

1 Change of Basis and Conjugation

Definition: Linear Endomorphisms

For the rest of the class, we will focus on linear maps from a vector space to itself

$$T : \mathbf{V} \rightarrow \mathbf{V}$$

These are sometimes called **linear endomorphisms** of \mathbf{V} .

Note: Notational Shortcuts

- \mathbf{F} denotes \mathbf{R} or \mathbf{C}
- \mathbf{V} and \mathbf{W} denote vector spaces over \mathbf{F}

1.1 The Vector Space of Linear Maps

Definition: Linear Map

A **linear map** from V to W is a function $T : V \rightarrow W$ with the following properties:

- **Additivity:** $T(u + v) = Tu + Tv$ for all $u, v \in V$;
- **Homogeneity:** $T(\lambda v) = \lambda Tv$ for all $\lambda \in \mathbf{F}$ and $v \in V$.

Some mathematicians use the term **linear transformation** or **vector space homomorphism**, which means the same thing as linear map.

Definition: $\mathcal{L}(V, W)$

The set of all linear maps from \mathbf{V} to \mathbf{W} is denoted $\mathcal{L}(\mathbf{V}, \mathbf{W})$.

It is easy to verify that each of the functions defined below is indeed a linear map.

Example: Linear Maps

Zero

In addition to other uses, the symbol 0 can be used to denote the function that takes each element of the vector space to zero. To be specific, $0 \in \mathcal{L}(V, W)$ is defined by

$$0v = 0$$

Identity

The **identity map** denoted I can be used to denote the function that takes each element of the vector space to itself. To be specific, $I \in \mathcal{L}(V, V)$ is defined by

$$Iv = v$$

Differentiation

Let $f, g \in \mathbf{R}^{\mathbf{R}}$ be once-differentiable functions in \mathbf{R} . Notice

$$\frac{d}{dx}(f + g)(x) = \frac{d}{dx}(f(x) + g(x)) = \frac{d}{dx}f(x) + \frac{d}{dx}g(x) = \left(\frac{d}{dx}f + \frac{d}{dx}g\right)(x).$$

$$\frac{d}{dx}(\lambda f)(x) = \frac{d}{dx}(\lambda f(x)) = \lambda \frac{d}{dx}f(x)$$

Integration

Let $f, g \in \mathbf{R}^{\mathbf{R}}$ be once-integrable functions over $(a, b) \subseteq \mathbf{R}$. Notice

$$\int_a^b (f + g)(x)dx = \int_a^b (f(x) + g(x))dx = \int_a^b f(x)dx + \int_a^b g(x)dx$$

$$\int_a^b (\lambda f)(x)dx = \int_a^b (\lambda f(x))dx = \lambda \int_a^b f(x)dx$$

The previous two assertions, that differentiation and integration are linear, are other ways of stating the basic results about differentiation and integration: the derivative or integral of a sum is the sum of the derivatives or integrals; the derivative or integral of a constant times a function is the constant multiple of the derivative or integral of the function.

Example: Linear Maps

Multiplication by x^2

Define $T \in \mathcal{L}(\mathcal{P}(\mathbf{F}), \mathcal{P}(\mathbf{F}))$ by,

$$(Tp)(x) = x^2p(x) \quad \text{for } x \in \mathbf{F}.$$

Backward shift

Recall that \mathbf{F}^∞ denotes the vector space of all sequences of elements in \mathbf{F} . Define $T \in \mathcal{L}(\mathbf{F}^\infty, \mathbf{F}^\infty)$ by

$$T(x_1, x_2, x_3, \dots) = (x_2, x_3, \dots)$$

From \mathbf{R}^3 to \mathbf{R}^2

Define $T \in \mathcal{L}(\mathbf{R}^3, \mathbf{R}^2)$ by

$$T(x, y, z) = (2x - y + 3z, 7x + 5y - 6z)$$

The uniqueness part of the next result means that a linear map is completely determined by its values on a basis.

Theorem: Linear maps and basis of domain

Suppose v_1, \dots, v_n is a basis of \mathbf{V} and $w_1, \dots, w_n \in W$. Then there exists a unique linear map $T : \mathbf{V} \rightarrow \mathbf{W}$ such that

$$Tv_j = w_j$$

for each $j = 1, \dots, n$.

