

Chapter 1

Review of matrices

1.1 Notions of vectors and matrices

In this part we give definitions and some concepts about vectors and matrices.
Let $\mathbb{R}^n, \mathbb{R}^p$ be two vector spaces of finite dimensions.

1.1.1 Vectors

Definition 1.1 : An element of the vector space $\mathbb{R}^n$ over $\mathbb{R}$ is called a vector v . It is an ordered set of elements of $\mathbb{R}$, as we denote $v = (v_1, v_2, \dots, v_n)^t$ or

$$v = \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix}$$

Remark 1.1 : Since $\mathbb{R}^n$ is a vector space on $\mathbb{R}$, it has the classic properties of a vector space: Let $x, y \in \mathbb{R}^n$ and $\alpha, \beta \in \mathbb{R}$, then

1. $(\alpha\beta)x = \alpha(\beta)x$.
2. $(\alpha + \beta)x = \alpha x + \beta x$.
3. $\alpha(x + y) = \alpha x + \alpha y$.
4. $1x = x$.

Definition 1.2

A dot product on E is an application of $E \times E$ on $\mathbb{R}$:

$$\begin{aligned} \langle \cdot, \cdot \rangle: E \times E &\longrightarrow \mathbb{R} \\ (x, y) &\longmapsto \langle x, y \rangle_E \end{aligned}$$

with the following properties:

1. $\langle x, x \rangle \geq 0$.
2. $\langle x, x \rangle = 0 \Rightarrow x = 0$.
3. $\langle x, y \rangle = \langle y, x \rangle$.
4. $\langle x, y + z \rangle = \langle x, y \rangle + \langle x, z \rangle$.
5. $\langle x, \alpha y \rangle = \alpha \langle x, y \rangle, \forall \alpha \in \mathbb{R}$.

Orthogonality of vectors and subspaces

- The angle θ between two vectors u and v given by $\cos(\theta) = \frac{u^t v}{\|u\| \|v\|}$.
- Two vectors u and v are orthogonal if $\theta = 90^\circ$ that is $u^t v = 0$.
- The symbol $\perp$ is used to denote orthogonality.

1.1.2 Matrices

Basic concepts

A collection of n vectors in $\mathbb{R}^n$ arranged in a rectangular array of n rows and m columns is called matrix.

A matrix A , there for has the form

$$\begin{pmatrix} a_{11} & a_{12} & \cdot & \cdot & \cdot & a_{1n} \\ a_{21} & a_{22} & \cdot & \cdot & \cdot & a_{2n} \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ a_{m1} & a_{m2} & \cdot & \cdot & \cdot & a_{mn} \end{pmatrix}$$

Definition 1.3 (Transpose and conjugate matrices)

Let A is matrix such that $A \in \mathbb{C}^{n \times n}$.

1. We say A^t is the transpose of A .
2. We say A^* is the conjugate of A and $A^* = \overline{(A^t)}$.

Definition 1.4 (Symmetric and Hermitian matrices)

Let A is matrix such that $A \in \mathbb{C}^{n \times n}$.

1. A is symmetric if $A = A^t$.
2. A is skew symmetric if $A = -A^t$.
3. A is Hermitian if $A^* = A$.
4. A is skew Hermitian if $A^* = -A$.
5. A is normal if $AA^* = A^*A$.

Inverse matrix

We want to determine an inverse with respect to matrix multiplication.

Definition 1.5

A matrix $A \in \mathbb{C}^{n \times n}$ is nonsingular (or invertible) if A has an inverse, that is if there is a matrix A^{-1} so that

$$AA^{-1} = I = A^{-1}A.$$

If A does not have an inverse it is singular.

Remark 1.2 The inverse is unique.

Properties of inverse matrix

Let A and B are two invertible matrices, then we have.

1. $(AB)^{-1} = B^{-1}A^{-1}$.
2. $(A^t)^{-1} = (A^{-1})^t$.
3. $(A^*)^{-1} = (A^{-1})^*$.

Definition 1.6 (Unitary and orthogonal matrices)

Let A is matrix such that $A \in \mathbb{C}^{n \times n}$.

1. A is unitary if $AA^* = A^*A = I$.
2. A is orthogonal if $AA^t = A^tA = I$.

Remarks 1.1 : Let A is matrix such that $A \in \mathbb{C}^{n \times n}$.

1. Diagonal matrices : $a_{ij} = 0$ for $j \neq i$, $A = \text{diag}(a_{11}, a_{22}, \dots, a_{nn})$.
2. Upper triangular matrices : $a_{ij} = 0$ for $i \geq j$.
3. Lower triangular matrices : $a_{ij} = 0$ for $i \leq j$
4. Upper bi-diagonal : $a_{ij} = 0$ for $j \neq i$ or $j \neq i + 1$
5. Lower bi-diagonal : $a_{ij} = 0$ for $j \neq i$ or $j \neq i - 1$
6. Tri-diagonal : $a_{ij} = 0$ for any pair i, j such that $|j - i| \geq 1$, $A = \text{tridiag}$

Trace and determinant of matrix

Definition 1.7 Let A is matrix such that $A \in \mathbb{C}^{n \times n}$.

1. The trace of matrix A is defined by $\text{tr}(A) = \sum_{i=1}^n a_{ii}$.
2. The determinant of matrix A is defined by

$$\det(A) = \sum_{\sigma \in S_n} \varepsilon_{\sigma} a_{\sigma(1)1} a_{\sigma(2)2} \dots a_{\sigma(n)n},$$

where σ_{ε} denotes the signature of the permutation.

