

- 1) Finish loose ends from Chapter 2
- 2) Cover Chapter 3

1) Let  $\vec{u}, \vec{v} \in \mathbb{R}^n$  be non-zero vectors.

We define  $\varphi_{\vec{u}}: \mathbb{R}^n \rightarrow \mathbb{R}$  and  $\varphi_{\vec{v}}: \mathbb{R} \rightarrow \mathbb{R}^n$   
 $\vec{x} \mapsto \vec{u}^T \vec{x} = \vec{u} \cdot \vec{x}$  and  $t \mapsto t\vec{v}$

These are lin. transf. as they are given by matrices  $u^t$  and  $v$  resp.

The lin transf.  $\varphi_{\vec{v}} \circ \varphi_{\vec{u}}: \mathbb{R}^n \rightarrow \mathbb{R}^n$  take  $\vec{x}$  to  $\vec{v}(\vec{x} \cdot \vec{u})$  or  $v u^T x$   
 Thus its matrix is  $v u^t$  (which is  $n \times n$ )

Q: What is rank of matrix  $v u^t$  (where  $v, u \neq \vec{0}$ )

Hint (rk of matrix  $A_{m \times n}$  is # d.o.f of solution of  $Ax = 0$ )

If  $v = u$  we called  $\varphi_{\vec{v}} \circ \varphi_{\vec{u}}$  as  $T_u$  in HW3 (when  $\vec{u} \cdot \vec{u} = 1$  additionally)

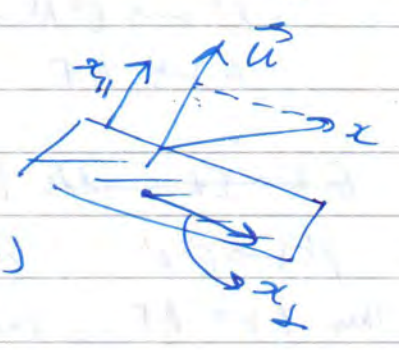
Basically  $\vec{x} \mapsto \frac{(\vec{u} \cdot \vec{x})}{\vec{u} \cdot \vec{u}} \vec{u}$  is projection of  $\vec{x}$  on direction of  $\vec{u}$

when  $\vec{u} \cdot \vec{u} = u^t u = 1$  projection is  $T_u$ .

$$x = T_u x + x - T_u x$$

$$= x_{||} + x_{\perp}$$

(where  $x_{||}$  is parallel to  $\vec{u}$ ,  $x_{\perp}$  is  $\perp$  to  $\vec{u}$ )



Clearly  $x_{\perp} - x_{||} = x - 2x_{||} = (I - 2T_u)x$  is reflection of  $\vec{x}$  in plane  $\perp$  to  $\vec{u}$ .

### Proposition

(2)

If  $A$  is a  $n \times n$  Sq matrix and  $\exists$  a  $n \times n$  matrix  $B$  s.t.  
one of  $AB = I_n$  ;  $BA = I_n$  then  $B = A^{-1}$

(original def<sup>n</sup> requires both these conditions)

Proof: If  $AB = I_n$  then  $ABx = x$  so  $ABx = 0$  only if  $x = 0$   
Thus  $Bx = 0$  only if  $x = 0 \Rightarrow B$  is invertible

Multiply  $AB = I_n$  by  $B^{-1}$  on right to get  $A = B^{-1}$ . Hence  $A^{-1} = B$ .

If  $BA = I_n$  above shows  $B = A^{-1}$ .

Q: Let  $A$  be  $m \times n$  matrices what does  $A \mapsto AE$  do where  $E$  is  $n \times n$  elem matrix?

As if  $E$  interchanges  $i$ -th and  $j$ -th row then  $E = E^t$  so  $E^t A^t = E A^t$   
interchanges  $i$ -th and  $j$ -th row of  $A^t$ . Thus  $A \mapsto AE$  interchanges  $i$ -th and  $j$ -th cols of  $A$ .

If  $E$  rescales  $i$ -th row by  $C$ , then  $E = E^t$  so  $E^t A^t = E A^t$   
 $A^t \mapsto E^t A^t = E A^t$  rescales  $i$ -th row of  $A^t$   
 $A \mapsto AE$  rescales  $i$ -th col of  $A$  by  $C$

If  $E \mapsto EA$  adds  $E$  corresponds to adding  $C \times j$ -th row to  $i$ -th row

th  $A^t \mapsto E^t A^t$  changes  $i$ -th row of  $A^t$  to  $i$ -th row +  $C \times j$ -th row  
Thus  $A \mapsto AE$  changes  $i$ -th col of  $A$  to  $i$ -th col +  $C \times j$ -th column.



(3)

## 3 definitions -

- 1) A subset  $W \subset \mathbb{R}^n$  is called a subspace if whenever  $w_1, w_2 \in W$  so does  $c_1 w_1 + c_2 w_2 \forall c_1, c_2 \in \mathbb{R}$ .