Proof:

First we show existence, then uniqueness. Define $T : \mathbf{V} \rightarrow \mathbf{W}$ by

$$T(c_1v_1 + \dots + c_nv_n) = c_1w_1 + \dots + c_nw_n$$

where c_1, \dots, c_n are arbitrary elements of \mathbf{F} . Since v_1, \dots, v_n is a basis, T is well-defined and is therefore a function from V to W . For each $j \in \{1, \dots, n\}$, taking $c_j = 1$ and the other c 's equal to 0 in the equation shows $Tv_j = w_j$. The map can be easily verified to be a vector space homomorphism. Thus, the desired map exists.

To prove uniqueness, suppose $S, T \in \mathcal{L}(\mathbf{V}, \mathbf{W})$ and that $\forall j \in \{1, \dots, n\}, Sv_j = Tv_j = w_j$. Then $\forall a_1, \dots, a_n \in \mathbf{F}$,

$$\begin{aligned} S(a_1v_1 + \dots + a_nv_n) &= a_1Sv_1 + \dots + a_nSv_n \\ &= a_1w_1 + \dots + a_nw_n \\ &= a_1Tv_1 + \dots + a_nTv_n \\ &= T(a_1v_1 + \dots + a_nv_n) \\ &\implies S = T \text{ for all elements in } \mathbf{span}(v_1, \dots, v_n). \end{aligned}$$

But $\mathbf{V} = \mathbf{span}(c_1, \dots, c_n)$, which implies $S = T \forall v \in \mathbf{V} \implies S = T \implies T$ is unique. \square

Algebraic Operations on $\mathcal{L}(\mathbf{V}, \mathbf{W})$

Definition: Addition and Scalar Multiplication on $\mathcal{L}(\mathbf{V}, \mathbf{W})$

Suppose $S, T \in \mathcal{L}(\mathbf{V}, \mathbf{W})$ and $\lambda \in \mathbf{F}$. The **sum** $S + T$ and the **product** λT are the maps from \mathbf{V} to \mathbf{W} defined by:

$$(S + T)(v) = Sv + Tv \quad \text{and} \quad (\lambda T)(v) = \lambda(Tv)$$

for all $v \in \mathbf{V}$.

Theorem: $\mathcal{L}(\mathbf{V}, \mathbf{W})$ is a vector space

With the operations of addition and scalar multiplication as defined above, $\mathcal{L}(\mathbf{V}, \mathbf{W})$ is a vector space.

Proof:

□

Definition: Product of Linear Maps

If $T \in \mathcal{L}(\mathbf{U}, \mathbf{V})$ and $S \in \mathcal{L}(\mathbf{V}, \mathbf{W})$, then the **product** $ST : \mathbf{U} \rightarrow \mathbf{W}$ is defined by

$$(ST)(u) = S(Tu)$$

for $u \in \mathbf{U}$.

Theorem: The product of linear maps is a linear map

If $T \in \mathcal{L}(\mathbf{U}, \mathbf{V})$ and $S \in \mathcal{L}(\mathbf{V}, \mathbf{W})$, then $ST \in \mathcal{L}(\mathbf{U}, \mathbf{W})$.

Proof:

Suppose $u, v \in \mathbf{U}$, $T \in \mathcal{L}(\mathbf{U}, \mathbf{V})$, and $S \in \mathcal{L}(\mathbf{V}, \mathbf{W})$

$$\begin{aligned} ST(u + v) &= S(Tu + Tv) & ST(\lambda u) &= S(\lambda Tu) \\ &= S(Tu) + S(Tv) & &= \lambda S(Tu) \\ &= ST(u) + ST(v) & &= \lambda ST(u) \end{aligned}$$

□

Property: Algebraic properties of products of linear maps

- **Associativity:**

$$(T_1 T_2) T_3 = T_1 (T_2 T_3)$$

whenever T_1, T_2 , and T_3 are linear maps such that products make sense.

- **Identity:**

$$TI = IT = T$$

whenever $T \in \mathcal{L}(V, W)$ and the left I is the identity map on V and the right I is the identity map on W .

- **Distributivity:**

$$(S_1 + S_2)T = S_1T + S_2T \quad \text{and} \quad S(T_1 + T_2) = ST_1 + ST_2$$

whenever $S, S_1, S_2 \in \mathcal{L}(V, W)$ and $T, T_1, T_2 \in \mathcal{L}(U, V)$.

Theorem: Linear maps take 0 to 0

Suppose $T \in \mathcal{L}(\mathbf{V}, \mathbf{W})$. Then $T(\mathbf{0}) = \mathbf{0}$.

