

Dominic Morchese  
Math 305  
Homework #2

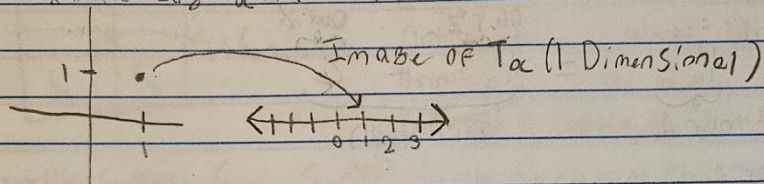
### Question #1

a) Rank is the Dimension of a linear maps Image,

So the rank of  $T_A: \mathbb{R}^n \rightarrow \mathbb{R}^1$ ,  $\text{rank}(T_A) = 1$

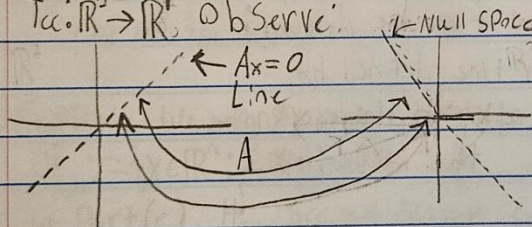
Nullity is the Dimension of the null space of our linear map  $T_A$ , So  $\text{nullity}(A) = n - \text{rank}(T_A) = n - 1$

For example say  $T_A: \mathbb{R}^2 \rightarrow \mathbb{R}^1$



Since we know  $\text{rank}(T_A) + \text{nullity}(T_A) = n$ , and we know  $n = 2$  in our example, then  $\text{nullity}(T_A) = 1$  for our mapping example.

b) Noting that null space =  $\{x \in \mathbb{R}^n : Ax = 0\}$ , Supposing  $T_A: \mathbb{R}^2 \rightarrow \mathbb{R}^1$ , Observe:



Imagine compressing the whole line down into a single point, the origin ( $\text{nullity}(A) = 1$ , and  $\text{rank}(A) = 1$ )

c) Assume that  $P_{A,b} := \{x \in \mathbb{R}^n : \sum_{i=1}^n a_i x_i = b\} \forall b \in \mathbb{R}$ , Suppose  $C \in \mathbb{R}^n$

for any  $P_{A,b}$  in the language of linear algebra we

have  $P_{A,b} = \{x \in \mathbb{R}^n : Ax = b\}$ , thus we can say that

for any  $b \in \mathbb{R}$ ,  $P: \mathbb{R}^n \rightarrow \mathbb{R}^1$ . Furthermore, we can observe that

$P_{A,b}$ , where  $b=0$ , is equivalent to  $P_{A,0} = \{x \in \mathbb{R}^n : Ax = 0\}$  which

is precisely the definition of the null space. Therefore,

since  $C \in \mathbb{R}^n$ , by the following property of a vector in the null space ( $Ax = A(x+z) = b$ , for any vector in the null space)

the we can say  $P_{A,b} = P_{A,0} + C$ , for any  $C \in \mathbb{R}^n$ , since

$P_{A,0} \in \text{nullspace}$ .

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# Question #1 - Continued

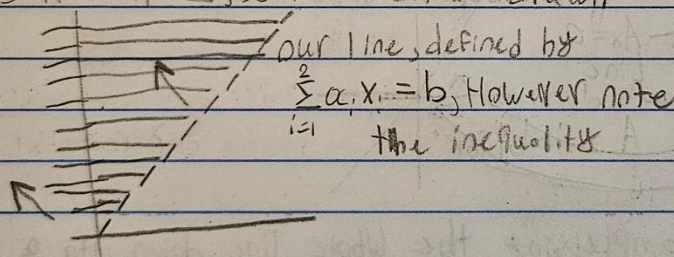
c) Since my explanation was verbose let me restate the reasoning:

For any linear map, call it  $A$ , any vector in  $\text{null}(A)$ , call it  $z$ , has the following property  $Ax = b \Rightarrow A(x+z) = b$ . Using that reasoning:

$$\underbrace{P_{A,b}}_{\text{our } Ax} = \underbrace{P_{A,0}}_{\text{our } z} + \underbrace{C}_{\text{our } x}$$

Note then that since  $P_{A,0} \in \text{null space}$ , if  $P_{A,b}$  such that  $b \neq 0$ , then  $C$  must also  $C \neq 0$ , our desired result.

d) Supposing  $H_{A,b} := \{x \in \mathbb{R}^n : \sum_{i=1}^n \alpha_i x_i \geq b\}$ , observe the following. Suppose  $n=2$ , so that it can be drawn:



Note that any region above the line, our at it, defines  $H_{A,b}$ . Expanding to higher dimensions, notice what we are doing more generally is separating  $\mathbb{R}^n$  into two parts. One part at or above the hyper plane defined by  $H_{A,b}$ , and the part strictly below the hyper plane.

