

115 A Reminder Chapter I + II

Vector spaces

Definition. A vector space V over a field F is a set with two operations, addition and scalar multiplication, (so for any x, y in V and $a \in F$, $x+y$ and ax are in V) such that the following conditions hold.

$$(VS1) \quad x+y = y+x \quad \text{for all } x, y \text{ in } V \quad (\text{commutativity})$$

$$(VS2) \quad (x+y)+z = x+(y+z) \quad \text{for all } x, y, z \text{ in } V \quad (\text{associativity})$$

$$(VS3) \quad \text{There exists an element } 0 \text{ in } V \text{ such that } x+0 = x \text{ for all } x \in V. \quad (\text{identity})$$

$$(VS4) \quad \text{For each } x \text{ in } V \text{ there is an element } y \text{ in } V \text{ such that } x+y = 0 \text{ (} y \text{ is an inverse of } x \text{)}$$

$$(VS5) \quad 1 \cdot x = x \text{ for all } x \text{ in } V \quad (\text{where } 1 \text{ is the multiplicative identity of } F).$$

$$(VS6) \quad a(bx) = (ab)x \quad \text{for all } x \text{ in } V \text{ and } a, b \text{ in } F$$

$$(VS7) \quad a(x+y) = ax+ay \quad \text{for all } a \text{ in } F \text{ and } x, y \text{ in } V. \quad \left. \begin{array}{l} \\ \\ \end{array} \right\} \text{ distributive laws}$$

$$(VS8) \quad (a+b)x = ax+bx \quad \text{for all } a, b \text{ in } F \text{ and } x \text{ in } V.$$

Elements of V are called **vectors**.

Elements of F are called **scalars**.

Theorem 1.1 (Cancellation law)

Let V be a v.s. and let $x, y, z \in V$.

If $x+z = y+z$, then $x=y$.

Corollary.

1) In any vector space V , there is a unique element 0 satisfying (VS3) — the **zero vector** of V .

2) For any v.s. V and any x in V , there is a unique element y in V satisfying (VS4).

It is called the **inverse** of x , and denoted by $-x$.

Theorem 1.2. Let V be a v.s. over F .

For all x in V and a in F we have:

1) $0 \cdot x = 0$ (Note: the 1st 0 is a scalar in F , the 2nd one is the zero vector in V).

2) $(-a) \cdot x = -(ax) = a(-x)$

3) $a \cdot 0 = 0$ (Note: this is the zero vector of V on both sides).

Subspaces.

Definition. Let V be a v.s. A subset $W \subseteq V$ is a **subspace** of V if W itself is a v.s. with respect to the addition and scalar multiplication defined on V .

Theorem 1.3. Let V be a v.s., and let $W \subseteq V$ be a subset of V .

Then W is a subspace of V if and only if all of the following conditions hold:

(a) $0 \in W$

(b) $x+y \in W$ for all $x, y \in W$ (W is closed under addition)

(c) $c \cdot x \in W$ for all $c \in F$ and $x \in W$ (W is closed under scalar multiplication).

Theorem 1.4. Let V be a v.s. over F .

If W_1, \dots, W_n are subspaces of V , then the set $W = W_1 \cap W_2 \cap \dots \cap W_n$ is also a subspace of V .

Linear combinations

Definition. Let V be a v.s., and let $S \subseteq V$ be a non-empty subset of V .

1) A vector v in V is a **linear combination** of S if one can write

$$v = a_1 u_1 + a_2 u_2 + \dots + a_n u_n$$

for some vectors u_1, \dots, u_n in S and some scalars a_1, \dots, a_n in F .

2) The **span of S** , denoted $\text{Span}(S)$, is the subset of V consisting precisely of all linear combinations of S . That is,

$$\text{Span}(S) = \{a_1 u_1 + \dots + a_n u_n : n \in \mathbb{N}, a_i \in F, u_i \in S\}.$$
 For convenience, we define $\text{Span}(\emptyset) = \{0\}$.

Theorem 1.5. Let S be any subset of a v.s. V . Then:

- 1) $\text{Span}(S)$ is a subspace of V .
- 2) Any subspace of V that contains S must also contain $\text{Span}(S)$.

