



MASTER IN ENTREPRENEURSHIP  
INNOVATION MANAGEMENT  
IN COLLABORATION WITH MIT SLOAN

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**MIT** MANAGEMENT  
SLOAN SCHOOL



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# DIGITAL TECH

# High Performance Computing

Lesson 1: Basic scientific computing tools

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## Lesson 1: Basic scientific computing tools

1. Vectors (arrays)
2. Matrices
3. Linear systems

## Vectors: preliminaries

Different perspective:

- Physics: length, direction (pointing from A to B)  $v \longrightarrow$
- Mathematician: couple (ordered pairs, tuples) of coordinates between commas

$$v = (v_1, v_2) \quad \text{couple,} \quad \text{two-dimensional vector}$$

$$w = (w_1, w_2, \dots, w_t) \quad \text{tuple,} \quad \text{t-dimensional vector}$$

- Computer science: list of numbers. Row and column vectors

$$w = [5, 3, 6, -4, \pi] \quad v = \begin{bmatrix} 6 \\ 2 \\ 4 \end{bmatrix}$$

$v_1, v_2, \dots$  are called components

## Vectors: preliminaries (2)

Let be  $v = [v_1, v_2, \dots, v_n]$   $w = [w_1, w_2, \dots, w_n]$  vectors and  $k \in \mathbf{R}$  a real number (scalar)

- norm (length)  $\|v\| = \sqrt{v_1^2 + v_2^2 + \dots + v_n^2}$
- sum  $v + w = [v_1 + w_1, v_2 + w_2, \dots, v_n + w_n]$
- inner (standard, scalar) product  $v \cdot w = v_1 \cdot w_1 + v_2 \cdot w_2 + \dots + v_n \cdot w_n$   
(please note: the result is a number).
- $kv = k \cdot v = [kv_1, kv_2, \dots, kv_n]$  is the scalar multiplication.  
(please note: the result is a vector). The resulting vector is k times as long as v (k times the original length)
- Pointwise (Hadamard) product  $v \circ w = [v_1 \cdot w_1, v_2 \cdot w_2, \dots, v_n \cdot w_n]$

## Vectors: preliminaries (3)

- if  $\|v\| = 1$  then  $v$  is called **Unit vector**
- vector  $\mathbf{0} = [0, 0, \dots, 0]$  is called **Zero vector**. Is the **neutral element** for the additive operation.

It is the unique vector of norm zero.

- $z = -v = [-v_1, -v_2, \dots, -v_n]$  is the **opposite vector** of  $v$ . Indeed  $v + z = v + (-v) = \mathbf{0}$ .
- the **angle** between vectors  $v, w \neq \mathbf{0}$  is given by  $\Theta_{v,w} = \arccos \frac{v \cdot w}{\|v\| \cdot \|w\|}$
- note that if  $v \cdot w = 0$  vectors are called **orthogonal** (perpendicular, normal)
- if  $v \neq \mathbf{0}$  the vector  $w = v/\|v\|$  is a unit vector

## Vectors: preliminaries (4)

- $V = V_n = \{[v_1, v_2, \dots, v_n] \mid v_i \in R\}$  is the (real) linear space of dimension  $n$ .
- a linear combination of  $k$  elements of  $V$  is of the form

$$a_1v_1 + a_2v_2 + \dots + a_kv_k$$

where  $a_1, \dots, a_k$  are scalar and  $v_1, \dots, v_k$  are vectors of  $V$

Scalars are also called the **coefficients of the linear combination**.

- Vectors  $S = \{v_1, v_2, \dots, v_k\}$  are said linearly independent, if no element of  $S$  can be written as a linear combination of the other elements of  $S$ .
- a basis  $B$  of  $V_n$  is a linearly independent set  $B$  of size  $n$
- standard basis is

$$e_1 = [1, 0, 0, \dots, 0, 0], \quad e_2 = [0, 1, 0, \dots, 0, 0], \quad \dots \quad e_n = [0, 0, 0, \dots, 0, 1]$$

## Vectors: preliminaries (5)

$$(i) \ v \cdot v = |v|^2;$$

$$(ii) \ v \cdot w = w \cdot v;$$

$$(iii) \ v \cdot (w + z) = v \cdot w + v \cdot z;$$

$$(iv) \ v \cdot (\alpha w) = (\alpha v) \cdot w = \alpha(v \cdot w);$$

$$(v) \ |v \cdot w| \leq |v||w|;$$

$$(vi) \ |v + w| \leq |v| + |w|;$$

Cauchy-Schwartz  
inequality

Triangle  
inequality

## Lesson 1: Basic scientific computing tools

1. Vectors (arrays)
2. **Matrices**
3. Linear systems



## Matrices: main concepts

- A matrix  $A \in R^{m \times n}$  is a two-dimensional table in two big (round or square) brackets with  $m$  rows and  $n$  columns

$$A = \begin{bmatrix} a_{1,1} & a_{1,2} & \dots & a_{1,n} \\ a_{2,1} & a_{2,2} & \dots & a_{2,n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m,1} & a_{m,2} & \dots & a_{m,n} \end{bmatrix}.$$

–  $A$  is a rectangular/square array of numbers, arranged in rows and columns.

