

Linear Algebra Done Right

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1 Vector Spaces

1.1 \mathbb{R}^n and \mathbb{C}^n

Definition 1.1.1 (Complex Number). A *complex number* is an ordered pair (a, b) , where $a, b \in \mathbb{R}$, but we write it as $a + bi$.

Notation 1.1.2. $\mathbb{C} := \{a + bi \mid a, b \in \mathbb{R}\}$

Definition 1.1.3 (Addition & Multiplication).

$$(a + bi) + (c + di) = (a + c) + (b + d)i$$

$$(a + bi)(c + di) = (ac - bd) + (ad + bc)i$$

Theorem 1.1.4 Properties of Complex Arithmetic

1. commutativity: $\alpha + \beta = \beta + \alpha$; $\alpha\beta = \beta\alpha$, $\forall \alpha, \beta \in \mathbb{C}$.
2. associativity: $(\alpha + \beta) + \lambda = \alpha + (\beta + \lambda)$; $(\alpha\beta)\lambda = \alpha(\beta\lambda)$, $\forall \alpha, \beta, \lambda \in \mathbb{C}$.
3. identities: $\lambda + 0 = \lambda$; $\lambda \cdot 1 = \lambda$, $\forall \lambda \in \mathbb{C}$.
4. additive inverse: $\forall \alpha \in \mathbb{C}, \exists$ unique $\beta \in \mathbb{C}$ s.t. $\alpha + \beta = 0$.
5. multiplicative inverse: $\forall \alpha \in \mathbb{C}, \alpha \neq 0, \exists$ unique $\beta \in \mathbb{C}$ s.t. $\alpha\beta = 1$.
6. distributivity: $\lambda(\alpha + \beta) = \lambda\alpha + \lambda\beta$, $\forall \lambda, \alpha, \beta \in \mathbb{C}$.

Definition 1.1.5 (Subtraction). If $-\alpha$ is the additive inverse of α , *subtraction* on \mathbb{C} is defined by

$$\beta - \alpha = \beta + (-\alpha).$$

Definition 1.1.6 (Division). For $\alpha \neq 0$, let $\frac{1}{\alpha}$ denote the multiplicative inverse of α . Then, *division* on \mathbb{C} is defined by

$$\frac{\beta}{\alpha} = \beta \cdot \left(\frac{1}{\alpha}\right)$$

Notation 1.1.7. \mathbb{F} is either \mathbb{R} or \mathbb{C} .

Definition 1.1.8 (List/Tuple). Suppose n is a non-negative integer. A list of length n is an ordered collection of n elements separated by commas and surrounded by parentheses: $(x_1, x_2, x_3, \dots, x_n)$. Two lists are equal if and only if they have the same length and the same elements in the same order.

Remark. Lists must have a FINITE length.

Definition 1.1.9 (\mathbb{F}^n and Coordinate). \mathbb{F}^n is the set of all lists of length n of elements of \mathbb{F} :

$$\mathbb{F}^n := \{(x_1, \dots, x_n) \mid x_i \in \mathbb{F} \forall i = 1, \dots, n\},$$

where x_i is the i^{th} coordinate of (x_1, \dots, x_n) .

Example 1.1.10 $\mathbb{R}^2 = \{(x, y) \mid x, y \in \mathbb{R}\}$ and $\mathbb{R}^3 = \{(x, y, z) \mid x, y, z \in \mathbb{R}\}$.

Definition 1.1.11 (Addition on \mathbb{F}^n). Addition on \mathbb{F}^n is defined by adding corresponding coordinates:

$$(x_1, \dots, x_n) + (y_1, \dots, y_n) = (x_1 + y_1, \dots, x_n + y_n).$$

Theorem 1.1.12 Commutativity of Addition on \mathbb{F}^n

If $x, y \in \mathbb{F}^n$, then $x + y = y + x$.

Proof 1. Suppose $x = (x_1, \dots, x_n)$ and $y = (y_1, \dots, y_n)$. Then

$$\begin{aligned} x + y &= (x_1 + y_1, \dots, x_n + y_n) \\ &= (y_1 + x_1, \dots, y_n + x_n) = y + x. \end{aligned}$$

■

Definition 1.1.13 (Zero). Let 0 denote the list of length n whose coordinates are all 0: $0 := (0, \dots, 0)$.

Definition 1.1.14 (Additive Inverse on \mathbb{F}^n). For $x \in \mathbb{F}^n$, the additive inverse of x , denoted $-x$, is the vector $-x \in \mathbb{F}^n$ s.t. $x + (-x) = 0$.

Definition 1.1.15 (Scalar Multiplication in \mathbb{F}^n). The product of a number $\lambda \in \mathbb{F}$ and a vector $x \in \mathbb{F}^n$ is computed by multiplying each coordinate of the vector by λ :

$$\lambda x = \lambda(x_1, \dots, x_n) = (\lambda x_1, \dots, \lambda x_n),$$

where $x = (x_1, \dots, x_n) \in \mathbb{F}^n$.

Theorem 1.1.16 Properties of Arithmetic Operations on \mathbb{F}^n

1. $(x + y) + z = x + (y + z) \quad \forall x, y, z \in \mathbb{F}^n$
2. $(ab)x = a(bx) \quad \forall x \in \mathbb{F}^n \text{ and } \forall a, b \in \mathbb{F}.$
3. $1 \cdot x = x \quad \forall x \in \mathbb{F}^n \text{ and } 1 \in \mathbb{F}.$
4. $\lambda(x + y) = \lambda x + \lambda y \quad \forall \lambda \in \mathbb{R} \text{ and } \forall x, y \in \mathbb{F}^n.$
5. $(a + b)x = ax + bx \quad \forall a, b \in \mathbb{F} \text{ and } \forall x \in \mathbb{F}^n.$

1.2 Definition of Vector Space

Definition 1.2.1 (Addition on V). An *addition* on V is a function $(u, v) \mapsto u + v$ for all $u, v \in V$.

Definition 1.2.2 (Scalar Multiplication on V). A *scalar multiplication* on V is a function $(\lambda, v) \mapsto \lambda v$ for all $\lambda \in \mathbb{F}$ and $v \in V$.

Definition 1.2.3 (Vector Space). A *vector space* is a set V along with an addition on V and a scalar multiplication s.t. the following properties hold:

1. commutativity: $u + v = v + u \quad \forall u, v \in V$
2. associativity: $(u + v) + w = u + (v + w)$ and $(ab)v = a(bv) \quad \forall u, v, w \in V$ and $\forall a, b \in \mathbb{F}$.
3. additive identity: $\exists 0 \in V$ s.t. $v + 0 = v \quad \forall v \in V$.
4. additive inverse: $\exists w \in V$ s.t. $v + w = 0 \quad \forall v \in V$.
5. multiplicative identity: $\exists 1 \in V$ s.t. $1 \cdot v = v \quad \forall v \in V$.
6. distributive properties: $a(u + v) = au + av$ and $(a + b)v = av + bv \quad \forall u, v \in V$ and $a, b \in \mathbb{F}$.

Definition 1.2.4 (Vector). Elements of a vector space are called *vectors* or *points*.

Notation 1.2.5. V is a vector space over \mathbb{F} .

Definition 1.2.6 (Real and Complex Vector Space). A vector space over \mathbb{R} is called a *real vector space*, and a vector space over \mathbb{C} is called a *complex vector space*.

Theorem 1.2.7 Unique Additive Identity of Vector Spaces

A vector space has a unique additive identity.

Proof 1. Suppose 0 and $0'$ are both additive identities for some vector space V . So,

$$\begin{aligned} 0' &= 0' + 0 && \text{Since } 0 \text{ is an additive identity} \\ &= 0 + 0' && \text{commutativity} \\ &= 0. && \text{Since } 0' \text{ is an additive identity} \end{aligned}$$

Then, $0' = 0$. ■

Theorem 1.2.8 Unique Additive Inverse of Vector Spaces

A vector in a vector space has a unique additive inverse.

Proof 2. Let V be a vector space. Suppose w and w' are additive inverses of v for some $v \in V$. Note that

$$\begin{aligned} w &= w + 0 \\ &= w + (v + w') \\ &= (w + v) + w' \\ &= 0 + w' = w'. \end{aligned}$$
■

Notation 1.2.9. Let $v, w \in V$. Then, $-v$ denotes the additive inverse of v .

Definition 1.2.10 (Subtraction). $w - v$ is defined to be $w + (-v)$.

Theorem 1.2.11

$$0 \cdot v = 0 \quad \forall v \in V.$$

Proof3. Since $v \in V$, we know

$$\begin{aligned} 0 \cdot v &= (0 + 0)v = 0 \cdot v + 0 \cdot v \\ 0 \cdot v + (-0 \cdot v) &= 0 \cdot v + 0 \cdot v + (-0 \cdot v) \\ 0 &= 0 \cdot v \end{aligned}$$

■

Theorem 1.2.12

$$a \cdot 0 = 0 \quad \forall a \in \mathbb{F}.$$

Proof4. For $a \in \mathbb{F}$, we have

$$\begin{aligned} a \cdot 0 &= a \cdot (0 + 0) = a \cdot 0 + a \cdot 0 \\ a \cdot 0 + (-a \cdot 0) &= a \cdot 0 + a \cdot 0 + (-a \cdot 0) \\ 0 &= a \cdot 0. \end{aligned}$$

■

Theorem 1.2.13

$$(-1)v = -v \quad \forall v \in V.$$

Proof5. For $v \in V$, we have

$$v + (-1)v = 1 \cdot v + (-1) \cdot v = (1 + (-1)) \cdot v = 0 \cdot v = 0.$$

Therefore, by definition, $(-1)v = -v$.

■

Notation 1.2.14. \mathbb{F}^S

1. If S is a set, then \mathbb{F}^S denotes the set of functions from S to \mathbb{F} .
2. For $f, g \in \mathbb{F}^S$, the sum $f + g \in \mathbb{F}^S$ is the function defined by $(f + g)(x) = f(x) + g(x) \quad \forall x \in S$.
3. For $\lambda \in \mathbb{F}$ and $f \in \mathbb{F}^S$, the product $\lambda f \in \mathbb{F}^S$ is the function defined by $(\lambda f)(x) = \lambda f(x) \quad \forall x \in S$.

Theorem 1.2.15

\mathbb{F}^S is a vector space.

1.3 Subspace

Definition 1.3.1 (Subspace). A subset U of V is called a *subspace* of V if U is also a vector space using the same addition and scalar multiplication as on V .

Theorem 1.3.2 Conditions for a Subspace

A subset U of V is a subspace of V if and only if U satisfies the following conditions:

1. additive identity: $0 \in U$;
2. closed under addition: $u, w \in U \implies u + w \in U$;
3. closed under scalar multiplication: $a \in \mathbb{F}$ and $u \in U \implies au \in U$.

Proof 1.

(\Rightarrow) Suppose U is a subspace of V . By definition, U is then a vector space, and so those conditions are automatically satisfied. \square

(\Leftarrow) Suppose U satisfies the three conditions. Since U is a subset of V , U automatically has *associativity*, *commutativity*, *multiplicative identity*, and *distributivity*. So, we want to check U has additive inverse and additive identities.

For additive identity, we know $0 \in U$, by assumption.

For additive inverse, by condition #3, we know $-u = (-1)u \in U$.

Then, U is a vector space. ■

Example 1.3.3 If $b \in \mathbb{F}$, then $\{(x_1, x_2, x_3, x_4) \in \mathbb{F}^4 \mid x_3 = 5x_4 + b\}$ is a subspace of \mathbb{F}^4 if and only if $b = 0$.

Proof 2.

(\Rightarrow) Suppose $U = \{(x_1, x_2, x_3, x_4) \in \mathbb{F}^4 \mid x_3 = 5x_4 + b\}$ is a subspace of \mathbb{F}^4 . Then, $0 = (0, 0, 0, 0) \in U$. So, $0 = 5 \cdot 0 + b$, or $b = 0$. \square

(\Leftarrow) Suppose $b = 0$. Then, $x_3 = 5x_4$. So, $U = \{(x_1, x_2, 5x_4, x_4) \in \mathbb{F}^4\}$

1. $0 = (0, 0, 0, 0) \in U$

2. Note that

$$(x_1, x_2, 5x_4, x_4) + (y_1, y_2, 5y_4, y_4) = (x_1 + y_1, x_2 + y_2, 5(x_4 + y_4), x_4 + y_4) \in U$$

So, addition is closed under U .

3. $\forall a \in \mathbb{F}$, we have

$$a(x_1, x_2, 5x_4, x_4) = (ax_1, ax_2, 5(ax_4), ax_4) \in U$$

Then, U is a subspace of \mathbb{F}^4 . ■

Example 1.3.4 The set of continuous real-valued functions on interval $[0, 1]$ is a subspace of $\mathbb{R}^{[0,1]}$.

Proof 3.

1. 0 (zero mapping) $\in U$
2. Set f and $g \in \mathcal{C}[0, 1]$, the set of continuous functions on interval $[0, 1]$. Then, $f + g \in \mathcal{C}[0, 1]$.
3. From Calculus, we know that $\forall a \in \mathbb{F}, \quad af \in \mathcal{C}[0, 1]$.

■

Definition 1.3.5 (Sum of Subspaces). Suppose U_1, \dots, U_m are subspaces of V . The *sum* of U_1, \dots, U_m , denoted as $U_1 + \dots + U_m$, is the set of all possible sums of elements of U_1, \dots, U_m :

$$U_1 + \dots + U_m = \{u_1 + \dots + u_m \mid u_i \in U_i \quad \forall i = 1, \dots, m\}.$$

Example 1.3.6 Suppose $U = \{(x, 0, 0) \in \mathbb{F}^3 \mid x \in \mathbb{F}\}$ and $W = \{(0, y, 0) \in \mathbb{F}^3 \mid y \in \mathbb{F}\}$, then

$$U + W = \{(x, y, 0) \in \mathbb{F}^3 \mid x, y \in \mathbb{F}\}.$$

Theorem 1.3.7

Suppose U_1, \dots, U_m are subspaces of V . Then, $U_1 + \dots + U_m$ is the *smallest subspace* of V containing U_1, \dots, U_m .

Proof 4. Suppose U_1, \dots, U_m are subspaces of U . Let $U_1 + \dots + U_m = \{u_1 + \dots + u_m \mid u_j \in U_j, j = 1, \dots, m\}$. Suppose $w_j \in U_j$, then $w_1 + \dots + w_m \in U_1 + \dots + U_m$.

1. $U_1 + \dots + U_m$ is a subspace of V .

(a) Note that

$$(u_1 + \dots + u_m) + (w_1 + \dots + w_m) = (u_1 + w_1) + \dots + (u_m + w_m) \in U_1 + \dots + U_m,$$

so $U_1 + \dots + U_m$ is closed under addition.

(b) Similarly, $U_1 + \dots + U_m$ is closed under scalar multiplication.

(c) Note that U_j is a subspace, so $0 \in U_j$. Hence, $(0, \dots, 0) = 0 \in U_1 + \dots + U_m$. □

2. Now, we want to show this subspace is the smallest subspace containing U_1, \dots, U_m . That is, we want to show $\forall W \supseteq U_1 \cup \dots \cup U_m$, we have $W \supseteq U_1 + \dots + U_m$.

Note that $U_j \subseteq U_1 + \dots + U_m$, so we have $(U_1 \cup U_2 \cup \dots \cup U_m) \subseteq U_1 + \dots + U_m$. This means $U_1 + \dots + U_m$ must contain U_1, \dots, U_m . Let W be some subspace containing U_1, \dots, U_m . Then, for $j = 1, \dots, m$, we have $u_j \in U_j$, which indicates $u_j \in W$. Therefore, $u_1 + \dots + u_m \in W$ and thus $U_1 + \dots + U_m \subseteq W$.

Since W was arbitrary, we've shown $\forall W$ that contains U_1, \dots, U_m , $U_1 + \dots + U_m \subseteq W$. Therefore, $U_1 + \dots + U_m$ is the smallest.

■

Definition 1.3.8 (Direct Sum). Suppose U_1, \dots, U_m are subspaces of V . $U_1 + \dots + U_m$ is called a *direct sum* if each element of $U_1 + \dots + U_m$ can be written in only one way as a sum $u_1 + \dots + u_m$, where $u_j \in U_j$.

Notation 1.3.9. If $U_1 + \dots + U_m$ is a direct sum, then we use $U_1 \oplus \dots \oplus U_m$ to denote it.

Example 1.3.10 Let $U = \{(x, y, 0) \in \mathbb{F}^3 \mid x, y \in \mathbb{F}\}$ and $W = \{(0, 0, z) \in \mathbb{F}^3 \mid z \in \mathbb{F}\}$. Then, $\mathbb{F}^3 = U \oplus W$.

Proof 5. Note that $U + W = \{(x, y, z) \mid x, y, z \in \mathbb{F}\} = \mathbb{F}^3$. Suppose

$$(x, y, z) = (x, y, 0) + (0, 0, z), \quad (1)$$

for some $x, y, z \in \mathbb{F}$ and

$$(x, y, z) = (x', y', 0) + (0, 0, z') \quad (2)$$

for some $x', y', z' \in \mathbb{F}$. Then, (1)–(2):

$$(0, 0, 0) = (x - x', y - y', 0) + (0, 0, z - z') = (x - x', y - y', z - z').$$

Then, $x - x' = y - y' = z - z' = 0$, which indicates $x = x'$, $y = y'$, $z = z'$. So, by definition $U + W$ is a direct sum, or $\mathbb{F}^3 = U \oplus W$. ■

Example 1.3.11 Suppose U_j is the subspace of \mathbb{F}^n s.t.

$$U_1 = \{x, 0, 0, \dots, 0 \mid x \in \mathbb{F}\}$$

$$U_2 = \{0, x, 0, \dots, 0 \mid x \in \mathbb{F}\}$$

$$\vdots$$

$$U_n = \{0, 0, 0, \dots, x \mid x \in \mathbb{F}\}$$

Then, $\mathbb{F}^n = U_1 \oplus U_2 \oplus \dots \oplus U_n$.

Proof 6. Note that $\mathbb{F}^n = U_1 + U_2 + \dots + U_n$ is evident. Now, we'll prove that $U_1 + U_2 + \dots + U_n$ is a direct sum. Consider $x = (x_1, x_2, \dots, x_n) \in \mathbb{F}^n$. Assume that

$$x = (x_1, 0, \dots, 0) + \dots + (0, \dots, 0, x_n) \quad (3)$$

and

$$x = (x'_1, 0, \dots, 0) + \dots + (0, \dots, 0, x'_n) \quad (4)$$

Then, from (3)–(4), we know that

$$0 = (x_1 - x'_1, \dots, x_n - x'_n) = (0, 0, \dots, 0).$$

Then, $\forall i = 1, \dots, n$ we have $x_i - x'_i = 0$, or $x_i = x'_i$. Therefore, by definition, we know $U_1 + \dots + U_n$ is a direct sum. ■

Example 1.3.12 Let

$$U_1 = \{(x, y, 0) \mid x, y \in \mathbb{F}\}$$

$$U_2 = \{(0, 0, z) \mid z \in \mathbb{F}\}$$

$$U_3 = \{(0, y, y) \mid y \in \mathbb{F}\}$$

Show that $U_1 + U_2 + U_3$ is not a direct sum.

Proof 7. Consider $(0, 0, 0) \in \mathbb{F}^3$. Note that

$$(0, 0, 0) = (0, 0, 0) + (0, 0, 0) + (0, 0, 0)$$

and

$$(0, 0, 0) = (0, 1, 0) + (0, 0, 1) + (0, -1, -1).$$

Then, $U_1 + U_2 + U_3$ is not a direct sum by definition. ■

Theorem 1.3.13

Suppose U_1, \dots, U_m are subspaces of V . Then, $U_1 + \dots + U_m$ is a direct sum if and only if the only way to write 0 as a sum $u_1 + \dots + u_m$ is by taking each $u_j = 0$.

