# MTH 102A - Linear Algebra - 2015-16-II Semester

## Arbind Kumar Lal \*

P. Field

- A field F is a set from which we choose our coefficients and scalars.
- Expected properties are
  - 1) a + b and  $a \times b$  should be defined in it.
  - 2) a + b and  $a \times b$  must be inside the field.
  - 3) Both operations are commutative: a + b = b + a;  $a \times b = b \times a$ .
  - 4) There should be identity elements for both operations. Identity element for + is called 0 and that for  $\times$  is called 1.
  - 5) Inverse for "a w.r.t. +:  $\forall a \in F, \exists b \in F \text{ s.t. } a+b=0$ "

"
$$a \neq 0$$
 w.r.t. ×:  $\forall a \in F \setminus \{0\}, \exists b \in F \text{ s.t. } a \times b = 1.$ "

6) Value of (a + b) + c and a + (b + c) are 'equal'.

Value of  $(a \times b) \times c$  and  $a \times (b \times c)$  are 'equal'.

7)  $\times$  distributes itself over +:  $a \times (b+c) = (a \times b) + (a \times c)$ .

Examples:  $\mathbb{R}$ ,  $\mathbb{C}$ ,  $\mathbb{Q}$ .

Also 
$$\mathbb{Z}_5 = \{0, 1, 2, 3, 4\}$$
 with

 $a + b := (a + b) \pmod{5}$  and  $a \times b := (a \times b) \pmod{5}$ . Here, 3 + 4 = 2,  $4 \times 2 = 3$  and  $4 \times 4 = 1$ .

$$\mathbb{Q}[\sqrt{2}] := \{a + b\sqrt{2} : a, b \in \mathbb{Q}\}.$$

$$(a + b\sqrt{2}) + (c + d\sqrt{2}) = (a + c) + (b + d)\sqrt{2}$$

$$(a + b\sqrt{2}) \cdot (c + d\sqrt{2}) = (ac + 2bd) + (ad + bc)\sqrt{2}$$

Also, whenever  $(a+b\sqrt{2}) \neq 0$ , we have  $(a+b\sqrt{2})^{-1} = \frac{a}{a^2-2b^2} - \frac{b}{a^2-2b^2}\sqrt{2}$  as  $\sqrt{2}$  is irrational.

• A linear equation (over some field  $\mathbb{F}$ ) is an expression of the form

$$a_1x_1 + a_2x_2 + \cdots + a_nx_n = b_1$$

<sup>\*</sup>Indian Institute of Technology Kanpur

• Example:  $2x_1 + 3x_2 + 7x_3 = 6$  is a linear equation. Over what? Over any field. What does 2 mean in  $\mathbb{F}$ ? 1 + 1!

Anyway we never mention that!! Come on, isn't it obvious?

 $\bullet$  A system of linear equations is a collection of m linear equations in the 'same' n variables. It has the form:

$$a_{11}x_1 + \dots + a_{1n}x_n = b_1$$

$$\vdots$$

$$a_{m1}x_1 + \dots + a_{mn}x_n = b_m$$

- $\begin{cases} x+y+z=1\\ 2y+3z=7 \end{cases}$  is a system of 2 linear equations in 3 variables.
- Matrix form of a system (of linear equations):

$$\begin{bmatrix} a_{11} & \cdots & a_{1n} \\ & \vdots & \\ a_{m1} & \cdots & a_{mn} \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ \vdots \\ b_m \end{bmatrix}.$$

• The product AB corresponds to operating on the *columns of the matrix* A, using entries of B. Thus,

$$\begin{bmatrix} a_{11} & \cdots & a_{1n} \\ & \vdots & \\ a_{m1} & \cdots & a_{mn} \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ \vdots \\ b_m \end{bmatrix}.$$

is same as

$$\begin{bmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{bmatrix} x_1 + \begin{bmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{m2} \end{bmatrix} x_2 + \dots + \begin{bmatrix} a_{1n} \\ a_{2n} \\ \vdots \\ a_{mn} \end{bmatrix} x_n = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}.$$

(Similar to 'does there exist integers x, y such that 13x + 5y = 1???")

P. The old story: A new way

Q Solve 
$$2x + y = 1$$
$$x + 2y = -1 .$$

A 
$$2x + y = 1$$
  $2x + y = 1$   $3y = -3$ . So...