Definition. Let V be a v.s. and S a subset of V .
 We say that S **generates (or spans) V** if $\text{Span}(S) = V$.

Definition. A subset S of a v.s. V is **linearly dependent** if there exist a finite number of distinct vectors u_1, \dots, u_n in S and scalars $a_1, \dots, a_n \in F$, with at least one $a_i \neq 0$, such that

$$a_1 u_1 + \dots + a_n u_n = 0.$$

We call S **linearly independent** if it is not linearly dependent.

Theorem 1.6. Let V be a v.s. and $S_1, S_2 \subseteq V$ be two subsets of V .

- 1) If S_1 is lin. dependent, then S_2 is also linearly dependent.
- 2) If S_2 is lin. indep., then S_1 is also lin. indep.

Theorem 1.7. Let S be a lin. indep. subset of a vector space V .

Let v be any vector in V **not contained in S** .

Then $S \cup \{v\}$ is lin. dep. if and only if $v \in \text{Span}(S)$.

Bases and dimension.

Definition. A **basis** for a v.s. V is a subset of V which is lin. indep. and generates V .

Theorem 1.8. A subset $\{u_1, \dots, u_n\}$ of a v.s. V is a basis if and only if every vector $v \in V$ can be written **uniquely** in the form

$$v = a_1 u_1 + \dots + a_n u_n,$$

where $a_i \in F$.

(so "uniquely" here means that there is only one possible choice of the scalars $a_1, \dots, a_n \in F$ satisfying the equality.)

Theorem 1.9. If a v.s. V is generated by a finite subset S , then some subset of S is a basis for V .
 It follows that every finitely generated v.s. has a basis.

Theorem 1.10. (Replacement Theorem)

Let V be a v.s. generated by a set $G \subseteq V$ with $|G| = n$, and let L be a lin. indep. subset of V , $|L| = m$.
 Then $m \leq n$, and there exists $H \subseteq G$ with $|H| = n - m$ such that $L \cup H$ generates V .

Corollary 1. Let V be a finitely generated v.s. Then every basis for V has the same number of elements.

Definition. A v.s. V is **finite-dimensional** if it has a finite basis.

The (unique) number of vectors in a basis for V is called the **dimension of V** , denoted $\dim(V)$.

If there is no finite basis, then V is **infinite-dimensional**.

Corollary 2. Let V be a v.s. of dimension n . Then:

- Any generating set for V must contain at least n vectors.
- Any lin. indep. subset of V with n elements is a basis.
- Every lin. indep. subset of V can be extended to a basis for V .

Theorem 1.11. Let W be a subspace of a v.s. V with $\dim(V) < \infty$.
 Then $\dim(W) \leq \dim(V)$.
 Moreover, if $\dim(W) = \dim(V)$, then $V = W$.

Linear transformations

Def Let V and W be v.s. over the same field of scalars F .

A lin. transformation from V to W is a function $T: V \rightarrow W$ satisfying

- $T(x+y) = T(x) + T(y)$ for all $x, y \in V$.
- $T(cx) = cT(x)$ for all $x \in V$ and $c \in F$.

Properties of lin. transformations

1) Let $T: V \rightarrow W$ be a lin. transf. Then:

- $T(0) = 0$
- $T\left(\sum_{i=1}^n a_i x_i\right) = \sum_{i=1}^n a_i T(x_i)$ for all $x_i \in V$, $a_i \in F$.

2) A function $T: V \rightarrow W$ is a lin. transf. $\Leftrightarrow T(cx+y) = cT(x) + T(y)$ for all $x, y \in V$, $c \in F$.

Thm 2.6 Let V, W be v.s. over a field F , and let $\{v_1, \dots, v_n\}$ be a basis for V .
 Then for any vectors $w_1, \dots, w_n \in W$ there exists exactly one lin. transf. $T: V \rightarrow W$ s.t.
 $T(v_i) = w_i$ for $1 \leq i \leq n$.

Def Let $T: V \rightarrow W$ be a lin. transf.

- T is **injective** if $T(v) = T(u)$ implies $v = u$, for all $u, v \in V$.
- T is **surjective** if for every $w \in W$ there is some $v \in V$ s.t. $T(v) = w$.
- T is **bijective** if it is both injective and surjective.

