

# ECE-GY 5253

# Applied Matrix Theory

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# About This Course

## 1. This is not a Math. Class.

Links to ECE courses:

- **Systems, Control, Robotics, Signal Processing:**  
ECE-GY 6113, 6243, 6253, 7133, 7253, ROB-GYxxxx, ...
- **Power Engineering:**  
ECE-GY 5613, 6603, 6623, 6633, 6653, 6663, ...
- **Communications, Networking, CompE:**  
ECE-GY 5363, 6023, 6033, 6313, 7353, 5483, ...
- **Machine learning and optimization**  
ECE-GY6143 / CS-GY6923, ECE-GY 6233, ...

# About This Course

2. Suitable for both upper-level undergraduates and graduate students from diverse fields of engineering & science:
  - **Electrical engineering** (wireless, control/robotics, communications, networking...)
  - **Mechanical and chemical engineering**
  - **Financial engineering**
  - **CS, Applied mathematics, etc**

# Background Knowledge

This course assumes only elementary knowledge about:

- o Algebra
- o Calculus

# Tips for Getting “A”

- **Well prepared:** Reading the recommended textbook before each class
- **Practice:** exercise, homework
- **Problem-driven:** applications
- **Engaging in teaching:** Ask questions

For advanced topics, consult the recommended textbooks of Horn & Johnson (2013) and of Gantmacher (1960)

# What is a Matrix?

A matrix is a rectangular collection of numbers (J.J. Sylvester, 1848). For example,

A  $m \times n$  matrix is often written as:

$$A = (a_{ij})_{m \times n} = \begin{pmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \cdots & a_{mn} \end{pmatrix}$$

# What is a Matrix?

So, a  $m \times n$  matrix has  $m$  **row vectors**:

$$(a_{i1}, \dots, a_{in}), \quad 1 \leq i \leq m$$

and  $n$  **column vectors** ( $1 \leq j \leq n$ ):

$$\begin{bmatrix} a_{1j} \\ \vdots \\ a_{mj} \end{bmatrix} \doteq \text{col}(a_{1j}, \dots, a_{mj})$$

$$= (a_{1j}, \dots, a_{mj})^T$$

# What is a Matrix?

It can be considered as a **linear mapping**  
from  $\mathbb{R}^n$  to  $\mathbb{R}^m$  (A. Cayley, 1855) :

$$x = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \in \mathbb{R}^n \mapsto y = Ax \in \mathbb{R}^m$$

with  $y_i \doteq \sum_{j=1}^n a_{ij}x_j = a_{i1}x_1 + \cdots + a_{in}x_n$

# Comment on the computation of $A$

Let  $e_i^n = \text{col}(0, \dots, 1, 0, \dots, 0)$ , with 1 as the  $i$ th element, be a vector in the coordinate basis of  $\mathbb{R}^n$ .

Then, for any linear mapping  $F$  (preserving the origin),

$$Fe_i^n = \sum_{j=1}^m a_{ji} e_j^m$$

These coefficients  $a_{ji}$  form the  $m \times n$  matrix

$$A = (a_{ji})_{m \times n}, \text{ associated with } F.$$

# Comment on the computation of $A$

**Proof :** Let  $x = \sum_{i=1}^n x_i e_i^n$ . By linearity,

$$y = Fx = \sum_{i=1}^n x_i F e_i^n = \sum_{i=1}^n x_i \sum_{j=1}^m a_{ji} e_j^m$$

