

# **Lineær algebra - 1**

## **Introduction**

**Lineær algebra og dynamiske  
systemer**

**Troels B. Sørensen /Gilberto Berardinelli**

# Welcome

- **Two course lecturers**
  - **Troels B. Sørensen (first part)**, tbs@es.aau.dk
  - **Gilberto Berardinelli (second part)**, gb@es.aau.dk
- **Course information**
  - Curricula [https://moduler.aau.dk/course/2023-2024/  
ESNESDB4K2?lang=en-GB](https://moduler.aau.dk/course/2023-2024/ESNESDB4K2?lang=en-GB)
  - **Lecture information – reading, exercises, slides, .. on  
Moodle course page**
- **Two lectures per week**
  - **For about five weeks**

# Literature

- Books
  - "Advanced Engineering Mathematics", 10th ed., 2011 af Erwin Kreyszig, ISBN: 978-0-470-64613-7 (part 1)
  - "Discrete-Time Signal Processing", 2nd (Prentice-Hall, Inc. 1999) or 3rd edition (Pearson 2014), A. V. Oppenheim and R. W. Schafer (part 2)
  - Older versions will do as well, but you need to "translate" - topics are quite standard in engineering

# Course plan

- **Linear Algebra (about 50% of the course)** *Tveks*
  - Matrix basics, determinants, rank, linear independence, Gauss elimination
  - Linear systems of equations, vector spaces and basis vectors, linear transformations, some numerical considerations
  - Eigenvalue problems, vector bases, numerical methods
  - Taylor and MacLaurin series *(Gilberto)*
  - Systems of differential equations, eigenvalue problems, stability
  - Systems of differential equations, linearization
- **Discrete LTI systems (about 50% of the course)** *Gilberto*
  - Continuous to discrete domain (link to previous courses)
    - Linearity and time invariance, impulse response and convolution, stability and causality
  - The z-transform
    - Definition and region of convergence (ROC), right- and left-sided sequences, ROC analysis
  - Inverse z\_transform
    - Defintion and inspection method, partial fraction expansion
  - Transform analysis
    - Linear constant coefficient difference equations, stability and causality, inverse systems

# Some prerequisites

- **Calculus (ESD1):**
  - complex numbers, first and second-order differential equations and Laplace as a transform
- **Kredsløbsteori (ESD1):**
  - application of first and second order differential equations and Laplace (with interpretation of time-frequency relation)
- **Instrumentering, interfacekredsløb og dynamiske systemer (ESD2):**
  - fundamentals on feedback (stability), basic linear dynamic systems

# **Course execution and work load**

- **Classical lecture in class**
  - Provides overview (hopefully), add supplementary information (a different view maybe), some examples
- **Exercises in groups/individual**
  - Important part of the curricula to apply theory and be confident with the topics
  - This is not (directly) exam preparation!
  - Advise to minimize the use of math programs
- **Course load**
  - A 5 ECTS course with expected 150 hours of effort (preparation for class, lecture in class, exercises in groups, follow up from lecture, preparation for the exam, exam attendance)

# Course exam

- **A written three hour exam, start of June**
  - The exam assignments will represent a subset of the course curricula – we sample
  - Allowed use of literature in physical and/or electronic format, personal notes, math programs – but ...
  - Online access to digital exam (only!)
- **No "type assignments"**
  - But, application of learned topic(s)
  - A typical exam exercise will combine different elements of what has been practized during exercise sessions
  - We will provide some example exercises during spring (a new course)

# Exercises

- **Support how?**
  - Pay a visit (in B1-101)?
  - Send an email?
  - ?
- **Solutions?**
  - Will become available (to check) but we would like you to work "from a blank sheet", as preparation for the exam situation



# Matrix notation

- columns n -

$\bar{\bar{A}} = \begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & a_{ij} & \vdots \\ a_{m1} & \dots & a_{mn} \end{bmatrix}$  rows m =  $[a_{ij}]$

$\text{size}(\bar{\bar{A}}) = m \times n$

$\bar{\bar{A}}$  by column vector

Square matrices  
will be particularly  
interesting  $m=n$

$$\bar{\bar{A}} = [\bar{b}_1, \bar{b}_2, \dots, \bar{b}_n], \bar{b}_i = \begin{bmatrix} b_{1i} \\ \vdots \\ b_{ni} \end{bmatrix}$$

