

## # Lecture 9 →

### □ spanning a space

we already know it right? vectors  $v_1, v_2, \dots, v_k$  span a space means the space consists of all combinations of those vectors.

For example, the columns of a matrix span the column space.

□ Basis for a space is a sequence of vectors  $v_1, v_2, v_3, \dots, v_k$  with 2 properties -

1. They are independent
2. They span the space.

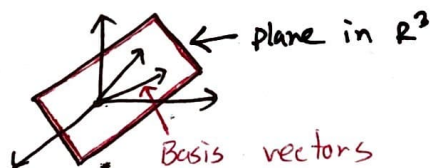
### Examples

For  $\mathbb{R}^3$ , one basis is  $\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$ . Why?

But this isn't the only basis? Another basis can be  $\begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 2 \\ 2 \\ 5 \end{bmatrix}, \begin{bmatrix} 3 \\ 3 \\ 8 \end{bmatrix}$ . Now if I took only two out of these ~~the~~ three, would that be a basis?

So, In  $\mathbb{R}^n$ ,  $n$  vectors give basis if the  $n \times n$  matrix with those columns is invertible.

Now, If I took  $\begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix}$  and  $\begin{bmatrix} 2 \\ 2 \\ 5 \end{bmatrix}$ , would they be a basis? Yes! They would be a basis for a subspace created by the plane (all linear combinations) in  $\mathbb{R}^3$ .



So, Basis is not unique. There can be a million possibilities of basis. But one thing common for all the basis are -

Every basis for the space has the same number of vectors.

So, In  $\mathbb{R}^6$ , if we got 7 vectors, that is too many for the basis, if we got 5 vectors, that would not be enough. So, 6 is just the perfect number.

And, this perfect number is the dimension of the space.

Example:

Any matrix,  $A = \begin{bmatrix} 1 & 2 & 3 & 1 \\ 1 & 1 & 2 & 1 \\ 1 & 2 & 3 & 1 \end{bmatrix}$

- (i) What is the span of these column vectors?  $\rightarrow C(A)$   
(ii) Are the columns independent?  $\rightarrow$  No. Because  $N(A)$  has a non-zero solution.  
(iii) What <sup>is</sup> ~~are~~ the basis of the column space?  $\rightarrow$  The first two columns / column 1 & 3  
(iv) What is the dimension of the column space?  $\rightarrow \text{rank}(A) = 2 = \# \text{ of pivots} = \text{dimension}$

$\dim C(A) = \text{rank}$

(v) What is the dimension of the nullspace?

$\rightarrow$  Give me two special solutions  $\rightarrow \begin{bmatrix} -1 \\ -1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ 0 \\ 1 \end{bmatrix} \rightarrow$  And these two are the basis of the nullspace.

This isn't the dimension of the matrix, but only the column space.

And,  $\dim N(A) = \# \text{ of free columns} = n - r = 2$

The four subspaces of  $R^n$   $\rightarrow$  Row Reduced Form  $\rightarrow$  Go to the last page

We are already familiar with 2 subspaces. Column space and Nullspace - these two ~~spaces~~ subspaces come from  $A$ . The other two subspaces, namely row space and left nullspace comes from  $A^T$ .

Let's know about all the subspaces through a specific example - Let's say we have a  $3 \times 5$  matrix  $A$  and its reduced to its row-echelon form  $R$ .

$R = \begin{bmatrix} \textcircled{1} & 3 & 5 & 0 & 7 \\ 0 & 0 & 0 & \textcircled{1} & 2 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$

$\uparrow$  Pivot column       $\uparrow$  Pivot column

$\leftarrow$  Pivot row       $\leftarrow$  Pivot row

$m = 3$   
 $n = 5$   
 $r = 2$

1. The column space  $C(A)$

As the  $\text{rank} = 2$ , dimension is also 2.

The pivot columns 1 & 4 form a basis for  $C(R)$ .

So, The dimension of the column space is  $\text{rank } r$ , & the pivot columns form a basis.

## 2. The Null Space $N(A)$

First, think about the special solutions of  $A$  that solves  $Ax=0$ . There are 3 free columns. So, setting the free columns to  $(1,0,0)$ ,  $(0,1,0)$ ,  $(0,0,1)$  gives the solution

$$x_{\text{special}} = \begin{bmatrix} -3 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} -5 \\ 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} -7 \\ 0 \\ 0 \\ -2 \\ 1 \end{bmatrix}$$

So, the dimension is  $n-r=5-2=3$ , and these 3  $x_{\text{special}}$  are a basis for the nullspace.

## 3. The row space $C(A^T)$

think about  $R^T$ . What is the rank? 2. And this is the dimension. And the first and the second row form a basis. Because the third row doesn't add anything with the row space.

So, The dimension of the row-space is  $r=2$ , The non-zero rows form a basis.

## 4. The Left-Nullspace $N(A^T)$

Let's say,  $Ay=0$   
 $\uparrow$  special solution

Now, if you transpose the whole thing -

$$\begin{aligned} (Ay)^T &= 0^T \\ \Rightarrow y^T A^T &= 0 \end{aligned}$$

$\uparrow$  Row space  
 $\uparrow$  Nullspace comes at left.

So, Left Nullspace is basically the nullspace of the row space

So,  $R^T = \begin{bmatrix} 1 & 0 & 0 \\ 3 & 0 & 0 \\ 5 & 0 & 0 \\ 0 & 1 & 0 \\ 7 & 2 & 0 \end{bmatrix}$ . So, How many special solution?