• Write the Augmented coefficient matrix of the system with a matrix:  $\begin{bmatrix} 2 & 1 & 1 \\ 1 & 2 & -1 \end{bmatrix}$ 

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• Above solution procedure is nothing but

$$\begin{bmatrix} 2 & 1 & 1 \\ 1 & 2 & -1 \end{bmatrix} \xrightarrow{2(2)} \begin{bmatrix} 2 & 1 & 1 \\ 2 & 4 & -2 \end{bmatrix} \xrightarrow{(2) - (1)} \begin{bmatrix} 2 & 1 & 1 \\ 0 & 3 & -3 \end{bmatrix}. \text{ Continue?}$$

$$\xrightarrow{\frac{1}{3}(2)} \begin{bmatrix} 2 & 1 & 1 \\ 0 & 1 & -1 \end{bmatrix} \xrightarrow{(1) - (2)} \begin{bmatrix} 2 & 0 & 2 \\ 0 & 1 & -1 \end{bmatrix} \xrightarrow{\frac{1}{2}(1)} \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & -1 \end{bmatrix}$$

- System:  $\frac{x+y+z=1}{2y+3z=7}$ . That is,  $\begin{bmatrix} 1 & 1 & 1 \\ 0 & 2 & 3 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 1 \\ 7 \end{bmatrix}$ .
- Augmented Coefficient Matrix:  $\begin{vmatrix} 1 & 1 & 1 & 1 \\ 0 & 2 & 3 & 7 \end{vmatrix}$ . Apply elimination.

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 2 & 3 & 7 \end{bmatrix} \xrightarrow{\frac{1}{2}(2)} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & \frac{3}{2} & \frac{7}{2} \end{bmatrix} \xrightarrow{(1)-(2)} \begin{bmatrix} 1 & 0 & -\frac{1}{2} & -\frac{5}{2} \\ 0 & 1 & \frac{3}{2} & \frac{7}{2} \end{bmatrix}$$

- Highlighted positions are called pivots/leading terms. Corresponding variables are basic variables. Others are free variables.
  - Here z is free. For solution: put z = t; then  $\begin{vmatrix} x \\ y \end{vmatrix} = \begin{vmatrix} -\frac{3}{2} + \frac{1}{2}t \\ \frac{7}{2} \frac{3}{2}t \end{vmatrix}$ .
  - For example, t = 0 gives  $\begin{bmatrix} -\frac{5}{2} \\ \frac{7}{2} \\ 0 \end{bmatrix}$  and t = 1 gives  $\begin{bmatrix} -2 \\ 2 \\ 1 \end{bmatrix}$ .

• In that case we have to provide an algorithm. What is an algorithm?

A 'step by step' instruction to carry out the task.

- Actually, we do not have to!! Gauss-Jordan have already done it!!
- To describe that we need 3 elementary row operations.
- To describe that we need 3 elementary row operations.
   Given the Augmented Coefficient Matrix (ACM) of a system:  $\begin{bmatrix} a_{11} & \cdots & a_{1n} & b_1 \\ & \vdots & & \vdots \\ a_{m1} & \cdots & a_{mn} & b_m \end{bmatrix}.$

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1)  $E_i(\alpha)$  Multiply ith row by  $\alpha \neq 0$ .

For example, 
$$\begin{bmatrix} 0 & 1 & 1 & 1 \\ 0 & 2 & 3 & 1 \\ 1 & 0 & 2 & -3 \end{bmatrix} \xrightarrow{E_2(\frac{1}{2})} \begin{bmatrix} 0 & 1 & 1 & 1 \\ 0 & 1 & \frac{3}{2} & \frac{1}{2} \\ 1 & 0 & 2 & -3 \end{bmatrix}.$$

Result is the same as:  $\begin{bmatrix} 1 & 0 & 0 \\ 0 & \frac{1}{2} & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 0 & 1 & 1 & 1 \\ 0 & 2 & 3 & 1 \\ 1 & 0 & 2 & -3 \end{bmatrix}$ • Elementary Matrix  $E_2(\frac{1}{2})$ 

P. Elementary row operations

 $E_{ij}$  Interchange rows i and j.