## Matrix notation

Matrix algebra:  $\bar{y} = \bar{A}\bar{x}$  "similar to numbers"

Matrix equality:  $\bar{A} = \bar{B} \Leftrightarrow [a_{ij}] = [b_{ij}]$   
size( $\bar{A}$ ) = size( $\bar{B}$ )

Matrix transpose:

$$\bar{A}^T = [a_{ij}]^T = [a_{ji}]$$

$\bar{A}$  by row vector

$$\bar{A} = [\bar{a}_1^T, \bar{a}_2^T, \dots, \bar{a}_m^T]^T, \text{ since } \bar{a}_i = (\bar{a}_i^T)^T$$
$$\bar{a}_i = [a_{i1}, \dots, a_{in}]$$

# Special matrices (square)

Diagonal

$$\begin{bmatrix} a & & & \\ 0 & b & & \\ & c & \ddots & \\ & & & d \end{bmatrix}$$

Triangular

lower/upper

$$\begin{bmatrix} & & & \\ 0 & & & d \\ & \ddots & & \\ & & \ddots & 0 \end{bmatrix}$$

Scalar

$$\begin{bmatrix} k & & & \\ 0 & k & & \\ & \ddots & \ddots & \\ & & 0 & k \end{bmatrix} = k \cdot \mathbb{I}$$

Identity

$$\begin{bmatrix} 1 & & & \\ 0 & 1 & & \\ & \ddots & \ddots & \\ & & & 1 \end{bmatrix} = \mathbb{I}$$

Null (in various forms)  $\bar{0}$  ?

# Special matrices (square)

Symmetric

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

e.g.

$$\begin{bmatrix} 5 & -1 & 0 \\ -1 & 9 & 5 \\ 0 & 5 & 7 \end{bmatrix}$$

Skew-Symmetric

$$\bar{A}^T = -\bar{A}$$

e.g.

$$\begin{bmatrix} 0 & -6 & 2 \\ 6 & 0 & 1 \\ -2 & -1 & 0 \end{bmatrix}$$

OBS ↗

what is particular  
about it?

slide

# Special matrices (square)

## Hankel

$$\begin{bmatrix} a & b & c & d & \dots \\ b & c & d & e & \\ c & d & e & f & \\ d & e & f & & \\ \vdots & \vdots & & & \end{bmatrix}$$

*Symmetric*

“stacking” measurements  
for subsequent  
processing/estimation

## Toeplitz

“convolving” a signal  
through a system  
(not necessarily square) →

$$\begin{bmatrix} a & b & c & d & e & f & g & h \\ h & a & b & c & d & e & f & g \\ g & h & a & b & c & d & e & f \\ f & g & h & a & b & c & d & e \end{bmatrix}$$

# Matrix algebra (addition)

## Addition

$$\bar{\bar{A}} + \bar{\bar{B}} = \bar{\bar{B}} + \bar{\bar{A}}$$

commutative

$$(\bar{\bar{U}} + \bar{\bar{V}}) + \bar{\bar{W}} = \bar{\bar{U}} + (\bar{\bar{V}} + \bar{\bar{W}})$$

associative

$$\bar{\bar{A}} + \bar{\bar{0}} = \bar{\bar{A}}$$

null element

$$\bar{\bar{A}} + (-\bar{\bar{A}}) = \bar{\bar{0}}$$

negative (anti)  
element

## Multiplication

$$\bar{\bar{A}} \bar{\bar{B}} \neq \bar{\bar{B}} \bar{\bar{A}}$$

can happen but  
in general not

# Matrix algebra (multiplication)

$m \times p$

$m \times n$

$n \times p$

$$\bar{\bar{A}} \bar{\bar{B}} = [\bar{a}_1^T, \bar{a}_2^T, \dots, \bar{a}_m^T]^T [\bar{b}_1, \bar{b}_2, \dots, \bar{b}_p]$$

