

## Linear Algebra

**Definition.** A vector space (over  $\mathbf{R}$ ) is an ordered quadruple

$$(V, \mathbf{0}, \alpha, \mu)$$

such that  $V$  is a set;  $\mathbf{0} \in V$ ;

$$\alpha : V \times V \rightarrow V \quad \text{and} \quad \mu : \mathbf{R} \times V \rightarrow V;$$

and the following eight axioms hold:

- (i)  $\alpha(\alpha(u, v), w) = \alpha(u, \alpha(v, w))$ ,  $u, v, w \in V$ ;
- (ii)  $\alpha(v, \mathbf{0}) = v = \alpha(\mathbf{0}, v)$ ,  $v \in V$ ;
- (iii) for each  $v \in V$  there is  $w \in V$  such that  $\alpha(v, w) = \mathbf{0} = \alpha(w, v)$ ;
- (iv)  $\alpha(u, v) = \alpha(v, u)$ ,  $u, v \in V$ ;
- (v)  $\mu(c + d, v) = \mu(c, v) + \mu(d, v)$ ,  $c, d \in \mathbf{R}$ ,  $v \in V$ ;
- (vi)  $\mu(c, \alpha(u, v)) = \alpha(\mu(c, u), \mu(c, v))$ ,  $c \in \mathbf{R}$ ,  $u, v \in V$ ;
- (vii)  $\mu(c, \mu(d, v)) = \mu(cd, v)$ ,  $c, d \in \mathbf{R}$ ,  $v \in V$ ;
- (viii)  $\mu(1, v) = v$ ,  $v \in V$ .

Axioms (i),(ii),(iii) say that  $(V, \mathbf{0}, \alpha)$  is an Abelian group. Axiom (iv) says that this group is Abelian. One calls the elements of  $V$  **vectors**. From now on we write

$$u + v$$

for  $\alpha(u, v)$  and call this operation **vector addition**, and we write

$$cv$$

for  $\mu(c, v)$ , with the latter binding more tightly than the former, and call this operation **scalar multiplication**. If  $\mathbf{0}_i$ ,  $i = 1, 2$  satisfy Axiom (ii) with  $\mathbf{0}$  there replaced by  $\mathbf{0}_i$ ,  $i = 1, 2$ , respectively, then

$$\mathbf{0}_1 = \mathbf{0}_1 + \mathbf{0}_2 = \mathbf{0}_2$$

so this element is unique; we call it the **zero** element of  $V$ . If  $w_i$ ,  $i = 1, 2$  satisfy Axiom (iii) for a given  $v \in V$  with  $w$  there replaced by  $w_i$ ,  $i = 1, 2$ , respectively, then

$$w_1 = w_1 + \mathbf{0} = w_1 + (v + w_2) = (w_1 + v) + w_2 = \mathbf{0} + w_2 = w_2$$

so the element  $\mathbf{w}$  is uniquely determined; we denote it

$$-v.$$

We also write

$$u - v$$

for  $u + (-v)$ ,  $u, v \in V$ . For any  $v \in V$  we have

$$0v = \mathbf{0} + 0v = (-0v + 0v) + 0v = -0v + (0v + 0v) = -0v + (0 + 0)v = -0v + 0v = \mathbf{0};$$

that is

$$0v = \mathbf{0}, \quad v \in V.$$

**Example.** Suppose  $S$  is a nonempty set. Then  $\mathbf{R}^S$  is a vector space where, given  $f, g \in \mathbf{R}^S$  and  $c \in \mathbf{R}$ , we set

$$(f + g)(s) = f(s) + g(s) \quad \text{and} \quad (cf)(s) = cf(s), \quad s \in S.$$

We call these operations **pointwise addition** and **pointwise scalar multiplication**, respectively.

**Example.** Since  $\mathbf{R}^n = \mathbf{R}^{\{1, \dots, n\}}$ , it is a vector space by virtue of the previous Example.

**Example.**  $\mathbf{R}$  is a vector space where vector addition is addition and where scalar multiplication is multiplication.

