

# Advanced Matrix Theory and Linear Algebra for Engineers

Prof. R. Vittal Rao

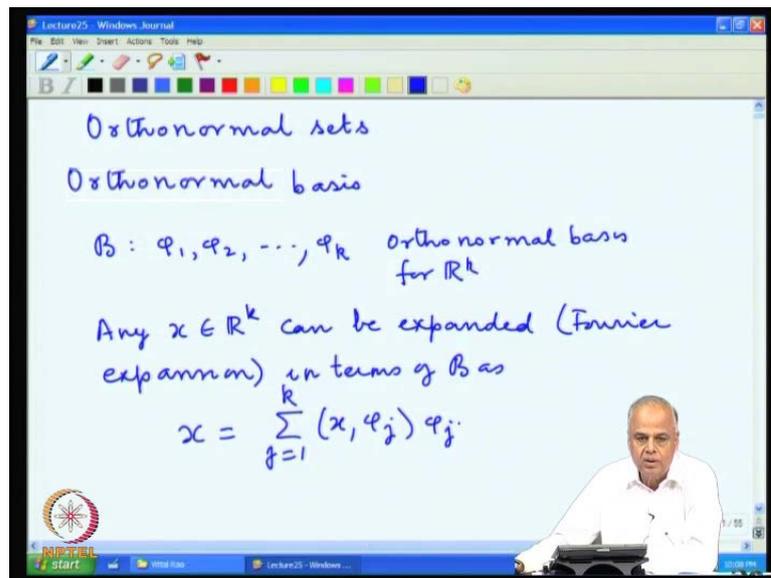
Centre for Electronics Design and Technology

Indian Institute of Science, Bangalore

Lecture No. # 25

Inner Product and Orthogonality- Part 4

(Refer Slide Time: 00:20)

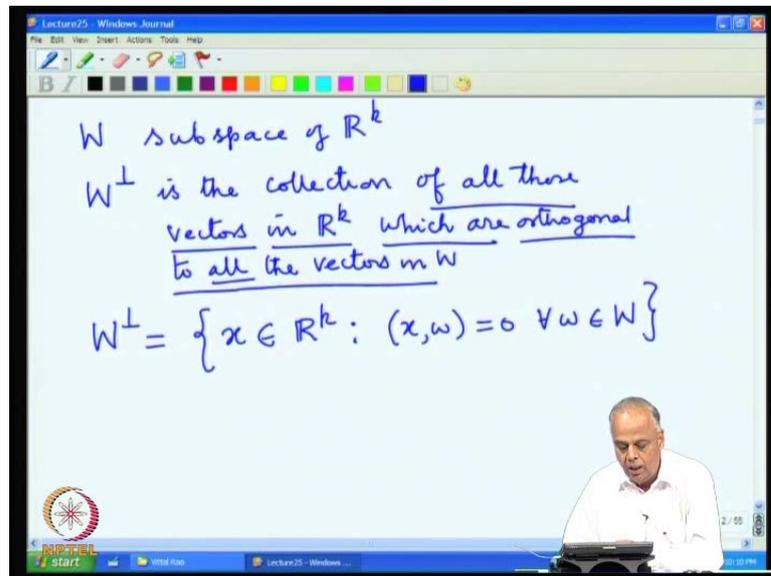


We have been discussing the notion of orthonormality in the space  $\mathbb{R}^k$ . We look at the notion of orthonormal sets that is these are sets, in which any two vectors are orthonormal to each other and every vector's length is 1. Then, we look at the notion of an orthonormal basis, and an important outcome of this notion of orthonormal basis was that, if you had an orthonormal basis  $\phi_1, \phi_2, \dots, \phi_k$  for  $\mathbb{R}^k$ , so this is an orthonormal basis. Suppose, this is an orthonormal basis for  $\mathbb{R}^k$ , so whenever we had an orthonormal basis for  $\mathbb{R}^k$ , we can expand any  $x$  in  $\mathbb{R}^k$ , and that expansion was called the Fourier expansion, expanded in terms of this basis, as  $x = \sum_{j=1}^k (x, \phi_j) \phi_j$ , where  $(x, \phi_j)$  denotes the inner product. So, if you recall the inner product  $(x, \phi_j)$  is also the same as  $\phi_j^T x$ .

So, every vector  $x$  can be expanded in terms of the orthonormal set as the linear combination of the orthonormal basis, and the coefficients are easily obtained as the inner

product of the vector  $x$  with  $\phi_j$ . The coefficients are simply the inner product of  $x$  with  $\phi_j$ . This was one of the important things that we got out of this idea of orthonormal basis.

(Refer Slide Time: 02:30)



Then, we introduce the notion of the orthogonal complement of a set. In particular, we were interested in an orthogonal complement of subspaces, so let say  $W$  is a subspace of  $\mathbb{R}^k$ .  $W$  is the subspace of  $\mathbb{R}^k$  then, we look at all the vectors which are orthogonal to every vector in  $W$ , so that set we call as  $W^\perp$ .  $W^\perp$  is the collection of all those vectors in  $\mathbb{R}^k$ , which are orthogonal to all the vectors in  $W$ . So, orthogonality is given by the fact that, the inner product is 0. So, mathematically, we can write in symbols as  $W^\perp = \{x \in \mathbb{R}^k : (x, w) = 0 \forall w \in W\}$ . So,  $x$  comma  $w$  must be 0 orthogonal to all the vectors, so should be for every  $w$  in  $W$ . So,  $W^\perp$  is the set of all the vectors in  $\mathbb{R}^k$ , which are orthogonal to all the vectors in  $W$ .

(Refer Slide Time: 04:24)

The screenshot shows a digital whiteboard with the following text:

$$W^\perp = \{x \in \mathbb{R}^k : (x, w) = 0 \forall w \in W\}$$

We saw that  $W^\perp$  is a subspace of  $\mathbb{R}^k$   
This subspace  $W^\perp$  is called the  
ORTHOGONAL COMPLEMENT of  $W$

The slide also features a toolbar at the top with various drawing tools and a small inset video of a lecturer in the bottom right corner.

We saw that,  $W^\perp$  is a subspace of  $\mathbb{R}^k$ . This subspace,  $W^\perp$  is called the orthogonal complement of  $W$ . So, every subspace as something called is orthogonal complements.

(Refer Slide Time: 05:08)

The screenshot shows a digital whiteboard with the following text:

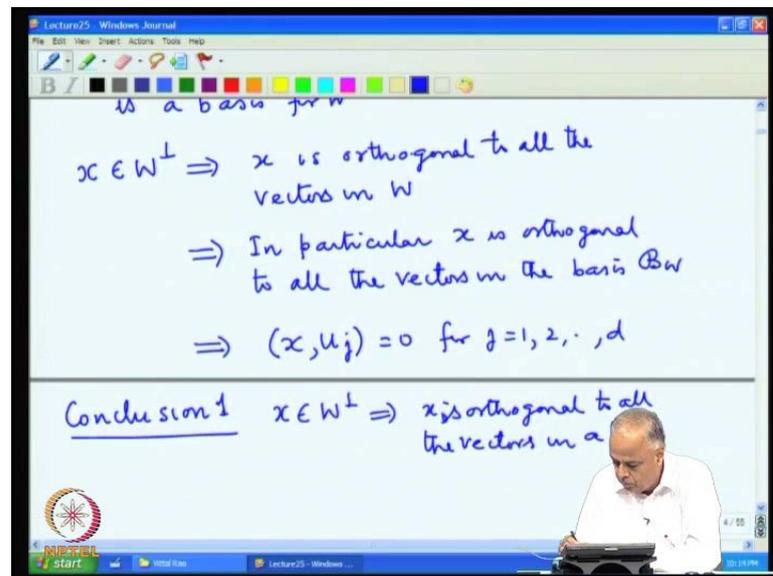
$W$ : subspace of  $\mathbb{R}^k$   $\dim W = d$   
 $W^\perp$ : orthogonal complement of  $W$   
Suppose  $B_W = \{u_1, u_2, \dots, u_d\}$   
is a basis for  $W$   
 $x \in W^\perp \Rightarrow x$  is orthogonal to all the  
vectors in  $W$

The slide also features a toolbar at the top and a small inset video of a lecturer in the bottom right corner.

We will make one simple observation. So, again we start with the subspace  $W$  of  $\mathbb{R}^k$  and we denote by  $W^\perp$ , the orthogonal complement of  $W$ . So, we have a subspace and it is orthogonal complement. Suppose, we take a basis for  $W$ , suppose  $B_W = \{u_1, u_2, \dots, u_d\}$  is a basis for  $W$ , so let say dimension of  $W$  is  $d$ .