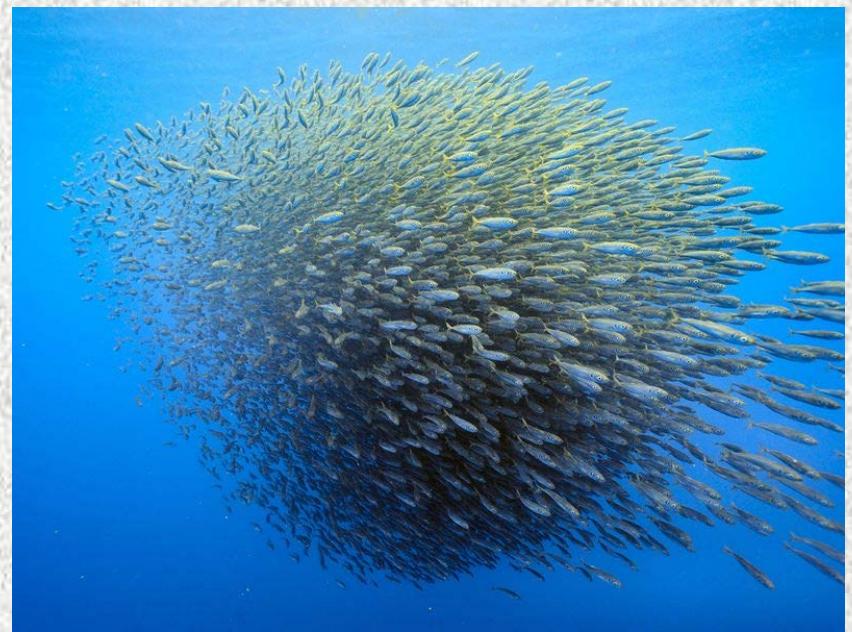
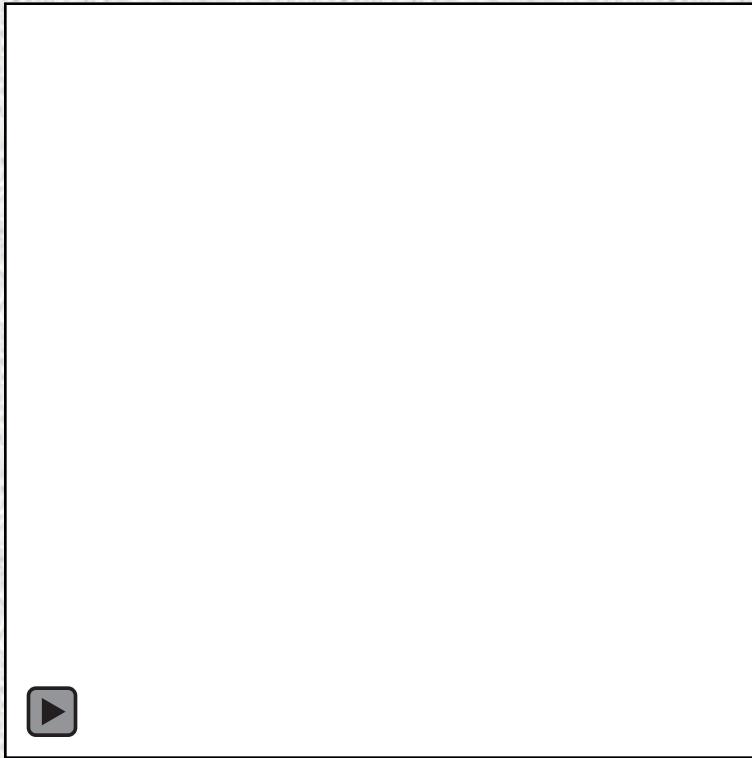
$$= \sum_{j=1}^m \left( \sum_{i=1}^n a_{ji} x_i \right) e_j^m \triangleq \sum_{j=1}^m y_j e_j^m$$

In other words,

$$y = Ax. \quad \text{DONE.}$$

# *A Modern Engineering Example*

We have observed many interesting biological group behaviors:  
Bird flocking, fish schooling, etc



- Why?
- How?

# A Modern Engineering Example

## Hot research topic:

Consider a group of “connected” autonomous agents (say, birds, robots or humans).

- What leads to the desired group behavior?
- Why do local interactions lead to emerging group behavior?
- How to take advantage of it?

Also, see the computer demonstration of *Reynolds's BOIDS model* of bird flocking at: <http://www.red3d.com/cwr/boids/>

# Motivation (cont'd)

**Assume that each agent updates her/his/its heading using the “average” of her/his/its nearby “neighbors”.**

Nearest neighbor rule:

$$\theta_i(t+1) = \frac{1}{1 + n_i(t)} \left( \theta_i(t) + \sum_{j \in N_i} \theta_j(t) \right)$$

$$\theta := \text{col}(\theta_1, \theta_2, \dots, \theta_n)$$

# Motivation (cont'd)

Nearest neighbor rule:

$$\theta_i(t+1) = \frac{1}{1+n_i(t)} \left( \theta_i(t) + \sum_{j \in N_i} \theta_j(t) \right)$$

$$\theta := \text{col}(\theta_1, \theta_2, \dots, \theta_n)$$

Equivalently, in compact matrix notation

$$\theta(t+1) = F_{\sigma(t)} \theta(t), \quad t = 0, 1, 2, \dots$$

$$F_{\sigma(t)} = \left( I + D_{\sigma(t)} \right)^{-1} \left( I + A_{\sigma(t)} \right)$$

