

Linear Algebra

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1 Vector Spaces

1.1 Definitions

For this lecture course, \mathbb{F} will always be field.

Definition. (Vector Space) A \mathbb{F} -vector space (or a vector space over \mathbb{F}) is an abelian group $(V, +, \mathbf{0})$ equipped with a function

$$\begin{aligned}\mathbb{F} \times V &\rightarrow V \\ (\lambda, v) &\rightarrow v\end{aligned}$$

which we call scalar multiplication such that $\forall v, w \in V, \forall \lambda, \mu \in \mathbb{F}$

- (i) $(\lambda + \mu)v = \lambda v + \mu v$
- (ii) $\lambda(v + w) = \lambda v + \lambda w$
- (iii) $\lambda(\mu v) = (\lambda\mu)v$
- (iv) $1 \cdot v = v \cdot 1 = v$

Remember that $\mathbf{0}$ and 0 are not the same thing. 0 is an element in the field \mathbb{F} and $\mathbf{0}$ is the additive identity in V .

For an example consider \mathbb{F}^n n -dimensional column vectors with entries in \mathbb{F} . We also have the example of a vector space \mathbb{C}^n which is a complex vector space, but also a real vector space (taking either \mathbb{C} or \mathbb{R} as the underlying scalar field).

We also can see that $M_{m \times n}(\mathbb{F})$ form a vector space with m rows and n columns.

For any non-empty set X , we denote \mathbb{F}^X as the space of functions from X to \mathbb{F} equipped with operations such that:

$$\begin{aligned}f + g \text{ is given by } (f + g)(x) &= f(x) + g(x) \\ \lambda f \text{ is given by } (\lambda f)(x) &= \lambda f(x)\end{aligned}$$

Proposition. For all $v \in V$ we have that $0 \cdot v = \mathbf{0}$ and $(-1) \cdot v = -v$ where $-v$ denotes the additive inverse of v .

Proof. Trivial.

Definition. (Subspace) A *subspace* of a \mathbb{F} -vector space V is a subset $U \subseteq V$ which is a \mathbb{F} -vector space itself under the same operations as V . Equivalently, $(U, +)$ is a subgroup of $(V, +)$ and $\forall \lambda \in \mathbb{F}, \forall u \in U$ we have that $\lambda u \in U$.

Remark. Axioms (i)-(iv) are always automatically inherited into all subspaces.

Proposition. (Subspace test) Let V be a \mathbb{F} -vector space and $U \subseteq V$ then U is a subspace of V if and only if,

- (i) U is nonempty.
- (ii) $\forall \lambda \in \mathbb{F}$ and $\forall u, w \in U$ we have that $u + \lambda w \in U$.

Proof. If U is a subspace then U satisfies (i) and (ii) since it contains 0 and is closed. Conversely suppose that $U \subseteq V$ satisfies (i) and (ii). Taking $\lambda = -1$ so $\forall u, w \in U, u - w \in U$ hence $(U, +)$ is a subgroup of $(V, +)$ by the subgroup test. Finally taking $u = 0$ so we have that $\forall w \in U, \forall \lambda \in \mathbb{F}$ we have that $\lambda w \in U$. So U is a subspace of V . \square

We notate U by $U \leq V$.

For some examples

(i)

$$\left\{ \begin{pmatrix} x \\ y \\ z \end{pmatrix} \in \mathbb{R}^3 : x + y + z = t \right\} \subseteq \mathbb{R}^3,$$

for fixed $t \in \mathbb{R}$ is a subspace of \mathbb{R}^3 iff $t = 0$.

(ii) Take $\mathbb{R}^{\mathbb{R}}$ as all the functions from \mathbb{R} to \mathbb{R} then the set of continuous functions is a subspace.

(iii) Also we have that $C^\infty(\mathbb{R})$, the set of infinitely differentiable functions from \mathbb{R} to \mathbb{R} is a subspace of $\mathbb{R}^{\mathbb{R}}$ and the subspace of continuous functions.

(iv) A further subspace of all of those subspaces is the set of polynomial functions.