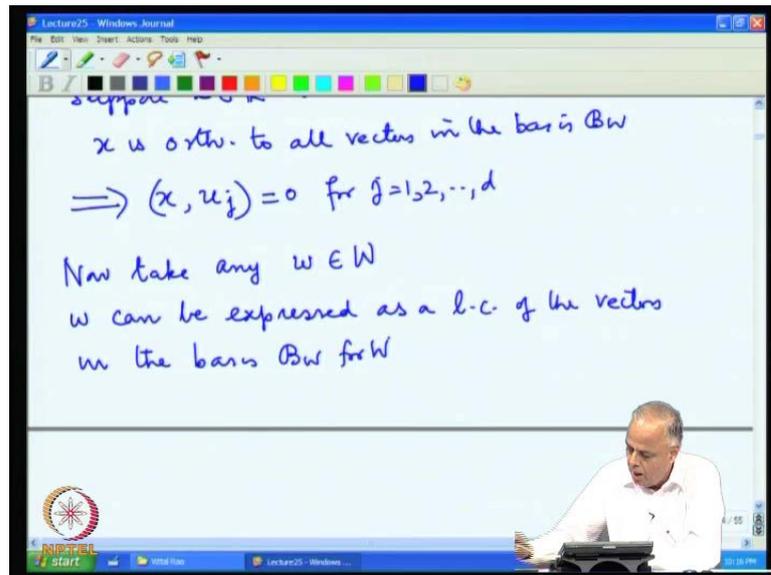
So, we have a subspace of dimension  $d$ , we call the  $w$  and  $W$  perp is the orthogonal complement of  $w$  and we are looking at a basis for  $w$ . It will consist of  $d$  linearly independent vectors let that, basis  $b$   $u_1, u_2, \dots, u_d$ . Now, suppose I have a vector  $x$  in  $W$  perp, this means  $x$  is orthogonal to all the vectors in  $W$ . That is the definition of  $W$  perp,  $W$  perp is the collection of all those vectors, which are orthogonal to all the vectors in  $W$ . So,  $x$  is orthogonal to all the vectors in  $W$ , this basis vectors  $u_1, u_2, \dots, u_d$  are all in  $w$ , and therefore  $x$  must be orthogonal to these vectors.

(Refer Slide Time: 07:05)



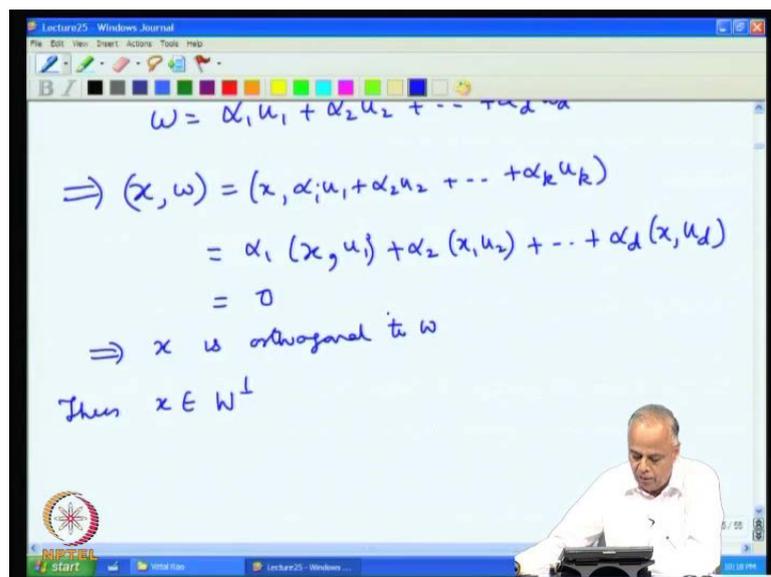
So, implies in particular  $x$  is orthogonal to all the vectors in the basis  $B_w$ . And therefore, what is meant by orthogonality? The inner product  $x$  comma these vectors  $u_1, u_2, \dots, u_k$ , so the inner product  $u_j$  must be equal to 0 for  $j$  equal to 1, 2,  $d$ . So, what these mean says that, if a vector is in  $W$  perp, it must be orthogonal to every one of the basis vectors. So, from it up call it conclusion 1,  $x$  belongs to  $W$  perp implies  $x$  is orthogonal to all the vectors in a basis for  $W$ .

(Refer Slide Time: 08:26)



Conversely, suppose  $x$  belongs to  $R_k$  is such that,  $x$  is orthogonal to all vectors in the basis  $B_w$ . What does that mean? This means,  $x \cdot u_j$  is equal to 0 for  $j$  equal to 1 2 up to  $d$ . Now, take any vector  $w$  in  $W$ . Now, since  $B_w$  is a basis, we can expand  $w$  in terms of these  $B_w$  vectors. So,  $w$  can be expressed as a linear combination of the vectors in the basis  $B_w$  **in the basis  $B_w$**  for  $W$ .

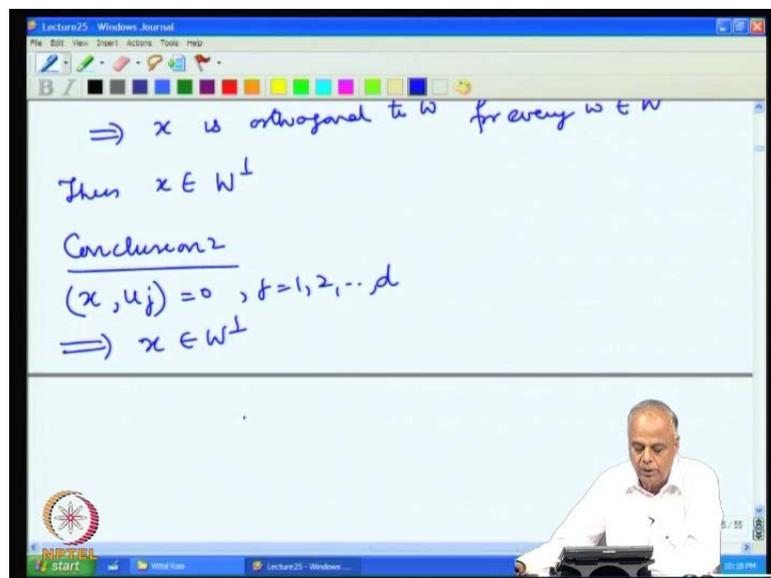
(Refer Slide Time: 09:52)



And therefore, we have  $w$  equal to  $\alpha_1 u_1 + \alpha_2 u_2 + \dots + \alpha_d u_d$ . Consequently,  $(x, w) = 0$ . This implies, if you now take the dot

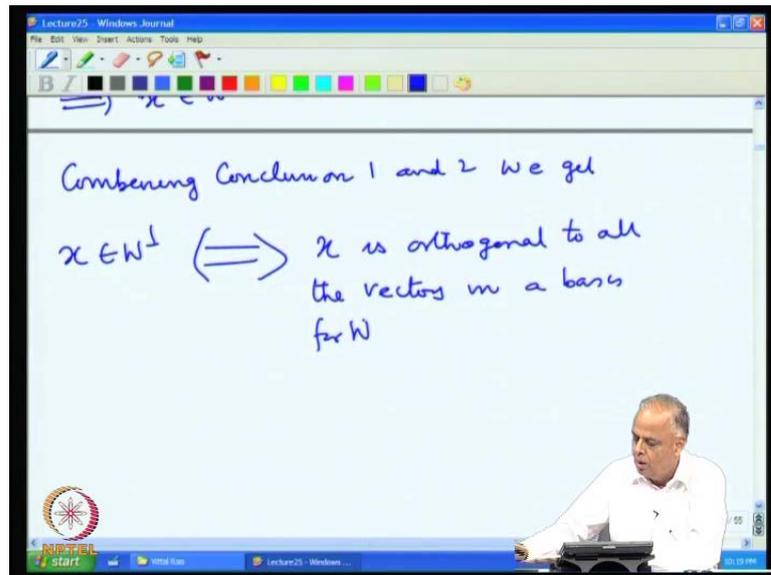
product of the inner product of  $x$  with  $w$ , I get  $x$  comma  $\alpha_1 u_1$  plus  $\alpha_2 u_2$  plus  $\alpha_k u_k$ , but the dot product is distributive and constant come out of the dot product, this is equal to  $\alpha_1$ ,  $x$  comma  $u_1$ ,  $\alpha_1 x$  comma  $u_1$  plus  $\alpha_2 x$  comma  $u_2$  and so on,  $\alpha_d x$  comma  $u_d$ . But now, we are given recall, that we had assume that,  $x$  is orthogonal to all the basis vectors. And therefore,  $x$  comma  $u_j$  is equal to 0 for every  $j$ , so this becomes 0. Which means,  $x$  is orthogonal to  $w$ , so what it says is that the moment the vector, if something is orthogonal to all the basis vectors then, it is automatically orthogonal to all the vectors, thus  $x$  belongs to  $W^\perp$ . This is true for any  $w$  for any  $w$  we can do this, and therefore  $x$  is orthogonal to  $w$  for every  $w$  in  $W$ .