Only 1.

So, the basis would be only one vector. Which is?  $\begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$ . Dimension = 1.

So,

If  $A$  is  $m \times n$  matrix of rank  $r$ , its left nullspace has dimension  $m-r$ .

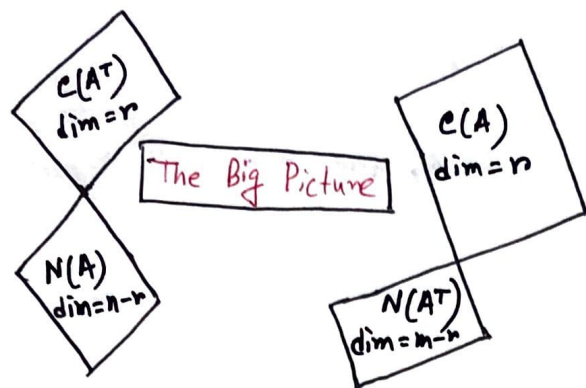


Summarizing - of a matrix  $m \times n$ ,

In  $\mathbb{R}^n$ , the row space and nullspace have dimensions  $r$  and  $n-r$ .

In  $\mathbb{R}^m$ , the column space and left-nullspace have dimensions  $r$  and  $m-r$ .

So, the big picture would be -



The Four Subspaces of  $A$ :

$$\begin{bmatrix} 1 & 2 & 3 & 1 \\ 1 & 1 & 2 & 1 \\ 1 & 2 & 3 & 1 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & 0 & 1 & 1 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$\textcircled{A}$   $\textcircled{R}$

See, the column space has changed!  $C(A) \neq C(R)$ .  $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$  is gone and you can't get it back by reversing the operations. But, their row space is same! Because, you only changed the rows by some row-operations!

$$EA = R \Rightarrow A = E^{-1}R$$

1.  $A$  has the same dimension as  $R$  and same basis. In fact,  $A$  has the same row-space as  $R$ .

2. The column space of  $A$  has dimension  $r \rightarrow$  rank. For Every matrix, the following is true -

The number of independent columns equals the number of independent rows.

Reason: The same combinations of the columns are zero for both  $A$  &  $R$ .  $AX=0$  exactly when  $RX=0$ .

③ A has the ~~same~~ nullspace as  $P$ . Same dimension  $n-r$  and ~~same~~ basis.

Reason:

③ A has the same nullspace as  $R$ .

Because nullspace is related to row space. Does elimination change the row space? If you think carefully, you can understand elimination steps actually ~~re~~ reveal which rows are dependent. so, as it doesn't change the row space, it wouldn't change the nullspace as well.

$$\begin{bmatrix} - & - & - & : & b_1 \\ - & - & - & : & b_1 - 2b_2 \\ - & - & - & : & b_3 - 2b_1 - b_2 \end{bmatrix}$$

so, the dimension would be  $n-r$  and the basis would be same.

④ The left nullspace is related to column space. so, it wouldn't be the same as well. The dimension would be  $m-r$ .

Example 1:  $A = \begin{bmatrix} 1 & 2 & 3 \end{bmatrix}$

$m=1, n=3, \text{rank}=1$

→ Row space is a line in  $\mathbb{R}^3$ .

⊗ ~~Geo~~ start with this matrix  $\rightarrow A = \begin{bmatrix} 2 & -1 & -3 \\ -4 & 2 & 6 \end{bmatrix}$

Columnspace:

$$\text{Span} = \left\{ \begin{bmatrix} 2 \\ -4 \end{bmatrix} \right\}, \text{Basis} = \begin{bmatrix} 2 \\ -4 \end{bmatrix}, \text{Dimension} = 1$$

Nullspace:

$$R = \left[ \begin{array}{ccc|c} 1 & -1/2 & -3/2 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right], \text{Span} = \left\{ \begin{bmatrix} 1/2 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 3/2 \\ 0 \\ 1 \end{bmatrix} \right\}, \text{Basis}, \text{Dimension} = 2.$$

$C(A^T)$ :

$$A^T = \begin{bmatrix} 2 & -4 \\ -1 & 2 \\ -3 & 6 \end{bmatrix}, \text{Span} = \left\{ \begin{bmatrix} 2 \\ -1 \\ -3 \end{bmatrix} \right\}, \text{Basis}, \text{Dimension} = 1.$$

↑

Can also say that this is the span of the row vector of  $A$ , or the rowspace of  $A$ ! so,  $C(A^T)$  is called the rowspace of  $A$ !

$N(A^T)$ :

$$\text{rref}(A^T) = \left[ \begin{array}{cc} 1 & -2 \\ 0 & 0 \\ 0 & 0 \end{array} \right], \text{Span} = \left\{ \begin{bmatrix} 2 \\ 1 \end{bmatrix} \right\}, \text{Basis}, \text{Dimension} = 1.$$

Now, If the nullspace of  $A^T$  is  $y$ ,

$$A^T y = 0$$

$$\Rightarrow (A^T y)^T = 0^T$$

$$\Rightarrow y^T A = 0^T$$

so, this is also called the left nullspace of  $A$ !

$\text{Rank}(A) = \text{Rank}(A^T) \rightarrow$  will always be the case.