Null space and range

Def Let V, W be v.s. and $T: V \rightarrow W$ a lin. transf.

- The **null space** of T is defined as
 $N(T) = \{x \in V : T(x) = 0\}$.
- The **range** of T is the image of V under T , that is the set
 $R(T) = \{y \in W : y = T(x) \text{ for some } x \in V\}$.

Thm 2.1

- $N(T)$ is a subspace of V .
- $R(T)$ is a subspace of W .

Thm 2.4 Let $T: V \rightarrow W$ be a lin. transf.

- T is injective $\Leftrightarrow N(T) = \{0\}$.
- T is surjective $\Leftrightarrow R(T) = W$.

Thm 2.2 Let $T: V \rightarrow W$ be a lin. transf.

Assume $\beta = \{v_1, \dots, v_n\}$ is a basis for V .
 Then $R(T) = \text{Span}(\{T(v_1), \dots, T(v_n)\})$.

Thm 2.3 (Dimension Theorem)

Let V, W be v.s., $T: V \rightarrow W$ a lin. transf., and $\dim(V) < \infty$. Then
 $\dim(V) = \dim(N(T)) + \dim(R(T))$.

Thm 2.5 Let $T: V \rightarrow W$ be a lin. transf., and assume $\dim(V) = \dim(W)$.

Then the following are equivalent:

- 1) T is injective.
- 2) T is surjective.
- 3) T is bijective.
- 4) $\dim(\mathcal{R}(T)) = \dim(V)$.

The vector space of linear transformations $\mathcal{L}(V, W)$

Def. Let V, W be v.s. over F , and let $T, U: V \rightarrow W$ be linear transformations.

Then we define the functions $T+U$ and aT , for every $a \in F$, by:

$$(T+U)(x) = T(x) + U(x) \text{ for all } x \in V.$$

$$(aT)(x) = a \cdot T(x) \text{ for all } x \in V.$$

Thm 2.7

If T and U are linear, then $T+U$ and aT are also linear.

Def. We denote the set of all lin. transf.'s from V to W by $\mathcal{L}(V, W)$.

Then it is a v.s. over F , with the operations of addition and scalar multiplication described above.

When $W = V$, we write $\mathcal{L}(V)$ instead of $\mathcal{L}(V, V)$.

Matrix representation of a lin. transf.

Def. Let V be a v.s. with $\dim(V) < \infty$. An **ordered basis** for V is a basis for V with a specified order on its vectors.

Def. Let $\beta = \{v_1, \dots, v_n\}$ be an ordered basis for V .

Then any vector $x \in V$ can be written as

$$x = a_1 v_1 + \dots + a_n v_n \text{ for some unique scalars } a_1, \dots, a_n \in F.$$

We define the **coordinate vector** of x relative to β by

$$[x]_{\beta} = \begin{pmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{pmatrix} \in F^n.$$

Def. Let V, W be v.s. with ordered bases $\beta = \{v_1, \dots, v_n\}$ and $\gamma = \{w_1, \dots, w_m\}$, respectively.

Let $T: V \rightarrow W$ be a lin. transformation.

Then the **matrix representation** of T in the ordered bases β and γ is defined as the matrix $[T]_{\beta}^{\gamma} \in M_{m \times n}(F)$ given by

$$[T]_{\beta}^{\gamma} = \begin{pmatrix} | & | & & | \\ [T(v_1)]_{\gamma} & [T(v_2)]_{\gamma} & \dots & [T(v_n)]_{\gamma} \\ | & | & & | \end{pmatrix},$$

where $[T(v_i)]_{\gamma}$ are the coordinates of the vector $T(v_i) \in W$ with respect to the basis γ .

If $V = W$ and $\beta = \gamma$, we simply write $[T]_{\beta}$.

Thm 2.8

Let V, W be fin. dim. v.s.'s with ordered bases β and γ , resp.

Let $T, U: V \rightarrow W$ be lin. transformations. Then:

- 1) $U = T$ (meaning that $U(x) = T(x)$ for all $x \in V$) $\iff [U]_{\beta}^{\gamma} = [T]_{\beta}^{\gamma}$.
- 2) $[T+U]_{\beta}^{\gamma} = [T]_{\beta}^{\gamma} + [U]_{\beta}^{\gamma}$.
- 3) $[aT]_{\beta}^{\gamma} = a \cdot [T]_{\beta}^{\gamma}$ for all $a \in F$.