For example, 
$$\begin{bmatrix} 0 & 1 & 1 & 1 \\ 0 & 2 & 3 & 1 \\ 1 & 0 & 2 & -3 \\ 1 & 2 & 3 & 0 \\ 5 & 2 & -1 & 2 \end{bmatrix} \xrightarrow{E_{14}} \begin{bmatrix} 1 & 2 & 3 & 0 \\ 0 & 2 & 3 & 1 \\ 1 & 0 & 2 & -3 \\ 0 & 1 & 1 & 1 \\ 5 & 2 & -1 & 2 \end{bmatrix}.$$

Result is the same as:  $\begin{bmatrix} 0 & & 1 & \\ & 1 & & \\ & & 1 & \\ 1 & & & 0 \\ & & & & 1 \end{bmatrix} \begin{bmatrix} 0 & 1 & 1 & 1 \\ 0 & 2 & 3 & 1 \\ 1 & 0 & 2 & -3 \\ 1 & 2 & 3 & 0 \\ 5 & 2 & -1 & 2 \end{bmatrix}$ 

• Elementary Matrix  $E_{14}$ 

 $E_{ij}(\alpha)$  Replace ith row  $R_i$  by  $R_i + \alpha R_j$ 

For example, 
$$\begin{bmatrix} 0 & 1 & 1 & 1 \\ 0 & 2 & 3 & 1 \\ 1 & 0 & 2 & -3 \\ 1 & 2 & 3 & 0 \\ 5 & 2 & -1 & 2 \end{bmatrix} \xrightarrow{E_{35}(-\frac{1}{5})} \begin{bmatrix} 1 & 2 & 3 & 0 \\ 0 & 2 & 3 & 1 \\ 0 & -\frac{2}{5} & \frac{11}{5} & -\frac{17}{5} \\ 0 & 1 & 1 & 1 \\ 5 & 2 & -1 & 2 \end{bmatrix}.$$

Result is the same as:  $\begin{bmatrix} 1 & & & & \\ & 1 & & & \\ & & 1 & & -\frac{1}{5} \\ & & & 1 \end{bmatrix} \begin{bmatrix} 0 & 1 & 1 & 1 \\ 0 & 2 & 3 & 1 \\ 1 & 0 & 2 & -3 \\ 1 & 2 & 3 & 0 \\ 5 & 2 & -1 & 2 \end{bmatrix}$ 

- Elementary Matrix  $E_{35}(-\frac{1}{\$})_{\text{Echelon-form}(\mathbf{EF})}$
- Let A be a matrix. A pivot/leading term is the first (from left) nonzero element of a nonzero row in A. We use  $a_{ij}$  to denote it.

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- A matrix A is in echelon form (EF) (ladder like) if
  - 1) Pivot of the  $i+1{\rm th}$  row comes to the right of the  $i{\rm th}$ .
  - 2) Entries below the pivot in a 'pivotal column' are 0.
  - 3) The zero rows are at the bottom.

• In EF: 
$$\begin{bmatrix} 3 & 1 & 3 & 0 \\ 0 & 2 & 0 & 1 \\ 0 & 0 & 0 & 3 \end{bmatrix}, \begin{bmatrix} 1 & 2 & 0 & 5 \\ 0 & 2 & 0 & 6 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}, \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}.$$
• Not in EF: 
$$\begin{bmatrix} 0 & 3 & 3 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 2 & 0 & 0 \end{bmatrix}$$
 (rule 1,2 fail); 
$$\begin{bmatrix} 0 & 1 & 3 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$
 (rule 3 fails).

P. Row-reduced-echelon-form(RREF)

- A matrix A is in RREF if
  - 1) It is in EF.
- 2) Pivot of each nonzero row is 1.
- 3) Other entries in a 'pivotal column' are 0.

• In RREF: 
$$\begin{bmatrix} 0 & 1 & 3 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}, \begin{bmatrix} 1 & 0 & 0 & 5 \\ 0 & 1 & 0 & 6 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}, \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 6 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}.$$
• Not in RREF: 
$$\begin{bmatrix} 0 & 3 & 3 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 & 3 & 1 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}.$$

• A is row equivalent to B if B is the result of k elementary row operations on A. That is, if there exists some elementary matrices  $E_1, \ldots, E_k$  s.t.

$$E_k E_{k-1} \cdots E_1 A = B.$$

P. Gauss-Jordan elimination(GJE)

GJE An algorithm. Uses row operations only. Input: A.

