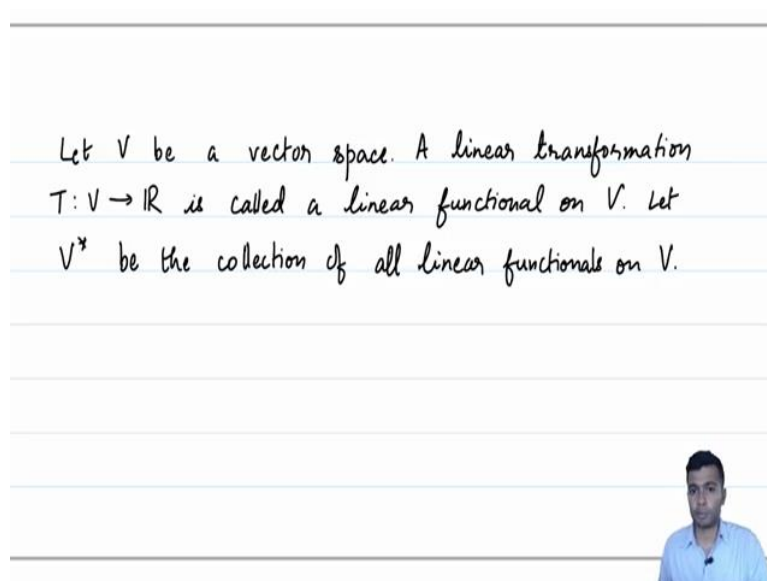


Linear Algebra
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Lecture - 5.3
Dual Spaces

So we have seen the product of vector spaces obtained from vector spaces v_1, v_2 up to v_n . We have also seen what is the quotient of a vector space V by a subspace u . Let us now look at what is meant by the dual of a vector space. Again we are in the process of obtaining newer and newer vector spaces from the given ones. So in order to study the dual, let us consider all linear transformations from a given vector space V into \mathbb{R} .

So remember that \mathbb{R} is a vector space over itself. It is a vector space of dimension 1. So it makes complete sense to talk about linear transformations from V into \mathbb{R} . Such linear transformations are called linear functions, functionals.

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So, let us begin with the definition. Let v be a vector space a linear transformation T from V to \mathbb{R} very specifically into \mathbb{R} , the codomain or the target or the range is \mathbb{R} , is called linear functional on V . Let V^* be the collection of all linear functionals on v . So in the previous week discussed, we how we can add two linear transformations. So if v and w are two vector spaces over \mathbb{R} and suppose s and t are linear transformations from v to w , we discussed what is the meaning of s plus t .

We also discussed the scalar multiplication of a scalar to a linear transformation. So if T is a linear transformation from v to w , we talked about c times T which also turned out to be a

linear transformation from v to w . So, if our w is our \mathbb{R} here, as given here, so all those things can be used, all those definitions can be used here to talk about addition of linear functions and scalar multiplication of linear functions. So let us do that and give a vector addition on v star and a scalar multiplication and on v star.

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V be the collection of all linear functionals on V .

Example: Let $V = C([0, 2\pi])$. Fix $h \in C([0, 2\pi])$

Define $T: V \rightarrow \mathbb{R}$ given by

$$T(x) = \frac{1}{2\pi} \int_0^{2\pi} x(t)h(t)dt.$$

Then T is a linear functional on $C([0, 2\pi])$.

So before we go into that, let us look at an example of a linear functional, which will tell us certain indicate us, why we should be considering these objects. After all, they do not look very natural or maybe they do, but there is a very concrete reason why these objects are studied. The dual is a very well studied object. So one example might help us in justifying that, so consider the vector space v of continuous functions on the closed interval 0 to pi.

So let v be c of 0 to pi and fix some continuous function, let us call it h on C of 0 to 2 pi. Let us define a linear functional on v . Define T from V to \mathbb{R} given by T of function x . This is equal to 1 by 2 pi times the definite integral from 0 to 2 pi of x of t h of t dt . You have already seen such examples of linear maps. And it is easy to check that this is indeed a linear map linear transformation.

So then T is a linear functional on C of 0 to 2 pi. It is quite well studied as mathematical objects, especially when we consider h to be the functions say \sin of nt , \cos of nt , where n is some natural number or where h is \cos of nt where n is some natural number. Such linear functionals will give us what is called as the n th Fourier coefficient of x . So in later analysis courses, later courses in mathematics you will see that these objects are very very well studied and it gives us a lot of information about the function itself.