e) Suppose  $H_{A,b} := \{x \in \mathbb{R}^n : \sum_{i=1}^n \alpha_i x_i \geq b\}$ . For any two points (chosen in  $H_{A,b}$ , say  $L$  and  $K$ ), if  $H_{A,b}$  is convex then  $\theta L + (1-\theta)K \in H_{A,b} \forall \theta \in [0,1]$ . Then,  $\theta \sum_{i=1}^n \alpha_i L_i \geq \theta b$  and  $(1-\theta) \sum_{i=1}^n \alpha_i K_i \geq (1-\theta)b$ , then we can observe  $\theta \sum_{i=1}^n \alpha_i L_i + (1-\theta) \sum_{i=1}^n \alpha_i K_i = \sum_{i=1}^n \alpha_i (\theta L_i + (1-\theta)K_i) \geq \theta b + (1-\theta)b = b$

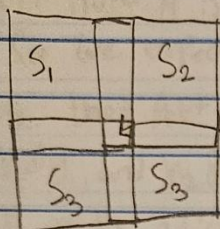
Thus, we have the desired result. ■



# Question #1-Continued

f) Suppose  $S_1, S_2, \dots, S_m \in \mathbb{R}^n$  are convex. First consider the geometric reasoning behind  $S_1 \cap S_2 \cap \dots \cap S_m$  being convex.

Consider 4 boxes (A convex shape)



$S_1 \cap S_2 \cap S_3 \cap S_4$  Notice how no matter we trim a set, if the portion removed is convex, the result is convex

Formally, Suppose  $S_1, \dots, S_m \in \mathbb{R}^n$ . Using an element chasing sort of approach, consider  $\vec{x}, \vec{y} \in S_1$ . Since  $S_1$  is convex  $\theta \vec{x} + (1-\theta) \vec{y} = \vec{z} \in S_1, \forall \theta \in [0,1]$  by definition. Thus, if  $\vec{z} \in S_2$ , and  $S_2$  is convex then  $\vec{z} \in S_1$  and  $\vec{z} \in S_2$ , which implies  $\vec{z} \in S_1 \cap S_2$ .

We can extend this reasoning to any  $S_1 \cap S_2 \cap \dots \cap S_m, \forall m \in \mathbb{N}$  giving us the desired result. ■

g) Suppose we are given a LP with constraints of the form

$$H_{a,b}^j := \{x \in \mathbb{R}^n : \sum_{i=1}^n a_i x_i \geq b\} \text{ for } j = \{1, \dots, m\}, \text{ by our reasoning}$$

in Part (e),  $H_{a,b}^j$  are all convex sets, and by our reasoning in Part (f), if all  $H_{a,b}^j$  sets are convex,

then  $H_{a,b}^1 \cap H_{a,b}^2 \cap \dots \cap H_{a,b}^m$  is a convex set. Since

the feasible region of our LP is the set of all points

satisfy all constraints, we can call our feasible region  $H_{a,b}^1 \cap \dots \cap H_{a,b}^m$ .

Since that set is convex, the feasible region of any LP with constraints of the form of  $H_{a,b}^j$  must be convex. ■



## Question #2 (Exercise 2-2-3)

$$A = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 3 & 1 & 3 & 0 \\ 1 & 3 & -3 & -8 \end{bmatrix}$$

	$x_1$	$x_2$	$x_3$	$x_4$
$y_1 =$	1	2	3	4
$T = y_2 =$	3	1	3	0
$y_3 =$	1	3	-3	-8

Jordan exchange as many  $y_i$  to the top, the number of  $y_i$  that can be exchanged is the number of linearly independent rows.

We can Jordan Exchange  $x_1$  &  $y_1$ ,  $x_2$  &  $y_2$ ,  $x_3$  &  $y_3$ , thus  
A has 3 linearly independent rows.

## Question #3 (Exercise 2-4-2)

We can find the solution  $X = (1, -1, 1)$  by first setting

up a Tableau

	$x_1$	$x_2$	$x_3$	1		$y_1$	$y_2$	$y_3$	1		
$y_1$	1	1	1	-1	$1 \times (T_{1,1})$	$x_1$	$1/2$	$1/2$	0	1	$\leftrightarrow x_1 = 1$
$T = y_2$	1	-1	-1	-1	$1 \times (T_{2,2})$	$x_2$	$1/2$	0	$-1/2$	-1	$\leftrightarrow x_2 = -1$
$y_3$	1	-1	1	-3	$1 \times (T_{3,3})$	$x_3$	0	$-1/2$	$1/2$	1	$\leftrightarrow x_3 = 1$

We then Jordan exchange All  $y_i$  to the top and All  $x_i$  to the side. We can then read off the 4<sup>th</sup> column as the associated  $x_i$  solution. Therefore, the solution  $X = (1, -1, 1)$ ,  
What was given in the beginning of the exercise.



