# Linear Algebra & Geometry LECTURE 11

- Systems of Linear Equations
- Kronecker-Cappelli Theorem
- Determinant

SYSTEMS OF LINEAR EQUATIONS  
A system of linear equations  
(\*)
$$\begin{cases}
a_{1,1}x_1 + a_{1,2}x_2 + \dots + a_{1,n}x_n = b_1 \\
a_{2,1}x_1 + a_{2,2}x_2 + \dots + a_{2,n}x_n = b_2 \\
\dots \\
a_{m,1}x_1 + a_{m,2}x_2 + \dots + a_{m,n}x_n = b_m
\end{cases}$$
can be represented as a single matrix equation  $AX = B$ , where  $A = [a_{i,j}]$  is called the coefficient matrix,

$$X = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \text{ and } B = \begin{bmatrix} b_1 \\ \vdots \\ b_m \end{bmatrix}. X \text{ and } B \text{ are single-column matrices.}$$

The system of linear equations (\*) can also be represented as a vector equation

$$x_{1}\begin{bmatrix}a_{1,1}\\a_{2,1}\\\vdots\\a_{m,1}\end{bmatrix} + x_{2}\begin{bmatrix}a_{1,2}\\a_{2,2}\\\vdots\\a_{m,2}\end{bmatrix} + \dots + x_{n}\begin{bmatrix}a_{1,n}\\a_{2,n}\\\vdots\\a_{m,n}\end{bmatrix} = \begin{bmatrix}b_{1}\\b_{2}\\\vdots\\b_{m}\\\vdots\\b_{m}\end{bmatrix}$$

Which means we are trying to find coefficients to express B as a linear combination of columns of A. This can only be done if

 $span\{C_1, C_2, ..., C_n\} = span\{C_1, C_2, ..., C_n, B\}.$ 

The matrix with columns  $C_1, C_2, ..., C_n$  and *B* is called *the augmented matrix* of the system of equations and is denoted by [A|B].

#### Theorem. (Kronecker, Cappelli)

A system AX = B of linear equations has a solution iff

r(A) = r([A|B]).

**Proof.** The vector-oriented approach form the previous slide together with properties of the *span* operation is proof enough.

# Remark.

Interchanging equations, multiplying both sides by a non-zero number and adding equations one to another do not affect the set of solutions of a system of equations. EROS are exactly these operations except that they are performed on rows of a matrix rather than on equations. This suggests a strategy for solving a system of equations. Start with a system (\*), represent it as its augmented matrix [A|B], row-reduce the matrix to a row echelon matrix [E|C], translate the matrix to the language of equations. Consider the system

$$\begin{cases} 2x + 4y - z = 11 \\ -4x - 3y + 3z = -20 \\ 2x + 4y + 2z = 2 \end{cases}$$

Its augmented matrix is

$$\begin{bmatrix} 2 & 4 & -1 & 11 \\ -4 & -3 & 3 & -20 \\ 2 & 4 & 2 & 2 \end{bmatrix} \sim (r_3 - r_1, r_2 + 2r_1) \begin{bmatrix} 2 & 4 & -1 & 11 \\ 0 & 5 & 1 & 2 \\ 0 & 0 & 3 & -9 \end{bmatrix}.$$

Clearly, the rank of both A and [A|B] is 3 which means the system is solvable. Let us reduce [A|B] to a row-canonical matrix.

$$\sim (r_1 + \frac{1}{3}r_3, r_2 - \frac{1}{3}r_3, \frac{1}{3}r_3) \begin{bmatrix} 2 & 4 & 0 & 8 \\ 0 & 5 & 0 & 5 \\ 0 & 0 & 1 & -3 \end{bmatrix} \sim \left(r_1 - \frac{4}{5}r_2, \frac{1}{5}r_2\right)$$
$$\begin{bmatrix} 2 & 0 & 0 & 4 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & -3 \end{bmatrix} \sim \left(\frac{1}{2}r_1\right) \begin{bmatrix} 1 & 0 & 0 & 2 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & -3 \end{bmatrix}, \text{ which is the matrix of }$$

x = 2, y = 1 and z = -3.

# **Definition.**

A system of linear equations AX = B is called *homogeneous* iff  $B = \Theta$ .

