## **Linear Transformations**

Def. Let V and W be vector spaces. We call a function  $T: V \to W$  a **linear** transformation from V to W if for all  $u, v \in V$  and  $c \in \mathbb{R}$ 

a. 
$$T(u + v) = T(u) + T(v)$$

b. 
$$T(cv) = cT(v)$$
.

Theorem: Let  $T: V \to W$  be a linear transformation from a vector space V to a vector space W. Then for  $u, v, u_1, ..., u_n \in V$ ,  $a, b, a_1, ..., a_n \in \mathbb{R}$ 

1. 
$$T(0) = 0$$

2. 
$$T(v - u) = T(v) - T(u)$$

3. 
$$T(au + bv) = aT(u) + bT(v)$$

4. 
$$T(\sum_{i=1}^{n} a_i u_i) = \sum_{i=1}^{n} a_i T(u_i)$$

Proof:

1. 
$$T(0) = T(2(0)) = 2T(0) \implies T(0) = 0$$
.

2. 
$$T(v-u) = T(v + (-u)) = T(v) + T(-u)$$
  
=  $T(v) - T(u)$ .

3. 
$$T(au + bv) = T(au) + T(bv) = aT(u) + bT(v)$$
.

4. 
$$T(\sum_{i=1}^{n} a_i u_i) = T(a_1 u_1 + a_2 u_2 + \cdots + a_n v_n)$$
  

$$= T(a_1 u_1) + T(a_2 u_2 + \cdots + a_n u_n)$$

$$= a_1 T(u_1) + T(a_2 u_2) + T(a_3 u_3 + \cdots + a_n u_n)$$

$$\vdots$$

$$= a_1 T(u_1) + a_2 T(u_2) + \cdots + a_n T(u_n)$$

$$= \sum_{i=1}^{n} a_i T(u_i).$$

Ex. Show that  $T: V \to W$  is a linear transformation if and only if T(cv + u) = cT(v) + T(u) for all  $u, v \in V$ ,  $c \in \mathbb{R}$ .

Proof: Case #3 of the previous theorem shows if *T* is linear then

$$T(cv + u) = cT(v) + T(u)$$
 for all  $u, v \in V$ ,  $c \in \mathbb{R}$ .

Now let's show that if T(cv+u)=cT(v)+T(u) for all  $u,v\in V,\ c\in\mathbb{R}$  then T is linear.

We must show:

- 1. T(u+v) = T(u) + T(v) for all  $u, v \in V$
- 2. T(cv) = cT(v) for all  $c \in \mathbb{R}$ .
- 1. Since T(cv + u) = cT(v) + T(u) for all  $u, v \in V$ ,  $c \in \mathbb{R}$ , it's true for c = 1.

Thus 
$$T(v + u) = T(v) + T(u)$$
 for all  $u, v \in V$ .

2. If we take 
$$u=0$$
 then  $T(cv+0)=cT(v)+T(0)$  
$$\Rightarrow T(cv)=cT(v)+0=cT(v).$$

Ex. Show that  $T: \mathbb{R}^2 \to \mathbb{R}^2$  by  $T(< a_1, a_2 >) = < a_1 + 2a_2, a_1 >$  is a linear transformation.

By the previous example we just need to show that T(cv + u) = cT(v) + T(u) for all  $u, v \in \mathbb{R}^2$ ,  $c \in \mathbb{R}$ .

For any 
$$u,v\in\mathbb{R}^2$$
, we have  $u=< x_1,y_1>$ ,  $v=< x_2,y_2>$  and 
$$T(c< x_1,y_1>+< x_2,y_2>)=T(< cx_1+x_2,\ cy_1+y_2>)$$
 
$$=< cx_1+x_2+2(cy_1+y_2),\ cx_1+x_2>.$$

$$\begin{split} cT() + T() &= c() + () \\ &= < cx_1+2cy_1+x_2+2y_2,\ cx_1+x_2> \\ &= < cx_1+x_2+2(cy_1+y_2),\ cx_1+x_2> \\ &= T(c+< x_2,y_2>). \end{split}$$

So T(cu + v) = cT(u) + T(v) for all  $u, v \in \mathbb{R}^2$ ,  $c \in \mathbb{R}$  and T is linear.