**Example.** Suppose  $V$  is a vector space and  $S$  is a nonempty set. Then  $V^S$  is a vector space where, given  $f, g \in V^S$  and  $c \in \mathbf{R}$ , we set

$$(f + g)(s) = f(s) + g(s) \quad \text{and} \quad (cf)(s) = cf(s), \quad s \in S.$$

We call these operations **pointwise addition** and **pointwise scalar multiplication**, respectively.

**Definition.** Suppose  $V$  is a vector space. We say a subset  $U$  of  $V$  is a **linear subspace (of  $V$ )**

- (i) if  $\mathbf{0} \in U$ ;
  - (ii)  $u + v \in U$  whenever  $u, v \in U$ ;
  - (iii)  $cu \in U$  whenever  $c \in \mathbf{R}$  and  $u \in U$ .
- Note that  $(U, \mathbf{0}, \alpha|(U \times U), \mu|(\mathbf{R} \times U))$  is a vector space.

**Proposition.** Suppose  $V$  is a vector space and  $\mathcal{U}$  is a nonempty family of linear subspaces of  $V$ . Then  $\bigcap \mathcal{U}$  is a linear subspace of  $V$ .

**Remark.** If  $\mathcal{U} = \emptyset$  then  $\bigcup \mathcal{U} = \emptyset$  and  $\bigcap \mathcal{U}$  is problematic.

**Proof.** Simple exercise for the reader.  $\square$

**Definition.** Suppose  $V$  and  $W$  are vector spaces and  $L : V \rightarrow W$ . We say  $L$  is **linear** if

- (i)  $L(v + w) = L(v) + L(w)$  whenever  $v, w \in V$ ;
- (ii)  $L(cv) = cL(v)$  whenever  $c \in \mathbf{R}$  and  $v \in V$ .

Note that the operations on the left are with respect to  $V$  and the operations on the right are with respect to  $W$ . We say  $L$  **carries  $V$  isomorphically onto  $W$**  if  $L$  is univalent and  $\text{rng } L = W$ .

We set

$$\ker L = \{v \in V : L(v) = \mathbf{0}\}$$

and call this subset of  $V$  the **kernel** or **null space** of  $L$ .

We let

$$\mathbf{L}(V; W)$$

be the set of linear maps from  $V$  into  $W$ . Note that  $\mathbf{L}(V; W)$  is a linear subspace of  $W^V$  and is therefore a vector space with respect to the operations of pointwise addition and scalar multiplication.

Of particular importance is the case when  $W = \mathbf{R}$ . We set

$$V^* = \mathbf{L}(V; \mathbf{R})$$

and call this vector space the **dual space** of  $V$ .

Suppose  $\omega \in V^*$  and  $w \in W$ . Define  $\omega w : V \rightarrow W$  by setting

$$\omega w(v) = \omega(v)w, \quad v \in V.$$

Note that  $\omega w$  is linear.

**Proposition.** Suppose  $V$  and  $W$  are vector spaces and  $L : V \rightarrow W$  is linear. Then

- (i)  $L(\mathbf{0}) = \mathbf{0}$ ;

- (ii)  $\ker L$  is a linear subspace of  $V$ ;
- (iii)  $L$  is univalent if and only if  $\ker L = \mathbf{0}$ ;
- (iv)  $\text{rng } L$  is a linear subspace of  $W$ .

**Proof.** Simple exercise which for the reader.

**Definition.** Suppose  $V$  is a vector space and  $S$  is a nonempty set. For each  $f \in V^S$  we set

$$\mathbf{spt } f = \{s \in S : f(s) \neq \mathbf{0}\}$$

and call the set the **support of  $f$** . We let

$$(V^S)_0 = \{f \in V^S : \mathbf{spt } f \text{ is finite}\}.$$

Note that

$$V^\emptyset = (V^\emptyset)_0 = \emptyset.$$

**Remark.** Suppose  $V$  is a vector space and  $S$  is a nonempty set. Since  $V$  is an Abelian group we know that

$$(V^S)_0 = \{f \in V^S : \mathbf{spt } f \text{ is finite}\}$$

is a subgroup of the Abelian group  $V^S$  and that there is one and only group homomorphism

$$\sum \cdot : (V^S)_0 \rightarrow V$$

such that  $\sum (s, v) = v$  whenever  $(s, v) \in S \times V$ . It is immediate that  $(V^S)_0$  is a linear subspace of  $V^S$ . We leave as a straightforward exercise for the reader to prove by induction that  $\sum \cdot$  is linear.