Composition of lin. transfs and matrix multiplication.

Def Let V, W, Z be v.s.'s over F . Let $T: V \rightarrow W$ and $U: W \rightarrow Z$ be lin. transfs.

Their **composition** is the function UT , from V to Z , defined by

$$(UT)(x) = U(T(x)) \text{ for all } x \in V.$$

Thm 2.9 If T and U are linear, then UT is also linear.

Def Given matrices $A \in M_{m \times n}(F)$ and $B \in M_{n \times p}(F)$, we define the **product** $AB \in M_{m \times p}(F)$ to be the matrix with the entries

$$(AB)_{ij} = \sum_{k=1}^n A_{ik} \cdot B_{kj}, \text{ for } 1 \leq i \leq m, 1 \leq j \leq p.$$

Thm 2.11 Let V, W, Z be fin. dim. v.s.'s with ordered bases α, β, γ respectively.

Let $T: V \rightarrow W$ and $U: W \rightarrow Z$ be lin. transformations. Then

$$[UT]_{\alpha}^{\gamma} = [U]_{\beta}^{\gamma} [T]_{\alpha}^{\beta}.$$

Corollary Let V be a fin. dim. v.s. with an ordered basis β .

Let $T, U \in \mathcal{L}(V)$.

$$\text{Then } [UT]_{\beta} = [U]_{\beta} [T]_{\beta}.$$

Thm 2.14 Let V, W be fin. dim. v.s.'s with ordered bases β and γ , resp.

Let $T: V \rightarrow W$ be a lin. transft. Then for each vector $u \in V$ we have:

$$[T(u)]_{\gamma} = [T]_{\beta}^{\gamma} [u]_{\beta}.$$

(so, we calculate the coordinates of the vector $T(u)$ from the coordinates of the vector u).

Def To every matrix $A \in M_{m \times n}(F)$, we associate a **linear transformation** $L_A: F^n \rightarrow F^m$ defined by

$$L_A(x) = Ax \text{ for every (column) vector } x \in F^n.$$

We call L_A the **left-multiplication transformation**.

Invertibility

Def Let V, W be v.s.'s and $T: V \rightarrow W$ linear.

1) A lin. transft. $U: W \rightarrow V$ is the **inverse of T** if

$$UT = I_W \text{ and } TU = I_V.$$

2) T is **invertible** if it has an inverse.

Basic facts

1) If T is invertible, then its inverse is **unique**, and is denoted by T^{-1} .

2) T is invertible $\Leftrightarrow T$ is a bijection.

3) If T, U are invertible, then

$$\bullet (TU)^{-1} = U^{-1}T^{-1}$$

$$\bullet (T^{-1})^{-1} = T.$$

Lemma Let $T: V \rightarrow W$ be lin. and invertible, and $\dim(V) < \infty$.

Then $\dim(V) = \dim(W)$.

Def A matrix $A \in M_{n \times n}(F)$ is **invertible** if there exist $B \in M_{n \times n}(F)$ s.t. $AB = BA = I$.

If such a B exists, then it is **unique**, called the **inverse of A** and denoted by A^{-1} .

Thm 2.18 Let V, W be fin. dim. v.s.'s with ordered bases β and γ , resp.

Let $T: V \rightarrow W$ be lin.

Then T is invertible \Leftrightarrow the matrix $[T]_{\beta}^{\gamma}$ is invertible.

Furthermore, $[T^{-1}]_{\gamma}^{\beta} = ([T]_{\beta}^{\gamma})^{-1}$.

Isomorphisms

Def Two v.s.'s V and W are **isomorphic** if there exists an invertible lin. transf. $T: V \rightarrow W$. Such a T is called an **isomorphism** from V onto W .

Thm 2.19 Two fin. dim. v.s.'s V and W are isomorphic $\Leftrightarrow \dim(V) = \dim(W)$.

Corollary Let V be a v.s. over F . Then V is isomorphic to F^n if and only if $\dim(V) = n$.