# Question # 4 (Exercise 2-4-6)

$$1.) Ax = a \quad A = \begin{bmatrix} 2 & -1 & 1 & 1 \\ -1 & 2 & -1 & -2 \\ 4 & 1 & 1 & -1 \end{bmatrix} \quad a = \begin{bmatrix} 1 \\ 1 \\ 5 \end{bmatrix}$$

First note  $A$  is  $m \times n$  such that  $m=3, n=4$ , so  $m < n$ ,  
So we expect infinite solutions.

Note that After constructing tableau and exchanging  $x_1$  &  $y_1$ , and  $x_2$  &  $y_2$ , we cannot exchange  $y_3$  with either  $x_3$  or  $x_4$ , indicating that  $y_3$  is a linearly dependent row.

More particularly, noting the values in the tableau

$$y_3 = 3y_1 + 2y_2$$

Moreover, there are infinite solutions as expected, in the form

$$x_1 = \frac{1}{3}x_3 + 1$$

$$x_2 = \frac{1}{3}x_3 + x_4 + 1$$

Therefore,

$$x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 0 \\ 0 \end{bmatrix} + \begin{bmatrix} -1/3 \\ 1/3 \\ 1 \\ 0 \end{bmatrix} \lambda_1 + \begin{bmatrix} 0 \\ 1 \\ 0 \\ 1 \end{bmatrix} \lambda_2 \quad \text{Such that } \lambda_1, \lambda_2 \text{ are arbitrary}$$

$$2.) Bx = b \quad B = \begin{bmatrix} 1 & -1 & 1 & 2 \\ 1 & 1 & 0 & -1 \\ 1 & -3 & 2 & 5 \end{bmatrix} \quad b = \begin{bmatrix} 2 \\ 1 \\ 1 \end{bmatrix}$$

Again note  $B$  is  $3 \times 4$ , so we expect infinitely many solutions.

Note After constructing tableau and exchanging  $x_1$  &  $y_1$  and  $x_2$  &  $y_2$ ,

We find we can't exchange  $y_3$  with  $x_3$  or  $x_4$ . Thus,  $y_3$  is

$$\text{linearly dependent on } y_1 \text{ and } y_2; y_3 = 2y_1 - y_2 + 2$$

Since we have a constant in the last column, we

can see that we have no solution since we can't set

$$y_1 = y_2 = y_3 = 0, \text{ because we have } y_3 = 2y_1 - y_2 + 2.$$



# Question #4 - Continued (Exercise 2-4-6)

$$3.) Cx = c \quad C = \begin{bmatrix} 1 & -1 & 1 \\ 2 & 1 & 1 \\ -1 & -1 & 2 \\ 1 & 1 & -1 \end{bmatrix} \quad c = \begin{bmatrix} 3 \\ 2 \\ 2 \\ -1 \end{bmatrix}$$

Note  $C$  is  $4 \times 3$ , So we expect no solution, unless one or more rows of  $C$  are linearly dependent.

After constructing the tableau and exchanging  $x_1$  &  $y_1$ ,  $x_2$  &  $y_2$ , and  $x_3$  &  $y_3$ , we find

$$X = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$$

So we check to insure a solution

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

Thus we do have a solution  $X$ .



## Question #5 (Exercise 2-4-8)

1. In order for  $Ax=b$  to have a single solution, regardless of  $b$ , we need two things. First we need  $\text{RANK}(A)=m$  so that a solution exists for all  $b$ . Second, we need  $\text{RANK}(A)=n$  to insure the uniqueness of our solution. Therefore, we need  $\text{RANK}(A)=m=n$  to insure one solution regardless of  $b$ .

Ex:  $A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 2 & 2 & 0 \end{bmatrix}$  A  $3 \times 3$  matrix with  $\text{RANK}(A)=3$

2. In order for  $Ax=b$  to have zero or infinitely many solutions, depending on  $b$ , we want two things. First, we want at least one "free" variable, one whose value does not affect the solution. Second, we want our not free variables to be linearly independent. In the language of linear algebra, we want  $\text{RANK}(A) < m$  and  $\text{RANK}(A) < n$ .

Ex:  $A = \begin{bmatrix} 1 & 4 & 7 \\ 3 & 12 & 21 \end{bmatrix}$

3. In order to insure  $Ax=b$  has either zero or one solution, depending on  $b$ , we need two things. We need to insure matrix  $A$  has  $\text{RANK}(A)=n$  and  $n < m$ .  $\text{RANK}(A)=n$  is to insure uniqueness, and  $n < m$  is to insure zero solutions at some values of  $b$ .

Ex:  $A = \begin{bmatrix} 4 & 9 \\ 1 & 3 \\ 3 & 6 \end{bmatrix}$

### Question # 5 - Continued (Exercise 2-4-8)

4.) To guarantee infinitely many solutions to  $Ax=b$ , independent of  $b$ , we need two things. First,  $\text{rank}(A)=m$  to insure existence of a solution. Second,  $n > m$  to insure infinitely many solutions.

Ex.  $A = \begin{bmatrix} 2 & 1 & 3 \\ 3 & -1 & 2 \end{bmatrix}$