Ex. Define  $T: \mathbb{R}^2 \to \mathbb{R}^2$  by  $T(< a_1, a_2 >) = < -a_1, a_2 >$ . T is a reflection about the y-axis. Show that T is a linear transformation.

For any 
$$u,v\in\mathbb{R}^2$$
, we have  $u=< x_1,y_1>$  ,  $v=< x_2,y_2>$  and 
$$T(c< x_1,y_1>+< x_2,y_2>)=T(< cx_1+x_2,\ cy_1+y_2>)$$
 
$$=<-(cx_1+x_2),\ cy_1+y_2>.$$

$$cT(\langle x_1, y_1 \rangle) + T(\langle x_2, y_2 \rangle) = c(\langle -x_1, y_1 \rangle) + \langle -x_2, y_2 \rangle$$

$$= \langle -cx_1 - x_2, y_1 + y_2 \rangle$$

$$= \langle -(cx_1 + x_2), cy_1 + y_2 \rangle$$

$$= T(c \langle x_1, y_1 \rangle) + \langle x_2, y_2 \rangle$$

So *T* is linear.

Ex. Show that  $T: \mathbb{R}^2 \to \mathbb{R}^2$  by T(< a, b >) = < a + 3, b > is not a linear transformation.

Notice that if  $c \neq 1$  then  $T(cv) \neq cT(v)$ :

$$T(c < a, b >) = T(< ca, cb >) = < ca + 3, cb >$$

$$cT(\langle a, b \rangle) = c(a+3, b) = (c(a+3), cb) \neq (ca+3, cb).$$

It's also true that  $T(<0,0>) \neq <0,0>$  and  $T(u+v)\neq T(u)+T(v)$  in general.

Ex. Show that  $T: \mathbb{R}^2 \to \mathbb{R}^2$  by  $T(<\alpha, b>) = <\alpha, 0>$  (called a **projection**) is a linear transformation.

If we let  $u=< a_1, b_1>$  ,  $\ v=< a_2, b_2>$  and  $c\in \mathbb{R}$  then we have:

$$T(cu + v) = T(c < a_1, b_1 > + < a_2, b_2 >)$$
  
=  $T(< ca_1 + a_2, b_1 + b_2 >)$   
=  $< ca_1 + a_2, 0 >$ 

$$cT(u) + T(v) = cT(\langle a_1, b_1 \rangle) + T(\langle a_2, b_2 \rangle)$$
  
=  $c < a_1, 0 \rangle + \langle a_2, 0 \rangle$   
=  $\langle ca_1 + a_2, 0 \rangle$ 

Thus T(cu + v) = cT(u) + T(v) for all  $u, v \in \mathbb{R}^2$ ,  $c \in \mathbb{R}$ , so T is linear.

Ex. Show that  $T: P_n(\mathbb{R}) \to P_n(\mathbb{R})$  by T(f(x)) = f'(x) is a linear transformation.

Let 
$$f(x), g(x) \in P_n(\mathbb{R})$$
 and  $c \in \mathbb{R}$ . 
$$T(cf(x) + g(x)) = (cf(x) + g(x))'$$
$$= cf'(x) + g'(x)$$
$$= cT(f(x)) + T(g(x)).$$

So T is a linear transformation.

Ex. Let V=C[a,b], the vector space of continuous real valued function on [a,b]. Define  $T:V\to\mathbb{R}$  by  $T\bigl(f(x)\bigr)=\int_a^b f(x)dx$ . Show that T is a linear transformation.

Let 
$$f(x), g(x) \in C[a, b]$$
 and  $c \in \mathbb{R}$  then 
$$T(cf(x) + g(x)) = \int_a^b (cf(x) + g(x)) dx$$
$$= c \int_a^b f(x) dx + \int_a^b g(x) dx$$
$$= cT(f(x)) + T(g(x)).$$

So *T* is a linear transformation.

