Linear Algebra I

Fall 2023

## § 8 Subspaces, Kernel & Image

In the previous lectures we considered subsets of  $\mathbb{R}^n$ , which arised when studying linear maps . For example, for a linear map  $F:\mathbb{R}^n \to \mathbb{R}^n$  we calculated the image  $\operatorname{in}(F) \subset \mathbb{R}^m$ . In the case m=3 we saw that the image could be everything  $(\mathbb{R}^3)$ , a plane, a line or just a point  $\underbrace{O \in \mathbb{R}^n}_{X\mapsto (3)}$ . These sets are examples of <u>subspaces</u>, which we define in a bit more abstract way by the following:

Definition 8.1 A subset 
$$U \subset \mathbb{R}^n$$
 is a subspace of  $\mathbb{R}^n$  if  
i)  $O \in \mathbb{R}^n$  (again: by O we mean  $\binom{9}{3}\binom{1}{7}$ )  
ii) For all  $u_1 v \in U$ :  $U + v \in U$  "U is closed under addition"  
iii) Fo all  $u \in U$  and  $\lambda \in \mathbb{R}$ :  $\lambda u \in U$  "U is closed under   
scalar multiplication"

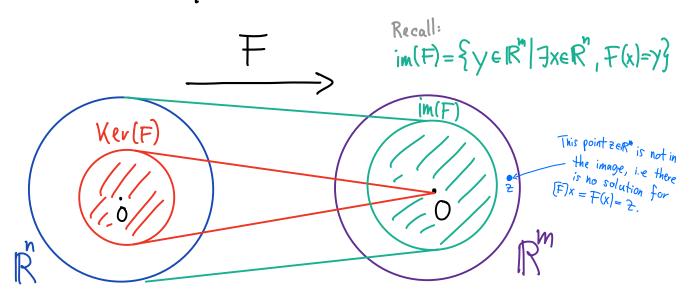
Example 29

1) 
$$U = \{0\}$$
 and  $U = |\mathbb{R}^n \text{ are always}$   
Subspaces of  $|\mathbb{R}^n \text{ for all } n \ge 1$ .  
2) Subspaces of  $|\mathbb{R}^n$ :  
 $n = 1$ :  $\{0\}$ ,  $\mathbb{R}$   
 $n = 2$ :  $\{0\}$ ,  $|\mathbb{R}^n|$   
 $n = 3$ :  $\{0\}$ ,  $|\mathbb{R}^n|$   
 $n = 3$ :  $\{0\}$ ,  $|\lim_{V \neq 0} \{|X| + |X| - |X| - |X|\}$   $|\mathbb{R}^2$   
 $n = 3$ :  $\{0\}$ ,  $|\lim_{V \neq 0} \{|X| + |X| - |X| - |X|\}$   $|\mathbb{R}^2$   
 $n = 3$ :  $\{0\}$ ,  $|\lim_{V \neq 0} \{|X| + |X| - |X| - |X|\}$   $|\mathbb{R}^3$   
3)  $U = \{x \in |\mathbb{R}^3 \mid X_1 + |X| - |X| - |X| - |X|\}$  is not a  
subspace because  $0 \notin U$ .  
4)  $U = \{|X| \in |\mathbb{R}^3| - 1| \le |X| \le 1\}$  is also not a  
subspace. We have  $(\frac{1}{2}) \in U$  but  $2 \cdot (\frac{1}{2}) \notin U$ .

A lot of subspaces come from linear maps (adually all of them). We will see that the image of a linear map is a subspace. Another important space coming from a linear map is its Kernel.

Definition 8.2 For a lin. map 
$$F:\mathbb{R}^n \to \mathbb{R}^n$$
 the  
Kernel of F is defined by  
 $Ker(F) = \{x \in \mathbb{R}^n \mid F(x) = 0\}.$ 

In other words: The Kernel of a linear map is the Set of all solutions of the linear system [F]X=0.