# Motivation (cont'd)

It is shown in (Jadbabaie, Lin and Morse, IEEE Transactions on Automat. Control, 2003); also see (Bertsekas-Tsitsiklis, 1986) that, under mild assumptions on graph connectivity,

$$\theta(t) = \text{col}(\theta_1, \theta_2, \dots, \theta_n) \rightarrow \theta_{ss} \text{col}(1, 1, \dots, 1)$$

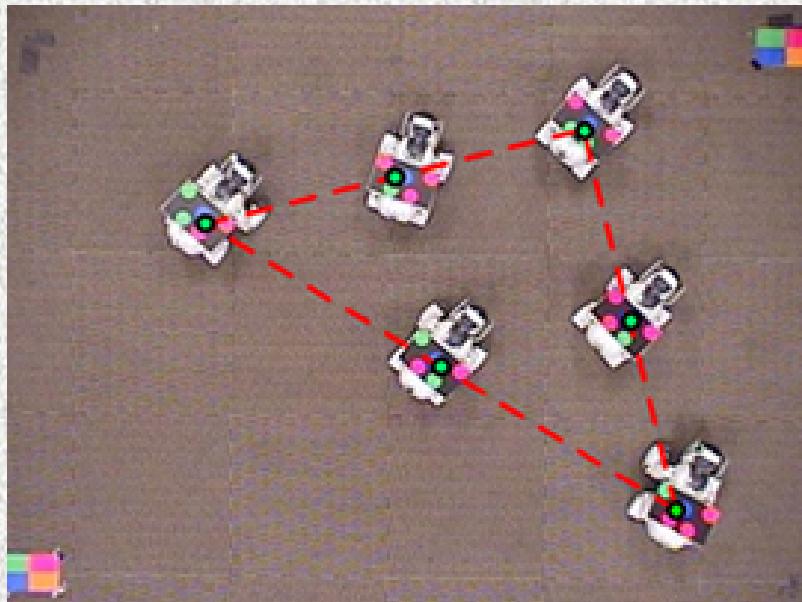
i.e.,

$$\theta_i(t) \rightarrow \theta_{ss}$$

for all  $i$  and all initial conditions  $\theta_i(0)$ !

# Engineering Applications

Coordinated Control of Groups of Unmanned Vehicles:



# Special Types of Matrices

$A \in \mathbb{R}^{m \times n}$  or  $\mathbb{C}^{m \times n}$  is a

- Square matrix, if  $n = m$ .
- Symmetric matrix, if  $n = m$  and  $a_{ij} = a_{ji}$ , or  $A = A^T$ .
- Hermitian matrix, if  $A = A^*$  ( $\doteq \overline{A}^T$ ). 
- Non-square matrix, if  $n \neq m$ .
- Any scalar number is a  $1 \times 1$  matrix.
- A column vector is a  $m \times 1$  matrix.
- A row vector is a  $1 \times n$  matrix.

# Matrix Addition and Multiplication

- 1) Addition of two matrices with the same dimensions:

$$A + B = \left( a_{ij} \right)_{m \times n} + \left( b_{ij} \right)_{m \times n} = \left( a_{ij} + b_{ij} \right)_{m \times n}.$$

- 2) Multiplication of two matrices  $A, B$  with matched dimensions:

$$AB = \left( a_{ij} \right)_{m \times n} \left( b_{ij} \right)_{n \times p} = \left( c_{ij} \right)_{m \times p}$$

with  $c_{ij} = \sum_{k=1}^n a_{ik} b_{kj}.$

**Extension (Semi-tensor product): D. Cheng, 2005**

# Matrix Addition and Multiplication

3) *Associative Property:*

$$(AB)C = A(BC).$$

4) *Distributive Property:*

$$(A + B)C = AC + BC,$$

$$A(B + C) = AB + AC.$$

# Notions about Vectors

- **Linear dependence**

A set of vectors  $\{x_1, x_2, \dots, x_k\}$ , of the same size, is said to be linearly dependent, if  $\exists$  constants  $\{\alpha_j\}_{j=1}^k$ , not all zero, s.t.

