

Lecture XXI: §3.7 Linear Transformations from  $\mathbb{R}^n$  to  $\mathbb{R}^m$  1

GOAL: Study functions between subspaces of  $\mathbb{R}^n$  &  $\mathbb{R}^m$  that respect the vector space structures on both sides (addition & scalar multiplication)

Ex. Examples ①  $F: \mathbb{R}^3 \rightarrow \mathbb{R}$   $F(\underline{x}) = F\left(\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}\right) = x_1 + 5x_3$  linear expression in  $x_1, x_2, x_3$   
 $= x_1 + 0 \cdot x_2 + 5x_3$

F linear:

$F(\underline{u} + \underline{v}) = F\left(\begin{bmatrix} u_1 \\ u_2 \\ u_3 \end{bmatrix} + \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix}\right) = F\left(\begin{bmatrix} u_1 + v_1 \\ u_2 + v_2 \\ u_3 + v_3 \end{bmatrix}\right) = u_1 + v_1 + 5(u_3 + v_3)$   
 $= (u_1 + 5u_3) + (v_1 + 5v_3) = F(\underline{u}) + F(\underline{v})$

$F(\alpha \underline{u}) = F\left(\begin{bmatrix} \alpha u_1 \\ \alpha u_2 \\ \alpha u_3 \end{bmatrix}\right) = \alpha u_1 + 5(\alpha u_3) = \alpha(u_1 + 5u_3) = \alpha F(\underline{u})$   
 $\alpha$  scalar in  $\mathbb{R}$   
 $\underline{u}$  in  $\mathbb{R}^3$

Notice:  $F(\underline{x}) = \begin{bmatrix} 1 & 0 & 5 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$

We can restrict  $F$  to a line  $L$  or a plane  $P$  through the origin in  $\mathbb{R}^3$ , and get 2 functions  $F: L \rightarrow \mathbb{R}$ ,  $F: P \rightarrow \mathbb{R}$  linear.

② We can get a function  $G: \mathbb{R}^3 \rightarrow \mathbb{R}^2$  by assigning  $F_1: \mathbb{R}^3 \rightarrow \mathbb{R}$  to each of the 2 coordinates in the target  $\mathbb{R}^2$ . The result will be a linear transformation from  $\mathbb{R}^3$  to  $\mathbb{R}^2$ .  
 $\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \rightarrow \begin{bmatrix} F_1(\underline{x}) \\ F_2(\underline{x}) \end{bmatrix}$   $F_2: \mathbb{R}^3 \rightarrow \mathbb{R}$  (2 linear functions)

Eg:  $F_1(\underline{x}) = \begin{bmatrix} 1 & 0 & 5 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$

$F_2(\underline{x}) = 3x_1 - 7x_2 + 8x_3 = \begin{bmatrix} 3 & -7 & 8 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$

Then  $G(\underline{x}) = \begin{bmatrix} x_1 + 5x_3 \\ 3x_1 - 7x_2 + 8x_3 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 5 \\ 3 & -7 & 8 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \rightarrow$  multiplication by a  $(2 \times 3)$ -matrix.

$G\left(\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}\right) = \begin{bmatrix} 1 \\ 3 \end{bmatrix} = 1^{\text{st}}$  column of  $A$

$G\left(\begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}\right) = \begin{bmatrix} 0 \\ -7 \end{bmatrix} = 2^{\text{nd}}$  \_\_\_\_\_

$G\left(\begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}\right) = \begin{bmatrix} 5 \\ 8 \end{bmatrix} = 3^{\text{rd}}$  \_\_\_\_\_

So: values on the canonical basis  $e_1, e_2, e_3$  for  $\mathbb{R}^3$  determine the matrix  $A$

& so  $G\left(\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}\right) = G\left(x_1 \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + x_2 \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} + x_3 \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}\right) \stackrel{G \text{ linear}}{=} x_1 G\left(\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}\right) + x_2 G\left(\begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}\right) + x_3 G\left(\begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}\right) = A \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$