(Refer Slide Time: 11:48)



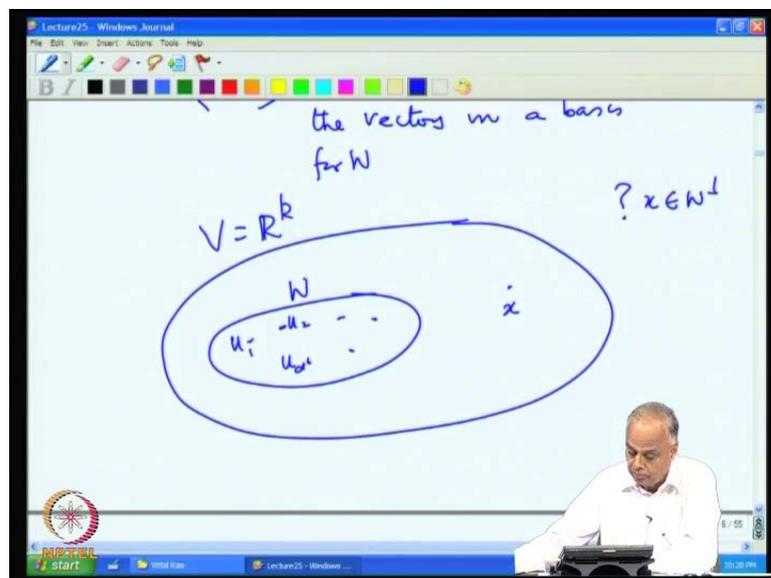
So therefore, conclusion 2 is, if we start with the fact that  $x$  is orthogonal to all the basis vectors then, that implies  $x$  is in  $W^\perp$ . Comparing, this conclusion in conclusion 1, we had that if  $x$  is in  $W^\perp$ , it must be orthogonal to all the basis vectors, and now we have say that if it conclusion orthogonal to all the basis vectors it must be in  $W^\perp$ .

(Refer Slide Time: 12:25)



So, combining conclusion 1 and 2, we get  $x$  belongs to  $W$  perp, if and only if  $x$  is orthogonal to all the vectors in a basis for  $W$ .

(Refer Slide Time: 13:02)



What means, what it means is the following, we have this vector space  $V$  and we have the subspace  $W$ . And we are going to pick a vector  $x$  in  $\mathbb{R}^k$  with the vector space  $V$ , we are dealing with  $\mathbb{R}^k$ . So, we are going to pick a vector in  $\mathbb{R}^k$  and we want to check, whether  $x$  belongs to  $W^\perp$ , so the question we are asking is, does  $x$  belong to  $W^\perp$ .

Now, basically in order to check that  $x$  belongs to  $W$  perp, we have to check whether  $x$  is orthogonal to every one of these vectors in this  $W$ .

However, what we have now shown is, you don't have to go on checking with every vector. You choose this  $(0)$   $d$  samples, namely the basis vectors, and just check whether  $x$  is orthogonal to  $u_1$ ;  $x$  is orthogonal to  $u_2$  and  $x$  is orthogonal to  $u_d$ . And therefore, what it says  $(0)$  go on checking, whether  $x$  is orthogonal to all the vectors is enough you check  $x$  is orthogonal to all the basis vectors then automatically you can conclude, whether it is  $x$  is orthogonal to all the vectors or not, and thus in order to check whether vector  $x$  is in  $W$  perp or not. We have to check only whether  $x$  is orthogonal to all the vectors.

(Refer Slide Time: 14:36)

(1)  $\mathbb{R}^3$   
 $W = \left\{ x = \begin{pmatrix} a \\ a \\ a \end{pmatrix} : a \in \mathbb{R} \right\}$

It is easy to check that  $W$  is a subspace of  $\mathbb{R}^k$ .

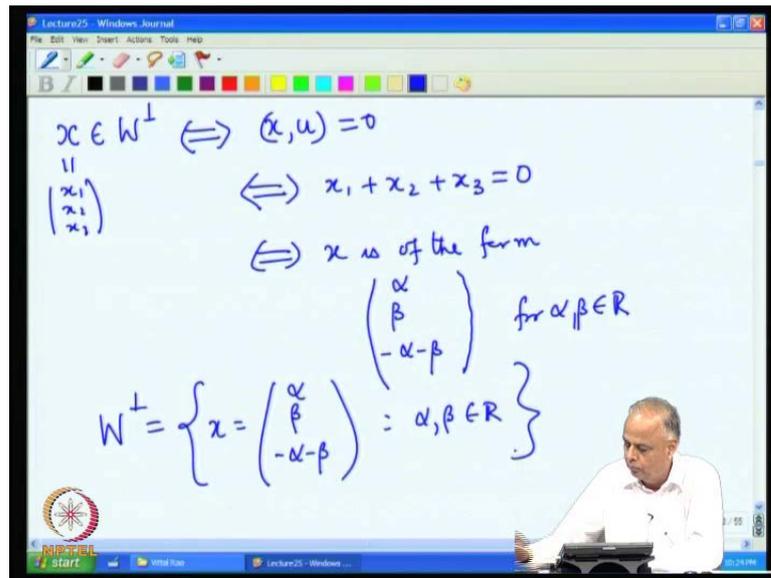
Let us find  $W^\perp$

Clearly  $u = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$  forms a basis for  $W$

Now, let us look at some examples, this idea we will use in the examples to find the  $W$  perp. Let us look at,  $\mathbb{R}^3$  and consider the subspace of all those vectors, which are the form  $a \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$ :  $a$  is in  $\mathbb{R}$ . In other words it consists of all those vectors everyone of this component is equal to the other. So, this is along the line  $1 \ 1 \ 1$ , so now consider this space and that it is easy to check that it is a subspace. We know how to check whether it is a subspace, we have to see it is non empty; we have to check whether it is closed under addition, whether it is closed under scalar multiplication all these are easy things, we have done this before, so it is easy to check that so  $W$  is in it a subspace of  $\mathbb{R}^k$ .

So, let us find  $W^\perp$  the orthogonal complement of  $w$ , the first of all in order to check whether something is  $W^\perp$  or not. We have to check whether everything is orthogonal to a basis vector, so in order to do that we will first find a basis for  $w$ . So, clearly  $(1, 1, 1)$  this vector  $u$  alone forms a basis for  $W$  why is this so, because this vector belongs to  $W$  and every vector in  $W$  is a times  $u$ .

(Refer Slide Time: 16:42)

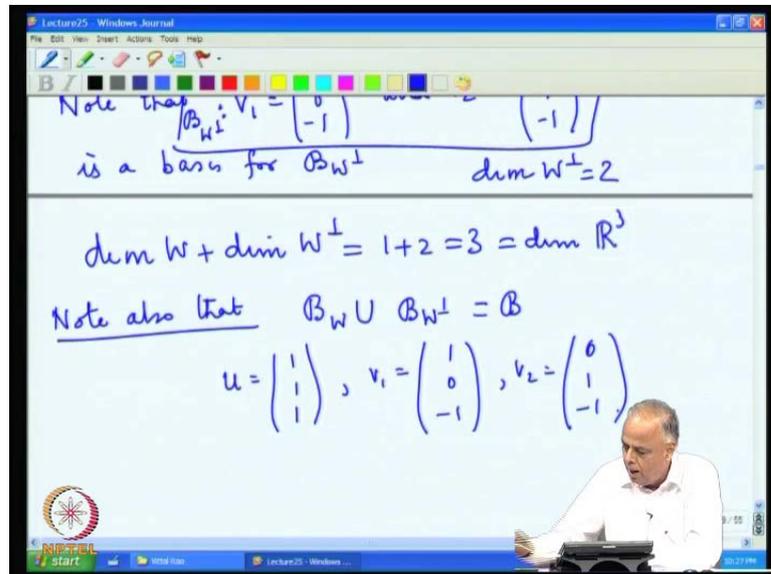


So Therefore, it is expands  $w$  and hence it forms a basis. So, of the  $B_w$  the basis for  $u$   $B_w$  consist of only one vector namely  $u$  equal to  $(1, 1, 1)$ . Now Therefore, a vector  $x$  will belong to  $W^\perp$  if and only if  $x$  is orthogonal to the basis vector and there is only one basis vector, and therefore we want  $x$  to be orthogonal to the basis vector. Suppose  $x$  is  $(x_1, x_2, x_3)$  then, this is  $x \cdot u$  the dot product between  $x$  and  $u$  is  $x_1 \cdot 1 + x_2 \cdot 1 + x_3 \cdot 1$  that must be a 0.