**Definition.** Suppose  $V$  is a vector space and  $S \subset V$ . In case  $S \neq \emptyset$  we define

$$\mathbf{s} : (\mathbf{R}^S)_0 \rightarrow V$$

by setting

$$\mathbf{s}(f) = \sum_{s \in S} f(s)s, \quad f \in (\mathbf{R}^S)_0.$$

Note that  $\mathbf{s}$  is linear because it is the composition of  $\sum$  with the linear map  $(\mathbf{R}^S)_0 \ni f \mapsto (S \ni s \mapsto f(s)s \in V) \in (V^S)_0$ . We let

$$\mathbf{span } S = \begin{cases} \{\mathbf{0}\} & \text{if } S = \emptyset, \\ \mathbf{rng } \mathbf{s} & \text{else} \end{cases}$$

and call this linear subspace of  $V$  the **(linear) span of  $S$** . We say  $S$  is **independent** if either  $S = \emptyset$  or  $S \neq \emptyset$  and  $\mathbf{s}$  is univalent. We say  $S$  is **dependent** if  $S$  is not independent. We say  $S$  is **basis for  $V$**  if  $S$  is independent and  $V = \mathbf{span } S$ . Evidently,

- (i) the empty set is independent;
- (ii) if  $\mathbf{0} \in S$  then  $S$  is dependent;
- (iii) a superset of a dependent set is dependent;
- (iv) a subset of an independent set is independent.

**Proposition.** Suppose  $V$  is a vector space and  $S \subset V$ . Then

$$\mathbf{span } S = \bigcap \{U : U \text{ is a linear subspace of } V \text{ and } S \subset U\}.$$

**Proof.** If  $U$  is a linear subspace of  $V$  and  $S \subset U$  then  $\mathbf{span } S \subset U$ . On the other hand,  $\mathbf{span } S$  is a linear subspace of  $V$  and  $S \subset \mathbf{span } S$ .  $\square$

**Definition.** Suppose  $V$  is a vector space and  $\mathcal{U}$  is a family of linear subspaces of  $V$ . Let

$$\sum \mathcal{U} = \mathbf{span} \bigcup \mathcal{U}.$$

**Proposition.** Suppose  $V$  is a vector space and  $S \subset V$ . Then  $S$  is dependent if and only if there is  $s_0 \in S$  such that  $s_0 \in \mathbf{span}(S \sim \{s_0\})$ .

**Proof.** Suppose  $S$  is dependent. Then  $S \neq \emptyset$  and there is  $f \in (\mathbf{R}^S)_0$  such that  $f$  is nonzero and  $\sum_{s \in S} f(s)s = \mathbf{0}$ . For any  $s_0 \in \mathbf{spt} f$  we have

$$f(s_0)s_0 + \sum_{s \in S \sim \{s_0\}} f(s)s = \mathbf{0}$$

so that

$$s_0 = -\frac{1}{f(s_0)} \sum_{s \in S \sim \{s_0\}} f(s)s \in \mathbf{span} S \sim \{s_0\}.$$

On the other hand, if  $s_0 \in S$  and  $s_0 \in \mathbf{span}(S \sim \{s_0\})$  then  $s_0 = \sum_{s \in S \sim \{s_0\}} g(s)s$  for some  $g \in (\mathbf{R}^{S \sim \{s_0\}})_0$ . Let  $f \in (\mathbf{R}^S)_0$  be such that

$$f(s) = \begin{cases} -1 & \text{if } s = s_0, \\ g(s) & \text{if } s \in S \sim \{s_0\}. \end{cases}$$

Then  $f$  is nonzero and  $\sum_{s \in S} f(s)s = \mathbf{0}$  so  $f \in \mathbf{ker} s$ . Thus  $S$  is dependent.  $\square$

**Proposition.** Suppose  $V$  is a vector space  $S$  is an independent subset of  $V$  and  $v \in V \sim \mathbf{span} S$ . Then  $S \cup \{v\}$  is independent.

