Some Notes on Bases and Dimension

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These notes, originally written for my Fall 2023 class, cover most of the content Section 1.6 of Friedberg, Insel, and Spence, *Linear Algebra*, 5th ed. (FIS), minus the examples. However, my order of presentation differs from the book's, and I include some material that's not in Section 1.6.

Recap of some earlier material that will be used in these notes

Proposition 0 below assembles some previously proven facts that will used in these notes. ("Previously proven" was correct for my Fall 2023 class. However, in Fall 2024, I don't recall mentioning in class the facts in part (e).)

Proposition 0 Let V be a vector space and let $S \subseteq V$.

- (a) Let $v \in \operatorname{span}(S)$. Then $\operatorname{span}(S \cup \{v\}) = \operatorname{span}(S)$.
- (b) If S is linearly independent, then so is every subset of S.
- (c) Let $S \subseteq V$. Then S is linearly dependent if and ony if there exists $v \in S$ such that $v \in \text{span}(S \setminus \{v\})$ (i.e., iff some element of S is in the span of the other elements). (Recall that " $S \setminus \{v\}$ " means "S with v removed.")
- (d) Let $S \subseteq V$ be a linearly independent set, and let v be an element of V that is not in S. Then $S \cup \{v\}$ is linearly independent if and only if $v \notin \operatorname{span}(S)$.
- (e) Let A and B be subsets of V.
 - (i) Let $v \in \text{span}(A \bigcup B)$. Then v = x + y for some $x \in \text{span}(A)$ and $y \in \text{span}(B)$.
 - (ii) Conversely, any $v \in V$ of the form (element of span(A)) + (element of span(B)) lies in span(A \cup B).

[Fall 2024: prove these facts as an exercise.]

(Equivalent to this bidirectional implication: $\operatorname{span}(A \cup B) = \operatorname{span}(A) + \operatorname{span}(B)$.

[End of recap.]

Basis: definition, existence, and equivalent characterizations

Definition 1 Let V be a vector space. A *basis* of V is a subset $\beta \subseteq V$ that spans V and is linearly independent.

A basis is *finite* if it is a finite set (possibly empty).

Note that using the letter " β " for some subset of a vector space V does not constitute an assumption that β is a basis of V!! In these notes, I often use " β " for sets that are going to **turn out to be** bases when I'm done proving something, but a hypothesis like "Let $\beta \subseteq V$ " tells you *nothing* about β other than that it is a subset of V.

Observe that the empty set \emptyset spans the trivial vector space $\{0\}$ and is linearly independent, hence is a basis of this vector space. Conversely, if V is a vector space for which \emptyset spans V, then $V = \{0\}$, and since \emptyset is linearly independent, \emptyset is a basis of V.

Observe also that if S is a linearly independent subset of V, then S is a basis of the subspace span(S), since S is linearly independent and is a spanning set for span(S).

Bases (the plural of "basis") can also be characterized as *minimal spanning sets* or as *maximal linearly independent sets*. We define both of these concepts next, and show that each is equivalent to the concept of *basis*.

Definition 2 Let V be a vector space and let $\beta \subseteq V$.

(a) We call β a minimal spanning set (for V) if β spans V, but for each $v \in \beta$, the set $\beta \setminus \{v\}$ (β with v removed) does not span V.

(b) We call β a maximal linearly independent set (in V) if β is linearly independent, but for each $v \in V$ that is not in β , the set $\beta \cup \{v\}$ is linearly <u>dependent</u>.

Note that Definition 2 makes no reference to *cardinality* ("size") of the sets involved. In the usage above, "minimality" (respectively, "maximality") simply means that we lose an indicated property if we remove any element of β (respectively, add any new element to β). Later, we will see how cardinality enters the picture.

Proposition 3 Let V be a vector space and let $\beta \subseteq V$.

(a) If β is a minimal spanning set for V, then β is linearly independent.

(b) If β is a maximal linearly independent set, then β spans V.