Ex. Show that  $T:\mathbb{R}^2\to\mathbb{R}^2$  by  $T(< a_1,a_2>)=< a_1a_2,a_2>$  is not a linear transformation.

Let 
$$v = < a_1, a_2 > \text{ and } c \in \mathbb{R}, \ c \neq 1 \text{ then}$$
 
$$T(cv) = T(c < a_1, a_2 >)$$
 
$$= T(< ca_1, \ ca_2 >)$$
 
$$= < c^2 a_1 a_2, \ ca_2 >$$

$$cT(\langle a_1, a_2 \rangle) = c \langle a_1 a_2, a_2 \rangle$$
  
=  $\langle ca_1 a_2, ca_2 \rangle \neq T(cv)$ .

Thus *T* is not a linear transformation.

In this case it also happens to be true that  $T(u+v) \neq T(u) + T(v)$  in general.

Ex: Suppose  $T: \mathbb{R}^2 \to \mathbb{R}^3$  is a linear transformation such that T(<2,3>) = <0,3,4> and T(<3,2>) = <-1,2,3>. Find T(<2,8>).

Notice that we can write < 2.8 > as a linear combination of < 2.3 > and < 3.2 >.

$$a < 2, 3 > +b < 3, 2 > = < 2,8 >$$
 $< 2a + 3b, 3a + 2b > = < 2,8 >$ 
 $\Rightarrow 2a + 3b = 2$ 
 $3a + 2b = 8.$ 

Solving these simultaneous equations we get a = 4, b = -2.

Thus we have: 4 < 2, 3 > -2 < 3, 2 > = < 2,8 >.

Hence we have:

$$T(\langle 2,8 \rangle) = T(4 \langle 2,3 \rangle -2 \langle 3,2 \rangle)$$
  
=  $4T(\langle 2,3 \rangle) - 2T(\langle 3,2 \rangle)$   
=  $4 \langle 0,3,4 \rangle -2 \langle -1,2,3 \rangle$   
=  $\langle 0,12,16 \rangle -\langle -2,4,6 \rangle$   
=  $\langle 2,8,10 \rangle$ .

In fact, given any  $< x, y > \in \mathbb{R}^2$  we can find T(< x, y >) by writing < x, y > as a linear combination of < 2,3 > and < 3,2 >. In this case we would need to solve:

$$a < 2, 3 > +b < 3, 2 > = < x, y >$$

$$\Rightarrow 2a + 3b = x$$

$$3a + 2b = y$$

for a and b in terms of x and y.

Two important linear transformations are:

- 1. The identity linear trasformation  $I: V \to V$ , where I(v) = v for all  $v \in V$ .
- 2. The zero linear transformation  $T_0: V \to V$ , where  $T_0(v) = 0$  for all  $v \in V$ .

Def. Let V and W be vector spaces, and let  $T:V\to W$  be linear. The **null space** or **kernel** of T,N(T), is the set of  $v\in V$  such that T(v)=0. The **range** or **image** of T is the subset of W given by  $R(T)=\{T(v)|v\in V\}$ .

Ex. Let  $I:V \to V$  and  $T_0:V \to W$  be the identity and zero transformations. Then

$$N(I) = \{0\}$$
  $N(T_0) = V$   
 $R(I) = V$   $R(T_0) = \{0\}.$ 

Ex. Let  $T \colon \mathbb{R}^3 \to \mathbb{R}^2$  be the linear transformation given by

$$T(\langle a_1, a_2, a_3 \rangle) = \langle a_1 + a_2, 3a_3 \rangle$$
. Find  $N(T)$  and  $R(T)$ .