–  $m \times n$  is the order of the matrix

– row vectors are matrices with  $m = 1$  of order  $1 \times n$

- Please note:

– column vectors are matrices with  $n = 1$  of order  $m \times 1$

## Matrices: main concepts

- Matrices are usually denoted by capital letters

(example) Matrix  $C = \begin{bmatrix} 1 & 2 & -3 \\ 3 & 4 & 0 \end{bmatrix}$  is a  $2 \times 3$  matrix.

- $A$  is also denoted as

$$A = (a_{i,j})$$

where real number  $a_{i,j}$  is called **entry** or **element** with row index  $i$  and column index  $j$

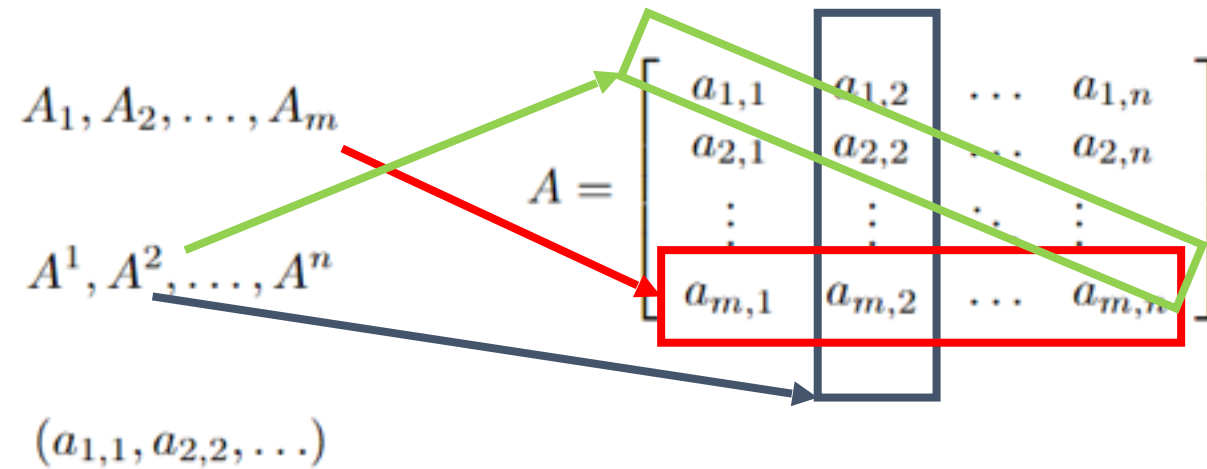
- A matrix with  $m = n$  is called **square matrix** of order  $n$

A matrix with  $m \neq n$  is called **rectangular matrix**

## Matrices: main concepts

A matrix  $A$  of order  $m \times n$  contains

- $m$  rows usually denoted by  $A_1, A_2, \dots, A_m$
- $n$  columns usually denoted by  $A^1, A^2, \dots, A^n$
- main diagonal is given by  $(a_{1,1}, a_{2,2}, \dots)$



$$A \equiv \begin{pmatrix} A_1 \\ A_2 \\ \dots \\ A_m \end{pmatrix}$$

$$A \equiv (A^1 | A^2 | \dots | A^n)$$

## Matrices: main concepts

The set  $R^{m,n} = \mathcal{M}_{m,n}(R)$  of all matrices of order  $m \times n$  is a linear space with two binary operations: **sum and scalar multiplication**.

- The **sum**  $C = A + B$  of two matrices is allowed only if they share the order (same number of rows and columns). It holds

$$c_{i,j} = a_{i,j} + b_{i,j}, \quad \forall i = 1, \dots, m, \quad j = 1, \dots, n$$

That is

$$C = A + B = (a_{ij}) + (b_{ij}) = (a_{ij} + b_{ij}).$$

- If  $A$  is a matrix and  $\alpha \in R$  is a scalar **scalar multiplication**, matrix  $C = \alpha \cdot A$  is a matrix of the same order of  $A$  with entries

$$C = \alpha \cdot A = \alpha \cdot (a_{ij}) = (\alpha a_{ij}).$$

## Matrices: main concepts

### Example

$$\text{Given } A = \begin{pmatrix} 1 & 2 \\ 3 & 1 \\ 4 & 1 \end{pmatrix}, \quad B = \begin{pmatrix} 0 & 2 \\ 1 & -2 \\ 3 & -2 \end{pmatrix} \quad \text{compute } C = A + B, \quad D = 6 \cdot A$$

- Sum

$$C = A + B = \begin{pmatrix} 1 & 2 \\ 3 & 1 \\ 4 & 1 \end{pmatrix} + \begin{pmatrix} 0 & 2 \\ 1 & -2 \\ 3 & -2 \end{pmatrix} = \begin{pmatrix} 1+0 & 2+2 \\ 3+1 & 1-2 \\ 4+3 & 1-2 \end{pmatrix} = \begin{pmatrix} 1 & 4 \\ 4 & -1 \\ 7 & -1 \end{pmatrix}$$

- scalar multiplication

$$D = 6 \cdot A = 6 \cdot \begin{pmatrix} 1 & 2 \\ 3 & 1 \\ 4 & 1 \end{pmatrix} = \begin{pmatrix} 6 \cdot 1 & 6 \cdot 2 \\ 6 \cdot 3 & 6 \cdot 1 \\ 6 \cdot 4 & 6 \cdot 1 \end{pmatrix} = \begin{pmatrix} 6 & 12 \\ 18 & 6 \\ 24 & 6 \end{pmatrix}.$$

## Matrices: main concepts

- matrix  $\mathbf{0} = \begin{pmatrix} 0 & 0 & \dots & 0 \\ 0 & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 0 \end{pmatrix}$  is called **zero matrix** or **null matrix**.

