

# Basic Notions

## Vector Space

Set  $\mathcal{D}$  with operations  $+$  and  $\alpha \cdot$ ,  $\alpha \in \mathbb{R}$ . Axioms:

1.  $x + (y + z) = (x + y) + z$
2.  $x + y = y + x$
3.  $\exists 0 \in \mathcal{D} : 0 + x = x$
4.  $\forall x \exists (-x) : x + (-x) = 0$
5.  $\alpha(x + y) = \alpha x + \alpha y$
6.  $(\alpha + \beta)x = \alpha x + \beta x$
7.  $\alpha(\beta x) = (\alpha\beta)x$
8.  $1x = x$

## Eucledian (Inner Product) Space

Vector space  $\mathcal{D}$  with inner product  $\langle x, y \rangle \in \mathbb{R}$ ,  $x, y \in \mathcal{D}$ .

1.  $\langle x, y \rangle = \langle y, x \rangle$
2.  $\langle x + y, z \rangle = \langle x, z \rangle + \langle y, z \rangle$
3.  $\langle \alpha x, y \rangle = \alpha \langle x, y \rangle$
4.  $\langle x, x \rangle \geq 0, \quad \langle x, x \rangle = 0 \iff x = 0$

$\|x\| = \sqrt{\langle x, x \rangle}$  - norm of  $x$ .

**Orthonormal basis (ONB)**  $e_1, \dots, e_n \in \mathcal{D}$ :

$$\langle e_i, e_j \rangle = \delta_{ij} = \begin{cases} 1, & i = j \\ 0, & i \neq j \end{cases}$$

**Coordinates** in the ONB  $e_1, \dots, e_n \in \mathcal{D}$ :

$$x_1, \dots, x_n \in \mathbb{R} : \quad x = \sum_{i=1}^n x_i e_i$$

$$x_i = \langle x, e_i \rangle, \quad \bar{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \text{ - column of coordinates.}$$

# Linear Transformation

$A : \mathcal{D} \rightarrow \mathcal{R}$  - Linear transformation (mapping, map) if  
 $\forall x, y \in \mathcal{D}$  and  $\forall \alpha \in \mathbb{R}$

1.  $A(x + y) = Ax + Ay$

2.  $A(\alpha x) = \alpha Ax$

If  $A : \mathcal{D} \rightarrow \mathcal{D}$  - linear operator.

**Matrix of a linear map**  $A : \mathcal{D} \rightarrow \mathcal{R}$  in ONBs

$$\varphi_1, \dots, \varphi_m \in \mathcal{D}, \quad e_1, \dots, e_n \in \mathcal{R}$$

$$\bar{A} = \begin{bmatrix} a_{11} & \cdots & a_{1m} \\ \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nm} \end{bmatrix} \quad a_{ij} = \langle e_i, A\varphi_j \rangle$$

$$y = Ax \iff \bar{y} = \bar{A}\bar{x}$$

**Adjoint linear map.** If  $A : \mathcal{D} \rightarrow \mathcal{R}$  then  $A^* : \mathcal{R} \rightarrow \mathcal{D}$  - adjoint to  $A$  if

$$\forall x \in \mathcal{D} \quad \forall y \in \mathcal{R} \quad \langle Ax, y \rangle = \langle x, A^*y \rangle$$

$$\overline{A^*} = \bar{A}^T$$

$$(AB)^* = B^*A^*$$

$S : \mathcal{R} \rightarrow \mathcal{R}$  is **self-adjoint operator** if  $S^* = S$ .

Its matrix is symmetric:  $\bar{S}^T = \bar{S}$ .

Let  $S$  - self-adjoint operator in  $\mathcal{R}$ . If  $\forall x \in \mathcal{R} : x \neq 0$

$\langle Sx, x \rangle \geq 0$  – **nonnegative definite**,  $S \geq 0$ .

