

# (ENKEMNA0302) Applied Linear Algebra

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#### **Actualities**

- ▶ I booked Room No. F07 in the Building F from 10:00 to 14:00 on May 15, 2025 for retake test.
- ► Are you interested in buildin a CubeSat?
- ► Recruit lecture will be held at UP Faculty of Engineering and Information Technology in Room B224 from 14:00 to 16:00 on Tuesday February 18, 2025.

# Requirements

- ➤ You will write tests based on the exercises of the practical courses. You can use everything during the test
- ▶ The minimum requirement is 41 % of both tests.
- Failed tests must be corrected
- You must take an oral written exam. You cannot use anything
- ► Grades: Insufficient/Fail (1): 0-40 %, Sufficient/Pass (2): 41-55 %, Average (3): 56-70 %, Good (4): 71-85 %, Excellent (5): 86-100 %.
- ▶ Mid-term test 1: March 13, mid-term test 2: May 8, retake tests: May 15, 2025.

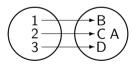
#### **Bibliography**

Bernard Kolman and David Hill: Elementary Linear Algebra with Applications, 9th ed., Person, 2007 Philip N. Klein: Coding the Matrix: Linear Algebra through Applications to Computer Science, Newtonian Press 2013

K. F. Riley, M. P. Hobson, S. J. Bence: Mathematical Methods for Physics and Engineering: A Comprehensive Guide, Cambridge University Press, 3rd. ed. (2006)

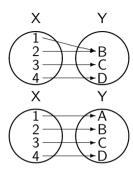
# Operators I

- ▶ <u>Definition</u>: The set is the sum of things. It is a fundamental term. You need a statement that collects the element. It means that you can decide whether an element is part of the set or not.
- <u>Definition</u>: The pair are sets consisting of two elements.
- ▶ <u>Definition</u>: Elements  $e_1$  and  $e_2$  consist of ordered pair if  $\{e_1, \{e_2\}\}$ . It sign is  $(e_1, e_2)$ .
- Definition: Relation is the set of ordered pairs.



▶ <u>Definition</u>: The injection orders different elements (X) to different elements (Y).

### Operators II



<u>Definition:</u> Surjections are those relations, that the values of the relation agree to the values of the set to order.

<u>Definition</u>: Bijection is an injection and a surjection. All elements are related to all elements of the other set.

# Operators III

▶ <u>Definition:</u> The functions are such a set of ordered pairs in that one element shows up only once:

$$(\forall x)(\forall y_1)(\forall y_2)[(x,y_1)\in f\wedge (x,y_2)\in f\Rightarrow y_1=y_2]$$

- ▶ <u>Definition</u>: V are U vectorspaces above  $\mathbb T$  body. The  $f:V\to U$  relation is linear if it is
  - 1. Additive, for all  $v_1, v_2 \in V$  vectors  $f(v_1 + v_2) = f(v_1) + f(v_2)$ .
  - 2. Homogen, for all  $v \in V$  vectors and  $\lambda \in \mathbb{T}$  elements  $f(\lambda v) = \lambda f(v)$ .
- <u>Definition</u>: Operators are the linear vector-vector functions.
- Például:
  - ldentical operator:  $\mathbf{A} \cdot \mathbf{1} = \mathbf{A}$ , for all  $\mathbf{A}$  operators.
  - Null operator:  $\mathbf{A} \cdot \mathbf{0} = \mathbf{0}$ , for all  $\mathbf{A}$  operators.