It is the **neutral element** for the additive operation. It holds  $\mathbf{0} + A = A + \mathbf{0} = A, \quad \forall A \in^{m,n}$ .

- Matrix  $B = (-a_{ij})$  is the opposite matrix of  $A$

- Difference of two matrices  $A - B$  is just sum of  $A$  and the the opposite of  $B$

$$A - B = A + (-B) = (a_{ij} - b_{ij}).$$

$$A = \begin{pmatrix} 1 & 1 \\ 2 & 0 \\ 3 & 5 \end{pmatrix} \quad A^T = \begin{pmatrix} 1 & 2 & 3 \\ 1 & 0 & 5 \end{pmatrix}$$

- matrix  $A^T = (a_{j,i})$  where rows and columns swap roles is called **transpose** of  $A$

## Matrices: main concepts

Structured matrices whose elements values depend on row and column indices

- $A$  is symmetric if  $a_{ij} = a_{ji}$   $A = \begin{pmatrix} 1 & 3 & 4 \\ 3 & 1 & 5 \\ 4 & 5 & 2 \end{pmatrix}$

- $A$  is anti-symmetric or skew-symmetric if  $a_{ij} = -a_{ji}$   $A = \begin{pmatrix} 0 & -3 & -1 \\ 3 & 0 & 5 \\ 1 & -5 & 0 \end{pmatrix}$

- $A$  is diagonal if  $a_{ij} = 0, \quad \forall i \neq j$   $A = \begin{pmatrix} d_1 & 0 & \dots & 0 \\ 0 & d_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & d_n \end{pmatrix}$

## Matrices: main concepts

- $A = I$  is the identity matrix if it is a diagonal matrix with  $a_{ij} = 1 \quad \forall i = j$

$$I_n = \begin{pmatrix} 1 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & 1 \end{pmatrix}$$

- $A = U$  is upper triangular if  $a_{ij} = 0 \quad \forall i > j$

$$U = \begin{pmatrix} u_{11} & u_{12} & \dots & u_{1n} \\ 0 & u_{22} & \dots & u_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & u_{nn} \end{pmatrix}$$

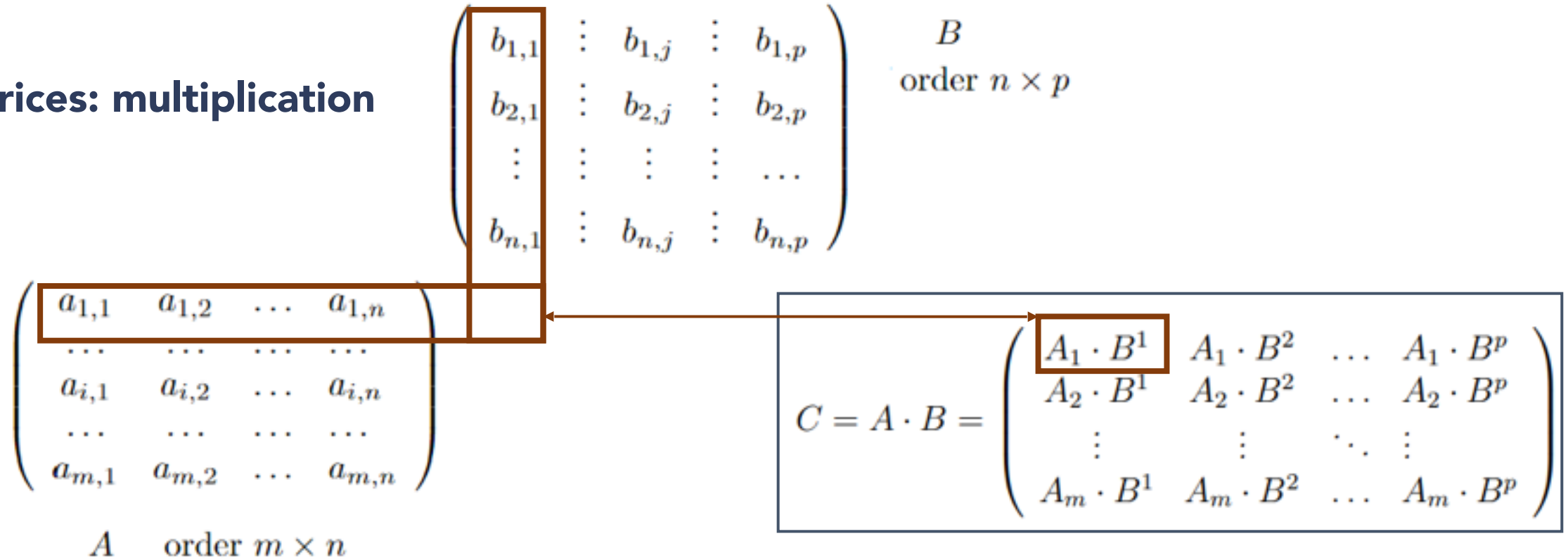
- $A = L$  is lower triangular if  $a_{ij} = 0 \quad \forall i < j$