$$= \begin{bmatrix} \bar{a}_1 \cdot \bar{b}_1, \bar{a}_1 \cdot \bar{b}_2, \dots, \bar{a}_1 \cdot \bar{b}_p \\ \vdots & \downarrow k^{\text{'th}} \text{ column} \\ \bar{a}_j \cdot \bar{b}_k \\ \vdots & \leftarrow j^{\text{'th}} \text{ row} \\ \bar{a}_m \cdot \bar{b}_1, \bar{a}_m \cdot \bar{b}_2, \dots, \bar{a}_m \cdot \bar{b}_p \end{bmatrix}$$

$$\bar{a} \cdot \bar{b} = \bar{a}^T \bar{b}$$

inner (dot) product  
outer ( $\otimes$ ) product

# Matrix algebra (mult. cont.)

Since  $\bar{A}\bar{B} \neq \bar{B}\bar{A}$  it means that

$$\begin{array}{l} \bar{B} = \bar{C} - \bar{D} \\ \downarrow \\ \bar{A}\bar{B} = \bar{A}\bar{C} - \bar{A}\bar{D} \end{array} \quad \cancel{\Rightarrow} \quad \bar{A} = 0, \bar{B} = 0, \bar{B}\bar{A} = 0$$
$$\cancel{\Rightarrow} \quad \bar{C} = \bar{D}$$

DOES NOT 

Multiplication order is important  
pre- OR post-multiply  
NOT THE SAME

$\bar{A}\bar{I} = \bar{I}\bar{A}$   
 $\bar{A}\bar{I} = \bar{A}$   
but

# Matrix algebra (mult. cont.)

---

$$\bar{A}(\bar{B}\bar{C}) = (\bar{A}\bar{B})\bar{C}$$

associative

$$(\bar{A} + \bar{B})\bar{C} = \bar{A}\bar{C} + \bar{B}\bar{C}$$

distributive

$$\bar{C}(\bar{A} + \bar{B}) = \bar{C}\bar{A} + \bar{C}\bar{B}$$

(wrt. addition)

⚠️ observe order

We can say more about the matrix product when we know the matrix rank (later)

# Matrix algebra (scalars)

---

$$c(\bar{A} + \bar{B}) = \bar{c}\bar{A} + \bar{c}\bar{B}$$

distributive  
(wrt. add./mult.)

$$(c+k)\bar{A} = c\bar{A} + k\bar{A}$$

$$c(k\bar{A}) = (ck)\bar{A}$$

$$(k\bar{A})\bar{B} = k(\bar{A}\bar{B})$$

associative

$$= \bar{A}(k\bar{B})$$

$$1\bar{A} = \bar{A}$$

identity

# Matrix algebra (transpose)

$$(\bar{A} + \bar{B})^T = \bar{A}^T + \bar{B}^T \quad \text{distributive}$$

$$(c \bar{A})^T = c \bar{A}^T \quad (\text{wrt. matrix add. scalar mult.})$$

$$(\bar{A} \bar{B})^T = \bar{B}^T \bar{A}^T \quad \text{NOT distributive}$$

$$(A^n)^T = (A^T)^n \quad (\text{wrt. matr. mult.})$$

we will see later that a similar reverse order applies to matrix inverse

# Matrix determinant ( $n \times n$ )

The determinant is defined for square  $\bar{A}$  ( $n \times n$ )

Row expansion (choose any row  $j = 1, \dots, n$ )

