# Linear Algebra & Geometry LECTURE 13

- Linear mappings
- Nullity and rank
- Matrices of linear mappings

# **Linear Mappings**

**Definition**. (reminder)

Let *V* and *W* be vector spaces over a field  $\mathbb{F}$ . A function  $\varphi: V \to W$  is called a *linear mapping* iff (a)  $(\forall u, v \in V) \varphi(u + v) = \varphi(u) + \varphi(v)$  (additivity),

(b)  $(\forall v \in V) (\forall p \in \mathbb{F}) \varphi(pv) = p\varphi(v)$  (homogeneity).

## Example.

For every  $n \times k$  matrix A the function  $f_A: \mathbb{F}^{k \times p} \to \mathbb{F}^{n \times p}$  defined as  $f_A(X) = AX$  is a linear mapping.

Additivity follows from distributivity of matrix multiplication over addition.

Homogeneity is obvious: for every *i* and *j*,  $f_A(\lambda X)(i, j) = A(\lambda X)(i, j) = \sum_{s=1}^n A(i, s)\lambda X(s, j) = \lambda \sum_{s=1}^n A(i, s)X(s, j) = \lambda (AX(i, j)) = \lambda f_A(X)(i, j).$ 

#### **Rank and nullity**

**Definition**. Let  $\varphi: V \to W$  be a linear mapping. The *image* of  $\varphi$  is the set im  $\varphi = \varphi(V)$  and the *kernel* of  $\varphi$  is the set  $\ker \varphi = \{v \in V : \varphi(v) = \Theta_W\}$ 

**Proposition.** im $\varphi$  is a subspace of W and ker $\varphi$  is a subspace of V. **Proof.** Let  $w_1, w_2 \in im\varphi$ . There exist  $v_1, v_2 \in V$  such that  $\varphi(v_i) = w_i, i = 1, 2$ . Then  $pw_1 + qw_2 = p\varphi(v_1) + q\varphi(v_2) = \varphi(pv_1 + qv_2) \in im\varphi$ , hence  $im\varphi$  is a subspace of W. For every  $v_1, v_2 \in ker\varphi$ ,  $\varphi(av_1 + bv_2) = a\varphi(v_1) + b\varphi(v_2) =$ 

 $a\Theta + b\Theta = \Theta$  hence, ker $\varphi$  is a subspace of V. QED

**Definition**.  $rank(\varphi) = \dim(im\varphi), nullity(\varphi) = \dim(ker\varphi)$ .

**Example**. Let  $V = W = \mathbb{R}_n[x]$  and  $\varphi(f(x)) = f'(x)$ . Then ker $\varphi = \mathbb{R}_0[x]$ , the space of all constant polynomials and  $\operatorname{im} \varphi = \mathbb{R}_{n-1}[x]$ . Hence,  $\operatorname{rank}(\varphi) = n$  and  $\operatorname{nullity}(\varphi) = 1$ . **Proposition**. For every set  $\{v_1, v_2, ..., v_n\} \subseteq V$  and for every linear mapping  $\varphi: V \to W$ ,  $\varphi(span(v_1, v_2, ..., v_n)) = span(\varphi(v_1), \varphi(v_2), ..., \varphi(v_n))$ . **Proof**. (Obvious)

**Example**. For every  $n \times k$  matrix A, f(X) = AX is a linear mapping,  $f: \mathbb{F}^k \to \mathbb{F}^n$ . ker f is the solution set (space, really) for  $AX = \Theta$  hence, *nullity* of f is the dimension of the solution space. On the other hand, denoting by  $e_1, e_2, \dots, e_k$  the unit vectors of  $\mathbb{F}^k$  we obtain that im f is spanned by  $\{f(e_1), f(e_2), \dots, f(e_k)\} = \{Ae_1, \dots, Ae_k\}$  i.e., by the columns of A. Hence, the rank of f is the number of linearly independent columns of A which is the same as rank of A.

