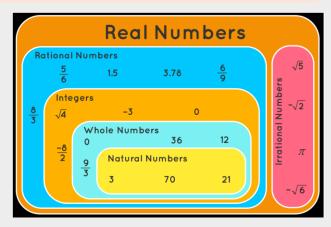
For any *n*-vector x and y, f(x + y) = f(x) + f(y)

■ Homogeneity:

For any *n*-vector *x* and any scalar  $\alpha \in R$ :  $f(\alpha x) = \alpha f(x)$ 

#### Counterexample:

$$f(a+\sqrt{5}b) \rightarrow a+b+\sqrt{5}b$$



## What are Linear Functions?



☐ If a function f is linear, superposition extends to linear combinations of any number of vectors:

$$f(\alpha_1 x_1 + \dots + \alpha_k x_k) = \alpha_1 f(x_1) + \dots + \alpha_k f(x_k)$$

# Inner product is Linear Function?



#### Theorem

A function defined as the inner product of its argument with some fixed vector is linear.

$$f(x) = a^T x = a_1 x_1 + a_2 x_2 + \dots + a_n x_n$$

# What are Linear Functions?



#### Theorem

If a function is linear, then it can be expressed as the inner product of its argument with some fixed vector.

## What are Linear Functions?



#### Theorem

The representation of a linear function f as  $f(x) = a^T x$  is unique, which means that there is only one vector a for which  $f(x) = a^T x$  holds for all x.

# Linear Form Examples



# Example

- Is average a linear function?
- Is maximum a linear function?

# Bilinear Form

# Bilinear Form over a real vector space



#### **Definition**

Suppose V and W are vector spaces over the same field  $\mathbb{F}$ . Then a function  $f: V \times W \to \mathbb{F}$  is called a bilinear form if it satisfies the following properties:

- a) It is linear in its first argument:
  - i.  $f(\mathbf{v_1} + \mathbf{v_2}, \mathbf{w}) = f(\mathbf{v_1}, \mathbf{w}) + f(\mathbf{v_2}, \mathbf{w})$  and
  - ii.  $f(c\mathbf{v_1}, \mathbf{w}) = cf(\mathbf{v_1}, \mathbf{w})$  for all  $c \in \mathbb{F}, \mathbf{v_1}, \mathbf{v_2} \in V$ , and  $\mathbf{w} \in W$ .
- b) It is linear in its second argument:
  - i.  $f(\mathbf{v}, \mathbf{w_1} + \mathbf{w_2}) = f(\mathbf{v}, \mathbf{w_1}) + f(\mathbf{v}, \mathbf{w_2})$  and
  - ii.  $f(\mathbf{v}, c\mathbf{w_1}) = cf(\mathbf{v}, \mathbf{w_1})$  for all  $c \in \mathbb{F}, \mathbf{v} \in V$ , and  $\mathbf{w_1}, \mathbf{w_2} \in W$ .

#### Bilinear Form



#### Note

Let V be a vector space over a field  $\mathbb{F}$ . Then the **dual** of V, denoted by  $V^*$ , is the vector space consisting of all linear forms on V.

# Example

Let V be a vector space over a field  $\mathbb{F}$ . Show that the function  $g: V^* \times V \to \mathbb{F}$  defined by

$$g(f, \mathbf{v}) = f(\mathbf{v})$$
 for all  $f \in V^*, \mathbf{v} \in V$ 

is a bilinear form.

#### Positive Definite Bilinear Form



#### **Definition**

A bilinear form function  $f: V \times V \to \mathbb{F}$  over a real vector space V is called positive definite if for all  $v \in V$ ,  $v \neq 0$ :

## Example

Which one is a positive definite bilinear form?