$\det(\bar{A}) = a_{j1} C_{j1} + a_{j2} C_{j2} + \dots + a_{jn} C_{jn}$ , where

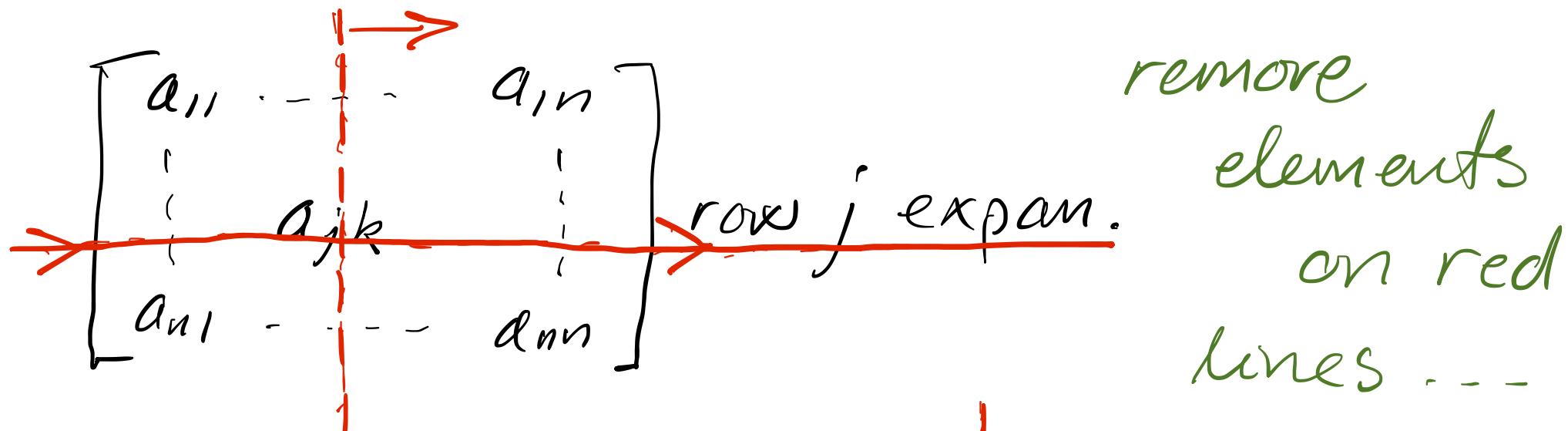
the co-factor of  $a_{jk}$  is  $C_{jk} = (-1)^{j+k} M_{jk}$

with  $M_{jk}$  the minor of  $a_{jk}$  (sub-determinant)

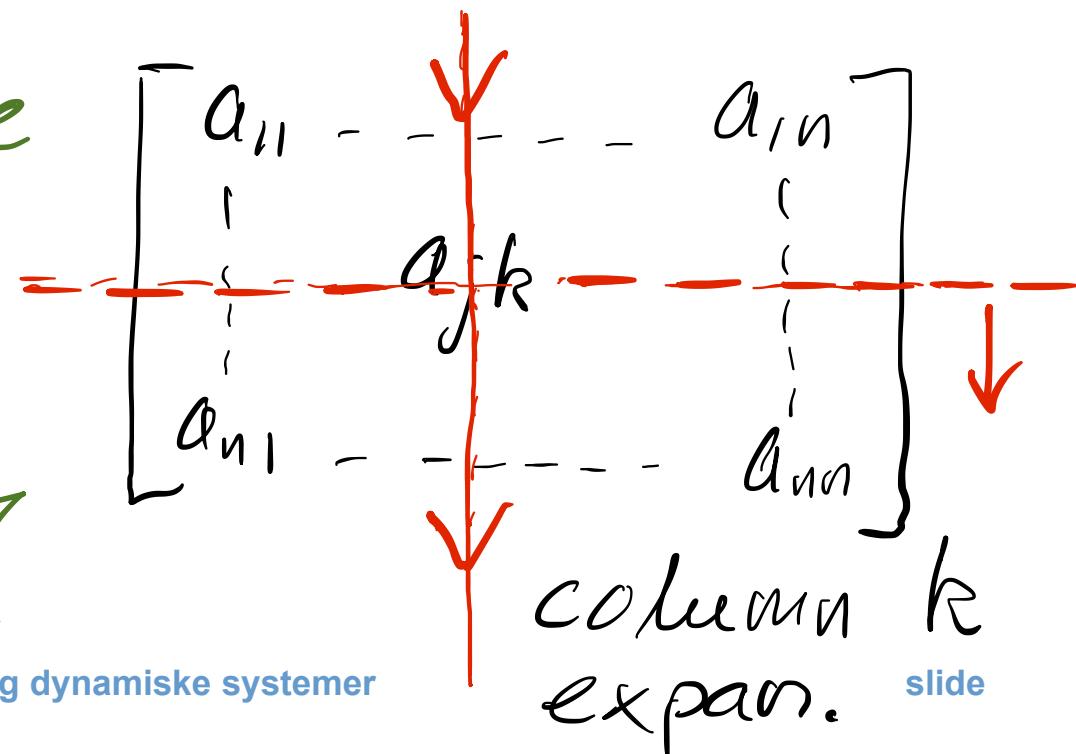
Column expansion (choose any column  $k$ )

$\det(\bar{A}) = a_{1k} C_{1k} + a_{2k} C_{2k} + \dots + a_{nk} C_{nk}$

# Matrix determinant (Minor)



--- and calculate  
subdeterminant,  
i.e. minor  $M_{jk}$   
of order  $n-1$  !