Lemma. For $U, W \leq V$ we have that $U \cap W \leq V$.

Proof. We'll use the subspace test. Both U, W are subspaces so they contain 0 hence $0 \in U \cap W$ so $U \cap W$ is nonempty. Secondly take $x, y \in U \cap W$ with $\lambda \in \mathbb{F}$. Then $U \leq V$ and $x, y \in U$ so $x + \lambda y \in U$. Similarly with W so $x + \lambda y \in W$ hence we have that $x + \lambda y \in U \cap W$ hence $U \cap W \leq V$ \square

Remark. This does not apply for subspaces, in fact from IA Groups, we know it doesn't even hold for the underlying abelian group.

Definition. (Subspace sum) For $U, W \leq V$, the *subspace sum* of U, W is

$$U + W = \{u + w : u \in U, w \in W\}.$$

Lemma. If $U, W \leq V$ then $U + W \leq V$.

Proof. Simple application of the subspace test.

Remark. $U + W$ is the smallest subgroup of U, W in terms of inclusion, i.e. if K is such that $U \subseteq K$ and $W \subseteq K$ then $U + W \subseteq K$.

1.2 Linear maps, isomorphisms, and quotients

Definition. (Linear map) For V, W \mathbb{F} -vector spaces. A *linear map* from V to W is a group homomorphism, φ , from $(V, +)$ to $(W, +)$ such that $\forall v \in V$

$$\varphi(\lambda v) = \lambda \varphi(v)$$

Equivalently to show any function $\alpha : V \rightarrow W$ is a linear map we just need to show that $\forall u, w \in V, \forall \lambda \in \mathbb{F}$ we have

$$\alpha(u + \lambda w) = \alpha(u) + \lambda \alpha(w).$$

For some examples of linear maps

- (i) $V = \mathbb{F}^n, W = \mathbb{F}^m, A \in M_{m \times n}(\mathbb{F})$. Then let $\alpha : V \rightarrow W$ be given by $\alpha(v) = Av$. Then α is linear.
- (ii) $\alpha : C^\infty(\mathbb{R}) \rightarrow C^\infty(\mathbb{R})$ defined by taking the derivative.
- (iii) $\alpha : C(\mathbb{R}) \rightarrow \mathbb{R}$ defined by taking the integral from 0 to 1.
- (iv) X any nonempty set, $x_0 \in X$,

$$\begin{aligned} \alpha : \mathbb{F}^X &\rightarrow \mathbb{F} \\ f &\rightarrow f(x_0) \end{aligned}$$

- (v) For any V, W the identity mapping from V to V is linear and so is the zero map from V to W .
- (vi) The composition of two linear maps is linear.
- (vii) For a non-example squaring in \mathbb{R} is not linear. Similarly adding constants is not linear, since linear maps preserve the zero vector.

Definition. (Isomorphism) A linear map $\alpha : V \rightarrow W$ is an *isomorphism* if it is bijective. We say that V and W are isomorphic, if there exists an isomorphism from $V \rightarrow W$ and denote this by $V \cong W$.

An example is the vector space $V = \mathbb{F}^4$ and $W = M_{2 \times 2}(\mathbb{F})$ we can define the map

$$\begin{aligned} \alpha : V &\rightarrow W \\ \begin{pmatrix} a \\ b \\ c \\ d \end{pmatrix} &\rightarrow \begin{pmatrix} a & b \\ c & d \end{pmatrix} \end{aligned}$$

Then α is an isomorphism.

Proposition. If $\alpha : V \rightarrow W$ is an isomorphism then $\alpha^{-1} : W \rightarrow V$ is also an isomorphism.