To find N(T) we need to find all vectors  $\langle a_1, a_2, a_3 \rangle$  such that

$$T(\langle a_1, a_2, a_3 \rangle) = \langle a_1 + a_2, 3a_3 \rangle = \langle 0, 0 \rangle.$$
 
$$a_1 + a_2 = 0 \quad \Rightarrow \quad a_1 = -a_2$$
 
$$3a_3 = 0 \quad \Rightarrow \quad a_3 = 0$$
 So  $N(T) = \{\langle a_1 - a_1 0 \rangle \in \mathbb{R}^3 | a \in \mathbb{R}\}.$ 

$$R(T) = \{T(\langle a_1, a_2, a_3 \rangle) = \langle a_1 + a_2, 3a_3 \rangle \in \mathbb{R}^2 | a_1, a_2, a_3 \in \mathbb{R}\}.$$

Let < x, y > be any vector in  $\mathbb{R}^2$ . Let's show that  $< x, y > \in R(T)$ .

$$< a_1 + a_2, 3a_3 > = < x, y >.$$
  
So  $a_1 + a_2 = x$   
 $3a_3 = y.$ 

In particular if  $a_1=x$ ,  $a_2=0$ ,  $a_3=\frac{y}{3}$  then

$$T\left(< x, 0, \frac{y}{3} > \right) = < x, y > \implies R(T) = \mathbb{R}^2.$$

Theorem: Let V and W be vector spaces and  $T:V\to W$  be linear. Then N(T) is a subspace of V and R(T) is a subspace of W.

Proof: First we show that N(T) is a subspace of V.

Suppose that  $v, w \in N(T)$  and  $c \in \mathbb{R}$  then

$$T(v+w) = T(v) + T(w) = 0 + 0 = 0 \implies v+w \in N(T).$$

$$T(cv) = cT(v) = c(0) = 0$$
  $\Rightarrow$   $cv \in N(T)$ .

So N(T) is a subspace of V.

Now we show that R(T) is a subspace of W.

Suppose that  $w_1, w_2 \in R(T)$  and  $c \in \mathbb{R}$  then there exist  $v_1, v_2 \in V$  such that  $T(v_1) = w_1$  and  $T(v_2) = w_2$ .

Thus we have:

$$T(v_1 + v_2) = T(v_1) + T(v_2) = w_1 + w_2 \implies w_1 + w_2 \in R(T).$$

$$T(cv_1) = cT(v_1) = cw_1 \implies cw_1 \in R(T).$$

So R(T) is a subspace of W.

Theorem: Let V and W be vector spaces and  $T:V\to W$  be linear. If  $B=\{v_1,\ldots,v_n\}$  is a basis for V then  $R(T)=span(T(B))=span\{T(v_1),\ldots,T(v_n)\}.$ 

Proof:  $T(v_i) \in R(T)$  for each i.

R(T) is a subspace of  $W \implies R(T)$  contains  $span\{T(v_1), ..., T(v_n)\}$ .

Now suppose  $w \in R(T)$  then w = T(v) for some  $v \in V$ .

Since  $\{v_1, \dots, v_n\}$  is a basis for V,  $v = \sum_{i=1}^n a_i v_i$  for some  $a_1, \dots a_n \in \mathbb{R}$ .

Since *T* is linear we have:

$$w = T(v) = T(\sum_{i=1}^{n} a_i v_i) = \sum_{i=1}^{n} a_i T(v_i) \in span(T(B)).$$

So R(T) is contained in the  $span\{T(v_1), ..., T(v_n)\}$ .

Since  $R(T) \supseteq span\{T(v_1), ..., T(v_n)\}$  and  $R(T) \subseteq span\{T(v_1), ..., T(v_n)\}$ ,  $\Rightarrow R(T) = span\{T(B)\} = span\{T(v_1), ..., T(v_n)\}$ .

Ex. Define the linear transformation  $T: P_2(\mathbb{R}) \to M_{2\times 2}(\mathbb{R})$  by

$$T(f(x)) = \begin{bmatrix} f(3) - f(1) & 0 \\ 0 & f(0) \end{bmatrix}.$$

Find a basis for R(T) and dim R(T).

Since  $B = \{1, x, x^2\}$  is a basis for  $P_2(\mathbb{R})$ 

$$R(T) = span\{T(1), T(x), T(x^{2})\}\$$

$$= span\{\begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 8 & 0 \\ 0 & 0 \end{bmatrix}\}\$$

$$= span\{\begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix}\}.$$

Since  $\begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$ ,  $\begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix}$  are linearly independent (one is not a nonzero multiple of the other) they form a basis for R(T). Hence dim R(T) = 2.

Def. Let V and W be vector spaces and  $T: V \to W$  be linear. If N(T) and R(T) are finite dimensional, then we define the nullity of , Nullity(T) = dimN(T), and the rank of T, Rank(T) = dimR(T).

Theorem (dimension theorem): Let V and W be vector spaces and  $T:V\to W$  be linear. If V is a finite dimensional vector space then

$$Nullity(T) + Rank(T) = \dim(V).$$

Proof: Suppose that  $\dim(V) = n$ ,

dimN(T) = k and  $\{v_1, \dots v_k\}$  is a basis for N(T).

Extend  $\{v_1, \dots v_k\}$  to a basis  $\{v_1, \dots v_n\}$  of V.

Claim:  $S = \{T(v_{k+1}), ..., T(v_n)\}$  is a basis for R(T).

Since  $T(v_1)=T(v_2)=\cdots=T(v_k)=0$ , we know from the previous theorem that S generates R(T) since

$$R(T) = span\{T(v_1), ..., T(v_n)\} = span\{T(v_{k+1}), ..., T(v_n)\}.$$

Now let's show that S is linearly independent. Suppose

$$b_{k+1}T(v_{k+1}) + \dots + b_nT(v_n) = 0.$$

Since T is linear:  $T(b_{k+1}(v_{k+1}) + \cdots + b_n(v_n)) = 0.$ 

So 
$$b_{k+1}(v_{k+1}) + \dots + b_n(v_n) \in N(T)$$
.

Since  $\{v_1, \dots v_k\}$  is a basis for N(T) there exist  $c_1, \dots, c_k \in \mathbb{R}$  such that

$$c_1 v_1 + \dots + c_k v_k = b_{k+1}(v_{k+1}) + \dots + b_n(v_n)$$
  
$$c_1 v_1 + \dots + c_k v_k - b_{k+1}(v_{k+1}) - \dots - b_n(v_n) = 0.$$

But  $\{v_1, \dots v_n\}$  are linearly independent so  $b_{k+1}, \dots, b_n = 0$ .

Hence S is linearly independent and a basis for R(T).

Thus Rank(T) = n - k and  $Nullity(T) + Rank(T) = \dim(V)$ .

Ex. Let  $T: \mathbb{R}^3 \to \mathbb{R}^2$  by  $T(< a_1, a_2, a_3 >) = < a_1 + a_2, 3a_3 >$ . Find the  $\dim(R(T))$ .

We found earlier that

$$N(T) = \{ < a, -a, 0 > | \ a \in \mathbb{R} \} = \{ a < 1, -1, 0 > | \ a \in \mathbb{R} \}.$$
 So  $N(T)$  has a basis of  $< 1, -1, 0 >$  and  $\dim(N(T)) = 1$ .

By the previous theorem we know that  $\dim(R(T)) = 2$  since  $\dim(\mathbb{R}^3) = 3$ .

$$\dim(N(T)) + \dim(R(T)) = \dim(\mathbb{R}^3)$$

$$1 + \dim(R(T)) = 3$$

$$\Rightarrow \dim(R(T)) = 2.$$

Def. Let V and W be vector spaces and  $T:V\to W$  be linear. T is called **one-to-one** if  $T(v_1)=T(v_2)$  implies  $v_1=v_2$ . T is called **onto** if given any  $w\in W$  there exists at least one  $v\in V$  such that T(v)=w.

Theorem: Let V and W be vector spaces and  $T: V \to W$  be linear. Then T is one-to-one if and only if  $N(T) = \{0\}$ .

Proof: Suppose T is one-to-one and  $v \in N(T)$ .

Then 
$$T(v) = 0 = T(0)$$
.

But T is one-to-one so v = 0.

Now suppose that  $N(T) = \{0\}$  and T(x) = T(y).

