## Linear Maps, Matrices and Linear Sytems

Let  $T : U \to V$  be linear. We defined ker $(T) = \{x | x \in U, T(x) = 0\}$  and  $im(T) = \{y | T(x) = y \text{ for some } x \in U\}$ . According to the dimension equality we have that dim(ker(T)) + dim(im(T)) = dim(U) = n. We have that the linear map *T* is one-one or injective iff ker $(T) = \{0\}$ . If U = V then *T* is injective iff *T* is surjective (onto). This is an easy consequence of the dimension equality.

For any linear map *T* we have a matrix representation. The matrix *A* of *T* depends on the chosen bases  $\alpha$  and  $\beta$  of *U* and *V*, repectively:

$$Mat(T; \alpha_1, \alpha_2, \dots, \alpha_n; \beta_1, \beta_2, \dots, \beta_m) = (a_{ij})$$
 where  $T(\alpha_j) = \sum_{i=1}^m a_{ij}\beta_i, j = 1, \dots, n; i = 1, \dots, m$ 

The matrix for *T* is an  $m \times n$  matrix where dim(V) = m, dim(U) = n. Each of the *n* columns of *A* contain the *m* components of  $T(\alpha_j)$  with respect to the basis  $\beta_j$ . Let *x* be a vector in *U*. If

 $x = \sum_{j=1}^{n} x_j \alpha_j, \text{then } T(x) = T(\sum_{j=1}^{n} x_j \alpha_j) = \sum_{j=1}^{n} x_j T(\alpha_j) = \sum_{j=1}^{n} x_j \sum_{i=1}^{m} a_{ij} \beta_i = \sum_{i=1}^{m} (\sum_{j=1}^{n} a_{ij} x_j) \beta_i = \sum_{j=1}^{m} y_i \beta_i$ where

$$y_{i} = \sum_{j=1}^{n} a_{ij} x_{j}$$
Thus
$$\begin{pmatrix} a_{11} & a_{12} & \dots & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & \dots & a_{2n} \\ \dots & \dots & \ddots & \dots & \dots & \dots \\ a_{m1} & a_{m2} & \dots & \dots & a_{mn} \end{pmatrix} \begin{pmatrix} x_{1} \\ x_{2} \\ \dots \\ \dots \\ x_{n} \end{pmatrix} = \begin{pmatrix} y_{1} \\ y_{2} \\ \dots \\ \dots \\ y_{m} \end{pmatrix}$$

Thus the map T(x) = y has a coordinate representation as Ax = y. with this in mind, we see that

 $\dim(im(T)) = \dim(\text{columnspace } A) = s, \dim(\ker(T)) = \dim(\text{solution space for } Ax = 0) = n - r$ 

.

where *r* is the row rank of *A*. By the dimension equality we have that s + (n - r) = n which is s = r that is **row rank=column rank**.

Example 1:

Let 
$$U = \mathbb{R}^3$$
,  $V = \mathbb{R}^4$  then the matrix  $A = \begin{pmatrix} 2 & 8 & -3 \\ 1 & 4 & 1 \\ -5 & 2 & 2 \\ 1 & -3 & 8 \end{pmatrix}$  stands for the linear map  $T$ 

where 
$$T\begin{pmatrix} 1\\ 0\\ 0 \end{pmatrix} = \begin{pmatrix} 2\\ 1\\ -5\\ 1 \end{pmatrix}, T\begin{pmatrix} 0\\ 1\\ 0 \end{pmatrix} = \begin{pmatrix} 8\\ 4\\ 2\\ -3 \end{pmatrix}, T\begin{pmatrix} 0\\ 0\\ 1 \end{pmatrix} = \begin{pmatrix} -3\\ 1\\ 2\\ 8 \end{pmatrix}$$

where we have chosen the unit vectors as bases for  $U = \mathbb{R}^3$  and  $V = \mathbb{R}^4$ . We have that

 $\begin{pmatrix} 2 & 8 & -3 \\ 1 & 4 & 1 \\ -5 & 2 & 2 \\ 1 & -3 & 8 \end{pmatrix}$  has rank 3. This means that ker(*T*) = {0}, the map is injective and

maps the 3 unit

vectors to three linearly independent vectors. The image of the vector  $x \in \mathbb{R}^3$  is a vector in  $\mathbb{R}^4$ :

$$\begin{pmatrix} x \\ y \\ z \end{pmatrix} \mapsto \begin{pmatrix} 2x + 8y - 3z \\ x + 4y + z \\ -5x + 2y + 2z \\ x - 3y + 8z \end{pmatrix} \text{ker}(T) \text{ is the solution space for the homogeneous}$$

system which as we already saw consists only of the zero-vector of  $\mathbb{R}^3$ . The map  $T : \mathbb{R}^3 \to \mathbb{R}^4$  is injective. We can find a linear map  $S : \mathbb{R}^4 \to \mathbb{R}^3$  such that  $S \circ T : \mathbb{R}^3 \to \mathbb{R}^3$  is the identity. Because  $T(e_1), T(e_2), T(e_3)$  are linearly independent, we can find a vector  $\beta_4$  such that  $\{\beta_1 = T(e_1), \beta_2 = T(e_2), \beta_3 = T(e_3), \beta_4\}$  form a basis of  $\mathbb{R}^4$ . We define *S* on this basis by  $\beta_1 \mapsto e_1, \beta_2 \mapsto e_2, \beta_3 \mapsto e_3, \beta_4 \mapsto \alpha$  where  $\alpha$  is any vector in  $\mathbb{R}^3$ . For example  $\alpha = 0 \in \mathbb{R}^3$  is fine. Then  $S(T(e_1)) = S(\beta_1) = e_1$  and similarly for the other unit vectors of  $\mathbb{R}^3$ . That is  $S(T(e_i)) = e_i$ . That is, the composition  $S \circ T$  is the identity on the unit vectors of  $\mathbb{R}^3$  and therefore the identity on  $\mathbb{R}^3$ . Because of  $S \circ T = id_{\mathbb{R}^3}$  we have for general reasons that *S* is surjective and *T* injective. Something we showed for arbitrary maps. We also see that *S* is not uniquely determined by *T*. First  $\beta_4$  is not unique and if we have found some  $\beta_4$  we can assign any vector  $\alpha$  in  $\mathbb{R}^3$  as its image under *S*. The map *S* is unique only on im(S) by assigning to  $T(\alpha)$ , the vector  $\alpha$ .

Example 2. This is a simple example which makes the logic quite transparent. Let

$$T: \mathbb{R}^2 \twoheadrightarrow \mathbb{R}^3, e_1^2 \mapsto e_1^3, e_2^2 \mapsto e_2^3.$$
 The matrix of  $T$  is  $\begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix}$  We add the third unit vector  $e_3^3$  to  $T(e_1^2) = e_1^3, T(e_2^2) = e_2^3$  and define the map  $S$  as  $S(T(e_1^2)) = e_1^2, S(T(e_2^2)) = e_2^3, S(e_3^3) = \alpha = \begin{pmatrix} a \\ b \end{pmatrix}, a, b$  arbitrarily chosen. Then

$$Mat(S; e_1^3, e_2^3, e_3^3; e_1^2, e_2^2) = \begin{pmatrix} 1 & 0 & a \\ 0 & 1 & b \end{pmatrix}$$

and

$$Mat(S \circ T; e_1^2, e_2^2; e_1^2, e_2^2) = \begin{pmatrix} 1 & 0 & a \\ 0 & 1 & b \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 1 \end{pmatrix}$$

This is an example where the product of two non-square matrices is square and invertible. Notice that

 $\begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 & a \\ 0 & 1 & b \end{pmatrix} = \begin{pmatrix} 1 & 0 & a \\ 0 & 1 & b \\ 0 & 0 & 0 \end{pmatrix}$  is also square but not the identity.