Proof 8.

(\Rightarrow) Since $U_1 + \dots + U_m$ is a direct sum, by definition, the only way to write $0 \in \mathbb{F}^n$ is to write it as

$$0 = 0 + \dots + 0 \quad \text{where } 0 \in U_i \forall i = 1, \dots, m. \quad \square$$

(\Leftarrow) Suppose the only way to write 0 as a sum $u_1 + \dots + u_m$ is by taking each $u_j = 0$. Assume that for some $v \in V$, we have

$$v = u_1 + \dots + u_m, \quad u_j \in U_j \tag{5}$$

and

$$v = u'_1 + \dots + u'_m, \quad u'_j \in U_j. \tag{6}$$

Then, by (5)-(6), and according to the conclusion from Example 1.3.11, we have

$$0 = (u_1 - u'_1) + \dots + (u_m - u'_m) = 0 + \dots + 0.$$

So, $\forall i \in 1, \dots, m$, we have $u_i - u'_i = 0$. that is, $u_i = u'_i$. So, $\forall v \in V$, there is only one way to write v as a sum of $u_1 + \dots + u_m$. Therefore, by definition, $U_1 + \dots + U_m$ is a direct sum. ■

Theorem 1.3.14

Suppose U and W are subspaces of V . Then, $U + W$ is a direct sum if and only if $U \cap W = \{0\}$.

Proof 9.

(\Rightarrow) Suppose $U + W$ is a direct sum. Assume $v \in U \cap W$. Then, $v \in U$ and $v \in W$. By definition of subspace, we know $-v \in W$ as well. Note that

$$0 = v + (-v) \in U \cap W.$$

Then, by Theorem 1.3.13, we know that the only representation of $0 \in U \cap W$ is $0 = 0 + 0$ since $U \cap W$

is a direct sum. Hence, it must be that $v = -v = 0$, and thus $U \cap W = \{0\}$. \square

(\Leftarrow) Suppose $U \cap W = \{0\}$. Let $u \in U$ and $w \in W$ s.t. $u + w = 0$. Then, we have $u = -w$. Since $-w \in W$, we know $u = -w \in W$. By $u \in U$ and $u \in W$, we know that $u \in U \cap W = \{0\}$. Therefore, $0 = 0 + 0$ is the only to represent $0 \in U + W$. By Theorem 1.3.13, we know $U + W$ is a direct sum. \blacksquare

Remark. When extending Theorem 1.3.14 to 3 subspaces U_1, U_2, U_3 , we cannot conclude $U_1 \oplus U_2 \oplus U_3$ if we have $U_1 \cap U_2 = U_1 \cap U_3 = U_2 \cap U_3 = \{0\}$. See Example 1.3.12 as a counterexample.

2 Finite-Dimensional Vector Spaces

2.1 Span and Linear Independence

Notation 2.1.1. We usually write list of vectors without using parentheses.

Example 2.1.2 $(4, 1, 6), (9, 5, 7)$ is a list of vectors of length 2 in \mathbb{R}^3 .

Definition 2.1.3 (Linear Combination). A *linear combination* of a list v_1, \dots, v_m of vectors in V is a vector of the form

$$a_1v_1 + \dots + a_mv_m,$$

where $a_1, \dots, a_m \in \mathbb{F}$.

Example 2.1.4 Since $(17, -4, 2) = 6(2, 1, -3) + 5(1, -2, 4)$, we say $(17, -4, 2)$ is a linear combination of $(2, 1, -3), (1, -2, 4)$.

Definition 2.1.5 (Span).

$$\text{span}(v_1, \dots, v_m) = \{a_1v_1 + \dots + a_mv_m \mid a_1 \dots a_m \in \mathbb{F}\}.$$

Example 2.1.6 Consider $\text{span}(e_1, e_2, e_3)$:

$$\begin{aligned} \text{span}(e_1, e_2, e_3) &= \{a_1e_1 + a_2e_2 + a_3e_3 \mid a_1, a_2, a_3 \in \mathbb{F}\} \\ &= \{(a_1, a_2, a_3) \mid a_1, a_2, a_3 \in \mathbb{F}\} = \mathbb{R}^3. \end{aligned}$$

Theorem 2.1.7

The span of a list of vectors in V is the smallest subspace of V containing all the vectors in the list.

Proof 1. To prove this theorem, we will prove two parts: span is a subspace and span is the smallest subspace.

1. Span is a subspace of V .

- (a) By definition of span, we know $\text{span}(v_1, \dots, v_m) = \{a_1v_1 + \dots + a_mv_m \mid a_1, \dots, a_m \in \mathbb{F}\}$. If we set $a_1, \dots, a_m = 0$, then we have $0 = 0v_1 + \dots + 0v_m$. So, $0 \in \text{span}(v_1, \dots, v_m)$.
- (b) Let $a_1v_1 + \dots + a_mv_m \in \text{span}(v_1, \dots, v_m)$ and $b_1v_1 + \dots + b_mv_m \in \text{span}(v_1, \dots, v_m)$. Then,

$$(a_1v_1 + \dots + a_mv_m) + (b_1v_1 + \dots + b_mv_m) = (a_1 + b_1)v_1 + \dots + (a_m + b_m)v_m.$$

Since $(a_1 + b_1), \dots, (a_m + b_m) \in \mathbb{F}$, we know $(a_1 + b_1)v_1 + \dots + (a_m + b_m)v_m \in \text{span}(v_1, \dots, v_m)$.

- (c) Let $\lambda \in \mathbb{F}$ and $a_1v_1 + \dots + a_mv_m \in \text{span}(v_1, \dots, v_m)$. Then,

$$\lambda(a_1v_1 + \dots + a_mv_m) = \lambda a_1v_1 + \dots + \lambda a_mv_m.$$

Since $\lambda a_1, \dots, \lambda a_m \in \mathbb{F}$, we know that $\lambda(a_1 v_1 + \dots + a_m v_m) \in \text{span}(v_1, \dots, v_m)$.

Therefore, we have proven that span is a subspace of V . \square

2. Now, we want to show that span is the smallest subspace.

Let U be a subspace of V containing v_1, \dots, v_m . If we can show that $\text{span}(v_1, \dots, v_m) \subseteq U$, we then know span is the smallest subspace containing v_1, \dots, v_m . Since U is a subspace containing v_1, \dots, v_m , it is closed under addition and scalar multiplication. So, $a_1 v_1 + \dots + a_m v_m \in \text{span}(v_1, \dots, v_m)$. Therefore, $\text{span}(v_1, \dots, v_m) \subseteq U$. \blacksquare

Definition 2.1.8 (Span as a Verb). If $\text{span}(v_1, \dots, v_m) = V$, we say v_1, \dots, v_m *spans* V .

Definition 2.1.9 (Finite-Dimensional Vector Space). A vector space V is called *finite-dimensional* if \exists a list of vectors, say v_1, \dots, v_m s.t. $\text{span}(v_1, \dots, v_m) = V$. In the following of this notes, we will use *f-d* as a shortcut for saying “finite-dimensional.”

Definition 2.1.10 (Infinte-Dimensional Vector Space). A vector space V is infinite-dimensional if it is not *f-d*. This is equivalent to say that \forall lists of vectors in V , they do not span V .

Definition 2.1.11 (Polynomial Functions). A function $p : \mathbb{F} \rightarrow \mathbb{F}$ is called a *polynomial* with coefficients in \mathbb{F} if $\exists a_0, \dots, a_m \in \mathbb{F}$ s.t. $p(z) = a_0 + a_1 z + a_2 z^2 + \dots + a_m z^m \quad \forall z \in \mathbb{F}$.

Notation 2.1.12. We use $\mathcal{P}(\mathbb{F})$ to denote the set of all polynomial with coefficients in \mathbb{F} .

Theorem 2.1.13

$\mathcal{P}(\mathbb{F})$ is a vector space over \mathbb{F} .

Proof2. Recall the definition of $\mathbb{F}^{\mathbb{F}}$. We will show $\mathcal{P}(\mathbb{F})$ is a subspace of $\mathbb{F}^{\mathbb{F}}$.

1. $0 = 0 + 0z + \dots + 0z^m \in \mathcal{P}(\mathbb{F})$.
2. Suppose $p(z) = a_m z^m + \dots + a_1 z + a_0$ and $q(z) = b_n z^n + \dots + b_1 z + b_0 \in \mathcal{P}(\mathbb{F})$. WLOG, suppose $m > n$, then we have $p(z) + q(z) = a_m z^m + \dots + (a_n + b_n) z^n + \dots + (a_0 + b_0) \in \mathcal{P}(\mathbb{F})$.
3. Suppose $\lambda \in \mathbb{F}$. Then, $\lambda p(z) = \lambda(a_m z^m + \dots + a_1 z + a_0) = \lambda a_m z^m + \dots + \lambda a_0 \in \mathcal{P}(\mathbb{F})$.

Hence, we've shown $\mathcal{P}(\mathbb{F})$ is a subspace over \mathbb{F} . \blacksquare

Definition 2.1.14 (Degree of a Polynomial). A polynomial $p \in \mathcal{P}(\mathbb{F})$ is said to have *degree* m if \exists scalars $a_0, \dots, a_m \in \mathbb{F}$ with $a_m \neq 0$ s.t. $p(z) = a_m z^m + \dots + a_1 z + a_0 \quad \forall z \in \mathbb{F}$. We write $\deg p = m$. Specially, $\deg 0 := -\infty$ and $\deg a_0 := 0$ when $a_0 \neq 0$.

Definition 2.1.15 ($\mathcal{P}_m(\mathbb{F})$). For $m \in \mathbb{N}^+$, $\mathcal{P}_m(\mathbb{F})$ denotes the set of all polynomial with coefficients in \mathbb{F} and degree $\leq m$. i.e.,

$$\mathcal{P}_m(\mathbb{F}) := \{p \in \mathcal{P}(\mathbb{F}) \mid \deg p \leq m\}.$$

Example 2.1.16 For each $m \in \mathbb{N}$, $\mathcal{P}_m(\mathbb{F})$ is a *f-d* vector space.

Proof3. Note that $\mathcal{P}_m(\mathbb{F})$ is a vector space because it is a subspace of $\mathcal{P}(\mathbb{F})$. Suppose $p(z) \in \mathcal{P}_m(\mathbb{F})$, then $p(z) = a_0 + a_1 z + \dots + a_m z^m \in \text{span}(1, z, \dots, z^m)$. Then, by definition, $\mathcal{P}_m(\mathbb{F})$ is *f-d*. \blacksquare

Remark. In this proof, we are abusing notation by letting z^k to denote a function.

Example 2.1.17 $\mathcal{P}(\mathbb{F})$ is infinite-dimensional.

Proof 4. For any list of vectors in $\mathcal{P}(\mathbb{F})$, by definition of list, the length of it is finite. Suppose the highest degree in this list is m . Consider a polynomial with degree of $m + 1$: z^{m+1} . Since z^{m+1} cannot be written as linear combinations of the list of polynomials, we know the list does not span $\mathcal{P}(\mathbb{F})$. So, $\mathcal{P}(\mathbb{F})$ is infinite-dimensional. ■

Definition 2.1.18 (Linear Independence). A list v_1, \dots, v_m of vectors in V is called *linearly independent* (L.I.) if the only choice of $a_1, \dots, a_m \in \mathbb{F}$ that makes $a_1v_1 + \dots + a_mv_m = 0$ is $a_1 = \dots = a_m = 0$. Specially, the empty list $()$ is declared to be L.I..

Definition 2.1.19 (Linear Dependence). v_1, \dots, v_m is called *linearly dependent* if it is not L.I.. Or, equivalently, v_1, \dots, v_m is *linearly dependent* if $\exists a_1, \dots, a_m \in \mathbb{F}$ not all 0 s.t. $\sum_{i=1}^m a_i v_i = 0$.

Example 2.1.20 Let $v_1, \dots, v_m \in V$. If v_j is a linear combination of other v 's, then v_1, \dots, v_m is linearly dependent.

Proof 5. By assumption, $v_j = a_1v_1 + \dots + a_{j-1}v_{j-1} + a_{j+1}v_{j+1} + \dots + a_mv_m$ for some a_i not all 0. So, $0 = a_1v_1 + \dots + a_{j-1}v_{j-1} + a_{j+1}v_{j+1} + \dots + a_mv_m - v_j$, a linear combination of v_1, \dots, v_m . Since $-v_j$ has a coefficient of $-1 \neq 0$, by definition, v_1, \dots, v_m is not L.I.. ■

Lemma 2.1.21 Linear Dependence Lemma Suppose v_1, \dots, v_m is a linearly dependent list in V . Then, $\exists j \in \{1, \dots, m\}$ s.t. the following hold:

1. $v_j \in \text{span}(v_1, \dots, v_{j-1})$
2. if the j^{th} term is removed from v_1, \dots, v_m , the span of the remaining list equals $\text{span}(v_1, \dots, v_m)$.

Proof 6.

1. Since v_1, \dots, v_m is linearly dependent, $a_1v_1 + \dots + a_mv_m = 0$, for some $a_i \neq 0$. Let j be the maximized index s.t. $a_j \neq 0$. Then, $a_{j+1} = \dots = a_m = 0$, by this assumption. Hence,

$$\begin{aligned} a_j v_j &= -a_1 v_1 - \dots - a_{j-1} v_{j-1} - a_{j+1} v_{j+1} - \dots - a_m v_m \\ &= -a_1 v_1 - \dots - a_{j-1} v_{j-1} \\ v_j &= -\frac{a_1}{a_j} v_1 - \dots - \frac{a_{j-1}}{a_j} v_{j-1}. \end{aligned}$$

Since $-\frac{a_1}{a_j}, \dots, -\frac{a_{j-1}}{a_j} \in \mathbb{F}$, we know $v_j \in \text{span}(v_1, \dots, v_{j-1})$. □

2. Consider

$$\begin{aligned} \text{span}(v_1, \dots, v_j, \dots, v_m) &= \text{span}\left(v_1, \dots, -\frac{a_1}{a_j} v_1 - \dots - \frac{a_{j-1}}{a_j} v_{j-1}, \dots, v_m\right) \\ &= \text{span}(v_1, \dots, v_{j-1}, v_{j+1}, \dots, v_m). \end{aligned}$$

■

Remark. By using this Lemma 2.1.21, we can do lots of proofs using the “step” strategy. Namely, we start to remove vectors from a list that are linearly dependent to obtain a L.I. list. However, this “step” strategy can only be used when dealing with FINITE-dimensional vector spaces.

Theorem 2.1.22

Let V be a f - d vector space. Let $\text{span}(w_1, \dots, w_n) = V$. Let u_1, \dots, u_m be L.I.. Then, $m \leq n$.

Proof 7.

Step 1 Note that u_1, w_1, \dots, w_n is linearly dependent because $u_1 \in V = \text{span}(w_1, \dots, w_n)$. Then, by Lemma 2.1.21, we can remove one of the w 's, say w_{j1} . Then, the list becomes

$$\{u_1, w_1, \dots, w_n\} \setminus \{w_{j1}\}.$$

Step 2 Adjoin u_2 . Apply the same reasoning, since $\text{span}(\{u_1, w_1, \dots, w_n\} \setminus \{w_{j1}\}) = V$, we know $\{u_1, u_2, w_1, \dots, w_n\} \setminus \{w_{j1}\}$ is linearly dependent. Since $u_2 \notin \text{span}(u_1)$, Lemma 2.1.21 is not applicable to u_2 . Now, we can remove another w from the list, say w_{j2} . The list becomes

$$\{u_1, u_2, w_1, \dots, w_n\} \setminus \{w_{j1}, w_{j2}\}.$$

\vdots

Step m After m steps, we list will become

$$\{u_1, \dots, u_m, w_1, \dots, w_n\} \setminus \{w_{j1}, \dots, w_{jm}\}.$$

Since $\text{span}(\{u_1, \dots, u_m, w_1, \dots, w_n\} \setminus \{w_{j1}, \dots, w_{jm}\}) = V$, this list is still linearly dependent, so by Lemma 2.1.21, we know $\exists w$ to be removed. Therefore, $n \geq m$. ■

Theorem 2.1.23

Every subspace of a f - d vector space is f - d .

Proof 8. Suppose V to be a f - d vector space and U to be a subspace of V .

Step 1 If $U = \{0\}$, then U is f - d . If $U \neq \{0\}$, then choose $v_1 \in U$ s.t. $v_1 \neq 0$.

\vdots

Step j If $U = \text{span}(v_1, \dots, v_{j-1})$, then U is f - d . If $U \neq \text{span}(v_1, \dots, v_{j-1})$, then choose $v_j \in U$ s.t. $v_j \notin \text{span}(v_1, \dots, v_{j-1})$.

By Lemma 2.1.21 and Theorem 2.1.22, we know this process will eventually terminate because the vector list that spans U cannot be longer than any spanning list of V . Therefore, U is f - d . ■

2.2 Bases

Definition 2.2.1 (Basis). A *basis* of V is a list of vectors in V that is L.I. and spans V .

Example 2.2.2

1. The standard basis of \mathbb{F}^n :

$$(1, 0, \dots, 0), (0, 1, 0, \dots, 0), \dots, (0, \dots, 0, 1).$$

2. $(1, 1, 0), (0, 0, 1)$ is a basis of V , where $V = \{(x, x, y) \in \mathbb{F}^3 \mid x, y \in \mathbb{F}\}$.

Proof 1.

- (a) Suppose $a_1(1, 1, 0) + a_2(0, 0, 1) = 0$, we have $(a_1, a_1, a_2) = 0$. So, it must be $a_1 = a_2 = 0$. Therefore, $(1, 1, 0), (0, 0, 1)$ is L.I. \square
- (b) Suppose $(x, x, y) \in V$. Note that $(x, x, y) = x(1, 1, 0) + y(0, 0, 1)$, then, $V = \text{span}((1, 1, 0), (0, 0, 1))$.

Therefore, we've proven $(1, 1, 0), (0, 0, 1)$ is a basis of V according to the definition of basis. \blacksquare

Theorem 2.2.3 Criterion for Basis

A list $v_1, \dots, v_n \in V$ is a basis list of V if and only if every $v \in V$ can be written uniquely in the form $v = a_1v_1 + \dots + a_nv_n$, where $a_i \in \mathbb{F}$.

Proof 2.