So, this might be thought of as a partial justification on why we should be studying linear should be studying okay I will say that again. So, this is one example of how the study of linear functional on a vector space tells us a lot about the vector space itself. So let us as I was discussing earlier, let us discuss how we can talk about the addition and the scalar multiplication.

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Vector addition on v

Consider $f, g \in V^*$. Recall

$(f+g)(v) := f(v) + g(v)$ is a linear functional.

scalar multiplication $(cf)(v) := c f(v)$ is v .

$\therefore V^*$ is closed under both these operations.

So let us talk about vector addition first on v star. So consider f and g belonging to v star. Then, f and g are linear transformations from v into \mathbb{R} . And we have already defined what it means to say f plus g . So recall, f plus g of x by definition, is equal to f of x plus g of x and we had checked that f plus g defined in this manner is a linear transformation from V to \mathbb{R} . Therefore, it is a linear functional.


Recall that this is linear functional. Similarly, CF acting on a vector v , so let me not use x here, is defined as C times f of P is also a linear functional. This we had discussed and therefore, v star is closed. So this is scalar multiplication. This is the vector addition and this is the scalar multiplication. So v star is closed under both these operations. And as usual, I am going to leave it as an exercise for you to check that v star is a vector space with these operations.

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also multiplication $(cf)(v) := c f(v)$ is v

$\therefore V^*$ is closed under both these operations.

Then V^* is a vector space with these operations.



The zero element of v star will just turn out to be the zero linear functional, every element being mapped to the zero element of R . And for f , minus f will turn out to be the additive inverse. So you should check that v star is indeed a vector space with the two operations we have just defined. And we did more, suppose we fixed an ordered basis. In the last week, we also discussed how the sum of two linear transformations manifests in the corresponding matrices.


We showed that matrix of s plus t is the matrix of s plus the matrix of t . So let us just observe that here.

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Let β be an ordered basis of V and δ be a basis of R .

Let $f, g \in V^*$

Let us define the map

$$\Phi: V^* \rightarrow M_{1 \times n}(R)$$
$$\Phi(f) = [f]_{\beta}^{\delta}. \text{ We have already checked}$$


$\Phi(f) = [f]_{\beta}^{\delta}$. We have already checked that Φ is a linear map and that it is invertible.

Hence V^* is isomorphic to $M_{1 \times n}(\mathbb{R})$ which has dimension $n = \dim V$.

Hence $\dim(V^*) = \dim(V)$.



dimension $n = \dim V$.

Hence $\dim(V^*) = \dim(V)$.

Hence V^* is isomorphic to V .



So let β be an ordered basis of V and δ be a basis of \mathbb{R} . Remember that \mathbb{R} is a one dimensional vector space over itself. Therefore, δ will consist of one element. It does not matter though. Let f and g be two elements in V^* . Remember, we did a lot more than just associating matrices and checking that the sum of our linear transformations will manifest as the sum of the corresponding matrices.

We showed that this operation is invertible in the sense that the matrix associated to the linear transformation, when we consider the linear transformation corresponding to this matrix, we get back the linear transformation. So let us define the map, let us call it Φ . From V^* into the matrices of size $1 \times n$ of \mathbb{R} which can be identified with \mathbb{R}^n , but we will come to that later, and what was this map? We defined it to be, $\Phi(f)$ is the matrix of f with respect to β δ which is a $1 \times n$ matrix.

And we checked that if we consider L subscript f corresponding to f , we get back f . So we checked, we have already checked that ϕ is a linear map and that it is invertible because this tells us that v^* can be identified with matrices of size $1 \times n$ which is of dimension n . So hence v^* , matrices of size $1 \times n$ is nothing but v is isomorphic to $M_{1 \times n}$ of \mathbb{R} which has dimension of n .

But what is n ? n is the dimension of v , it is equal to the dimension of v . Hence, dimension of v^* is equal to the dimension of v , but what do we know about two vector spaces V and w of equal dimension? We know that if two vector spaces V and w have the same dimension then they are isomorphic and since v and v^* have the same dimension, v^* is isomorphic to v . Now let us consider some very specific linear functions which will be of interest to us.

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dimension $n = \dim V$.

Hence $\dim(V^*) = \dim(V)$.

Hence V^* is isomorphic to V .

Let V be a finite dimensional vector space.

Let $\beta = (v_1, \dots, v_n)$ be an ordered basis of V .

The linear functional f_i is called the i^{th} coordinate function of β .


So in order to do that, let us let v be a finite dimensional vector space, let us fix an ordered basis of V . So let β which is equal to say v_1 to v_n be an ordered basis of v , then for any v , there exists a corresponding column vector corresponding to β . Then for v in capital V consider v β to be equal to the column vector a_1 to a_n . Now let us define f_i from v into \mathbb{R} given by f_i of v is equal to a_i .