# Operators IV

- Mirror operators:  $(\mathbf{A} \cdot \mathbf{M}) \cdot \mathbf{M} = \mathbf{A}$ , for all **A** operators.
- Projection operator:  $\mathbf{A} \cdot \mathbf{P} = \mathbf{P}$ , for all **A** operator.
- Rotational operator: later.
- Operators could be multiplied on both sides.
- The representation of operators is the matrixes. See  $\alpha_{ij} \in \mathbb{R}$  for all  $i \in \{1, 2, ..., m\}$  and  $j \in \{1, 2, ..., n\}$ , where  $m, n \in \mathbb{N}^+$ . The

$$A = \begin{pmatrix} \alpha_{11} & \alpha_{12} & \cdots & \alpha_{1n} \\ \alpha_{21} & \alpha_{22} & \cdots & \alpha_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ \alpha_{m1} & \alpha_{m2} & \cdots & \alpha_{mn} \end{pmatrix}$$

table is called  $m \times n$  type matrix. The set of the  $m \times n$  type matrixes is  $M_{m \times n}$ .

# Operators V

- ▶ The spur of the matrix is the set of  $\{\alpha_{11}, \alpha_{22}, \dots, \alpha_{nn}\}$ .
- ▶ The first index of the elements  $\alpha_{ij}$  is the rowindex (i), the 2nd index is the column index (j).
- ▶ The Row *i* of the Matrix is  $A_i$ , and the Column *j* of the matrix is  $A_i$ .
- Determinant!!!

### Transpose I

▶ <u>Definition</u>: The transpose of the  $A = (\alpha_{ij})_{m \times n}$  matrix is the  $A^T = (\alpha_{ji})_{m \times n}$ . This means the change of the rows and columns. The transpose is the mirror of a square matrix.

$$A_{m\times n} = \begin{pmatrix} \alpha_{11} & \alpha_{12} & \cdots & \alpha_{1n} \\ \alpha_{21} & \alpha_{22} & \cdots & \alpha_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ \alpha_{m1} & \alpha_{m2} & \cdots & \alpha_{mn} \end{pmatrix} \quad A_{n\times m}^{T} = \begin{pmatrix} \alpha_{11} & \alpha_{21} & \cdots & \alpha_{1m} \\ \alpha_{12} & \alpha_{22} & \cdots & \alpha_{2m} \\ \vdots & \vdots & \ddots & \vdots \\ \alpha_{1n} & \alpha_{2n} & \cdots & \alpha_{mn} \end{pmatrix}$$

Examples for transpose.

### Matrix Operations I

Definition:  $A = (\alpha_{ij})_{m \times n}$  and  $B = (\beta_{ij})_{m \times n}$  are two matrixes with same type,  $\lambda \in \mathbb{R}$  a scalar. The sum of Matrixes A and B is Matrix  $A + B = (\alpha_{ij} + \beta_{ij})_{m \times n}$ , the  $\lambda$  times Matrix A is Matrix  $\lambda A = (\lambda \alpha_{ij})_{m \times n}$ .

$$A_{m \times n} = \begin{pmatrix} \alpha_{11} & \alpha_{12} & \cdots & \alpha_{1n} \\ \alpha_{21} & \alpha_{22} & \cdots & \alpha_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ \alpha_{m1} & \alpha_{m2} & \cdots & \alpha_{mn} \end{pmatrix} B_{m \times n} = \begin{pmatrix} \beta_{11} & \beta_{12} & \cdots & \beta_{1n} \\ \beta_{21} & \beta_{22} & \cdots & \beta_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ \beta_{m1} & \beta_{m2} & \cdots & \beta_{mn} \end{pmatrix}$$

$$A_{m \times n} + B_{m \times n} = \begin{pmatrix} \alpha_{11} + \beta_{11} & \alpha_{12} + \beta_{12} & \cdots & \alpha_{1n} + \beta_{1n} \\ \alpha_{21} + \beta_{21} & \alpha_{22} + \beta_{22} & \cdots & \alpha_{1n} + \beta_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ \alpha_{m1} + \beta_{m1} & \alpha_{m2} + \beta_{m2} & \cdots & \alpha_{mn} + \beta_{mn} \end{pmatrix}$$

# Matrix Operations II

$$\lambda A_{m \times n} = \begin{pmatrix} \lambda \alpha_{11} & \lambda \alpha_{12} & \cdots & \lambda \alpha_{1n} \\ \lambda \alpha_{21} & \lambda \alpha_{22} & \cdots & \lambda \alpha_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ \lambda \alpha_{m1} & \lambda \alpha_{m2} & \cdots & \lambda \alpha_{mn} \end{pmatrix}$$

The elements of the matrixes are added and multiplying by a scalar means to multiply all elements of the matrix by the scalar.