(\Rightarrow) Let v_1, \dots, v_n be a basis of V . Let $v \in V$. By definition of basis, $V = \text{span}(v_1, \dots, v_n)$. So, $v \in \text{span}(v_1, \dots, v_n)$, and thus $v = a_1v_1 + \dots + a_nv_n$ for some $a_i \in \mathbb{F}$. Assume for the sake of contradiction that $v = b_1v_1 + \dots + b_nv_n$ for some $b_i \neq a_i \in \mathbb{F}$. Then,

$$\begin{aligned} v - v &= (a_1 - b_1)v_1 + \dots + (a_n - b_n)v_n \\ 0 &= (a_1 - b_1)v_1 + \dots + (a_n - b_n)v_n. \end{aligned}$$

Since v_1, \dots, v_n is a basis, it is L.I.. So, $0 = 0v_1 + \dots + 0v_n$. Therefore, we know $a_1 - b_1 = \dots = a_n - b_n = 0$. That is, $a_1 = b_1, \dots, a_n = b_n$. * This is a contradiction with the assumption that $\exists a_i \neq b_i$. Hence, it must be that $v = a_1v_1 + \dots + a_nv_n$ is unique. \square

(\Leftarrow) Suppose $v = a_1v_1 + \dots + a_nv_n$ is the unique representation $\forall v \in V$. Then, $v \in \text{span}(v_1, \dots, v_n)$. Since $v \in V$, then $V \subseteq \text{span}(v_1, \dots, v_n)$. However, $v_1, \dots, v_n \in V$, so $\text{span}(v_1, \dots, v_n) \subseteq V$. Therefore, $\text{span}(v_1, \dots, v_n) = V$. To show v_1, \dots, v_n is L.I., further consider $0 = a_1v_1 + \dots + a_nv_n$. Since $0 \in V$, by assumption, \exists a unique way to write 0 as $a_1v_1 + \dots + a_nv_n$, and that unique way is to take every $a_i = 0$. Hence, by definition, we know v_1, \dots, v_n is L.I.. Since v_1, \dots, v_n is L.I. and $\text{span}(v_1, \dots, v_n) = V$, we know v_1, \dots, v_n is a basis list of V . \blacksquare

Theorem 2.2.4

Every spanning list can be reduced to a basis of the vector space.

Proof 3. Suppose $V = \text{span}(v_1, \dots, v_n)$. If $v_i = 0$, we just remove v_i . So, let's suppose $v_i \neq 0$.

Step 1 If $v_2 \in \text{span}(v_1)$, delete it. If $v_2 \notin \text{span}(v_1)$, keep it.

\vdots

Step j If $v_j \in \text{span}(v_1, \dots, v_{j-1})$, delete it. If $v_j \notin \text{span}(v_1, \dots, v_{j-1})$, keep it.

\vdots

Step n After n steps, we will have a “sub-list” from the original list *s.t.* it spans V and is L.I.. Therefore, the basis list is contained in the spanning list. ■

Corollary 2.2.5 Every f - d vector space has a basis.

Proof 4. By definition, f - d vector space always has a spanning list. By Theorem 2.2.4, a spanning list contain a basis. ■

Theorem 2.2.6

Every linearly independent list of vectors in a f - d vector space can be extended to a basis of the vector space.

Proof 5. Suppose u_1, \dots, u_m is L.I. in a f - d vector space of V . Let w_1, \dots, w_n be a basis of V . Then, $u_1, \dots, u_m, w_1, \dots, w_n$ spans V . According to Lemma 2.1.21 and Theorem 2.1.22, we can reduce $u_1, \dots, u_m, w_1, \dots, w_n$ to some list of u_1, \dots, u_m and some w 's. ■

Theorem 2.2.7

Suppose V is f - d and U is a subspace of V . Then, there is a subspace W of V *s.t.* $V = U \oplus W$.

Proof 6. Since V is f - d , U , as V 's subspace, is also f - d . So, \exists a basis of U , say u_1, \dots, u_m . Then, u_1, \dots, u_m is L.I. and $\in V$. By Theorem 2.2.6, this list can be extended to a basis

$$u_1, \dots, u_m, w_1, \dots, w_n \text{ of } V.$$

Let $W = \text{span}(w_1, \dots, w_n)$. We'll show $V = U \oplus W$.

1. WTS: $V = U + W$. Suppose $v \in V$. Then,

$$v = \underbrace{a_1 u_1 + \dots + a_m u_m}_{\in U} + \underbrace{b_1 w_1 + \dots + b_n w_n}_{\in W}.$$

So, $v \in U + W$, or $V = U + W$. □

2. WTS: $U \cap W = \{0\}$. Suppose $v \in U \cap W$. Then, $v \in U$ and $v \in W$. So,

$$v = a_1 u_1 + \dots + a_m u_m = b_1 w_1 + \dots + b_n w_n.$$

Hence,

$$a_1 u_1 + \dots + a_m u_m - b_1 w_1 - \dots - b_n w_n = 0. \quad (7)$$

Since by assumption, $u_1, \dots, u_m, w_1, \dots, w_n$ is a basis of V , so $u_1, \dots, u_m, w_1, \dots, w_n$ is L.I.. Therefore, the only way for Equation (7) to hold is when $a_1 = \dots = a_m = b_1 = \dots = b_n = 0$. Hence, $v = 0u_1 + \dots + 0u_m = 0$. That is, $U \cap W = \{0\}$.

Therefore, we've shown that $V = U \oplus W$. ■

2.3 Dimension

Theorem 2.3.1

Let B_1 and B_2 be two bases of V , then B_1 and B_2 have the same length.

Proof 1. Since B_1 is L.I. in V and B_2 spans V , by Theorem 2.1.22, we know $\text{len}(B_1) \leq \text{len}(B_2)$. Interchanging the roles of B_1 and B_2 , we have $\text{len}(B_2) \leq \text{len}(B_1)$. So, we have $\text{len}(B_1) = \text{len}(B_2)$. ■

Definition 2.3.2 (Dimension). The *dimension* of a f -d vector space V is the length of any basis of V .

Notation 2.3.3. We use $\dim V$ to denote the dimension of a f -d vector space V .

Example 2.3.4 $\dim \mathbb{F}^n = n$ and $\dim \mathcal{P}_m(\mathbb{F}) = m + 1$ ($1, z, z^2, \dots, z^m$).

Theorem 2.3.5

If V is f -d and U is a subspace of V , then $\dim U \leq \dim V$.

Proof 2. Let B_1 be a basis of U and B_2 be a basis of V . Then, B_1 is a L.I. list of V and B_2 spans V . Then, By Theorem 2.1.22, we know that $\text{len}(B_1) \leq \text{len}(B_2)$. So, by definition of dimension, we know $\dim U \leq \dim V$. ■

Extension. If V is f -d and U is a subspace of V , given $U \subsetneq V$, then $\dim U < \dim V$.

Proof 3. Let u_1, \dots, u_m be a basis of U . Since $U \subsetneq V$, we know $V - U \neq \emptyset$. So, choose $v \in V - U$. Then, $v \notin \text{span}(u_1, \dots, u_m)$. Therefore, u_1, \dots, u_m, v is L.I. in V . That is

$$\begin{aligned} \dim V &\geq \dim(\text{span}(u_1, \dots, u_m, v)) \\ &> \dim(\text{span}(u_1, \dots, u_m)) \\ &= \dim U. \end{aligned}$$

■

Theorem 2.3.6

Let V be f -d, then every L.I. list of vectors in V with length $\dim V$ is a basis of V .

Proof 4. Let $v_1, \dots, v_n \in V$ be L.I.. Let $n = \dim V$. When extending the list to basis, we get

$$\{v_1, \dots, v_n\} \cup \emptyset$$

as a basis of V . That is, v_1, \dots, v_n has already been a basis of V . ■

Remark. The proof given above is not that straight-forward, so we are giving an easier-understanding proof as follows.

Proof 5. Suppose for the sake of contradiction that $\exists v_1, \dots, v_n \in V$ not a basis of V for $n = \dim V$. Then, $\text{span}(v_1, \dots, v_n) \neq V$. That is, $\exists v_{n+1}$ s.t. $v_{n+1} \notin \text{span}(v_1, \dots, v_n)$. Adding v_{n+1} to the vector list, we have v_1, \dots, v_n, v_{n+1} is L.I.. By Theorem 2.3.5, we know $\text{len}(v_1, \dots, v_{n+1}) = n + 1 \leq \dim V$. * This contradicts with the fact that $\dim V = n < n + 1$. So, our assumption is incorrect, and it must be that v_1, \dots, v_n is a basis of V . ■

Theorem 2.3.7

Suppose V is f - d . Then, every spanning list of vectors in V with length $\dim V$ is a basis of V .

Example 2.3.8 Show that $1, (x-5)^2, (x-5)^3$ is a basis of the subspace U of $\mathcal{P}_3(\mathbb{R})$ defined by

$$U = \{p \in \mathcal{P}_3(\mathbb{R}) \mid p'(5) = 0\}.$$

Proof 6. Consider $a_1 + a_2(x-5)^2 + a_3(x-5)^3 = 0$, we will get $a_1 = a_2 = a_3 = 0$ easily from the equation. Then, $1, (x-5)^2, (x-5)^3$ is L.I.. So, by Theorem 2.3.5, we know $\dim U \geq 3$. Since $U \subsetneq \mathcal{P}_3(\mathbb{R})$, we have $\dim U < \dim \mathcal{P}_3(\mathbb{R}) = 4$. Therefore, $\dim U = 3 = \text{len}(1, (x-5)^2, (x-5)^3)$. By Theorem 2.3.6, we know $1, (x-5)^2, (x-5)^3$ is a basis of U . ■

Theorem 2.3.9

If U_1 and U_2 are subspaces of a f - d vector space, then

$$\dim(U_1 + U_2) = \dim(U_1) + \dim(U_2) - \dim(U_1 \cap U_2).$$

Proof 7. Let u_1, \dots, u_m be a basis of $U_1 \cap U_2$, then $\dim(U_1 \cap U_2) = m$. Also, u_1, \dots, u_m is L.I. in U_1 , so we can extend it to a basis of U_1 as $u_1, \dots, u_m, v_1, \dots, v_j$. Then, $\dim(U_1) = m + j$. Similarly, extending u_1, \dots, u_m to a basis of U_2 , we will get $u_1, \dots, u_m, w_1, \dots, w_k$. So, $\dim(U_2) = m + k$. Now, we want to show $u_1, \dots, u_m, v_1, \dots, v_j, w_1, \dots, w_k$ is a basis of $U_1 + U_2$.

1. Since $U_1, U_2 \subseteq \text{span}(u_1, \dots, u_m, v_1, \dots, v_j, w_1, \dots, w_k)$, we know that

$$\text{span}(u_1, \dots, u_m, v_1, \dots, v_j, w_1, \dots, w_k) = U_1 + U_2. \quad \square$$

2. Suppose $a_1u_1 + \dots + a_mu_m + b_1v_1 + \dots + b_jv_j + c_1w_1 + \dots + c_kw_k = 0$. Then we know that

$$c_1w_1 + \dots + c_kw_k = -a_1u_1 - \dots - a_mu_m - b_1v_1 - \dots - b_jv_j.$$

Since $c_1w_1 + \dots + c_kw_k \in U_2$, and $-a_1u_1 - \dots - a_mu_m - b_1v_1 - \dots - b_jv_j \in U_1$, we know that $c_1w_1 + \dots + c_kw_k \in U_1 \cap U_2$. Therefore, $c_1w_1 + \dots + c_kw_k = d_1u_1 + \dots + d_mu_m$. Since $u_1, \dots, u_m, w_1, \dots, w_k$ is L.I., we know $c_1 = \dots = c_k = 0$. So, $-a_1u_1 - \dots - a_mu_m - b_1v_1 - \dots - b_jv_j = 0$. Since $u_1, \dots, u_m, v_1, \dots, v_j$ is L.I., we have $a_1 = \dots = a_m = b_1 = \dots = b_j = 0$. Therefore, we've proven $u_1, \dots, u_m, v_1, \dots, v_j, w_1, \dots, w_k$ is L.I. and thus is a basis of $U_1 + U_2$. ■

Since $u_1, \dots, u_m, v_1, \dots, v_j, w_1, \dots, w_k$ is a basis of $U_1 + U_2$, we know $\dim(U_1 + U_2) = m + j + k$. Further note that

$$\begin{aligned} \dim(U_1) + \dim(U_2) - \dim(U_1 \cap U_2) &= (m + j) + (m + k) - m \\ &= m + j + k \\ &= \dim(U_1 + U_2). \end{aligned}$$

■

3 Linear Maps

Notation 3.0.1. In this section, we use V and W to denote vector spaces over \mathbb{F} .

3.1 The Vector Space of Linear Maps

Definition 3.1.1 (Linear Map). A *linear map* from V to W is a function $T : V \rightarrow W$ with the following properties:

- additivity: $T(u + v) = Tu + Tv \quad \forall u, v \in V$.
- homogeneity: $T(\lambda v) = \lambda(Tv) \quad \forall \lambda \in \mathbb{F} \text{ and } \forall v \in V$.

Notation 3.1.2. The set of all linear maps from V to W is denoted by $\mathcal{L}(V, W)$.

Example 3.1.3

1. Zero-mapping: $0 \in \mathcal{L}(V, W)$ is defined by $0v = 0$.
2. Identity-mapping: $I \in \mathcal{L}(V, V)$ is defined by $Iv = v$.
3. Differentiation: $D \in \mathcal{L}(\mathcal{P}(\mathbb{R}), \mathcal{P}(\mathbb{R}))$ is defined by $Dp = p'$.

Proof 1. Note that $(f + g)' = f' + g'$ and $(\lambda f)' = \lambda f'$. ■

4. Integration: $T \in \mathcal{L}(\mathcal{P}(\mathbb{R}), \mathbb{R})$ is defined by $Tp = \int_0^1 p(x) dx$

Proof 2. Note that $\int_0^1 (f + g) = \int_0^1 f + \int_0^1 g$ and $\int_0^1 \lambda f = \lambda \int_0^1 f$. ■

5. Backward shift: $T \in \mathcal{L}(\mathbb{F}^\infty, \mathbb{F}^\infty)$ as $T(x_1, x_2, x_3, \dots) = (x_2, x_3, \dots)$.

Proof 3. Note that

$$\begin{aligned} T(x_1, x_2, x_3, \dots) + T(y_1, y_2, y_3, \dots) &= (x_2, x_3, \dots) + (y_2, y_3, \dots) \\ &= (x_2 + y_2, x_3 + y_3, \dots) \\ &= T(x_1 + y_1, x_2 + y_2, x_3 + y_3, \dots). \end{aligned}$$

Therefore, T is additive. Homogeneity of T is trivial and thus omitted here. ■

6. From \mathbb{F}^n to \mathbb{F}^m , we define $T \in \mathcal{L}(\mathbb{F}^n, \mathbb{F}^m)$ as

$$T(x_1, \dots, x_n) = (A_{1,1}x_1 + \dots + A_{1,n}x_n, \dots, A_{m,1}x_1 + \dots + A_{m,n}x_n),$$

where $A_{j,k} \in \mathbb{F} \quad \forall j = 1, \dots, m \text{ and } k = 1, \dots, n$.

Theorem 3.1.4

Suppose v_1, \dots, v_n is a basis of V and $w_1, \dots, w_n \in W$. Then, \exists a unique linear map $T : V \rightarrow W$ s.t. $Tv_j = w_j \quad \forall j = 1, \dots, n$.

Remark. If T in Theorem 3.1.1 is a linear mapping, we should have

1. $T(v_1 + \cdots + v_n) = Tv_1 + \cdots + Tv_n = w_1 + \cdots + w_n$, by additivity of T , and
2. $T(\lambda_j v_j) = \lambda_j Tv_j$, by homogeneity of T .

Combine the two properties, we should have

$$T(\lambda_1 v_1 + \cdots + \lambda_n v_n) = \lambda_1 Tv_1 + \cdots = \lambda_n Tv_n = \lambda_1 w_1 + \cdots + \lambda_n w_n.$$

This remark will be very helpful in our following proof of the theorem.

Proof 4. Let's define $T : V \rightarrow W$ by $T(c_1 v_1 + \cdots + c_n v_n) = c_1 w_1 + \cdots + c_n w_n$, where c_1, \dots, c_n are arbitrary elements of \mathbb{F} . Now, we want to show that T is a linear mapping.

Suppose $u, v \in V$, $u = a_1 v_1 + \cdots + a_n v_n$, and $v = c_1 v_1 + \cdots + c_n v_n$. Then, we have

$$\begin{aligned} T(u + v) &= T((a_1 + c_1)v_1 + \cdots + (a_n + c_n)v_n) \\ &= (a_1 + c_1)w_1 + \cdots + (a_n + c_n)w_n \\ &= (a_1 w_1 + \cdots + a_n w_n) + (c_1 w_1 + \cdots + c_n w_n) \\ &= Tu + Tv. \quad \square \end{aligned}$$

Now, we want to show T has homogeneity. Suppose $\lambda \in \mathbb{F}$. Then, we know

$$\begin{aligned} T(\lambda v) &= T(\lambda c_1 v_1 + \cdots + \lambda c_n v_n) \\ &= \lambda c_1 w_1 + \cdots + \lambda c_n w_n \\ &= \lambda(c_1 w_1 + \cdots + c_n w_n) \\ &= \lambda Tv. \quad \square \end{aligned}$$

Also, we want to show that this T satisfy the condition the theorem is asking (i.e., $Tv_j = w_j$). Note that when $c_j = 0$ and other c 's equal 0, we will get $Tv_j = w_j$. \square

Finally, we will prove the uniqueness of this T . Suppose that $T' \in \mathcal{L}(V, W)$ and $T'v_j = w_j$. Let $c_1, \dots, c_n \in \mathbb{F}$. Then, $T'(c_j v_j) = c_j w_j$. So, we know that $T'(c_1 v_1 + \cdots + c_n v_n) = c_1 w_1 + \cdots + c_n w_n$. However, by definition, we know $c_1 w_1 + \cdots + c_n w_n = T(c_1 v_1 + \cdots + c_n v_n)$. So, we can conclude that $T'(c_1 v_1 + \cdots + c_n v_n) = T(c_1 v_1 + \cdots + c_n v_n)$. Thus, $T' = T$, and thus the T we defined above is unique in $\mathcal{L}(V, W)$. \blacksquare

Definition 3.1.5 (Addition and Scalar Multiplication on $\mathcal{L}(V, W)$). Suppose $S, T \in \mathcal{L}(V, W)$ and $\lambda \in \mathbb{F}$. Then, the *addition* is defined as $(S + T)(v) := Sv + Tv$, and the *scalar multiplication* is defined as $(\lambda T)(v) := \lambda(Tv) \quad \forall v \in V$.

Theorem 3.1.6

$\mathcal{L}(V, W)$ is a vector space.

Proof 5.

1. additive identity: Note that the zero-mapping $0 \in \mathcal{L}(V, W)$ satisfies the following equation:

$$(0 + T)(v) = 0v + Tv = 0 + Tv = Tv. \quad \square$$

2. commutativity: Note that

$$(S + T)(v) = Sv + Tv = Tv + Sv = (T + S)(v). \quad \square$$

3. associativity: Let $S, T, R \in \mathcal{L}(V, W)$. Then,

$$\begin{aligned} ((S + T) + R)(v) &= (S + T)(v) + Rv = Sv + Tv + Rv \\ &= Sv + (Tv + Rv) \\ &= Sv + (T + R)(v) \\ &= (S + (T + R))(v). \end{aligned}$$

Let $a, b \in \mathbb{F}$. Then,

$$((ab)T)(v) = T(abv) = T(a(bv)) = aT(bv) = (a(bT))(v). \quad \square$$

4. multiplicative identity: Note we have $1 \in \mathbb{F}$ s.t.

$$(1 \cdot T)(v) = T(1 \cdot v) = Tv. \quad \square$$

5. additive inverse: Note that

$$(T + (-T))(v) = Tv + (-T)(v) = Tv + T(-v) = T(v - v) = T0 = 0. \quad \square$$

6. distributivity: Note that

$$a(T + S)(v) = a(Tv + Sv) = aTv + aSv,$$

and

$$(a + b)Tv = T((a + b)v) = T(av + bv) = T(av) + T(bv) = aTv + bTv.$$

■

Definition 3.1.7 (Product of Linear Maps). If $T \in \mathcal{L}(U, V)$ and $S \in \mathcal{L}(V, W)$, then the *product* $ST \in \mathcal{L}(U, W)$ is defined by $(ST)(u) = S(Tu) \quad \forall u \in U$.