(Informally a subset is a subspace if it is closed w.r.t. <sup>Taking</sup> linear combinations)

Note taking  $w_1 = w_2$  and  $c_1 = c_2 = 0$  we see  $\vec{0} \in W$

Ex: 1)  $\mathbb{R}^k \subset \mathbb{R}^n$  since  $\{(x_1, \dots, x_k, 0, \dots, 0)^t : x_i \in \mathbb{R}\}$

2) Let  $\vec{v} \in \mathbb{R}^n$  and consider  $W = \{c\vec{v} : c \in \mathbb{R}\}$

3) Let  $v_1, \dots, v_m \in \mathbb{R}^n$  consider  $W = \{c_1 \vec{v}_1 + \dots + c_m \vec{v}_m : c_i \in \mathbb{R}\}$

2) If  $T: \mathbb{R}^m \rightarrow \mathbb{R}^n$  is a lin. transf we define  
 $\text{Im}(T) = \{Tx : x \in \mathbb{R}^m\} \subset \mathbb{R}^n$  "Image of T"

$\text{Ker } T = \{x \in \mathbb{R}^m : Tx = 0\} \subset \mathbb{R}^m$  "Kernel of T".

Claim:  $\text{Im } T$  is subspace of  $\mathbb{R}^n$ ,  $\text{Ker } T$  is subspace of  $\mathbb{R}^m$

Pf: • If  $w_1, w_2 \in \text{Im } T$ , write  $w_1 = Tv_1$ ,  $w_2 = Tv_2$ , then  
 $c_1 \vec{w}_1 + c_2 \vec{w}_2 = c_1 Tv_1 + c_2 Tv_2 = T(c_1 v_1 + c_2 v_2) \in \text{Im } T$ .

• If  $Tv_1 = 0$ ,  $Tv_2 = 0$  then  $T(c_1 \vec{v}_1 + c_2 \vec{v}_2) = c_1 Tv_1 + c_2 Tv_2 = \vec{0} + \vec{0} = \vec{0}$ .

Q: What are Subsp of  $\mathbb{R}$ ?

A: We certainly have  $\{0\}$ ,  $\mathbb{R}$ . If  $W$  is a subsp,  $W \neq \{0\}$   
 pick a vector  $\vec{a} \in W$ ,  $\vec{a} \neq \vec{0}$  then clearly  $\{c\vec{a} : c \in \mathbb{R}\} \subset W$   
 $= \mathbb{R}$ .

So  $\{0\}$ ,  $\mathbb{R}$  only Subsp. of  $\mathbb{R}$

Q: What are Subsp. of  $\mathbb{R}^2$ ?

A:  $\{0\}$ ,  $\{\mathbb{R}^2\}$ ,  $\{t\vec{v} : t \in \mathbb{R}\}$  are Subsp. We claim there are  
 $\vec{v} \neq \vec{0}$  no others.

Let  $W \neq \{0\}$  be Subsp of  $\mathbb{R}^2$ . Pick  $\vec{v} \in W$ , if  $W \neq \{c\vec{v} : c \in \mathbb{R}\}$

$\exists \vec{u} \in W$  s.t.  $\vec{u} \neq c\vec{v}$  for any  $c \in \mathbb{R}$ .

Consider  $A = [\vec{u}, \vec{v}]_{2 \times 2}$ ,  $A\vec{x} = \vec{0}$  has only  $\vec{x} = \vec{0}$  as sol<sup>n</sup>  
why?

So for every  $\vec{b} \in \mathbb{R}^2$   $A\vec{x} = \vec{b}$  has unique sol<sup>n</sup> ( $\vec{x} = A^{-1}\vec{b}$ )  
 In other words  $\text{Im}(A) = \mathbb{R}^2$ . But  $\text{Im}(A) \subset W$ , so  $W = \mathbb{R}^2$

Q: Consider  $W = \{(x_1, x_2, x_3)^T \in \mathbb{R}^3 : x_1 + 2x_2 + 3x_3 = 0\}$

• Find  $T_A$  s.t.  $\text{Im}(T_A) = W$

• Find matrix  $B$  s.t.  $\text{Ker}(T_B) = W$

A: Any  $3 \times m$  matrix, whose cols are ~~to~~ <sup>Span</sup>  $\left(\frac{1}{3}\right)^\perp$   
 e.g.  $\begin{pmatrix} -2 & 0 \\ 1 & -3 \\ 0 & 2 \end{pmatrix}$   $m=2$  works.

$B = \begin{bmatrix} 1 & 2 & 3 \end{bmatrix}$



(5)

Lemma: If  $W \subset \mathbb{R}^n$  Subsp,  $\exists$   $n \times m$  matrix  $A$  with  $m \leq n$

such that  $\text{Im}(T_A) = W$  and  $T_A$  is 1-1.