Proof:

Let $v \in \mathbf{V}$. Then

$$T(\mathbf{0}) = T(0v) = 0T(v) = \mathbf{0}$$

as desired □

1.2 Null Spaces and Ranges

Null Spaces and Injectivity

In this section, we will learn about two subspaces intimately connected with each linear map. The first subspace exists in the domain; the second exists in the codomain.

Definition: Null Space

For $T \in \mathcal{L}(\mathbf{V}, \mathbf{W})$, the **null space** of T , denoted $\text{null } T$, is the subset of \mathbf{V} consisting of all the vectors $v \in \mathbf{V}$ that T maps to $0 \in \mathbf{W}$:

$$\text{null } T = \{v \in \mathbf{V} : Tv = 0 \in \mathbf{W}\}$$

Some mathematicians use the term **kernel** instead of null space.

Theorem: The null space is a subspace

Suppose $T \in \mathcal{L}(\mathbf{V}, \mathbf{W})$. Then $\text{null } T$ is a subspace of \mathbf{V} .

Proof:

Suppose $u, v \in \text{null } T$.

1. The additive identity $\mathbf{0} \in \text{null } T$ because $T(\mathbf{0}) = \mathbf{0}$.
2. $T(u + v) = Tu + Tv = \mathbf{0} + \mathbf{0} = \mathbf{0} \implies u + v \in \text{null } T$
3. $T(\lambda v) = \lambda Tv = \lambda \mathbf{0} = \mathbf{0} \implies \lambda v \in \text{null } T$

Hence, $\text{null } T$ is a subspace of \mathbf{V} . □

Definition: Injective, one-to-one

A function $T : \mathbf{V} \rightarrow \mathbf{W}$ is called injective if $Tu = Tv$ implies $u = v$.

Theorem: Injectivity is equivalent to null space equals $\{\mathbf{0}\}$

Let $T \in \mathcal{L}(\mathbf{V}, \mathbf{W})$. Then T is injective if and only if $\text{null } T = \{\mathbf{0}\}$

Proof:

T is injective \implies only one vector in \mathbf{V} is sent to $\mathbf{0}$ in $\mathbf{W} \implies \text{null } T = \{\mathbf{0}\}$.

$\text{null } T = \{\mathbf{0}\}$ and $Tv = Tu \implies \mathbf{0} = Tv - Tu = T(v - u) \implies v - u \in \text{null } T \implies v = u$. □

Range and Surjectivity

Definition: Range

For a function $T : \mathbf{V} \rightarrow \mathbf{W}$, the range of T is the subset of \mathbf{W} consisting of those vectors that are of the form Tv for some $v \in \mathbf{V}$:

$$\text{range } T = \{Tv : v \in \mathbf{V}\}$$

Theorem: The range is a subspace

Suppose $T \in \mathcal{L}(\mathbf{V}, \mathbf{W})$. Then $\text{range } T$ is a subspace of \mathbf{W} .

Proof:

Suppose $u, w \in \text{range } T$ and $T(v_1) = u$ and $T(v_2) = w$.

1. The additive identity $\mathbf{0} \in \text{range } T$ because $T(\mathbf{0}) = \mathbf{0}$.
2. $u + w = Tv_1 + Tv_2 = T(v_1 + v_2) \implies u + w \in \text{range } T$
3. $\lambda u = \lambda Tv_1 = T(\lambda v_1) \implies \lambda u \in \text{range } T$

□

Definition: Surjective, onto

A function $T : V \rightarrow W$ is called **surjective** or **onto** if its range is equal to W .

Fundamental Theorem of Linear Maps

The next result is so important that it gets a dramatic name.

Theorem: Fundamental Theorem of Linear Maps

Suppose \mathbf{V} is finite-dimensional and $T \in \mathcal{L}(\mathbf{V}, \mathbf{W})$. Then $\text{range } T$ is finite-dimensional and

$$\dim \mathbf{V} = \dim \text{null } T + \dim \text{range } T$$

Proof:

Because $\text{null } T$ is a subspace of \mathbf{V} , $\text{null } T$ has a basis v_1, \dots, v_m . Because v_1, \dots, v_m is linearly independent in \mathbf{V} , the list can be extended to a basis of $v_1, \dots, v_m, w_1, \dots, w_n$ of \mathbf{V} . Therefore $\dim V = m + n = \dim \text{null } T + n$, and the above proposition holds if it can be demonstrated that $\dim \text{range } T = n$.

