

What about systems  $AX=Y$ ? non-homogeneous systems.

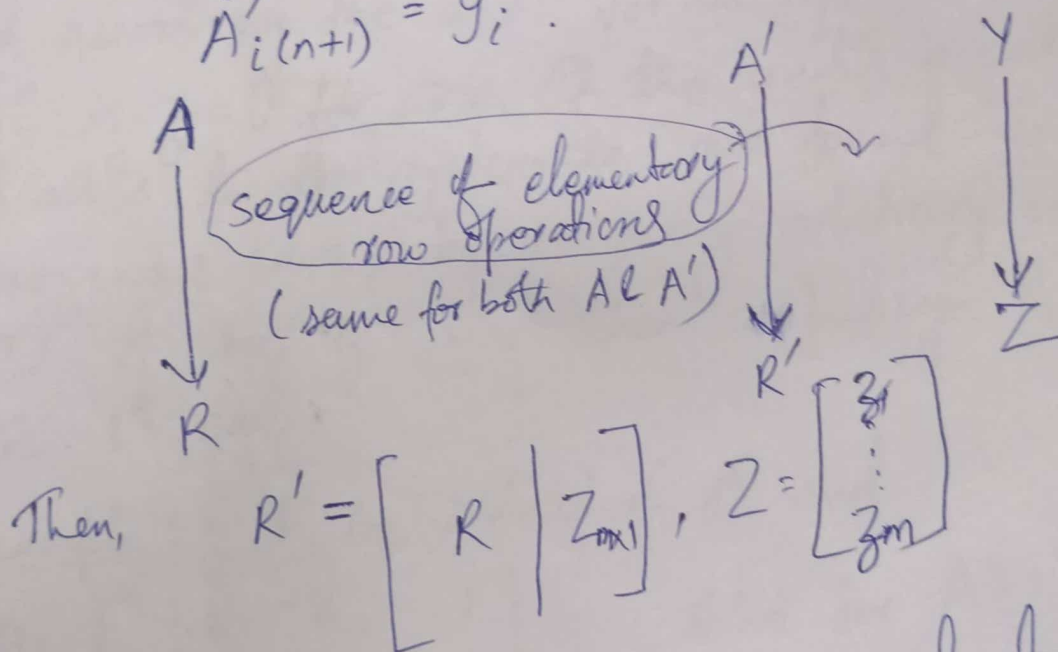
→ While  $AX=0$  always has a trivial sol<sup>n</sup>, systems  $AX=Y$  for  $Y \neq 0$  need not have a sol<sup>n</sup>.

How to find solutions for  $AX=Y, Y \neq 0$ ?

→ Form the augmented matrix  $A'$  of the system  $AX=Y$ .  $A'$  is the  $m \times (n+1)$  matrix whose 1<sup>st</sup>  $n$  columns are the columns of  $A$  and whose last column is  $Y$ .

$$A'_{ij} = A_{ij} \quad \forall j \leq n$$

$$A'_{i(n+1)} = y_i.$$



$AX=Y$  and  $RX=Z$  are equivalent and hence have same solutions.

Whether  $RX=Z$  has any solutions? To determine all the sol<sup>n</sup>s if any exist.

If  $R$  has  $r$  non-zero rows, with leading non-zero entry of row  $i$  occurring in column  $k_i, i=1, \dots, r$ , then the first  $r$  eq<sup>s</sup> of  $RX=Z$  effectively express  $x_{k_1}, \dots, x_{k_r}$  in the terms of the  $(n-r)$  remaining  $x_j$  and the scalars  $z_1, \dots, z_r$ . The last  $(m-r)$  eq<sup>s</sup> are:

$$\begin{aligned} 0 &= z_{r+1} \\ &\vdots \\ 0 &= z_m \end{aligned}$$

and accordingly the cond<sup>n</sup> for the system to have a sol<sup>n</sup> is  $z_i = 0$  for  $i > r$ . If this cond<sup>n</sup> is satisfied, all sol<sup>n</sup>s to the system are found as in the homogenous case, by assigning arbitrary values to  $(n-r)$  of the  $x_j$  and then computing  $x_{k_i}$  from the  $i^{\text{th}}$  eq<sup>n</sup>.