Example 2:  $G: \mathbb{R}^3 \rightarrow \mathbb{R}^2$   $G(\underline{x}) = \begin{bmatrix} 1 & 0 & 5 \\ 2 & 0 & 10 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} x_1 + 5x_3 \\ 2x_1 + 10x_3 \end{bmatrix}$  [2]

Note:  $G(\underline{x}) = \begin{bmatrix} \text{"scalar"} \ x_1 + 5x_3 \\ 2(x_1 + 5x_3) \end{bmatrix} = (x_1 + 5x_3) \begin{bmatrix} 1 \\ 2 \end{bmatrix}$

So points in the image lie in the line [with direction  $\begin{bmatrix} 1 \\ 2 \end{bmatrix}$ ]  
 $\Rightarrow$  We can write  $G: \mathbb{R}^3 \rightarrow L$  (image is in the subspace  $L$  of  $\mathbb{R}^2$ )

### §2 General definition:

Def: Let  $V$  be a subspace of  $\mathbb{R}^n$  &  $W$  be a subspace of  $\mathbb{R}^m$ , Consider a function  $T$  from  $V$  to  $W$ :

$$T: V \rightarrow W$$

We say  $T$  is a linear transformation if for all  $\underline{u}$  &  $\underline{v}$  in  $V$  & all scalars  $\alpha$  in  $\mathbb{R}$  we have:

$$(1) T(\underline{u} + \underline{v}) = T(\underline{u}) + T(\underline{v})$$

$$(2) T(\alpha \underline{u}) = \alpha T(\underline{u})$$

Remark (later) Every linear transformation  $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$  corresponds to multiplication by a matrix  $A$  of size  $m \times n$ , namely  $T\left(\begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}\right) = A \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$

We will give a similar definition for abstract vector spaces, but the remark no longer holds necessarily, only works when the spaces involve finite bases (hence we can identify them with subspaces of  $\mathbb{R}^n$  &  $\mathbb{R}^m$  via coordinates with respect to a basis  $B$  of  $V$  & a basis  $B'$  of  $W$  & let  $T: V \rightarrow W$  linear)

Non-example:  $F: \mathbb{R}^2 \rightarrow \mathbb{R}^3$   $F(\underline{x}) = \begin{bmatrix} x_1 - x_2 + 1 \\ x_2 \\ -x_2 \end{bmatrix}$  is NOT linear

$$F\left[\begin{bmatrix} 1 \\ 0 \end{bmatrix}\right] = \begin{bmatrix} 2 \\ 0 \\ 0 \end{bmatrix}, \quad F\left[\begin{bmatrix} 0 \\ 1 \end{bmatrix}\right] = \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix}, \quad F\left[\begin{bmatrix} 1 \\ 1 \end{bmatrix}\right] = \begin{bmatrix} 1 \\ 1 \\ -1 \end{bmatrix}$$

$$\text{BUT } F\left[\begin{bmatrix} 1 \\ 0 \end{bmatrix}\right] + F\left[\begin{bmatrix} 0 \\ 1 \end{bmatrix}\right] = \begin{bmatrix} 2 \\ 1 \\ -1 \end{bmatrix} \neq \begin{bmatrix} 1 \\ 1 \\ -1 \end{bmatrix}$$

What's wrong? The non-linear factor in the first coordinate!

Non-example 2:  $F: \mathbb{R} \rightarrow \mathbb{R}$   $F(x) = e^x$  is non-linear.

$$F(0) = 1, F(1) = e \quad \text{But} \quad F(0) + F(1) = 1 + e \neq e = F(0+1)$$



Prop:  $F: \mathbb{R} \rightarrow \mathbb{R}$  linear if and only if  $F(x) = ax$  for a fixed scalar  $a$ . □

Why? Use Property (2) in the definition:

$$F(x) = x \underbrace{F(1)}_a = ax$$

↙ scalar

↪ clearly linear!  

$$\begin{cases} F(x+y) = a(x+y) = F(x) + F(y) \\ F(\alpha x) = a(\alpha x) = \alpha F(x) \end{cases}$$

Prop 2:  $F: \mathbb{R}^n \rightarrow \mathbb{R}$  linear if and only if  $F(x) = u^T x$  for some vector  $u$  in  $\mathbb{R}^n$ .