Remark. Compare this definition with composite functions. ST is only defined when T maps into the domain of S .

Theorem 3.1.8 Algebraic Properties of Products of Linear Maps

1. associativity: $(T_1 T_2) T_3 = T_1 (T_2 T_3)$.
2. identity: $TI = IT = T$, where I is the identity mapping
3. distributive properties: $(S_1 + S_2)T = S_1 T + S_2 T$ and $S(T_1 + T_2) = ST_1 + ST_2$.

Proof 6. First, we want to show the associativity. Note that

$$[(T_1 T_2) T_3](v) = (T_1 T_2)(T_3 v) = (T_1)(T_2(T_3 v)) = (T_1)[(T_2 T_3)(v)]. \quad \square$$

Then, we want to show the identity. This proof can be done using the following diagram:

$$\begin{array}{ccc}
 V & \xrightarrow{T} & W \\
 I_V \uparrow & & \downarrow I_W \\
 V & & W
 \end{array}
 \quad \square$$

Finally, we will show the distributive properties. Note that

$$\begin{aligned}
 [(S_1 + S_2)T](v) &= (S_1 + S_2)(Tv) = S_1(Tv) + S_2(Tv) \\
 &= (S_1T)(v) + (S_2T)(v) \\
 &= (S_1T + S_2T)(v).
 \end{aligned}$$

Similarly, we can show

$$\begin{aligned}
 [S(T_1 + T_2)](v) &= S[(T_1 + T_2)(v)] = S(T_1v + T_2v) \\
 &= S(T_1v) + S(T_2v) \\
 &= (ST_1)(v) + (ST_2)(v) \\
 &= (ST_1 + ST_2)(v).
 \end{aligned}$$

■

Example 3.1.9 Suppose $D \in \mathcal{L}(\mathcal{P}(\mathbb{R}), \mathcal{P}(\mathbb{R}))$ is the differentiation map, and $T \in \mathcal{L}(\mathcal{P}(\mathbb{R}), \mathcal{P}(\mathbb{R}))$ be defined by $(Tp)(x) = x^2p(x)$. Show that $DT \neq TD$.

Proof 7. Note that $(DT)p = D(Tp) = D(x^2p(x)) = 2xp(x) + x^2p'(x)$. Similarly, we can compute a general formula for TD : $(TD)p = T(Dp) = T(p') = x^2p'(x)$. Since $2xp(x) + x^2p'(x) \neq x^2p'(x)$, we know $DT \neq TD$. ■

Theorem 3.1.10

Let $T \in \mathcal{L}(V, W)$, then $T(0) = 0$.

Proof 8. Since $T(0) = T(0 + 0) = T(0) + T(0)$, we know $0 = T(0)$, or $T(0) = 0$. ■

Corollary 3.1.11 If $T(0) \neq 0$, then $T \notin \mathcal{L}(V, W)$.

3.2 Null Spaces and Ranges

Definition 3.2.1 (Null Space/Kernel). For $T \in \mathcal{L}(V, W)$, the *null space* of T , denoted $\text{null } T$, is the subset of V consisting of those vectors that T maps to 0: $\text{null } T = \{v \in V \mid Tv = 0\}$.

Remark. Sometimes, null space of T is also called the kernal of T , denoted as $\ker T$.

Example 3.2.2

1. Null space of zero-mapping: Let T be the zero mapping from V to W . Since $Tv = 0 \quad \forall v \in V$, we know $\text{null } T = V$.
2. $D \in \mathcal{L}(\mathcal{P}(\mathbb{R}), \mathcal{P}(\mathbb{R}))$ as $Dp = p'$: $\text{null } D = \{a \mid a \in \mathbb{R}\}$.
3. $T \in \mathcal{L}(\mathbb{F}^\infty, \mathbb{F}^\infty)$ as $T(x_1, x_2, x_3, \dots) = (x_2, x_3, \dots)$: $\text{null } T = \{(a, 0, 0, \dots) \mid a \in \mathbb{F}\}$.

Theorem 3.2.3

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

Proof 1.

1. Note that $T(0) = 0$, so $0 \in \text{null } T$. \square
2. Suppose $u, v \in \text{null } T$. Then, $Tu = Tv = 0$. So, $T(u + v) = Tu + Tv = 0 + 0 = 0$. Hence, $u + v \in \text{null } T$. \square
3. Suppose $u \in \text{null } T$ and $\lambda \in \mathbb{F}$. Then, $Tu = 0$. So, $T(\lambda u) = \lambda Tu = \lambda \cdot 0 = 0$. Therefore, $\lambda u \in \text{null } T$. \blacksquare

Definition 3.2.4 (Injective/Injection). A function $T : V \rightarrow W$ is called *injective* if $Tu = Tv$ implies $u = v$.

Remark. Sometimes, the contrapositive will be much more helpful: T is injective if $u \neq v$, then $Tu \neq Tv$.

Theorem 3.2.5

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

Proof 2.

(\Rightarrow) Suppose T is an injective. We've already known that $\{0\} \subseteq \text{null } T$. Then, we need to show $\text{null } T \subseteq \{0\}$. Suppose $v \in \text{null } T$, then $Tv = 0$. However, since T is an injection, and $Tv = T0 = 0$, then we have $v = 0$. So, $\text{null } T \subseteq \{0\}$. Therefore, it's sufficient to say $\text{null } T = \{0\}$. \square

(\Leftarrow) Suppose $\text{null } T = \{0\}$. Suppose $u, v \in V$ and $Tu = Tv$. Then, $Tu - Tv = T(u - v) = 0$. Hence, $u - v \in \text{null } T$. By $\text{null } T = \{0\}$, we know $u - v = 0$, so $u = v$. Then, T is an injection. \blacksquare

Definition 3.2.6 (Range/Image). For $T \in \mathcal{L}(V, W)$, the range of T is the subset of W consisting of those vectors that are of the form Tv for some $v \in V$: $\text{range } T = \{Tv \mid v \in V\}$.

Theorem 3.2.7

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

Proof 3.

1. Since $T(0) = 0$, we know $0 \in \text{range } T$. \square
2. Suppose $w_1, w_2 \in \text{range } T$. Then, $\exists v_1, v_2 \in V$ s.t. $Tv_1 = w_1$ and $Tv_2 = w_2$. Then, $w_1 + w_2 = Tv_1 + Tv_2 = T(v_1 + v_2)$. Since $v_1 + v_2 \in V$, we have $w_1 + w_2 = T(v_1 + v_2) \in \text{range } T$. \square
3. Suppose $w \in \text{range } T$ and $\lambda \in \mathbb{F}$. Then, $\exists v \in V$ s.t. $w = Tv$. So, $\lambda w = \lambda(Tv) = T(\lambda v)$. Since $\lambda v \in V$, $\lambda w = T(\lambda v) \in \text{range } T$. \blacksquare

Definition 3.2.8 (Surjective/Surjection). A function $T : V \rightarrow W$ is called *surjective* if $\text{range } T = W$.

Remark. A function $T : V \rightarrow W$ is called a *bijection*, or is *bijjective*, if it is both injective and surjective.

Theorem 3.2.9 Fundamental Theorem of Linear Maps

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

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

Proof 4. Let u_1, \dots, u_m be a basis of $\text{null } T$. Then, $\dim \text{null } T = m$. By Theorem 3.2.3, we know $\text{null } T$ is a basis of V , so we can extend the basis to a basis of V : $u_1, \dots, u_m, v_1, \dots, v_n$. Thus, $\dim V = m + n$. WTS: $\dim \text{range } T = n$. Further WTS: Tv_1, \dots, Tv_n is a basis of $\text{range } T$.

Suppose $v \in V$. Then

$$v = a_1u_1 + \dots + a_mu_m + b_1v_1 + \dots + b_nv_n.$$

Since $u_1, \dots, u_m \in \text{null } T$, we know $Tu_1, \dots, Tu_m = 0$. Therefore,

$$Tv = a_1Tu_1 + \dots + a_mTu_m + b_1Tv_1 + \dots + b_nTv_n = b_1Tv_1 + \dots + b_nTv_n.$$

Hence, $\text{span}(Tv_1, \dots, Tv_n) = \text{range } T$, and thus $\text{range } T$ is f - d . Now, WTS: Tv_1, \dots, Tv_n is L.I..

Consider $c_1Tv_1 + \dots + c_nTv_n = 0$. Then, $T(c_1v_1 + \dots + c_nv_n) = 0$. Hence, $c_1v_1 + \dots + c_nv_n \in \text{null } T$. Since u_1, \dots, u_m is a basis of $\text{null } T$, we know

$$c_1v_1 + \dots + c_nv_n = d_1u_1 + \dots + d_mu_m \quad f.s. d_i \in \mathbb{F}.$$

So,

$$c_1v_1 + \dots + c_nv_n - d_1u_1 - \dots - d_mu_m = 0. \quad (8)$$

However, by assumption, we know $v_1, \dots, v_n, u_1, \dots, u_m$ is a basis of V , and thus it is L.I.. So, the only way to make Equation (8) hold is by taking $c_1 = \dots = c_n = -d_1 = \dots = -d_m = 0$. Therefore, we've shown Tv_1, \dots, Tv_n is L.I., and thus is a basis of $\text{range } T$. Then, $\dim \text{range } T = n$.

So, we've shown that $\dim \text{null } T + \dim \text{range } T = m + n = \dim V$. \blacksquare

Theorem 3.2.10

Suppose V and W are f - d vector spaces s.t. $\dim V > \dim W$. Then, no linear map from V to W is injective.

Proof 5. Let $T \in \mathcal{L}(V, W)$. By the Fundamental Theorem of Linear Maps, we have $\dim V = \dim \text{null } T + \dim \text{range } T$. Then, we know

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

This implies that $\text{null } T \neq \{0\}$. So, T is not injective by Theorem 3.2.5. ■

Theorem 3.2.11

Suppose V and W are f - d vector space s.t. $\dim V < \dim W$. Then, no linear map from V to W is surjective.

Proof 6. We know

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

Then, T cannot be surjective by definition. ■

Example 3.2.12 Solving Linear Systems Using Linear Maps I

For a homogenous system of linear equations,

$$\begin{cases} A_{1,1}x_1 + \cdots + A_{1,n}x_n = 0 \\ \vdots \\ A_{m,1}x_1 + \cdots + A_{m,n}x_n = 0 \end{cases},$$

where $A_{j,k} \in \mathbb{F}$ and $(x_1, \dots, x_n) \in \mathbb{F}^n$, we can defined a linear map $T : \mathbb{F}^n \rightarrow \mathbb{F}^m$ as

$$T(x_1, \dots, x_n) = \left(\sum_{k=1}^n A_{1,k}x_k, \dots, \sum_{k=1}^n A_{m,k}x_k \right).$$

Apparently, $(x_1, \dots, x_n) = 0$ is a solution to the system, but the question is “If there are any non-zero solutions for this linear system?”

Theorem 3.2.13

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

Proof 7. Suppose $T \in \mathcal{L}(V, W)$. Then, $\dim V = n$ and $\dim W = m$. Suppose $n > m$. So, $\dim V > \dim W$. By the Theorem 3.2.5, we know T is not injective. ■

Example 3.2.14 Solving Linear Systems Using Linear Maps II

For an inhomogeneous system of linear equations

$$\begin{cases} \sum_{k=1}^n A_{1,k}x_k = c_1 \\ \vdots \\ \sum_{k=1}^n A_{m,k}x_k = c_m \end{cases},$$

where $A_{j,k} \in \mathbb{F}$ and $(c_1, \dots, c_m) \in \mathbb{F}^m$ and $(x_1, \dots, x_n) \in \mathbb{F}^n$, we can define $T : \mathbb{F}^n \rightarrow \mathbb{F}^m$ by

$$T(x_1, \dots, x_n) = \left(\sum_{k=1}^n A_{1,k}x_k, \dots, \sum_{k=1}^n A_{m,k}x_k \right).$$

However, in this case, $(x_1, \dots, x_n) = 0$ may not be a solution to the system.

Theorem 3.2.15

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

Proof 8. Suppose $T \in \mathcal{L}(V, W)$. So, $\dim V = n$ and $\dim W = m$. Suppose $n < m$. Then, $\dim V < \dim W$. By Theorem 3.2.11, we know T is not surjective. ■

3.3 Matrices

Definition 3.3.1 (Matrix). Let $m, n \in \mathbb{Z}^+$. An m -by- n *matrix* A is a rectangular array of elements of \mathbb{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 .

Definition 3.3.2 (Matrix of a Linear Map). Suppose $T \in \mathcal{L}(V, W)$ and v_1, \dots, v_n is a basis of V and w_1, \dots, w_m is a basis of W . The *matrix of T* with respect to these bases is the $m \times n$ matrix $\mathcal{M}(T)$ whose $A_{j,k}$ are defined by

$$Tv_k = A_{1,k}w_1 + \cdots + A_{m,k}w_m.$$

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.

Example 3.3.3 Suppose $T \in \mathcal{L}(\mathbb{F}^2, \mathbb{F}^3)$ is defined by $T(x, y) = (x + 3y, 2x + 5y, 7x + 9y)$. Find the matrix of T with respect to the standard bases of \mathbb{F}^2 and \mathbb{F}^3 .

Answer 1.

Note that $T(1, 0) = (1, 2, 7)$ and $T(0, 1) = (3, 5, 9)$. Then,

$$\mathcal{M}(T) = \begin{pmatrix} 1 & 3 \\ 2 & 5 \\ 7 & 9 \end{pmatrix}.$$

□

Example 3.3.4 Suppose $D \in \mathcal{L}(\mathcal{P}_3(\mathbb{R}), \mathcal{P}_2(\mathbb{R}))$ is the differentiation map defined by $Dp = p'$. Find the matrix of D with respect to the standard bases of $\mathcal{P}_3(\mathbb{R})$ and $\mathcal{P}_2(\mathbb{R})$.

Answer 2.

Standard bases of $\mathcal{P}_3(\mathbb{R})$: $1, x, x^2, x^3$. Standard bases of $\mathcal{P}_2(\mathbb{R})$: $1, x, x^2$. Since $(x^n)' = nx^{n-1}$, so we have

$$D(1) = 0 = 0 \cdot 1 + 0 \cdot x + 0 \cdot x^2$$

$$D(x) = 1 = 1 \cdot 1 + 0 \cdot x + 0 \cdot x^2$$

$$D(x^2) = 2x = 0 \cdot 1 + 2 \cdot x + 0 \cdot x^2$$

$$D(x^3) = 3x^2 = 0 \cdot 1 + 0 \cdot x + 3 \cdot x^2$$

So, we have

$$\mathcal{M}(D) = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 3 \end{pmatrix}.$$

□

Definition 3.3.5 (Matrix Addition). The *sum of two matrices of the same size* is the matrix obtained by

adding corresponding entries in the matrices:

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

Theorem 3.3.6

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

Proof 3. Let v_1, \dots, v_n be a basis of V and w_1, \dots, w_m be a basis of W . Suppose $\mathcal{M}(S) = A$ and $\mathcal{M}(T) = C$. Then, if $1 \leq k \leq n$, we have

$$\begin{aligned} (S + T)v_k &= Sv_k + Tv_k \\ &= (A_{1,k}w_1 + \cdots + A_{m,k}w_m) + (C_{1,k}w_1 + \cdots + C_{m,k}w_m) \\ &= (A_{1,k} + C_{1,k})w_1 + \cdots + (A_{m,k} + C_{m,k})w_m. \end{aligned}$$

Hence, we have $\mathcal{M}(S + T) = \mathcal{M}(S) + \mathcal{M}(T)$. ■

Definition 3.3.7 (Scalar Multiplication of a Matrix). The *product of a scalar and a matrix* is the matrix obtained by multiplying each entry in the matrix by the scalar:

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

In other words, $(\lambda A)_{j,k} = \lambda A_{j,k}$.

Theorem 3.3.8

Suppose $\lambda \in \mathbb{F}$ and $T \in \mathcal{L}(V, W)$. Then, $\mathcal{M}(\lambda T) = \lambda \mathcal{M}(T)$.

Proof 4. Let v_1, \dots, v_n be a basis of V and $\mathcal{M}(T) = A$. When $1 \leq k \leq n$, note that

$$\begin{aligned} (\lambda T)v_k &= \lambda(Tv_k) \\ &= \lambda(A_{1,k}w_1 + \cdots + A_{m,k}w_m) \\ &= (\lambda A_{1,k})w_1 + \cdots + (\lambda A_{m,k})w_m. \end{aligned}$$

So, $\mathcal{M}(\lambda T) = \lambda \mathcal{M}(T)$. ■

Notation 3.3.9. $\mathbb{F}^{m,n} :=$ the set of all $m \times n$ matrices with entries in \mathbb{F} .

Theorem 3.3.10

Suppose $m, n \in \mathbb{Z}^+$. With addition and scalar multiplication defined above, $\mathbb{F}^{m,n}$ is a vector space and $\dim \mathbb{F}^{m,n} = mn$.

Proof 5. It is trivial to prove $\mathbb{F}^{m,n}$ is a vector space. □

Define $A_{j,k}$ as the matrix with 1 on its j^{th} row, k^{th} column and 0 elsewhere. Then, we can see that $A_{j,k}$ for $j = 1, \dots, m$ and $k = 1, \dots, n$ is a basis for $\mathbb{F}^{m,n}$. So, $\dim \mathbb{F}^{m,n} = m \cdot n$. ■

Definition 3.3.11 (Matrix Multiplication). Suppose A is an $m \times n$ matrix and C is an $n \times p$ matrix. Then,

AC is defined to be the $m \times p$ matrix whose entry in row j , column k is given by

$$(AC)_{j,k} = \sum_{r=1}^n A_{j,r} C_{r,k}.$$

Remark. Matrix multiplication is not commutative. i.e., $AC \neq CA$. However, it is distributive and associative.

Theorem 3.3.12

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

Notation 3.3.13. Suppose A is an $m \times n$ matrix.

1. If $1 \leq j \leq m$, then $A_{j,\cdot}$ denotes the $1 \times n$ matrix consisting of row j of A .
2. If $1 \leq k \leq n$, then $A_{\cdot,k}$ denotes the $m \times 1$ matrix consisting of column k of A .

In other words,

$$A = \begin{pmatrix} A_{1,1} & \cdots & A_{1,n} \\ \vdots & & \vdots \\ A_{m,1} & \cdots & A_{m,n} \end{pmatrix}; \quad A_{j,\cdot} = (A_{j,1} \quad \cdots \quad A_{j,n}) \in \mathbb{F}^{1,n}; \quad A_{\cdot,k} = \begin{pmatrix} A_{1,k} \\ \vdots \\ A_{m,k} \end{pmatrix} \in \mathbb{F}^{m,1}.$$

Theorem 3.3.14 Practical Interpretations of Matrix Multiplication

1. Suppose A is an $m \times n$ matrix and C is an $n \times p$ matrix. Then, $(AC)_{j,k} = A_{j,\cdot} C_{\cdot,k}$ for $1 \leq j \leq m$ and $1 \leq k \leq p$.
2. Suppose A is an $m \times n$ matrix and C is an $n \times p$ matrix. Then, $(AC)_{\cdot,k} = AC_{\cdot,k}$ for $1 \leq k \leq p$.

3. Suppose A is an $m \times n$ matrix and $C = \begin{pmatrix} c_1 \\ \vdots \\ c_n \end{pmatrix}$ is an $n \times 1$ matrix. Then,

$$AC = c_1 A_{\cdot,1} + \cdots + c_n A_{\cdot,n}.$$

In other words, AC is a linear combination of the columns of A , with the scalars that multiply the columns coming from C .