Example:  $F$  be a field of  $\mathbb{Q}$  and

$$A = \begin{bmatrix} 1 & -2 & 1 \\ 2 & 1 & 1 \\ 0 & 5 & -1 \end{bmatrix}.$$

$$\text{Solve for } AX=Y = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix}.$$

We perform a sequence of row operations on the augmented matrix  $A'$  which row-reduces  $A$ :

$$\begin{bmatrix} 1 & -2 & 1 & y_1 \\ 2 & 1 & 1 & y_2 \\ 0 & 5 & -1 & y_3 \end{bmatrix} \xrightarrow{2-2\otimes\textcircled{1}} \begin{bmatrix} 1 & -2 & 1 & y_1 \\ 0 & 5 & -1 & (y_2-2y_1) \\ 0 & 5 & -1 & y_3 \end{bmatrix}$$

$$\downarrow \textcircled{3}-\textcircled{1} \quad \begin{bmatrix} 1 & -2 & 1 & y_1 \\ 0 & 5 & -1 & (y_2-2y_1) \\ 0 & 0 & 0 & (y_3-y_2+2y_1) \end{bmatrix} \xleftarrow{\textcircled{2} \cdot \frac{1}{5}} \begin{bmatrix} 1 & -2 & 1 & y_1 \\ 0 & 1 & -\frac{1}{5} & \frac{1}{5}(y_2-2y_1) \\ 0 & 0 & 0 & (y_3-y_2+2y_1) \end{bmatrix}$$

$$\downarrow \textcircled{1} + 2\otimes\textcircled{2} \quad \begin{bmatrix} 1 & 0 & \frac{3}{5} & \frac{1}{5}(y_1+2y_2) \\ 0 & 1 & -\frac{1}{5} & \frac{1}{5}(y_2-2y_1) \\ 0 & 0 & 0 & (y_3-y_2+2y_1) \end{bmatrix}$$

Cond<sup>n</sup>: that  $AX=Y$  has a sol<sup>n</sup> is

$$2y_1 - y_2 + y_3 = 0$$

and if scalars  $y_i$  satisfy this cond<sup>n</sup>, all sol<sup>n</sup>s are obtained by assigning a value  $c$  to  $x_3$  & then computing

$$x_1 = -\frac{3}{5}c + \frac{1}{5}(y_1+2y_2)$$

$$x_2 = \frac{1}{5}c + \frac{1}{5}(y_2-2y_1)$$



# Lecture 3: Matrix multiplication

Suppose  $B$  is an  $n \times p$  matrix over a field  $F$  with rows  $\beta_1, \dots, \beta_n$  and that from  $B$  we construct a matrix  $C$  with rows  $\gamma_1, \dots, \gamma_m$  by forming certain linear combinations:

$$\gamma_i = A_{i1}\beta_1 + A_{i2}\beta_2 + \dots + A_{in}\beta_n \quad (1.4)$$

The rows of  $C$  are determined by the ~~mxn~~  $mn$  scalars  $A_{ij}$  which are themselves the entries of an  $m \times n$  matrix  $A$ .  
From (1.4),

$$(\gamma_1 \quad \dots \quad \gamma_i \quad \dots \quad \gamma_p) = \sum_{r=1}^n (A_{ir}\beta_{r1} \quad \dots \quad A_{ir}\beta_{rp}),$$

$$\text{entries of } C: \quad C_{ij} = \sum_{r=1}^n A_{ir} B_{rj}.$$

Def<sup>n</sup>: Let  $A$  be an  $m \times n$  matrix over the field  $F$  and let  $B$  be an  $n \times p$  matrix over  $F$ . The product  $AB$  is the  $m \times p$  matrix  $C$  whose  $i, j$  entry is

$$C_{ij} = \sum_{r=1}^n A_{ir} B_{rj}.$$

Example: (a) Consider  $\begin{bmatrix} 5 & -1 & 2 \\ 0 & 7 & 2 \end{bmatrix} \cdot \begin{bmatrix} 1 & 0 \\ -3 & 1 \end{bmatrix} \begin{bmatrix} 5 & -1 & 2 \\ 15 & 4 & 8 \end{bmatrix}$

$$\text{Here, } \gamma_1 = (5 \quad -1 \quad 2) = 1 \cdot (5 \quad -1 \quad 2) + 0 \cdot (15 \quad 4 \quad 8)$$

$$\gamma_2 = (0 \quad 7 \quad 2) = -3 \cdot (5 \quad -1 \quad 2) + 1 \cdot (15 \quad 4 \quad 8).$$