Notice that when T acts on any $v \in \mathbf{V}$, we have

$$\begin{aligned} Tv &= T(c_1v_1 + \dots + c_mv_m + d_1w_1 + \dots + d_nw_n) \\ &= c_1Tv_1 + \dots + c_mTv_m + d_1Tw_1 + \dots + d_nTw_n \\ &= \mathbf{0} + \dots + \mathbf{0} + d_1Tw_1 + \dots + d_nTw_n \\ &= d_1Tw_1 + \dots + d_nTw_n \\ &\implies Tw_1, \dots, Tw_n \text{ spans range } T \end{aligned}$$

If we select a $v \in \mathbf{V}$ such that $Tv = \mathbf{0}$, then

$$\begin{aligned} Tv = d_1Tw_1 + \dots + d_nTw_n = \mathbf{0} &\implies v \in \text{null } T \\ &\implies v = c_1v_1 + \dots + c_mv_m \\ &\implies d_1 = \dots = d_n = 0 \\ &\implies Tw_1, \dots, Tw_n \text{ is linearly independent} \\ &\implies Tw_1, \dots, Tw_n \text{ is a basis for range } T \end{aligned}$$

Because Tw_1, \dots, Tw_n is a basis, it follows that $\dim \text{range } T = n$, as desired. By this logic, we conclude that if \mathbf{V} is finite-dimensional and $T \in \mathcal{L}(\mathbf{V}, \mathbf{W})$ then,

$$\dim \mathbf{V} = \dim \text{null } T + \dim \text{range } T$$

□

Now we can show that no injective linear map exists from a finite-dimensional vector space to a “smaller” (is measured in dimension).

Theorem: A map to a smaller dimensional subspace is not injective

Suppose \mathbf{V} and \mathbf{W} are finite-dimensional vector spaces such that $\dim \mathbf{V} > \dim \mathbf{W}$. Then no linear map from \mathbf{V} to \mathbf{W} is injective.

Proof:

Take any $T \in \mathcal{L}(\mathbf{V}, \mathbf{W})$ and suppose $\dim \mathbf{V} > \dim \mathbf{W}$. By the Fundamental Theorem of Linear Maps,

$$\begin{aligned}\dim \text{null } T &= \dim \mathbf{V} - \dim \text{range } T \\ \dim \text{null } T &\geq \dim \mathbf{V} - \dim \mathbf{W} \\ \dim \text{null } T &> 0\end{aligned}$$

This means $\text{null } T$ contains vectors other than $\mathbf{0}$. Therefore, T is not injective. \square

Theorem: A map to a larger dimensional subspace is not surjective

Suppose V and W are finite-dimensional vector spaces such that $\dim V < \dim W$. Then no linear map from V to W is surjective.

Proof:

Take any $T \in \mathcal{L}(\mathbf{V}, \mathbf{W})$ and suppose $\dim \mathbf{V} < \dim \mathbf{W}$. By the Fundamental Theorem of Linear Maps,

$$\begin{aligned}\dim \text{range } T &= \dim \mathbf{V} - \dim \text{null } T \\ &\leq \dim \mathbf{V} \\ &< \dim \mathbf{W}\end{aligned}$$

Since $\dim \text{range } T < \dim \mathbf{W}$, $\text{range } T$ does not span \mathbf{W} . Therefore, T is not surjective. \square

Theorem: Homogenous system of linear equations

A homogenous system of linear equations with more variables than equations has non-zero solutions.

Proof:

\square

Theorem: Inhomogeneous system of linear equations

An inhomogeneous system of linear equations with more equations than variables has no solution for some choice of the constant terms.

Proof:

\square

1.3 Matrices

Suppose $T \in \mathcal{L}(\mathbf{V}, \mathbf{W})$ and v_1, \dots, v_n is a basis of \mathbf{V} , then the values of Tv_1, \dots, Tv_n determine the action of T on any arbitrary vectors in \mathbf{V} . Matrices became popular for recording the values of the Tv_j 's in terms of a basis of \mathbf{W} .

Definition: Matrix, $A_{j,k}$

Let m and n denote positive integers. An m -by- n **matrix** A is a rectangular array of elements of \mathbf{F} with m rows and n columns:

$$A = \begin{pmatrix} A_{1,1} & \cdots & A_{1,n} \\ \vdots & & \vdots \\ A_{m,1} & \cdots & A_{m,n} \end{pmatrix}$$

The notation $A_{j,k}$ denotes the entry in row j , column k of A . In other words, the first index refers to the row number and the second index refers to the column number.

