

Lecture V: §1.3 Consequences of Gauss-Jordan  
§1.5 Matrix Operations

§1. Consequences of Gauss-Jordan:

- ① A linear system has 0, 1 or  $\infty$ -many solutions
- ② If # equations  $<$  # variables, the linear system cannot have a unique solution  
(Reason: # dependent variables  $\leq$  # eqns  $<$  # variables, so we will have at least one independent variable)

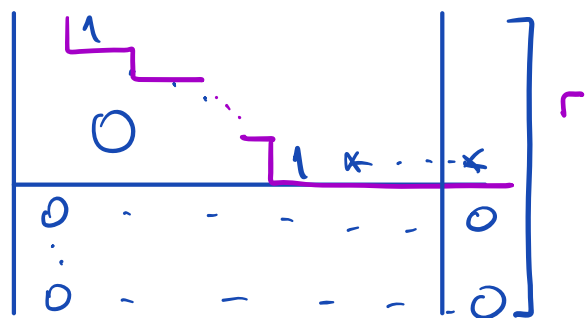
Definition: Assume the matrix  $B'$  is in reduced echelon form. We define  $\text{Rank}(B') = r = \# \text{ nonzero rows of } B' = \# \text{ dependent vars.}$

For arbitrary matrices  $B$   $\text{Rank}(B) = \text{Rank}(B')$  where  $B'$  is the unique reduced echelon form matrix equivalent to  $B$ .

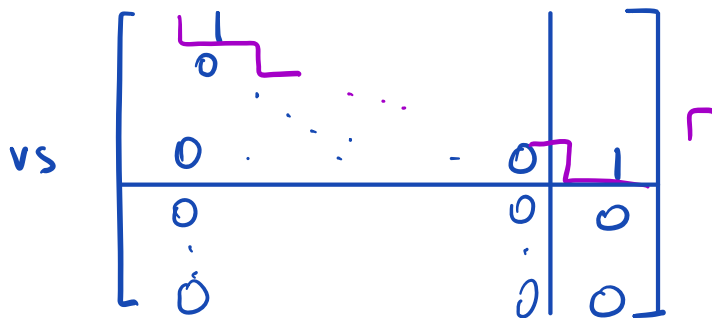
- Consequences:
- ①  $\text{Rank}(B') \leq m = \# \text{ rows of } B'$
  - ②  $\text{Rank}(B') \leq n+1 = \# \text{ columns of } B'$
  - ③  $\text{Rank}(B') \leq n$  if the system is consistent
  - ④ # Free parameters =  $n - \text{Rank}(B')$ , so we have a unique solution if and only if the system is consistent &  $\text{Rank}(B') = n$ .  
(note that  $n \leq m$  by ①)

Why ③?  $B' = \left[ \begin{array}{ccc|c} \overset{\text{1}}{\circ} & \dots & & \\ \hline \circ & \dots & \circ & \circ \\ \circ & \dots & \circ & \circ \end{array} \right] \left. \vphantom{\begin{array}{ccc|c} \overset{\text{1}}{\circ} & \dots & & \\ \hline \circ & \dots & \circ & \circ \\ \circ & \dots & \circ & \circ \end{array}} \right\} \begin{array}{l} r = \text{Rank}(B') \\ \text{zero rows} \end{array}$   
 $\underbrace{\hspace{10em}}_n$

If  $B'$  has  $[0, \dots, 0, 1]$  as a row, we are inconsistent. This means by our assumptions that the right-most starting 1 is within the first  $n$  columns. The staircase shape forces  $r \leq n$ .



consistent



inconsistent

## §2. Operations on Matrices:

Matrices = rectangular arrays of numbers

NEXT GOALS: ① Define 3 operations on matrices

Addition

Scalar Multiplication

Multiplication

(these 3 will be used to define abstract vector spaces)

② Study the algebra behind these operations (rules that will help us compute faster, like we do with  $+$  &  $\times$  in  $\mathbb{R}$ )