Why? We can explicitly find  $u$ !

$$\begin{aligned} F\left(\begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}\right) &= F\left(x_1 \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} + \dots + x_n \begin{bmatrix} 0 \\ \vdots \\ 1 \end{bmatrix}\right) \\ &\stackrel{F \text{ linear}}{=} F\left(x_1 \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}\right) + F\left(x_2 \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix}\right) + \dots + F\left(x_n \begin{bmatrix} 0 \\ \vdots \\ 1 \end{bmatrix}\right) \\ &\stackrel{\text{↙}}{=} x_1 \underbrace{F\left(\begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}\right)}_{=u_1} + x_2 \underbrace{F\left(\begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix}\right)}_{=u_2} + \dots + x_n \underbrace{F\left(\begin{bmatrix} 0 \\ \vdots \\ 1 \end{bmatrix}\right)}_{=u_n} \\ &= [u_1 \ u_2 \ \dots \ u_n] \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = \underline{u}^T \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \end{aligned}$$

Note:  $\underline{u}^T = [F(e_1) \ \dots \ F(e_n)]$  values of  $F$  at the canonical basis of  $\mathbb{R}^n$  □  
 $\{e_1, \dots, e_n\}$

For a general  $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$  linear, we can build it from  $m$  functions

$$\begin{cases} T_1: \mathbb{R}^n \rightarrow \mathbb{R} \text{ linear} \\ \vdots \\ T_m: \mathbb{R}^n \rightarrow \mathbb{R} \text{ linear} \end{cases} \rightsquigarrow \begin{cases} T_1(x) = v_1^T \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \text{ for a vector } v_1 \text{ in } \mathbb{R}^n \\ \vdots \\ T_m(x) = v_m^T \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \text{ } \underline{\hspace{2cm}} \text{ } v_m \text{ in } \mathbb{R}^n \end{cases}$$

So  $T(x) = \begin{bmatrix} v_1^T \\ \vdots \\ v_m^T \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$

We conclude:

Thm: Every linear transformation  $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$  is obtain as  $m \times n$  matrix

$$\begin{cases} T(x) = A x \text{ for an } (m \times n) \text{ matrix.} \\ \text{Moreover: } \begin{matrix} 1^{\text{st}} \text{ column } (A) = T\left(\begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}\right) \\ \vdots \\ n^{\text{th}} \text{ } (A) = T\left(\begin{bmatrix} 0 \\ \vdots \\ 1 \end{bmatrix}\right) \end{matrix} \text{ so } A = \begin{bmatrix} \text{col 1} & & & \\ \downarrow & & & \\ T(e_1) & \dots & & \\ & & \text{col n} & \\ & & \downarrow & \\ & & T(e_n) & \end{bmatrix} \end{cases}$$

Notice: Only thing we use about  $\{e_1, \dots, e_n\}$  is the basis property.

Exercise: Find a transformation  $F: \mathbb{R}^2 \rightarrow \mathbb{R}^3$  such that

Sol:  $F\left(\begin{bmatrix} 1 \\ 1 \end{bmatrix}\right) = \begin{bmatrix} 1 \\ 4 \\ 3 \end{bmatrix}$ ,  $F\left(\begin{bmatrix} 1 \\ 2 \end{bmatrix}\right) = \begin{bmatrix} 1 \\ 0 \\ 5 \end{bmatrix}$ . Is  $F$  unique?