$\langle Sx, x \rangle > 0$  – **positive definite**,  $S > 0$ .

$$S \geq 0 \implies s_{ii} \geq 0, \quad S > 0 \implies s_{ii} > 0.$$

$$\forall A : \mathcal{D} \rightarrow \mathcal{R} \quad AA^* \geq 0, \quad A^*A \geq 0.$$

Let  $S, T : \mathcal{R} \rightarrow \mathcal{R}$  - self-adjoint. Define:

$$S \geq T \quad \text{if} \quad S - T \geq 0, \quad S > T \quad \text{if} \quad S - T > 0.$$

**Trace of  $S : \mathcal{R} \rightarrow \mathcal{R}$**

$$\operatorname{tr} S = \sum_{i=1}^n s_{ii} = \sum_{i=1}^n \langle e_i, Se_i \rangle.$$

- Linear:  $\operatorname{tr} (\alpha S + \beta T) = \alpha \operatorname{tr} S + \beta \operatorname{tr} T$ .

- Monotone:  $S \geq T \implies \operatorname{tr} S \geq \operatorname{tr} T$ .

**Invertible operator**  $S : \mathcal{R} \rightarrow \mathcal{R}$  if

$$\exists S^{-1} : \quad S^{-1}S = I = SS^{-1}.$$

$S$  is invertible iff  $Sx = 0 \implies x = 0$ .

$S > 0 \implies S$  – is invertible.

$$(S^*)^{-1} = (S^{-1})^* \quad (ST)^{-1} = T^{-1}S^{-1}.$$

# Random Vector

$\nu$  - random vector in  $\mathcal{R}$  if its coordinates in (any) basis are random variables.

$$\nu_i = \langle \nu, e_i \rangle, \quad \begin{bmatrix} \nu_1 \\ \vdots \\ \nu_n \end{bmatrix} - \text{column of coordinates.}$$

**Mathematical Expectation** of  $\nu$ ,  $E\nu$ :

$$E \begin{bmatrix} \nu_1 \\ \vdots \\ \nu_n \end{bmatrix} = \begin{bmatrix} E\nu_1 \\ \vdots \\ E\nu_n \end{bmatrix}$$

**Linearity:**  $\nu, \mu$  - random vectors in  $\mathcal{R}$ ,  $\alpha, b \in \mathbb{R}$ ,

$$E(\alpha\nu + \beta\mu) = \alpha E\nu + \beta E\mu.$$

$$x \in \mathcal{R} \quad E \langle \nu, x \rangle = \langle E\nu, x \rangle.$$

$$A : \mathcal{R} \rightarrow \mathcal{D} \quad EAx = AEx.$$

**Independence:**  $\nu, \mu$  - rand.vec. in  $\mathcal{R}$ ,  $\alpha$  - rand.var.

$$\nu, \alpha \text{ independent} \implies E(\alpha\nu) = E\alpha E\nu.$$

$$\nu, \mu \text{ independent} \implies E \langle \nu, \mu \rangle = \langle E\nu, E\mu \rangle.$$

# Variance operator

of a random vector  $\nu \in \mathcal{R}$ .  $S = \text{Var}(\nu) : \mathcal{R} \rightarrow \mathcal{R}$

$$\forall x \in \mathcal{R} \quad Sx = E \langle \nu - E\nu, x \rangle (\nu - E\nu).$$

$$s_{ij} = \text{Cov}(\nu_i, \nu_j).$$

$$\bar{S} = \begin{bmatrix} \text{Var}(\nu_1) & \text{Cov}(\nu_1, \nu_2) & \cdots & \text{Cov}(\nu_1, \nu_n) \\ \text{Cov}(\nu_2, \nu_1) & \text{Var}(\nu_2) & \cdots & \text{Cov}(\nu_2, \nu_n) \\ \vdots & \vdots & \ddots & \vdots \\ \text{Cov}(\nu_n, \nu_1) & \text{Cov}(\nu_n, \nu_2) & \cdots & \text{Var}(\nu_n) \end{bmatrix}$$

- variance-covariance matrix of  $\begin{bmatrix} \nu_1 \\ \vdots \\ \nu_n \end{bmatrix}$ .

- $S \geq 0$  ( $\implies s_{ii} \geq 0 \implies \text{tr } S \geq 0$ ).
- If  $B : \mathcal{R} \rightarrow \mathcal{D}$   $\text{Var}(B\nu) = BSB^*$ .
- If  $E\nu = 0$   $E \|\nu\|^2 = \text{tr } S$ .