It is an easy check to show that f_i is a linear map. Check. It is an exercise that f_i belongs to v^* . So given any ordered basis, for each basis vector, we have now constructed a corresponding linear functional. This linear functional has a name, it is called the i th coordinate function. Then f_i , the functional f_i , the linear functional f_i is called the i th coordinate function as should be the case, is called the i th coordinate function.

But it very heavily depends on the ordered basis β . So this is of β or with respect to β . So these linear functionals f_i that we just defined the very explicitly used our knowledge of β . It so happens that these f_i 's which are n vectors in an n dimensional vector space happens to be a basis of our v^* as well.

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Theorem: Let V be a finite dim vector space. Let $\beta = (v_1, \dots, v_n)$ be an ordered basis of V . Let $\beta^* = \{f_1, \dots, f_n\}$. Then β^* is a basis of V^* . Further, for $f \in V^*$

$$f = \sum_{i=1}^n f(v_i) f_i.$$


So let us give a theorem here. So let me just rework the entire setup so that it is completely clear. Let v be a finite dimensional vector space and suppose β is an ordered basis, v_1 to v_n be an ordered basis of v . Then for each v_i we have, for β we have a i th coordinate vector. So let the dual \mathbb{R} be given by f_1 up to f_n which we just defined or β f_i captures the i th coordinate of any vector v when written in the column representation.


Let β be f_1, f_2 up to f_n , then statement says that β is a basis of V , of V^* . Further, f is equal to f in V^* , f is equal to summation i is equal to 1 to n , f of v_i times f_i . Let us prove this theorem.

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Theorem: Let V be a finite dim vector space. Let $\beta = (v_1, \dots, v_n)$ be an ordered basis of V . Let $\beta^* = \{f_1, \dots, f_n\}$. Then β^* is a basis of V^* . Further, for $f \in V^*$

$$f = \sum_{i=1}^n f(v_i) f_i.$$

Proof: Enough to show that β^* is a spanning set.
Let $f \in V^*$.



Let us give a proof. So we have n vectors f_1, f_2 up to f_n in an n dimensional vector space V^* . So, if we show that f_1, f_2 up to f_n is a spanning set, then by a corollary, it will be replacement theorem. This is necessarily linearly independent and hence a basis. So first observation is that enough to show that β^* is a spanning set. The theorem already describes how to show that it is a spanning set.

For any f , we know that we have to prove that f is equal to summation f of v_i times f_i . So, we are being prescribed exactly the method to say how f is in the span of f_1, f_2 up to f_n . So, let us take an element in V^* . Let us show that it is exactly this, but we do not know a priori whether the right hand side is equal to f , is not it? So let us do one thing.


Let us say that this right hand side is some element, g in V^* , of course, it is an element in V^* . Let us say that it is some g in V^* and we will show that g and f are equal on every element of V . So that is the strategy.

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Proof: Enough to show that β^* is a spanning set.
Let $f \in V^*$.
Suppose $g = \sum f(v_i) f_i$. We want to show that $g=f$.


i.e. $g(v) = f(v) \forall v \in V$.

Recall that f, g are linear.
 \therefore it is enough to show that $g(v_i) = f(v_i)$



Recall that f, g are linear.
 \therefore it is enough to show that $g(v_i) = f(v_i)$ for every basis vector in β .

By defn $g(v_i) = \left(\sum_{j=1}^n f(v_j) f_j \right) (v_i)$

$$= \sum_{j=1}^n f(v_j) f_j(v_i).$$


So suppose g is equal to summation f of v_i times f_i . F is a function, a functional, linear functional on v and therefore summation f of v_i times f is a very valid honest element in v star. Let us call it g . We want to show that g is equal to f . What is the meaning of g is equal to f ? i.e. g of v is equal to f of v for all v in capital V . This is what we would like to show.

But remember, recall that g and f are both linear maps. Recall that f comma g are linear. Therefore, it is enough to show that g of V_i is equal to f of V_i for every basis vector in β because if we do that the linearity of f and g will tell us that g of v is equal to f of v for all v . But then what is g of v_i ? By definition g of v_i , this is equal to summation j is equal to 1 to n , f

of v_j times f_j , the whole at v_i . This is exactly what g of v_i is, but by definition again this is equal to summation j is equal to 1 to n , f_j of v_j of v_i .

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The basis β^* is called the dual basis corresponding to β .