Proposition 1.1 Let A, B be two square matrices.

1. $\text{tr}(A) = \sum_{i=1}^n \lambda_i(A)$
2. $\det(A) = \prod_{i=1}^n \lambda_i(A)$
3. $\text{tr}(AB) = \text{tr}(BA)$
4. $\det(AB) = \det(BA) = \det(A) \cdot \det(B)$

1.2 Vector norms and matrix norms

1.2.1 Vector norms

Let V an vector space over the field $\mathbb{K}$.

A Norm on V is an application $\|\cdot\| : V \rightarrow \mathbb{R}$ that verifies the following properties :

1. $\|v\| = 0 \Leftrightarrow v = 0$ and $\|v\| \geq 0$ for all $v \in V$.
2. $\|\alpha v\| = |\alpha| \|v\|$, for all $\alpha \in \mathbb{K}$ and $v \in V$.
3. $\|u + v\| \leq \|u\| + \|v\|$, for all $u, v \in V$.

Definition 1.8 Let V be a finite-dimensional vector space.

1. $\|v\|_1 = \sum_{i=1}^n |v_i| = |v_1| + |v_2| + \dots + |v_n|$.
2. $\|v\|_2 = (\sum_{i=1}^n |v_i|^2)^{\frac{1}{2}} = (|v_1|^2 + |v_2|^2 + \dots + |v_n|^2)^{\frac{1}{2}}$.
3. $\|v\|_\infty = \max_i |v_i|$.

Theorem 1.1 Let V be a finite-dimensional vector space, for all real number $p \geq 1$ the application $\|\cdot\|_p$ defined by

$$\|v\|_p = \left(\sum_{i=1}^n |v_i|^p \right)^{\frac{1}{p}}$$

is a norm.

1.2.2 Matrix norms

Let A is matrix such that $A \in \mathbb{C}^{n \times n}$, we define the following special set of norms

$$\|A\|_{pq} = \max_{x \in \mathbb{C}^n, x \neq 0} \frac{\|Ax\|_p}{\|x\|_q}$$

Definition 1.9 A matrix norm is an application $\|\cdot\| : \mathcal{A}_n \leftarrow \mathbb{R}$ that verifies the following properties

1. $\|A\| = 0 \Leftrightarrow A = 0$ and $\|A\| \geq 0$ for all $A \in \mathcal{A}_n$.
2. $\|\alpha A\| = |\alpha| \|A\|$, for all $\alpha \in \mathbb{K}$ and $A \in \mathcal{A}_n$.
3. $\|A + B\| \leq \|A\| + \|B\|$, for all $A, B \in \mathcal{A}_n$.
4. $\|AB\| \leq \|A\| \|B\|$, for all $A, B \in \mathcal{A}_n$.

Remarks 1.2

TD Series: On Matrices

Exercise 01

- Let the vectors $v_i, i = 1, 2, 3$ such that $v_1 = (1, 0, 0)^t$, $v_2 = (1, 3, 5)^t$ and $v_3 = (11, 5, 10)^t$
 - Compute the norms $\|v_i\|_1, \|v_i\|_2$ and $\|v_i\|_\infty$
 - Compute the dot product of v_1 and v_2 , v_1 and v_3 , and v_2 and v_3 .
- Consider the following matrices:

$$A = \begin{pmatrix} 5 & -4 & 2 \\ -1 & 2 & 3 \\ -2 & 1 & 0 \end{pmatrix}, \quad B = \begin{pmatrix} 3 & 6 & -1 \\ 3 & 1 & 0 \\ 2 & 4 & -7 \end{pmatrix}, \quad C = \begin{pmatrix} 1 & 7 & 3 \\ 4 & -2 & -2 \\ -2 & -1 & 1 \end{pmatrix}$$

- Compute the norms $\|A\|_1, \|A\|_2$ and $\|A\|_\infty$, and similarly for matrices B and C .
- Compute the Frobenius norm $\|I\|_F$

Exercise 02

Let A and B two matrices. Show that:

- Diagonal matrices are symmetric.
- Diagonal matrices are Hermitian.
- Diagonal matrices commute, i.e., if A and B are two diagonal matrices, then AB is diagonal and $AB = BA$.
- Let D be a diagonal matrix. If $D = (I + A)^{-1}A$ then A is diagonal.
- Let $A \in \mathbb{C}^{m \times n}$ be a matrix. Prove that $A + A^t$ is symmetric and $A + A^*$ is Hermitian.

Exercise 03

Let matrix Z such that

$$Z = \begin{pmatrix} 1+2i & 3-5i & 7 \\ 8+i & 6i & 2-i \end{pmatrix}.$$

And let $U = A + iB$ and $U^* = A - iB$ where A and B are real matrices.

- Compute Z^t , Z^* and Z^{**} .
- Show that $tr(U^*) \neq tr$

Exercise 04

Let Z be a matrix defined in the form

$$Z = \begin{pmatrix} A & B \\ C & D \end{pmatrix}$$

1. Case: Matrices A and D are not of the same size.

(a) Show that $Tr(Z) = tr(A) + tr(D)$

2. Case: Matrices A and D are square matrices.

(a) Show that, if Z is symmetric then A and D are also symmetric.

(b) Show that, if Z is diagonal then A and D are also diagonal.

(c) Show that, if Z is upper triangular then A and D are also upper triangular.