$$L = \begin{pmatrix} l_{11} & 0 & \dots & 0 \\ l_{21} & l_{22} & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ l_{n1} & l_{n2} & \dots & l_{nn} \end{pmatrix}$$





## Matrices: multiplication

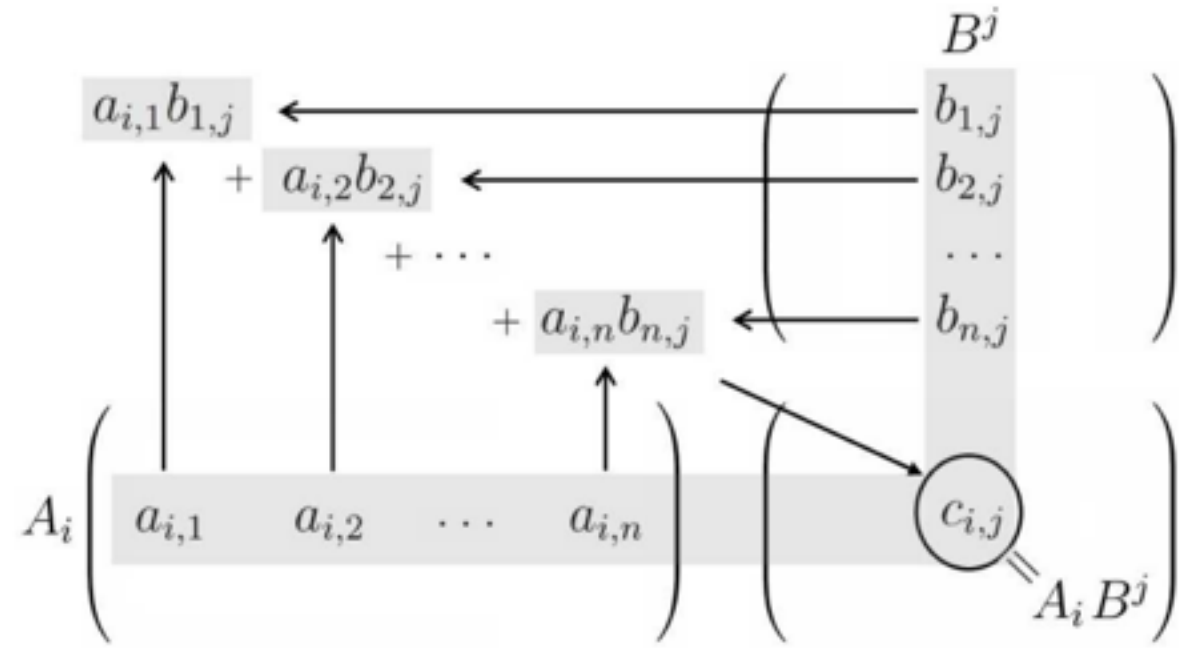


matrix multiplication (matrix-matrix product) is a matrix  $C$  of order  $m \times p$  whose entries are obtained as the dot product of  $A_i$  and  $B^j$ , that is

$$C = (c_{ij}) = (A_i \cdot B^j), \quad \forall i = 1, \dots, m. \quad j = 1, \dots, p$$

# Matrices: multiplication

$$\begin{matrix}
 A_1 \\
 A_2 \\
 A_3
 \end{matrix}
 \begin{bmatrix}
 1 & 2 & 2 \\
 3 & 4 & 0 \\
 5 & -6 & 1
 \end{bmatrix}
 \begin{matrix}
 B^1 & B^2 \\
 \left[ \begin{array}{cc}
 1 & 2 \\
 0 & 1 \\
 1 & 4
 \end{array} \right] \\
 C
 \end{matrix}
 \begin{matrix}
 A_1 B^1 & A_1 B^2 \\
 A_2 B^1 & A_2 B^2 \\
 A_3 B^1 & A_3 B^2
 \end{matrix}
 \quad \checkmark$$