Example 3.3.15

$$\begin{pmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \end{pmatrix} \begin{pmatrix} 5 \\ 1 \end{pmatrix} = 5 \begin{pmatrix} 1 \\ 3 \\ 5 \end{pmatrix} + 1 \begin{pmatrix} 2 \\ 4 \\ 6 \end{pmatrix} = \begin{pmatrix} 7 \\ 19 \\ 31 \end{pmatrix}.$$

3.4 Invertibility and Isomorphic Vector Spaces

Definition 3.4.1 (Invertible). A linear map $T \in \mathcal{L}(V, W)$ is called *invertible* if \exists a linear map $S \in \mathcal{L}(W, V)$ s.t. ST equals the identity map on V and TS equals the identity map on W .

Definition 3.4.2 (Inverse). A linear map $S \in \mathcal{L}(W, V)$ satisfying $ST = I$ and $TS = I$ is called an *inverse* of T .

Theorem 3.4.3

An invertible linear map has a unique inverse.

Proof 1. Suppose $T \in \mathcal{L}(V, W)$ is invertible. Let S_1 and S_2 be inverses of T . Then,

$$S_1 = S_1 I = S_1 (TS_2) = (S_1 T) S_2 = I S_2 = S_2.$$

Thus, $S_1 = S_2$, and so inverse is unique. ■

Notation 3.4.4. If T is invertible, then its inverse is denoted by T^{-1} .

Theorem 3.4.5

A linear map is invertible if and only if it is injective and surjective.

Proof 2.

(\Rightarrow) Let $T \in \mathcal{L}(V, W)$ be invertible. Then, $TT^{-1} = I_W$ and $T^{-1}T = I_V$. Let $Tv = 0$. Note that $(T^{-1}T)v = 0$, so $Iv = 0$ and thus $v = 0$. Therefore, $\text{null } T = \{0\}$, and so T is an injection.

To show T is surjective, suppose $w \in W$. Note that since $T^{-1} \in \mathcal{L}(W, V)$, $T^{-1}w \in V$. So,

$$T(T^{-1}w) = (TT^{-1})w = I_W w = w \in W.$$

Therefore, $T^{-1}w$ is the $v \in V$ we intend to find. Hence, T is also a surjection. □

(\Leftarrow) Let T be surjective and injective. For $w \in W$, define $Sw \in V$ s.t. $T(Sw) = w$. So, we know Sw is unique. Since $(T \circ S)w = w$, we know $(T \circ S) = I_W$. Consider $(S \circ T)v = S(Tv)$, we have $T(S(Tv)) = Tv$, by definition of S . Since T is injective, we know $S(Tv) = v$. So, $(S \circ T)v = v$, and thus $ST = I_V$. Therefore T is invertible.

Now, we want to show S is a linear map. Let $w_1, w_2 \in W$, then

$$T(S(w_1 + w_2)) = (TS)(w_1 + w_2) = I_W(w_1 + w_2) = w_1 + w_2.$$

By definition, $w_1 + w_2 = T(Sw_1) + T(Sw_2) = T(Sw_1 + Sw_2)$. So, $T(S(w_1 + w_2)) = T(Sw_1 + Sw_2)$. By T is an injection, we have $S(w_1 + w_2) = Sw_1 + Sw_2$. So, S is additive. Further consider

$$T(S(\lambda w)) = \lambda w = \lambda(T(Sw)) = T(\lambda Sw)$$

for some $w \in W$. Again, since T is injective, $S(\lambda w) = \lambda Sw$. So, S has homogeneity. Then, S is a linear map. ■

Definition 3.4.6 (Isomorphism). An *isomorphism* is an invertible linear map.

Definition 3.4.7 (Isomorphic). Two vector spaces are called *isomorphic* if there is an isomorphism from one vector space onto the other one.

Notation 3.4.8. If two vector spaces V and W are isomorphic, we denote them as $V \cong W$.

Theorem 3.4.9

Suppose V and W are f -d vector spaces, then $V \cong W$ if and only if $\dim V = \dim W$.

Proof 3.

(\Rightarrow) Suppose $V \cong W$. By Fundamental Theorem of Linear Maps, we know

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

Since $V \cong W$, T is invertible and thus is injective and surjective. So, $\dim \text{null } T = 0$ and $\dim \text{range } T = \dim W$. Therefore, $\dim V = 0 + \dim W = \dim W$. \square

(\Leftarrow) Suppose $\dim V = \dim W$. Suppose v_1, \dots, v_n and w_1, \dots, w_n are bases of V and W , respectively. Then, $\dim V = \dim W = n$. Here, we want to define a bijection between V and W . Let T be defined as $Tv_i = w_i$ ($i = 1, \dots, n$).

Let $Tv = 0$. Then, $T(a_1v_1 + \dots + a_nv_n) = 0$. So, by definition, $a_1w_1 + \dots + a_nw_n = 0$. Since w_1, \dots, w_n is a basis, we have $a_1 = \dots = a_n = 0$. So, $\text{null } T = \{0\}$, and thus T is an injection.

Let $w \in W$ be any vector. Then, we know $w = c_1w_1 + \dots + c_nw_n$. Note that, by definition of T , we have $T(c_1v_1 + \dots + c_nv_n) = c_1w_1 + \dots + c_nw_n$. Hence, $\forall w \in W, \exists v = c_1v_1 + \dots + c_nv_n \in V$ s.t. $Tv = w$. Therefore, T is a surjection.

Finally, it is trivial to show that T is indeed a linear map, and so the proof is complete. \blacksquare

Theorem 3.4.10

Suppose v_1, \dots, v_n is a basis of V and w_1, \dots, w_m is a basis of W . then, \mathcal{M} is an isomorphism between $\mathcal{L}(V, W)$ and $\mathbb{F}^{m,n}$.

Proof 4. We already know \mathcal{M} is linear, so we just need to show \mathcal{M} is a bijection.

To prove \mathcal{M} is injective, consider $\mathcal{M}(T) = 0$ for some $T \in \mathcal{L}(V, W)$. So, we get $Tv_k = 0$. Since v_1, \dots, v_n is a basis of V , we know $Tv = 0 \quad \forall v \in V$. Then, T is the zero-mapping, or $T = 0$. Therefore, $\text{null } \mathcal{M} = \{0\}$.

To show \mathcal{M} is surjective, suppose $A \in \mathbb{F}^{m,n}$. Let T be a linear map from V to W s.t.

$$Tv_k = \sum_{j=1}^m A_{j,k} w_j, \quad k = 1, \dots, n.$$

Obviously, $\mathcal{M}(T) = A$, and thus $\text{range } \mathcal{M} = \mathbb{F}^{m,n}$. So, \mathcal{M} is also a surjection. \blacksquare

Theorem 3.4.11

Suppose V and W are f -d. Then, $\mathcal{L}(V, W)$ is f -d and $\dim \mathcal{L}(V, W) = (\dim V)(\dim W)$.

Proof 5. By Theorem 3.4.10 and Theorem 3.4.9, we know $\dim \mathcal{L}(V, W) = \dim \mathbb{F}^{m,n}$. Further by Theorem 3.3.10, we know $\dim \mathbb{F}^{m,n} = (m)(n)$. As $\dim V = n$ and $\dim W = m$, so we have

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

Definition 3.4.12 (Matrix of a Vector, $\mathcal{M}(v)$). Suppose $v \in V$ and v_1, \dots, v_n is a basis of V . The *matrix*

of v with respect to this basis is the $n \times 1$ matrix

$$\mathcal{M}(v) = \begin{pmatrix} c_1 \\ \vdots \\ c_n \end{pmatrix},$$

where c_1, \dots, c_n are scalars s.t. $v = c_1v_1 + \dots + c_nv_n$.

Theorem 3.4.13 $\mathcal{M}(T)_{\cdot,k} = \mathcal{M}(v_k)$

Suppose $T \in \mathcal{L}(V, W)$ and v_1, \dots, v_n is a basis of V and w_1, \dots, w_m is a basis of W . Let $1 \leq k \leq n$. Then, the k^{th} column of $\mathcal{M}(T)$, which is denoted by $\mathcal{M}(T)_{\cdot,k}$, equals $\mathcal{M}(v_k)$.

Proof 6. This theorem is an immediate result by definitions of matrix of a linear mapping and a vector. ■

Theorem 3.4.14

Suppose $T \in \mathcal{L}(V, W)$ and $v \in V$. Suppose v_1, \dots, v_n is a basis of V and w_1, \dots, w_m is a basis of W . Then, $\mathcal{M}(Tv) = \mathcal{M}(T)\mathcal{M}(v)$.

Proof 7. Note that $v = c_1v_1 + \dots + c_nv_n$, so we have $Tv = c_1Tv_1 + \dots + c_nTv_n$. So, by Theorem 3.4.13, we know

$$\begin{aligned} \mathcal{M}(Tv) &= c_1\mathcal{M}(Tv_1) + \dots + c_n\mathcal{M}(Tv_n) \\ &= c_1\mathcal{M}(T)_{\cdot,1} + \dots + c_n\mathcal{M}(T)_{\cdot,n} \\ &= \mathcal{M}(T)\mathcal{M}(v). \end{aligned}$$

The final equality holds due to our interpretation of matrix multiplication as column linear combinations (Theorem 3.3.14(3)) ■

Remark. $\mathcal{M} : \mathbb{F}^n \rightarrow \mathbb{F}^{n,1}$ is an isomorphism:

$$v = c_1v_1 + \dots + c_nv_n \mapsto \begin{pmatrix} c_1 \\ \vdots \\ c_n \end{pmatrix}.$$

Proof 8. Suppose $\mathcal{M}(v) = 0$: $\mathcal{M}(c_1v_1 + \dots + c_nv_n) = 0$. So, we have $c_1w_1 + \dots + c_nw_n = 0$. Since w_1, \dots, w_n is a basis, $c_1 = \dots = c_n = 0$. So, $v = 0$. Therefore, $\text{null } \mathcal{M} = \{0\}$, and so \mathcal{M} is injective. □

Now, prove \mathcal{M} is surjective. Note that $\forall \begin{pmatrix} c_1 \\ \vdots \\ c_n \end{pmatrix}$, we have $\mathcal{M}(c_1v_1 + \dots + c_nv_n) = \begin{pmatrix} c_1 \\ \vdots \\ c_n \end{pmatrix}$. So, \mathcal{M} is a surjection. □

Finally, it's trivial to prove \mathcal{M} is a linear map. □

Since \mathcal{M} is both surjective and injective, \mathcal{M} is an isomorphism. ■

Definition 3.4.15 (Operator). A linear map from a vector space to itself is called an *operator*.

Notation 3.4.16. The notation $\mathcal{L}(V)$ denotes the set of all operators on V . So, $\mathcal{L}(v) = \mathcal{L}(V, V)$.

Theorem 3.4.17

Suppose V is f - d and $T \in \mathcal{L}(V)$. Then, the following are equivalent: (a) T is invertible; (b) T is injective; and (c) T is surjective.

Proof 9.

1. Clearly (a) implies (b). \square

2. Suppose (b): T is injective. So, $\text{null } T = \{0\}$. Then, by Fundamental Theorem of Linear Maps, we know

$$\dim V = \dim \text{null } T + \dim \text{range } T = 0 + \dim \text{range } T.$$

Since $\dim \text{range } T = \dim V$, we know T is surjective. \square

3. Suppose (c): T is surjective. So, $\text{range } T = V$. Then, by Fundamental Theorem of Linear maps, we have

$$\dim \text{null } T = \dim V - \dim \text{range } T = 0.$$

So, $\text{null } T = \{0\}$, and thus T is injective. Since T is surjective and injective, T is invertible. ■

Example 3.4.18 Show that for each polynomial $q \in \mathcal{P}(\mathbb{R})$, there exists a polynomial $p \in \mathcal{P}(\mathbb{F})$ such that $((x^2 + 5x + 7)p)'' = q$.

Proof 10. We know that every non-zero polynomial must have a degree of m . So, we can think of this problem under $\mathcal{P}_m(\mathbb{R})$. Note that

$$((x^2 + 5x + 7)p)'' = 2p + (4x + 10)p' + (x^2 + 5x + 7)p'' = q.$$

Therefore, the degree of p and q should be the same. Define $T : \mathcal{P}_m(\mathbb{R}) \rightarrow \mathcal{P}_m(\mathbb{R})$ as

$$Tp = ((x^2 + 5x + 7)p)'.$$

Then, T is an operator on $\mathcal{P}_m(\mathbb{R})$. Consider $Tp = 0$. We have $ax + b = (x^2 + 5x + 7)p$. Note that only when $p = 0$, the equation above holds. So, it must be that $p = 0$ when $Tp = 0$. That is, $\text{null } T = \{0\}$, and so T is injective. By Theorem 3.4.18, we know T is also surjective, and so our proof is complete. ■

3.5 Duality

Definition 3.5.1 (Linear Functional). A *linear functional* on V is a linear map from V to \mathbb{F} . That is, a linear functional is an element of $\mathcal{L}(V, \mathbb{F})$.

Example 3.5.2

1. Fix $(c_1, \dots, c_n) \in \mathbb{F}^n$. Define $\varphi : \mathbb{F}^n \rightarrow \mathbb{F}$ by $\varphi(x_1, \dots, x_n) = c_1x_1 + \dots + c_nx_n$. Then, φ is a linear functional on \mathbb{F}^n .
2. Define $\varphi : \mathcal{P}(\mathbb{R}) \rightarrow \mathbb{R}$ as $\varphi(p) = 3p''(5) + 7p(4)$.
3. Define $\varphi : \mathcal{P}(\mathbb{R}) \rightarrow \mathbb{R}$ as $\varphi(p) = \int_0^1 p(x)dx$.

Definition 3.5.3 (Dual Space/ V'/V^*). The *dual space* of V , denoted as V' , is the vector space of all linear functionals on V . In other words, $V' = \mathcal{L}(V, \mathbb{F})$.

Theorem 3.5.4

Suppose V is f - d . Then, V' is also f - d and $\dim V' = \dim V$.

Proof 1. Note that for a general linear map, $\mathcal{L}(V, W) \cong \mathbb{F}^{m,n}$. So, $\mathcal{L}(V, \mathbb{F}) = V' \cong \mathbb{F}^{1,n}$. Hence,

$$\dim V' = \dim \mathbb{F}^{1,n} = 1 \cdot n = n = \dim V.$$

■

Definition 3.5.5 (Dual Basis). If v_1, \dots, v_n is a basis of V , then the *dual basis* of v_1, \dots, v_n is the list $\varphi_1, \dots, \varphi_n$ of elements of V' , where each φ_j is the linear functional on V s.t.

$$\varphi_j(v_k) = \begin{cases} 1 & \text{if } k = j \\ 0 & \text{if } k \neq j \end{cases}.$$

Example 3.5.6 Find the dual basis of $e_1, \dots, e_n \in \mathbb{F}^n$

Answer 2.

$$\begin{array}{cccc} \varphi_1(e_1) = 1 & \varphi_2(e_1) = 0 & \cdots & \varphi_n(e_1) = 0 \\ \varphi_1(e_2) = 0 & \varphi_2(e_2) = 1 & \cdots & \varphi_n(e_2) = 0 \\ \vdots & \vdots & \ddots & \vdots \\ \varphi_1(e_n) = 0 & \varphi_2(e_n) = 0 & \cdots & \varphi_n(e_n) = 1 \end{array}$$

Define φ_j as

$$\varphi_j(x) = \varphi_j(x_1, \dots, x_n) = x_1\varphi_j(e_1) + \dots + x_j\varphi_j(e_j) + \dots + x_n\varphi_j(e_n) = x_j.$$

□

Theorem 3.5.7

Suppose V is f -d. Then, the dual basis of a basis of V is a basis of V' .

Proof3. Suppose v_1, \dots, v_n is a basis of V and $\varphi_1, \dots, \varphi_n$ denotes the dual basis. Since we've shown $\dim V = \dim V'$ in Theorem 3.5.4, we only need to show $\varphi_1, \dots, \varphi_n$ is L.I.. Select $c_1\varphi_1 + \dots + c_n\varphi_n = 0$. Then,

$$(c_1\varphi_1 + \dots + c_n\varphi_n)(v) = 0 \quad \forall v \in V.$$

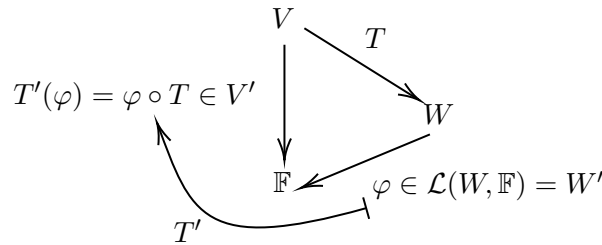
Suppose $v = v_1 + \dots + v_n$, then

$$(c_1\varphi_1 + \dots + c_n\varphi_n)(v_j) = c_j \quad \text{for } j = 1, \dots, n.$$

So, $(c_1\varphi_1 + \dots + c_n\varphi_n)(v) = c_1 + \dots + c_n = 0$. So, it must be that $c_1 = \dots = c_n = 0$. Therefore, $\varphi_1, \dots, \varphi_n$ is L.I. and our proof is complete. ■

Definition 3.5.8 (Dual Map). If $T \in \mathcal{L}(V, W)$, then the *dual map* of T is the linear map $T' \in \mathcal{L}(W', V')$ defined by $T'(\varphi) = \varphi \circ T$ for $\varphi \in W'$.

Remark. The following diagram represents dual map (but not an exact representation).



Also, dual map is a linear map, so it is additive and homogeneous.

1. $T'(\varphi + \psi) = (\varphi + \psi) \circ T = \varphi \circ T + \psi \circ T = T'(\varphi) + T'(\psi)$.
2. $T'(\lambda\varphi) = (\lambda\varphi) \circ T = \lambda(\varphi \circ T) = \lambda T'(\varphi)$.

Example 3.5.9 Suppose $D : \mathcal{P}(\mathbb{R}) \rightarrow \mathcal{P}(\mathbb{R})$ as $Dp = p'$.

1. Define a linear functional $\varphi : \mathcal{P}(\mathbb{R}) \rightarrow \mathbb{R}$ as $\varphi(p) = p(3)$. Find $D'(\varphi)$.

Answer 4.

$$(D'(\varphi))(p) = (\varphi \circ D)(p) = \varphi(Dp) = \varphi(p') = p'(3).$$

□

2. Define $\varphi : \mathcal{P}(\mathbb{R}) \rightarrow \mathbb{R}$, a linear functional, as $\varphi(p) = \int_0^1 p(x) dx$. Find $D'(\varphi)$.

Answer 5.

$$(D'(\varphi))(p) = (\varphi \circ D)(p) = \varphi(Dp) = \varphi(p') = \int_0^1 p'(x) dx = p(1) - p(0).$$

□

Theorem 3.5.10 Algebraic Properties of Dual Maps

1. $(S + T)' = S' + T' \quad \forall S, T \in \mathcal{L}(V, W)$
2. $(\lambda T)' = \lambda T' \quad \forall T \in \mathcal{L}(V, W)$
3. $(ST)' = T'S' \quad \forall T \in \mathcal{L}(U, V) \text{ and } S \in \mathcal{L}(V, W)$

Proof 6.