Example: Let $V = \mathbb{R}^2$ and $\beta = \{(1, 1), (1, -1)\}$

Let $\{f_1, f_2\}$ be the dual basis.

We know that $f_1(1, 1) = 1$ & $f_1(1, -1) = 0$

i.e. 1. $f_1(1, 0) + 1 f_1(0, 1) = 1$
 1 $f_1(1, 0) - 1 f_1(0, 1) = 0$

We know that $f_1(1, 1) = 1$ & $f_1(1, -1) = 0$

i.e. 1. $f_1(1, 0) + 1 f_1(0, 1) = 1$
 1 $f_1(1, 0) - 1 f_1(0, 1) = 0$

$\Rightarrow f_1(1, 0) = 1/2, f_1(0, 1) = 1/2.$

$\therefore f_1(x, y) = (x+y)/2.$

Similarly $f_2(x, y) = (x-y)/2.$

Let us now look at an example. So an example, let us consider the simple case of v being \mathbb{R}^2 . So let v be equal to \mathbb{R}^2 and suppose β is the ordered basis given by 1 comma 1 , and 1 comma minus 1 . v_1 v_2 . So let f_1 comma f_2 be the dual basis. We would like to explicitly find out what f_1 and f_2 are.

So what we know is that f_1 of 1 comma 1 is 1 and we know that f_1 of 1 comma minus 1 is 0 . Using this, let us try to see whether we can find out f_1 entirely. So we know that f_1 of 1 comma 1 is equal to 1 , but f_1 is

a linear function, so in particular i.e, 1 times f_1 of 1 comma 0, plus 1 times f_1 of 1 comma, 1 0 comma 1 is equal to 1.

We also know that f_1 of 1 comma minus 1 is 0, right? And that is telling us that 1 times f_1 of 1 comma 0 minus 1 times f_1 of 0 comma 1 is equal to 0. These two equations together tell us that f_1 of 1 comma 0 is equal to half, f_1 of 0 comma 1 is also equal to half. Similarly, you can check. Before we go into what f_2 is, we now know what to do with f_1 of x comma y .


f_1 of x comma y is x times f_1 of 1 comma 0 plus y times f_1 of 0 comma 1, which is equal to x by 2 plus y by 2 which is equal to x plus y by 2. Similarly, it is an easy check to see that f_2 of x comma y is x minus y by 2. You have seen something similar already and you see that many of these things are related. That is why when we consider 1, 1 and 1 minus 1 we are getting f_1 and f_2 like this.

So we have explicitly computed what is the dual basis corresponding to 1, 1 and 1 minus 1. Maybe in fact, it is a good exercise for you to check what the dual basis corresponding to the standard basis is. It will be a good exercise to get hold of concrete basis of \mathbb{R}^2 and see what the dual basis is. Now, let us consider a linear map from V into w . Let us call it T . Let T be a linear map from V into w .

We now know that corresponding to v , we have a dual basis v star; corresponding to w , there is a linear dual vector space w star. Can we somehow using t obtain a natural linear transformation between these two? That is going to be the content of the next theorem.

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Theorem: Let $T : V \rightarrow W$ be a linear transformation. Then, the map $T^t : W^* \rightarrow V^*$ given by $T^t h := hT$ is a linear transformation. Suppose V & W are finite dim. & let α, β be ordered bases of V & W resp. Suppose α^*, β^* are the dual bases of α & β resp. Then

$$[T^t]_{\beta^*}^{\alpha^*} = \left([T]_{\alpha}^{\beta} \right)^t$$


So let us have a look at what the next theorem is. So let T from V to W be a linear transformation. So let us assume that okay, let us not assume as of now that it is finite dimensional. We do not need finite dimensionality yet. Then the map T^* which is from W^* to V^* . Observe that T^* is a map from W^* to V^* , not from V^* to W^* .

And what is this map? The map given by T^* of let us say h , defined to be h composed with T is a linear transformation. Note that we have not assumed any finite dimensionality of V or W , but now let us consider the case when V and W are finite dimensional. Suppose, V and W are finite dimensional and suppose α and β and let α, β be ordered basis of V and W respectively.

Now, given any ordered basis of a vector space, there is a corresponding dual basis as well in the dual space. So α^* and β^* is the dual basis of α and β . Suppose, α^* and β^* or the dual basis of α and β respectively, then the matrix of T^* , so T^* is from W^* to V^* . The dual basis in W^* is β^* and the dual basis in V^* is α^* .

So this is a matrix corresponding to these basis. Now this, how is this related to our matrix of T ? The relationship is quite simple, it turns out to be T^* , transpose of the matrix and that is precisely why the map is being denoted by T^* .

