# Symmetric, Positive Definite Matrices

#### Definition

A symmetric matrix  $A \in \mathbb{R}^{n \times n}$  is called **positive definite**, if

$$\mathbf{x}^T A \mathbf{x} > 0, \quad \mathbf{x} \in \mathbb{R}^n, \mathbf{x} \neq \mathbf{0}$$

The matrix A is called positive semi-definite, if

$$\mathbf{x}^T A \mathbf{x} \ge 0, \quad \mathbf{x} \in \mathbb{R}^n.$$

#### Example

Which of the following matrices is positive definite?

$$A = \begin{bmatrix} 4 & -4 \\ -4 & 5 \end{bmatrix}, \quad B = \begin{bmatrix} 4 & -4 \\ -4 & 2 \end{bmatrix}$$



#### Proposition

Let V be an n-dimensional vector space with an inner product  $\langle \cdot, \cdot \rangle$  and a basis  $B = (\mathbf{b}_1, \dots, \mathbf{b}_n)$ . Prove that

$$\langle \mathbf{x}, \mathbf{y} \rangle = [\mathbf{x}]_B^T A[\mathbf{y}]_B,$$

where  $A_{ij} = \langle \mathbf{b}_i, \mathbf{b}_j \rangle$ . Conclude that the matrix A is positive definite.

#### Theorem

For a real-valued, finite-dimensional vector space V and a basis B of V it holds that  $\langle \cdot, \cdot \rangle : V \times V \to \mathbb{R}$  is an inner product if and only if there exists a symmetric, positive definite matrix  $A \in \mathbb{R}^{n \times n}$  with

$$\langle \mathbf{x}, \mathbf{y} \rangle = [\mathbf{x}]_B^T A[\mathbf{y}]_B,$$



## Proposition

Let  $A = (a_{ij})$  be a positive definite matrix, then

- $\ker(A) = \{0\}$
- $a_{ii} > 0$

## Lengths and Distances

Any inner product induces a norm

$$||x|| := \sqrt{\langle \mathbf{x}, \mathbf{x} \rangle}$$

However, not every norm is induced by an inner product (e.g. Manhattan norm or  $\ell_1$  norm).

#### Example

Compute the norm of the vector  $[2 \ 2]^T$  w.r.t. the following inner products:

a) 
$$\langle \mathbf{x}, \mathbf{y} \rangle = \mathbf{x}^T \mathbf{y}$$

b) 
$$\langle \mathbf{x}, \mathbf{y} \rangle = \mathbf{x}^T \begin{bmatrix} 3 & -1 \\ -1 & 2 \end{bmatrix} \mathbf{y}$$



### Distance and metric

Definition

A function  $d: V \times V \to \mathbb{R}$  is metric. if

- d is positive definite, i.e.,  $d(\mathbf{x}, \mathbf{y}) \ge 0$  for all  $\mathbf{x}, \mathbf{y} \in V$  and  $d(\mathbf{x}, \mathbf{y}) = 0 \Leftrightarrow \mathbf{x} = \mathbf{y}$
- d is symmetric, i.e.,  $d(\mathbf{x}, \mathbf{y}) = d(\mathbf{y}, \mathbf{x})$  for all  $\mathbf{x}, \mathbf{y} \in V$
- Triangular inequality:  $d(\mathbf{x}, \mathbf{z}) \leq d(\mathbf{x}, \mathbf{y}) + d(\mathbf{y}, \mathbf{z})$ .

### **Proposition**

Every norm (inner product) induces a metric:

$$d(\mathbf{x}, \mathbf{y}) := \|\mathbf{x} - \mathbf{y}\| = \sqrt{\langle \mathbf{x} - \mathbf{y}, \mathbf{x} - \mathbf{y} \rangle}$$



# Angles and Orthogonality

By Cauchy-Schwarz inequality

$$-1 \le \frac{\langle \mathbf{x}, \mathbf{y} \rangle}{\|\mathbf{x}\| \|\mathbf{y}\|} \le 1.$$

Therefore, there exists a unique  $w \in [0,\pi]$  with

$$\cos w = \frac{\langle \mathbf{x}, \mathbf{y} \rangle}{\|\mathbf{x}\| \|\mathbf{y}\|}$$

The number w is the **angle** between the vectors  $\mathbf{x}$  and  $\mathbf{y}$ .

#### Example

Find the angle between  ${\bf x}$  and  $\pm {\bf x}$ .



## Example

Find the angle between  $\mathbf{x} = [1, \ 1]^T$  and  $\mathbf{y} = [1, -1]^T$  w.r.t. the inner products given by

a) 
$$\langle \mathbf{x}, \mathbf{y} \rangle = \mathbf{x}^T \mathbf{y}$$

b) 
$$\langle \mathbf{x}, \mathbf{y} \rangle = \mathbf{x}^T \begin{bmatrix} 1 & -1 \\ -1 & 2 \end{bmatrix} \mathbf{y}$$