## Determinant (further)

$$\det(\bar{A}) = \det(\bar{A}^T) \quad \text{as per co-factor expansion}$$

and,

$$\det(\bar{A}\bar{B}) = \det(\bar{A})\det(\bar{B}) = \det(\bar{B}\bar{A})$$

despite  $\bar{A}\bar{B} \neq \bar{B}\bar{A}$  in general

# Matrix inverse ( $n \times n$ )

Singular ("singular") if  $\det(\bar{A}) = 0$

Non-singular ("regular") if  $\det(\bar{A}) \neq 0$

If  $\bar{A} (n \times n)$  is non-singular (why?):

$$\underline{\text{inv}(\bar{A})} = \bar{A}^{-1} = \frac{1}{\det(\bar{A})} [C_{jk}]^T = [C_{kj}]$$

!  $C_{jk}$  is the co-factor of  $a_{jk}$  placed in element  $kj$  position

# Matrix Inverse (mult.)

The inverse is that for which

$$\bar{A} \bar{A}^{-1} = \bar{A}^{-1} \bar{A} = \bar{I}$$

from which we see by substitution

$$\begin{aligned} \bar{I} &= (\bar{A} \bar{C})(\bar{A} \bar{C})^{-1} && \text{pre-multiply } \bar{A}^{-1} \\ \bar{A}^{-1} \bar{I} &= \bar{A}^{-1} = (\bar{A}^{-1} \bar{A}) \bar{C} (\bar{A} \bar{C})^{-1} = \bar{C} (\bar{A} \bar{C})^{-1} && \text{pre-multiply } \bar{C}^{-1} \\ \bar{C}^{-1} \bar{A}^{-1} &= (\bar{C}^{-1} \bar{C})(\bar{A} \bar{C})^{-1} = (\bar{A} \bar{C})^{-1} && \end{aligned}$$

# Matrix algebra (inverse)

$$\bar{\bar{I}} \bar{\bar{I}}^{-1} = \bar{\bar{I}} \iff \bar{\bar{I}}^{-1} = \bar{\bar{I}}$$

$$(\bar{\bar{A}}^{-1})^{-1} = \bar{\bar{A}}, \text{ since } \bar{\bar{A}}^{-1}(\bar{\bar{A}}^{-1})^{-1} = \bar{\bar{A}}^{-1}\bar{\bar{A}} = \bar{\bar{I}}$$

$$(\bar{\bar{A}}^n)^{-1} = \bar{\bar{A}}^{-1}\bar{\bar{A}}^{-1}\dots\bar{\bar{A}}^{-1} = (\bar{\bar{A}}^{-1})^n \quad \text{cf. prev. slide}$$

$$(\bar{\bar{A}}^{-1})^T = (\bar{\bar{A}}^T)^{-1}, \text{ since } \bar{\bar{I}} = \bar{\bar{I}}^T = (\bar{\bar{A}}\bar{\bar{A}}^{-1})^T \\ = (\bar{\bar{A}}^{-1})^T\bar{\bar{A}}^T$$

$$\det(\bar{\bar{I}}) = 1$$

$$\det(\bar{\bar{I}}) = \det(\bar{\bar{A}}\bar{\bar{A}}^{-1}) = \det(\bar{\bar{A}}) \cdot \det(\bar{\bar{A}}^{-1}) \\ \iff \det(\bar{\bar{A}}^{-1}) = 1 / \det(\bar{\bar{A}})$$



# Echelon form

A triangular matrix, such that:

- all 0's are below, left
- first  $a'_{ij} \neq 0$  in each row is to the right of the  $a'_{ij} (\neq 0)$  in the row above AND has only 0's before it

