### LECTURE 5

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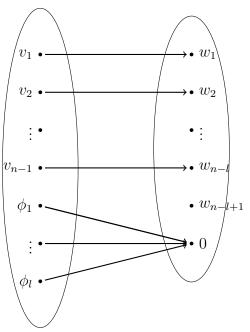
### 1. Rank Nullity Theorem

It is also called then fundamental theorem of Linear Maps because of its importance in linear transformation.

## 1.1. Statement and Proof.

**Theorem 1.** Rank-Nullity Theorem: Suppose V is finite-dimensional and  $T \in \mathcal{L} : (V, W)$ . Then range T is finite-dimensional and

$$Rank(T) + Nullity(T) = \dim(V), or$$
  
 $\dim(Range(T)) + \dim(Ker(T)) = \dim(V)$ 



Let  $\phi_1, \phi_2, \ldots, \phi_l$  be the minimum vectors will that span Ker(T), where l is the  $\dim(Ker(T))$ .

 $Andv_1, v_2, \ldots, v_{n-l}$  be the vectors that will span the remaining vector space V where, n is the  $\dim(V)$ . Their linear transformation

Proof.

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This paper is in final form.

 $(w_1, w_2, \ldots, w_{n-l})$  must be independent and should span the entire range of T in W, let  $\dim(Range(T))$  be k.

We will prove this claim later.

Then it becomes quite easy to see why the rank nullity theorem holds. Because,

$$n-l=k$$
 or  $n=l+k$ 

Hence, Proved.

Claim 1.  $w_1, w_2, \ldots, w_{n-l}$  are linearly independent and span the entire Range(T).

Proving Claim.  $w_1, w_2, \ldots, w_{n-l}$  are linear transformation of basis  $v_1, v_2, \ldots, v_{n-l}$ . So,

$$\alpha_1 w_1 + \alpha_2 w_2 + \ldots + \alpha_{n-l} w_{n-l} = 0$$

$$or, \sum_{i=1}^{n-l} \alpha_i w_i = 0$$

$$or, \sum_{i=1}^{n-l} \alpha_i T v_i = 0$$

$$or, T(\sum_{i=1}^{n-l} \alpha_i v_i) = 0$$

So  $\alpha_1 v_1 + \alpha_2 v_2 + \ldots + \alpha_{n-l} v_{n-l} \in \dim(nullT)$ . Since,  $\phi_1, \phi_2, \ldots, \phi_l$  spans null T, we can write:

$$\alpha_1 v_1 + \alpha_2 v_2 + \ldots + \alpha_{n-l} v_{n-l} = \beta_1 \phi_1 + \beta_2 \phi_2 + \ldots + \beta_l \phi_l$$

But we also know that  $\phi_1 to \phi_l$  and  $v_1 to v_{n-l}$  are linearly independent, which means the only way to satisfy above equation is to have  $\alpha_1 = \alpha_2 = \ldots = \alpha_{n-1} = \beta_1 = \ldots = \beta_l = 0$ .

ThusTv<sub>1</sub>,  $Tv_2, \ldots, Tv_{n-l}$  are linearly independent and span the entire Range(T). Hence  $w_1, w_2, \ldots, w_{n-l}$  will span the entire Range(T).  $\square$ 

1.2. **Applications.** We can now prove that some of the mapping can not be surjective or injective.

**Theorem 2** (A map to a smaller dimensional space is not injective). Suppose V and W are finite-dimensional vector spaces such that  $\dim(V)$  and  $\dim(W)$ , then there is no injective linear mapping from V to W.

*Proof.* Let  $T \in \mathcal{L}:(V, W)$ . Then

$$\dim(null T) = \dim(V) - \dim(Range(T))$$

$$\geq \dim(V) - \dim(W)$$

$$> 0$$

The equality above comes from the fundamental theorem of Linear Maps and the strict inequality at the end shows the dimension of T is greater than 0, means vector other than 0 are present in T. Hence the mapping is not injective.  $\Box$ 

**Theorem 3** (A map to a larger dimensional space is not surjective). Suppose V and W are finite-dimensional vector spaces such that  $\dim(V)$  and  $\dim(W)$ , then there is no surjective linear mapping from V to W.

*Proof.* Let  $T \in \mathcal{L}:(V, W)$ . Then

$$\dim(range T) = \dim(V) - \dim(null T)$$

$$\leq \dim(V)$$

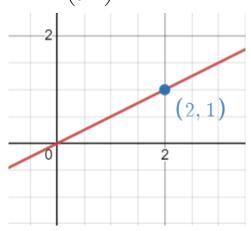
$$< \dim(W)$$

Here again the equality comes from the Rank-Nullity Theorem, and the rest is just normal inference.

At the end we have range of T ; the dimension of W. Hence the mapping is not surjective.  $\hfill\Box$ 

#### 2. Matrix Representation of Linear Transformation

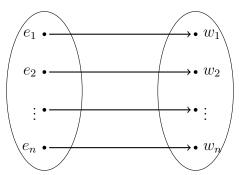
A vector is represented in different ways on the coordinate system depending on what its basis is. Suppose the vector (2,1) in  $\mathbb{R}^2$  is represented with basis as  $\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$ 



But when the basis is changed the vector is represented differently and on a different basis.