So, the only requirement for a vector to get qualify to be in  $W^\perp$  is that the sum of each components is 0. And therefore, the vector  $x$  is of the form  $(\alpha, \beta, -\alpha - \beta)$  and  $\alpha, \beta$  can be anything, but the moment, you **will** make give any values for  $x_1$  and  $x_2$  to make the total sum is 0, we have to make the third component as minus alpha minus beta for alpha beta in  $\mathbb{R}$ . And therefore, we get  $W^\perp$  to be the collection of all vectors of the form  $(\alpha, \beta, -\alpha - \beta)$  for alpha and beta belong to  $\mathbb{R}$ . So, the original space  $w$  which was all a a a now has this orthogonal component which complement

which consist of vectors is third component. This is the negative of the sum of the first two components.

(Refer Slide Time: 18:40)



Note, that  $v_1$  equal to  $1 \ 0 \ -1$  and  $v_2$  is equal to  $0 \ 1 \ -1$  is a basis for let us call this a  $\beta_{W^\perp}$  is a basis for  $\beta_{W^\perp}$ . Because these 2 vectors are in  $W^\perp$ , and any other any vector  $\alpha \beta - \alpha - \beta$  in  $W^\perp$  can be written as  $\alpha v_1 + \beta v_2$ ; therefore, the span and there obviously linearly independent; and therefore, the they are form a independent set in  $W^\perp$  and span  $W^\perp$ , and therefore they form a basis for  $W^\perp$ . Now, notice the  $w$  had as basis consisting of the single vector, and therefore the dimension of  $W$  was 1 and the basis consist for  $W^\perp$  consist of two vectors; and therefore, dimension of  $W^\perp$  is 2 and in notice that the dimension of  $W$  plus dimension of  $W^\perp$  is equal to 1 plus 2 is 3, which is the dimension of the original vector space in these all the subspaces live, they are subspaces of  $\mathbb{R}^3$  so in this case  $k$  is 3.

So, the dimension of  $w$  plus dimension of  $W^\perp$  tense out to be the dimension of this whole space. Note also, that we have this basis for  $w$  and we had the basis for  $W^\perp$  and if you put them together we get a set  $\mathcal{B}$ , which consist of this vector  $u$  which was  $1 \ 1 \ 1$  and  $v_1$  which consisted  $1 \ 0 \ -1$  and  $v_2$  which was  $0 \ 1 \ -1$ .

(Refer Slide Time: 21:04)

$B = \left\{ \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ -1 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \\ -1 \end{pmatrix} \right\}$

This is l.i. & has 3 vectors &  $\therefore$  a basis for  $\mathbb{R}^3$

The union of a basis for  $W$  and a basis for  $W^\perp$  is a basis for  $\mathbb{R}^3$

And this B is this is linearly independent and has 3 vectors, and we know that in  $\mathbb{R}^3$  any linearly independent set having 3 vectors is a basis, and therefore a basis for  $\mathbb{R}^3$ . So, in this example we observe a lot of things are happening, namely the dimension of  $W$  and dimension of  $W^\perp$  added together to give the dimension of the whole space, and the basis of  $W$  and the basis of  $W^\perp$  added together to give now a basis for  $\mathbb{R}^3$ . So, the union of a basis for  $W$  and a basis for  $W^\perp$  is the basis for  $\mathbb{R}^3$ ; that is in this example, we will see the more general version that this is in general true is later.

(Refer Slide Time: 22:12)

$W = \left\{ x = \begin{pmatrix} a \\ b \\ a+b \\ a-b \end{pmatrix} : a, b \in \mathbb{R} \right\}$

It is easy to check that  $W$  is a subspace of  $\mathbb{R}^4$ .

Let us find  $W^\perp$

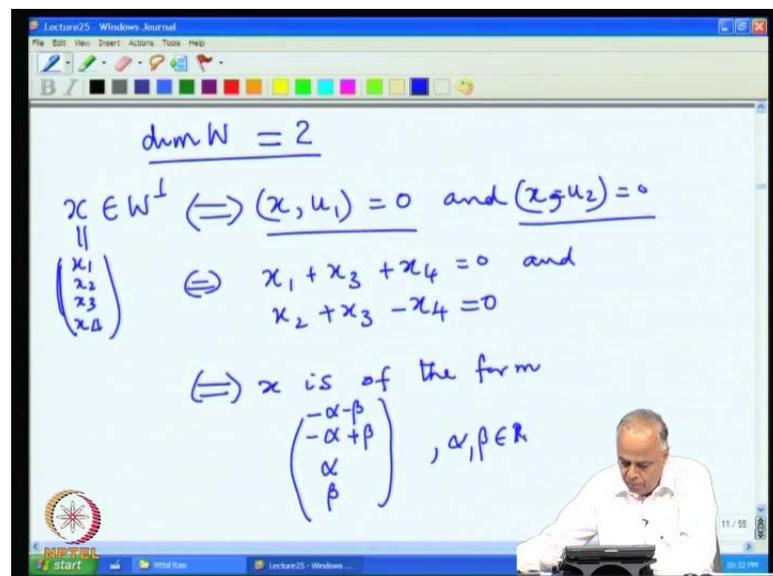
It is easy to see that

$B_{W^\perp} = \left\{ u_1 = \begin{pmatrix} 0 \\ 1 \\ 1 \\ 1 \end{pmatrix}, u_2 = \begin{pmatrix} 0 \\ 1 \\ -1 \\ -1 \end{pmatrix} \right\}$

Let us now, look at another example of this construction of this orthogonal complement. Let us again, now consider  $\mathbb{R}^4$  straight bigger dimensional space than before, and now consider this space subspace  $w$ , which consist of all those vectors of the form  $a \begin{pmatrix} 1 \\ 0 \\ 1 \\ 1 \end{pmatrix} + b \begin{pmatrix} 0 \\ 1 \\ 1 \\ 1 \end{pmatrix}$ ; where  $a, b$  are in  $\mathbb{R}$ . Now once again, it is easy to check that  $w$  is a subspace of  $\mathbb{R}^4$ . So,  $w$  is a subspace of  $\mathbb{R}^4$ , and therefore we will find let us find the orthogonal complement  $W^\perp$ . Now, again as before in order to find the orthogonal complement we first found a basis for  $w$  and then looked at all those vectors which are orthogonal to each one of these basis vectors. So, to this vector we will now go and find a basis for  $w^\perp$ , now from the structure.

It is easy now to see if I take  $a$  equal to 1 and  $b$  equal to 0 I get this vector  $\begin{pmatrix} 1 \\ 0 \\ 1 \\ 1 \end{pmatrix}$ , and if I take  $a$  equal to 0 and  $b$  equal to 1 I get this vector, and this forms a base this vectors are in  $w$  they are linearly independent, and any vector  $w$  is  $a$  times  $u_1$  of the form  $a$  times  $u_1$  plus  $b$  times  $u_2$ , and therefore they are linearly independent vectors in  $w$  which form  $w$ , and therefore they form a basis, so this is a basis for  $w$ . So, what we observe from this about dimension of  $w^\perp$ .