$$\Box f(x,y) = x_1y_1 + 2x_1y_2 + 2x_2y_1 + 3x_2y_2$$

# Symmetric Bilinear Form



#### **Definition**

A bilinear form function  $f: V \times V \to \mathbb{F}$  over a real vector space V is called symmetric if for all  $v, w \in V$ :

$$f(v,w) = f(w,v)$$

# Bilinear Form arises from a matrix



#### Theorem

Every **bilinear form** function  $f: V \times V \to \mathbb{F}$  over a real vector space V arises from a matrix for all  $v, w \in V$ :

$$f(v, w) = v^T A w$$

#### **Associated Matrices**



#### **Definition**

If V is a finite-dimensional vector space,  $B = \{b_1, ..., b_n\}$  is a basis of V, and  $f: V \times V \to \mathbb{F}$  be a **bilinear form** function the associated matrix A of f with respect to B is the matrix  $[f]_B \in \mathbb{F}^{n \times n}$  whose (i, j)-entry is the value  $f(b_i, b_i)$ .

$$f(v, w) = v^T A w = v^T [f]_B w$$

$$[f]_{\mathcal{B}} = \begin{pmatrix} f(b_1, b_1) & \dots & f(b_1, b_n) \\ \vdots & & \vdots \\ f(b_n, b_1) & \dots & f(b_n, b_n) \end{pmatrix}$$

#### **Associated Matrices**



#### Note

The associated matrix changes if we use a different basis.

## Example

For the bilinear form  $f \begin{pmatrix} a \\ b \end{pmatrix}, \begin{bmatrix} c \\ d \end{pmatrix} = 2ac + 4ad - bc$  on  $\mathbb{F}^2$ , find  $[f]_B$  for basis  $B = \{\begin{bmatrix} 2 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ 4 \end{bmatrix}\}$  and  $[f]_P$  for basis  $P = \{\begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix}\}$ 

# Bilinear Form Over Complex Vector Space

# Bilinear Form over a real vector space



#### **Definition**

Suppose V and W are vector spaces over the same field  $\mathbb{C}$ . Then a function  $f: V \times W \to \mathbb{C}$  is called a bilinear form if it satisfies the following properties:

- a) It is linear in its first argument:
  - i.  $f(\mathbf{v_1} + \mathbf{v_2}, \mathbf{w}) = f(\mathbf{v_1}, \mathbf{w}) + f(\mathbf{v_2}, \mathbf{w})$  and
  - ii.  $f(\lambda \mathbf{v_1}, \mathbf{w}) = \lambda f(\mathbf{v_1}, \mathbf{w})$  for all  $\lambda \in \mathbb{C}, \mathbf{v_1}, \mathbf{v_2} \in V$ , and  $\mathbf{w} \in W$ .
- b) It is conjugate linear in its second argument:
  - i.  $f(\mathbf{v}, \mathbf{w_1} + \mathbf{w_2}) = f(\mathbf{v}, \mathbf{w_1}) + f(\mathbf{v}, \mathbf{w_2})$  and
  - ii.  $f(\mathbf{v}, \lambda \mathbf{w}_1) = \overline{\lambda} f(\mathbf{v}, \mathbf{w}_1)$  for all  $\lambda \in \mathbb{C}, \mathbf{v} \in V$ , and  $\mathbf{w}_1, \mathbf{w}_2 \in W$ .

# Bilinear forms (real vs complex vector space)



Bilinear forms on $\mathbb{R}^n$	Bilinear forms on $\mathbb{C}^n$
<u>Linear</u> in the first variable	<u>Linear</u> in the first variable
<u>Linear</u> in the second variable	Conjugate linear in the second variable

# Inner product

# Inner product over real vector space



#### Definition

An inner product is a positive-definite symmetric bilinear form.

- □ An inner product on V is a function  $\langle , \rangle : V \times V \to \mathbb{R}$  such that  $v, w \in V, c \in \mathbb{R}$ :
  - 1.  $\langle v, v \rangle = 0$  if and only if v = 0.
  - 2.  $\langle w, v \rangle = \langle v, w \rangle$ .
  - 3.  $\langle u + v, w \rangle = \langle u, w \rangle + \langle v, w \rangle$  for all  $u, v, w \in V$ .
  - 4.  $\langle cw, u \rangle = c \langle w, u \rangle$  for all  $u, w \in V$  and  $c \in \mathbb{R}$ .
  - 5.  $\langle v, v \rangle \ge 0$  for all  $v \in V$ .