Examples for matrix operations.

### Matrix Operations III

▶ <u>Definition:</u>  $A = (\alpha_{ij})_{m \times n}$  and  $B = (\beta_{ij})_{n \times k}$  are two matrixes. The product of Matrixes A and B is Matrix  $A \cdot B = (\gamma_{ij})_{m \times k}$ , where

$$\gamma_{ij} = \sum_{l=1}^{n} \alpha_{il} \beta_{lj}.$$
Or:
$$A_{m \times n} = \begin{pmatrix}
\alpha_{11} & \alpha_{12} & \cdots & \alpha_{1n} \\
\alpha_{21} & \alpha_{22} & \cdots & \alpha_{2n} \\
\vdots & \vdots & \ddots & \vdots \\
\alpha_{i1} & \alpha_{i2} & \cdots & \alpha_{in} \\
\vdots & \vdots & \ddots & \vdots \\
\alpha_{m1} & \alpha_{m2} & \cdots & \alpha_{mn}
\end{pmatrix}$$

$$B_{n \times k} = \begin{pmatrix}
\beta_{11} & \beta_{12} & \cdots & \beta_{1j} & \cdots & \beta_{1k} \\
\beta_{21} & \beta_{22} & \cdots & \beta_{2j} & \cdots & \beta_{2k} \\
\vdots & \vdots & \ddots & \vdots & \ddots & \vdots \\
\beta_{n1} & \beta_{n2} & \cdots & \beta_{nj} & \cdots & \beta_{nk}
\end{pmatrix}$$

$$A \cdot B_{m \times k} = \begin{pmatrix}
\alpha_{11}\beta_{11} + \alpha_{12}\beta_{21} + \cdots + \alpha_{1n}\beta_{n1} & \alpha_{11}\beta_{12} + \alpha_{12}\beta_{22} + \cdots + \alpha_{1n}\beta_{n2} & \cdots & \alpha_{11}\beta_{1k} + \alpha_{12}\beta_{2k} + \cdots + \alpha_{1n}\beta_{nk} \\
\alpha_{21}\beta_{11} + \alpha_{22}\beta_{21} + \cdots + \alpha_{2n}\beta_{n1} & \alpha_{21}\beta_{12} + \alpha_{22}\beta_{22} + \cdots + \alpha_{2n}\beta_{n2} & \cdots & \alpha_{21}\beta_{1k} + \alpha_{22}\beta_{2k} + \cdots + \alpha_{2n}\beta_{nk} \\
\vdots & \vdots & \ddots & \vdots & \ddots & \vdots \\
\alpha_{m1}\beta_{11} + \alpha_{m2}\beta_{21} + \cdots + \alpha_{mn}\beta_{n1} & \alpha_{m1}\beta_{12} + \alpha_{m2}\beta_{22} + \cdots + \alpha_{mn}\beta_{n2} & \cdots & \alpha_{m1}\beta_{1k} + \alpha_{m2}\beta_{2k} + \cdots + \alpha_{mn}\beta_{nk}
\end{pmatrix}$$

Examples for matrix multiplications.

# Matrix Operations IV

► For similar square matrixes the condition of matrix production is fulfilled and the product will be the same type. Therefore, there is an exponentiation of matrixes:

$$A^1 = A$$
 és  $A^m = AA^{m-1}$ 

where  $(m \ge 2)$  és  $A \in \mathcal{M}_{n \times n}$ . Let us consider  $A^0 = E_m$ .