Application: We can write linear systems  $\begin{cases} a_{11}x_1 + \dots + a_{1n}x_n = b_1 \\ \vdots \\ a_{m1}x_1 + \dots + a_{mn}x_n = b_m \end{cases}$

using matrix multiplication

$$\underset{m \times n}{A} \underset{n \times 1}{\begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}} = \underset{m \times 1}{\begin{bmatrix} b_1 \\ \vdots \\ b_m \end{bmatrix}}$$

$$A = (a_{ij})_{\substack{1 \leq i \leq m \\ 1 \leq j \leq n}}$$

Definition: Two matrices are equal when they have the same size & the same entries. More explicitly:

if  $A = (a_{ij})_{\substack{1 \leq i \leq m \\ 1 \leq j \leq n}}$  &  $B = (b_{ij})_{\substack{1 \leq i \leq r \\ 1 \leq j \leq s}}$  are two matrices of sizes  $m \times n$  &  $r \times s$ , respectively then  $A = B$  means  $m = r$ ,  $n = s$  and  $a_{ij} = b_{ij}$  for every  $i = 1, \dots, m$  &  $j = 1, \dots, n$

Examples:  $\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \neq \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$  because (1,2) entries don't agree

$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \neq \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$  because the number of columns is different.

Definition: A scalar is a number (real or complex)

• Next we define addition and scalar multiplication, and study their properties.

### § 3. Matrix Addition & Scalar Multiplication:

• Fix two matrices  $A$  &  $B$  of the same size. We want to build a new matrix  $A+B$

Definition 1: If  $A$  &  $B$  have size  $m \times n$ , then  $A+B$  is a matrix of size  $m \times n$  with entries

$$(A+B)_{i,j} = A_{i,j} + B_{i,j}$$

$$\text{for } \begin{matrix} i=1, \dots, m \\ j=1, \dots, n \end{matrix}$$

(add entry by entry)

Examples: (1)  $\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + \begin{bmatrix} 3 & 1 \\ 2 & 0 \end{bmatrix} = \begin{bmatrix} 1+3 & 0+1 \\ 0+2 & 1+0 \end{bmatrix} = \begin{bmatrix} 4 & 1 \\ 2 & 1 \end{bmatrix}$

(2)  $\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \end{bmatrix}$  is not defined (the matrices have different sizes!)

• Given a matrix  $A$  and a scalar  $r$  we want to build a new matrix  $rA$

Definition 2: If  $A$  has size  $m \times n$ , then the scalar product  $rA$  is a matrix of size  $m \times n$  with entries

$$(rA)_{i,j} = rA_{i,j} \text{ for } \begin{matrix} i=1, \dots, m \\ j=1, \dots, n \end{matrix}$$

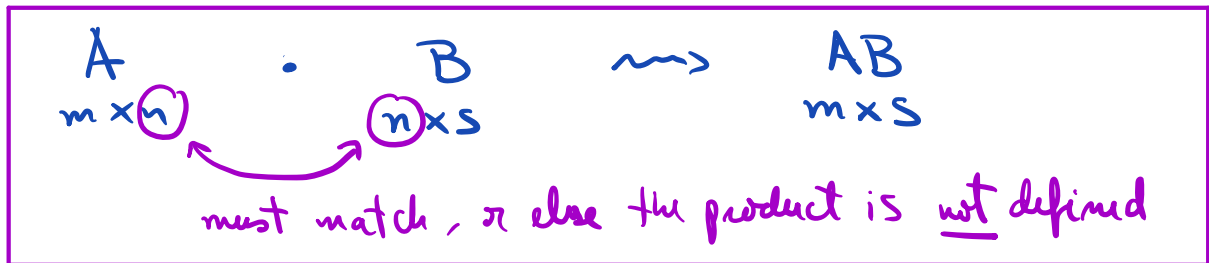
(multiply all entries by  $r$ )

Example  $2 \begin{bmatrix} 1 & 3 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 2 \cdot 1 & 2 \cdot 3 \\ 2 \cdot 0 & 2 \cdot 1 \end{bmatrix} = \begin{bmatrix} 2 & 6 \\ 0 & 2 \end{bmatrix}$

## §4. Matrix multiplication:

• Next, we define multiplication of matrices. We will discuss the definition & do some examples. Next time, we will see why this is the "right definition".