Thm 2.20 Let V, W be v.s.'s over F , $\dim(V) = n$, $\dim(W) = m$.

Let β, γ be ordered bases for V, W , resp.

Then the map $\phi: \mathcal{L}(V, W) \rightarrow M_{m \times n}(F)$ defined by

$$\phi(T) = [T]_{\beta}^{\gamma} \text{ for all } T \in \mathcal{L}(V, W)$$

is an isomorphism.

Corollary If $\dim(V) = n$, $\dim(W) = m$ then $\dim(\mathcal{L}(V, W)) = \dim(M_{m \times n}(F)) = mn$.

Dual spaces

Definition Let V be a vector space over a field of scalars F (which is itself a vector space of dim 1 over F). A linear transformation from V to F is called a **linear functional on V** .

Ex 1 Let V be the v.s. ^{over $F = \mathbb{R}$} of continuous real-valued functions on the interval $[0, 2\pi]$. Fix a function $g \in V$. Then the function $h: V \rightarrow \mathbb{R}$ defined by $h(x) = \frac{1}{2\pi} \int_0^{2\pi} x(t)g(t) dt$ for $x \in V$ is a lin. functional on V . } - Definite integral is one of the most important lin. functionals!

Ex 2 Let $V = M_{n \times n}(F)$, and define $f: V \rightarrow F$ by $f(A) = \text{tr}(A)$, the trace of the matrix $A \in V$. (recall: $\text{tr}(A) = A_{11} + A_{22} + \dots + A_{nn}$). Then f is a lin. functional.

Ex 3 Let V be a finite dim. v.s., and let $\beta = \{x_1, \dots, x_n\}$ be an ordered basis for V . For each $i = 1, 2, \dots, n$ we define $f_i(x) = a_i$, where $[x]_{\beta} = \begin{pmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{pmatrix}$ is the coordinate vector of x relative to β . Then each f_i is a lin. functional on V called the **i^{th} coordinate function w.r.t. the basis β** .

Def For a v.s. V over F , we define the **dual space of V** to be the v.s. $\mathcal{L}(V, F)$, denoted by **V^*** . We define the **double dual V^{**}** of V to be the dual of V^* .

Thus V^* is the v.s. consisting of all lin. functionals on V .

Remark If V is finite dimensional, then by Theorem 2.20 we have:

$$\dim(V^*) = \dim(\mathcal{L}(V, F)) = \dim(M_{\dim(V) \times \dim(F)}(F)) = \dim(V) \cdot \dim(F) = \dim(V) \cdot 1 = \dim(V).$$

Hence, by Thm 2.19, V and V^* are isomorphic. (this can be false when V is infinite dimensional!)

Thm 2.24 Supp. that V is a fin. dim. v.s. over F with the ordered basis $\beta = \{x_1, \dots, x_n\}$. Let $f_i, 1 \leq i \leq n$ be the i^{th} coordinate function w.r.t. β (as in Example 3 above). Let $\beta^* = \{f_1, \dots, f_n\}$.

Then β^* is an ordered basis for V^* , and for any $f \in V^*$ we have $f = \sum_{i=1}^n f(x_i) f_i$.

Proof. Let $f \in V^*$ be arbitrary. Since $\dim(V^*) = n$ by the remark above, we only need to show that $f = \sum_{i=1}^n f(x_i) f_i$ (as this means that β^* generates V^* , and is of size n , hence is a basis for V^* by the Replacement Theorem).

Let $g = \sum_{i=1}^n f(x_i) f_i$. Then for $1 \leq j \leq n$ we have: $g(x_j) = \left(\sum_{i=1}^n f(x_i) f_i \right) (x_j) = \sum_{i=1}^n f(x_i) f_i(x_j)$.

Since by definition, $f_i(x_j) = \begin{cases} 1 & \text{if } i=j \\ 0 & \text{if } i \neq j \end{cases}$, we get $g(x_j) = f(x_j)$.

Hence f and g are two lin. transformations that agree on every vector in a basis, hence they are equal on the whole space (by Corollary to Thm 2.6).

Def Using the notation of Thm 2.24, we call the ordered basis $\beta^* = \{f_1, \dots, f_n\}$ of V^* that satisfies $f_i(x_j) = \begin{cases} 1 & \text{if } i=j \\ 0 & \text{if } i \neq j \end{cases}$ the **dual basis of β** .