Proposition 8.3 Let F: IR"-> IR" be a linearmap. i) The Kernel Ker(F) is a subspace of IR". ii) The image im(F) is a subspace of IR"

Proof: To show that a subset U is a sahspace  
We need to check the 3 conditions  
1) 
$$O \in \mathbb{R}^{n}$$
  
1i) For all  $u_{n,v \in U}$ :  $u_{tv \in U}$   
1ii) For all  $u_{n,v \in U}$ :  $u_{tv \in U}$   
1ii) For all  $u_{n,v \in U}$ :  $u_{tv \in U}$   
1ii) We learned before that for any linear map  
we have  $F(0)=0$ .  
 $\Rightarrow O \in Ker(F)$ .  
1ii) Let  $u_{1}v \in Kev(F)$ . Then we have  
 $F(u_{tv}) = F(u) + F(v) = O + O = O$   
 $\Rightarrow U + v \in Kev(F)$ .  
1ii) Let  $u \in Kev(F)$ .  
1iii) Let  $u \in Kev(F)$ .  
1iii) Let  $u \in Kev(F)$ .  
1iii) Let  $u \in Kev(F)$ .  
 $F(\lambda u) = \lambda F(u) = \lambda O = O$   
 $\Rightarrow \lambda u \in Ker(F)$ .  
 $\Rightarrow Ker(F)$  is a subspace of  $\mathbb{R}^{n}$ .  
 $\underbrace{im(F)}$ : 1) Since  $F(0)=O$  we have  $O \in im(F)$ .  
1i) Let  $u_{1}v \in im(F)$ , i.e.  $u=F(x), v=F(y)$  for  
some  $x, y \in \mathbb{R}^{n}$ . Then we have  
 $U+v = F(x) + F(y) = F(x+y)$   
 $\Rightarrow U+v \in im(F)$ .

Example 30 1) Let 
$$u \in \mathbb{R}^{n}, u \neq 0$$
.  
ovthosonal proj:  $P_{u}: \mathbb{R}^{n} \longrightarrow \mathbb{R}^{n}$   
The Kernel of  $P_{u}$  is given  
by all vectors  $v \in \mathbb{R}^{n}$  with  
 $u \cdot v = 0$ , because  
 $P_{u}(v) = \frac{u \cdot v}{u \cdot u} u = 0$ .  
 $N = 2: Ker(P_{u}) = line$ ,  $N = 3: Ker(P_{u}) = plane$ 

2) Consider the linear map  

$$F: \mathbb{R}^2 \longrightarrow \mathbb{R}^3$$
  
 $\chi \longmapsto \begin{pmatrix} 1 & 1 \\ 2 & 1 \end{pmatrix} \chi$ 

## <u>Kernel:</u>

$$\frac{\text{Image:}}{\text{Check for which } Y \in \text{IR we have a } X \in \mathbb{R}^{2} \text{ with}}_{F(x) = Y}.$$

$$\left( [F] \begin{vmatrix} Y_{1} \\ Y_{2} \end{vmatrix} = \frac{G}{2} \begin{pmatrix} i & | & Y_{1} \\ 2 & | & Y_{2} \\ 0 & | & Y_{3} \end{pmatrix} \sim \begin{pmatrix} 1 & | & Y_{1} \\ 0 & | & Y_{2} - 2Y_{1} \\ 0 & | & Y_{2} - 2Y_{1} \end{pmatrix} \sim \begin{pmatrix} 0 & | & 1 & | & Y_{1} \\ 0 & -1 & | & Y_{2} - 2Y_{1} \\ 0 & | & Y_{3} + Y_{2} - 2Y_{1} \end{pmatrix} \sim \begin{pmatrix} 1 & 0 & | & -Y_{1} + Y_{2} \\ 0 & | & Y_{3} + Y_{2} - 2Y_{1} \end{pmatrix} \sim \begin{pmatrix} 1 & 0 & | & -Y_{1} + Y_{2} \\ 0 & | & 2Y_{1} - Y_{2} \\ 0 & | & -2Y_{1} + Y_{2} \\ -2Y_{1} + Y_{2} + Y_{3} \end{pmatrix} \sim \begin{pmatrix} 1 & 0 & | & -Y_{1} + Y_{2} \\ 0 & | & 2Y_{1} - Y_{2} \\ 0 & | & -2Y_{1} + Y_{2} + Y_{3} \end{pmatrix}$$

$$=) \quad \text{im} (F) = \begin{cases} Y_{1} \begin{pmatrix} Y_{1} \\ Y_{2} \end{pmatrix} \in \mathbb{R}^{3} \\ Y_{3} \in \mathbb{R}^{3} \end{cases} \quad Y_{3} = 2Y_{1} - Y_{2} \end{cases}.$$