Proof. Clearly α^{-1} is a bijection. We need to prove that α^{-1} is linear. Take $w_1, w_2 \in W$ and $\lambda \in \mathbb{F}$. So we can write $w_i = \alpha(v_i)$ for $i = 1, 2$. Then

$$\alpha^{-1}(w_1 + \lambda w_2) = \alpha^{-1}(\alpha(v_1) + \lambda \alpha(v_2)) = \alpha^{-1}(\alpha(v_1 + \lambda v_2)) = v_1 + \lambda v_2 = \alpha^{-1}(w_1) + \lambda \alpha^{-1}(w_2)$$

. Hence α^{-1} is linear, so α^{-1} is an isomorphism. □

Definition. (Kernal) Let V, W be \mathbb{F} -vector spaces. Then the *kernal* of the linear map $\alpha : V \rightarrow W$ is

$$\ker(\alpha) = \{v \in V : \alpha(v) = \mathbf{0}_W\} \subseteq V$$

Definition. (Image) Let V, W be \mathbb{F} -vector spaces. Then the *image* of a linear map $\alpha : V \rightarrow W$ is

$$\text{im}(\alpha) = \{\alpha(v) : v \in V\} \subseteq W$$

Lemma. For a linear map $\alpha : V \rightarrow W$ the following hold.

- (i) $\ker \alpha \leq V$ and $\text{im } \alpha \leq W$
- (ii) α is surjective if and only if $\text{im } \alpha = W$
- (iii) α is injective if and only if $\ker \alpha = \{\mathbf{0}_V\}$

Proof. $\mathbf{0}_V + \mathbf{0}_V = \mathbf{0}_V$, so applying α to both sides any using the fact that α is linear gives that $\alpha(\mathbf{0}_V) = \mathbf{0}_W$. So $\ker \alpha$ is nonempty. The rest of the proof is a simple application of the subspace test.

The second statement is immediate from the definition.

For the final statement suppose α injective. Suppose $v \in \ker \alpha$. Then $\alpha(v) = \mathbf{0}_W = \alpha(\mathbf{0}_V)$ so $v = \mathbf{0}_V$ by injectivity. Hence $\ker \alpha$ is trivial. Conversely suppose that $\ker \alpha = \{\mathbf{0}_V\}$. Let $u, v \in V$ and suppose that $\alpha(u) = \alpha(v)$. Then $\alpha(u - v) = \mathbf{0}_W$, so $u - v \in \ker \alpha$, so $u = v$. \square

For V a \mathbb{F} -vector space, $W \leq V$ write

$$\frac{V}{W} = \{v + W : v \in V\}$$

as the left cosets of W in V . Recall that two cosets $v + W$ and $u + W$ are the same coset if and only if $v - u \in W$.

Proposition. V/W is an \mathbb{F} -vector space under operations

$$\begin{aligned} (u + W) + (v + W) &= (u + v) + W \\ \lambda(v + W) &= (\lambda v) + W \end{aligned}$$

We call V/W the quotient space of V by W .

Proof. The proof is long and requires a lot of vector space axioms so we'll just sketch out the proof.

We check that operations are well-defined, so for $u, \bar{u}, v, \bar{v} \in V$ and $\lambda \in \mathbb{F}$ if

$$u + W = \bar{u} + W, \quad v + W = \bar{v} + W$$

then

$$(u + v) + W = (\bar{u} + \bar{v}) + W$$

and

$$(\lambda u) + W = (\lambda \bar{u}) + W$$

The vector space axioms are inherited from V . \square

Proposition. (Quotient map) The function $\pi_W : V \rightarrow \frac{V}{W}$ called a *quotient map* is given by

$$\pi_W(v) = v + W$$

is a well-defined, surjective, linear map with $\ker \pi_W = W$.

Proof. Surjectivity is clear. For linearity let $u, v \in V$ and $\lambda \in \mathbb{F}$. Then

$$\begin{aligned}\pi_W(u + \lambda v) &= (u + \lambda v) + W \\ &= (u + W) + (\lambda v + W) \\ &= (u + W) + \lambda(v + W) \\ &= \pi_W(u) + \lambda\pi_W(v)\end{aligned}$$

For $v \in V$, we have that $v \in \ker \pi_W \iff \pi_W(v) = \mathbf{0}_{V/W}$. So $v + W = \mathbf{0}_V + W$ so finally $v = v - \mathbf{0}_V \in W$. \square

