

## MTH 100: Lecture 26

Definition: A linear transformation  $T: V \rightarrow W$  is called an isomorphism if it is injective and surjective (i.e. if  $\text{Range } T = W$ )

Proposition: Let  $V$  and  $W$  be finite dimensional spaces.

- (a) An isomorphism  $T: V \rightarrow W$  takes any arbitrary basis of  $V$  to a basis of  $W$ .
- (b) Conversely, if a linear transformation  $T: V \rightarrow W$  takes some basis of  $V$  to a basis of  $W$ , then it is an isomorphism.

Sketch of a Proof:

(a): Given:  $T: V \rightarrow W$  is an isomorphism

Now,  $T$  is onto  $\Rightarrow \text{Range } T = W \Rightarrow \text{Rank } T = \dim W$

$T$  is 1-1  $\Rightarrow \text{Ker } T = \{0\} \Rightarrow \text{nullity } T = 0$

By the Rank Theorem,  
 $\text{Rank } T + \text{nullity } T = \dim V = n$  (say)  
 $\Rightarrow \dim W + 0 = n$   
 $= \dim W = \dim V = n$

Let  $\{v_1, \dots, v_n\}$  be a Basis of  $V$ .

Then  $\{Tv_1, \dots, Tv_n\}$  is a spanning set of  $\text{Range } T = W$ .  
Since  $\dim W = n$ ,  $\{Tv_1, \dots, Tv_n\}$  forms a basis of  $W$ .

(b)  $\Leftarrow$  :

Assume that  $\dim V = n$  and  
 $T$  takes some basis  $\{v_1, \dots, v_n\}$  of  $V$   
to a basis  $\{Tv_1, \dots, Tv_n\}$  of  $W$   
Therefore  $\dim W = n$  and  $\text{Rank } T = n$   
Hence  $\text{Range } T = W$  and  $T$  is onto.

Now using nullity Theorem :

$$\begin{aligned}\text{Rank } T + \text{nullity } T &= \dim V = n \\ \Rightarrow n + \text{nullity } T &= n \\ \Rightarrow \text{nullity } T &= 0 \\ \Rightarrow \ker T &= \{0\} \\ \Rightarrow T &\text{ is 1-1.}\end{aligned}$$

Therefore  $T$  is an isomorphism.

Proposition: Two finite dimensional vector spaces  $V$  and  $W$  (over the same field  $F$ ) are isomorphic if and only if  $\dim V = \dim W$ .

Proof:  $\Rightarrow$ : Assume  $T: V \rightarrow W$  is an isomorphism.  
Want to show  $\dim V = \dim W$ .

Suppose  $\dim V = n$  and let  $\{v_1, \dots, v_n\}$  be a basis of  $V$ .  
Then by the previous proposition  $\{Tv_1, \dots, Tv_n\}$  is a basis of  $W$ . Hence  $\dim W = n$ .  
So,  $\boxed{\dim V = \dim W}$

$\Leftarrow$ : Assume that  $\dim V = \dim W$ .  
Want to show that  $V$  and  $W$  are isomorphic.

Let  $\{v_1, \dots, v_n\}$  be a basis of  $V$   
and  $\{w_1, \dots, w_n\}$  be a basis of  $W$ .  
Consider the unique linear transformation  
 $T: V \rightarrow W$  such that  $Tv_i = w_i$   
for  $i = 1, 2, \dots, n$ .

Since  $T$  takes a basis of  $V$  to a basis of  $W$ ,  
by the previous proposition,  $T$  is an isomorphism  
and hence  $V$  and  $W$  are isomorphic.

Remark: Every vector space of dimension  $n$  over a field  $F$  is isomorphic to  $F^n$ . In particular, every vector space of dimension  $n$  over  $\mathbb{R}$  is isomorphic to  $\mathbb{R}^n$ .

## An Important Linear Transformation:

### Left multiplication by a Matrix:

Let  $A$  be a  $m \times n$  matrix.

Define  $T_A : \mathbb{R}^n \longrightarrow \mathbb{R}^m$

by  $T_A(x) = Ax$

[Note that  $\underbrace{A_{m \times n} x_{n \times 1}}_{m \times 1}$  is a  $m \times 1$  matrix]

- $T_A$  is a linear transformation.

- For  $x, y \in \mathbb{R}^n$ ,

$$\begin{aligned} T_A(x+y) &= A(x+y) = Ax + Ay \\ &= T_A(x) + T_A(y) \end{aligned}$$

- For  $x \in \mathbb{R}^n$  and  $c \in \mathbb{R}$ ,

$$T_A(cx) = A(cx) = cAx = cT_A(x)$$

Hence  $T_A$  is a linear transformation.

### Consider the Reverse Problem:

Suppose  $V$  and  $W$  are finite dimensional vector spaces over the field  $F$ .

Suppose  $T: V \rightarrow W$  is a linear transformation  
we will associate a matrix with this  
linear transformation.

### Coordinate Systems:

Suppose  $V$  is a finite dimensional vector space.

An ordered basis for a finite dimensional  
vector space  $V$  is a finite sequence of  
vectors which is linearly independent  
and spans  $V$ .

In other words,

an ordered basis is a basis with  
the vectors taken in a specified fixed  
order.

Thus,  
given an ordered basis of  $V$

$$B = \{u_1, \dots, u_n\}, \text{ we can}$$

express any vector  $u \in V$   
uniquely in the form  $u = x_1 u_1 + x_2 u_2 + \dots + x_n u_n$

The scalars  $x_i$  are called the coordinates of  $u$   
relative to the (ordered) basis  $B$

Remark: Given a fixed ordered basis  $B$  for a finite  
dimensional vector space  $V$ , we can find an  $n$ -tuple in  $F^n$

(usually  $F$  is  $\mathbb{R}$  or  $\mathbb{C}$ ) corresponding to any vector  $u$  in  $V$  as follows:  $u \longrightarrow (x_1, x_2, \dots, x_n)$  where  $x_i$  are the coordinates of  $u$  relative to  $B$

Rather than the  $n$ -tuple  $(x_1, x_2, \dots, x_n)$  we express it as a column vector

$$\begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

• This vector is called the coordinate vector of  $u$  (relative to  $B$ ) and is written  $[u]_B = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$

Ex: Let  $V = R_n(t)$  be the vector space of polynomials of degree  $\leq n$

Then  $B = \{1, t, t^2, \dots, t^n\}$  is an ordered basis of  $R_n(t)$

Let  $v = 2t^3 \in R_n(t)$

Then  $v$  can be written as:

$$v = 0 \cdot 1 + 0 \cdot t + 0 \cdot t^2 + 2 \cdot t^3 + 0 \cdot t^4 + \dots + 0 \cdot t^n$$

Hence the coordinate vector of  $v$  relative to  $B$

is  $[v]_B = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 2 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$