Idea:



Definition: Given  $A = (a_{ij})$   $m \times n$  matrix, the product  $A \cdot B$   
 $B = (b_{ij})$   $r \times s$  —

is only defined when  $n = r$  (# cols  $A =$  # rows  $B$ ). Assuming this is the case, then  $A \cdot B$  is a matrix of size  $m \times s$  with

$$\begin{aligned} \text{entries} \quad (A \cdot B)_{ij} &= a_{i1} b_{1j} + a_{i2} b_{2j} + \dots + a_{in} b_{nj} \\ &= \sum_{k=1}^n a_{ik} b_{kj} \quad (\text{summation notation}) \end{aligned}$$

Pictorially: To get  $(i,j)^{\text{th}}$  entry of  $A \cdot B$ , we focus on the  $i^{\text{th}}$  row of  $A$  & the  $j^{\text{th}}$  column of  $B$

$$\begin{bmatrix} \underline{a_{i1}} & \underline{a_{i2}} & \dots & a_{in} \end{bmatrix} \quad \begin{bmatrix} b_{1j} \\ \underline{b_{2j}} \\ \vdots \\ b_{nj} \end{bmatrix} \quad . \quad \text{We multiply the corresponding}$$

entries & add them up:

$$\rightsquigarrow a_{i1} b_{1j} + a_{i2} b_{2j} + \dots + a_{in} b_{nj} = (ij) \text{ entry of } AB$$

Examples: (1)  $\begin{bmatrix} 1 & 0 & 3 \\ 1 & 2 & 1 \end{bmatrix} \cdot \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$  is NOT defined

$2 \times 3$   $2 \times 2$

(2)  $\begin{bmatrix} 1 & 0 & 3 \\ 1 & 2 & 1 \end{bmatrix} \begin{bmatrix} 1 & -1 & 3 \\ 0 & 2 & 7 \\ -2 & 4 & 5 \end{bmatrix}$  is defined and gives a  $2 \times 3$  matrix

$2 \times 3$

$3 \times 3$

• (1,1) entry:  $\begin{bmatrix} 1 & 0 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ -2 \end{bmatrix} = 1 \cdot 1 + 0 \cdot 0 + 3 \cdot (-2) = -5$

• (1,2) — :  $\begin{bmatrix} 1 & 0 & 3 \end{bmatrix} \begin{bmatrix} -1 \\ 2 \\ 4 \end{bmatrix} = 1 \cdot (-1) + 0 \cdot 2 + 3 \cdot 4 = -1 + 12 = 11$

• (1,3) — :  $\begin{bmatrix} 1 & 0 & 3 \end{bmatrix} \begin{bmatrix} 3 \\ 7 \\ 5 \end{bmatrix} = 1 \cdot 3 + 0 \cdot 7 + 3 \cdot 5 = 3 + 15 = 18$

similarly, we can compute the (2,1), (2,2) & (2,3) entries.

Result:  $\begin{bmatrix} 1 & 0 & 3 \\ 1 & 2 & 1 \end{bmatrix} \begin{bmatrix} 1 & -1 & 3 \\ 0 & 2 & 7 \\ -2 & 4 & 5 \end{bmatrix} = \begin{bmatrix} -5 & 11 & 18 \\ -1 & 7 & 22 \end{bmatrix}$

Note: The formula for each entry of  $A \cdot B$  is exactly how we define dot products in  $\mathbb{R}^2$  &  $\mathbb{R}^3$ . This is more general: