## Notes 15 – Linear Mappings and Matrices

In this lecture, we turn attention to linear mappings that may be neither surjective nor injective. We show that once bases have been chosen, a linear map is completely determined by a matrix. This approach provides the first real justification for the definition of matrix multiplication that we gave in the first lecture.

**L15.1 The linear mapping associated to a matrix.** First, we point out that any matrix defines a linear mapping.

**Lemma.** A matrix  $A \in \mathbb{R}^{m,n}$  defines a linear mapping  $f: \mathbb{R}^n \to \mathbb{R}^m$  by regarding elements of  $\mathbb{R}^n$  as column vectors and setting

$$f(\mathbf{v}) = A\mathbf{v}, \qquad \mathbf{v} \in \mathbb{R}^{n,1}.$$

*Proof.* The conditions (LM1) and (LM2) are obvious consequences of the rules of matrix algebra. QED

Our preference for column vectors means that an  $m \times n$  matrix defines a mapping from  $\mathbb{R}^n$  to  $\mathbb{R}^m$ , so that m, n are 'swapped over'. Here is an example:

$$A = \begin{pmatrix} 0 & 2 & 4 \\ 3 & 5 & 1 \end{pmatrix} \in \mathbb{R}^{2,3} \quad \text{defines} \quad f \colon \mathbb{R}^3 \to \mathbb{R}^2$$

with

$$f\begin{pmatrix}1\\0\\0\end{pmatrix} = \begin{pmatrix}0\\3\end{pmatrix}, \quad f\begin{pmatrix}0\\1\\0\end{pmatrix} = \begin{pmatrix}2\\5\end{pmatrix}, \quad f\begin{pmatrix}0\\0\\1\end{pmatrix} = \begin{pmatrix}4\\1\end{pmatrix}.$$

The *j*th column of A represents the image of the *j*th element of the canonical basis of  $\mathbb{R}^{n,1}$ .

This example shows that at times we can use *A* and *f* interchangeably. But there is a subtle difference: in applying *f*, we are allowed to represent elements of  $\mathbb{R}^n$  and  $\mathbb{R}^m$  by row vectors. Thus it is also legitimate to write

$$f(1,0,0) = (0,3), \quad f(0,1,0) = (2,5), \quad f(0,0,1) = (4,1),$$

and more memorably,

$$f(x_1, x_2, x_3) = (2x_2 + 4x_3, 3x_1 + 5x_2 + x_3).$$
(1)

The last equation shows us how to pass from the *rows* of *A* to the definition of *f*. It turns out that any linear mapping  $f: \mathbb{R}^n \to \mathbb{R}^m$  has a form analogous to (1), from which we can construct the rows of an associated matrix.

## L15.2 The matrix associated to a linear mapping. Let *V*, *W* be vector spaces.

**Lemma.** Let *V* be a vector space with basis  $\{\mathbf{v}_1, ..., \mathbf{v}_n\}$ . A linear mapping  $f: V \to W$  is completely determined by vectors  $f(\mathbf{v}_1), ..., f(\mathbf{v}_n)$ , which (in order to define f) can be assigned arbitrarily.

*Proof.* Suppose that  $\{\mathbf{v}_1, ..., \mathbf{v}_n\}$  is a basis of *V*. Then any element  $\mathbf{v}$  of *V* can be written in a *unique* way as  $a_1\mathbf{v}_1 + \cdots + a_n\mathbf{v}_n$ , and

$$f(\mathbf{v}) = f(a_1\mathbf{v}_1 + \dots + a_n\mathbf{v}_n) = a_1f(\mathbf{v}_1) + \dots + a_nf(\mathbf{v}_n).$$
(2)

Thus,  $f(\mathbf{v})$  is determined by the *n* images  $f(\mathbf{v}_i)$  of the basis elements. Moreover, *any* choice of *n* such vectors allows us to define a linear mapping *f* by means of (2). QED

We have seen that a matrix determines a linear mapping between vector spaces, namely from  $\mathbb{R}^n$  to  $\mathbb{R}^m$ . The Lemma allows us to go backwards and associate a matrix to *any* linear mapping from *V* to *W*, once we have chosen bases

$$\{\mathbf{v}_1,\cdots,\mathbf{v}_n\},\qquad \{\mathbf{w}_1,\cdots,\mathbf{w}_m\}\tag{3}$$

of *V* and *W* (we are assuming that dim V = n and dim W = m).

**Definition.** Let  $g: V \to W$  be a linear mapping. The matrix of g with respect to the bases (3) is the matrix  $B \in \mathbb{R}^{m,n}$  (also written  $M_g$  or M(g) if the choice (3) is understood) whose j th column gives the coefficients of  $g(\mathbf{v}_j)$  in terms of  $\mathbf{w}_1, \ldots, \mathbf{w}_m$ .

If  $B = (b_{ij})$ , then we are asserting that

$$g(\mathbf{v}_j) = b_{1j}\mathbf{w}_1 + \dots + b_{mj}\mathbf{w}_m = \sum_{i=1}^m b_{ij}\mathbf{w}_i, \quad \text{for each } j = 1, \dots, n.$$
(4)

By the previous lemma, g is completely determined by B.  $\mathfrak{Marning}$ : In this definition, we are really thinking of the bases (3) as *ordered* sets.