⚠ first  $a'_{ij}$ 's are not necessarily on one diagonal

⚠ there can be rows with all 0's

$$\begin{bmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & & \vdots \\ a_{mn} & \cdots & a_{mn} \end{bmatrix} (= \bar{A})$$

pivot - "the central element"

$$\begin{bmatrix} a'_{11} & \cdots & a'_{1n} \\ 0 & \cdots & 0 \\ a'_{ij} & \cdots & a'_{in} \\ 0 & \cdots & 0 \\ 0 & \cdots & 0 \end{bmatrix} (= \bar{A}')$$

row equivalent BUT  
different  $\bar{A}$

# Echelon form - examples ?

---

$$\checkmark \times \begin{bmatrix} 1 & 0 & 0 & 0 & 7 & 5 & 3 \\ 0 & 1 & 0 & 0 & 3 & 1 & 9 \\ 0 & 0 & 5 & 0 & 8 & 0 & 1 \\ 0 & 0 & 0 & 1 & 7 & 7 & 7 \end{bmatrix}$$

$$\checkmark \times \begin{bmatrix} 9 & 1 & 3 & 5 & 6 & 1 & 5 \\ 0 & 0 & 0 & 6 & 2 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 & 9 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

$$\checkmark \times \begin{bmatrix} 0 & 1 & 0 & 9 & 6 & 3 \\ 0 & 5 & 1 & 6 & 9 & 3 \\ 0 & 0 & 0 & 1 & 2 & 3 \\ 0 & 0 & 0 & 0 & 1 & 1 \end{bmatrix}$$

## Reduced echelon form

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A triangular matrix, such that:

- it fulfills the echelon form
- in addition, the first  $a'_{ij}$  in each row is 1 with 0's below AND above

↙  
a the pivot is 1

$$\left[ \begin{array}{cccccc} 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \end{array} \right] \quad \sum$$

## Reduced echelon form - examples

---

z.

$$\begin{matrix} \checkmark & \times \\ \begin{bmatrix} 1 & 0 & 0 & 0 & 7 & 5 & 3 \\ 0 & 1 & 0 & 0 & 3 & 1 & 9 \\ 0 & 0 & 1 & 0 & 8 & 0 & 1 \\ 0 & 0 & 0 & 1 & 7 & 7 & 7 \end{bmatrix} \end{matrix}$$

z.

$$\begin{matrix} \times & \checkmark \\ \begin{bmatrix} 5 & 7 & 9 & 13 & 1 \\ 0 & 3 & 3 & 3 & 3 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \end{matrix}$$

# Gauss elimination

---

Elementary row operations:

1) multiplication of a row  
by a constant  $k \neq 0$

2) swapping rows

3) adding a weighted sum  
of rows to another row

! these operations forms the  
basis of Gauss elimination

Reduces a matrix to  
the row equivalent  
echelon form, and/or  
reduced echelon  
form

! can be done on  
° rows OR columns

# Gauss elimination - by matrix ops.

Operation  $\begin{cases} \text{rows : } EA \text{ (pre)} \\ \text{columns : } AE \text{ (post)} \end{cases}$

Case illustration of  $3 \times 3$  for

Multiply by a constant

$$E = \begin{bmatrix} 1 & 0 & 0 \\ 0 & k & 0 \\ 0 & 0 & 1 \end{bmatrix} \leftarrow \begin{array}{l} \text{this row} \\ \text{OR} \\ \uparrow \text{this column} \end{array}$$

Swap rows / columns

$$E = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \leftarrow \begin{array}{l} \text{these rows} \\ \text{OR} \\ \text{these columns} \end{array}$$

Multiply and add

$$E = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -3 & 1 \end{bmatrix} \leftarrow \begin{array}{l} \text{this row} \\ \text{OR} \\ \uparrow \text{this column slide} \end{array}$$



## Determinant (Gauss)

---

From the echelon form, we can calculate

$$\det(\bar{A}) = (-1)^r \cdot \prod_i p_i$$

where  $p_i$  are the (diagonal) pivots  
and  $r$  is the number of row inter-  
changes.