► <u>Thesis:</u> Equiations of matrix exponentation:

$$A^m A^k = A^{m+k}$$
  
$$(A^m)^k = A^{mk},$$

ahol  $m, k \in \mathbb{N}$ .

<u>Deduction:</u> It is trivial based on the definition of matrixproduct.

Examples for matrix exponentation.

# **Identity Matrix**

▶ <u>Definition</u>: The  $n^{th}$  order identity matrix is:

$$E_n = egin{pmatrix} 1 & 0 & \cdots & 0 \ 0 & 1 & \cdots & 0 \ dots & dots & \ddots & dots \ 0 & 0 & \cdots & 1 \end{pmatrix}$$

▶ Thesis: For all  $A \in \mathcal{M}_{n \times n}$ :  $A \cdot E_n = E_n \cdot A = A$ , or matrix  $E_n$  is identity element of the  $n \times n$  square matrixes for matrix production.

<u>Deduction:</u>  $A = (\alpha_{ij})_{n \times n}$  and  $E_n = (\beta_{ij})_{n \times n}$  are two matrixes, where  $\beta_{ij} = 1$ , if i = j, otherwise it is zero. The product of Matrixes A and  $E_n$  is Matrix  $A \cdot E_n = (\sum_{i=1}^n \alpha_{ij} \beta_{ij})$ . It is Matrix  $A = (\alpha_{ij})$  because the definition of

 $A \cdot E_n = (\sum_{l=1}^n \alpha_{il} \beta_{lj})_{n \times n}$ . It is Matrix  $A = (\alpha_{ij})_{n \times n}$ , because the definition of  $b_{ij}$  erases all other elements than  $\alpha_{ij}$ .

### Matrix Rank I

- ▶ <u>Definition:</u>  $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_s \in V$  are vectors. The rank of the  $\{\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_s\}$  vector system is the dimension of the  $\mathcal{L}(\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_s)$  subspace. Its sign is  $\rho(\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_s)$ .
- ▶ Thesis: The following transformation do not change the order of the  $\{a_1, a_2, ..., a_s\}$  vector system:
  - 1. Multiplying a vectors by a  $\lambda \neq 0$  scalar.
  - 2. Adding the vector multiplied by  $\lambda$  to another vector.
  - 3. Eliminating a vector that is a linear combination of the remaining vectors.
  - 4. Changing the order of vectors.
- ▶ <u>Definition</u>: The rank of Matrix  $A \in \mathcal{M}_{m \times n}$  is the rank of its row vector system.
- ► The rank of a matrix is equal to the common rank of the maximal ranked non-disappearing subdeterminants.

### Matrix Rank II

The rank of a matrix is determined by transforming the matrix to trapezoid form by rank invariant transformations. You can change the columns. (A matrix has trapesoid shape if  $\alpha_{ij} = 0$ , i > j, and  $\alpha_{ii} \neq 0$ , where  $(1 \leq i \leq \min\{m,n\})$ .) Rows and columns containing 0 could be deleted. The rank of the trapezoid matrix is the number of its rows.

Examples of determination of the rank of a matrix.

# Image processing I





- Matrix  $\mathbf{A}_{m \times n}$  is a representation of an  $m \times n$  greyscale image.
- All matrix elements gives color from the  $\{0, 1, ..., k\}$  range, where 0 is black, k-1 is white and k means transparency.
- ▶ The  $\mathbf{B}_{m \times n}$  background is transparent.
- Let us construct the  $\mathbf{A} \odot \mathbf{B} = [a_{ij} \odot b_{ij}]$  operations that copies the 2nd image to the background of the first image.
- Formula:  $[a_{ij} \odot b_{ij}] = \begin{cases} b_{ij}, & \text{if } a_{ij} = k. \\ a_{ij}, & \text{otherwise.} \end{cases}$
- Let us use the  $x \mapsto \lfloor x \rfloor$  function, that means to rounding to the lower integer value.
- ▶ If  $a \in [0, k]$ , then  $0 \le a/k \le 1$ , therefore  $\lfloor a/k \rfloor$  is 0, or 1.