Now we come to the key definition in this section.

Definition: Matrix of a Linear Map, $\mathcal{M}(T)$

Suppose $T \in \mathcal{L}(\mathbf{V}, \mathbf{W})$ and v_1, \dots, v_n is a basis of \mathbf{V} and w_1, \dots, w_m is a basis of \mathbf{W} . The **matrix of** T with respect to these bases is the m -by- n matrix $\mathcal{M}(T)$ whose entries $A_{j,k}$ are defined by,

$$Tv_k = A_{1,k}w_1 + \cdots + A_{m,k}w_m \quad \text{for } k = 1, \dots, n.$$

If the bases are not clear from the context, then the notation $\mathcal{M}(T, (v_1, \dots, v_n), (w_1, \dots, w_m))$ is used.

To remember how $\mathcal{M}(T)$ is constructed from T , you might write across the top of the matrix the basis vectors v_1, \dots, v_n for the domain and along the left the basis vectors w_1, \dots, w_m for the vector space into which T maps as follows:

Example: Matrix encoding $(v_1, \dots, v_n) \mapsto (w_1, \dots, w_m)$

Only the k^{th} column is shown. Thus the second index of each displayed entry of the matrix is k .

$$\mathcal{M}(T) = \begin{matrix} & \begin{matrix} v_1 & \cdots & v_k & \cdots & v_n \end{matrix} \\ \begin{matrix} w_1 \\ \vdots \\ w_m \end{matrix} & \begin{pmatrix} & & A_{1,k} & & \\ & & \vdots & & \\ & & A_{m,k} & & \end{pmatrix} \end{matrix}$$

The above picture should remind you that Tv_k **can be computed from $\mathcal{M}(T)$ by multiplying each entry in the k^{th} column by the corresponding w_j then adding up the resulting vectors.**

If T is a linear map from \mathbf{F}^n to \mathbf{F}^m , then unless stated otherwise, assume the bases in question are the standard ones (where the k^{th} basis vector is 1 in the k^{th} slot and 0 in all the other slots).

Example: Matrix of $T \in \mathcal{L}(\mathbf{F}^n, \mathbf{F}^m)$

If $T \in \mathcal{L}(\mathbf{F}^n, \mathbf{F}^m)$ and you think of elements of \mathbf{F}^m as columns of numbers, then you can think of the k^{th} column of $\mathcal{M}(T)$ as T applied to the k^{th} standard basis vector of \mathbf{F}^n .

Addition and Scalar Multiplication of Matrices

Suppose $S, T \in \mathcal{L}(\mathbf{V}, \mathbf{W})$. For the rest of the section, assume \mathbf{V} and \mathbf{W} are finite-dimensional and have a basis chosen for them.

We are interested in whether $\mathcal{M}(S + T) = \mathcal{M}(S) + \mathcal{M}(T)$ and whether $\mathcal{M}(\lambda T) = \lambda \mathcal{M}(T)$?

Definition: Matrix Addition

The sum of two matrices of the same size is the matrix obtained by the corresponding entries of two matrices.

$$\begin{pmatrix} A_{1,1} & \cdots & A_{1,n} \\ \vdots & & \vdots \\ A_{m,1} & \cdots & A_{m,n} \end{pmatrix} + \begin{pmatrix} B_{1,1} & \cdots & B_{1,n} \\ \vdots & & \vdots \\ B_{m,1} & \cdots & B_{m,n} \end{pmatrix} = \begin{pmatrix} A_{1,1} + B_{1,1} & \cdots & A_{1,n} + B_{1,n} \\ \vdots & & \vdots \\ A_{m,1} + B_{m,1} & \cdots & A_{m,n} + B_{m,n} \end{pmatrix}$$

In the following result, the assumption is that the same bases are used for $\mathcal{M}(S)$, $\mathcal{M}(T)$, and $\mathcal{M}(S + T)$.

Theorem: The Matrix of the Sum of Linear Maps

Suppose $S, T \in \mathcal{L}(\mathbf{V}, \mathbf{W})$. Then $\mathcal{M}(S + T) = \mathcal{M}(S) + \mathcal{M}(T)$.

Proof:

□

Still assuming that the same bases are used, is the matrix of a scalar times a linear map the same as the scalar times the matrix of the linear map? Again we must provide a definition for scalar multiplication.