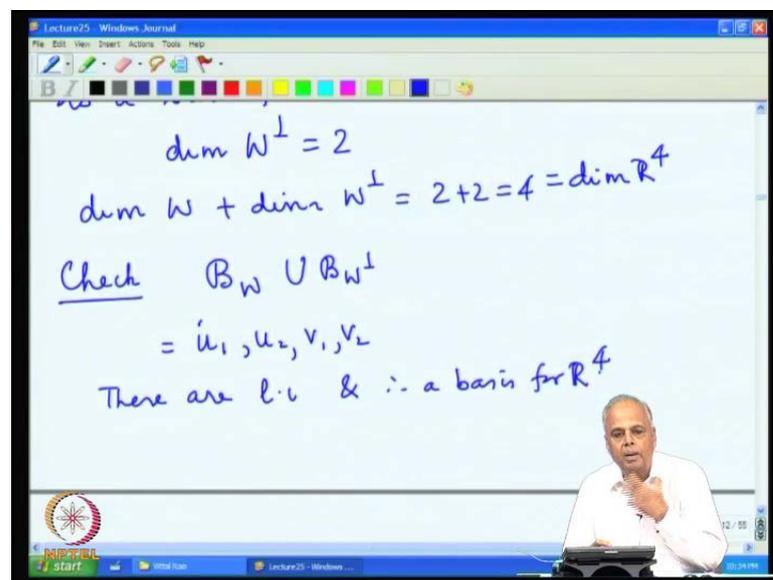
(Refer Slide Time: 24:24)



Since, there are two vectors in this basis the dimensional  $w$  is in it 2. Now, let us find  $W^\perp$   $x$  belongs to  $W^\perp$ , we have seen that a vector belongs to the orthogonal complement, if and only if it is orthogonal to every vector in a basis for  $W$  and therefore,  $x$  must be orthogonal to  $u_1$ , and  $x$  must be orthogonal to  $u_2$ . Now, suppose  $x$  is  $\begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix}$

3 x 4. Now, we are looking in the space  $\mathbb{R}^4$ , so  $x$  will have 4 components. Now, what does the first condition say  $x$  must be orthogonal to  $u_1$ . So, the dot product of  $x$  with  $u_1$  must be 0  $u_1$  is this vector. So,  $x_1$  plus 0 times  $x_2$  plus 1 times  $x_3$  plus 1 times  $x_4$  is 0, so we get  $x_1$  plus  $x_3$  plus  $x_4$  equal to 0. That is this fact then, this fact tells us that  $x_2$  plus  $x_3$  minus  $x_4$  is equal to 0. Now, that says that there are two equations that before component should satisfy, and they are going to be two floating variables and for  $x$  is of the form. We can choose  $x_3$  and  $x_4$  arbitrarily then,  $x_1$  has to be minus  $x_3$  minus  $x_4$  and  $x_2$  has to be  $x_3$  plus  $x_4$ , so  $\alpha$  and  $\beta$  can be anything.

(Refer Slide Time: 26:30)



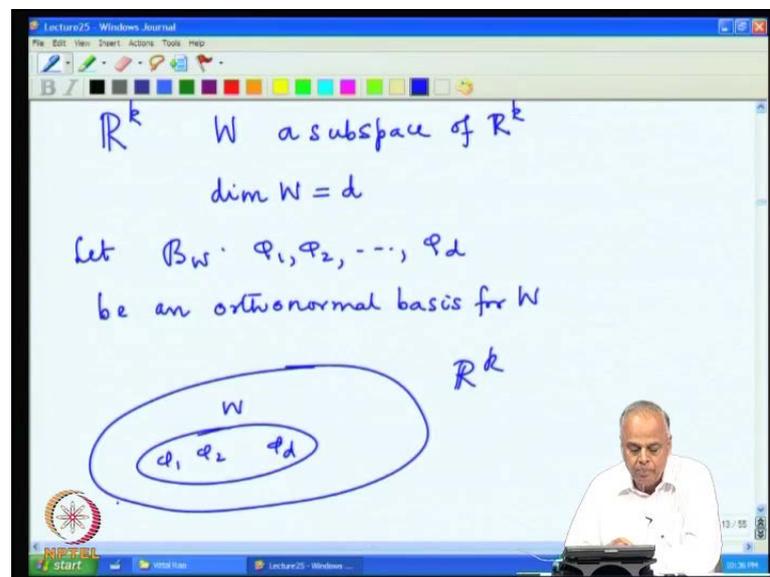
And therefore, we get  $W^\perp$  consist of all the vectors of the form  $-\alpha u_1 - \beta u_2 + \alpha v_1 + \beta v_2$  where  $\alpha$  and  $\beta \in \mathbb{R}$ . Now, clearly again taking  $\alpha$  as 1 and  $\beta$  is 0, we get this vector and  $\alpha$  is 0 and  $\beta$  is 1, we get this vector and that will be a basis for  $W^\perp$ , so is a basis for  $W^\perp$ . And therefore so, in the basis for  $W^\perp$  consist of two vectors, we also observe the  $W^\perp$  as dimension 2 and it turns out  $u_1$ , in this case dimension of  $W$  plus dimension of  $W^\perp$  is 2 plus 2 which is 4 this is the dimension of this space  $\mathbb{R}^4$ .

So, this dimension plus dimension  $W^\perp$  in both these examples added up to the dimension of the whole space, again check which is to check if you put the basis of  $W$  union basis of  $W^\perp$ , we get these vectors  $u_1, u_2, v_1, v_2$  where  $u_1, u_2, v_1, v_2$  are as we observed before, now these are linearly independent check that these are linearly

independent, and these are four vectors which are linearly independent in  $\mathbb{R}^4$ , and therefore a basis for  $\mathbb{R}^4$ .

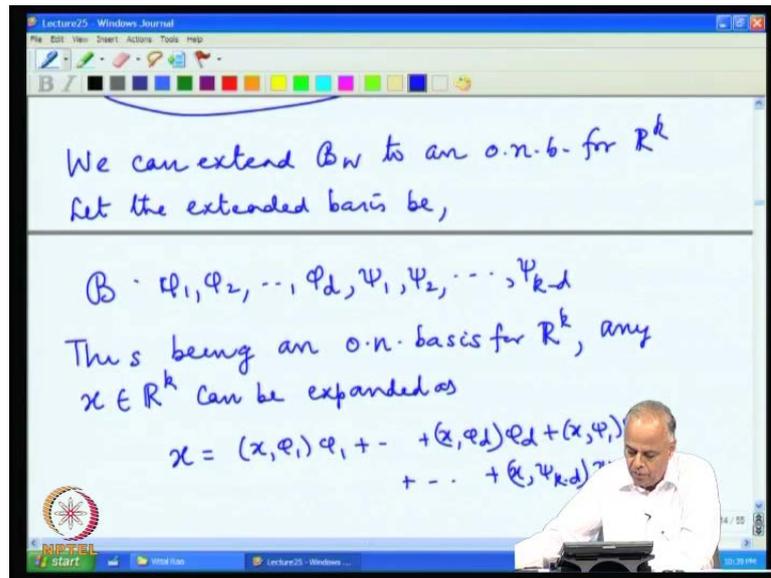
So, basis for  $W$  and a basis for  $W^\perp$  put together gave as a basis for  $\mathbb{R}^4$ , and that it turns essentially gives as the fact that  $W$  the dimension of  $w$  plus dimension of  $W^\perp$  is indeed is equal to the dimension of the whole space. Now, let us perceive this further and analyse the two orthogonal complement all these analysis will be used eventually to study our questions about system of equations diagonalisation etcetera.

(Refer Slide Time: 29:01)



So now, let us get back to  $\mathbb{R}^k$  and let say  $W$  a subspace of  $\mathbb{R}^k$ , let them dimension of  $w$  be equal to  $d$ , and therefore we will have an orthonormal basis, if you take any orthonormal basis for  $w$ , it will have a exactly  $d$  orthonormal vectors. So, let  $B_w$  be  $\phi_1 \phi_2 \phi_d$  and be an orthonormal basis for  $W$ . Now, we have seen that every orthonormal set is also linearly independent, and therefore  $\phi_1 \phi_2 \phi_d$  is in linearly independent set and it is an orthonormal set. It is an orthonormal set sitting in the big space  $\mathbb{R}^k$ , so here is  $\mathbb{R}^k$  and here is  $W$  in this  $\phi_1 \phi_2 \phi_d$  are sitting. Now, since  $\phi_1 \phi_d \phi_2 \phi_d$  is an orthonormal set we have seen that any orthonormal set in  $\mathbb{R}^k$  can be extended to a orthonormal basis in  $\mathbb{R}^k$ .

(Refer Slide Time: 30:35)



So, we can extend  $B_w$  to an orthonormal basis for  $\mathbb{R}^k$  let how many vectors we will have to append the dimension of  $\mathbb{R}^k$  is  $k$ . So, we would have to pick some vectors how many of them  $k$  minus  $d$  call them  $\psi_1, \psi_2, \dots, \psi_{k-d}$ , and these are going to be orthonormal there outside. So, let the extension the extended basis be,  $B$  we will have the entire  $\psi_1$  is that it what it on the extension to this. We will add this append this  $\psi_1, \psi_2, \dots, \psi_{k-d}$ .

So, this is now a basis for  $\mathbb{R}^k$ , and it is an orthonormal basis. So, we have an orthonormal basis for so this being an orthonormal basis for  $\mathbb{R}^k$ , we have seen that any vector can be expanded in a Fourier expansion or as a linear combination in terms of the orthonormal basis vectors, where the coefficients in the linear combination is simply the inner product of the vector with that corresponding basis vector. So, let us some this up, this being an orthonormal basis for  $\mathbb{R}^k$  any  $x$  in  $\mathbb{R}^k$  can be expanded as  $x = (x, \psi_1) \psi_1 + \dots + (x, \psi_d) \psi_d + (x, \psi_{d+1}) \psi_{d+1} + \dots + (x, \psi_{k-d}) \psi_{k-d}$ .