**Proof.** Were  $S \cup \{v\}$  dependent there would be  $c \in \mathbf{R}$  and  $f \in (\mathbf{R}^S)_0$  such that not both  $c$  and  $f$  are zero and

$$cv + \sum_{s \in S} f(s)s = \mathbf{0}.$$

But  $c \neq 0$  since  $S$  is independent. Thus

$$v = -\frac{1}{c} \sum_{s \in S} f(s)s \in \mathbf{span} S$$

which is a contradiction.  $\square$

**Corollary.** Suppose  $V$  is a vector space. Any maximal independent subset of  $V$  is a basis for  $V$ .

**Proof.** This is immediate.  $\square$

**Theorem.** Suppose  $V$  is a vector space. Then  $V$  has a basis.

**Proof.** Suppose  $\mathcal{S}$  is a nested family of independent subsets of  $V$ . Then  $\bigcup \mathcal{S}$  is independent. Thus, by the Hausdorff Maximal Principle, there is a maximal independent subset of  $V$ .  $\square$

**Remark.** Suppose  $V = \mathbf{span} S$  where  $S$  is a finite subset of  $V$ . Then  $S$  has a maximal independent subset which, by the previous Proposition is a basis for  $V$ . Thus, in this case, we can avoid using the Hausdorff Maximal Principle to show that  $V$  has a basis.

**Corollary.** Suppose  $V$  is a vector space and  $S$  is an independent subset of  $V$  then  $S$  is a subset of a basis for  $V$ .

**Proof.** Argue as in the proof of the preceding Corollary that there is a maximal independent subset of  $V$  which contains  $S$ .  $\square$

**Definition.** Suppose  $V$  is a nontrivial vector space and  $S$  is a basis for  $V$ . We define

$$\cdot^* : S \rightarrow V^*$$

at  $s \in S$  by requiring that

$$s^*(\mathbf{s}(f)) = f(s), \quad f \in (\mathbf{R}^S)_0.$$

One easily verifies that that for any  $v \in \mathbf{span} S$  the set  $\{s \in S : s^*(v) \neq 0\}$  is finite and that

$$(2) \quad v = \sum_{s \in S} s^*(v)s, \quad v \in V;$$

simply represent  $v$  by  $\mathbf{s}(f)$  for some  $f \in (\mathbf{R}^S)_0$ .

If the set  $S$  is indexed by the set  $A$  we will frequently write

$$s^a \quad \text{for} \quad s_a^* \quad \text{whenever} \quad a \in A.$$

**Theorem.** Suppose  $V$  is a vector space and  $T$  is a finite independent subset of  $V$ . If  $S \subset \mathbf{span} T$  and  $\mathbf{card} S > \mathbf{card} T$  then  $S$  is dependent.

**Proof.** We induct on  $\mathbf{card} T$ . The Theorem holds trivially in case  $\mathbf{card} T = 0$ .

Suppose  $\mathbf{card} T > 0$  and choose  $\tilde{t} \in T$ . Then

$$(3) \quad v = \tilde{t}^*(v)\tilde{t} + \sum_{t \in T \sim \{\tilde{t}\}} t^*(v)t, \quad v \in \mathbf{span} T.$$

In case  $\tilde{t}^*(s) = 0$  for all  $s \in S$  we infer from (3) that  $S \subset \mathbf{span}(T \sim \{\tilde{t}\})$  which implies by the inductive hypothesis that  $S$  is dependent.

So suppose  $\tilde{s} \in S$  and  $\tilde{t}^*(\tilde{s}) \neq 0$ . Define  $F : S \sim \{\tilde{s}\} \rightarrow V$  by letting

$$F(s) = s - \frac{\tilde{t}^*(s)}{\tilde{t}^*(\tilde{s})}\tilde{s}, \quad s \in S \sim \{\tilde{s}\};$$

we infer from (3) and the linearity of  $\tilde{t}^*$  that

$$(4) \quad S' \subset \mathbf{span}(T \sim \{\tilde{t}\}).$$

where we have set  $S' = \mathbf{rng} F$ .