## Matrices: multiplication

$$\begin{matrix}
 A_1 \\
 A_2 \\
 A_3
 \end{matrix}
 \begin{bmatrix}
 1 & 2 & 2 \\
 3 & 4 & 0 \\
 5 & -6 & 1
 \end{bmatrix}
 \begin{matrix}
 B^1 & B^2 \\
 \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} & \begin{bmatrix} 2 \\ 1 \\ 4 \end{bmatrix}
 \end{matrix}
 =
 \begin{matrix}
 A_1 B^1 & A_1 B^2 \\
 A_2 B^1 & A_2 B^2 \\
 A_3 B^1 & A_3 B^2
 \end{matrix}
 =
 \begin{matrix}
 C \\
 \begin{bmatrix} 3 & 12 \\ 3 & 10 \\ 6 & 8 \end{bmatrix}
 \end{matrix}$$

$$c_{1,1} = A_1 \cdot B^1 = \begin{bmatrix} 1 & 2 & 2 \\ A_1 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 0 \\ 1 \\ B^1 \end{bmatrix} = 1 \cdot 1 + 2 \cdot 0 + 2 \cdot 1 = 3, \quad \text{DL E}$$

$$c_{1,2} = A_1 \cdot B^2 = \begin{bmatrix} 1 & 2 & 2 \\ A_1 \end{bmatrix} \cdot \begin{bmatrix} 2 \\ 1 \\ 4 \\ B^2 \end{bmatrix} = 1 \cdot 2 + 2 \cdot 1 + 2 \cdot 4 = 12.$$

$$c_{2,1} = A_2 \cdot B^1 = \begin{bmatrix} 3 & 4 & 0 \\ A_2 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 0 \\ 1 \\ B^1 \end{bmatrix} = 3 \cdot 1 + 4 \cdot 0 + 0 \cdot 1 = 3,$$

$$c_{2,2} = A_2 \cdot B^2 = \begin{bmatrix} 3 & 4 & 0 \\ A_2 \end{bmatrix} \cdot \begin{bmatrix} 2 \\ 1 \\ 4 \\ B^2 \end{bmatrix} = 3 \cdot 2 + 4 \cdot 1 + 0 \cdot 4 = 10.$$

$$c_{3,1} = A_3 \cdot B^1 = \begin{bmatrix} 5 & -6 & 1 \\ A_3 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 0 \\ 1 \\ B^1 \end{bmatrix} = 5 \cdot 1 - 6 \cdot 0 + 1 \cdot 1 = 6,$$

$$c_{3,2} = A_3 \cdot B^2 = \begin{bmatrix} 5 & -6 & 1 \\ A_3 \end{bmatrix} \cdot \begin{bmatrix} 2 \\ 1 \\ 4 \\ B^2 \end{bmatrix} = 5 \cdot 2 - 6 \cdot 1 + 1 \cdot 4 = 8.$$

## Matrices: multiplication (properties)

- (i) **Associativity**  $A \in^{m,n}, B \in^{n,p}, C \in^{p,q}$  then  $(AB)C = A(BC)$ .
- (ii) **Left distributivity**  $A \in^{m,n}, B, C \in^{n,p}$  then  $A(B + C) = AB + AC$ .
- (iii) **Right distributivity**  $A \in^{n,p}, B, C \in^{m,n}$ . then  $(B + C)A = BA + CA$ .
- (iv)  $A \in^{m,n}, B \in^{n,p}, \alpha \in$  then  $\alpha(AB) = (\alpha A)B = A(\alpha B)$ .
- (v) **Neutral element**  $I_n$  and  $I_m$  identity matrices (of order  $n$  and  $m$ ),  $A \in^{m,n}$  then  $I_m A = A I_n = A$
- (vi) **Transpose multiplication**  $A \in^{m,n}$  and  $B \in^{n,p} \implies (AB)^T = B^T A^T$ .

## Matrices: matrix-vector multiplication as a linear combination

Matrix  $A$  by column vector  $x$  produces a column  $Ax$  that is the linear combination of columns of  $A$  with components of  $x$  as coefficients

$$\begin{pmatrix} A^1 & A^2 & \dots & A^k \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \dots \\ x_k \end{pmatrix} = x_1 \begin{pmatrix} A^1 \end{pmatrix} + x_2 \begin{pmatrix} A^2 \end{pmatrix} + \dots + x_k \begin{pmatrix} A^k \end{pmatrix}$$

$$A \quad x \quad = \quad x_1 A^1 + x_2 A^2 + \dots + x_k A^k$$

$$A = \begin{pmatrix} 1 & 4 & \sqrt{3} \\ 3 & 1 & 19 \end{pmatrix}, \quad x = \begin{pmatrix} 2 \\ 1 \\ 0 \end{pmatrix}$$

$$Ax = \begin{pmatrix} 1 & 4 & \sqrt{3} \\ 3 & 1 & 19 \end{pmatrix} \cdot \begin{pmatrix} 2 \\ 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 6 \\ 7 \end{pmatrix}$$

$$= 2 \cdot \begin{pmatrix} 1 \\ 3 \end{pmatrix} + 1 \cdot \begin{pmatrix} 4 \\ 1 \end{pmatrix} + 0 \cdot \begin{pmatrix} \sqrt{3} \\ 19 \end{pmatrix}$$

$$= x_1 A^1 + x_2 A^2 + x_3 A^3$$

## Matrices: vector – matrix multiplication as a linear combination

Row vector  $x$  by matrix  $A$  produces a row  $xA$  that is the linear combination of rows of  $A$  with components of  $x$  as coefficients