So let us give a proof of the statement. So let me show you the statement once more. The first part said that if you define T^* as $T^*(h)$ is equal to h composed with T , then it is a linear transformation from W^* to V^* . In fact, it is not even clear yet that T^* maps W^* to V^* , but that will be clear in a moment, the moment we start writing down the proof.

And then further it says that if you fix two ordered basis, α, β and let α^*, β^* be the corresponding dual basis, then the matrix of T^* with respect to β^* and α^* will be the transpose of the matrix of T with respect to α and β .

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$[T^t]_{\beta^*}^{\alpha^*} = ([T]_{\alpha}^{\beta})^t$

Proof: Let $h \in W^*$. Then $T^t h = hT$

Since h & T are linear, hT is also a linear transformation
 χ $hT: V \rightarrow \mathbb{R} \Rightarrow T^t h \in V^*$.

Let $h, g \in W^*$ $T^t(g+h) = (g+h)T = gT + hT$
 $= T^t g + T^t h.$

Likewise $T^t(ch) = cT^t h.$

Let $h, g \in W^*$ $T^t(g+h) = (g+h)T = gT + hT$
 $= T^t g + T^t h.$

Likewise $T^t(ch) = cT^t h.$

Hence T^t is a linear transform

So let us first prove the first part of the theorem. So let h be an element of w star, then T transpose h is equal to h times h transpose T . It is actually h times T , it is a linear transformation, both are linear and since, h and T are linear, ht also linear. Composition of linear transformations is a linear transformation. Further ht maps w into \mathbb{R} .

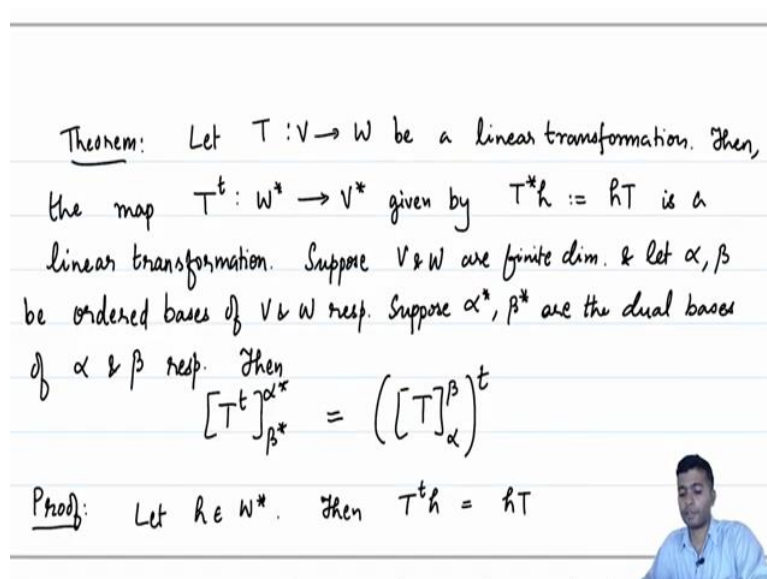
From linear transformation and ht is a map from w into \mathbb{R} . This implies that T transpose of h belongs to w star. w into sorry, it is from V into \mathbb{R} , therefore this belongs to, so note that t is a map from v to w and h is a map from w to \mathbb{R} . So ht is a map from v into \mathbb{R} . It goes through w though. T transpose h is an element of v star hence.

Let us now check that it is a linear transformation. We checked that T transpose h is a linear transformation, but why is T transpose a linear transmission from w star to v star? So that is

actually a simple check. So let h comma g be in w star. Let us see what T transpose of g plus h is. In fact see g plus h is. I would have done it in one go. I have just proved the additive part. You can show the scalar multiplication also is linear.

So this by definition is g plus hT but what is the meaning of g plus hT ? It says gT plus hT , but gT is precisely equal to T transpose g and hT is precisely equal to T transpose h and that is precisely what we wanted to show. Similarly, T transpose of C times h is equal to c times T transpose h . So that tells us that T transpose is a linear transformation. So we have proved the first part of the theorem.