(Refer Slide Time: 33:05)

The screenshot shows a whiteboard with the following content:

$$x = (x_1, \psi_1) \psi_1 + \dots + (x_{k-d}, \psi_{k-d}) \psi_{k-d}$$
$$x = \sum_{j=1}^d (x, \psi_j) \psi_j + \sum_{j=1}^{k-d} (x, \psi_j) \psi_j \quad \dots \quad (I)$$

Some observations:

i) Note since  $\mathcal{B}_0$  is an orthonormal basis for  $\mathbb{R}^k$  every vector in

We will write this, a simply  $j$  equal to 1 to  $d$   $x$   $\psi_j$   $\psi_j$  plus  $j$  equal to 1 to  $k$  minus  $d$   $x$   $\psi_j$   $\psi_j$  is call this 1. Now, we are going to observe in facts from this examination, this expansion now for this we make some certain observations. Now, some observations that we make are the about this vector  $\psi_j$ . Now, since this is an orthonormal basis **this is an orthonormal basis.**

(Refer Slide Time: 34:30)

The screenshot shows a whiteboard with the following content:

In particular

$$(\psi_i, \psi_j) = 0 \text{ for } j=1, 2, \dots, d$$

$\Rightarrow \psi_i$  is orthogonal to all the vectors in the Basis  $\mathcal{B}_0$  for  $W$

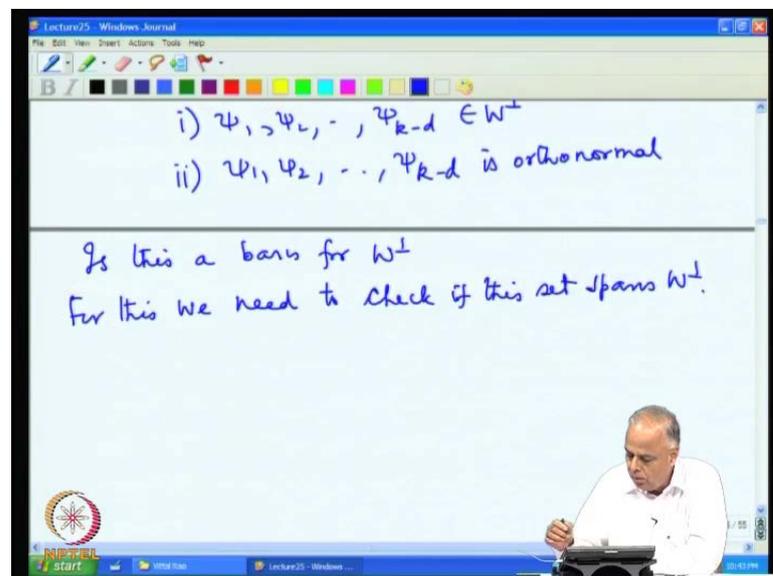
$\Rightarrow \psi_i \in W^\perp$

Similarly  $\psi_2, \psi_3, \dots, \psi_{k-d} \in W^\perp$

So, every vector in this is orthogonal to any other vector, and therefore  $\psi_1$  will be orthogonal to  $\psi_2$  and  $\psi_1$  is orthogonal to  $\psi_d$ . So,

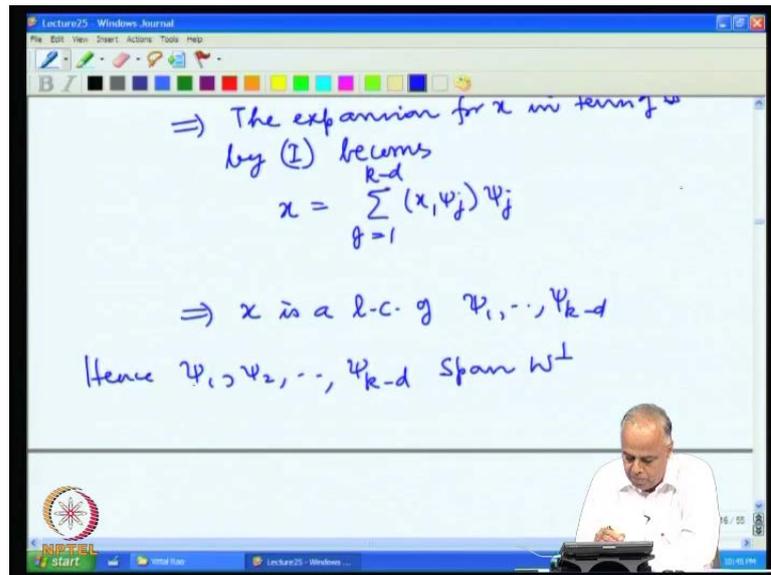
note since,  $B$  is an orthonormal basis for  $\mathbb{R}^k$ , every vector in it is orthogonal to every other vector in it. And therefore, in particular  $\psi_1$  will be orthogonal to  $\phi_1, \phi_2, \dots, \phi_j$  for  $j = 1$  to  $d$ , so  $\psi_1$  is orthogonal to all the vectors in the  $\phi_1, \phi_2, \dots, \phi_j$  group, but these  $\phi_1, \phi_2, \dots, \phi_j$  form a basis for  $W$ , this means  $\psi_1$  is orthogonal to all the vectors in the basis  $B_W$  for  $W$ , and just a 2 minutes back we saw that if a vector is orthogonal to all the vectors in a basis then, it must belong to the orthogonal complement, so that says,  $\psi_1$  belongs to  $W^\perp$ . Similarly,  $\psi_2$  is orthogonal to all the vectors  $\phi_j$ , and therefore  $\psi_2$  belongs to  $W^\perp$  and so on. Similarly,  $\psi_2, \psi_3, \dots, \psi_{k-d}$  all these belong to  $W^\perp$ .

(Refer Slide Time: 36:08)



And therefore, we have the first observation  $\psi_1, \psi_2, \dots, \psi_{k-d}$  are all in  $W^\perp$  and second since the whole set is orthonormal since, the whole set  $B$  is orthonormal even this part must be orthonormal and therefore, we get  $\psi_1, \psi_2, \dots, \psi_{k-d}$  is orthonormal. So, we have an orthonormal set in  $W^\perp$ , we would like to know whether it is an orthonormal basis for  $W^\perp$  is this basis for  $W^\perp$ , for this we have to check whether every vector which is in  $W^\perp$  is a linear combination of this, because it is already orthonormal and hence linearly independent. Therefore, the only requirement to be check for this to form a basis is whether it spans the space. So, for this we need to check if this set spans  $W^\perp$ .

(Refer Slide Time: 37:37)



The screenshot shows a whiteboard with the following handwritten text:

$\Rightarrow$  The expansion for  $x$  in terms of  $\psi$   
by (I) becomes  
$$x = \sum_{j=1}^{k-d} (x, \psi_j) \psi_j$$

$\Rightarrow x$  is a l.c. of  $\psi_1, \dots, \psi_{k-d}$   
Hence  $\psi_1, \psi_2, \dots, \psi_{k-d}$  span  $W^\perp$

The whiteboard is part of a software interface titled "Lecture25 - Windows Journal". In the bottom right corner, a lecturer is visible, looking at a laptop. The bottom of the screen shows a Windows taskbar with the "NPTL Start" logo and a system tray.

So, let check that let  $x$  be in  $W$  perp, if since  $x$  is in  $W$  perp  $x \psi_j$ , since  $x$  is in  $W$  perp  $x \psi_j$  must be equal to 0 for  $j$  equal to 1 to  $d$ . Why is this, because some vector is in  $W$  perp if and only if it is orthogonal to all the vectors in the basis for  $w$ ? Now  $\psi_1 \psi_2 \dots \psi_d$  forms a basis for  $w$ , and therefore  $x$  must be orthogonal to  $\psi_1 \psi_2 \dots \psi_d$ ; and that means, the expansion for  $x$  in terms of  $B$ , remember we had this equation 1 which said that any vector can be expanded of this form. Now, in particular if I have taken the  $x$  for which all the  $x \psi_j$  is 0, the only thing I will get is the remaining part so, by 1 by 1 becomes  $x$  is equal to summation  $j$  equal to 1 to  $k$  minus  $d$   $x \psi_j \psi_j$ , which means  $x$  is a linear combination of this  $\psi_1 \psi_2 \dots \psi_{k-d}$ ,  $x$  is a linear combination of  $\psi_1 \psi_2 \dots \psi_{k-d}$ . Hence,  $\psi_1 \psi_2 \dots \psi_{k-d}$  span  $W$  perp, so we have seen, but they are in  $W$  perp they are orthonormal and now, they span  $W$  perp.