$$\begin{pmatrix} x_1 & x_2 & \cdots & x_k \end{pmatrix} \begin{pmatrix} A_1 \\ A_2 \\ \cdots \\ A_k \end{pmatrix} = \begin{matrix} + x_1 \begin{pmatrix} A_1 \end{pmatrix} \\ + x_2 \begin{pmatrix} A_2 \end{pmatrix} \\ + \cdots \\ + x_k \begin{pmatrix} A_k \end{pmatrix} \end{matrix}$$

$$x A = x_1 A_1 + x_2 A_2 + \cdots + x_k A_k$$

$$\begin{aligned} & \begin{matrix} x & A \\ xA = \begin{pmatrix} 1 & 2 & 0 \end{pmatrix} \cdot \begin{pmatrix} 1 & 0 & 5 & 2 \\ 1 & 4 & 4 & 0 \\ 5 & 5 & 2 & 3 \end{pmatrix} = \\ & = \begin{pmatrix} 3 & 8 & 13 & 2 \end{pmatrix} \\ & = \begin{matrix} 1 \cdot \begin{pmatrix} 1 & 0 & 5 & 2 \end{pmatrix} & x_1 A_1 \\ + & + \\ 2 \cdot \begin{pmatrix} 1 & 4 & 4 & 0 \end{pmatrix} & = x_2 A_2 \\ + & + \\ 0 \cdot \begin{pmatrix} 5 & 5 & 2 & 3 \end{pmatrix} & x_3 A_3. \end{matrix} \end{matrix} \end{aligned}$$

## Matrices: inverse matrix

- A square matrix of order  $n$ .  $B = A^{-1}$  is called inverse of  $A$  if  $BA = AB = I_n$ .
- inverse is denoted by  $inv(A)$  Alert: not all matrices have an inverse
- matrix

$A = (a_{ij})$  of order 2 is invertible iff  $a_{11}a_{22} - a_{12}a_{21} \neq 0$ . In this case it is

$$A^{-1} = \frac{1}{a_{11}a_{22} - a_{12}a_{21}} \begin{pmatrix} a_{22} & -a_{12} \\ -a_{21} & a_{11} \end{pmatrix}$$

$$A = \begin{bmatrix} 9 & -4 \\ 3 & 2 \end{bmatrix} \quad a_{11}a_{22} - a_{12}a_{21} = 9 \cdot 2 - (-4) \cdot 3 = 30 \neq 0.$$

$$A^{-1} = \frac{1}{9 \cdot 2 - (-4) \cdot 3} \begin{bmatrix} 2 & 4 \\ -3 & 9 \end{bmatrix} = \frac{1}{30} \begin{bmatrix} 2 & 4 \\ -3 & 9 \end{bmatrix} = \begin{bmatrix} \frac{1}{15} & \frac{2}{15} \\ -\frac{1}{10} & \frac{3}{10} \end{bmatrix}$$



## Matrices: inverse matrix (properties)

- (i) if  $A^{-1}$  is the inverse of  $A$ , then  $A$  is the inverse of  $A^{-1}$ .
- (ii) If  $A$  and  $B$  are invertible  $AB$  is also invertible and  $(AB)^{-1} = B^{-1}A^{-1}$
- (iii) If  $A$  is invertible then also  $A^T$  is invertible and  $(A^T)^{-1} = (A^{-1})^T$
- (iv) if  $A = \text{diag}(d_1, d_2, \dots, d_n)$  is diagonal, then it is invertible iff  $d_1, d_2, \dots, d_n \neq 0$ .

In this case  $A^{-1} = \text{diag}(d_1^{-1}, d_2^{-1}, \dots, d_n^{-1})$ .

$$A = \begin{pmatrix} \frac{2}{3} & 0 & 0 & 0 \\ 0 & -\frac{1}{2} & 0 & 0 \\ 0 & 0 & \frac{4}{\pi} & 0 \\ 0 & 0 & 0 & \frac{3}{e} \end{pmatrix} \quad A^{-1} = \begin{pmatrix} \frac{3}{2} & 0 & 0 & 0 \\ 0 & -2 & 0 & 0 \\ 0 & 0 & \frac{\pi}{4} & 0 \\ 0 & 0 & 0 & \frac{e}{3} \end{pmatrix}$$

## Matrices: determinant (small order $n=1, n=2, n=3$ )

Symbols  $\det(A), \det A, |A|$ .

Case  $n = 1, A = (a_{11}) \Rightarrow \det A \stackrel{\text{def}}{=} a_{11}$ .

Case  $n = 2, A = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \Rightarrow \det A \stackrel{\text{def}}{=} a_{11}a_{22} - a_{12}a_{21}$ .