$$\alpha_1 x_1 + \alpha_2 x_2 + \dots + \alpha_k x_k = 0.$$

Or, equivalently,  $\exists j \in \{1, \dots, k\}$  such that

$$x_j = \sum_{l \neq j} c_l x_l$$

$$:= c_1 x_1 + \dots + c_{j-1} x_{j-1} + c_{j+1} x_{j+1} + \dots + c_k x_k.$$

# Notions about Vectors

- **Linear independence**

A set of vectors  $\{x_1, x_2, \dots, x_k\}$ , of the same size, is said to be **linearly independent**, if they are **not** linearly dependent.

Or, equivalently,

$$\alpha_1 x_1 + \alpha_2 x_2 + \dots + \alpha_k x_k = 0, \quad \forall \alpha_i$$
$$\Rightarrow \alpha_1 = \dots = \alpha_k = 0.$$

# **Remark**

- 1)  $Ax$  is a linear combination of the columns of  $A$ ,  
with the coordinates of  $x$  as the coefficients.
  
- 2)  $y^T A$  is a linear combination of the rows of  $A$ ,  
with the coordinates of  $y$  as the coefficients.

# Examples

Are the following sets of vectors linearly dependent or independent?

$$1) \left\{ \begin{pmatrix} 1, 0, 0 \end{pmatrix}^T, \begin{pmatrix} 1, 1, 0 \end{pmatrix}^T, \begin{pmatrix} 1, 1, 1 \end{pmatrix}^T \right\}$$

$$2) \{0\}$$

$$3) \{1\}$$

$$4) \left\{ \begin{pmatrix} 1, 1, 1 \end{pmatrix}^T, \begin{pmatrix} 1, 2, 3 \end{pmatrix}^T, \begin{pmatrix} 2, 0, -2 \end{pmatrix}^T \right\}$$

# Notions about Vectors

- **Basis**

Consider a **subspace**  $V$  of  $\mathbb{R}^n$  or  $\mathbb{C}^n$  (being itself a vector space).

A set of vectors  $\{x_1, \dots, x_k\}$  is said to be a **basis** for  $V$ , if :

- 1) They are linearly independent;
- 2) They span  $V$ , i.e.

$$\text{span}\{x_1, \dots, x_k\} = \left\{ a_1x_1 + \dots + a_kx_k : \forall a_j \right\} = V.$$

# Standard Basis

The basis  $\{e_1, e_2, \dots, e_n\}$  is called the **standard basis** of  $\mathbb{R}^n$ , where

$$e_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \quad e_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \quad \dots, \quad e_n = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 1 \end{bmatrix}.$$

# *Comments on Basis*

- 1) In this case, the **dimension** of  $V$  is  $k$ .
- 2)  $k \leq n$ .
- 3) A subspace has an **infinite** number of bases.  
Nonetheless, all bases must be composed of  
the **same number**,  $k$ , of vectors.

# Remark

It is worth noting that *some* vector spaces may be **infinite-dimensional**, i.e. there does not exist any basis consisting of a finite number of elements. Such an example is

$$V = \{ \text{all continuous functions } f : [0,1] \rightarrow \mathbb{R} \}.$$

Note: such a space occurs in several optimization problems!

# Illustrative Example

Consider the following subspace of  $\mathbb{R}^3$ :

$$V = \left\{ \begin{pmatrix} x_1, x_2, 0 \end{pmatrix}^T : \forall x_1, x_2 \in \mathbb{R} \right\}.$$

Examples of a basis for  $V$  include:

- 1)  $\left\{ \begin{pmatrix} 1, 0, 0 \end{pmatrix}^T, \begin{pmatrix} 0, 1, 0 \end{pmatrix}^T \right\};$
- 2)  $\left\{ \begin{pmatrix} 1, 1, 0 \end{pmatrix}^T, \begin{pmatrix} 0, 1, 0 \end{pmatrix}^T \right\};$
- 3)  $\left\{ \begin{pmatrix} 1, 2, 0 \end{pmatrix}^T, \begin{pmatrix} 3, 1, 0 \end{pmatrix}^T \right\}.$

# 1<sup>st</sup> Application: Solving Linear Equations

Solving  $n$  equations for  $n$  unknowns  $x_i$ ,  $i = 1, 2, \dots, n$ :

$$\left. \begin{array}{l} a_{11}x_1 + \cdots + a_{1n}x_n = b_1 \\ a_{21}x_1 + \cdots + a_{2n}x_n = b_2 \\ \vdots \\ a_{n1}x_1 + \cdots + a_{nn}x_n = b_n \end{array} \right\} Ax = b.$$

When does a solution exist? When unique?