#### Theorem.

For every linear mapping  $\varphi: V \to W$ 

 $rank(\varphi) + nullity(\varphi) = dimV$ 

**Proof.** Let  $\{v_1, v_2, ..., v_n\}$  be a basis for ker $\varphi$ . There exist vectors  $w_1, w_2, ..., w_k$  such that  $\{v_1, v_2, ..., v_n, w_1, w_2, ..., w_k\}$  is a basis for *V*. It is enough to show that  $S = \{\varphi(w_1), \varphi(w_2), ..., \varphi(w_k)\}$  is a basis for im  $\varphi$ . Since  $span(S) = im \varphi$ , it is enough to show that *S* is linearly independent. Let  $\sum_{i=1}^{k} a_i \varphi(w_i) = \Theta_W$ . By linearity of  $\varphi$ ,  $\Theta_W = \sum_{i=1}^{k} a_i \varphi(w_i) = \varphi(\sum_{i=1}^{k} a_i w_i)$  i.e.,  $\sum_{i=1}^{k} a_i w_i \in \text{ker}\varphi$  so, for some scalars  $b_1, b_2, ..., b_n$  we have  $\sum_{i=1}^{k} a_i w_i = \sum_{j=1}^{n} b_j v_j$ . This implies  $\sum_{i=1}^{k} a_i w_i - \sum_{j=1}^{n} b_j v_j = \Theta_V$  hence, all  $a_i$  and  $b_j$  are zeroes. QED

# Corollary.

If A is an  $n \times k$  matrix then the dimension of the solution space of the homogeneous system of equations  $AX = \Theta$  is k - rank(A).

### Note.

Every *n*-dimensional vector space V over a field  $\mathbb{F}$  is *isomorphic* with  $\mathbb{F}^n$ .

**Proof.** Let  $R = \{v_1, v_2, ..., v_n\}$  be a basis of *V*. The function  $\Phi(x) = (a_1, a_2, ..., a_n)$ , where  $a_1, a_2, ..., a_n$  are unique scalars such that  $x = \sum_{s=1}^n a_s v_s$  maps *V* into  $\mathbb{F}^n$  is an isomorphism. QED

The vector  $(a_1, a_2, ..., a_n)$  is called the *coordinate vector* of x with respect to R and is often denoted by  $[x]_R$ . From now on we will use  $\mathbb{F}^n$ , rather than general symbol V of a vector space. **Definition**. Let  $\varphi: \mathbb{F}^k \to \mathbb{F}^n$  be a linear mapping and let  $R = \{v_1, v_2, \dots, v_k\}$  and  $S = \{w_1, w_2, \dots, w_n\}$  be bases for  $\mathbb{F}^k$  and  $\mathbb{F}^n$ , respectively. For each  $v_i$  there exist unique scalars  $a_{1,i}, a_{2,i}, \dots$ ,  $a_{n,i}$  such that  $\varphi(v_i) = a_{1,i}w_1 + a_{2,i}w_2 + \dots + a_{n,i}w_n = \sum_{s=1}^n a_{s,i}w_s$ . The  $k \times n$  matrix  $M_S^R(\varphi) = [a_{i,j}]$  is called the matrix of  $\varphi$  in bases R and S.

**Remark**. In other words, the *i*-th column of the matrix  $M_S^R(\varphi)$  is equal to  $[\varphi(v_i)]_S$ , the coordinate vector of  $\varphi(v_i)$  in the basis *S*. **Remark**. The matrix of a linear mapping  $\varphi$  depends on the choice of the bases *R* and *S*, but the size of the matrix depends only on the dimensions of the domain and the range of  $\varphi$ .

It may happen that two different linear mappings may have the same matrix but with respect to two pairs of bases.

#### Example.

Let  $\varphi : \mathbb{R}^3 \to \mathbb{R}^2$ ,  $\varphi(x, y, z) = (x + 2y - 2z, 3x - y + 2z)$ . Find the matrix for  $\varphi$  in bases  $R = \{(0, 1, 1), (1, 0, 1), (1, 1, 0)\}$  and  $S = \{(1, 1), (1, 0)\}$ .

We must calculate the values of  $\varphi$  on vectors from R and represent them as linear combinations of vectors from S.

$$\begin{aligned} \varphi(0,1,1) &= (0,1) = 1(1,1) + (-1)(1,0), \\ \varphi(1,0,1) &= (-1,5) = 5(1,1) + (-6)(1,0), \\ \varphi(1,1,0) &= (3,2) = 2(1,1) + 1(1,0). \end{aligned}$$

Finally, we form the matrix placing the coefficients of the linear combinations in consecutive columns,  $M_S^R(\varphi) = \begin{bmatrix} 1 & 5 & 2 \\ -1 & -6 & 1 \end{bmatrix}$ .