#### Inner Product



#### Why for bilinear form I wrote just two properties instead of four properties?

☐ Using properties (2) and (4) and again (2)

$$\langle w, cu \rangle = \langle cu, w \rangle = c \langle u, w \rangle = c \langle w, u \rangle$$

Using properties (2), (3) and again (2)  $\langle w, u + v \rangle = \langle u + v, w \rangle = \langle u, w \rangle + \langle v, w \rangle = \langle w, u \rangle + \langle w, v \rangle$ 

- 1.  $\langle v, v \rangle = 0$  if and only if v = 0.
- 2.  $\langle w, v \rangle = \langle v, w \rangle$ .
- 3.  $\langle u + v, w \rangle = \langle u, w \rangle + \langle v, w \rangle$  for all  $u, v, w \in V$ .
- **4**.  $\langle cw, u \rangle = c \langle w, u \rangle$  for all  $u, w \in V$  and  $c \in \mathbb{R}$ .
- 5.  $\langle v, v \rangle \ge 0$  for all  $v \in V$ .

# Inner Products



## Note

# General Inner product



#### Definition

Suppose that  $\mathbb{F} = \mathbb{R}$  or  $\mathbb{F} = \mathbb{C}$  and that V is a vector space over  $\mathbb{F}$ . Then an inner product on V is a function  $\langle \cdot, \cdot \rangle : V \times V \to \mathbb{F}$  such that the following three properties hold for all  $c \in \mathbb{F}$  and all  $\mathbf{v}, \mathbf{w}, \mathbf{x} \in V$ :

- a)  $\langle \mathbf{v}, \mathbf{w} \rangle = \overline{\langle \mathbf{w}, \mathbf{v} \rangle}$  (conjugate symmetry)
- b)  $\langle v+cx,w\rangle = \langle v,w\rangle + c\langle x,w\rangle$  (linearity)
- c)  $\langle \mathbf{v}, \mathbf{v} \rangle \geq 0$ , with equality if and only if  $\mathbf{v} = \mathbf{0}$ . (pos. definiteness)

# Inner Products for vectors



#### Note

 $\square$  The standard inner product between vectors is:  $(x, y \in \mathbb{R}^n)$ 

$$\langle x, y \rangle = x^T y = \sum x_i y_i$$

 $\square$  Euclidean inner product: The function  $\langle \cdot, \cdot \rangle : \mathbb{C}^n \times \mathbb{C}^n \to \mathbb{C}$  defined by

$$\langle v, w \rangle = v w^* = \sum_{i=1}^n v_i \overline{w_i}$$

for all  $v, w \in \mathbb{C}^n$  is an inner product on  $\mathbb{C}^n$ .

# Inner Product for matrices



#### Note

 $\square$  The standard inner product between two matrices is:  $(X,Y \in \mathbb{R}^{m \times n})$ 

$$\langle X, Y \rangle = trace(X^TY) = \sum_{i} \sum_{j} X_{ij} Y_{ij}$$

# Example

$$U = \begin{bmatrix} 1 & 2 \\ 1 & 0 \end{bmatrix}, \qquad V = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}$$

# Inner Product for functions



#### Note

Let a < b be real numbers and let C[a, b] be the vector space of continuous functions on the real interval [a, b]. The function  $\langle \cdot, \cdot \rangle : C[a, b] \times C[a, b] \to \mathbb{R}$  defined by

$$\langle f, g \rangle = \int_a^b f(x)g(x) dx$$

for all 
$$f, g \in C[a, b]$$

is and inner product on C[a, b].

# Inner Product for polynomials



#### Note

 $\square$  For p(x) and q(x) with at most degree n:

$$\langle p(x), q(x) \rangle = p(0)q(0) + p(1)q(1) + \dots + p(n)q(n)$$

- $\square$  For p(x) and q(x):  $\langle p(x), q(x) \rangle = p(0)q(0) + \int_{-1}^{1} p'q'$
- $\square$  For p(x) and q(x):  $\langle p(x), q(x) \rangle = \int_0^\infty p(x)q(x)e^{-x}dx$

# Inner product space



#### **Definition**

An inner product space is a finite-dimensional  $\underline{real}$  or complex vector space V along with an inner product on V.

**Euclidean Space Unitary Space** 

#### References



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