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Theorem: Let $T : V \rightarrow W$ be a linear transformation. Then, the map $T^t : W^* \rightarrow V^*$ given by $T^t h := hT$ is a linear transformation. Suppose V & W are finite dim. & let α, β be ordered bases of V & W resp. Suppose α^*, β^* are the dual bases of α & β resp. Then

$$[T^t]_{\beta^*}^{\alpha^*} = \left([T]_{\alpha}^{\beta} \right)^t$$

Proof: Let $h \in W^*$. Then $T^t h = hT$

Let me show you the statement of the theorem wants to recall what we were proving. We were proving that given any linear transformation T from v to w , T transpose from w star to v star is a linear transformation. So what we just showed was that T transpose indeed maps w star into v star. Not just that, we further showed that T star is indeed a linear map. Therefore, the first part, the part till here has been proved. Now, let us move on to the case when both are finite dimensional.

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$$\chi \quad hT: V \rightarrow \mathbb{R} \Rightarrow T^t h \in V^*$$


$$\text{Let } h, g \in W^* \quad T^t(g+h) = (g+h)T = gT + hT$$

$$= T^t g + T^t h.$$

$$\text{Likewise } T^t(ch) = cT^t h.$$

Hence T^t is a linear transformation

Suppose V, W are finite dim. Let $\alpha = (v_1, \dots, v_n)$
and $\beta = (w_1, \dots, w_m)$ be ordered basis.




$$\text{Suppose } [T]_{\alpha}^{\beta} = A = \begin{pmatrix} a_{11} & \dots & a_{1n} \\ \vdots & & \vdots \\ a_{m1} & \dots & a_{mn} \end{pmatrix}$$

$$B = [T^t]_{\beta^*}^{\alpha^*} \quad \alpha^* = (f_1, \dots, f_n) \quad \beta^* = (g_1, \dots, g_m)$$

The i th column of B is the column repⁿ of $[T^t g_i]^{\alpha^*}$

$$T^t g_i = \sum_{j=1}^n T^t g_i(v_j) f_j$$

$$\text{Hence } [T^t g_i]^{\alpha^*} = \begin{pmatrix} T^t g_i(v_1) \\ \vdots \\ T^t g_i(v_n) \end{pmatrix} = \begin{pmatrix} g_i T v_1 \\ \vdots \\ g_i T v_n \end{pmatrix} = \begin{pmatrix} a_{i1} \\ \vdots \\ a_{in} \end{pmatrix}$$


So suppose, v and w are finite dimensional and let α comma β be two basis, two order basis. Let α equal to say v_1 to v_n and β be equal to w_1 two w_m be ordered basis and let us now consider the corresponding dual basis. So let α^* be equal to f_1 to it will be n of them because v is n dimensional by assumption and β^* be equal to g_1 to g_m .

And suppose the matrix of T corresponding to α, β be equal to a matrix A which is given by a_{11} to a_{1n} , a_{m1} to a_{mn} . So we would like to see what is, so we want to calculate what is T^t beta star alpha star. This is what we would like to calculate, but what is our beta star and alpha beta star? Beta star is g_1 to g_m and alpha star is f_1 to f_m . So this is going to be an n cross m matrix for sure.

T transpose with respect to this has to be an n cross m matrix and the i th column of, so let us call it B , of B is the column presentation of T transpose g_i with respect to α star. Remember, so let me just write it down here α star is f_1 to f_n and β star is equal to g_1 to g_m . Now, what is T transpose of g_i ?

That is precisely what we would like to see. T transpose of g_i we know from the previous theorem. This is equal to summation j is going from 1 to n , T transpose g_i of v_j and then f_j the n vectors and this is, so hence what is T transpose g_i with respect to α star? That is nothing but T transpose g_i of v_1 , T transpose g_i of v_n . But by definition what is T transpose g_i ? It is g_i^T .

This is equal to $g_i^T v_1$ up to $g_i^T v_n$ but again what is g_i ? g_i 's are the dual basis corresponding to β . So hence it is the i th coordinate with respect to w_1, w_2 up to w_n . So this is going to give you the i th coordinate. So this is going to be your, a_{i1} up to a_{in} . But what is a_{i1}, a_{i2} up to a_{in} ?

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Handwritten notes on lined paper:

$$\left(T^t g_i(v_n) \right) \quad \left(g_i^T v_n \right) \quad \left(a_{in} \right)$$

$\Rightarrow [T^t g_i]^{\alpha^*}$ is the i^{th} row of A .

Hence $B = A^t$. □

That is the is the i th row of A , which implies T transpose g_i α star is the i th row of A . Hence, the i th column of B is the i th row of A and the dimensions match. Hence, B is equal to A transpose. That is precisely what we were trying to prove. So, we have taken a vector space v , looked at its dual v star, and for every ordered basis β , we gave a corresponding dual basis and there is a natural correspondence once we fix a given basis.

However, the correspondence between v and the dual of its dual is far more natural. So let us consider what I just said.