Theorem. (First isomorphism theorem) Let V, W be \mathbb{F} -vector spaces and $\alpha : V \rightarrow W$ linear. Then there is an isomorphism

$$\bar{\alpha} : \frac{V}{\ker \alpha} \rightarrow \text{im } \alpha$$

given by $\bar{\alpha}(v + \ker \alpha) = \alpha(v)$

Proof. For $u, v \in V$,

$$u + K = v + K \iff u - v \in K \iff \alpha(u - v) = \mathbf{0}_W \iff \alpha(u) = \alpha(v) \iff \bar{\alpha}(u + \ker \alpha) = \bar{\alpha}(v + \ker \alpha)$$

The forward direction shows that $\bar{\alpha}$ is well-defined, and the converse shows that $\bar{\alpha}$ is injective. For surjectivity given $w \in \text{im } \alpha$, there exists some $v \in V$ s.t. $w = \alpha(v)$. Then $w = \bar{\alpha}(v + \ker \alpha)$. Finally for linearity given $u, v \in V$, $\lambda \in \mathbb{F}$,

$$\begin{aligned}\bar{\alpha}((u + \ker \alpha) + \lambda(v + \ker \alpha)) &= \bar{\alpha}((u + \lambda v) + \ker \alpha) \\ &= \alpha(u + \lambda v) \\ &= \alpha(u) + \lambda\alpha(v) \\ &= \bar{\alpha}(u + \ker \alpha) + \lambda\bar{\alpha}(v + \ker \alpha)\end{aligned}$$

So $\bar{\alpha}$ is linear hence is an isomorphism \square

1.3 Basis

Definition. (Span) Let V be a \mathbb{F} -vector space. Then the *span* of some subset $S \subseteq V$ is

$$\langle S \rangle = \left\{ \sum_{s \in S} \lambda_s \cdot s : \lambda_s \in \mathbb{F} \right\}$$

where \sum denotes finite sums. An expression the form above is called a *linear combination* of S .

We say that S spans V if $\langle S \rangle = V$

Definition. (Finite-dimensional) For a vector space V we say that it is *finite-dimensional* if there exists a finite spanning set.

We'll give some simple remarks without proof.

- (i) $\langle S \rangle \leq V$ and conversely if $W \leq V$ and $S \subseteq W$ then $\langle S \rangle \leq W$.
- (ii) If $S, T \subseteq W$ and S spans V and $S \subseteq \langle V \rangle$ then T spans V .
- (iii) By convention $\langle \emptyset \rangle = \{\mathbf{0}_V\}$.
- (iv) $\langle S \cup T \rangle = \langle S \rangle + \langle T \rangle$

For an example consider $V = \mathbb{R}^3$ and consider the sets

$$S = \left\{ \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix} \right\}$$

$$T = \left\{ \begin{pmatrix} 2 \\ 1 \\ 2 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 2 \end{pmatrix}, \begin{pmatrix} -1 \\ 2 \\ 4 \end{pmatrix} \right\}$$

Then $\langle S \rangle = \langle T \rangle = \left\{ \begin{pmatrix} x \\ y \\ 2y \end{pmatrix} : x, y \in \mathbb{R} \right\} \leq \mathbb{R}^3$.

For a second example consider $V = \mathbb{R}^{\mathbb{N}}$ and set $T = \{\delta_n : n \in \mathbb{N}\}$. This is not a spanning set, since we require infinitely many elements from T to make an element in V . In fact we can write that

$$\langle T \rangle = \{f \in \mathbb{R}^{\mathbb{N}} : f(n) = 0 \text{ for all but finitely many terms}\}.$$

Definition. (Linear Independence) A subset $S \subseteq V$ is called *linearly independent* if, for all finite linear combinations

$$\sum_{s \in S} \lambda_s s \text{ of } S$$

if the sum is the zero vector in V the $\lambda_s = 0$ for all $s \in S$.

If S is not linearly independent we say that S is linearly dependent.

We'll make some more remarks

- (i) If $\mathbf{0} \in S$ then S is not linearly independent.
- (ii) If we have a finite set, then to show linearly independent, we only need to consider the linear combination of all elements, not all finite linear combinations.
- (iii) However if S is infinite, then we have to consider every possible finite subset of S and show it's linearly independent.
- (iv) Every subset of a linearly independent set is itself linearly independent.