$$(b) \begin{bmatrix} 0 & 6 & 1 \\ 9 & 12 & -8 \\ 12 & 62 & -3 \\ 3 & 8 & -2 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ -2 & 3 \\ 5 & 4 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0 & 6 & 1 \\ 3 & 8 & -2 \end{bmatrix}$$

$$\gamma_3 = 5 \cdot (0 \quad 6 \quad 1) + 4 \cdot (3 \quad 8 \quad -2)$$

$B$  is an  $n \times p$  matrix,  $B = [B_1, \dots, B_p]$ ,  
 $B_j = \begin{bmatrix} B_{1j} \\ \vdots \\ B_{nj} \end{bmatrix}$ ,  $1 \leq j \leq p$ .  $B_j$  is  $1 \times n$  matrix.  
 column matrix.

Check that  $AB = [AB_1, \dots, AB_p]$ .

Thm: If  $A, B, C$  are matrices over the field  $F$  such that the products  $BC$  and  $A(BC)$  are defined, then so are the products  $AB, (AB)C$  and  $(AB)C = A(BC)$ .

Proof: —————

Remark: For a square matrix  $A$ ,  $A^n$  is well-defined  
 $A^p A^q A^r = A^s A^t A^u$  for all  $p+q+r=s+t+u=n$ .

$A(BC) = (AB)C \rightarrow$  linear combinations of linear combinations of the rows of  $C$  are again linear combinations of the rows of  $C$ .

If  $B \xrightarrow[\text{row operations}]{\text{elementary}} C$ , then each row of  $C$  is a linear combination of the rows of  $B$ , and so  $\exists$  a matrix  $A$  s.t.  $AB = C$ .  
 (There can be many such  $A$ 's in general.)

Def: An  $m \times m$  matrix is said to be an elementary matrix if it can be obtained from the  $m \times m$  identity matrix  $I_{m \times m}$  by means of a single elementary row operation.

Example:  $2 \times 2$  elementary matrices:

$$\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 1 & c \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ c & 1 \end{bmatrix},$$

$$\begin{bmatrix} c & 0 \\ 0 & 1 \end{bmatrix} \text{ for } c \neq 0, \begin{bmatrix} 1 & 0 \\ 0 & c \end{bmatrix} \text{ for } c \neq 0.$$

Thm: Let  $e$  be an elementary row operation and let  $E$  be the  $m \times m$  elementary matrix  $E = e(I)$ . Then, for every  $m \times n$  matrix  $A$ ,

$$e(A) = EA.$$

Proof: Type (1)  $E_{ik} = \begin{cases} \delta_{ik}, & i \neq r \\ \delta_{ik} + c\delta_{sk}, & i = r \end{cases}$  (To replace row  $r$  with row  $r$  +  $c$  row  $s$ )

$$(EA)_{ij} = \sum_{k=1}^m E_{ik} A_{kj} = \begin{cases} A_{ik}, & i \neq r \\ A_{rj} + cA_{sj}, & i = r. \end{cases}$$

Check for other types.

$E_{rr}$   $\begin{bmatrix} 1 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & \textcircled{r} & \dots & 1 \end{bmatrix}$   $E_{rs}$   $\begin{bmatrix} 1 & \dots & 1 & \dots & \textcircled{r} & \dots \\ \vdots & \ddots & \vdots & \ddots & \vdots & \ddots \\ r \rightarrow r + c \cdot s & \dots & \dots & \dots & \dots & \dots \end{bmatrix}$   $E_{rs}$   $\begin{bmatrix} 1 & \dots & 1 & \dots & \textcircled{r} & \dots \\ \vdots & \ddots & \vdots & \ddots & \vdots & \ddots \\ r \rightarrow r + c \cdot s & \dots & \dots & \dots & \dots & \dots \end{bmatrix}$



Corollary: Let  $A$  and  $B$  be  $n \times n$  matrices over the field  $F$ . Then  $B$  is row-equivalent to  $A$  if and only if  $B = PA$ , where  $P$  is a product of  $n \times n$  elementary operations.

## IV Invertible matrices.

Defn: Let  $A$  be an  $n \times n$  matrix over the field  $F$ . An  $n \times n$  matrix  $B$  such that  $BA = I$  is called a left inverse of  $A$ ; an  $n \times n$  matrix  $B$  such that  $AB = I$  is called a right inverse of  $A$ . If  $AB = BA = I$  then  $B$  is called a two-sided inverse of  $A$  and  $A$  is said to be invertible.