Output: a matrix B in RREF s.t. B is row equivalent to A.

- 0) Put 'region' = A.
- 1) If all entries in the region are 0, STOP.

Else, in the region, find the leftmost nonzero column and find its topmost nonzero entry.

Suppose it is  $a_{ij}$ . Box it. This is a pivot.

Take it to the top row of the region. Make it 1.

Make other entries of the whole matrix in it's column 0.

2) Put region = the submatrix below and to the right of the current pivot. Go to step 1).

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• The process will stop, as we can get at most  $\min\{m, n\}$  pivots.

• Apply GJE: 
$$\begin{bmatrix} 0 & 2 & 3 & 7 \\ \hline 1 & 1 & 1 & 1 \\ 1 & 3 & 4 & 8 \\ 0 & 0 & 0 & 1 \end{bmatrix} \xrightarrow{E_{12}} \begin{bmatrix} \boxed{1} & 1 & 1 & 1 \\ 0 & 2 & 3 & 7 \\ 1 & 3 & 4 & 8 \\ 0 & 0 & 0 & 1 \end{bmatrix} \xrightarrow{E_{31}(-1)} \begin{bmatrix} \boxed{1} & 1 & 1 & 1 \\ 0 & 2 & 3 & 7 \\ 0 & 0 & 0 & 1 \end{bmatrix} \xrightarrow{E_{2}(\frac{1}{2})} \xrightarrow{E_{2}(\frac{1}{2})} \xrightarrow{E_{31}(-1)} \begin{bmatrix} \boxed{1} & 1 & 1 & 1 \\ 0 & 2 & 3 & 7 \\ 0 & 0 & 0 & 1 \end{bmatrix} \xrightarrow{E_{2}(\frac{1}{2})} \xrightarrow{E_{2}(\frac{1}{2})} \xrightarrow{E_{31}(-1)} \xrightarrow{E_{31}(-1)$$

Th Each matrix  $A_{m \times n}$  is row equivalent to some matrix in RREF.!!

#### P. Gauss elimination

GE It is the same as GJE except that

- 1) pivots need not be made 1 and
- 2) entries above the pivots need not be made 0.

$$\bullet \begin{bmatrix} 1 & 1 & 1 \\ 1 & 2 & 3 \\ 1 & 4 & 9 \end{bmatrix} \xrightarrow{E_{21}(-1), E_{31}(-1)} \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \\ 0 & 3 & 8 \end{bmatrix} \xrightarrow{E_{32}(-3)} \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \\ 0 & 0 & 2 \end{bmatrix}.$$

• GJE may be viewed as an extension of GE.

P. Inverses of elementary matrices

Sometimes it helps to imagine an elementary matrix as an elementary row operation.

Recall Invertibility:  $A_{n\times n}$  is invertible if there is a B s.t. AB = BA = I.

What is the inverse of  $E_i(\alpha)$ ? That is, 'if I have multiplied *i*th row by  $\alpha$ , how do i get back the original'? Must be  $E_i(\frac{1}{\alpha})$ .

What is the inverse of  $E_{ij}$ ? That is, 'if I have interchanged  $R_i$  and  $R_j$ , how do i get back the original'? Must be  $E_{ji} = E_{ij}$ .

What is the inverse of  $E_{ij}(\alpha)$ ? That is, 'if I have added  $\alpha R_j$  to  $R_i$ , how do i get back the original'? Must be  $E_{ij}(-\alpha)$ .

**Ex** Inverse of an elementary matrix is an elementary matrix.

 $\mathbf{E}\mathbf{x}$  If A is invertible, then it has a unique inverse.

• Recall that for any invertible matrix  $A(i,:) \neq \mathbf{0}$ .

Th Let  $A_{n\times n}$  be invertible. Then, the RREF of A is I.

Po. A RREF of A is nothing but  $E_k \cdots E_1 A = EA$  (say), where  $E_i$ 's are elementary.

As EA is invertible, it has no zero row. How many pivots should it have? As EA is in RREF, it must be I.

P. RREF

Th If A is row equivalent to B then, the systems Ax = 0 and Bx = 0 have the same solution set. !!

Q Is 
$$A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$
 row equivalent to  $B = \begin{bmatrix} 1 & 0 & a \\ 0 & 1 & b \\ 0 & 0 & 0 \end{bmatrix}$ ?