Suppose  $F$  is not univalent. Choose  $s_i \in S$ ,  $i = 1, 2$ , such that  $s_1 \neq s_2$  and  $F(s_1) = F(s_2)$ . Then

$$s_1 - s_2 - \frac{\tilde{t}^*(s_1 - s_2)}{\tilde{t}^*(\tilde{s})}\tilde{s} = \mathbf{0}$$

which implies  $S$  is dependent.

Suppose  $F$  is univalent. Then

$$\mathbf{card} S' = \mathbf{card} S - 1 > \mathbf{card} T - 1 = \mathbf{card}(T \sim \{\tilde{t}\}).$$

By (4) and the inductive hypothesis we infer that  $S'$  is dependent. Thus there is  $f \in (\mathbf{R}^{S \sim \{\tilde{s}\}})$  such that  $f$  is nonzero and

$$\sum_{s \in S \sim \{\tilde{s}\}} f(s)F(s) = \mathbf{0}.$$

But this implies that

$$\sum_{s \in S \sim \{\tilde{s}\}} f(s)s - \frac{\tilde{t}^*(\sum_{s \in S \sim \{\tilde{s}\}} f(s)s)}{\tilde{t}^*(\tilde{s})}\tilde{s} = \mathbf{0}$$

so  $S$  is dependent.  $\square$

**Theorem.** Suppose  $V$  is a vector space. Then any two bases have the same cardinality.

**Remark.** An infinite basis is not a very useful thing. At least that's my opinion.

**Proof.** This is a direct consequence of the previous Theorem if  $V$  has a finite basis.

More generally, Suppose  $A$  and  $B$  are bases for  $V$  and  $B$  is infinite. Let  $F$  be the set of finite subsets of  $B$ . Define  $f : A \rightarrow F$  by letting

$$f(a) = \{b \in B : b^*(a) \neq 0\}, \quad a \in A.$$

By the preceding Theorem we find that

$$\mathbf{card} \{a \in A : f(a) = F\} \leq \mathbf{card} F.$$

That  $\mathbf{card} A \leq \mathbf{card} B$  now follows from the theory of cardinal arithmetic.  $\square$

**Definition.** Suppose  $V$  is a vector space. We let  $\mathbf{dim} V$  be the cardinality of a basis for  $V$ . We say  $V$  is **finite dimensional** if  $\mathbf{dim} V$  is finite.

**Remark.** If  $S$  is a finite subset of  $V$  and  $\mathbf{span} S = V$  then  $V$  is finite dimensional.

**Corollary.** Suppose  $V$  is a finite dimensional vector space and  $S$  is an independent subset of  $V$ . Then

$$\mathbf{card} S \leq \mathbf{dim} V$$

with equality only if  $S$  is a basis for  $V$ .

**Proof.** The inequality follows directly from the preceding Theorem.

Suppose  $\mathbf{card} S \leq \mathbf{dim} V$ . Were there  $v \in V \sim \mathbf{span} S$  then  $S \cup \{v\}$  would be an independent subset of  $V$  with cardinality exceeding  $\mathbf{dim} V$ .  $\square$

**Corollary.** Suppose  $V$  is finite dimensional and  $U$  is a linear subspace of  $V$ . Then  $U$  is finite dimensional.

**Proof.** Let  $S$  be a maximal independent subset of  $U$ ; such an  $S$  exists because any independent subset of  $V$  has at most  $\mathbf{dim} V$  elements. Were there  $v \in U \sim \mathbf{span} S$  then  $S \cup \{v\}$  would be an independent subset of  $U$  with cardinality exceeding that of  $S$ .  $\square$

**Corollary.** Suppose  $V$  and  $W$  are vector spaces and  $L \in \mathbf{L}(V; W)$ . Then there are  $\omega \in V^*$  and  $w \in W \sim \{0\}$  such that  $L = \omega w$  if and only if  $\mathbf{dim} \mathbf{rng} L = 1$ .

**Proof.** If there are  $\omega \in V^*$  and  $w \in W \sim \{0\}$  such that  $L = \omega w$  then  $\{w\}$  is a basis for  $\mathbf{rng} L$ .