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$(T^t g_i(v_n)) \quad (g_i T v_n) \quad (a_i n)$

$\Rightarrow [T^t g_i]^{\alpha^t}$ is the i^{th} row of A .

Hence $B = A^t$. \square

Consider $V^{**} = (V^*)^*$.

We know that $\dim V^{**} = \dim V^* = \dim V$.

So consider, v^{**} . Consider v^{**} in fact, which is the dual space of the dual space of v , which is the v^* is the dual space and you look at the dual space of v . So in other words it is all the linear functionals on v^* . So, what we will do is, so we know that dimension of V^* is n and the dual of dimension, the dimension of the dual of v^* will also be n . So we know that dimension of v^{**} is equal to the dimension of v^* by what we have done.


And this is nothing but the dimension of B . Hence, we know that v^{**} and v are isomorphic to each other. The fact is that this isomorphism is far more natural than we would expect. In fact, we can talk about an isomorphism here without going down to a basis. Without referring to any basis we can talk about an isomorphism here.

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Consider $v \in V$.

We know that $\dim V^{**} = \dim V^* = \dim V$.

Let $v \in V$. Define

$$v^{**} : V^* \rightarrow \mathbb{R} \text{ given by } v^{**}(f) := f(v).$$

$$v^{**}(f+g) = (f+g)(v) = f(v) + g(v) = v^{**}(f) + v^{**}(g).$$

Defn: Define $\bar{\Psi} : V \rightarrow V^{**}$ defined by

$$\bar{\Psi}(v) = v^{**}. \text{ Then } \bar{\Psi} \text{ is an isomorphism.}$$


So let us start the groundwork for doing that. So let v be an element of capital V . Define, v star star, small v star star are mapped from v star into \mathbb{R} given by v star star of f is equal to by definition, or is defined to be f of v . It is an easy check to see that v star star is a linear functional on v star. Let us just have a quick look at it, V star star of say f plus g is just going to be equal to f plus g of V , which by definition is equal to f of v plus g of v , which is equal to v star star of f plus v star star.

And similarly, v star star of cf is equal to c times v star. c times v star star of f . So let us now define a map. So define let us call it ψ , from v into v star star defined by ψ of v to be equal to v star star. Then so let me just call it a theorem. As you can see, to talk about ψ , we do

not even have to worry about whether, we do not have to refer to any ordered basis of V , directly from V , we are able to get hold of Ψ , for every element v , we can talk about a corresponding element in the double dual, or V^{**} . So the theorem says that Ψ is an isomorphism. So to show that Ψ is an isomorphism first we need to check that Ψ is a linear map, so how to do that?

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$\underline{\text{prf:}}$ Define $\Psi : V \rightarrow V^{**}$ defined by
 $\Psi(v) = v^{**}$. Then Ψ is an isomorphism.
 Let $v_1, v_2 \in V$. For $h \in V^*$
 $\Psi(v_1 + v_2)(h) = h(v_1 + v_2) = h(v_1) + h(v_2)$
 $= \Psi(v_1)(h) + \Psi(v_2)(h)$
 $\Rightarrow \Psi(v_1 + v_2) = \Psi(v_1) + \Psi(v_2)$.

Let us consider v_1 and v_2 , two vectors in capital V . Let us look at Ψ of v_1 plus v_2 . What is this? This is basically, so this will be vector in V^{**} . So in other words, this will act on some element, h . So for h in V^* , we would like to see what is Ψ of v_1 plus v_2 of h . So this is by definition equal to v_1 . This is going to v_1 plus v_2 star star of h , which is just h of v_1 plus v_2 , but h is a linear map, h is an element in V^* and therefore, h of v_1 plus v_2 is h of v_1 plus v_2 .

But that is v_1 star star of h_1 , sorry, v_1 star star of h plus v_2 star star of h which is just Ψ , v_1 of h plus Ψ , v_2 of h , which hence implies that Ψ of v_1 plus v_2 , yes, this is equal to Ψ , v_1 plus Ψ , v_2 of h . So, this is equal to Ψ , v_1 plus Ψ , v_2 of h . So Ψ hence is a linear map. You can check that the scalar properties also satisfied in a very similar manner, but we have only obtained linear map from V to V^{**} .

But notice that V and V^{**} , both have dimension equal to the dimension of V , let us say n . So if dimension of V is n , V and V^{**} both have dimension n , and we already know that if you have an injective map from vector space V to another vector space W of the same

dimension, then by the dimension theorem, we can conclude that this is a surjective map. So let us see, let us try to conclude that ψ is an injective map.

