

## W05-08

**Slide 01:** In this unit, we continue our discussion on the dimension of a vector space.

**Slide 02:** The following is a theorem which will be stated without proof. Let  $V$  be a vector space of dimension  $k$  and suppose  $S$  is a subset of  $V$ . The following three statements are logically equivalent, meaning that they are either all true or all false. The first statement is  $S$  is a basis for  $V$ .

(#)

The second statement states that  $S$  is a linearly independent set which contains exactly  $k$  vectors.

(#)

The third statement states that  $S$  spans the vector space  $V$  and it contains exactly  $k$  vectors.

(#)

What is the significance of this theorem and how can we use it?

**Slide 03:** Basically, the theorem allows us to make certain conclusions when we know that the dimension of the vector space is  $k$ .

(#)

Firstly, any subset  $S$  of  $V$  with the correct number of vectors, which is  $k$ , and are linearly independent will be a basis for  $V$ . Note that we can make this conclusion even though we have not verified that  $S$  spans  $V$ . Having the correct number of vectors, together with the linear independence property would be good enough to conclude that  $S$  is a basis for  $V$ .

(#)

Secondly, any subset  $S$  of  $V$  with the correct number of vectors, which is  $k$ , and can be shown to span  $V$  will be a basis for  $V$ . Note that we can make this conclusion even though we have not verified that  $S$  is a linearly independent set. Once again, having the correct number of vectors, together with the knowledge that  $S$  spans  $V$  would be good enough to conclude that  $S$  is a basis for  $V$ . In a way, this theorem tells us that if we have knowledge of the dimension of the vector space, verifying whether a set, with the correct number of vectors, is a basis for the vector space or not, can be done by checking either linear independence or linear span without having to do both.

**Slide 04:** Consider the following example. We would like to show that  $\mathbf{u}_1, \mathbf{u}_2$  and  $\mathbf{u}_3$  forms a basis for  $\mathbb{R}^3$ .

(#)

As mentioned in the previous slide, if we go strictly by definition, we would need to show that the three vectors are linearly independent, and they span  $\mathbb{R}^3$ .

(#)

However, since we know that the dimension of  $\mathbb{R}^3$  is 3, meaning that we have the correct number of vectors in  $\mathbf{u}_1, \mathbf{u}_2$  and  $\mathbf{u}_3$ ,

(#)

it suffices for us to show either the 3 vectors are linearly independent or they span  $\mathbb{R}^3$ .

(#)

We will proceed to show that the three vectors are linearly independent.

**Slide 05:** To do so, we go through the usual procedure of first setting up a vector equation as shown here.

(#)

Writing down the corresponding homogeneous linear system,

(#)

and the augmented matrix,

(#)

we proceed to obtain the reduced row-echelon form of the augmented matrix as shown.

(#)

The conclusion that  $\mathbf{u}_1, \mathbf{u}_2$  and  $\mathbf{u}_3$  are linearly independent is thus obtained.

(#)

Following what we have discussed earlier, since the dimension of  $\mathbb{R}^3$  is 3,

(#)

and 3 linearly independent vectors in  $\mathbb{R}^3$  will always form a basis for  $\mathbb{R}^3$ .

**Slide 06:** The next theorem relates the dimension of subspaces to the dimension of the vector space where the subspace belongs to. Let  $U$  be a subspace of a vector space  $V$ . Then the dimension of  $U$  does not exceed the dimension of  $V$ . Furthermore, if  $U$  is not equal to  $V$ , then the dimension of  $U$  will be strictly smaller than the dimension of  $V$ .

(#)

Essentially, given a vector space  $V$

(#)

and a subspace  $U$  found inside  $V$ ,

(#)

the relationship between the dimension of the two spaces is immediate.

(#)

In addition, if  $U$  is not equal to  $V$ , meaning that there are vectors in  $V$  that are not in  $U$ , then

(#)

the dimension of  $U$  will be strictly smaller than the dimension of  $V$ . In other words, the only subspace of  $V$  that can have the same dimension as  $V$  is  $V$  itself.

**Slide 07:** Consider the following example where  $V$  is a plane in  $\mathbb{R}^3$  containing the origin. We can treat  $V$  as a vector space and the dimension of  $V$  is 2.

(#)

Suppose  $U$  is a subspace of  $V$  and  $U$  is not equal to  $V$ . Then by the preceding theorem, the dimension of  $U$  is strictly less than 2, meaning that it can either be 0 or 1.

(#)

If the dimension of  $U$  is 0, then  $U$  is the zero subspace of  $\mathbb{R}^3$ , which is simply the origin, a single point.

(#)

If the dimension of  $U$  is 1, then  $U$  will be a straight line that passes through the origin in  $\mathbb{R}^3$ .

**Slide 08:** To summarise this unit,

(#)

We saw that knowing the dimension of a vector space  $V$  helps in determining whether a set  $S$  is a basis for  $V$ .

(#)

We also saw that the dimension of all subspaces of a vector space  $V$  cannot exceed the dimension of  $V$  itself.

(#)

In fact, the only subspace of  $V$  that has the same dimension as  $V$  is  $V$  itself.