To recover our previous correspondence, we need to take  $V = \mathbb{R}^n$  and  $W = \mathbb{R}^m$  with their *canonical bases.* For let  $A \in \mathbb{R}^{m,n}$ . If  $f: \mathbb{R}^n \to \mathbb{R}^m$  is the linear mapping defined by setting  $f(\mathbf{v}) = A\mathbf{v}$  and we choose (3) to be the canonical bases, then  $B = M_f = A$ .

**Example.** We are perfectly at liberty to apply the Definition to the same vector space with different bases. Let  $V = W = \mathbb{R}_2[x]$ . Choose the basis  $\{1, x+1, (x+1)^2\}$  for V and the basis  $\{1, x, x^2\}$  for W. Let D be the linear mapping defined by differentiation:  $D\mathbf{p} = \mathbf{p}'$ . Then the matrix of D with respect to the chosen bases is

$$A = \begin{pmatrix} 0 & 1 & 2 \\ 0 & 0 & 2 \\ 0 & 0 & 0 \end{pmatrix};$$

the null column tells us that D(1) = 0 and the null row tells us that Im D consists of polynomials of degree no greater than 1.

**L15.3 Compositions and products.** With the link between linear mappings and matrices now established, we shall see that composition of matrices corresponds to the product of matrices. Suppose that  $B \in \mathbb{R}^{m,n}$ ,  $A \in \mathbb{R}^{n,p}$ , and consider the associated linear mappings

$$\mathbb{R}^{m,1} \stackrel{g}{\leftarrow} \mathbb{R}^{n,1} \stackrel{f}{\leftarrow} \mathbb{R}^{p,1}$$

defined by  $f(\mathbf{u}) = A\mathbf{u}$  and  $g(\mathbf{v}) = B\mathbf{v}$ . (It is easier to understand what follows by writing the mappings from right to left.) The composition  $g \circ f$  is obviously

$$BA\mathbf{u} \leftarrow A\mathbf{u} \leftarrow \mathbf{u}_{\mu}$$

and is therefore associated to the matrix *BA*.

More generally, given vector spaces *U*, *V*, *W* and linear mappings

$$W \stackrel{g}{\leftarrow} V \stackrel{f}{\leftarrow} U, \tag{5}$$

choose bases for each of U, V, W, and let  $M_f, M_g$  be the associated matrices (the same basis of V being used for both matrices). Then we state without proof the

**Proposition.** Let  $h = g \circ f$  be the composition (5). Then  $M_h = M_g M_f$ .

This result is especially useful in the case of a single vector space *V* of dimension *n*, and a linear mapping  $f: V \rightarrow V$ . Such a linear mapping (between the same vector space) is called a *linear transformation* or *endomorphism*. In these circumstances, we can fix the same basis  $\{\mathbf{v}_1, \ldots, \mathbf{v}_n\}$  of *V*, and consider compositions of *f* with itself:

**Example.** Define  $f : \mathbb{R}^3 \to \mathbb{R}^3$  by  $f(\mathbf{e}_1) = \mathbf{e}_2$ ,  $f(\mathbf{e}_2) = \mathbf{e}_3$  and  $f(\mathbf{e}_3) = \mathbf{e}_1$ . Check that the matrix  $A = M_f$  (taken with respect to the canonical basis  $\{\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3\}$ ) satisfies  $A^3 = I_3$ .

## L15.4 Nullity and rank. Important examples of subspaces are provided by the

**Lemma.** Let  $g: V \rightarrow W$  be a linear mapping. Then

(*i*) *g*<sup>-1</sup>(**0**) *is a subspace of V ,*(*ii*) Im *g is a subspace of W .*

*Proof.* We shall only prove (ii). If  $\mathbf{w}_1, \mathbf{w}_2 \in \text{Im } g$  then we may write  $\mathbf{w}_1 = g(\mathbf{v}_1)$  and  $\mathbf{w}_2 = g(\mathbf{v}_2)$  for some  $\mathbf{v}_1, \mathbf{v}_2 \in V$ . If  $a \in F$  then

$$a\mathbf{w}_1 + \mathbf{w}_2 = ag(\mathbf{v}_1) + g(\mathbf{v}_2) = g(a\mathbf{v}_1 + \mathbf{v}_2),$$

and so  $a\mathbf{w}_1 + \mathbf{w}_2 \in \text{Im } g$ . Part (i) is similar.

**Example.** In the case of a linear mapping  $f: \mathbb{R}^n \to \mathbb{R}^m$  defined by  $f(\mathbf{v}) = A\mathbf{v}$  with  $A \in \mathbb{R}^{m,n}$ ,

$$f^{-1}(\mathbf{0}) = \{ \mathbf{v} \in \mathbb{R}^{n,1} : A\mathbf{v} = \mathbf{0} \}$$

is the solution space of the homogeneous linear system  $A\mathbf{x} = \mathbf{0}$ . We already know that this is a subspace and we labelled it Ker *A*. On the other hand, the image of *f* is generated by the vectors  $f(\mathbf{e}_i)$  that are the columns of *A*:

Im 
$$f = \mathscr{L} \{ f(\mathbf{e}_1), \dots, f(\mathbf{e}_n) \} = \mathscr{L} \{ \mathbf{c}_1, \dots, \mathbf{c}_n \} = \operatorname{Col} A.$$

It follows that the dimension of Im f equals the rank, r(A), the common dimension of Row A and Col A.