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Claim: $\bar{\psi}$ is injective.

Suppose $v \in V$ s.t. $v \neq 0$. Let $\beta = (v_1, \dots, v_n)$ be an ordered basis with $v_1 = v$ & $\beta^* = (f_1, \dots, f_n)$ be the dual basis.

Then $f_1(v_1) = 1$.

$\Rightarrow v^{**}(f_1) = 1$.

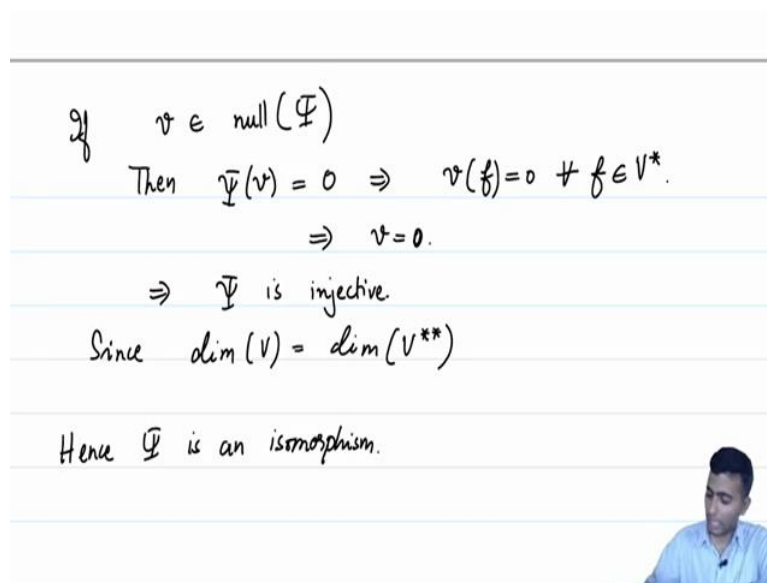
$\Rightarrow v^{**} \neq 0$.

So claim, ψ is injective, if we show this, notice that it is enough to show that ψ is injective to show that ψ is surjective as well, and therefore it turns out to be an isomorphism. Why is ψ injective? Let us consider some element in the null space. So let ψ of v be equal to 0, so you will show that that forces v to be equal to 0.

So suppose, v is an element such that v is not equal to 0. Suppose v is a non-zero element. And then this is the first time we will be going down to our basis in this proof. Let β equal to v_1 to v_n be an ordered basis with v_1 equal to v , extend v to a basis of V and β^* be equal to f_1 to f_n be the dual basis corresponding to β .

But what is the property of the dual basis? If you consider f_1 , f_1 satisfies some nice properties. Then f_1 of v_1 is equal to 1, but our v_1 is v and f_1 of v_1 or f_1 of v is precisely equal to $v^{**}(f_1)$. This implies $v^{**}(f_1)$ is equal to 1, and what does this imply? The meaning of this is that this implies v^{**} is not equal to the 0 element, but what do we want? Why are we considering this? Why are we doing this?

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if $v \in \text{null}(\Psi)$
Then $\Psi(v) = 0 \Rightarrow v(f) = 0 \forall f \in V^*$
 $\Rightarrow v = 0$
 $\Rightarrow \Psi$ is injective.
Since $\dim(V) = \dim(V^{**})$
Hence Ψ is an isomorphism.

So if v belongs to the null space of ψ , we want to show that ψ is injective. So suppose v belongs to the null space of ψ , then what do we have? Then ψ of v is equal to, identically equal to or is equal to the 0 element of v star star, which implies that v of f is equal to 0 for all f in v star, but we just checked that if v is non-zero, there exist at least one element in v star where v of f is equal to 1, but that will not happen if, in fact, if v of f is equal to zero for all f , then it forces because of the observation we just did, this forces our v to be identically equal to or v to equal to the 0 vector.

This implies that ψ is injective because null space being the 0 vector space (50:09) ψ is injective, since dimension of v is equal to dimension of v star star. Argument using the dimension theorem which we have already proved shows that the image of ψ or the range of ψ has dimension n and v star star also has dimension n , therefore the range has to be equal to v star star and hence surjective.

What do we know about injective and surjective or bijective linear transformation? We know that it is invertible and hence an isomorphism. Hence ψ is an isomorphism. So indeed, when proving that ψ is injective we did use a basis, an ordered basis and considered its dual, but if you observe carefully, this correspondence between v and v star star does not refer to any ordered basis. There is a very natural correspondence between v and v star star. So this is a notion which is quite essential when we do more analysis of function spaces.