# Image processing II





- Or  $\lfloor a/k \rfloor$  is 1, if only a=k, or the pixel is transparent, otherwise it is 0.
- ▶  $1 \lfloor a/k \rfloor$  is 0, if only a = k, or otherwise it is 1.
- ▶ Therefore,  $a \odot b = \left\lfloor \frac{a}{k} \right\rfloor b + \left(1 \left\lfloor \frac{a}{k} \right\rfloor\right) a$  is the desired operation.
- (Top) 32 × 24 matrix. (Bottom left) Original image. (Bottom middle) Background. (Bottom right) The result of the operation.

### Moment of inertia

- Mass times square of distance from the axis.
- $\triangleright$  For a rigid object of N point masses  $m_k$ , the moment of inertia tensor is given by

$$\mathbf{I} = \begin{pmatrix} I_{11} & I_{12} & I_{13} \\ I_{21} & I_{22} & I_{23} \\ I_{31} & I_{32} & I_{33} \end{pmatrix},$$

where

$$I_{ij} = \sum_{k=1}^{N} m_k \left( \|\mathbf{r}_k\|^2 \, \delta_{ij} - x_i^{(k)} x_j^{(k)} \right)$$

and  $i, j \in \{1, 2, 3\}$ ,  $\mathbf{r}_k = \left(x_1^{(k)}, x_2^{(k)}, x_3^{(k)}\right)$  is the vector to the point mass  $m_k$  from the point about which the tenzor is calculated, and  $\delta_{ij}$  is the Kronecker delta.

#### Transformation matrixes I

Rotational matrix in 2D:

$$\begin{pmatrix} \cos \alpha & -\sin \alpha \\ \sin \alpha & \cos \alpha \end{pmatrix}.$$

▶ Rotational matrixes in 3D around z, x, y axes:

$$\begin{pmatrix} \cos\alpha & -\sin\alpha & 0 \\ \sin\alpha & \cos\alpha & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 1 \\ 0 & \cos\alpha & -\sin\alpha \\ 0 & \sin\alpha & \cos\alpha \end{pmatrix} \begin{pmatrix} \cos\alpha & 0 & \sin\alpha \\ 0 & 1 & 0 \\ -\sin\alpha & 0 & \cos\alpha \end{pmatrix}.$$

▶ Mirror of the vectors of the plan for the  $\alpha/2$  angular line:

$$\begin{pmatrix} \cos \alpha & \sin \alpha \\ \sin \alpha & -\cos \alpha \end{pmatrix}.$$

#### Transformation matrixes II

▶ Mirror of the vectors of the 3D space for the **n** normal vector planes:

$$\mathbf{M} = \mathbf{I} - 2\mathbf{n} \otimes \mathbf{n}^T$$
.

Perpendicular projection to a line with **b** direction vector:

$$\mathsf{P} = \frac{1}{\mathsf{b}\mathsf{b}^{\mathsf{T}}}\mathsf{b} \otimes \mathsf{b}^{\mathsf{T}}.$$

Perpendicular projection to the plan with **n** normal vector:

$$P = I - n \otimes n^T$$
.

### Transformation matrixes III

 $\triangleright$  Shift by (a, b) vector in 2D:

$$\begin{pmatrix} 1 & 0 & a \\ 0 & 1 & b \\ 0 & 0 & 1 \end{pmatrix}.$$

▶ Shift by (a, b, c) vector in 3D:

$$\begin{pmatrix} 1 & 0 & 0 & a \\ 0 & 1 & 0 & b \\ 0 & 0 & 1 & c \\ 0 & 0 & 0 & 1 \end{pmatrix}.$$

# Comming soon...

- Diagonal matrixes
- ▶ Permutation matrixes and snakes
- ► Triangular matrixes
- Symmetric and skew-symmetric matrixes

# The End

Thank you for your attention!