

## Linear Maps

### Definition

**Examples:** Identity, differentiation, integration, multiplication by  $x^2$ , backward shift

**Algebra:** Vector space with an associative product of vectors. Moreover  $X(aY) = a(XY)$ .

**The algebra of linear operators from  $V$  to  $V$ :** multiplication given by function composition. Notation:  $\mathcal{L}(V)$ .

**Nullspace and range of a linear operator.** These are subspaces of domain and codomain. Dimensions are called rank and nullity.

**Theorem:** A linear operator is injective iff its nullspace is trivial.

**Rank-Nullity Theorem:** Given a linear operator  $T : V \rightarrow W$ ,

$$\dim V = \text{rank} + \text{nullity}.$$

**Proof:** Find a basis  $\{n_1, \dots, n_a\}$  for the nullspace and a basis  $\{y_1, \dots, y_b\}$  for the range. Pull the range basis back to the codomain, yielding  $\{x_1, \dots, x_b\}$  where  $Tx_i = y_i$ . Verify that a basis for  $V$  is  $\{n_1, \dots, n_a, x_1, \dots, x_b\}$ : Let  $v \in V$ . Then  $Tv = \sum_i a_i y_i$ , therefore  $T \sum_i a_i x_i = Tv$ , therefore  $\sum_i a_i x_i - v \in$  nullspace, therefore  $\sum_i a_i x_i - v = \sum_i b_i n_i$ , hence  $v$  is spanned by these vectors. Moreover  $\sum_i a_i n_i + \sum_i b_i x_i = 0$  implies  $\sum_i b_i y_i = 0$ , hence each  $b_i = 0$ , hence each  $a_i = 0$ .

**Corollary:** A linear map is injective if and only if the dimension of its domain is equal to the dimension of its range. A linear map is surjective if and only if the dimension of its range is equal to the dimension of its codomain. A linear map is bijective if and only if domain, range, and codomain have the same dimension. For a linear map  $T : V \rightarrow V$ , if  $T$  is injective then  $\dim V = \text{rank}$ , hence domain, range, and codomain have the same dimension, hence  $T$  is bijective. The same is true of a surjective map  $T : V \rightarrow V$ . Hence injective, surjective, and bijective are all equivalent for  $T \in \mathcal{L}(V, V)$ . Note that the inverse of a bijective linear map is also linear.

**Isomorphic Vector Spaces:** A pair of vector spaces with a bijective linear map from one to the next. Two vector spaces are isomorphic if and only if they have the same dimension. To construct an isomorphism, map a basis to another basis.

**Counting Solutions to a linear system of equations:** If  $A$  is an  $m \times n$  matrix and  $x$  is an  $n \times 1$  column vector, then  $Ax$  is the  $m \times 1$  column vector consisting of the linear combination of the columns of  $A$  using the entries of  $x$ . Hence any linear system of equations can be expressed in the form  $Ax = b$ . This gives rise to a linear map  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  via  $Tx = Ax$ . The range of  $T$  is the span of the columns of  $A$ , hence  $\text{rank} \leq n$ . So if  $m > n$  the map is not surjective. In other words, too many equations implies some solutions to  $Ax = b$  don't exist. Also, if  $m < n$  then the map can't be injective, so there must be some non-trivial solution to  $Ax = 0$ . (Or: there are more column vectors than the dimension of the space they live in, so they are linearly dependent.) Solutions to  $Ax = b$  always exist and are unique (both) if and only if  $m = n$ .

**The matrix of a linear map:** Let  $T : V \rightarrow W$  be linear with  $V$ -basis  $\{v_1, \dots, v_m\}$  and  $W$ -basis  $\{w_1, \dots, w_n\}$ . Suppose that for each  $v_j$ ,

$$Tv_j = \sum_i a_{ij}w_i.$$

We say that  $m(T) = (a_{ij})$ . For a vector  $v = \sum_i x_i v_i \in V$  we write

$$m(v) = \begin{bmatrix} x_1 \\ \vdots \\ x_m \end{bmatrix}.$$

For a vector  $w = \sum_i y_i w_i \in W$  we write

$$m(w) = \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix}.$$

With these conventions we have

$$m(Tv) = m(T)m(v).$$

**Proof:** Write  $m(v) = \begin{bmatrix} x_1 \\ \vdots \\ x_m \end{bmatrix}$  and  $m(Tv) = \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix}$ . Then

$$Tv = T\left(\sum_j x_j v_j\right) = \sum_j x_j Tv_j = \sum_j x_j \left(\sum_i a_{ij} w_i\right) = \sum_i \left(\sum_j a_{ij} x_j\right) w_i,$$

which implies

$$y_i = \sum_j a_{ij}x_j$$

for each  $i$ . In other words,

$$m(Tv) = \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} = (a_{ij}) \begin{bmatrix} x_1 \\ \vdots \\ x_m \end{bmatrix} = m(T)m(v).$$

**The matrix of a composition of linear maps:** Let  $T : U \rightarrow V$  and  $S : V \rightarrow W$  be linear maps with  $U$ -basis  $\{u_i\}$ ,  $V$ -basis  $\{v_i\}$ , and  $W$ -basis  $\{w_i\}$ . Then  $ST : U \rightarrow W$  satisfies  $m(ST) = m(S)m(T)$ .

**Proof:**  $m(STu) = m(S)m(Tu) = m(S)m(T)m(u)$  for all  $u$ . In particular,  $m(STu_i) = m(S)m(T)m(u_i) = m(S)m(T)e_i = \text{column } i \text{ of } m(S)m(T)$ . This implies that  $m(ST) = m(S)m(T)$ .

The dimension of  $\mathcal{L}(V, W)$ : There is an isomorphism between  $\mathcal{L}(V, W)$  and the appropriate size matrices which represent these linear maps. This yields

$$\dim \mathcal{L}(V, W) = \dim V \cdot \dim W.$$