1. $(S + T)' \in \mathcal{L}(W', V')$. Let $\varphi \in W'$. Then,

$$(S + T)'(\varphi) = \varphi \circ (S + T) = \varphi \circ S + \varphi \circ T = S'(\varphi) + T'(\varphi) = (S' + T')(\varphi). \quad \square$$

2. $(\lambda T)' \in \mathcal{L}(W', V')$. Let $\varphi \in W'$. Then,

$$(\lambda T)'(\varphi) = \varphi \circ (\lambda T) = \lambda(\varphi \circ T) = \lambda T'(\varphi) = (\lambda T')(\varphi). \quad \square$$

3. $(ST)' \in \mathcal{L}(W', U')$. Let $\varphi \in W'$. Then,

$$(ST)'(\varphi) = \varphi \circ (ST) = \varphi \circ (S \circ T) = (\varphi \circ S) \circ T = (S'(\varphi)) \circ T = T'(S'(\varphi)) = (T'S')(\varphi).$$

■

Definition 3.5.11 (Transpose/ A^t). The transpose of a matrix A , denoted A^t , is the matrix obtained from A by interchanging the rows and columns. i.e., $(A^t)_{k,j} = A_{j,k}$.

Remark. Transpose is additive and homogeneous. That is, $(A + C)^t = A^t + C^t$ and $(\lambda A)^t = \lambda A^t$.

Theorem 3.5.12

If A is an $m \times n$ matrix and C is an $n \times p$ matrix, then $(AC)^t = C^t A^t$.

Proof 7. Note that

$$(AC)^t_{k,j} = (AC)_{j,k} = \sum_{r=1}^n A_{j,r} C_{r,k} = \sum_{r=1}^n (C^t)_{k,r} (A^t)_{r,j} = (C^t A^t)_{k,j}$$

■

Theorem 3.5.13

Suppose $T \in \mathcal{L}(V, W)$. Then, $\mathcal{M}(T') = (\mathcal{M}(T))^t$.

Proof 8. Suppose v_1, \dots, v_n is a basis of V , w_1, \dots, w_m is a basis of W , $\varphi_1, \dots, \varphi_n$ is a basis of V' , and ψ_1, \dots, ψ_m is a basis of W' . Let $A = \mathcal{M}(T)$ and $C = \mathcal{M}(T')$. Since $T'(\psi_j) = C_{1,j}\varphi_1 + \dots + C_{n,j}\varphi_n$ and $T'(\psi_j) = \psi_j \circ T$, we have $\psi_j \circ T = C_{1,j}\varphi_1 + \dots + C_{n,j}\varphi_n$. Consider

$$(\psi_j \circ T)(v_k) = (C_{1,j}\varphi_1 + \dots + C_{n,j}\varphi_n)(v_k) = C_{k,j}\varphi_k(v_k) = C_{k,j}.$$

Also, we have

$$(\psi_j \circ T)(v_k) = \psi_j(Tv_k) = \psi_j(A_{1,k}w_1 + \cdots + A_{m,k}w_m) = \psi_j(A_{j,k}w_j) = A_{j,k}(\varphi_j(w_j)) = A_{j,k}.$$

Therefore, we have $A_{j,k} = C_{k,j}$, and thus $A = C^t$. So, $\mathcal{M}(T) = (\mathcal{M}(T'))^t$. ■

Definition 3.5.14 (Annihilator/ U^0). For $U \subseteq V$, the *annihilator* of U , denoted as U^0 , is defined by

$$U^0 = \{\varphi \in V' \mid \varphi(u) = 0 \quad \forall u \in U\}.$$

Theorem 3.5.15

Suppose $U \subseteq V$. Then U^0 is a subspace of V' .

Proof 9.

1. $0 \in U^0$: Since $0(u) = 0 \quad \forall u \in U$, then $0 \in U^0$. □

2. Let $\varphi, \psi \in U^0$. Then,

$$(\varphi + \psi)(u) = \varphi(u) + \psi(u) = 0.$$

So, $\varphi + \psi \in U^0$. □

3. Let $\lambda \in \mathbb{F}$ and $\varphi \in U^0$. Then

$$(\lambda\varphi)(u) = \lambda\varphi(u) = \lambda \cdot 0 = 0.$$

So, $\lambda\varphi \in U^0$. ■

Lemma 3.5.16 Suppose V is f - d vector space. If U is a subspace of V and $S \in \mathcal{L}(U, W)$, then there exists $T \in \mathcal{L}(V, W)$ s.t. $Tu = Su \quad \forall u \in U$.

Proof 10. Suppose u_1, \dots, u_m is a basis of U . Then, we can extend it to a basis of V as $u_1, \dots, u_m, v_{m+1}, \dots, v_n$. Define $T \in \mathcal{L}(V, W)$ as $Tu_i = Su_i, Tv_j = 0$, where $i = 1, \dots, m$ and $j = m+1, \dots, n$. Note that

$$\begin{aligned} Tu &= T(a_1u_1 + \cdots + a_mu_m) \\ &= a_1Tu_1 + \cdots + a_mTu_m \\ &= a_1Su_1 + \cdots + a_mSu_m \\ &= S(a_1u_1 + \cdots + a_mu_m) = Su. \end{aligned}$$

Therefore, we've found such a T . ■

Theorem 3.5.17

Let V be f - d and U be a subspace of V , then $\dim U + \dim U^0 = \dim V$.

Proof 11. Let $i \in \mathcal{L}(U, V)$ as $i(u) = u \quad \forall u \in U$. Then, $i' \in \mathcal{L}(V', U')$. So, by Fundamental Theorem of Linear Map, we know

$$\dim V' = \dim \text{null } i' + \dim \text{range } i'. \quad (9)$$

By Theorem 3.5.4, we know $\dim V = \dim V'$. Note that $U^0 = \{\varphi \in V' \mid \varphi(u) = 0 \quad \forall u \in U\}$ and

$$\begin{aligned} \text{null } i' &= \{\varphi \in V' \mid i'(\varphi) = 0\} \\ &= \{\varphi \in V' \mid \varphi \circ i = 0\} \\ &= \{\varphi \in V' \mid (\varphi \circ i)(u) = 0 \quad \forall u \in U\} \\ &= \{\varphi \in V' \mid \varphi(u) = 0 \quad \forall u \in U\} \end{aligned}$$

So, $U^0 = \text{null } i'$, and thus $\dim \text{null } i' = \dim U^0$.

Further, if $\varphi \in U'$, then $\varphi : U \rightarrow \mathbb{F}$. By Lemma 3.5.16, φ can be extended to $\psi \in V'$ with $\psi(u) = \varphi(u) \quad \forall u \in U$. Note that $i'(\psi) = \psi \circ i$, so $(\psi \circ i)(u) = \psi(u) = \varphi(u) \quad \forall u \in U$. Then, $\exists \psi \in V'$ s.t. $i'(\psi) = \varphi$. So, $\varphi \in \text{range } i'$. So, $\dim \text{range } i' = \dim U' = \dim U$.

Substitute $\dim V' = \dim V$, $\dim \text{null } i' = \dim U^0$, and $\dim \text{range } i' = \dim U$ to Equation (9), we get

$$\dim V = \dim U^0 + \dim U.$$

Theorem 3.5.18 The Null Space of T'

Suppose V and W are f -d and $T \in \mathcal{L}(V, W)$. Then,

1. $\text{null } T' = (\text{range } T)^0$
2. $\dim \text{null } T' = \dim \text{null } T + \dim W - \dim V$

Proof 12.

1. (\subseteq) Suppose $\varphi \in \text{null } T' \subseteq W'$. Then, $T'(\varphi) = \varphi \circ T = 0 \in V'$. So, we know

$$(\varphi \circ T)(v) = 0 \quad \forall v \in V. \quad \text{i.e., } \varphi(Tv) = 0.$$

Note that $Tv \in \text{range } T$. By definition, we have $\varphi \in (\text{range } T)^0$ \square

(\supseteq) Suppose $\varphi \in (\text{range } T)^0$. Then, $\varphi(w) = 0 \quad \forall w \in \text{range } T$. That is, $\varphi(Tv) = 0 \quad \forall v \in V$. So, $(\varphi \circ T)(v) = 0 \quad \forall v \in V$. Hence, we know $\varphi \circ T = T'(\varphi) = 0 \in V'$. Thus, $\varphi \in \text{null } T'$ \blacksquare

- 2.

$$\begin{aligned} \dim \text{null } T' &= \dim(\text{range } T)^0 \\ &= \dim W - \dim \text{range } T \\ &= \dim W - (\dim V - \dim \text{null } T) \\ &= \dim W - \dim V + \dim \text{null } T. \end{aligned}$$

Theorem 3.5.19

Suppose V and W are f -d and $T \in \mathcal{L}(V, W)$. Then, T is surjective if and only if T' is injective.

Proof 13.

(\Rightarrow) Suppose T is surjective. Then, $\dim \text{range } T = W$. So, $(\text{range } T)^0 = \{0\}$. Hence,

$$\dim \text{null } T' = \dim(\text{range } T)^0 = 0.$$

Thus, T' is injective. \square

(\Leftarrow) Suppose T' is injective. Then,

$$\dim \text{null } T' = 0.$$

So, $\dim(\text{range } T)^0 = \dim \text{null } T' = 0$. Then, $(\text{range } T)^0 = \{0\}$. So, $\dim \text{range } T = W$, and thus T is surjective. \blacksquare

Theorem 3.5.20 The Range of T'

Suppose V and W are f -d and $T \in \mathcal{L}(V, W)$. Then,

1. $\dim \text{range } T' = \dim \text{range } T$
2. $\text{range } T' = (\text{null } T)^0$

Proof 14.

1. By Fundamental Theorem of Linear Map, we have

$$\begin{aligned} \dim \text{range } T' &= \dim W' - \dim \text{null } T' \\ &= \dim W' - \dim(\text{range } T)^0 \\ &= \dim W' - \dim W' + \dim \text{range } T \\ &= \dim \text{range } T. \end{aligned}$$

2. Suppose $\varphi \in \text{range } T' \subseteq V'$. Then, $\exists \psi \in W'$ s.t. $T'(\psi) = \psi \circ T = \varphi$. Let $v \in \text{null } T$. Then,

$$\varphi(v) = (\psi \circ T)(v) = \psi(Tv) = \psi(0) = 0.$$

Then, $\varphi \in (\text{null } T)^0$. So, $\text{range } T' \subseteq (\text{null } T)^0$. \square

Note that

$$\dim \text{range } T' = \dim \text{range } T = \dim V - \dim \text{null } T = \dim(\text{null } T)^0.$$

Then, $\text{range } T' \subseteq (\text{null } T)^0$ and $\dim \text{range } T' = \dim(\text{null } T)^0$, so it must be that $\text{range } T' = (\text{null } T)^0$. \blacksquare

Theorem 3.5.21

Suppose V and W are f -d and $T \in \mathcal{L}(V, W)$. Then, T is injective if and only if T' is surjective.

Proof 15.

(\Rightarrow) If T is injective, $\text{null } T = \{0\}$. So,

$$\dim \text{null } T = \dim V - \dim(\text{null } T)^0 = \dim V - \dim \text{range } T' = 0.$$

So, $\dim \text{range } T' = \dim V = \dim V'$. Then, T' is surjective. \square

(\Leftarrow) If T' is surjective, $\dim \text{range } T' = \dim V' = \dim V$. So,

$$\dim \text{null } T = \dim V - \dim(\text{null } T)^0 = \dim V - \dim \text{range } T' = 0.$$

Then, $\text{null } T = \{0\}$, and so T is injective. \blacksquare

Definition 3.5.22 (Row Rank & Column Rank). Suppose A is an $m \times n$ matrix with entries in \mathbb{F} .

1. The *row rank* of A is the dimension of the span of the rows of A in $\mathbb{F}^{1,n}$.
2. The *column rank* of A is the dimension of the span of the columns of A in $\mathbb{F}^{m,1}$.

Theorem 3.5.23

Suppose V and W are f -d and $T \in \mathcal{L}(V, W)$. Then, $\dim \text{range } T$ equals the column rank of $\mathcal{M}(T)$.

Proof 16. Suppose v_1, \dots, v_n is a basis of V and w_1, \dots, w_m is a basis of W . Then,

$$Tv_k = A_{1,k}w_1 + \dots + A_{m,k}w_m$$

and thus

$$\mathcal{M}(Tv_k) = \begin{pmatrix} A_{1,k} \\ \vdots \\ A_{m,k} \end{pmatrix} \in \mathbb{F}^{m,1}$$

Therefore, $\mathcal{M}(T) = \begin{pmatrix} \mathcal{M}(Tv_1) & \dots & \mathcal{M}(Tv_n) \end{pmatrix}$. Note that $\text{range } T = \text{span}(Tv_1, \dots, Tv_n)$.

Define $\mathcal{M} : \text{span}(Tv_1, \dots, Tv_n) \rightarrow \text{span}(\mathcal{M}(Tv_1), \dots, \mathcal{M}(Tv_n))$ as $w \mapsto \mathcal{M}(w)$.

1. \mathcal{M} is surjective: Note that

$$c_1\mathcal{M}(Tv_1) + \dots + c_n\mathcal{M}(Tv_n) = \mathcal{M}(c_1Tv_1 + \dots + c_nTv_n).$$

Since $c_1Tv_1 + \dots + c_nTv_n \in \text{range } T$, we know \mathcal{M} is surjective. \square

2. \mathcal{M} is injective: Let

$$\mathcal{M}(c_1Tv_1 + \dots + c_nTv_n) = 0. \tag{10}$$

We can reduce $c_1Tv_1 + \dots + c_nTv_n$ to a basis $Tv_{j_1}, \dots, Tv_{j_m}$. Then, Equation (10) becomes

$$\mathcal{M}(a_1Tv_{j_1} + \dots + a_mTv_{j_m}) = 0. \text{ By definition of matrix, we know } \begin{pmatrix} a_1 \\ \vdots \\ a_m \end{pmatrix} = 0. \text{ So, } a_1 = \dots = a_m = 0$$

and $a_1Tv_{j_1} + \dots + a_mTv_{j_m} = 0$. So, \mathcal{M} is injective. \square

Since \mathcal{M} is both surjective and injective, \mathcal{M} is a bijection. Thus, \mathcal{M} is an isomorphism between $\text{span}(Tv_1, \dots, Tv_n)$ and $\text{span}(\mathcal{M}(Tv_1), \dots, \mathcal{M}(Tv_n))$. In other words,

$$\text{span}(Tv_1, \dots, Tv_n) \cong \text{span}(\mathcal{M}(Tv_1), \dots, \mathcal{M}(Tv_n)).$$

Then, $\dim \text{span}(Tv_1, \dots, Tv_n) = \dim \text{span}(\mathcal{M}(Tv_1), \dots, \mathcal{M}(Tv_n))$. That is,

$$\dim \text{range } T = \text{column rank of } T.$$

■

Theorem 3.5.24 Row Rank Equals Column Rank

Suppose $A \in \mathbb{F}^{m,n}$. Then, the row rank of A equals the column rank of A .

Proof 17. Define $T : \mathbb{F}^{n,1} \rightarrow \mathbb{F}^{m,1}$ by $Tx = Ax$. Then, $\mathcal{M}(T) = A$, where $\mathcal{M}(T)$ is computed with respect to the standard basis of $\mathbb{F}^{n,1}$ and $\mathbb{F}^{m,1}$. Note that

$$\begin{aligned}
 \text{column rank of } A &= \text{column rank of } \mathcal{M}(T) \\
 &= \dim \text{range } T && \text{Theorem 3.5.23} \\
 &= \dim \text{range } T' && \text{Theorem 3.5.20(1)} \\
 &= \text{column rank of } \mathcal{M}(T') \\
 &= \text{column rank of } A^t && \text{Theorem 3.5.13} \\
 &= \text{row rank of } A
 \end{aligned}$$

■

Definition 3.5.25 (Rank). The *rank* of a matrix $A \in \mathbb{F}^{m,n}$ is the column rank of A , denoted as $\text{rank } A$.

3.6 Quotients of Vector Spaces

Definition 3.6.1 ($v + U$ /Affine Subset). Suppose $v \in V$ and U is a subspace of V . Then

$$v + U := \{v + u \mid u \in U\}.$$

An *affine subset* of V is a subset of V of the form $v + U$ for some $v \in V$ and some subspace U of V . The affine subset is said to be *parallel* to U .

Definition 3.6.2 (Quotient Space, V/U). Suppose U is a subspace of V . Then the quotient space V/U is the set of all affine subsets of V parallel to U . In other words,

$$V/U := \{v + U \mid v \in V\}.$$

Example 3.6.3 If $U = \{(x, 2x) \in \mathbb{R}^2 \mid x \in \mathbb{R}\}$, then \mathbb{R}^2/U is the set of all lines in \mathbb{R}^2 with slope of 2.

Theorem 3.6.4

Suppose U is a subspace of V and $v, w \in V$. Then, the following are equivalent:

1. $v - w \in U$
2. $v + U = w + U$
3. $(v + U) \cap (w + U) \neq \emptyset$

Proof 1.

1. We want to show (1) \implies (2). Suppose $v - w \in U$. Note that $v + u = w + ((v - w) + u)$. Since $v - w$ and $u \in U$, we have $(v - w) + u \in U$. So, $v + u \in w + U$. Similarly, we can show that $w + u \in v + U$. Then, we have $v + U = w + U$. \square
2. Now, we want to show (2) \implies (3): Suppose $v + U = w + U$. Then, we have $(v + U) \cap (w + U) \neq \emptyset$, which is evident from the assumption. \square
3. Finally, we will show (3) \implies (1). Suppose $(v + U) \cap (w + U) \neq \emptyset$. Then, $\exists u_1, u_2 \in U$ s.t. $v + u_1 = w + u_2$. So we have $v - w = u_2 - u_1 \in U$. \blacksquare

Definition 3.6.5 (Addition & Scalar Multiplication on V/U). Suppose U is a subspace of V . Then, *addition* and *scalar multiplication* is defined on V/U by

$$(v + U) + (w + U) = (v + w) + U$$

and

$$\lambda(v + U) = (\lambda v) + U$$

for $v, w \in U$ and $\lambda \in \mathbb{F}$.

Theorem 3.6.6

Suppose U is a subspace of V . Then, V/U , with the operations of addition and scalar multiplication defined above, is a vector space.

Proof 2.