Pf: We will come up with a sequence of matrices  $A(1), A(2), \dots, A(m)$  where  $A(i)$  has size  $n \times i$ .  $W$  will be  $\text{Im}(T_{A(m)})$

Let  $A(1) = \begin{bmatrix} v_1 \end{bmatrix}_{n \times 1}$  where  $v_1$  is any nonzero vector. If  $\text{Im}(T_{A(1)}) = W$  stop.

Let  $A(2) = \begin{bmatrix} v_1 & v_2 \end{bmatrix}_{n \times 2}$  where  $v_2 \neq c v_1$  for any  $c \in \mathbb{R}$ ,  $v_2 \in W$

If  $\text{Im}(T_{A(2)}) = W$  stop.

$\vdots$

Let  $A(i+1) = \begin{bmatrix} v_1 & \dots & v_i & v_{i+1} \end{bmatrix}$  where

$v_{i+1}$  is not a lin. comb of  $v_1, \dots, v_i$ , ( $v_{i+1} \in W$ )

This process must terminate at some stage  $m \leq n$   
Why?: because if we are at stage  $i$  it means

$A(i) \vec{x} = \vec{0}$  has only  $\vec{0}$  sol<sup>n</sup> i.e.  $\text{Rank}(A(i)) = i \leq \min\{n, i\}$

But  $\text{Rank}(A(i)) \leq \min\{n, i\} \Rightarrow i \leq n$ .

( $T_A$  1-1  $\Leftrightarrow \text{Rank}(A) = \# \text{Cols of } A$ . At stage  $m$   $\text{Im } T_A = W$ ,  $T_A$  is 1-1)

Def<sup>n</sup>: given Subsp  $W \subset \mathbb{R}^n$ , we define  $\dim(W) = m$ .  
 where  $m$  is as in Lemma above.

Q: Can we have  $\dim(W) = m$ ,  $\dim(W) = p$ ,  $m \neq p$ ?

A: No. Sp8  $T: \mathbb{R}^m \rightarrow \mathbb{R}^n$  both one-to-one and have image  $\rightarrow W$  (6)  
 $\uparrow S$   
 $\mathbb{R}^p$

If  $f: \mathbb{R}^m \rightarrow W$  is the (another name for) function defined by  $T$  then by hypothesis  $f$  is bijective so  $f^{-1}: W \rightarrow \mathbb{R}^m$  exists

We note that  $f^{-1}: W \rightarrow \mathbb{R}^m$  satisfies  $f^{-1}(aw_1 + bw_2) = af^{-1}(w_1) + bf^{-1}(w_2)$

Why? pick  $v_1, v_2 \in \mathbb{R}^m$  s.t.  $f(v_i) = w_i$ ,  $i=1,2$ , in other words  $v_i = f^{-1}(w_i)$

then  $af^{-1}(w_1) + bf^{-1}(w_2) = av_1 + bv_2$ . Hence  $f(av_1 + bv_2) =$

$$f(av_1 + bv_2) = T(av_1 + bv_2) = aTv_1 + bTv_2 = af(v_1) + bf(v_2) = aw_1 + bw_2$$

$$\text{Thus } f^{-1}(aw_1 + bw_2) = av_1 + bv_2 = af^{-1}(w_1) + bf^{-1}(w_2)$$

Similarly if  $g: \mathbb{R}^p \rightarrow W$  denotes the function from  $\mathbb{R}^p \rightarrow W$  determined by  $S$ , then  $g^{-1}: W \rightarrow \mathbb{R}^p$  satisfies linearity.

$$\text{Thus } \mathbb{R}^p \xrightarrow{f \circ g^{-1}} \mathbb{R}^m \text{ and } \mathbb{R}^m \xrightarrow{g \circ f^{-1}} \mathbb{R}^p$$

$\mathbb{R}^m \xrightarrow{g^{-1} \circ f} \mathbb{R}^p$  is a lin. transf. which is bijective (its inverse is  $f^{-1} \circ g$ )

If  $C_{p \times m}$  is matrix of  $g^{-1} \circ f$  it is invertible, hence  $p=m$  ■

Note: A basis for a subspace  $W \subseteq \mathbb{R}^n$  is a collection of vectors  $\{w_1, w_2, \dots, w_m\}$  s.t. every element of  $W$  is a lin. combo of  $w_1, \dots, w_m$  in a unique way.