#### Theorem.

Let  $\varphi \colon \mathbb{F}^k \to \mathbb{F}^n$  be a linear mapping, let  $R = \{v_1, v_2, \dots, v_k\}, S = \{w_1, w_2, \dots, w_n\}$  be bases for  $\mathbb{F}^k$  and  $\mathbb{F}^n$ , respectively, and let A be an  $n \times k$  matrix. Then:

 $\left( \left( \forall x \in \mathbb{F}^k \right) [\varphi(x)]_S = A[x]_R \right) \Leftrightarrow A = M_S^R(\varphi).$  **Proof.** ( $\Leftrightarrow$ ) Let  $M_S^R(\varphi) = A = [a_{i,j}]$  and  $[x]_R = [x_1, x_2, \dots, x_k]$  i.e.,  $x = \sum_{i=1}^k x_i v_i$ . Then,  $\varphi(x) = \varphi\left(\sum_{i=1}^k x_i v_i\right) = \sum_{i=1}^k x_i \varphi(v_i) = \sum_{i=1}^k x_i \sum_{s=1}^n a_{s,i} w_s = \sum_{i=1}^k \sum_{s=1}^n x_i (a_{s,i} w_s) = \sum_{s=1}^n (\sum_{i=1}^k x_i a_{s,i}) w_s$ (comprehension: why can we change the order of summation?) which means  $[\varphi(x)]_S = (\sum_{i=1}^k x_i a_{1,i}, \sum_{i=1}^k x_i a_{2,i}, \dots, \sum_{i=1}^k x_i a_{k,i}). M_S^R(\varphi)[x]_R = A [x]_R = (\sum_{i=1}^k a_{1,i} x_i, \sum_{i=1}^k a_{2,i} x_i, \dots, \sum_{i=1}^k a_{k,i} x_i),$ 

(except that the vectors  $[\varphi(x)]_S$  and  $M_S^R(\varphi)[x]_R$  should be written as columns). ( $\Rightarrow$ ) Replacing x with  $v_i$  one gets *i*-th column of  $M_S^R(\varphi)$ . QED

# **Fact.** All linear mappings are functions of the form $\varphi(X) = AX$ .

#### Theorem.

Let  $\varphi \colon \mathbb{F}^k \to \mathbb{F}^n$  and  $\psi \colon \mathbb{F}^n \to \mathbb{F}^p$  be linear maps and let R, Sand T be bases for  $\mathbb{F}^k, \mathbb{F}^n$  and  $\mathbb{F}^p$ , respectively. Then  $M_T^R(\psi \circ \varphi) = M_T^S(\psi) M_S^R(\varphi).$ 

#### **Proof.**

For every 
$$x \in \mathbb{F}^k$$
,  $M_T^R(\psi \circ \varphi)[x]_R = [(\psi \circ \varphi)(x)]_T = [\psi(\varphi(x))]_T = M_T^S(\psi)[\varphi(x)]_S = M_T^S(\psi)(M_S^R(\varphi)[x]_R) = (M_T^S(\psi)M_S^R(\varphi))[x]_R$ .

From the *if* part of the previous theorem we get  $M_T^R(\psi \circ \varphi) = M_T^S(\psi)M_S^R(\varphi)$ . QED **Corollary.** 

Matrix multiplication is associative.

#### **Linear operators**

From now on, we will study linear maps which map a vector space into itself. They are called *linear operators*.

If  $\varphi$  is a linear operator and *R* is a basis then  $M_R^R(\varphi)$  is called the matrix of  $\varphi$  in (with resp. to) *R* and is denoted by  $M_R(\varphi)$ .

#### Example.

Consider  $\varphi \colon \mathbb{R}^2 \to \mathbb{R}^2$ ,  $\varphi(x, y) = (x + y, x - y)$ . Find  $M_S(\varphi)$  and  $M_R(\varphi)$ , where  $S = \{(1,0), (0,1)\}$  and  $R = \{(1,1), (2,1)\}$ . To find  $M_R(\varphi)$ :  $\varphi(1,1) = (2,0) = a(1,1) + b(2,1)$ . Solving the system of equations  $\begin{cases} a+2b=2\\ a+b=0 \end{cases}$  we get a = -2, b = 2. $\varphi(2,1) = (3,1) = c(1,1) + d(2,1)$ . Solving the system of equations  $\begin{cases} c+2d=3\\ c+d=1 \end{cases}$  we get c = -1, d = 2. Hence,  $M_R(\varphi) =$  $\begin{bmatrix} -2 & -1 \\ 2 & 2 \end{bmatrix}$ . Obviously,  $M_S(\varphi) = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$  (because  $\varphi(1,0) = (1,1) =$ 1(1,0) + 1(0,1) and  $\varphi(0,1) = (1,-1) = 1(1,0) + (-1)(0,1)$ .