Ex 4 Let $\beta = \{(2,1), (3,1)\}$ be an ordered basis for $V = \mathbb{R}^2$. Suppose that the dual basis of β is given by $\beta^* = \{f_1, f_2\}$. To determine a formula for f_1 , we know that by def. it must satisfy the equations:

$$1 = f_1(2,1) = f_1(2e_1 + e_2) = 2f_1(e_1) + f_1(e_2),$$

$$0 = f_1(3,1) = f_1(3e_1 + e_2) = 3f_1(e_1) + f_1(e_2).$$

Solving the equations, we obtain $f_1(e_1) = -1$ and $f_1(e_2) = 3$. Hence $f_1(x,y) = -x + 3y$.

Similarly, we get $f_2(x,y) = x - 2y$.

Assume now that V, W are fin. dim. v.s. over F with ordered bases β and γ , respectively, $\dim(V) = m, \dim(W) = n$. Recall (Section 2.4): there exists a one-to-one correspondence between lin. transformations $T: V \rightarrow W$ and $m \times n$ matrices over F given by $T \leftrightarrow [T]_{\beta}^{\gamma}$.

Question: Given a matrix $A = [T]_{\beta}^{\gamma}$, when is it possible to find a lin. transformation U represented by the matrix A^t in some basis?

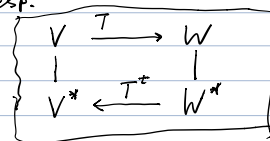
Of course, if $m \neq n$ then it is impossible for U to be a lin. transformation from V to W . Dual spaces help!

Thm 2.25 Let V and W be fin. dim. v.s. over F with ordered bases β and γ , resp.

For any lin. transformation $T: V \rightarrow W$, the map $T^t: W^* \rightarrow V^*$ defined by

$$T^t(g) = gT \text{ for all } g \in W^*$$

is a lin. transformation such that $[T^t]_{\beta^*}^{\gamma^*} = ([T]_{\beta}^{\gamma})^t$.



Proof.

1) For any $g \in W^*$, we defined $T^t(g) = gT$, i.e. the composition of linear maps $T: V \rightarrow W$ and $g: W \rightarrow F$. Hence $T^t(g)$ is a linear map from $V \rightarrow F$, so it is an element of V^* . Thus indeed T^t maps W^* into V^* . Given any $g, h \in W^*$ and $a \in F$, we have that for any $x \in V$:

$$T^t(\underbrace{ag+h}_{g \in W^*})(x) \stackrel{\text{def}}{=} (ag+h)(T(x)) = (ag+h)(T(x)) = a(g(T(x))) + h(T(x)) \stackrel{\text{def}}{=} (a \cdot T^t(g))(x) + (T^t(h))(x),$$

hence $T^t(ag+h) = aT^t(g) + T^t(h)$, so $T^t: W^* \rightarrow V^*$ is a lin. transformation.

2) Let $\beta = \{x_1, \dots, x_n\}$ and $\gamma = \{y_1, \dots, y_m\}$ be as given, with dual bases $\beta^* = \{f_1, \dots, f_n\}$ and $\gamma^* = \{g_1, \dots, g_m\}$, resp. Let $A = [T]_{\beta}^{\gamma} = (A_{ij})$.

To find the j^{th} column of the matrix $[T^t]_{\beta^*}^{\gamma^*}$, we begin by expressing the vector $T^t(g_j) \in V^*$ as a linear combination of the vectors of β^* . By Thm 2.24 we have:

$$T^t(g_j) = g_j T = \sum_{i=1}^n (g_j T)(x_i) f_i.$$

So the row i , column j entry of the matrix $[T^t]_{\beta^*}^{\gamma^*}$ is

$$(g_j T)(x_i) = g_j(T(x_i)) = g_j\left(\sum_{k=1}^m A_{ki} y_k\right) = \sum_{k=1}^m A_{ki} g_j(y_k) = A_{ji}.$$

Hence $[T^t]_{\beta^*}^{\gamma^*} = A^t$.

$$= 1 \text{ if } j=k \\ = 0 \text{ if } j \neq k$$