!  $\det(\bar{A}) \neq 0 \Rightarrow$  (diagonal)  $p_i \neq 0$

# Gauss elimination (det.)

---

1) Multiplication by  $k \Rightarrow k \cdot \det(\bar{A})$

OBS!  $\det(k\bar{A}) = k^n \det(\bar{A})$

2) Swapping rows  $\Rightarrow -1 \cdot \det(\bar{A})$

3) Adding a weighted sum  $\Rightarrow$  no change

⚠ Proportional rows/columns  $\Rightarrow$  zero rows/columns  $\Rightarrow \det(\bar{A}) = 0$



# Linear independence

---

Given  $P$  vectors (  $n$  columns  $\bar{b}_i$ 's or  $m$  rows  $\bar{a}_i$ 's )

$$c_1 \bar{v}_1 + c_2 \bar{v}_2 + \dots + c_p \bar{v}_p = \bar{0}$$

If only solution is for scalars

$c_i = 0$ ,  $1 \leq i \leq P$ , then  $\bar{v}_i$ 's

are linearly independent

"the essential vectors"

## Linear independence (cont.)

In Gauss elimination we find the "essential vectors" - usually row vectors.

$\text{rank}(\bar{A})$

The maximum number of linearly independent vectors (row or column) is called the rank of  $\bar{A}$

$\nabla \bar{A} = \bar{0} \Rightarrow \text{rank}(\bar{A}) = 0 \quad \nabla \text{rank}(\bar{A}) = \text{rank}(\bar{A}^T)$

# Matrix rank

---

Row equivalent matrices have the same rank  $\Leftrightarrow$  we can find the rank by Gauss elimination,  
i.e. how many non-zero rows (or columns) result in the end.

! We need to check both rows and columns to get  $\text{rank}(\bar{A})$

# Rank implication

$$\bar{A} = \begin{bmatrix} \bar{a}_1 \\ \vdots \\ \bar{a}_m \end{bmatrix}, \quad \bar{a}_i = [a_{i1}, \dots, a_{in}]$$

↑ n columns

If  $\text{rang}(\bar{A}) = p$ , then:

$$c_1 \bar{a}_1' + c_2 \bar{a}_2' + \dots + c_p \bar{a}_p' = \bar{o} \iff$$

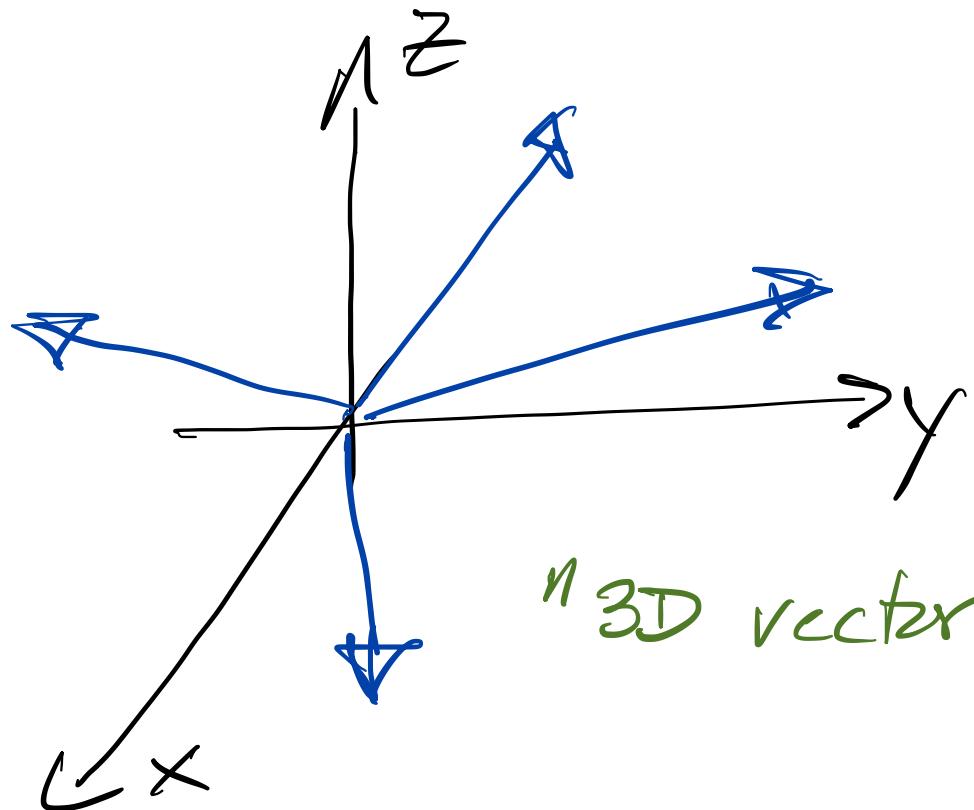
$$\forall c_i = 0, \quad 1 \leq i \leq p$$