Definition. (Basis) A subset $S \subseteq V$ is a *basis* for V if S is linearly independent and a spanning set.

For an example consider $e_i \in \mathbb{F}^n$ be given by

$$e_i = \begin{pmatrix} 0 \\ \vdots \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \quad \text{with the 1 in the } i\text{th entry}$$

then the set $\{e_i : 1 \leq i \leq n\}$ is the standard basis for \mathbb{F}^n .

For $P(\mathbb{R})$ the set of real polynomial functions and let $p_n \in P(\mathbb{R})$ be given by $p_n(x) = x^n$, then $\{p_n : n \in \mathbb{Z}_{\geq 0}\}$ is a basis for $P(\mathbb{R})$.

Proposition. If $S \subseteq V$ is a finite spanning set, then there exists a subset $S' \subseteq S$ such that S' is a basis.

Proof. If S is linearly independent then we're done. Otherwise write $S = \{v_1, \dots, v_n\}$. Then there exists $\lambda_1, \dots, \lambda_n$ such that $\lambda_1 v_1 + \dots + \lambda_n v_n = \mathbf{0}$ wlog suppose that λ_n is nonzero. Then

$$v_n = -\frac{1}{\lambda_n} \sum_{i=1}^{n-1} \lambda_i v_i$$

so v_n is in the span of the other vectors. Hence $S \setminus \{v_n\}$ is still a spanning set. Repeat which the set is linearly independent, must terminate since the set is finite and the empty set is not a spanning set. \square

Corollary. Every finite-dimensional vector space has a finite basis.

Proof. Trivial application of the proposition \square

Theorem. (Steinitz Exchange Lemma) Let $S, T \subseteq V$ finite with S linearly independent and T a spanning set of V . Then

- (i) $|S| \leq |T|$,
- (ii) and there exists $T' \subseteq T$ which has size $|T'| = |T| - |S|$ and $S \cup T'$ spans V .

Proof. To come later...

Let's look at some consequences of the lemma first.

Corollary. For a finite-dimensional vector space V ,

- (i) Every basis for V is finite.
- (ii) All finite basis have the same size.

Proof. V has a finite basis B , suppose we have some other basis B' infinite. Let $B'' \subseteq B'$ with $|B''| = |B| + 1$ then $|B''|$ is linearly independent, so applying (i) of the Steinitz exchange lemma with $S = B''$ and $T = B$ we get a contradiction.

For the second part, let B_1, B_2 be finite basis for V then apply Steinitz symmetrically since both are spanning set and linearly independent, so we get that $|B_1| \geq |B_2|$ and $|B_1| \leq |B_2|$ so $|B_1| = |B_2|$. \square

Definition. (Dimension) For a vector space V the *dimension* of V is the size of any basis. We write this as $\dim V$.

This definition is well-defined by the previous corollary.

For an example $\dim \mathbb{F}^n = n$ since we've shown the standard basis has size n . As a complex vector space \mathbb{C} is one-dimensional as a complex vector space and two-dimension as a real vector space, with basis $\{1\}$ and $\{1, i\}$ respectively.

Corollary. For a vector space V let $S, T \subseteq V$ finite, with S linearly independent and T a spanning set, then

$$|S| \leq \dim V \leq |T|$$

with equality if and only if S spans or V is linearly independent respectively.

Proof. The inequalities are immediate from Steinitz. If S is a basis then $|S| = \dim V$ from the previous corollary. Conversely if $|S| = \dim V$ and let B be a basis for V so we have that $|B| = |S|$ so B is a spanning set. So we can apply Steinitz (ii) to B so there exists $B' \subseteq B$ with $|B'| = |B| - |S| = 0$ and $S \cup B' = S \cup \emptyset$ spans V . So S is a basis. Similiar we have a very similiar proof for equality in V . \square

We will not prove that every vector space has a basis, however some non-finitely dimensional vector spaces have an infinite basis, for example $P(\mathbb{R})$.