# Special Case 1: $n=1$

In this case, the equation becomes

$$a_{11}x_1 = b_1$$

Clearly,

$$a_{11} \neq 0 \Rightarrow x_1 = \frac{b_1}{a_{11}}, \text{ unique.}$$

However, when  $a_{11} = b_1 = 0$ ,

an infinite number of solutions exist!

## Special Case 2: $n=2$

In this case, the equation becomes

$$a_{11}x_1 + a_{12}x_2 = b_1$$

$$a_{21}x_1 + a_{22}x_2 = b_2.$$

Solving for  $x_1$  and  $x_2$  gives

$$(a_{11}a_{22} - a_{12}a_{21})x_1 = b_1a_{22} - b_2a_{12}$$

$$(a_{11}a_{22} - a_{12}a_{21})x_2 = a_{11}b_2 - a_{21}b_1$$

# Special Case 2: $n=2$

Denote  $a_{11}a_{22} - a_{12}a_{21} \doteq \det \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}$ .

When  $\det A \neq 0$ , the equation has the **unique** solution:

$$x_1 = \det \begin{pmatrix} b_1 & a_{12} \\ b_2 & a_{22} \end{pmatrix} / \det \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}$$

$$x_2 = \det \begin{pmatrix} a_{11} & b_1 \\ a_{21} & b_2 \end{pmatrix} / \det \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}$$

Cramer's Rule,  
1750

# Exercise

Solve

$$\begin{pmatrix} 1 & 2 \\ 3 & 7 \end{pmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 4 \\ 9 \end{bmatrix}.$$

# Determinant

Given any square matrix  $A = \left( a_{ij} \right)_{n \times n}$ ,

its **determinant** is defined by

$$\det A \triangleq \sum_{(j_1, \dots, j_n)} s(j_1, \dots, j_n) a_{1j_1} a_{2j_2} \cdots a_{nj_n}$$

where  $(j_1, \dots, j_n)$  is one of the  $n!$  permutations of  $1, \dots, n$ , whose sign is given as

$$s(j_1, \dots, j_n) = sign \prod_{1 \leq p < q \leq n} (j_q - j_p).$$

# Facts about Permutations

**Fact 1:** If two numbers in a permutation are inter-changed, the sign of the permutation is reversed.

For example,  $s(3,1,4,2) = -s(4,1,3,2)$ .

# Examples

*Case 1:  $n = 2$*

$$s(1, 2) = 1, \quad s(2, 1) = -1.$$

*Case 2:  $n = 3$*

$$s(1, 2, 3) = s(3, 1, 2) = s(2, 3, 1) = 1,$$

$$s(1, 3, 2) = s(2, 1, 3) = s(3, 2, 1) = -1.$$

# Computing a Determinant

$$\begin{aligned} & \det \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \\ &= \sum_{(j)=(j_1 j_2)} s(j) a_{1j_1} a_{2j_2} \\ &= s(1\ 2) a_{11} a_{22} + s(2\ 1) a_{12} a_{21} \\ &= a_{11} a_{22} - a_{12} a_{21}. \end{aligned}$$

# Facts about Permutations

**Fact 2:** Let the permutation  $j_1, \dots, j_n$  be formed from  $1, 2, \dots, n$  by  $k$  successive inter-changes of pairs of numbers.

$$\text{Then, } s(j_1, \dots, j_n) = (-1)^k.$$

The permutation  $j$  is **even**, if  $k$  is even.  
Otherwise, it is called an **odd permutation**.

# A Puzzle?

Is it possible to rearrange the letters of the alphabet a, b, ..., z in reverse order z, y, ..., a by exactly 100 successive interchanges of pairs of letters?