In view of this key example, we state the

**Definition.** (*i*) The kernel of an arbitrary linear mapping  $g: V \to W$  is the subspace  $g^{-1}(\mathbf{0})$ , more usually written Ker g or ker g. Its dimension is called the nullity of g. (*ii*) The dimension of Im g is called the rank of g.

**Rank-Nullity Theorem.** *Given an arbitrary linear mapping*  $g: V \rightarrow W$ *,* 

 $\dim V = \dim(\operatorname{Ker} g) + \dim(\operatorname{Im} g).$ 

This important result can also be expressed in the form

$$\operatorname{nullity}(g) + \operatorname{rank}(g) = n,$$

*n* being the dimension of the space of 'departure'.

OED

*Proof.* By choosing bases for *V* and *W*, we may effectively replace *g* by a linear mapping  $f: \mathbb{R}^n \to \mathbb{R}^m$ . But in this case, the previous example shows that Ker *f* = Ker *A* and Im *f* = Col *A*. We know that dim(Col *A*) = dim(Row *A*) = *r*(*A*), and (essentially by (RC2)),

$$\dim(\operatorname{Ker} A) = n - r(A).$$

The result follows.

The following result is for class discussion:

**Corollary.** Given a linear mapping  $g: V \to W$  with dim V = n and dim W = m,

 $g \text{ is injective } \Leftrightarrow \operatorname{rank}(g) = n$   $g \text{ is surjective } \Leftrightarrow \operatorname{rank}(g) = m$  $g \text{ is bijective } \Leftrightarrow \operatorname{rank}(g) = m = n.$ 

## L15.5 Further exercises.

1. Let  $f : \mathbb{R}^4 \to \mathbb{R}^3$  be the linear mapping with

$$f(x_1, x_2, x_3, x_4) = (x_1, x_1 + x_2, x_1 + x_2 + x_3).$$

Write down that associated matrix *A*, and find  $\mathbf{v} \in \mathbb{R}^4$  such that  $f(\mathbf{v}) = (0, 1, 1)$ .

2. Let  $f : \mathbb{R}^3 \to \mathbb{R}^2$  and  $g : \mathbb{R}^2 \to \mathbb{R}^3$  be the two mappings:

$$f(x, y, z) = (x + y, x + y - z), \qquad g(s, t) = (3s - t, 3t - s, s).$$

(i) Complete: g(f(x, y, z)) = (2x+2y+z, ?)

(ii) Find the matrices  $M_f$ ,  $M_g$ ,  $M_{g\circ f}$  with respect to the canonical bases of  $\mathbb{R}^2$  and  $\mathbb{R}^3$ , and check that  $M_{g\circ f} = M_g M_f$ .

3. Let  $V = W = \mathbb{R}_2[x]$ , and let *D* be the linear mapping given by  $D(\mathbf{p}(x)) = \mathbf{p}'(x)$ . Find the matrix  $M_D$  with respect to the bases:  $\{1, x, x^2\}$  for *V* and  $\{1, x+1, (x+1)^2\}$  for *W*.

4. Let  $f : \mathbb{R}^3 \to \mathbb{R}^3$  be a linear transformation such that

Ker 
$$f = \mathscr{L}\{(1,0,0), (0,1,0)\}$$
 and Im  $f \subseteq \mathscr{L}\{(0,1,0), (0,0,1)\}$ 

Find all possible matrices  $M_f$  associated to f with respect to the canonical basis of  $\mathbb{R}^3$ .

5. Find bases for the kernel and the image of each of the following linear mappings:

$$f: \mathbb{R}^3 \to \mathbb{R}^{2,2}, \quad f(x, y, z) = \begin{pmatrix} x & -y \\ y & x \end{pmatrix},$$
$$g: \mathbb{R}^{2,2} \to \mathbb{R}^4, \quad g\begin{pmatrix} x & y \\ z & t \end{pmatrix} = (x - 2y, x - 2z, y + t, x + 2t).$$

6. Let  $f : \mathbb{R}^{3,1} \to \mathbb{R}^{3,1}$  be the linear transformation defined by  $A = \begin{pmatrix} 1 & 2 & -1 \\ 1 & 1 & 3 \\ 3 & 5 & 1 \end{pmatrix}$ . (i) Find a vector  $\mathbf{v}_1$  such that Ker  $f = \mathscr{L} \{ \mathbf{v}_1 \}$ .

(ii) *Choose*  $\mathbf{v}_2, \mathbf{v}_3$  such that  $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$  is a basis of  $\mathbb{R}^{3,1}$ .

(iii) Check that  $\{f(\mathbf{v}_2), f(\mathbf{v}_3)\}$  is a basis of the subspace Im f (it always will be!).

QED