Suppose  $\mathbf{dim} \mathbf{rng} L = 1$ . Let  $w \in W$  be such that  $\{w\}$  is a basis for  $\mathbf{rng} L$ . Then, as  $L(v) = w^*(L(v))w$  for  $v \in V$  we can take  $\omega = w^* \circ L$ .  $\square$

**Theorem.** Suppose  $V$  and  $W$  are vector spaces,  $L : V \rightarrow W$  is linear and  $V$  is finite dimensional. Then  $\mathbf{rng} L$  is finite dimensional and

$$\mathbf{dim} V = \mathbf{dim} \mathbf{ker} L + \mathbf{dim} \mathbf{rng} L.$$

**Proof.** Let  $A$  be a basis for  $\mathbf{ker} L$ . Let  $B$  be a maximal independent subset of  $V$  containing  $A$  and note that  $B$  is a basis for  $V$ . Note that  $L|(B \sim A)$  is univalent and that  $C = \{L(v) : v \in B \sim A\}$  is a basis for  $\mathbf{rng} L$ . The assertion to be proved follows from that fact that

$$\mathbf{card} B = \mathbf{card} A + \mathbf{card} C.$$

$\square$

**Definition.** Suppose  $V$  is a finite dimensional vector space and  $B$  is a basis for  $V$ . Then

$$B^* = \{b^* : b \in B\}$$

is a basis for  $V^*$  which we call the **dual basis (to  $B$ )**; the independence of this set is clear and the fact that it spans  $V^*$  follows from the fact that

$$\omega = \sum_{b \in B} \omega(b)b^*, \quad \omega \in V^*,$$

which follows immediately from (2). In particular,  $\dim V^* = \dim V$ .

**Remark.** The  $\cdot^*$  notation is quite effective but must be used with care because of the following ambiguity. Suppose  $V$  is a finite dimensional vector space and  $B_i$ ,  $i = 1, 2$  are two different bases for  $V$ . Show that if  $b \in B_1 \cap B_2$  then the  $v^*$ 's corresponding to  $B_1$  and  $B_2$  are different if  $\mathbf{span} B_1 \sim \{b\} \neq \mathbf{span} B_2 \sim \{b\}$ .

**Remark.** Suppose  $S$  is a nonempty set. We have linear maps

$$\mathbf{R}^S \ni g \mapsto ((\mathbf{R}^S)_0 \ni f \mapsto \sum_{s \in S} f(s)g(s)) \in (\mathbf{R}^S)_0^*$$

and

$$(\mathbf{R}^S)_0^* \ni \omega \mapsto (S \ni s \mapsto \omega(\delta_s)) \in \mathbf{R}^S.$$

These maps are easily seen to be linear and inverse to each other. Thus  $(\mathbf{R}^S)_0^*$  is isomorphic to  $\mathbf{R}^S$ . Now suppose  $S$  is a basis for the vector space  $V$ . Since  $\mathbf{s}$  carries  $(\mathbf{R}^S)_0$  isomorphically onto  $V$  we find that

$$V^* \equiv (\mathbf{R}^S)_0^* \equiv \mathbf{R}^S.$$

Chasing through the above isomorphisms one finds that  $\{b^* : b \in B\}$  is independent but, in case  $S$  is infinite, does not span  $V^*$ . In fact,  $V^*$  is not of much use when  $V$  is not finite dimensional.

**Definition.** Let

$$\iota : V \rightarrow V^{**}$$

be the map

$$V \ni v \mapsto (V^* \ni \omega \mapsto \omega(v)) \in V^{**}.$$

Evidently, this map is linear and univalent.

Now suppose  $V$  is finite dimensional. Let  $B$  be a basis for  $V$ . One easily verifies that

$$\iota(b) = b^{**}, \quad b \in B.$$

Thus, since  $\iota$  carries the basis  $B$  to the basis  $B^{**}$  it must be an isomorphism. It is called the **canonical isomorphism from  $V$  onto  $V^{**}$** .

**Definition.** Suppose  $U$  is a linear subspace of the vector space  $V$ . We let

$$U^\perp = \{\omega \in V^* : \omega|U = 0\}$$

and note that  $U^\perp$  is a linear subspace of  $V^*$ .