Proposition. If V is a finite-dimensional vector space, then if $U \leq V$ then U is finite-dimensional, namely, $\dim U \leq \dim V$ with equality if and only if $U = V$.

Proof. If $U = \{\mathbf{0}\}$, we're done. Otherwise let $\mathbf{0} \neq u_1 \in U$. Then $\{u_1\} \subseteq U$ is linearly indepedent. Repeating, after repeating k times suppose we have $\{u_1, \dots, u_k\}$ linearly indepedent with $k \leq \dim(V)$ by the previously corollary. If the set spans U we're done, if not we'll add another vector, u_{k+1} outside of the span of our space. If $\{u_1, \dots, u_{k+1}\}$ is not linearly indepedent, we can write $\mathbf{0}$ non-trivially, so

$$\sum_{i=1}^{k+1} \lambda_i u_i = \mathbf{0}$$

with $\lambda_{k+1} \neq 0$ since $\{u_1, \dots, u_k\}$ linearly indepedent. Thus we have that

$$u_{k+1} = -\frac{1}{\lambda_{k+1}} \left(\sum_{i=1}^k \lambda_i u_i \right)$$

this process must terminate after at most $\dim V$ many steps, by the previous corollary. If $\dim U = \dim V$ apply the previous corollary with S being any basis for U . \square

Proposition. (Extending a basis) Let $U \leq V$. For any basis B_U of U there exists a basis B_V of V such that $B_U \subseteq B_V$.

Proof. Apply the second result from Steinitz with $S = B_U$ and T is any basis for V . We obtain that $T' \subseteq T$ s.t.

$$|T'| = |T| - |S| = \dim V - \dim U$$

and $B_V = B_U \cup T'$ spans V . But we have that

$$|B_V| \leq |B_U| + |T'| = \dim V$$

so by the previous corollary, B_V is a basis for V . \square

Now we'll finally prove the Steinitz exchange lemma.

Proof. Let $S = \{u_1, \dots, u_m\}$, $T = \{v_1, \dots, v_n\}$ with $|T| = m$ and $|T| = n$. If S is empty then we're done. Otherwise there exists $\lambda_i \in \mathbb{F}$ such that

$$u_1 = \sum_{i=1}^n \lambda_i v_i$$

so by renumbering we can say that $\lambda_1 \neq 0$. Then

$$v_1 = \frac{1}{\lambda_1} \left(u_1 - \sum_{i=2}^n \lambda_i v_i \right)$$

So $\{u_1, v_2, \dots, v_n\}$ spans V . After repeating k times with $k < m$ suppose $\{u_1, \dots, u_k, v_{k+1}, \dots, v_n\}$ spans V , then there exists $\lambda_i, \mu_j \in \mathbb{F}$ such that

$$u_{k+1} = \sum_{j=1}^k \mu_j u_j + \sum_{i=k+1}^n \lambda_i v_i$$

If for all $\lambda_i = 0$ then

$$\left(\sum_{j=1}^k \mu_j u_j \right) - u_{k+1} = \mathbf{0}$$

which is a contradiction since S is linearly independent. So by relabeling we have that $\lambda_{k+1} \neq 0$ such that

$$v_{k+1} = \frac{1}{\lambda_{k+1}} \left(u_{k+1} - \sum_{j=1}^k \mu_j u_j - \sum_{i=k+1}^n \lambda_i v_i \right)$$

so $\{u_1, \dots, u_{k+1}, v_{k+2}, \dots, v_n\}$ spans V . So we can conclude that $m \neq n$ and $\{u_1, \dots, u_m, v_{m+1}, \dots, v_n\}$ spans V hence the set $T' = \{v_{m+1}, \dots, v_n\}$ exists as claimed. \square