(Notice: This calculation can also be used to  
calculate the kernel by setting 
$$Y_{1}=Y_{1}=Y_{2}=0$$
)  
We can also write the image as follows:  
 $im(F) = \{\lambda_{1} \begin{pmatrix} 0\\ 2 \end{pmatrix} + \lambda_{2} \begin{pmatrix} 0\\ -1 \end{pmatrix} \mid \lambda_{11}\lambda_{2} \in \mathbb{R}\}$   
3) Consider the linear map  
 $G: \mathbb{R}^{4} \rightarrow \mathbb{R}^{2}$   
 $X \longmapsto \begin{pmatrix} 1 & 2 & 0 & 1 \\ 1 & 1 & 1 & 1 \end{pmatrix} X$   
 $\underbrace{Vernel}: Solve [G] X = O$   
 $(G] \begin{pmatrix} 0\\ 0 \end{pmatrix} = \underbrace{G} \begin{pmatrix} 1 & 2 & 0 & 1 & 0 \\ 1 & 1 & 1 & 1 & 0 \end{pmatrix} \xrightarrow{P} \begin{pmatrix} 1 & 2 & 0 & 1 & 0 \\ 0 & -1 & 1 & 0 & 0 \end{pmatrix}$   
 $\sim \begin{pmatrix} 1 & 0 & 2 & 1 & 0 \\ 0 & 1 - 1 & 0 & 0 \end{pmatrix}$   
Solution:  $X_{1} = -2t_{1} - t_{2}$   
 $X_{4} = t_{2}$ 

Another way of writing:  

$$X = t_{1} \begin{pmatrix} -2 \\ 1 \\ 0 \end{pmatrix} + t_{2} \begin{pmatrix} -1 \\ 0 \\ 0 \\ 1 \end{pmatrix}.$$

Therefore:  $Ver(G) = \{ t_1(-2) + t_2(-2) \} \ t_1, t_2 \in \mathbb{R}^2 \}.$ 

Check yourself:  $im(G) = \mathbb{R}^2$ 

span 
$$\{V_1, \dots, V_n\} = \{\lambda_1, V_1 + \dots + \lambda_n, V_n \in \mathbb{R}^n \mid \lambda_1, \dots, \lambda_n \in \mathbb{R}\}$$
.

Example 31  
Since 
$$\begin{pmatrix} 1\\ 2\\ 3 \end{pmatrix} = 2 \begin{pmatrix} 0\\ 0\\ 1 \end{pmatrix} + \begin{pmatrix} 1\\ 3\\ 3 \end{pmatrix}$$
.  
 $\Rightarrow \begin{pmatrix} 1\\ 2\\ 3 \end{pmatrix} \in \text{Span} \left\{ \begin{pmatrix} 0\\ 0\\ 2 \end{pmatrix} + \begin{pmatrix} 1\\ 3\\ 3 \end{pmatrix} \right\}$ .  
2) Any  $x = \begin{pmatrix} x_1\\ 3x_1 \end{pmatrix}$  is a lin. comb. of  $e_{11..., e_n}$ :  
 $\begin{pmatrix} 1\\ 2\\ 3 \end{pmatrix} = x_1 e_1 + \dots + x_n e_n$   
and therefore  $\mathbb{R}^n = \text{Span} \left\{ e_{11..., e_n} \right\}$ .  
3) In Example 16 we have  
 $\inf(\mathbb{F}) = \text{Span} \left\{ \begin{pmatrix} 1\\ 2\\ 0 \end{pmatrix}, \begin{pmatrix} 0\\ -1\\ 1 \end{pmatrix} \right\}$ .  
 $\text{Notice}: \quad \text{If } \mathbb{F}: \mathbb{R}^n \rightarrow \mathbb{R}^n \text{ with } A = \begin{pmatrix} 1\\ 1\\ 1\\ 1 \end{pmatrix} \begin{pmatrix} 1\\ 1\\ 1\\ 1 \end{pmatrix}$   
then  $\inf(\mathbb{F}) = \text{Span} \left\{ v_{1, \dots, 1} v_n \right\}$ .