The vector V can be represented as  $a_1b_1 + a_2b_2$  and in matrix form can be represented as  $V\begin{pmatrix} a_1 \\ a_2 \end{pmatrix}$ .

Suppose there are is a linear transformation T that maps a vector V from  $R^n$  to W in  $R^m$ . Let the basis be  $B = \{e_1, \ldots, e_n\}$  and  $B' = \{e'_1, \ldots, e'_m\}$  The map can be represented as



Suppose the basis for B is  $\begin{pmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{pmatrix}$  and the upon applying the transfor-

mation the basis for B' is  $\begin{pmatrix} \beta_1 \\ \vdots \\ \beta_m \end{pmatrix}$ 

Now the question is how should we connect  $\alpha_i$  to the  $\beta_j$  such that we reach to our answer.

We know that Tv = w

$$v = \sum_{i=1}^{n} \alpha_i e_i$$

Applying linear transformation T on both sides we get

$$Tv = \sum_{i=1}^{n} \alpha_i T(e_i)$$

Now what is  $Te_i$ ?

In this case  $e_i \to w_i$ 

$$\therefore Tv = \sum_{i=1}^{n} \alpha_i w_i$$

But we want  $w_i$  to be represented in terms of Standard Basis.

$$w_i = \begin{pmatrix} w_{1i} \\ w_{2i} \\ \vdots \\ w_{mi} \end{pmatrix}$$
 (This is the representation of  $w_i$  in standard basis)

$$\therefore w_i = \sum_{i=1}^m w_{ij} e'_j$$

Hence now the Tv can be transformed as follows

$$Tv = \sum_{i=1}^{n} \sum_{j=1}^{m} w_{ij} e_{j}$$

$$Tv = \sum_{j=1}^{m} \left(\sum_{i=1}^{n} w_{ij} \alpha_{i}\right) e'_{j}$$

$$\text{Call the quantity } \left(\sum_{i=1}^{n} w_{ij} \alpha_{i}\right) = \beta_{j}$$

$$\sum_{j=1}^{m} \beta_{j} e'_{j}$$

This can now be represented in the matrix form as follows

$$Tv = \begin{pmatrix} w_{1,1} & w_{1,2} & \cdots & w_{1,n} \\ \vdots & \vdots & \vdots & \vdots \\ w_{m,1} & w_{m,2} & \cdots & w_{m,n} \end{pmatrix} \begin{pmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{pmatrix}$$

Here we can clearly see that  $Te_i = \begin{pmatrix} w_{1,i} \\ \vdots \\ w_{m,i} \end{pmatrix}$  That is the  $i^{\text{th}}$  column in the matrix.

Hence to get the first column and eventually  $Te_1$  we must place the

$$\begin{pmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{pmatrix}$$
 equal to  $\begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}$ . Similarly to get the second column we must

place the  $\begin{pmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{pmatrix}$  equal to  $\begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix}$  and so on.

**Example 1:**  $\mathbb{R}^2 \xrightarrow{T} \mathbb{R}^2$ . Given

$$\begin{pmatrix} 1 \\ 2 \end{pmatrix} \xrightarrow{T} \begin{pmatrix} -3 \\ 1 \end{pmatrix}$$
$$\begin{pmatrix} -3 \\ 1 \end{pmatrix} \xrightarrow{T} \begin{pmatrix} 1 \\ 2 \end{pmatrix}$$

Find the linear transformation  $\begin{pmatrix} \alpha_1 \\ \alpha_2 \end{pmatrix} \xrightarrow{T} \begin{pmatrix} \beta_1 \\ \beta_2 \end{pmatrix}$ .

**Solution:** The first question that arises is where the standard basis goes. The vector on the LHS be in the standard basis (1,0) and (0,1).

$$\begin{pmatrix} 1 \\ 2 \end{pmatrix} \xrightarrow{T} \begin{pmatrix} -3 \\ 1 \end{pmatrix}$$

$$-2 \cdot \begin{pmatrix} \begin{pmatrix} -3 \\ 1 \end{pmatrix} \xrightarrow{T} \begin{pmatrix} 1 \\ 2 \end{pmatrix} \end{pmatrix}$$

$$= \begin{pmatrix} 7 \\ 0 \end{pmatrix} \xrightarrow{T} \begin{pmatrix} -5 \\ -3 \end{pmatrix}$$

$$\implies \begin{pmatrix} 1 \\ 0 \end{pmatrix} \xrightarrow{T} \frac{1}{7} \begin{pmatrix} -5 \\ -3 \end{pmatrix}$$

Similarly we find for the vector  $\begin{pmatrix} 0 \\ 1 \end{pmatrix}$  and we get it to be  $\begin{pmatrix} 0 \\ 1 \end{pmatrix} \xrightarrow{T} \begin{pmatrix} -8 \\ 5 \end{pmatrix}$ .

Now we know the linear transformation for both the standard basis vectors. Now we can push these values in the matrix and get the value for the corresponding  $\beta$  depending on the value of  $\alpha$ .

$$\begin{pmatrix} \beta_1 \\ \beta_2 \end{pmatrix} = \begin{pmatrix} \frac{-5}{7} & \frac{-8}{7} \\ \frac{-3}{7} & \frac{5}{7} \end{pmatrix} \begin{pmatrix} \alpha_1 \\ \alpha_2 \end{pmatrix}$$

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