1. Addition on V/U makes sense.

Note the addition can be written in the language of mapping as $+: V/U \times V/U \rightarrow V/U$. So, we have $(v + U, w + U) \mapsto (v + w) + U$. Suppose $\exists \hat{v}, \hat{w} \in V$ s.t. $v + U = \hat{v} + U$ and $w + U = \hat{w} + U$. Note that $v - \hat{v} \in U$ and $w - \hat{w} \in U$ by Theorem 3.6.4. Then, $(v - \hat{v}) + (w - \hat{w}) \in U$. So, we have $(v + w) - (\hat{v} + \hat{w}) \in U$. Further, by Theorem 3.6.4, we have

$$(v + w) + U = (\hat{v} + \hat{w}) + U. \quad \square$$

2. Scalar multiplication on V/U makes sense.

We can write the scalar multiplication on V/U as a mapping: $\cdot : \mathbb{F} \times V/U \rightarrow V/U$ defined as $(\lambda, v + U) \mapsto \lambda v + U$. Suppose $\exists \hat{v} \in V$ s.t. $v + U = \hat{v} + U$. So we know $v - \hat{v} \in U$, and thus $\lambda(v - \hat{v}) = \lambda v - \lambda \hat{v} \in U$. By Theorem 3.6.4, we then have $(\lambda v) + U = (\lambda \hat{v}) + U$. Thus, the scalar multiplication makes sense. \square

3. additive identity: $0 + U = U$. \square

4. additive inverse: $(-v) + U$. \square

5. commutativity:

$$\begin{aligned} (v + U) + (w + U) &= (v + w) + U = (w + v) + U \\ &= (w + U) + (v + U). \end{aligned} \quad \square$$

6. associativity:

$$\begin{aligned} [(v + U) + (w + U)] + (x + U) &= [(v + w) + U] + (x + U) \\ &= [(v + w) + x] + U \\ &= [v + (w + x)] + U \\ &= (v + U) + [(w + x) + U] \\ &= (v + U) + [(x + U) + (w + U)]. \end{aligned} \quad \square$$

7. multiplicative identity: $1 \cdot (v + U) = (1 \cdot v) + U = v + U$. \square

8. distributivity:

$$\begin{aligned} a[(v + U) + (w + U)] &= a[(v + w) + U] \\ &= a(v + w) + U \\ &= (av + aw) + U \\ &= (av + U) + (aw + U) \\ &= a(v + U) + a(w + U). \end{aligned}$$

$$\begin{aligned}
(a+b)(v+U) &= (a+b)v + U \\
&= (av + bv) + U \\
&= (av + U) + (bv + U) \\
&= a(v+U) + b(v+U)
\end{aligned}$$

Definition 3.6.7 (Quotient Map). Suppose U is a subspace of V . The *quotient map* π is the linear map $\pi : V \rightarrow V/U$ defined by $\pi(v) := v + U \quad \forall v \in V$. ■

Remark. Here are some properties of the quotient map:

1. $\pi(v)$ is defined $\forall v \in V$. Thus, π is surjective.
2. $\text{null } \pi = \{v \in V \mid \pi(v) = 0\}$. If $\pi(v) = 0$, then $v + U = U = 0 + U$. So, $v - 0 \in U$ by Theorem 3.6.4. Then, $v \in U$. So, $\text{null } \pi \subseteq U$. Further, $\forall v \in U$, if $\pi(v) = 0$, then $v \in \text{null } \pi$, then $U \subseteq \text{null } \pi$. So, $U = \text{null } \pi$.
3. $\pi(v + w) = (v + w) + U = (v + U) + (w + U) = \pi(v) + \pi(w)$.
4. $\pi(\lambda v) = (\lambda v) + U = \lambda(v + U) = \lambda\pi(v)$.

Theorem 3.6.8

Suppose V is f - d and U is a subspace of V . Then

$$\dim V/U = \dim V - \dim U.$$

Proof 3. By Fundamental Theorem of Linear Map, we have

$$\dim V = \dim \text{null } \pi + \dim \text{range } \pi. \quad (11)$$

Since $\text{null } \pi = U$ from the Remark, we have $\dim \text{null } \pi = \dim U$. Further, since π is surjective as mentioned in the Remark, $\text{range } \pi = V/U$. Hence, $\dim \text{range } \pi = \dim V/U$. Therefore, Equation (11) becomes

$$\dim V = \dim U + \dim V/U,$$

or we have

$$\dim V/U = \dim V - \dim U$$

Definition 3.6.9 (\tilde{T}). Suppose $T \in \mathcal{L}(V, W)$. Define $\tilde{T} : V/(\text{null } T) \rightarrow W$ by $\tilde{T}(v + \text{null } T) = Tv$. ■

Proof 4.

1. This definition makes sense

Suppose $u, v \in V$ s.t. $u + \text{null } T = v + \text{null } T$. By Theorem 3.6.4, we know $u - v \in \text{null } T$. Then, $T(u - v) = 0$, or $Tu = Tv$. □

2. \tilde{T} is a linear map.

$$\begin{aligned}
\tilde{T}[(u + \text{null } T) + (v + \text{null } T)] &= \tilde{T}[(u + v) + \text{null } T] \\
&= T(u + v) \\
&= Tu + Tv = \tilde{T}(u + \text{null } T) + \tilde{T}(v + \text{null } T). \quad \square
\end{aligned}$$

$$\begin{aligned}
\tilde{T}[\lambda(u + \text{null } T)] &= \tilde{T}(\lambda u + \text{null } T) \\
&= T(\lambda u) \\
&= \lambda T u \\
&= \lambda T(u + \text{null } T).
\end{aligned}$$

■

Theorem 3.6.10

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

1. \tilde{T} is injective.
2. $\text{range } \tilde{T} = \text{range } T$.
3. $V/(\text{null } T) \cong \text{range } T$.

Proof 5.

1. Suppose $v \in V$ and $\tilde{T}(v + \text{null } T) = 0$. Then, $Tv = 0$. So, $v \in \text{null } T$, or $v - 0 \in \text{null } T$. By Theorem 3.6.4, we then have $v + \text{null } T = 0 + \text{null } T$. Then, it implies $\text{null } \tilde{T} = 0$. So, \tilde{T} is injective. \square
2. By definition of \tilde{T} , it must be $\text{range } \tilde{T} = \text{range } T$. \square
3. Note that $\dim V/(\text{null } T) = \dim \text{null } \tilde{T} + \dim \text{range } \tilde{T} = 0 + \dim \text{range } T$. Then, by Theorem 3.4.9, we know two vector spaces are isomorphic if and only if their dimensions are equal. Then,

$$V/(\text{null } T) \cong \text{range } T.$$

■

4 Eigenvectors and Invariant Subspaces

4.1 Invariant Subspaces

Theorem 4.1.1

Suppose V is f - d with $\dim V = n \geq 1$. Then, \exists 1-dimensional subspaces U_1, \dots, U_n of V s.t.

$$V = U_1 \oplus \dots \oplus U_n.$$

Proof 1. Choose a basis v_1, \dots, v_n of V . Then, we know $V = \text{span}(v_1) + \dots + \text{span}(v_n)$. Also, $\forall v \in V$, we have $v = a_1 v_1 + \dots + a_n v_n$ with $a_j v_j \in \text{span}(v_j)$. Set $a_1 v_1 + \dots + a_n v_n = 0$. Since v_1, \dots, v_n is a basis, it must be $a_1 = \dots = a_n = 0$. Then,

$$V = \text{span}(v_1) \oplus \dots \oplus \text{span}(v_n).$$

■

Theorem 4.1.2

Suppose U_1, \dots, U_m are f - d subspaces of V s.t. $U_1 + \dots + U_m$ is a direct sum. Then, $U_1 \oplus \dots \oplus U_m$ is f - d and

$$\dim U_1 \oplus \dots \oplus U_m = \dim U_1 + \dots + \dim U_m.$$

Proof 2. Suppose $u_{k,1}, \dots, u_{k,j_k}$ is a basis of the subspace U_k . Then, any vector in $\bigoplus_{i=1}^m U_i$ is in the form of $u_1 + \dots + u_m$, $u_j \in U_j$. Also,

$$u_i = \sum_{k=1}^{j_i} a_{i,k} u_{i,k}.$$

So,

$$u_1 + \dots + u_m = \sum_{k=1}^{j_1} a_{1,k} u_{1,k} + \dots + \sum_{k=1}^{j_m} a_{m,k} u_{m,k}.$$

Then, $u_1 + \dots + u_m$ is a linear combination of $u_{1,1}, \dots, u_{j,m}$. So, the direct sum is f - d . \square

Further, suppose

$$\sum_{k=1}^{j_1} a_{1,k} u_{1,k} + \dots + \sum_{k=1}^{j_m} a_{m,k} u_{m,k} = 0.$$

Since $U_1 + \dots + U_m$ is a direct sum, it must be

$$\sum_{k=1}^{j_1} a_{1,k} u_{1,k} = \dots = \sum_{k=1}^{j_m} a_{m,k} u_{m,k} = 0.$$

Since we selected bases, $a_{1,k} = \dots = a_{m,k} = 0$. So, $u_{1,1}, \dots, u_{j,m}$ is a basis of $U_1 \oplus \dots \oplus U_m$. Then,

$$\dim U_1 \oplus \dots \oplus U_m = \dim U_1 + \dots + \dim U_m.$$

■

Definition 4.1.3 (Invariant Subspace). Suppose $T \in \mathcal{L}(V)$. A subspace U of V is called *invariant* under T if $u \in U$ implies $Tu \in U$.

Example 4.1.4 Suppose $T \in \mathcal{L}(V)$. Show that each of the following subspaces of V is invariant under T :

1. $\{0\}$

Proof 3. $T0 = 0 \in \{0\}$ ■

2. V

Proof 4. $u \in V \implies Tu \in V$ ■

3. $\text{null } T$

Proof 5. $u \in \text{null } T \implies Tu = 0 \in \text{range } T$ ■

4. $\text{range } T$

Proof 6. $u \in \text{range } T \implies Tu \in \text{range } T$ ■

Example 4.1.5 Suppose $T \in \mathcal{L}(\mathcal{P}(\mathbb{R}))$ is defined by $Tp = p'$. Then, $\mathcal{P}_4(\mathbb{R})$ is invariant under T .

Proof 7. Note that $Tp_4 \in \mathcal{P}_4(\mathbb{R})$. Then, $\mathcal{P}_4(\mathbb{R})$ is invariant under T . ■

Definition 4.1.6 (Eigenvalue). Suppose $T \in \mathcal{L}(V)$. A number $\lambda \in \mathbb{F}$ is called an *eigenvalue* of T if $\exists v \in V$ s.t. $v \neq 0$ and $Tv = \lambda v$.

Corollary 4.1.7 T has a 1-dimensional invariant subspace if and only if T has an eigenvalue.

Proof 8.

(\implies) Suppose $\text{span}(v)$ is invariant under T . Let U be defined as $U = \{\lambda v \mid \lambda \in \mathbb{F}\} = \text{span}(v)$. Then, U is the invariant subspace under T and $\dim U = 1$. Then, $\forall v \in V$, we have $Tv \in U$. Hence, $\exists \lambda \in \mathbb{F}$ s.t. $Tv = \lambda v$. Then, λ is an eigenvalue. □

(\impliedby) Suppose $\lambda \in \mathbb{F}$ is an eigenvalue. Then, $Tv = \lambda v$. Hence, $\text{span}(v)$ is a 1-dimensional invariant subspace under T . ■

Theorem 4.1.8 Equivalent Conditions to be an Eigenvalue

Suppose V is f -d, $T \in \mathcal{L}(V)$, and $\lambda \in \mathbb{F}$. Then, the following are equivalent:

1. λ is an eigenvalue of T .
2. $T - \lambda I$ is not injective.
3. $T - \lambda I$ is not surjective.
4. $T - \lambda I$ is not invertible.

Proof 9.

1. (1) \implies (2): Suppose λ is an eigenvalue of T . Then, $\exists v \in V$ s.t. $v \neq 0$ and $Tv = \lambda v$. So, $Tv - \lambda v = (T - \lambda I)v = 0$. Since $v \neq 0$, $\text{null}(T - \lambda I) \neq \{0\}$, and thus T is not injective. □
2. Note that $T - \lambda I$ is an operator by itself. By Theorem 3.4.17, we know (2), (3), and (4) are equivalent.

3. (4) \implies (1): Suppose $T - \lambda I$ is not invertible. Then, it is not injective. So, $\exists v \neq 0$ s.t. $(T - \lambda I)v = 0$. That is, $Tv - \lambda Iv = Tv - \lambda v = 0$. So, $Tv = \lambda v$. Then, λ is an eigenvalue of T . ■

Definition 4.1.9 (Eigenvector). Suppose $T \in \mathcal{L}(V)$ and $\lambda \in \mathbb{F}$ is an eigenvalue of T . A vector $v \in V$ is called an *eigenvector* of T corresponding to λ if $v \neq 0$ and $Tv = \lambda v$.

Corollary 4.1.10 A vector $v \in V$ with $v \neq 0$ is an eigenvector of T with respect to λ if and only if $v \in \text{null}(T - \lambda I)$.

Proof 10. Note that $Tv = \lambda v$ if and only if $(T - \lambda I)v = 0$. ■

Example 4.1.11 Suppose $T \in \mathcal{L}(\mathbb{F}^2)$ is defined by $T(w, z) = (-z, w)$.

1. Find the eigenvalues and eigenvectors of T if $\mathbb{F} = \mathbb{R}$.

Answer 11.

Let $T(w, z) = \lambda(w, z)$. So, $(-z, w) = (\lambda w, \lambda z)$. Then, solve $\begin{cases} -z = \lambda w \\ w = \lambda z \end{cases}$.

Then, we have $\lambda^2 z + z = 0$. If $z \neq 0$, $\lambda^2 + 1 = 0$. This equation has no solutions on \mathbb{R} . So T has no eigenvalues. If $w = 0, z = 0$, then $T(w, z) = T(0, 0) = T0$. By definition, T has no eigenvalues. □

2. Find the eigenvalues and eigenvectors of T if $\mathbb{F} = \mathbb{C}$.

Answer 12.

Applying similar rational, $z \neq 0$ and solve $\lambda^2 + 1 = 0$. Then, we have $\lambda = \pm i$. If $\lambda = i$, then $-z = iw$. So, $v = (w, z) = (w, -iw)$. If $\lambda = -i$, then $-z = -iw$, or $z = iw$. So, $v = (w, iw)$. □

Theorem 4.1.12

Let $T \in \mathcal{L}(V)$. Suppose $\lambda_1, \dots, \lambda_m$ are distinct eigenvalues of T and v_1, \dots, v_m are corresponding eigenvectors. Then, v_1, \dots, v_m is L.I..

Proof 13. Suppose for the sake of contradiction that v_1, \dots, v_m is linearly dependent. Let k be the smallest positive integer s.t. $v_k \in \text{span}(v_1, \dots, v_{k-1})$. Then, $v_k = a_1 v_1 + \dots + a_{k-1} v_{k-1}$. Applying T , we have

$$\lambda_k v_k = a_1 \lambda_1 v_1 + \dots + a_{k-1} \lambda_{k-1} v_{k-1}. \quad (12)$$

Since $v_k = a_1 v_1 + \dots + a_{k-1} v_{k-1}$, we also have

$$\lambda_k v_k = a_1 \lambda_k v_1 + \dots + a_{k-1} \lambda_k v_{k-1}. \quad (13)$$

So, by Equation (13)-(12), we have

$$0 = a_1 (\lambda_k - \lambda_1) v_1 + \dots + a_{k-1} (\lambda_k - \lambda_{k-1}) v_{k-1}.$$

By assumption, v_1, \dots, v_{k-1} is L.I.. Then, it must be that $a_1 = \dots = a_{k-1} = 0$ since $\lambda_1, \dots, \lambda_k$ are distinct eigenvalues. Therefore, $v_k = a_1 v_1 + \dots + a_{k-1} v_{k-1} = 0$. * This contradicts with the fact that v_k is an eigenvector, which cannot be 0. So, it must be that v_1, \dots, v_m are L.I. ■

Theorem 4.1.13

Suppose V is f - d . Then, each operator on V has at most $\dim V$ distinct eigenvalues.

Proof 14. Let $T \in \mathcal{L}(V)$. Suppose $\lambda_1, \dots, \lambda_m$ are distinct eigenvalues of T . Let v_1, \dots, v_m be corresponding eigenvectors. By Theorem 4.1.12, we know v_1, \dots, v_m is L.I.. Further by Theorem 2.3.5, we know $\dim \text{span}(v_1, \dots, v_m) \leq \dim V$. That is, $m \leq \dim V$ as desired. ■

4.2 Eigenvectors and Upper-Triangular Matrices

Definition 4.2.1 (T^m). Suppose $T \in \mathcal{L}(V)$ and m is a positive integer. Then, T^m is defined by

$$T^m := \underbrace{T \cdots T}_{m \text{ times}}.$$

Specially, T^0 is defined to be the identity operator I on V . Further, if T is invertible with inverse T^{-1} , then T^{-m} is defined by $T^{-m} := (T^{-1})^m$.

Theorem 4.2.2

$$T^m T^n = T^{m+n}; \quad (T^m)^n = T^{mn}.$$

Definition 4.2.3 ($p(T)$). Suppose $T \in \mathcal{L}(V)$ and $p \in \mathcal{P}(\mathbb{F})$ is a polynomial given by

$$p(z) = a_0 + a_1 z + a_2 z^2 + \cdots + a_m z^m, \quad z \in \mathbb{F}.$$

Then, $p(T)$ is the operator defined by

$$p(T) := a_0 I + a_1 T + a_2 T^2 + \cdots + a_m T^m.$$

Example 4.2.4 Suppose $D \in \mathcal{L}(\mathcal{P}(\mathbb{R}))$ is the differentiation operator defined by $Dq = q'$ and p is the polynomial defined by $p(x) = 7 - 3x + 5x^2$. Find $p(D)$ and $(p(D))q$.

Answer 1.

$$\begin{aligned} p(D) &= 7I - 3D + 5D^2 \\ (p(D))q &= (7I - 3D + 5D^2)q \\ &= 7Iq - 3Dq + 5D^2q \\ &= 7q - 3q' + 5q''. \end{aligned}$$

□

Theorem 4.2.5

If we fix an operator $T \in \mathcal{L}(V)$, then the function from $\mathcal{P}(\mathbb{F})$ to $\mathcal{L}(V)$ given by $p \mapsto p(T)$ is linear.

Proof 2. Suppose $f : \mathcal{P}(\mathbb{F}) \rightarrow \mathcal{L}(V)$ is defined by $p \mapsto p(T)$. Suppose

$$p = a_0 + a_1 z + \cdots + a_m z^m \mapsto a_0 I + a_1 T + \cdots + a_m T^m$$

and

$$q = b_0 + b_1 z + \cdots + b_m z^m \mapsto b_0 I + b_1 T + \cdots + b_m T^m.$$

Then,

$$\begin{aligned} f(p+q) &= (a_0 + b_0)I + (a_1 + b_1)T + \cdots + (a_m + b_m)T^m \\ &= (a_0 I + a_1 T + \cdots + a_m T^m) + (b_0 I + b_1 T + \cdots + b_m T^m) \\ &= f(p) + f(q). \end{aligned}$$

Further, suppose $\lambda \in \mathbb{F}$, then

$$\begin{aligned} f(\lambda p) &= \lambda a_0 I + \lambda a_1 T + \cdots + \lambda a_m T^m \\ &= \lambda(a_0 I + a_1 T + \cdots + a_m T^m) \\ &= \lambda f(p). \end{aligned}$$

Definition 4.2.6 (Product of Polynomials). If $p, q \in \mathcal{P}(\mathbb{F})$, then $pq \in \mathcal{P}(\mathbb{F})$ is the polynomial defined by $(pq)(z) := p(z)q(z)$ for $z \in \mathbb{F}$. ■

Remark. $(pq)(z) = p(z)q(z) = q(z)p(z) = (qp)(z)$ for $z \in \mathbb{F}$.

Theorem 4.2.7 Multiplicative Properties

Suppose $p, q \in \mathcal{P}(\mathbb{F})$ and $T \in \mathcal{L}(V)$. Then

1. $(pq)(T) = p(T)q(T)$
2. $p(T)q(T) = q(T)p(T)$

Proof 3.

1. Suppose $p(z) = \sum_{j=0}^m a_j z^j$ and $q(z) = \sum_{k=0}^n b_k z^k$. Then

$$(pq)(z) = p(z)q(z) = \sum_{j=0}^m a_j z^j \sum_{k=0}^n b_k z^k = \sum_{j=0}^m \sum_{k=0}^n a_j b_k z^{j+k}$$

So, by definition, we have

$$p(T)q(T) = \sum_{j=0}^m \sum_{k=0}^n a_j b_k T^{j+k} = \left(\sum_{j=0}^m a_j T^j \right) \cdot \left(\sum_{k=0}^n b_k T^k \right) = p(T)q(T). \quad \square$$

2. Similar to the Remark,

$$p(T)q(T) = (pq)(T) = (qp)(T) = q(T)p(T).$$

Theorem 4.2.8 Fundamental Theorem of Algebra

Every non-constant polynomial with complex coefficients has a zero. ■

Theorem 4.2.9 Existence of Eigenvalues

Every operator on a f -d, non-zero, complex vector space has an eigenvalue.