No,  $\begin{bmatrix} a \\ b \\ -1 \end{bmatrix}$  is a solution of Bx = 0, not of Ax = 0.

Th Let A and B be two row equivalent matrices in RREF. Then A = B.

#### P. Rank of a Matrix

Cor Each matrix A is row equivalent to a unique matrix in RREF.!!

- We use RREF A to denote this matrix.
- The rank of a matrix A is the number of pivots in RREF A. Notation: rank A.
- Thus, rank  $A_{m \times n} \leq m, n \text{ and } rank(\mathbf{0}) = 0.$

P. Gauss-Jordan say it

The Take a system Ax = b and an invertible matrix B. Then, y is a solution of Ax = b if and only if y is a solution of BAx = Bb.!!

Gauss-Jordan idea Take the ACM [A|b] of a system. Keep on applying elementary row operations. Solution space stays same!!!

STOP at the RREF[A'|b'].

If A'(i,:) = 0 and  $b'_i \neq 0$ , then conclude that the system has no solution (inconsistent). Note that, here,  $\operatorname{rank}(A) < \operatorname{rank}([A|b])$ .

Otherwise, a general solution is obtained by assigning the free-variables arbitrary values and by evaluating the values of the basic variables. Note that, here, rank(A) = rank(A|b|).

P. Very Important Ideas

Th Let A be a matrix of RANK r. Then, there exist(s)

- Invertible P s.t.  $PA = \begin{bmatrix} B \\ \mathbf{0} \end{bmatrix}$ .
- $\bullet$  Invertible P and Q such that

$$P_{m \times m} A_{m \times n} Q_{n \times n} = PAQ == \begin{bmatrix} I_r & \mathbf{0} \\ \mathbf{0} & \mathbf{0} \end{bmatrix}.$$

Ex[Rank Factorization] Let rank  $A_{m\times n}=r$ . Then there exist  $B_{m\times r}, C_{r\times n}$  s.t. rank  $B=\operatorname{rank} C=r$  and A=BC as

$$A = P^{-1} \begin{bmatrix} I_r & \mathbf{0} \\ \mathbf{0} & \mathbf{0} \end{bmatrix} Q^{-1} = \begin{bmatrix} P_1 & P_2 \end{bmatrix} \begin{bmatrix} I_r & \mathbf{0} \\ \mathbf{0} & \mathbf{0} \end{bmatrix} \begin{bmatrix} Q_1 \\ Q_2 \end{bmatrix} = P_1 Q_1.$$

Th Consider the homogeneous linear system  $A\mathbf{x} = \mathbf{0}$ . Then,

- The zero vector,  $\mathbf{0} = (0, \dots, 0)^t$ , is always a solution, called the TRIVIAL solution.
- Suppose  $\mathbf{x}_1, \mathbf{x}_2 \neq \mathbf{0}$  are two solutions of  $A\mathbf{x} = \mathbf{0}$ . Then,  $k_1\mathbf{x}_1 + k_2\mathbf{x}_2$  is also a solution of  $A\mathbf{x} = \mathbf{0}$  for any  $k_1, k_2 \in \mathbb{R}$ .

Th Consider the linear system  $A\mathbf{x} = \mathbf{b}$ , where A is an  $m \times n$  matrix and  $\mathbf{x}^t = (x_1, \dots, x_n)$ . If  $[C \mid \mathbf{d}] = \mathbf{rref}([A \mid \mathbf{b}])$  then,

•  $A\mathbf{x} = \mathbf{b}$  is inconsistent (has no solution) if  $[C \ \mathbf{d}]$  has a row of the form  $[\mathbf{0}^t | 1]$ , where  $\mathbf{0}^t = (0, \dots, 0)$ . That is,

$$\operatorname{Rank}(A) < \operatorname{Rank}([A \mid \mathbf{b}]) = \operatorname{Rank}([C \mid \mathbf{d}]).$$

•  $A\mathbf{x} = \mathbf{b}$  is consistent (has a solution) if  $[C \mid \mathbf{d}]$  has **NO** row of the form  $[\mathbf{0}^t \mid 1]$ . That is,

$$\operatorname{Rank}(A) = \operatorname{Rank}([A \mid \mathbf{b}]) = \operatorname{Rank}([C \mid \mathbf{d}]).$$

Furthermore, (recall n = Number of unknowns implies)

- if  $\mathbf{Rank}(A) = n$  then,  $A\mathbf{x} = \mathbf{b}$  has a unique solution.
- if  $\mathbf{Rank}(A) < n$  then,  $A\mathbf{x} = \mathbf{b}$  has infinite number of solutions.

P. Example

• System x + y + z = 1 y + 3z = 2 -x + 2z = 2

• ACM: 
$$A = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 \\ 0 & 1 & 3 & 2 \\ -1 & 0 & 2 & 2 \end{bmatrix}$$
 RREF $(A) = \begin{bmatrix} 1 & 0 & -2 & 0 \\ 0 & 1 & 3 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$   $\rightarrow$  Matlab command rref(a)

- System is inconsistent (has no solution).
- ACM:  $A = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 3 & 2 \\ -1 & 0 & 2 & 1 \end{bmatrix}$  RREF $(A) = \begin{bmatrix} 1 & 0 & -2 & -1 \\ 0 & 1 & 3 & 2 \\ 0 & 0 & 0 & 0 \end{bmatrix}$
- x, y are basic, z is free. General solution:  $\left\{ \begin{bmatrix} -1 + 2t \\ 2 3t \\ t \end{bmatrix} : t \in \mathbb{R} \right\}$ .
- Let  $[C|\mathbf{d}] = \text{RREF}([A|\mathbf{b}])$ . Assume  $C(i,:) \neq \mathbf{0}$  and the system is consistent. Notice that in C(i,:) all entries are zero, except the pivot and entries corresponding to free variables. Thus, if we assign all free variables 0, then we get  $x_i = d_i$ .
- What happens if we put z = t = 0, in the above example?
- If  $rank(A_{m \times n}) < n$ , then  $A\mathbf{x} = \mathbf{0}$  has a nonzero solution.!!

P. Invertibility and Gauss-Jordan

Let A be a square matrix of order n. Then the following statements are equivalent.

- A is invertible.
- The homogeneous system  $A\mathbf{x} = \mathbf{0}$  has only the trivial solution.
- The (rref) row-reduced echelon form of A is  $I_n$ .
- A is a product of elementary matrices.
- If A is invertible then rref(A) = I. Thus,  $E_k E_{k-1} \cdots E_2 E_1 A = I$  for some elementary matrices. Hence,  $A^{-1} = E_k E_{k-1} \cdots E_2 E_1$ .

matrices. Hence,  $A^{-1} = E_k E_{k-1} \cdots E_2 E_1$ .

Implication. Given a matrix  $A_{n \times n}$ , apply GJE to  $[A|I_n]$ . Then, we get elementary matrices  $E_1, E_2, \ldots, E_k$  such that

$$E_k E_{k-1} \cdots E_1 [A|I] = [E_k E_{k-1} \cdots E_1 A|E_k E_{k-1} \cdots E_1 I] = [I|A^{-1}].$$

P. Inverse - Example

$$\begin{bmatrix} 1 & 2 & 1 & 0 \\ 2 & 1 & 0 & 1 \end{bmatrix} \underbrace{E_{21}(-2)}_{0} \begin{bmatrix} 1 & 2 & 1 & 0 \\ 0 & -3 & -2 & 1 \end{bmatrix} \underbrace{E_{2}(-1/3)}_{0} \begin{bmatrix} 1 & 2 & 1 & 0 \\ 0 & 1 & \frac{2}{3} & \frac{-1}{3} \end{bmatrix}$$

$$\underbrace{E_{12}(-2)}_{0} \begin{bmatrix} 1 & 0 & \frac{-1}{3} & \frac{2}{3} \\ 0 & 1 & \frac{2}{3} & \frac{-1}{3} \end{bmatrix}$$

$$\begin{bmatrix} 1 & 2 & 3 & 1 & 0 & 0 \\ 2 & 4 & 7 & 0 & 1 & 0 \\ 1 & 1 & 1 & 0 & 0 & 1 \end{bmatrix} \underbrace{E_{21}(-2)}_{E_{31}(-1)} \begin{bmatrix} 1 & 2 & 3 & 1 & 0 & 0 \\ 0 & 0 & 1 & -2 & 1 & 0 \\ 0 & -1 & -2 & -1 & 0 & 1 \end{bmatrix}$$