Definition. (Nullity) For a linear map $\alpha : V \rightarrow W$ we define the *nullity* of α as

$$n(\alpha) = \dim \ker \alpha.$$

Definition. (Rank) For a linear map $\alpha : V \rightarrow W$ we define the *rank* of α as

$$\text{rk}(\alpha) = \dim \text{im } \alpha.$$

Theorem. (Rank-nullity theorem) If V is a finite dimensional \mathbb{F} -vector space and W is a \mathbb{F} -vector space. Then if $\alpha : V \rightarrow W$ is linear then $\text{im } \alpha$ is finite dimensional and

$$\dim V = \text{n}(\alpha) + \text{rk}(\alpha).$$

Proof. Recall the first isomorphism theorem so

$$\frac{V}{\ker \alpha} \cong \text{im } \alpha$$

It is sufficient to prove the lemma

Lemma. For $U \leq V$,

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

Proof. Let $B_U = \{u_1, \dots, u_m\}$ be a basis of U . Extend to a basis $B_V = \{u_1, \dots, u_m, v_{m+1}, \dots, v_n\}$ of V where $m = \dim U$ and $n = \dim V$.

Set $B_{V/U} = \{v_i + U : m+1 \leq i \leq n\}$. Then we claim that $B_{V/U}$ is a basis for V/U of size $n - m$. To show spanning, for $v \in V$ write

$$v = \sum_i \lambda_i v_i + \sum_j \mu_j v_j$$

Then $v + U = \sum_i \lambda_i (v_i + U) \in \langle B_{V/U} \rangle$. For linear independence, suppose

$$\sum_i \lambda_i (v_i + U) = \mathbf{0} + U$$

hence

$$\begin{aligned} &= \left(\sum_i \lambda_i v_i \right) + U \\ &\quad \sum_i \lambda_i v_i \in U \\ &\quad \sum_i \lambda_i v_i = \sum_j \mu_j v_j \end{aligned}$$

since B_V is linearly independent, we have that all λ_i and μ_j are zero. Similarly if $v_i + U = v_j + U$ with $i \neq j$ then we can write $v_i - v_j = \sum_j \mu_j v_j$ which is a contradiction. \square

Remark. We can make a direct proof without quotient spaces by rearranging some of the arguments of the proof.

Corollary. (Linear Pigeonhole principle) If $\dim V = \dim W = n$ and $\alpha : V \rightarrow W$ then the following conditions are equivalent.

- (i) α is injective,
- (ii) α is surjective,
- (iii) α is an isomorphism.

Proof. If α injective then $n(\alpha) = 0$ so by rank nullity we have that $\text{rk}(\alpha) = n$ so α is surjective. If α is surjective then $\text{rk}(\alpha) = n$ so by rank nullity, the dimension of the kernel is 0 hence the kernel is trivial, so α injective, hence α is an isomorphism. If α is an isomorphism, clearly it's injective, so all equivalent. \square

Proposition. Suppose V is a vector space with a basis B . For any vector space W and any function $f : B \rightarrow W$ there is a unique linear map $F : V \rightarrow W$ such that $F(B) = W$.

Proof. First we'll show existence. For $v \in V$ write $v = \sum_b \lambda_b b$ for a finite sum. Then define

$$F(v) = \sum_b \lambda_b f(b).$$

This is well-defined, since B is a basis the λ_b are uniquely determined by v . For $u, v \in V$ and $\lambda \in \mathbb{F}$ we write

$$u = \sum_b \mu_b b, \quad \sum_b \lambda_b b.$$

Then

$$\begin{aligned} F(u + \lambda v) &= F\left(\sum_b (\mu_b + \lambda \lambda_b) f(b)\right) \\ &= \sum_b \mu_b f(b) + \lambda \sum_b \lambda_b f(b) \\ &= F(u) + \lambda F(v). \end{aligned}$$

So F is linear. To show uniqueness $\bar{F} : V \rightarrow W$ is another linear map extending f then,

$$\bar{F}\left(\sum_b \lambda_b b\right) = \sum_b \lambda_b \bar{F}(b)$$

which is the same as our definition for F hence they are the same function.

Corollary. For a vector space, V , with $\dim V = n$ with a basis $B = \{v_1, \dots, v_n\}$ for V then there is a isomorphism

$$F_B : V \rightarrow \mathbb{F}^n$$

$$\sum_{i=1}^n \lambda_i v_i \rightarrow \begin{pmatrix} \lambda_1 \\ \vdots \\ \lambda_n \end{pmatrix}$$