**Theorem.** Suppose  $U$  is a linear subspace of the finite dimensional vector space  $V$ . Then

$$\dim U^\perp = \dim V - \dim U.$$

**Proof.** Let  $B$  be a basis for  $U$  and let  $C$  be a basis for  $V$  such that  $B \subset C$ . Evidently,  $\{b^* : b \in C \sim B\} \subset U^\perp$ . Moreover, for any  $\omega \in U^\perp$  we have

$$\omega = \sum_{b \in C} \omega(b)b^* = \sum_{b \in C \sim B} \omega(b)b^*.$$

□

**Definition.** Suppose  $V$  and  $W$  are vector spaces and  $L : V \rightarrow W$  is linear. We define

$$L^* \in \mathbf{L}(W^*; V^*)$$

by letting

$$L^*(\omega) = \omega \circ L, \quad \omega \in W^*.$$

One easily verifies that  $\cdot^*$  carries  $\mathbf{L}(V; W)$  linearly into  $\mathbf{L}(W^*; V^*)$  and that, under appropriate hypotheses,

- (i)  $L^{**} = L$  and
- (ii)  $(L \circ M)^* = M^* \circ L^*$ .

The map  $\cdot^*$  is called the **adjoint**. Note that this term is used in the context of inner product spaces in a similar but different way and that it occurs in the theory of determinants in a totally different way.

**Theorem.** Suppose  $V$  and  $W$  are finite dimensional vector spaces and  $L : V \rightarrow W$  is linear. Then

$$(\mathbf{rng} L)^\perp = \mathbf{ker} L^* \quad \text{and} \quad (\mathbf{ker} L)^\perp = \mathbf{rng} L^*.$$

**Remark.** It is evident that the right hand side in each case is contained in the left hand side.

**Proof.** Let  $C$  be a basis for  $\mathbf{rng} L$  and let  $D$  be a basis for  $W$  containing  $C$ . Let  $A$  be a subset of  $V$  such that  $\{L(v) : v \in A\} = C$  and note that  $A$  is independent. Let  $B$  be the union of  $A$  with a basis for  $\mathbf{ker} L$  and note that  $B$  is a basis for  $V$ .

Note that  $\{d^* : d \in D \sim C\}$  is a basis for  $(\mathbf{rng} L)^\perp$  and that  $\{b^* : b \in A\}$  is a basis for  $(\mathbf{ker} L)^\perp$ .

For any  $d \in D$  we have

$$\begin{aligned} L^*(d^*) &= \sum_{b \in B} L^*(d^*)(b)b^* \\ &= \sum_{b \in B} d^*(L(b))b^* \\ &= \sum_{b \in B \sim A} d^*(L(b))b^* \\ &= \begin{cases} b^* & \text{if } d = L(b) \text{ for some } b \in A, \\ 0 & \text{else.} \end{cases} \end{aligned}$$

Thus  $\{b^* : b \in A\}$  is a basis for  $\mathbf{rng} L^*$  and  $\{d^* : d \in D \sim C\}$  is a basis for  $\mathbf{ker} L^*$ .  $\square$

**Theorem.** Suppose  $V$  is a finite dimensional vector space. There is one and only one

$$\mathbf{trace} \in \mathbf{L}(V; V)^*$$

such that

$$(5) \quad \mathbf{trace}(\omega(v)) = \omega(v), \quad \omega \in V^*, v \in V.$$

Moreover, if  $B$  is a basis for  $V$  then

$$(6) \quad \mathbf{trace} L = \sum_{b \in B} b^*(L(b)).$$

**Proof.** Let  $B$  be a basis for  $V$ . Then the formula in (6) defines a member of  $\mathbf{L}(V; V)^*$  which satisfies (5).  $\square$

**Theorem.** Suppose  $U$  and  $W$  are subspaces of  $V$ . Then

$$\mathbf{dim} U + \mathbf{dim} W + \mathbf{dim} U \cap W = \mathbf{dim} U + \mathbf{dim} W.$$

**Proof.** Let  $A$  be a basis for  $U \cap W$ . Extend  $A$  to a basis  $B$  for  $U$  and a basis  $C$  for  $W$ . Verify that  $A \cup B \cup C$  is a basis for  $U + W$ .  $\square$