(Refer Slide Time: 39:31)

$\Rightarrow \varphi_1, \varphi_2, \dots, \varphi_{k-d}$  an orthonormal  
basis for  $W^\perp$

$$\dim W^\perp = k - d = \dim R^k - \dim W$$

$\Rightarrow$   $\boxed{\dim W + \dim W^\perp = \dim R^k}$

And hence, they form a basis an orthonormal basis for  $W$  perp so, what it says is that if we now start with a basis for  $w$  and orthonormal basis for  $w$ . Extended to an orthonormal basis for the whole space, the part that is coming in the extension; namely in this case this will automatically be a basis for the orthogonal complement. Now, we get some simply facts from this, what do you get from this as the dimension of  $W$  perp. This is a basis for  $W$  perp it has  $k$  minus  $d$  vectors.

The dimension is precisely the number of vectors in a basis, so dimension of  $W$  perp is  $k$  minus  $d$ , which is dimension of  $R^k$  in  $d$  was the dimension of  $w$ , which implies dimension of  $w$  plus dimension of  $W$  perp is equal to dimension of  $R^k$ . So, this is a very important fact namely that, if you have any subspace  $w$  of  $R^k$  then, the dimension of  $w$  and the dimension of is orthogonal complement add up to the total dimension of the whole space. This is very useful important fact, which we will use shortly.

(Refer Slide Time: 41:26)

The screenshot shows a whiteboard with the following content:

$$x \in \mathbb{R}^k, \quad x = \sum_{j=1}^d (x, \phi_j) \phi_j + \sum_{j=1}^{k-d} (x, \psi_j) \psi_j$$

$$x = x_W + x_{W^\perp}$$

where

$$x_W = \sum_{j=1}^d (x, \phi_j) \phi_j \in W$$

$$x_{W^\perp} = \sum_{j=1}^{k-d} (x, \psi_j) \psi_j \in W^\perp$$

Then let us look at, (( )) the equation 1 which says that  $x$  is in  $\mathbb{R}^k$  then  $x$  can be expanded as  $\sum_{j=1}^d (x, \phi_j) \phi_j + \sum_{j=1}^{k-d} (x, \psi_j) \psi_j$ , where  $\phi_1, \phi_2, \dots, \phi_d$  is an orthonormal basis for  $W$  and  $\psi_1, \psi_2, \dots, \psi_{k-d}$  is an orthonormal basis for  $W^\perp$ . We will call this as  $x_W + x_{W^\perp}$ , where  $x_W$  is the first sum  $x_W = \sum_{j=1}^d (x, \phi_j) \phi_j$ , and  $x_{W^\perp} = \sum_{j=1}^{k-d} (x, \psi_j) \psi_j$ . Now, observe that since  $\phi_1, \phi_2, \dots, \phi_d$  are all in  $W$ , the  $x_W$  is a linear combination of the  $\phi_1, \phi_2, \dots, \phi_d$  and hence belongs to  $W$ . Similarly,  $\psi_1, \psi_2, \dots, \psi_{k-d}$  are in  $W^\perp$ , and therefore this belongs to  $W^\perp$ . So, what this says is the vector  $x$  has been decomposed, added this sum of two vectors the first vector in  $W$  and the second vector is  $W^\perp$ .

(Refer Slide Time: 43:20)

Thus  
Every  $x \in \mathbb{R}^k$  can be written as the  
sum of a vector  $x_W \in W$  and a vector  
 $x_{W^\perp} \in W^\perp$   
Is this decomposition unique?

So, thus the conclusion is every  $x$  in  $\mathbb{R}^k$  can be written or decomposed as the sum of a vector  $x_W$  belong to  $W$ , and a vector  $x_{W^\perp}$  belong into  $W^\perp$ . Now, this show that we can decompose it, and to two parts any vector can be decomposed into two parts; one in  $W$  and other one in  $W^\perp$ . Now or there many ways of doing this decomposition, if so would like to choose the correct one or the useful one or if there is only one way of doing it, then we denote of much choice, so this decomposition unique.

(Refer Slide Time: 44:34)

Suppose  $x = x_W + x_{W^\perp}$   
&  $x = x'_W + x'_{W^\perp}$   
 $x_W, x'_W \in W$   
 $x_{W^\perp}, x'_{W^\perp} \in W^\perp$   
 $0_k = (x_W - x'_W) + (x_{W^\perp} - x'_{W^\perp})$   
 $\Rightarrow (x'_W - x_W) = (x_{W^\perp} - x'_{W^\perp})$

So, suppose we have two decomposition, suppose  $x$  equal to  $x_w$  plus  $x_{w^\perp}$  and  $x$  equal to  $x_{w'} + x_{w'^\perp}$ , where  $x_w, x_{w'}$  are in  $w$  and  $x_{w^\perp}, x_{w'^\perp}$  are in  $w^\perp$ . So, we have two such the composition say of a vector  $x$  into a  $w^\perp$ , and a  $w$  part. Then subtracting we get  $\theta k$  is equal to  $x_w - x_{w'} + x_{w^\perp} - x_{w'^\perp}$  or if you take 1 of these terms to the other side, we get  $x_{w'} - x_w$  is equal to  $x_{w^\perp} - x_{w'^\perp}$ . Now, what that says is this vector, and this vector is the same let us call this common vector as  $z$ .

Then this says, because  $z$  is equal to this, and  $x_{w'}$  is in  $w$   $x_w$  is in  $w$  this vector belongs to  $w$ . Similarly, this is in  $w^\perp$  and this is in  $w^\perp$  the difference of two  $w^\perp$  vector is in  $w^\perp$  and therefore, this is in  $w^\perp$ , because  $w$  and  $w^\perp$  are subspaces difference of two vectors in  $w$  will be  $w$ . Difference of two vectors in  $w^\perp$  will be in  $w^\perp$ ; therefore, this say  $z$  is if a  $w$  vector it is also a  $w^\perp$  vector that says  $z$  belongs to  $w$  into section  $w^\perp$ .

Now, whenever you have  $w$  and  $w^\perp$  the intersection can contain only the  $0$  vector, because if you now look at,  $z$ ,  $z$  treat  $z$  as a vector in  $w^\perp$ , and treat this  $z$  as a vector in  $w$ , we can do that, because that belongs to both  $w$  and  $w^\perp$  now, we have an inner product of vector in  $w$  and  $w^\perp$ , and that must be  $0$ , but the inner product is the length squared, the length is  $0$  means  $z$  is equal to  $\theta k$ . And that says  $x_w$  must be equal to  $x_{w'}$  and  $x_{w^\perp}$  must be equal to  $x_{w'^\perp}$ . So, not only we can decompose a vector  $x$  of the sum of a vector in  $w$  and vector in  $w^\perp$  that this decomposition is unique.

(Refer Slide Time: 47:26)

Theorem: Let  $W$  be any subspace of  $\mathbb{R}^k$ .  
Then every vector  $x \in \mathbb{R}^k$  can be decomposed uniquely as the sum  
$$x = x_W + x_{W^\perp}$$
where  $x_W \in W$  and  $x_{W^\perp} \in W^\perp$ .

The  $x_W$  is called the ORTHOGONAL PROJECTION of  $x$  onto  $W$  and  $x_{W^\perp}$  is:

The screenshot shows a man in a white shirt sitting at a desk in front of a whiteboard. The whiteboard contains the text and equation above. The software interface includes a toolbar with drawing tools and a Windows taskbar at the bottom.

So, what do we get let  $w$  be any subspace of  $\mathbb{R}^k$ . Then every vector  $x$  in  $\mathbb{R}^k$  can be decomposed uniquely as the sum  $x$  is equal to  $x_w$  plus  $x_w$  perp, where  $x_w$  belongs to  $w$  and  $x_w$  perp belongs to  $w$  perp. There is a unique decomposition of every vector with respect to every subspace, this old decomposition depends on the subspace  $w$  with which you start, so moment you give me a subspace, there is a breaking of the vector into two parts one from  $w$  and one from  $w$  perp the  $x_w$  is now, I consider the  $x_w$ , because the whole decomposition is unique.