Matrices and linear maps can be identified. If *A* is an  $m \times n$ -matrix, then  $L_A : \mathbb{R}^n \twoheadrightarrow \mathbb{R}^m, X \mapsto AX = Y$  is linear and  $Mat(L_A) = A$ . This is because  $Ae_j^n = A_j$  where  $A_j$  is the *j*<sup>th</sup> -column vector of *A*.

## Example 3.

The matrix

 $\begin{pmatrix} 2 & 3 & 1 \\ 1 & 1 & 4 \end{pmatrix}$  stands for a map,*A*, from  $\mathbb{R}^3$  into  $\mathbb{R}^2$ . The map is onto (why?), and

therefore ker(A) + 2 = 3, which gives us a one-dimensional null-space. How can we compute the kernel, that is find a basis? We have for

$$\begin{pmatrix} 2 & 3 & 1 \\ 1 & 1 & 4 \end{pmatrix}$$
 as row echelon form: 
$$\begin{pmatrix} 1 & 0 & 11 \\ 0 & 1 & -7 \end{pmatrix}$$
 This is  $x = -11z, y = 7z$  or ker(A) is the span of the vector 
$$\begin{pmatrix} -11 \\ 7 \\ 1 \end{pmatrix}$$

We have that the matrix of the composition of maps corresponds to the product of the matrices

If 
$$S: U \rightarrow V, T: V \rightarrow W, A = Mat(S; \alpha_1, \alpha_2, \dots, \alpha_n; \beta_1, \beta_2, \dots, \beta_m), B = Mat(T; \beta_1, \beta_2, \dots, \beta_m; \gamma_1, \gamma_2, \dots, \gamma_l)$$

then

$$Mat(T \circ S; \alpha_1, \alpha_2, \dots, \alpha_n; \gamma_1, \gamma_2, \dots, \gamma_l) = BA$$

In particular,  $Mat(L_{BA}) = BA : L_{BA}(X) = (BA)X = B(AX) = L_B(L_A(X))$ . Thus  $L_{BA} = L_B \circ L_A$ and therefore  $Mat(L_{BA}) = Mat(L_B \circ L_A) = Mat(L_B)Mat(L_A) = BA$ 

The linear map  $T : U \to V$  is invertible if there is a linear map  $S : V \to U$  such that  $S \circ T = id_U$  and  $T \circ S = id_V$ . For a linear map  $T : U \to V$  to have an inverse,  $T^{-1}$ , it is necessary that dim  $U = \dim V$ . (Proof?)

If  $A = Mat(T; \alpha_1, \alpha_2, ..., \alpha_n; \beta_1, \beta_2, ..., \beta_n)$  and  $B = Mat(T^{-1}; \beta_1, \beta_2, ..., \beta_n; \alpha_1, \alpha_2, ..., \alpha_n)$  then  $B = A^{-1}$  where  $AA^{-1} = A^{-1}A = I_n$ , the identity  $n \times n$  -matrix.

We have the following important result:

For an  $n \times n$  –matrix the following are equivalent:

Ax = 0 has only the trivial solution;

Ax = y has for every y exactly one solution x

A has an inverse.

All of this follows from the theorem that a linear map on a finite dimensional vector space is injective if an only if it is surjective.

Now, how can we find the inverse of a matrix? While the book postpones this up to a later chapter, see p.100, Example 2, using our current knowledge on linear maps this is actually quite trivial to do. Let us explain this on that example:

$$A = \begin{pmatrix} 5 & 7 \\ 2 & 3 \end{pmatrix} \text{ stands for the linear map } T = L_A : \mathbb{R}^2 \to \mathbb{R}^2 \text{. where}$$

$$\begin{pmatrix} 1 \\ 0 \end{pmatrix} \mapsto \begin{pmatrix} 5 \\ 2 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \end{pmatrix} \mapsto \begin{pmatrix} 7 \\ 3 \end{pmatrix}. \text{ That is } A = Mat(T; e_1, e_2; e_1, e_2) \text{ where the } e_i$$
are the unit vectors in  $\mathbb{R}^2$ . We have
$$T^{-1}\begin{pmatrix} 5 \\ 2 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \text{ and } T^{-1}\begin{pmatrix} 7 \\ 3 \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \end{pmatrix}. \text{ Then obviously,}$$