# Facts about Determinants

**Fact 1:** If  $A$  is a square matrix,  $\det A = \det A^T$ .

**Proof.** It follows directly from the definition of determinant.  $\square$

# An Application of Fact 1

For any given  $m \times n$  matrix  $A$ , the "**row rank** of  $A$ " is equal to its "**column rank**".

# Facts about Determinants

**Fact 2:** If two rows (or columns) of a square matrix  $A$  are interchanged, the sign of the determinant is reversed.

# Examples

$$\det \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} = - \det \begin{pmatrix} a_{21} & a_{22} \\ a_{11} & a_{12} \end{pmatrix}$$

$$\det \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} = - \det \begin{pmatrix} a_{12} & a_{11} \\ a_{22} & a_{21} \end{pmatrix}$$

# Exercise

What is the determinant of

$$\begin{pmatrix} 3 & 2 & 1 \\ 7 & 5 & 4 \\ 3 & 2 & 1 \end{pmatrix}$$

# Proof of Fact 2

Let  $B$  be generated by inter-changing the rows  $r$  and  $s$  of  $A$ . That is,  $b_{rj} = a_{sj}$ ,  $b_{sj} = a_{rj}$ ,  $b_{ij} = a_{ij}$  if  $i \neq r, s$ .

By definition,

$$\det B = \sum_{(j)} s(j) b_{1j_1} \dots \color{blue}{b_{rj_r}} \dots \color{red}{b_{sj_s}} \dots b_{nj_n}$$

Let  $k$  be the permutation produced from  $j$  by interchanging  $j_r$  and  $j_s$ . Then,  $s(k) = -s(j)$ .

$$\begin{aligned} \text{Thus, } \det B &= \sum_{(j)} -s(k) b_{1j_1} \dots \color{red}{b_{sj_s}} \dots \color{blue}{b_{rj_r}} \dots b_{nj_n} \\ &= -\sum_{(k)} s(k) a_{1j_1} \dots \color{red}{a_{rj_s}} \dots \color{blue}{a_{sj_r}} \dots a_{nj_n} = -\det A. \end{aligned}$$

# An Implication of Fact 2

If a square matrix has two **identical** rows (or columns), then the determinant must be zero.

# Facts about Determinants

**Fact 3:** If a row (or column) of a square matrix is multiplied by a constant  $c$ , the determinant is also multiplied by  $c$ .

**Fact 4:** If a multiple of one row (or column) is subtracted from another row (or column) of a square matrix, the determinant is unchanged.

# Exercise

Show that

$$\det \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} = 0.$$

# Proof of Fact 4

Because of Fact 1, we only need to prove the row part.

Let  $\tilde{A}$  be the new matrix produced from subtracting row  $s$  by  $\lambda$  times row  $r$ . So,

$$\begin{aligned}\det \tilde{A} &= \sum_{(j)} s(j) a_{1j_1} \dots \color{blue}{a_{rj_r}} \dots \left( a_{sj_s} - \lambda \color{blue}{a_{rj_r}} \right) \dots a_{nj_n} \\ &= \sum_{(j)} s(j) a_{1j_1} \dots \color{blue}{a_{rj_r}} \dots a_{sj_s} \dots a_{nj_n} \\ &\quad - \lambda \sum_{(j)} s(j) a_{1j_1} \dots \color{blue}{a_{rj_r}} \dots \color{blue}{a_{rj_r}} \dots a_{nj_n} \\ &= \det A - 0 := \det A.\end{aligned}$$

# Facts about Determinants

**Fact 5:** For any "upper-triangular" square matrix

$$A = \left( a_{ij} \right)_{n \times n}, \text{ i.e., } a_{ij} = 0 \text{ for } i > j,$$

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

**Fact 6:** For any "lower-triangular" square matrix

$$A = \left( a_{ij} \right)_{n \times n}, \text{ i.e., } a_{ij} = 0 \text{ for } i < j,$$

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

# Facts about Determinants

**Fact 7 (Row expansion):** Consider any row  $i$  of a matrix  $A = (a_{ij})_{n \times n}$ . Then,

$$\det A = c_{i1}a_{i1} + c_{i2}a_{i2} + \cdots + c_{in}a_{in}$$

where  $c_{ik} = (-1)^{i+k} \det A_{ik}$ ,

$A_{ik}$  =  $(n - 1) \times (n - 1)$  matrix formed by deleting row  $i$  and column  $k$  from  $A$ .

**Remark:** Because of Fact 1, the same can be stated for “column expansion”.

# An Exercise

Give a simple expression for *Vandermonde's Determinant*:

$$V_n(x_1, \dots, x_n) = \det \begin{bmatrix} 1 & x_1 & x_1^2 & \cdots & x_1^{n-1} \\ 1 & x_2 & x_2^2 & \cdots & x_2^{n-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_n & x_n^2 & \cdots & x_n^{n-1} \end{bmatrix}$$

# Solution

$$V_n(x_1, \dots, x_n) = \prod_{i>j} (x_i - x_j).$$

Idea of Proof :

If  $x_i = x_j$  for some  $i \neq j$ , then the above is obvious.

Otherwise,  $x_i$ ,  $i = 1, \dots, n$ , are distinct.

Then,  $V_n$  is a polynomial of degree  $n-1$  in  $x = x_n$ ,  
with distinct roots  $x_1, x_2, \dots, x_{n-1}$ . That is,

$$V_n = \alpha (x - x_1) \cdots (x - x_{n-1}),$$

$$\text{with } \alpha \doteq \begin{cases} \text{the coefficient of } x^{n-1} = x_n^{n-1}, \text{ i.e.} \\ V_{n-1}(x_1, \dots, x_{n-1}). \end{cases}$$

Then, the identity follows by induction.

# RREF: Row-reduced Echelon Form

This is a canonical form, useful for solving the system of linear equations  $Ax=b$ .

Indeed, it suffices to bring the augmented matrix  $[A, b]$  down to a RREF.

# What is a RREF?

Any RREF must meet the following requirements:

- 1) Each nonzero row has 1 as its first nonzero entry;
- 2) All other entries in the column of such a leading 1 are equal to 0;
- 3) Any rows consisting entirely of zeroes occur at the bottom of the matrix;
- 4) The leading 1's occur in a "stairstep" pattern, left to right.

# An Example of RREF

$$\left( \begin{array}{ccccc} 1 & -1 & 0 & 0 & 3 \\ 0 & 0 & 1 & 0 & -1 \\ 0 & 0 & 0 & 1 & 5 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right)$$

# An Example of RREF

$$\left( \begin{array}{ccccc} 1 & -1 & 0 & 0 & 3 \\ 0 & 0 & 1 & 0 & -1 \\ 0 & 0 & 0 & 1 & 5 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right) \quad \underbrace{\qquad\qquad\qquad}_{A} \quad \underbrace{\qquad\qquad\qquad}_{b}$$

So, the solutions are:

$$\left\{ \begin{array}{l} x_1 - x_2 = 3, \\ x_3 = -1, \\ x_4 = 5. \end{array} \right.$$

# Conversion to RREF

Any matrix can be converted into a (**unique**) RREF, via the following elementary (row!) operations:

Type 1) Interchange of two rows;

Type 2) Multiplication of a row by a nonzero scalar;

Type 3) Addition of a scalar multiple of one row to another row.

# An Exercise

Give the RREFs of the following matrices:

$$\begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} \text{ and } \begin{pmatrix} 3 & 2 & 1 \\ 1 & 3 & 2 \\ 2 & 1 & 3 \end{pmatrix}.$$

Circulant matrix

Or, Toeplitz matrix

# Homework #1

1. Prove the following identities:

$$(A + B)^T = A^T + B^T,$$

$$(AB)^T = B^T A^T,$$

$$(A_1 A_2 \cdots A_n)^T = A_n^T \cdots A_2^T A_1^T.$$

2. Show that  $AB$  is not necessarily symmetric if  $A$  and  $B$  are symmetric.

# Homework #1

3. If  $A + jB$  is Hermitian,  $A, B$  real, then

$$A^T = A, \quad B^T = -B.$$

4. For any square matrix  $A = \begin{pmatrix} A_1 & * \\ O & A_2 \end{pmatrix}$

with  $A_1, A_2$  two square submatrices,  
show that  $\det A = \det A_1 \cdot \det A_2$ .

# Homework #1

5. (*Optional*) Give a simple expression for

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