Lemma: If  $A$  has a left inverse  $B$  and a right inverse  $C$ , then  $B = C$ .

Proof:  $B = BI = BAC = IC = C$ .

Thm: Let  $A$  and  $B$  be  $n \times n$  matrices over the field  $F$ .

(i) If  $A$  is invertible, so is  $A^T$  and  $(A^{-1})^T = A^T$ .

(ii) If both  $A$  and  $B$  are invertible, so is  $AB$  and  $(AB)^T = B^T A^T$ .

Corollary: A product of invertible matrix is invertible.

Theorem: An elementary matrix is invertible.

Proof: \_\_\_\_\_

Thm: If  $A$  is an  $n \times n$  matrix, the following are equivalent.

- (i)  $A$  is invertible.
- (ii)  $A$  is row-equivalent to  $I_{n \times n}$ .
- (iii)  $A$  is product of elementary operations.

Proof: \_\_\_\_\_

Thm: For an  $n \times n$  matrix  $A$ , the following are equivalent.

- (i)  $A$  is invertible.
- (ii) The homogenous system  $AX=0$  has only the trivial sol<sup>n</sup>.
- (iii) The system of eq's  $AX=Y$  has a sol<sup>n</sup>  $X$  for each  $n \times 1$  matrix  $Y$ .

Proof: \_\_\_\_\_

Column-equivalent

Column-reduced echelon matrix

~~Column~~ Elementary column operations:  $AE$

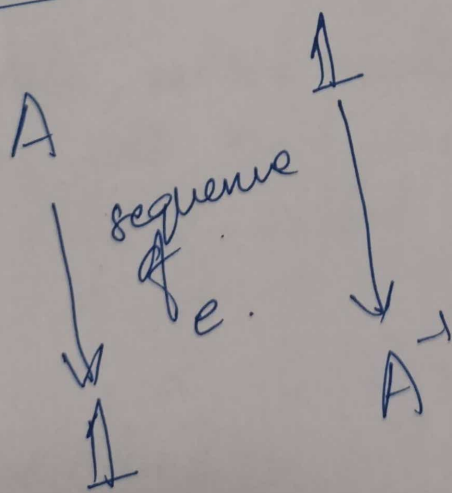
$$\begin{aligned} AX &= Y \\ X^T A^T &= Y^T \end{aligned}$$



Corollary: A square matrix with either a left inverse or right inverse is invertible.

Proof:  $A_{n \times n}$ . Suppose left inverse of  $A$  exists,  $BA = I$ . Then,  $AX = 0$  has only the trivial sol<sup>n</sup>, because  $X = I X = B(AX)$ .  $\therefore A$  is invertible.

If  $A$  has a right inverse,  $AC = I$ . Then  $C$  has a left inverse & is therefore invertible. It then follows  $A = C^{-1}$ , so  $A$  is invertible w/ inverse  $C$ .



$$AX = Y$$

$$RX = Z,$$

$$\text{if } R = PA$$

$$\text{then } Z = PY$$

invertible matrix.

# Lecture 4

## Vector spaces

Def<sup>n</sup>. A vector space (or linear space) consists of the following:

Vector space over the field

1. a field  $F$  of scalars;
2. a set  $V$  of objects, called vectors;
3. a rule (or operation), called vector addition, which associates with each pair of vectors  $\vec{\alpha}, \vec{\beta} \in V$  a vector  $\vec{\alpha} + \vec{\beta} \in V$ , called the sum of  $\vec{\alpha}$  &  $\vec{\beta}$ , in such a way that

- (a) addition is commutative,  $\vec{\alpha} + \vec{\beta} = \vec{\beta} + \vec{\alpha}$ ;
- (b) addition is associative,  $\vec{\alpha} + (\vec{\beta} + \vec{\gamma}) = (\vec{\alpha} + \vec{\beta}) + \vec{\gamma}$ ;
- (c)  $\exists$  a unique vector  $\vec{0} \in V$ , called the zero vector, s.t.  $\vec{\alpha} + \vec{0} = \vec{\alpha} \forall \vec{\alpha} \in V$ ;
- (d) for each  $\vec{\alpha} \in V \exists$  a unique vector  $-\vec{\alpha} \in V$  s.t.  $\vec{\alpha} + (-\vec{\alpha}) = \vec{0}$ .