Then 
$$0 = T(x) - T(y) = T(x - y)$$
, so  $x - y \in N(T)$ .

Thus x - y = 0 and x = y. Thus T is one-to-one.

Theorem: Let V and W be vector spaces of equal (finite) dimension, and let  $T: V \to W$  be linear. Then the following are equivalent:

- a. T is one-to-one
- b. T is onto
- c. Rank(T) = dim(V).

Proof: Recall that  $Nullity(T) + Rank(T) = \dim(V)$ .

T is one-to-one  $\Leftrightarrow N(T) = \{0\}.$ 

$$N(T) = \{0\} \iff Nullity(T) = 0.$$

$$Nullity(T) = 0 \iff Rank(T) = \dim(V).$$

$$Rank(T) = \dim(V) \iff Rank(T) = \dim(W).$$

$$Rank(T) = dim(W) \iff R(T) = W$$
, i.e.  $T$  is onto.

Note: The previous theorem does not hold if V and W are infinite dimensional.

For example, let 
$$V=W=P(\mathbb{R})$$
 and  $T:P(\mathbb{R})\to P(\mathbb{R})$  by

1. 
$$T(f(x)) = \int_0^x f(t)dt$$
.

T is one-to-one because T(f(x)) = T(g(x)) means

$$\int_0^x f(t)dt = \int_0^x g(t)dt \text{ for all } x.$$

$$\Rightarrow \int_0^x (f(t) - g(t)) dt = 0 \text{ for all } x.$$

$$\Rightarrow f(x) = g(x).$$

However, T is not onto as  $T(f(x)) \neq \text{constant function}$ .

$$2. T(f(x)) = f'(x).$$

T is not one-to-one because  $T\big(f(x)\big) = T(g(x))$  means  $f'(x) = g'(x) \implies f(x) = g(x) + \mathcal{C} \text{ for any constant } \mathcal{C}.$ 

However, T is onto because given any  $g(x)=a_0+a_1x+\cdots+a_nx^n\in P(\mathbb{R})$  then  $f(x)=\int g(x)dx=a_0x+\frac{1}{2}a_1x^2+\cdots+\frac{a_n}{n}x^{n+1}+C$  has the property that  $T\big(f(x)\big)=g(x)$ .

Ex. Let  $T: P_2(\mathbb{R}) \to P_2(\mathbb{R})$  by T(f(x)) = xf'(x). Show that T is a linear transformation. Determine if T is one-to-one and/or onto.

To show that T is a linear transformation, let f(x),  $g(x) \in P_2(\mathbb{R})$  and  $c \in \mathbb{R}$ . Then we have:

$$T(cf(x) + g(x)) = x[cf(x) + g(x)]'$$

$$= x[cf'(x) + g'(x)]$$

$$= cxf'(x) + xg'(x)$$

$$= cT(f(x)) + T(g(x)).$$

Thus *T* is linear.

T is one-to-one  $\iff N(T) = \{0\}.$ 

$$T(f(x)) = 0$$
$$xf'(x) = 0$$
$$\Rightarrow x = 0 \text{ or } f'(x) = 0.$$

But 
$$f'(x) = 0 \implies f(x) = constant$$
.

Thus all constant functions  $f(x) \in N(T)$ .

Hence  $N(T) \neq \{0\}$  and T is not one-to-one.

In fact N(T) is spanned by f(x) = 1 hence  $\dim(N(T)) = 1$ .

Since 
$$Nullity(T) + Rank(T) = \dim(P_2(\mathbb{R})) = 3$$
  
  $Rank(T) = 2$  so  $R(T) \neq P_2(\mathbb{R})$  and  $T$  is not onto.

R(T) is spanned by T(1), T(x),  $T(x^2)$  because  $\{1, x, x^2\}$  is a basis for  $P_2(\mathbb{R})$ .  $R(T) = span\{T(1), T(x), T(x^2)\}.$ 

$$T(1) = x(0) = 0$$
 since if  $f(x) = 1$ ,  $f'(x) = 0$ .

$$T(x) = x(1) = x$$
 since if  $f(x) = x$ ,  $f'(x) = 1$ .

$$T(x^2) = x(2x) = 2x^2$$
 since if  $f(x) = x^2$ ,  $f'(x) = 2x$ .