$$[2,4,7,0,1,0] - 2[1,2,3,1,0,0] = [0,0,1,-2,1,0]$$

$$[1,1,1,0,0,1] - [1,2,3,1,0,0] = [0,-1,-2,-1,0,1].$$

$$\xrightarrow{E_{32}} \begin{bmatrix} 1 & 2 & 3 & 1 & 0 & 0 \\ 0 & -1 & -2 & -1 & 0 & 1 \\ 0 & 0 & 1 & -2 & 1 & 0 \end{bmatrix} \xrightarrow{E_2(-1)} \begin{bmatrix} 1 & 0 & -1 & -1 & 0 & 2 \\ 0 & 1 & 2 & 1 & 0 & -1 \\ 0 & 0 & 1 & -2 & 1 & 0 \end{bmatrix}$$

$$\xrightarrow{E_{23}(-2)} \begin{bmatrix} 1 & 0 & 0 & -3 & 1 & 2 \\ 0 & 1 & 0 & 5 & -2 & -1 \\ 0 & 0 & 1 & -2 & 1 & 0 \end{bmatrix}$$

Thus, we have solved three linear systems simultaneously, namely

$$A \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, A \begin{bmatrix} \alpha \\ \beta \\ \gamma \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \text{ and } A \begin{bmatrix} u \\ v \\ w \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}, \text{ where } A = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 4 & 7 \\ 1 & 1 & 1 \end{bmatrix}.$$
$$(x, y, z) = (-3, 5, -2), (\alpha, \beta, \gamma) = (1, -2, 1) \text{ and } (u, v, w) = (2, -1, 0).$$

P. Determinant

Notation:- Let 
$$A = \begin{bmatrix} 1 & 2 & 3 \\ 1 & 3 & 2 \\ 2 & 4 & 7 \end{bmatrix}$$
.  
Then,  $A(1|2) = \begin{bmatrix} 1 & 2 \\ 2 & 7 \end{bmatrix} = \begin{bmatrix} 1 & 2 & 3 \\ 1 & 3 & 2 \\ 2 & 4 & 7 \end{bmatrix}$ ,  

$$A(1|3) = \begin{bmatrix} 1 & 3 \\ 2 & 4 \end{bmatrix} = \begin{bmatrix} 1 & 2 & 3 \\ 1 & 3 & 2 \\ 2 & 4 & 7 \end{bmatrix}$$
,  
and  $A(1, 2|1, 3) = [4]$ .

Determinant: Determinant of a square matrix  $A = [a_{ij}]$ , denoted det(A) (or |A|) is defined by

$$\det(A) = \begin{cases} a, & \text{if } A = [a] \ (n = 1), \\ \sum_{j=1}^{n} (-1)^{1+j} a_{1j} \det(A(1|j)), & \text{otherwise.} \end{cases}$$

• Let A = [-2]. Then det(A) = |A| = -2.

• Let 
$$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$
. Then,  $\det(A) = |A| = a |A(1|1)| - b |A(1|2)| = ad - bc$ .  
For  $A = \begin{bmatrix} 1 & 2 \\ 3 & 5 \end{bmatrix}$ ,  $\det(A) = \begin{vmatrix} 1 & 2 \\ 3 & 5 \end{vmatrix} = 1 \cdot 5 - 2 \cdot 3 = -1$ .

Singular, Non-Singular: A matrix A is said to be a SINGULAR if  $\det(A) = 0$ . It is called NON-SINGULAR if  $\det(A) \neq 0$ .

Th Let A be an  $n \times n$  matrix. If

- $B = E_{ij}A$  then, det(B) = -det(A),
- $B = E_i(c), c \neq 0$  then, det(B) = c det(A),
- $B = E_{ij}(c), c \neq 0$  then, det(B) = det(A),
- all the elements of one row of A are 0 then, det(A) = 0,
- two rows of A are equal then det(A) = 0.
- A is a triangular matrix then det(A) is product of diagonal entries.