(Refer Slide Time: 49:31)

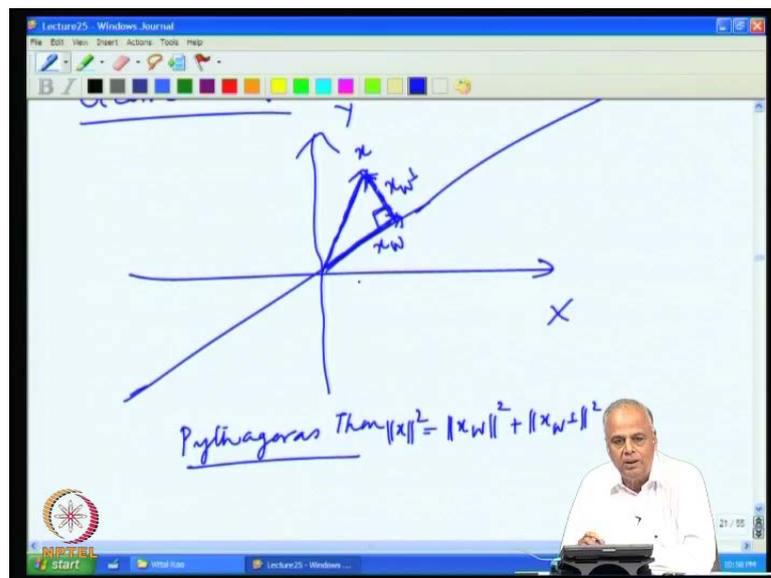
$$x = x_W + x_{W^\perp}$$
where  $x_W \in W$  and  $x_{W^\perp} \in W^\perp$ .

The  $x_W$  is called the ORTHOGONAL PROJECTION of  $x$  onto  $W$  and  $x_{W^\perp}$  is called the ORTHOGONAL PROJECTION of  $x$  onto  $W^\perp$ .

The screenshot shows the same man and whiteboard as the previous slide. The whiteboard text is updated to include the definition for  $x_{W^\perp}$ . The software interface remains the same.

The  $x_w$  is called the orthogonal projection of  $x$  onto  $W$ , and  $x_{W^\perp}$  is called the orthogonal projection of  $x$  onto  $W^\perp$ , we see there if we know one orthogonal projection suppose, we know  $x_w$  we can get  $x_{W^\perp}$  as  $x$  minus  $x_w$ , and alternately, if we know the projection  $x_{W^\perp}$  then we can get  $x_w$  as  $x$  minus  $x_{W^\perp}$ . So, that is we have the notion of the orthogonal projection of this.

(Refer Slide Time: 50:03)



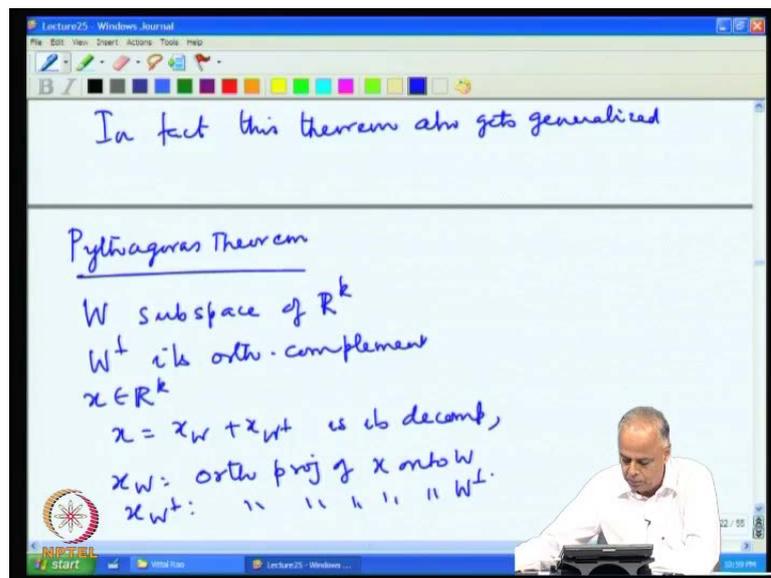
Let us look at what this says geometrically what this does is the generalization of the following fact that, we know into dimensions let us say 2 dimensions. Let us take this  $x$  and  $y$  axis suppose, what is as the notion of a subspace is the line passing through the origin. So, let us say this is the subspace, a line passing through the origin. Take any vector  $x$ , now draw the perpendicular from  $x$  to this. Now, this vector is what now,  $x_w$  is and this vector is  $x_{W^\perp}$  and what it says we know, but the resolution of force is that force  $x$  is equal to the velocity or force, whichever vector you want to use, if the sum of these two vectors, and this is what is now generalize in the set up of  $\mathbb{R}^k$  the line is replace by a subspace  $w$ , and this orthogonal dropping is the orthogonal projection, and their difference is the projection on to the other space.

So, this is the generalization of the simple projection idea, which we have been geometry. Now, in geometry when we have such a projection idea this angle being on

orthogonal ninety degrees, we know by Pythagoras theorem, the length of the hypotenuse squared is equal to sum of the length of the squares of the other 2 sides.

So, the Pythagoras theorem says in  $\mathbb{R}^2$ , but the length of this vector  $x$  squared plus the length of the vector  $x$  perp squared is the length of hypotenuse squared. Now, since we have generalized the projection idea can be generalize this Pythagoras theorem also and in fact this is true.

(Refer Slide Time: 52:15)



In fact this Pythagoras theorem, this theorem also gets generalized, so we have this Pythagoras theorem in the set up of  $\mathbb{R}^k$  is general version of the Pythagoras theorem. So,  $W$  subspace of  $\mathbb{R}^k$ ,  $W^\perp$  its orthogonal complement  $x$  belongs to  $\mathbb{R}^k$ ;  $x$  is equal to  $x_w$  plus  $x_{w^\perp}$  is its decomposition, where  $x_w$  is the orthogonal projection of  $x$  onto  $W$  and  $x_{w^\perp}$  is the orthogonal projection of  $x$  onto  $W^\perp$ , this  $x$  belongs to  $W \times W^\perp$  and  $W^\perp$  belongs to  $W^\perp$ .

(Refer Slide Time: 53:26)

$$\|x\|^2 = \|x_w\|^2 + \|x_{w^\perp}\|^2$$

Proof:

$$\|x\|^2 = (x, x) = (x_w + x_{w^\perp}, x_w + x_{w^\perp})$$
$$= (x_w, x_w) + (x_w, x_{w^\perp}) + (x_{w^\perp}, x_w) + (x_{w^\perp}, x_{w^\perp})$$
$$= \|x_w\|^2$$

Now, let us look at the length of  $x$  squared, so and then length of  $x$  squared is equal to the length of  $x_w$  squared plus the length of  $x_{w^\perp}$  squared. This is what the Pythagoras theorem is lets proof that, so very simple proof the length of any vector is given by the dot product of the vector with itself, but now we are given that  $x$  is  $x_w$  plus  $x_{w^\perp}$ . This will be writing like that, and the dot product is distributive, so we can take out the individual products this will be  $x_w$  dot product  $x_w$  plus  $x_w$  dot product with  $x_{w^\perp}$  plus  $x_{w^\perp}$  dot product with  $x_w$  plus  $x_{w^\perp}$  dot product with  $x_{w^\perp}$ . The first term is the length of  $x_w$  squared, because the dot product of any vector with itself is the length squared and the second term is 0, because  $x_w$  and  $x_{w^\perp}$  are in orthogonal spaces and similarly,  $x_w$  belongs to  $W$  and  $x_{w^\perp}$  belongs to  $W^\perp$  so, that is 0 and that gives me the length of  $x_{w^\perp}$  squared.

(Refer Slide Time: 54:57)

$$= \|x_w\|^2 + 0 + 0 + \|x_{w^\perp}\|^2$$
$$\Rightarrow \|x\|^2 = \|x_w\|^2 + \|x_{w^\perp}\|^2$$

$w$   $w^\perp$

And hence, we get the Pythagoras theorem, length of  $x$  squared is equal to **length of  $x$  square is equal to** length of  $w$  squared plus length of  $x$   $w$  perp squared. So now, if look at this theorem that, any vector can be decomposed as  $x$   $w$  plus  $x$   $w$  perp. Now, we can add and also norm  $x$  squared is equal to norm of  $x$   $w$  of norm of  $x$   $w$  perp squared. So we have, what are all the things that, we have seen you have  $w$  and you have  $w$  perp, and everything can be decomposed in terms of these two. We can decompose a vector, we can decompose the basis into two parts one from  $w$ , and one from  $w$  perp, and we can decompose the vector and it is length into two parts. It is this sort of decompositions that, we will be using in our analysis. We shall look at an example, and see how would use it in your matrix context in the next lecture.