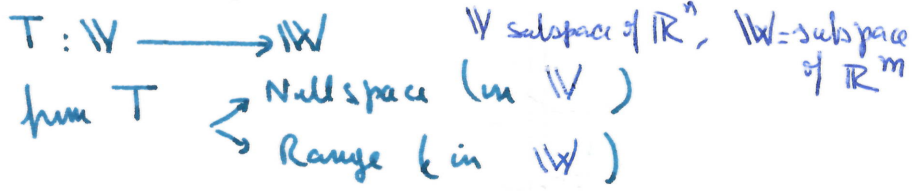
$e_1 = 2\begin{bmatrix} 1 \\ 1 \end{bmatrix} - \begin{bmatrix} 1 \\ 2 \end{bmatrix} \implies F(e_1) = 2F\left(\begin{bmatrix} 1 \\ 1 \end{bmatrix}\right) - F\left(\begin{bmatrix} 1 \\ 2 \end{bmatrix}\right)$

$e_2 = \begin{bmatrix} 1 \\ 2 \end{bmatrix} - \begin{bmatrix} 1 \\ 1 \end{bmatrix} \implies F(e_2) = F\left(\begin{bmatrix} 1 \\ 2 \end{bmatrix}\right) - F\left(\begin{bmatrix} 1 \\ 1 \end{bmatrix}\right) = \begin{bmatrix} 1 \\ 0 \\ 5 \end{bmatrix} - \begin{bmatrix} 1 \\ 4 \\ 3 \end{bmatrix} = \begin{bmatrix} 0 \\ -4 \\ 2 \end{bmatrix}$

Conclusion:  $F\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = \begin{bmatrix} 1 & 0 \\ 8 & -4 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} x \\ 8x-4y \\ x+2y \end{bmatrix} \implies$  so it is UNIQUE.

Check:  $F\left(\begin{bmatrix} 1 \\ 1 \end{bmatrix}\right) = \begin{bmatrix} 1 \\ 4 \\ 3 \end{bmatrix} \checkmark$ ,  $F\left(\begin{bmatrix} 1 \\ 2 \end{bmatrix}\right) = \begin{bmatrix} 1 \\ 0 \\ 5 \end{bmatrix} \checkmark$

§3 Null Space & Range:



- Def:  
• The Nullspace of  $T$  is  $\mathcal{N}(T) = \{v : v \text{ in } V \text{ \& } T(v) = 0\}$   
• The Range of  $T$  is  $\mathcal{R}(T) = \{w : w \text{ in } W \text{ \& } w = T(v) \text{ for some } v \text{ in } V\}$

Def: If  $V = \mathbb{R}^n$ ,  $W = \mathbb{R}^m$ , then  $T(x) = Ax$  for some  $(n \times m)$  matrix  $A$   
&  $\mathcal{N}(T) = \mathcal{N}(A)$  &  $\mathcal{R}(T) = \mathcal{R}(A)$ .

- dimension of  $\mathcal{R}(T) =: \text{rank}(T)$
- dimension of  $\mathcal{N}(T) =: \text{nullity}(T)$ .

Thm: If  $V = \mathbb{R}^n$  &  $W = \mathbb{R}^m$ ,  $\text{rank}(T) + \text{nullity}(T) = n$ .

Example above:

$T: \mathbb{R}^2 \rightarrow \mathbb{R}^3$

•  $\text{rank}(T) = \text{rank}(A) = 2$  ( $A = \begin{bmatrix} 1 & 8 & 1 \\ 0 & -4 & 2 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 5 \\ 0 & 1 & -\frac{1}{2} \end{bmatrix}$  REF  $\implies \text{rank} = 2$ . (2 non zero rows!))

•  $\text{nullity}(T) = \dim \mathcal{N}(T) = \dim \{x : Ax = 0\} = 0$  (so the system  $\begin{cases} x = 0 \\ 8x - 4y = 0 \\ x + 2y = 0 \end{cases}$  has unique solution  $= \begin{bmatrix} 0 \\ 0 \end{bmatrix}$  so  $\text{nullity} = 0$ )

•  $2 + 0 = 2$  ( $= n$ ).