Th As  $\det(I_n) = 1$ ,  $\det(E_{ij}) = -1$   $\det(E_i(c)) = c$  whenever  $c \neq 0$   $\det(E_{ij}(c)) = 1$ , whenever  $c \neq 0$ 

• Let  $A_{n \times n} = [a_{ij}]$ . Then, for any  $k, 1 \le k \le n$ 

$$\det(A) = \sum_{j=1}^{n} (-1)^{k+j} a_{kj} \det(A(k|j)).$$

• Let  $A_{n\times n}$  matrix. Then,  $|\det(A)|$  equals the volume of the *n*-dimensional parallelepiped formed by the rows of A.

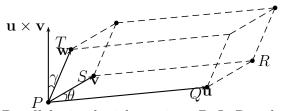


Figure 3: Parallelepiped with vertices P, Q, R and S as base

• Let 
$$\mathbf{u}^t = (u_1, u_2, u_3), \mathbf{v}^t = (v_1, v_2, v_3)$$
 and  $\mathbf{w}^t = (w_1, w_2, w_3)$ .  
Then,  $\mathbf{u} \times \mathbf{v} = (u_2v_3 - u_3v_2, u_3v_1 - u_1v_3, u_1v_2 - u_2v_1)$  and

volume 
$$(P) = \text{Area}(PQRS) \cdot \text{height} = |\mathbf{w} \bullet (\mathbf{u} \times \mathbf{v})| = \pm \begin{vmatrix} w_1 & w_2 & w_3 \\ u_1 & u_2 & u_3 \\ v_1 & v_2 & v_3 \end{vmatrix}.$$

(i, j)<sup>th</sup> minor of A Denoted  $A_{ij}$  equals  $\det (A(i|j))$ .

(i,j)<sup>th</sup> cofactor of A: Denoted  $C_{ij}$  equals  $(-1)^{i+j}A_{ij}$ .

Adjoint of a Matrix: Denoted  $Adj(A) = [C_{ji}]$ 

• Example 
$$A = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 3 & 1 \\ 2 & 2 & 2 \end{bmatrix}$$
. Then,  $Adj(A) = \begin{bmatrix} 4 & 2 & -7 \\ -2 & -4 & 5 \\ -2 & 2 & -1 \end{bmatrix}$  as

$$C_{11} = (-1)^{1+1}A_{11} = 4, C_{21} = (-1)^{2+1}A_{21} = 2, \dots, C_{33} = (-1)^{3+3}A_{33} = -1.$$

$$A \cdot Adj(A) = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 3 & 1 \\ 1 & 2 & 2 \end{bmatrix} \begin{bmatrix} 4 & 2 & -7 \\ -2 & -4 & 5 \\ -2 & 2 & -1 \end{bmatrix}$$
$$= -6 \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = Adj(A) \cdot A.$$

Note that det(A) = -6. So, Important Observation:

$$A \cdot Adj(A) = Adj(A) \cdot A = \det(A)I.$$

Is it always TRUE?

### P. Determinant-cofactor

Th Let A be an  $n \times n$  matrix. Then,

• for 
$$1 \le i \le n$$
,  $\sum_{i=1}^{n} a_{ij} C_{ij} = \sum_{i=1}^{n} a_{ij} (-1)^{i+j} A_{ij} = \det(A)$ ,

• for 
$$i \neq \ell$$
,  $\sum_{j=1}^{n} a_{ij} C_{\ell j} = \sum_{j=1}^{n} a_{ij} (-1)^{\ell+j} A_{\ell j} = 0$ ,

•  $A(Adj(A)) = \det(A)I_n$ . Thus,

whenever 
$$det(A) \neq 0$$
 one has  $A^{-1} = \frac{1}{det(A)} Adj(A)$ .

Th A square matrix A is non-singular if and only if A is invertible.

Th Let A and B be square matrices of order n. Then

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

Th Let A be a square matrix. Then  $det(A) = det(A^t)$ .

P. Cramer's Rule

Th The following statements are equivalent for a square matrix A:

- $\bullet$  A is invertible.
- The linear system  $A\mathbf{x} = \mathbf{b}$  has a unique solution for every  $\mathbf{b}$ .
- $\det(A) \neq 0$ .

Cramer's Rule Let A be an  $n \times n$  matrix. If  $det(A) \neq 0$  then, the unique solution of the linear system  $A\mathbf{x} = \mathbf{b}$  is

$$x_j = \frac{\det(A_j)}{\det(A)}, \text{ for } j = 1, 2, \dots, n,$$

where  $A_j$  is the matrix obtained from A by replacing the jth column of A by the column vector **b**.