$$Mat(T^{-1}; Te_1, Te_2; e_1, e_2) = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \text{ and this is not what we want. We want}$$

$$Mat(T^{-1}; e_1, e_2; e_1, e_2) = A^{-1}. \text{ For this we need to find}$$

$$T^{-1}(e_1) = \begin{pmatrix} x_{11} \\ x_{21} \end{pmatrix}, T^{-1}(e_2) = \begin{pmatrix} x_{12} \\ x_{22} \end{pmatrix} \text{ But for this we need to express the unit}$$
vectors as linear combinations of  $\begin{pmatrix} 5 \\ 2 \\ 3 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$  This is the inhomogeneous linear system of

2 equations in 2 unknowns with augmented matrix

$$\begin{pmatrix} 5 & 7 & 1 \\ 2 & 3 & 0 \end{pmatrix}$$
 which has the row echelon form: 
$$\begin{pmatrix} 1 & 0 & 3 \\ 0 & 1 & -2 \end{pmatrix}$$
. Hence:  
$$x = 3, y = -2.$$
 And from:  
$$\begin{pmatrix} 1 \\ 0 \end{pmatrix} = 3\begin{pmatrix} 5 \\ 2 \end{pmatrix} - 2\begin{pmatrix} 7 \\ 3 \end{pmatrix}$$
 we get  
$$T^{-1}\begin{pmatrix} 1 \\ 0 \end{pmatrix} = 3T^{-1}\begin{pmatrix} 5 \\ 2 \end{pmatrix} - 2T^{-1}\begin{pmatrix} 7 \\ 3 \end{pmatrix} = 3\begin{pmatrix} 1 \\ 0 \end{pmatrix} - 2\begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 3 \\ -2 \end{pmatrix}$$

and similarly for the second column of the inverse. Actually we can work out simultaneously both inhomogeneous systems where the right hand sides are the unit vectors

$$\begin{pmatrix} 5 & 7 & 1 & 0 \\ 2 & 3 & 0 & 1 \end{pmatrix}, \text{ row echelon form:} \begin{pmatrix} 1 & 0 & 3 & -7 \\ 0 & 1 & -2 & 5 \end{pmatrix}$$
  
Thus,  $Mat(T^{-1}; e_1, e_2; e_1, e_2) = \begin{pmatrix} 3 & -7 \\ -2 & 5 \end{pmatrix}$ 

Let *A* be an  $n \times n$  –matrix where  $L_A$  has an inverse. Then *A* has an inverse  $A^{-1}$ . Then if  $A_j$  is the  $j^{th}$  –column of *A* then  $L_A(e_j) = A_j$ . Then if

$$x_{1j}A_1 + x_{2j}A_2 + \ldots + x_{nj}A_n = e_j$$
, where  $e_j$  is the  $j$  - th unit vector

then applying  $L_A^{-1}$  to this equation gives:

$$x_{1j}e_1 + x_{2j}e_2 + \dots + x_{nj}e_n = L_A^{-1}(e_j)$$

This is  $L_A^{-1}(e_j) = \begin{pmatrix} x_{1j} \\ x_{2j} \\ \vdots \\ x_{nj} \end{pmatrix} = X_j$  and the matrix with columns  $X_j$  is the matrix of  $L_A^{-1}$ 

and the inverse of A.

Let  $(A | I_n)$  be the matrix A augmented by the n-columns of unit vectors  $e_1, e_2, \ldots, e_n$ . Then using the elementary row operations transforms A into  $I_n$  and  $I_n$  into  $A^{-1}$ .

$$(A \mid I_n) \stackrel{\text{elementary row operations}}{\Rightarrow} (I_n \mid A^{-1})$$

Actually,  $AA^{-1} = I_n$ . This tells us that  $L_{A^{-1}}$  is injective. But then it must be also surjective, that is  $AA^{-1}$  also.