Proof 4. Let V be a complex vector space with dimension $n > 0$. Suppose $T \in \mathcal{L}(V)$. Choose $v \in V$ s.t. $v \neq 0$. Then, v, Tv, T^2v, \dots, T^nv is linearly dependent because $\dim V = n$ but the length of the list is $n + 1 > n$. Hence, $\exists a_0, a_1, \dots, a_n$ not all 0 $\in \mathbb{C}$ s.t.

$$0 = a_0 v + a_1 Tv + \cdots + a_n T^n v \tag{14}$$

By Fundamental Theorem of Algebra (Theorem 4.2.8), we have

$$a_0 + a_1z + \cdots + a_nz^n = c(z - \lambda_1) \cdots (z - \lambda_m)$$

with $c \in \mathbb{C}$, $c \neq 0$, and $\lambda_j \in \mathbb{C}$. Then, Equation (14) becomes

$$\begin{aligned} 0 &= a_0v + a_1Tv + \cdots + a_nT^n v \\ &= (a_0I + a_1T + \cdots + a_nT^n)v \\ &= c(T - \lambda_1I) \cdots (T - \lambda_mI)v \end{aligned}$$

Since $v \neq 0$ and $c \neq 0$, it must be some $T - \lambda_iI = 0$. Thus, $T = \lambda_iI$, and λ_i is an eigenvalue of T . ■

Definition 4.2.10 (Diagonal of a Matrix). The *diagonal of a square matrix* consists of the entries along the line from the upper left corner to the bottom right corner.

Definition 4.2.11 (Upper-Triangular Matrix). A matrix is called *upper-triangular* if all the entries below the diagonal equal 0. Typically, we present an upper triangular matrix in the form

$$\begin{pmatrix} \lambda_1 & & * \\ & \ddots & \\ 0 & & \lambda_n \end{pmatrix}.$$

Theorem 4.2.12 Conditions for Upper-Triangular Matrix

Suppose $T \in \mathcal{L}(V)$ and v_1, \dots, v_n is a basis of V . Then, the following are equivalent:

1. the matrix of T with respect to v_1, \dots, v_n is upper triangular.
2. $Tv_j \in \text{span}(v_1, \dots, v_j)$ for each $j = 1, \dots, n$
3. $\text{span}(v_1, \dots, v_j)$ is invariant under T for each $j = 1, \dots, n$.

Proof 5.

1. First, we will show (1) \iff (2).

Suppose $\mathcal{M}(T) = \begin{pmatrix} A_{1,1} & \cdots & A_{1,n} \\ & \ddots & \vdots \\ 0 & & A_{n,n} \end{pmatrix}$. Then,

$$\begin{aligned} Tv_1 &= A_{1,1}v_1 \\ Tv_2 &= A_{1,2}v_1 + A_{2,2}v_2 \\ &\vdots \\ Tv_j &= A_{1,j}v_1 + \cdots + A_{j,j}v_j. \end{aligned}$$

So, $Tv_j \in \text{span}(v_1, \dots, v_j)$. The reverse implication is trivial to prove. □

2. (3) \implies (2) is obvious and trivial to prove.

3. Lastly, we want to show (2) \implies (3).

Note that for each fixed $j = 1, \dots, n$, we have

$$\begin{aligned} Tv_1 &\in \text{span}(v_1) \subseteq \text{span}(v_1, \dots, v_j) \\ Tv_2 &\in \text{span}(v_1, v_2) \subseteq \text{span}(v_1, \dots, v_j) \\ &\vdots \\ Tv_j &\in \text{span}(v_1, \dots, v_j) \end{aligned}$$

Let $v \in \text{span}(v_1, \dots, v_j)$. Then, v is a linear combination of v_1, \dots, v_j , then

$$Tv \in \text{span}(v_1, \dots, v_j).$$

That is, $\text{span}(v_1, \dots, v_j)$ is invariant under T . ■

Definition 4.2.13 (Quotient Operator). Suppose $T \in \mathcal{L}(V)$ and U is a subspace of V invariant under T . The *quotient operator* $T/U \in \mathcal{L}(V/U)$ is defined by $(T/U)(v + U) := Tv + U$.

Proof 6. The definition makes sense, and here is the proof. If $v + U = w + U$, then $v - w \in U$. So, $T(v - w) \in U$ since U is invariant. That is, $Tv - Tw \in U$. Then, $Tv + U = Tw + U$. ■

Theorem 4.2.14

Suppose U is a subspace of V . Let $v_1 + U, \dots, v_m + U$ be a basis of V/U and u_1, \dots, u_n be a basis of U . Then, $v_1, \dots, v_m, u_1, \dots, u_n$ is a basis of V .

Proof 7. Let $v \in V$. Then $v + U \in V/U$. So, $v + U = a_1v_1 + \dots + a_mv_m + U$, uniquely. Then, by Theorem 3.6.4, we have $v - (a_1v_1 + \dots + a_mv_m) \in U$. Therefore, $v - (a_1v_1 + \dots + a_mv_m) = b_1u_1 + \dots + b_nu_n$, uniquely. So, $v = a_1v_1 + \dots + a_mv_m + b_1u_1 + \dots + b_nu_n$, uniquely. By definition, $v_1, \dots, v_m, u_1, \dots, u_n$ is a basis of V . ■

Theorem 4.2.15

Suppose V is a f - d complex vector space and $T \in \mathcal{L}(V)$. Then, T has an upper-triangular matrix with respect to some basis of V .

Proof 8.

Base Case When $\dim V = 1$, the implication holds.

Inductive Steps Suppose the implication is true for some complex vector space with dimension of $n - 1$. Let $\dim V = n$ and v_1 be any eigenvector of T . Suppose $U = \text{span}(v_1)$. Then, U is invariant under T . Note that $\dim V/U = \dim V - \dim U = n - 1$, so we can use the inductive hypothesis on the quotient operator $T/U \in \mathcal{L}(V/U)$. Then, \exists a basis $v_2 + U, \dots, v_n + U \in V/U$ s.t. T/U has an upper-triangular matrix. By Theorem 4.2.12, we have

$$(T/U)(v_j + U) \in \text{span}(v_2 + U, \dots, v_j + U) \quad \text{for } j \in \{2, \dots, n\}.$$

So, $Tv_j + U = (c_2v_2 + \dots + c_jv_j) + U$. Then,

$$Tv_j - (c_2v_2 + \dots + c_jv_j) \in U = \text{span}(v_1).$$

So, $Tv_j - (c_2v_2 + \dots + c_jv_j) = c_1v_1$ for some $c_1 \in \mathbb{F}$. Then, $Tv_j = c_1v_1 + c_2v_2 + \dots + c_jv_j$. So, $Tv_j \in \text{span}(v_1, \dots, v_j)$ for $j \in \{1, \dots, n\}$. Since by Theorem 4.2.14, v_1, \dots, v_n is a basis of V , further

by Theorem 4.2.12, T has an upper-triangular matrix with respect to v_1, \dots, v_n . So, the implication is true for $\dim V = n$.

Since the implication is true for $\dim V = 1$ and is true for $\dim V = n$ whenever it is hold for $\dim V = n - 1$, by the Principle of Mathematical Induction, the implication is true for all positive integers n . Hence, the proof is complete. ■

4.3 Eigenspaces and Diagonal Matrices

5 Inner Product Spaces

5.1 Inner Products and Norms

5.2 Orthonormal Bases

5.3 Orthogonal Complements and Minimization Problems

6 Operators on Inner Product Spaces

6.1 Self-Adjoint and Normal Operators

6.2 The Spectral Theorem

6.3 Positive Operators and Isometries

6.4 Polar Decomposition and SVD

7 Operators on Complex Vector Spaces

7.1 Generalized Eigenvectors, Nilpotent Operators

7.2 Decomposition of an Operator

7.3 Characteristic and Minimal Polynomials

7.4 Jordan Form

8 Operators on Real Vectors Spaces

8.1 Complexification

8.2 Operators on Real Inner Product Spaces

9 Trace and Determinant

9.1 Trace

9.2 Determinant

10 Exercises

10.1 Span and Linear Independence

1. Suppose v_1, v_2, v_3, v_4 spans V . Prove that the list $v_1 - v_2, v_2 - v_3, v_3 - v_4, v_4$ also spans V .
2. Prove that if \mathbb{C} is a vector space on \mathbb{R} , then the list $1 + i, 1 - i$ is L.I..
3. Prove that if \mathbb{C} is a vector space on \mathbb{C} , then the list $1 + i, 1 - i$ is linearly dependent.
4. Prove or give a counterexample: Suppose v_1, v_2, \dots, v_m is L.I. in V and $\lambda \in \mathbb{F}$ with $\lambda \neq 0$. Then $\lambda v_1, \lambda v_2, \dots, \lambda v_m$ is L.I..
5. Suppose v_1, \dots, v_m is L.I. in V and $w \in V$. Prove that if $v_1 + w, \dots, v_m + w$ is linearly dependent, then $w \in \text{span}(v_1, \dots, v_m)$.

10.2 Bases

1. Find all the vectors spaces that consist of only one basis.

Hint. $\{0\}$.

2. Suppose U is a subspace of \mathbb{R}^5 s.t. $U = \{(x_1, x_2, x_3, x_4, x_5) \in \mathbb{R}^5 \mid x_1 = 3x_2, x_3 = 7x_4\}$. Find a basis of U . Extend this basis into a basis of \mathbb{R}^5 . Then, find a subspace W of \mathbb{R}^5 s.t. $\mathbb{R}^5 = U \oplus W$.
3. Suppose v_1, v_2, v_3, v_4 is a basis of V . Prove that $v_1 + v_2, v_2 + v_3, v_3 + v_4, v_4$ is also a basis of V .
4. **Prove** or disprove: $\mathcal{P}_3(\mathbb{F})$ has a basis p_0, p_1, p_2, p_3 s.t. no one from p_0, p_1, p_2, p_3 has a degree of 2.

Hint. Use the conclusion from #3.

5. Prove or **give a counterexample**: If v_1, v_2, v_3, v_4 is a basis of V and U is a subspace of V s.t. $v_1, v_2 \in U, v_3 \notin U, v_4 \notin U$, then v_1, v_2 is basis of U .

10.3 Dimension

1. Suppose V is f - d and U is a subspace of V s.t. $\dim U = \dim V$. Prove that $U = V$.
2. Prove that the subspaces of \mathbb{R}^2 are exactly the following: $\{0\}, \mathbb{R}^2$, and all the lines passing through the origin in \mathbb{R}^2 .
3. Suppose v_1, \dots, v_m is L.I. in V and $w \in V$. Prove $\dim \text{span}(v_1 + w, \dots, v_m + w) \geq m - 1$.
4. Suppose $p_0, p_1, \dots, p_m \in \mathcal{P}(\mathbb{F})$ s.t. $\deg p_j = j$. Prove p_0, p_1, \dots, p_m is a basis of $\mathcal{P}_m(\mathbb{F})$.
5. Suppose U and W are subspaces of \mathbb{R}^8 s.t. $\dim U = 3, \dim W = 5$, and $U + W = \mathbb{R}^8$. Prove that $\mathbb{R}^8 = U \oplus W$.
6. Suppose U and W are 5-dimensional subspaces of \mathbb{R}^9 . Prove $U \cap W \neq \{0\}$.
7. Suppose U and W are 4-dimensional subspaces of \mathbb{C}^6 . Prove that \exists two vectors in $U \cap W$ s.t. any one of which is not a scalar multiple of another one.

8. Suppose U_1, \dots, U_m are f - d vector spaces of V . Prove that $U_1 + \dots + U_m$ is f - d and

$$\dim(U_1 + \dots + U_m) \leq \dim U_1 + \dots + \dim U_m.$$

9. Suppose V is f - d and $\dim V = n \geq 1$. Prove that \exists 1-dimensional subspaces of V , U_1, \dots, U_n s.t.

$$V = U_1 \oplus \dots \oplus U_n.$$

10. Suppose U_1, \dots, U_m are f - d vector subspaces of V s.t. $U_1 + \dots + U_m$ is a direct sum. Prove that $U_1 \oplus \dots \oplus U_m$ is f - d and

$$\dim U_1 \oplus \dots \oplus U_m = \dim U_1 + \dots + \dim U_m.$$

Hint. Use mathematical induction.

Remark. This problem deepens the analogy between direct sums of subspaces and disjoint unions of subsets. Specifically, compare this problem to the following obvious statement: if a set is written as a disjoint union of finite subsets, then the number of elements in the set equals the sum of the numbers of elements in the disjoint subsets.

11. Prove or give a counter example:

$$\begin{aligned} \dim(U_1 + U_2 + U_3) &= \dim U_1 + \dim U_2 + \dim U_3 \\ &\quad - \dim(U_1 \cap U_2) - \dim(U_1 \cap U_3) - \dim(U_2 \cap U_3) \\ &\quad + \dim(U_1 \cap U_2 \cap U_3). \end{aligned}$$

Hint. Consider $U_1 = \{(x, 0) \mid x \in \mathbb{R}\}$, $U_2 = \{(0, y) \mid y \in \mathbb{R}\}$, $U_3 = \{(x, x) \mid x \in \mathbb{R}\}$.

10.4 The Vector Space of Linear Maps

1. Suppose $T \in \mathcal{L}(V, W)$ and v_1, \dots, v_m is a vector list in V s.t. Tv_1, \dots, Tv_m is L.I. in W . Prove that v_1, \dots, v_m is L.I.
2. Prove that $\mathcal{L}(V, W)$ is a vector space.
3. Prove the algebraic properties of products of linear maps.
4. Show that every linear map from a 1-dimensional vector space to itself is multiplication by some scalar. More precisely, prove that if $\dim V = 1$ and $T \in \mathcal{L}(V, V)$, then $\exists \lambda \in \mathbb{F}$ s.t. $Tv = \lambda v \quad \forall v \in V$.

10.5 Null Spaces and Range

1. Suppose V is a vector space and $S, T \in \mathcal{L}(V, V)$ s.t. $\text{range } S \subset \text{null } T$. Prove that $(ST)^2 = 0$.
2. Prove that \nexists a linear map $T : \mathbb{R}^5 \rightarrow \mathbb{R}^5$ s.t. $\text{range } T = \text{null } T$.
3. Suppose $T \in \mathcal{L}(V, W)$ is injective and v_1, \dots, v_n is L.I. in V . Prove that Tv_1, \dots, Tv_n is L.I. in W .

4. Suppose v_1, \dots, v_n spans V and $T \in \mathcal{L}(V, W)$. Prove that Tv_1, \dots, Tv_n spans $\text{range } T$.
5. Suppose U is a 3-dimensional subspace of \mathbb{R}^8 and T is a linear map from \mathbb{R}^8 to \mathbb{R}^5 s.t. $\text{null } T = U$. Prove that T is surjective.
6. Suppose V and W are f -d. Prove that \exists an injective linear map from V to $W \iff \dim V \leq \dim W$.
7. Suppose U and V are f -d vector spaces, $S \in \mathcal{L}(V, W)$, and $T \in \mathcal{L}(U, V)$. Prove

$$\dim \text{null } ST \leq \dim \text{null } S + \dim \text{null } T.$$

8. Suppose U and V are f -d vector spaces, $S \in \mathcal{L}(V, W)$, and $T \in \mathcal{L}(U, V)$. Prove

$$\dim \text{range } ST \leq \min \{ \dim \text{range } S, \dim \text{range } T \}.$$

9. Suppose $D \in \mathcal{L}(\mathcal{P}(\mathbb{R}), \mathcal{P}(\mathbb{R}))$ s.t. $\deg Dp = (\deg p) - 1 \forall$ non-constant polynomial $p \in \mathcal{P}(\mathbb{R})$.

Remark. The notation D is used above to remind you of the differentiation map that sends a polynomial p to p' . Without knowing the formula for the derivative of a polynomial (except that it reduces the degree by 1), you can use the exercise above to show that for every polynomial $q \in \mathcal{P}(\mathbb{R})$, \exists a polynomial $p \in \mathcal{P}(\mathbb{R})$ s.t. $p' = q$.

10. Suppose $p \in \mathcal{P}(\mathbb{R})$. Prove that $\exists q \in \mathcal{P}(\mathbb{R})$ s.t. $5q'' + 3q' = p$.

Remark. This problem can be solved without using knowledge in Linear Algebra, but it is more interesting to solve with Linear Algebra.

11. Suppose $T \in \mathcal{L}(V, W)$ and let w_1, \dots, w_m be a basis of $\text{range } T$. Prove that $\exists \varphi_1, \dots, \varphi_m \in \mathcal{L}(V, \mathbb{F})$ s.t. $Tv = \varphi_1(v)w_1 + \dots + \varphi_m(v)w_m \quad \forall v \in V$.

10.6 Matrices

1. Suppose V and W are f -d and $T \in \mathcal{L}(V, W)$. Prove that for any basis in V and W , the matrix for T has at least $\dim \text{range } T$ non-zero entries.
2. If $S, T \in \mathcal{L}(V, W)$, then $\mathcal{M}(S + T) = \mathcal{M}(S) + \mathcal{M}(T)$.
3. Suppose $\lambda \in \mathbb{F}$ and $T \in \mathcal{L}(V, W)$. Then, $\mathcal{M}(\lambda T) = \lambda \mathcal{M}(T)$.

10.7 Invertibility and Isomorphism

1. Suppose $T \in \mathcal{L}(U, V)$ and $S \in \mathcal{L}(V, W)$ are invertible linear maps. Prove that $ST \in \mathcal{L}(U, W)$ is invertible and $(ST)^{-1} = T^{-1}S^{-1}$.
2. Suppose V is f -d and $\dim V > 1$. Prove that the set of non-invertible operators on V is not a subspace of $\mathcal{L}(V)$.
3. Suppose V is f -d and U is a subspace of V . Let $S \in \mathcal{L}(U, V)$. Prove that \exists invertible operator $T \in \mathcal{L}(V)$ s.t. $Tu = Su \quad \forall u \in U \iff S$ is injective.

4. Suppose W is f - d and $T_1, T_2 \in \mathcal{L}(V, W)$. Prove that $\text{null } T_1 = \text{null } T_2 \iff \exists$ invertible operator $S \in \mathcal{L}(W)$ s.t. $T_1 = ST_2$.
5. Suppose V is f - d and $T_1, T_2 \in \mathcal{L}(V, W)$. Prove that $\text{range } T_1 = \text{range } T_2 \iff \exists$ invertible operator $S \in \mathcal{L}(V)$ s.t. $T_1 = T_2S$.
6. Suppose V is f - d and $S, T \in \mathcal{L}(V)$. Prove that ST is invertible \iff both S and T are invertible.
7. Suppose V is f - d and $S, T \in \mathcal{L}(V)$. Prove $ST = I \iff TS = I$.
8. Suppose V is f - d and $S, T, U \in \mathcal{L}(V)$ s.t. $STU = I$. Prove T is invertible and $T^{-1} = US$.
9. Suppose V is f - d and $R, S, T \in \mathcal{L}(V)$ s.t. RST is a surjection. Prove that S is an injection.
10. Suppose v_1, \dots, v_n is a basis of V . Define a linear map $T : V \rightarrow \mathbb{F}^{n,1}$ as $Tv = \mathcal{M}(v)$, where $\mathcal{M}(v)$ is the matrix of $v \in V$ with respect to the basis v_1, \dots, v_n . Prove that T is an isomorphism from V to $\mathbb{F}^{n,1}$.
11. Prove that $V \cong \mathcal{L}(\mathbb{F}, V)$.

10.8 Duality

- 1.