4. a rule, called scalar multiplication, which associates with each scalar  $c \in F$  &  $\vec{\alpha} \in V$  a vector  $c\vec{\alpha} \in V$ , called the product of  $c$  &  $\vec{\alpha}$  s.t.

- (a)  $1\vec{\alpha} = \vec{\alpha} \forall \vec{\alpha} \in V$ ;
- (b)  $(c_1 c_2)\vec{\alpha} = c_1(c_2\vec{\alpha})$ ;
- (c)  $c(\vec{\alpha} + \vec{\beta}) = c\vec{\alpha} + c\vec{\beta}$ ;
- (d)  $(c_1 + c_2)\vec{\alpha} = c_1\vec{\alpha} + c_2\vec{\alpha}$ .

A vector space is a composite object consisting of a field, a set of 'vectors', & two operations w/ certain properties.

### Examples

① The  $n$ -tuple space,  $F^n$ . Let  $F$  be any field. Let  $V$  be the set of all  $n$ -tuples  $\bar{x} = (x_1, \dots, x_n)$  of scalars  $x_i \in F$ . If  $\bar{y} = (y_1, y_2, \dots, y_n)$  w/  $y_i \in F$ , the sum of  $\bar{x}$  &  $\bar{y}$  is defined by

$$\bar{x} + \bar{y} = (x_1 + y_1, x_2 + y_2, \dots, x_n + y_n) \quad \text{--- (2.1)}$$

The product of a scalar  $c$  and vector  $\bar{x}$  is defined by

$$c\bar{x} = (cx_1, \dots, cx_n) \quad \text{--- (2.2)}$$

② The space of  $m \times n$  matrices,  $F^{m \times n}$ . Let  $F$  be any field and let  $m$  &  $n$  be the integers. Let  $F^{m \times n}$  be the set of all  $m \times n$  matrices over the field  $F$ .

$\bar{A}, \bar{B} \in F^{m \times n}$  then

$$(\bar{A} + \bar{B})_{ij} = A_{ij} + B_{ij}$$

$c \in F, \bar{A} \in F^{m \times n}$  then

$$(c\bar{A})_{ij} = cA_{ij}$$

③ The space of functions from a set to a field.  $F$  be field,  $S$  be any non-empty set.  $V$  be the set of all  $f$ 's from the set  $S$  into  $F$ .



For  $f, g \in V$ ,

$$(f+g)(s) = f(s) + g(s).$$

For  $c \in F$ ,  $f \in V$ ,

$$(cf)(s) = cf(s).$$

④ The space of polynomial  $f^n$ s over a field  $F$ .  
Let  $F$  be a field and let  $V$  be the set of all  $f$ 's  $f$  from  $F$  into  $F$  which have the rule of the form

$$f(x) = c_0 + c_1x + \dots + c_nx^n,$$

where  $c_0, c_1, \dots, c_n \in F$  are independent of  $x$ .

⑤ The field  $\mathbb{C}$  of complex nos.  $\rightarrow$  a vector space over the field  $\mathbb{R}$  of real nos.

✓ From the def<sup>n</sup> of a vector space, we observe: ~~that~~ If for a scalar  $c \in F$  and vector  $\bar{\alpha} \in V$  we have  $c\bar{\alpha} = \bar{0}$  then either  $c = 0$  or  $\bar{\alpha} = \bar{0}$ .

For any  $\bar{\alpha} \in V$ ,  $-\bar{\alpha} \in V$  since  $\bar{0} \in V$  and

$$\bar{0} = 0\bar{\alpha} = (1-1)\bar{\alpha} = 1\bar{\alpha} + (-1)\bar{\alpha} = \bar{\alpha} + (-1)\bar{\alpha},$$

$$(-1)\bar{\alpha} = -\bar{\alpha}.$$

For any  $\bar{\alpha}_1, \bar{\alpha}_2, \bar{\alpha}_3, \bar{\alpha}_4 \in V$ ,  $(\bar{\alpha}_1 + \bar{\alpha}_2) + (\bar{\alpha}_3 + \bar{\alpha}_4) = (\bar{\alpha}_2 + (\bar{\alpha}_1 + \bar{\alpha}_3)) + \bar{\alpha}_4$ .