#### Fact.

Every homogeneous system of linear equations has a solution, namely  $x_1 = 0, x_2 = 0, ..., x_n = 0$ . Any other solution (if there is one) is called a *non-trivial* or *non-zero* solution.

### Theorem.

Let  $AX = \Theta$  be a homogeneous system of *m* linear equations with *n* unknowns. Then the set  $W = \{v \in \mathbb{K}^n | Av = \Theta\}$  of all solutions of the system is a subspace of the vector space  $\mathbb{K}^n$ . Moreover,

 $\dim(W) = n - r(A).$ 

**Proof.** (of the first statement)

Take  $u, v \in W$ . This means that  $Au = \Theta$  and  $Av = \Theta$ . Since matrix multiplication is distributive over addition, we have  $A(u + v) = Au + Av = \Theta + \Theta = \Theta$  i.e.,  $u + v \in W$ .

Similarly, we prove that for every  $k \in \mathbb{K}$  we have  $A(ku) = k(Au) = k\Theta = \Theta$ .

We skip the proof of the second statement. QED

#### Example.

$$\begin{cases} x + y - z = 0\\ 2x - 3y + z = 0\\ x - 4y + 2z = 0 \end{cases} A = \begin{bmatrix} 1 & 1 & -1\\ 2 & -3 & 1\\ 1 & -4 & 2 \end{bmatrix} \sim r_2 - 2r_1, r_3 - r_1 \sim$$

 $\begin{bmatrix} 1 & 1 & -1 \\ 0 & -5 & 3 \\ 0 & -5 & 3 \end{bmatrix} \sim r_3 - r_2 \sim \begin{bmatrix} 1 & 1 & -1 \\ 0 & -5 & 3 \\ 0 & 0 & 0 \end{bmatrix}$ . The rank of the last matrix is 2. Hence the dimension of the solution space is 3 - 2 = 1. We shall find a basis for the space reducing the matrix further. $\begin{bmatrix} 1 & 1 & -1 \\ 0 & -5 & 3 \\ 0 & 0 & 0 \end{bmatrix} \sim \frac{1}{-5}r_2 \sim \begin{bmatrix} 1 & 1 & -1 \\ 0 & 1 & \frac{-3}{5} \\ 0 & 0 & 0 \end{bmatrix} \sim r_1 - r_2 \sim \begin{bmatrix} 1 & 0 & \frac{2}{5} \\ 0 & 1 & \frac{-3}{5} \\ 0 & 0 & 0 \end{bmatrix}$ .

In the language of equations this reads

$$\begin{cases} x + \frac{2}{5}z = 0\\ y - \frac{3}{5}z = 0\\ 0z = 0 \end{cases}$$

The bottom equation really says, "*z* may be anything you like" and the top two say  $x = -\frac{2}{5}z$  and  $y = \frac{3}{5}z$ . Hence every vector (x, y, z)belonging to the solution space looks like  $\left(-\frac{2}{5}z, \frac{3}{5}z, z\right) =$ 

 $z(-\frac{2}{5},\frac{3}{5},1)$  and the set  $\{(-\frac{2}{5},\frac{3}{5},1)\}$  is a one-element basis for the space.

#### Example.

 $\begin{cases} x + ay + az = 1 \\ ax + ay + z = 1 \\ ax + y + az = 1 \\ ax + ay + az = 1 \end{cases}$  Discuss solvability of the system in terms of *a*.

$$\begin{bmatrix} 1 & a & a & 1 \\ a & a & 1 & 1 \\ a & 1 & a & 1 \\ a & a & a & 1 \end{bmatrix} \xrightarrow{r_2 - ar_1}_{r_3 - ar_1} \xrightarrow{r_3 - ar_1}_{r_4 - ar_1} \begin{bmatrix} 1 & a & a & 1 \\ 0 & a - a^2 & 1 - a^2 & 1 - a \\ 0 & a - a^2 & a - a^2 & 1 - a \\ 0 & a - a^2 & a - a^2 & 1 - a \end{bmatrix}$$
. Since for  $a = 1$  everything except the top row becomes nil, it looks like a good idea to split cases

In this case r(A) = r(A|B) = 1. This means the system is solvable. It is reduced to x + y + z = 1 hence, x = 1 - y - z; y and z are free.

Case 2, 
$$a \neq 1$$
.  

$$\begin{bmatrix} 1 & a & a & 1 \\ 0 & a - a^2 & 1 - a^2 & 1 - a \\ 0 & 1 - a^2 & a - a^2 & 1 - a \\ 0 & a - a^2 & a - a^2 & 1 - a \end{bmatrix} \sim divide \ rows \ 2,3, and \ 4 \ by(1 - a) \sim$$

$$\begin{bmatrix} 1 & a & a & 1 \\ 0 & a & 1 + a & 1 \\ 0 & 1 + a & a & 1 \\ 0 & a & a & 1 \end{bmatrix} \sim subtract \ row \ 4 \ from \ other \ rows \sim$$

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & a \\ 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & a & a & 1 \end{bmatrix} \sim r_2 \leftrightarrow r_3 \sim \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & a & a & 1 \end{bmatrix} \sim r_4 - ar_2 - ar_3 \sim \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} r(A) = 3, \ r(A|B) = 4.$$
 The system is inconsistent.

#### Theorem.

Let AX = B be an arbitrary system of linear equations. Let U be the solution set and let  $v_0 \in U$  be a solution to the system.

Then  $U = v_0 + W = \{v_0 + w | w \in W\}$ , where W is the solution space of the corresponding homogeneous system  $AX = \Theta$ .

## **Proof.**

Each vector  $v = v_0 + w$  from  $v_0 + W$  is a solution to AX = B. Indeed,

 $A(v_0 + w) = Av_0 + Aw = B + \Theta = B$ . Hence  $v_0 + W \subseteq U$ Moreover, for every vector  $z \in U$  we may put  $t = z - v_0$  so that  $z = v_0 + t$ . Then

 $At = A(z - v_0) = Az - Av_0 = B - B = \Theta$ which means  $t \in W$ . Hence,  $U \subseteq v_0 + W$ . QED

#### **Illustration.**

(1)  $\{-x + y = 1$  (a system of equation, one equation two unknowns)

**V**<sub>0</sub>

(2)  $\{-x + y = 0$  (the corresponding homogeneous system)  $v_0 - a$  solution of (1)

# **Definition.**

Determinant (det for short) is a function defined on the set of all square matrices ( $n \times n$  matrices, n=1,2,...) over a field K into K. The definition is inductive with respect to n: 1. if n = 1, ( $A = [a_{1,1}]$ ) then det(A) =  $a_{1,1}$ 2. if n > 1

$$\det(A) = \sum_{i=1}^{n} (-1)^{i+1} a_{i,1} \det(A_{i,1})$$

where  $A_{i,j}$  denotes the matrix obtained from A by the removal of row number i and column j. det(A) is also denoted by |A|. The formula is known as *Laplace expansion on column 1*.

Notice that we only use the symbol  $A_{i,j}$  in the case j = 1.

Example.  
1. Find 
$$det \begin{bmatrix} a_{1,1} & a_{1,2} \\ a_{2,1} & a_{2,2} \end{bmatrix}$$
.  
 $det(A) = \sum_{i=1}^{2} (-1)^{i+1} a_{i,1} det(A_{i,1}) = a_{1,1}a_{2,2} - a_{2,1}a_{1,2}$ 

In particular, 
$$\begin{vmatrix} 2 & 3 \\ 1 & -2 \end{vmatrix} = 2 \cdot (-2) - 1 \cdot 3 = -7$$

## Example.

2.  $det \begin{bmatrix} a & b & c \\ p & q & r \\ x & y & z \end{bmatrix} = (-1)^{1+1} a det \begin{bmatrix} q & r \\ y & z \end{bmatrix} + (-1)^{2+1} p det \begin{bmatrix} b & c \\ y & z \end{bmatrix}$  $+ (-1)^{3+1} x det \begin{bmatrix} b & c \\ q & r \end{bmatrix} = a(qz - ry) - p(bz - cy) + x(br - qc) =$ aqz + pyc + xbr - cqx - rya - zbp. The last formula is known as the Sarrus Rule.

**BEWARE !.** It only works for  $3 \times 3$  matrices.

$$\begin{array}{c} + \begin{bmatrix} a & b & c \\ p & q & r \\ + \begin{bmatrix} x & y & z \end{bmatrix} - \\ x & y & z \end{bmatrix} - \\ \begin{array}{c} a & b & c \\ p & q & r \end{array}$$

#### Theorem.

For every 
$$j = 1, 2, ..., n$$
 and for every  $n \times n$  matrix  $A$   
$$det(A) = \sum_{i=1}^{n} (-1)^{i+j} a_{i,j} det(A_{i,j})$$

**Proof.** Omitted.

**Remark.** The theorem says that instead of Laplace expansion on column 1 we can do Laplace expansion on column *j*, for every *j*.

#### Example.

1. Find 
$$det \begin{bmatrix} a_{1,1} & a_{1,2} \\ a_{2,1} & a_{2,2} \end{bmatrix}$$
 by Laplace expansion on column 2.  
$$det(A) = \sum_{i=1}^{2} (-1)^{i+2} a_{i,2} det(A_{i,2}) = -a_{1,2}a_{2,1} + a_{2,2}a_{1,1}$$

**Theorem.** (determinant versus transposition)

For every matrix  $A \det(A) = \det(A^T)$ 

**Proof.** Omitted.

**Remark.** The theorem says (indirectly) that instead of Laplace expansion on columns we can do Laplace expansion on rows.

Example.

1. Find 
$$det \begin{bmatrix} a_{1,1} & a_{1,2} \\ a_{2,1} & a_{2,2} \end{bmatrix}$$
 by Laplace expansion on row 1.  
$$det(A) = \sum_{i=1}^{2} (-1)^{1+i} a_{1,i} det(A_{1,i}) = a_{1,1}a_{2,2} - a_{1,2}a_{2,1}$$

## **Theorem.** (*determinant* versus EROS)

For every matrix A

- 1. If  $A \sim (r_i \leftrightarrow r_j)B$  then  $\det(B) = -\det(A) \ (i \neq j)$
- 2. If  $A \sim (r_i \leftarrow cr_i)B$  then det(B) = cdet(A)
- 3. If  $A \sim (r_i \leftarrow r_i + r_j)B$  and  $i \neq j$  then det(B) = det(A)
- 4. Combining 3 with 2 we get

If  $A \sim (r_i \leftarrow r_i + cr_j)B$  and  $i \neq j$  then det(B) = det(A).

Proof. Omitted.

**Remark.** Thanks to the transposition law the theorem applies also to column rather than row operations.

## Remark.

" $A \sim (r_i \leftarrow cr_i)B$ " means "*B* has been obtained from *A* by replacing  $r_i$  of *A* with  $cr_i$ ".

Theorem. (determinant versus not-quite-matrix-addition)

Suppose  $s \in \{1, 2, ..., n\}$  and A[i, j] = B[i, j] = C[i, j] for every i,j such that  $j \neq s$  and C[i, s] = A[i, s] + B[i, s]. Then det(C) = det(A) + det(B).

## **Proof.**

 $\det \begin{bmatrix} c_{1,1} & \dots & a_{1,s} + b_{1,s} & \dots & c_{1,n} \\ c_{2,1} & \dots & a_{2,s} + b_{2,s} & \dots & c_{2,n} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ c_{n,1} & \dots & a_{m,s} + b_{n,s} & \dots & c_{n,n} \end{bmatrix} \xrightarrow{\text{By Laplace}}_{\text{expansion}} = \xrightarrow{\text{on column s}}_{\text{on column s}}$   $\sum_{i=1}^{n} (-1)^{i+s} (a_{i,s} + b_{i,s}) \det(C_{i,s}) = \sum_{i=1}^{n} (-1)^{i+s} a_{i,s} \det(C_{i,s}) + \sum_{i=1}^{n} (-1)^{i+s} b_{i,s} \det(C_{i,s}) = \det(A) + \det(B).$ 

**Warning.** This is NOT about determinant of the sum of two matrices being equal to the sum of their determinants; **that is not true.** This is about determinant of a matrix whose ONE column is the sum of two vectors.