Case  $n = 3$

$A \quad A^1 \quad A^2$

+

$a_{11}$	$a_{12}$	$a_{13}$	$a_{11}$	$a_{12}$
$a_{21}$	$a_{22}$	$a_{23}$	$a_{21}$	$a_{22}$
$a_{31}$	$a_{32}$	$a_{33}$	$a_{31}$	$a_{32}$

↓ ↓ ↓

$A \quad A^1 \quad A^2$

-

$a_{11}$	$a_{12}$	$a_{13}$	$a_{11}$	$a_{12}$
$a_{21}$	$a_{22}$	$a_{23}$	$a_{21}$	$a_{22}$
$a_{31}$	$a_{32}$	$a_{33}$	$a_{31}$	$a_{32}$

↓ ↓ ↓

$\det A \stackrel{\text{def}}{=} a_{11}a_{22}a_{33} + a_{12}a_{23}a_{31} + a_{13}a_{21}a_{32} - a_{13}a_{22}a_{31} - a_{11}a_{23}a_{32} - a_{12}a_{21}a_{33}$

## Matrices: determinant (Laplace expansion, recursive definition)

Consider submatrices obtained by removing 1 row and 1 column

$$A_{ij} = \begin{pmatrix} a_{1,1} & \dots & a_{1,j-1} & * & a_{1,j+1} & \dots & a_{1,n} \\ \dots & \dots & \dots & * & \dots & \dots & \dots \\ a_{i-1,1} & \dots & a_{i-1,j-1} & * & a_{i-1,j+1} & \dots & a_{i-1,n} \\ * & * & * & * & * & * & * \\ a_{i+1,1} & \dots & a_{i+1,j-1} & * & a_{i+1,j+1} & \dots & a_{i+1,n} \\ \dots & \dots & \dots & * & \dots & \dots & \dots \\ a_{n,1} & \dots & a_{n,j-1} & * & a_{n,j+1} & \dots & a_{n,n} \end{pmatrix}$$

$$\det A \stackrel{\text{def}}{=} a_{i1}C_{i1} + a_{i2}C_{i2} + \dots + a_{in}C_{in} = \sum_{j=1}^n a_{ij}C_{ij}.$$

or

$$\det A \stackrel{\text{def}}{=} a_{1j}C_{1j} + a_{2j}C_{2j} + \dots + a_{nj}C_{nj} = \sum_{i=1}^n a_{ij}C_{ij}.$$

where

$$C_{ij} = (-1)^{i+j} \det A_{ij}.$$

## Matrices: determinant (properties)

$A = (A^1 | A^2 | \dots | A^n)$  square of order  $n$ .

(i)  $\alpha \in A^k$  column of  $A$  then

$$\det (A^1 | \dots | A^{k-1} | \alpha A^k | A^{k+1} | \dots | A^n) = \alpha \det A.$$

(iii) If  $A$  has two equal columns  $\det A = 0$ .

(iv)  $\det I_n = 1$ .

(v) If  $A$  has a zero column  $\det A = 0$ .

(vi) If  $A$  has two parallel columns  $\det A = 0$ .

(vii) if some columns are linearly dependent  $\det A = 0$ .

(same properties for rows)

## Matrices: determinant (properties)

$A = (A^1 | A^2 | \dots | A^n)$  square of order  $n$ .

(viii) Swapping columns produces change in sign of the determinant

(ix) Adding to a column a multiple of another one does not change the determinant

(x) If  $A$  is diagonal or triangular the determinant is the product of diagonal elements:

$$\det A = a_{11}a_{22} \cdots a_{nn}.$$

(xi)  $\det A = 0$  iff columns are linearly dependent

(xii)  $\det(AB) = \det A \det B$ .

(same properties for rows)

## Lesson 1: Basic scientific computing tools

1. Vectors (arrays)
2. Matrices
3. **Linear systems**

## Linear systems: a linear equation ...

$b$  is the constant.

- A linear equation has the form

$$a_1x_1 + a_2x_2 + \cdots + a_nx_n = b$$

$a_1, a_2, \dots$  are called coefficients

$x_i$  values are called  
unknowns (variables)

- a solution in tuple which verifies equation

$$x_1 + 3x_2 - x_5 = 2$$

Some solution vectors

$$(1, 1, 27, 53, 2), \quad (2, 3, \pi, 3, 9), \quad (0, 3, -15\pi, 0, 7)$$

## Linear systems: more linear equations, i.e linear systems

Consider  $m \geq 1$  linear equations sharing same  $n \geq 1$  unknowns  $x_1, x_2, \dots, x_n$ :

$$a_{i1}x_1 + a_{i2}x_2 + \dots + a_{in}x_n = b_i \quad (i = 1, 2, \dots, m)$$

$a_{ij}$  coefficient of  $x_j$  in equation  $i$ ,                       $b_i$  constant of equation  $i$

- a linear system of  $m$  equations in  $n$  unknowns has the form

$$\Sigma : \begin{cases} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1 \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2 \\ \dots & \dots & \dots & \dots & \dots \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n = b_m \end{cases}$$

- a vector  $(\bar{x}_1, \bar{x}_2, \dots, \bar{x}_n)$  is called solution of  $\Sigma$  iff is a solution of all equations



## Linear systems: matrix form

- Let observe that if we denote by  $A = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ \dots & \dots & \dots & \dots \\ a_{i1} & a_{i2} & \dots & a_{in} \\ \dots & \dots & \dots & \dots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix}$   $x = \begin{pmatrix} x_1 \\ x_2 \\ \dots \\ x_n \end{pmatrix}$   $b = \begin{pmatrix} b_1 \\ \dots \\ b_i \\ \dots \\ b_m \end{pmatrix}$

it is  $Ax = b.$

- please note that to solve the system means to find coefficients used to express  $b$  as a linear combination of columns of  $A$