If  $n < p$ , since  $\text{rank}(\bar{A}) = \text{rank}(\bar{A}^T)$   
then  $\text{rank}(\bar{A}) \leq n < p$

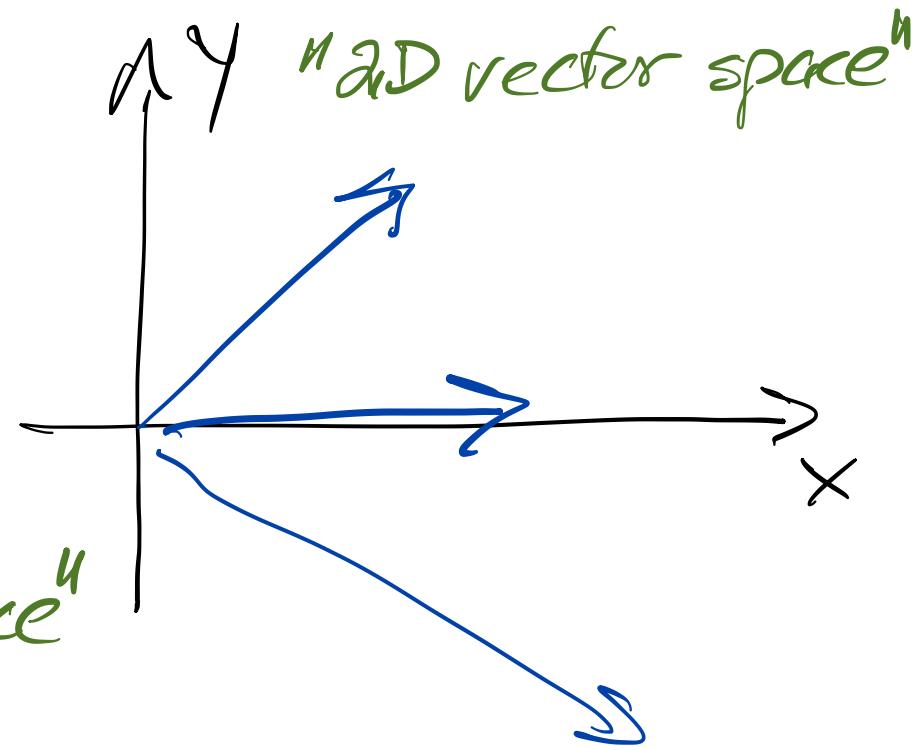
# Rank implication (cont.)

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$p$  vectors with  $n < p$  components  
are linearly dependent, always



"3D vector space"



"2D vector space"

## Rank implications (cont.)

An  $m \times n$  matrix  $\bar{A}$  has rank  $r \Leftrightarrow$   
 $\bar{A}$  has an  $r \times r$  submatrix  $\bar{A}_r$  with  
 $\det(\bar{A}_r) \neq 0$  and  $\det(\bar{A}_{r+1}) = 0$

$\bar{A}$  is said to be rank deficient  
if  $\text{rank}(\bar{A}) < \min(m, n)$  (singular)

- ?
- A square matrix  $\bar{A}$ ,  $n \times n$ , is full rank  $\Leftrightarrow \text{rank}(\bar{A}) = n$  and  $\det(\bar{A}) \neq 0$

# Rank multiplication (cont.)

$$\text{inv}(\bar{A}) = \bar{A}^{-1} \quad (n \times n) \iff \det(\bar{A}) \neq 0$$

$$\iff \underline{\text{rank}(\bar{A}) = n} \quad \text{Also :}$$

$$\bar{A} \bar{B} = \bar{A} \bar{C} \iff \bar{B} = \bar{C}$$

$$\bar{A} \bar{B} = \bar{0} \iff \bar{B} = \bar{0}$$

In general  
 $\bar{A} \bar{B} \neq \bar{B} \bar{A}$