Thus 
$$R(T) = span\{0, x, 2x^2\} = span\{x, 2x^2\}$$
  
=  $\{p(x) \in P_2(\mathbb{R}) | p(x) = ax + 2bx^2, a, b \in \mathbb{R}\}$ 

Since x and  $2x^2$  are linearly independent, dimR(T) = 2.

Ex. Suppose that  $T: \mathbb{R}^2 \to \mathbb{R}^2$  is linear and T(<0,1>) = <2,3> and T(<2,-1>) = <1,2>. Is T one-to-one?

By the previous theorem with  $V=W=\mathbb{R}^2$ , we have dimV=dimW=2. T is one-to-one if T is onto (i.e.  $R(T)=\mathbb{R}^2$ ).

But since < 0,1 > and < 2,-1 > are linearly independent (one is not a multiple of the other), they form a basis for  $V = \mathbb{R}^2$ . Thus we have:

$$R(T) = span\{T(<0,1>), T(<2,-1>)\} = span\{<2,3>,<1,2>\}.$$

But < 2,3 > and < 1,2 > are also linearly independent and thus a basis for  $W = \mathbb{R}^2$ .

Hence  $R(T) = \mathbb{R}^2$  and T is onto  $\implies T$  is one-to-one

Ex. Let V and W be vector spaces of equal (finite) dimension, and let  $T: V \to W$  be linear. Show that if  $\dim(V) > \dim(W)$ , then T can't be one-to-one.

$$Nullity(T) + Rank(T) = \dim(V) > \dim(W).$$

R(T) is a subspace of W so  $\dim(R(T)) \leq \dim(W)$ .

Thus  $\dim(N(T)) + Rank(T) > \dim(W) \implies \dim(N(T)) \ge 1$ .

Hence  $N(T) \neq \{0\}$  and T is not one-to-one.

Theorem: Let V and W be vector spaces and suppose that  $\{v_1, \dots, v_n\}$  is a basis for V. For  $w_1, \dots, w_n \in W$ , there exists exactly one linear transformation  $T: V \to W$  such that  $T(v_i) = w_i$ .

Proof: Given any  $v \in V$ ,  $v = a_1v_1 + \cdots + a_nv_n$ , where  $a_1, \dots, a_n$  are unique.

Define 
$$T: V \to W$$
 by  $T(v) = a_1 w_1 + \cdots + a_n w_n$ .

Notice that T is linear since if  $u, s \in V$ ,  $d \in \mathbb{R}$  we have:

$$u = b_1 v_1 + \cdots b_n v_n$$
 and  $s = c_1 v_1 + \cdots c_n v_n$  so we get:  $du + s = (db_1 + c_1)v_1 + \cdots + (db_n + c_n)v_n$  and  $T(du + s) = (db_1 + c_1)T(v_1) + \cdots + (db_n + c_n)T(v_n)$  
$$= (db_1 + c_1)w_1 + \cdots + (db_n + c_n)w_n$$
 
$$= d(b_1 w_1 + \cdots + b_n w_n) + (c_1 w_1 + \cdots + c_n w_n)$$
 
$$= dT(u) + T(s).$$

T is unique because if  $U: V \to W$  is a linear transformation with  $U(v_i) = w_i$  then

$$U(v) = a_1 U(v_1) + \dots + a_n U(v_n)$$

$$= a_1 w_1 + \dots + a_n w_n$$

$$= T(v). \qquad (Since  $T(v_i) = w_i$ )$$

Hence U = T.

Corollary: Let V and W be vector spaces and suppose that  $\{v_1, ..., v_n\}$  is a basis for V. If  $U, T: V \to W$  are linear transformations with  $U(v_i) = T(v_i)$  for i = 1, ..., n, then U = T.

In other words, a linear transformation is defined by what it does to a set of basis vectors.