## Linear systems: linear systems with a unique solution

- if  $A$  is square  $\Sigma$  is said to be square. Rectangular otherwise.
- if  $A$  is square the system is said to be
  - singular if  $\det A = 0$
  - nonsingular if  $\det A \neq 0$ ,
- Theorem (Cramer): if  $A$  is nonsingular, linear systems  $Ax = b$  will have a unique solution.
- methods for computing solution should take in account the structure of  $A$
- We show three methods: one for diagonal systems, one for triangular systems and one for "full" systems

## Linear systems: DIAGONAL SYSTEMS

What linear systems are fast to solve?

### DIAGONAL SYSTEMS

$$\begin{aligned} a_{11}x_1 &= b_1 \\ a_{22}x_2 &= b_2 \\ a_{33}x_3 &= b_3 \\ a_{44}x_4 &= b_4 \end{aligned}$$

$$\begin{pmatrix} a_{11} & & & \\ & a_{22} & & \\ & & a_{33} & \\ & & & a_{44} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{pmatrix}$$

Immediate  
solution



$$\begin{aligned} x_1 &= \frac{b_1}{a_{11}} \\ x_2 &= \frac{b_2}{a_{22}} \\ x_3 &= \frac{b_3}{a_{33}} \\ x_4 &= \frac{b_4}{a_{44}} \end{aligned}$$

# Linear systems: TRIANGULAR SYSTEMS

## TRIANGULAR SYSTEMS

$$\begin{aligned}
 a_{11}x_1 + a_{12}x_2 + a_{13}x_3 + a_{14}x_4 &= b_1 \\
 a_{22}x_2 + a_{23}x_3 + a_{24}x_4 &= b_2 \\
 a_{33}x_3 + a_{34}x_4 &= b_3 \\
 a_{44}x_4 &= b_4
 \end{aligned}$$



$$\begin{pmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ & a_{22} & a_{23} & a_{24} \\ & & a_{33} & a_{34} \\ & & & a_{44} \end{pmatrix}
 \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix}
 =
 \begin{pmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{pmatrix}$$

1)

$$x_4 = \frac{b_4}{a_{44}}$$

2)

$$x_3 = \frac{b_3 - a_{34}x_4}{a_{33}}$$

3)

$$x_2 = \frac{b_2 - a_{23}x_3 - a_{24}x_4}{a_{22}}$$

4)

$$x_1 = \frac{b_1 - a_{12}x_2 - a_{13}x_3 - a_{14}x_4}{a_{11}}$$

## Linear systems: TRIANGULAR SYSTEMS

IN GENERAL → **Backward Substitution Algorithm (matlab code)**

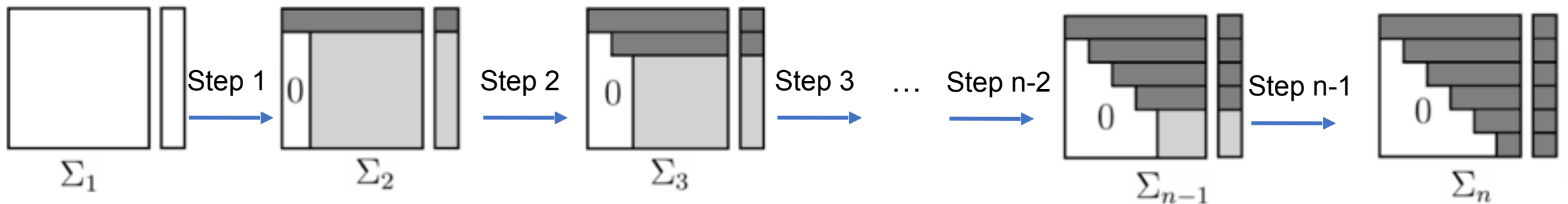
```
x(n) = b(n) / A(n, n)
for i = n-1:-1:1
    x(i) = b(i)
    for j = i+1:n
        x(i) = x(i) - A(i, j) * x(j)
    end
    x(i) = x(i) / A(i, i)
end
```

## Linear systems: FULL SYSTEMS (Gaussian elimination)

Gaussian elimination method transforms a full system of  $n$  equations in a equivalent triangular one in  $n-1$  steps

### Idea

At step  $k$ , to subtract suitable vectors, multiple of equation  $k$ , to all equations at the bottom so that elements in column  $k$  under the diagonal element become 0



How to do this?

## Linear systems: Gaussian Elimination

IN GENERAL → **Gaussian Elimination Algorithm (matlab code)**

```
for k=1:n-1
    for i=k+1:n
        M(i) = A(i,k)/A(k,k)
        for j=k+1:n
            A(i,j) = A(i,j) - M(i)*A(k,j)
        end
        b(i) = b(i) - M(i)*b(k)
    end
end
```



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**Thanks for your attention**  
**..... Digital Tech HPC continues**