(This is same as saying that  $TA$  is bijective onto  $W$  where  $A$  is the  $n \times m$  matrix with columns  $w_1, w_2, \dots, w_m$ )



# Rank- Nullity theorem.

(7)

Lemma: 1) If  $A = \begin{bmatrix} I_r & 0 \\ 0 & 0 \end{bmatrix}_{m \times n}$  then  $\text{Rank}(A) = r$   
~~nullity(A) =~~

$$\dim(\text{Ker } A) = \dim\{e_{r+1}, \dots, e_n\} = \dim \left\{ \begin{array}{ccc} \mathbb{R}^{n-r} & \xrightarrow{\quad} & \mathbb{R}^n \\ \vec{x} & \mapsto & (0, \vec{x}) \end{array} \right\}$$

is 1-1

Hence  $\dim \text{Ker}(A) = n-r.$

Thus calling  $\text{nullity}(A) = \dim(\text{Ker } A)$  we note

$$\boxed{\text{nullity}(A) + \text{Rank}(A) = n}$$

Thm: If  $A$  is  $m \times n$  ; <sup>and</sup>  $Q_{m \times m}$  and  $P_{n \times n}$  are invertible matrices

then  $\text{nullity}(A) = \text{nullity}(QAP)$ ,  $\text{Rank}(A) = \text{Rank}(QAP)$

Proof: Lemma: 2) Let  $W \subset \mathbb{R}^n$  and  $T_A: \mathbb{R}^n \rightarrow \mathbb{R}^n$  be lin trans with  $A$  invertible then  $\dim(W) = \dim(T_A W)$

Pf: If  $w_1, \dots, w_r$  is basis for  $W$ , then  $Tw_1, \dots, Tw_r$  spans  $T_A W$  and if  $\sum c_i Tw_i = 0$  then  $T(\sum c_i w_i) = 0$ . But  $T$  is invertible so  $\sum c_i w_i = 0 \Rightarrow c_i = 0 \forall i$  because  $w_1, \dots, w_n$  are l.i.

$$\text{Ker}(QAP) = \{x: QAPx = 0\} = Q \cdot \text{Ker}(AP) \quad \text{By Lemma 2, } \text{nullity}(QAP) = \text{nullity}(AP)$$

So  $\text{Ker}(AP) = P^{-1} \text{Ker}(A)$  So by Lemma 2 again,  $\text{nullity}(AP) = \text{nullity } A$

Thus we have shown  $\text{nullity}(A) = \text{nullity}(QAP)$

$$\text{rk}(QAP) = \dim(Q \cdot \text{Im}(AP)) \xrightarrow{\text{Lemma 2}} \dim(\text{Im}(AP)) = \dim(\text{Im } A) = \text{Rank}(A)$$

becau  $AP: \mathbb{R}^n \rightarrow A\mathbb{R}^n$



(8)

Let  $A$  be an  $m \times n$  matrix. We have seen that there exists an invertible  $m \times m$  matrix  $Q = E_s E_{s-1} \dots E_1$  s.t.

$QA$  is RREF. Let  $P = E_1' E_2' \dots E_t'$  be a sequence of  $n \times n$  elem matrices then by Thm above  $\text{rk}(A) = \text{rk}(QAP)$  and  $\text{nullity}(A) = \text{nullity}(QAP)$

Q How to Pick  $P$  wisely so that  $\overset{\text{RREF}}{QAP}$  is as simple as possible  
 $M$  where  $M = \text{RREF}_{m \times n}$  matrix

We can Pick  $P$  s.t.  $M \cdot P = \left[ \begin{array}{c|c} I_r & 0 \\ \hline 0 & 0 \end{array} \right]$  where  $r = \text{rk}(M)$

Subtract  $\beta$  suitable multiple of pivot columns from non pivot columns. Then for  $l=1 \dots r$  the  $l$ -th row looks like  $(e_{l_i})^T$ . Now use column interchanges to bring all pivot columns to the left, in the form  $\left[ \begin{array}{c|c} I_r & 0 \\ \hline 0 & 0 \end{array} \right]$ .

$$\text{Thus } \text{rk}(A) = \text{rk}(QAP) = \text{rk} \left( \left[ \begin{array}{c|c} I_r & 0 \\ \hline 0 & 0 \end{array} \right] \right) = r$$

$$\text{nullity}(A) = \text{nullity}(QAP) = \text{nullity} \left( \left[ \begin{array}{c|c} I_r & 0 \\ \hline 0 & 0 \end{array} \right] \right) = n - r$$

Thus  $\text{rk}(A) + \text{nullity}(A) = n$  for  $m \times n$  matrix  $A$

**RANK-NULLITY THEOREM**

Another def<sup>n</sup>:  $v_1, v_2, \dots, v_k \in \mathbb{R}^n$  are called lin indep if matrix  $A_{n \times k} = [v_1 \ v_2 \ \dots \ v_k]$  has  $\text{rank } k \Leftrightarrow (A x = 0 \Rightarrow x = 0)$

$$\Leftrightarrow \left( \sum_{i=1}^k c_i \vec{v}_i = 0 \Rightarrow \text{all } c_i = 0 \right)$$