Proof: (a) Assume that β is a minimal spanning set for V.

Suppose that β is linearly dependent. Then by Proposition 0(c) there exists $v \in \beta$ such that $v \in \operatorname{span}(\beta \setminus \{v\})$. Select such a v. Then from Proposition 0(a) (applied with $S = \beta \setminus \{v\}$), it follows that $\operatorname{span}(\beta \setminus \{v\}) = \operatorname{span}(\beta) = V$, contradicting the assumption that β is a *minimal* spanning set for V. Hence β is linearly independent.

(b) Assume that β is a maximal linearly independent set.

Suppose that β does not span V, and let v be an element of V that is not in span (β) . Then by Proposition 0(d), $\beta \cup \{v\}$ is linearly independent, contradicting the assumption that β is a *maximal* linearly independent set. Hence β spans V.

Corollary 4 Let V be a vector space and let $\beta \subseteq V$. Then the following are equivalent:

- (i) β is a minimal spanning set for V.
- (ii) β is maximal linearly independent set.
- (iii) β is a basis of V.

Proof: $((i) \implies (iii) \text{ and } (ii) \implies (iii))$ By Proposition 3, if either (i) or (ii) is satisfied, then β is both linearly independent and a spanning set for V, hence is a basis of V.

((iii) \implies (i) and (ii)). Assume that β is a basis of V. Thus, span(β) = V and β is linearly independent.

First suppose that β is not minimal (as a spanning set for V). Let $v \in \beta$ be such that $\operatorname{span}(\beta \setminus \{v\}) = V$. Then, in particular, $v \in \operatorname{span}(\beta \setminus \{v\})$, so by Proposition 0(d), β is linearly dependent, a contradiction. Hence β is minimal (as a spanning set for V).

Next suppose that β is not maximal (as a linearly independent set). Let $v \in V$ be such that $v \notin \beta$ but $\beta \cup \{v\}$ is linearly independent. Then by Proposition 0(c), $v \notin \text{span}(\beta)$, contradicting the assumption that β spans V. Hence β is maximal (as a linearly independent set).

We have now shown that each of (i) and (iii) implies the other, and that each of (ii) and (iii) implies the other. Hence each of the statements (i), (ii), and (iii) implies the other two.

Remark 5 In most proofs that three assertions (i), (ii), and (iii) are equivalent, the most efficient way to prove that each assertion is equivalent to every other is to establish a "cycle of implications" such as "(i) \implies (ii) \implies (iii) \implies (iii) \implies (i)," (conventional shorthand for "(i) implies (ii), (ii) implies (iii), and (iii) implies (i)"). The short-hand for the logic of the proof we gave for Corollary 4 is "(i) \iff (ii) \iff (iii)". Although this logical

structure requires that four separate implications be shown, rather than just the three needed for a "cycle of implications," for proving Corollary 4 the "(i) \iff (ii) \iff (iii)" structure seemed more natural to me. The general efficiency-advantage of the "cycle of implications" logical structure becomes more significant when you're showing that four or more statements are equivalent.

Question: Does every vector space *have* a basis?

The answer is yes, but this is not easy to show (and we will not show it). For example, the vector space $\mathcal{F}(\mathbf{R}, \mathbf{R})$ has a basis, but you will never succeed in writing one down! There is one *very* important class of vector spaces for which we can (and will) establish existence of a basis—in fact, a finite basis—fairly easily: the *finitely generated* vector spaces, defined below.

Definition 6 A vector space V is *finitely generated* if V has a finite spanning set.

Notation 7 (for these notes) For any finite set A, the notation |A| will denote the cardinality of A (a non-negative integer).

Proposition 8 Every finitely generated vector space V has a finite basis. In fact, any finite spanning set $S \subseteq V$ has a subset S that is a (finite) basis of V.

Proof: Let V be a finitely generated vector space, let $S \subseteq V$ be a finite spanning set, and let \mathcal{G}_S be the collection of all subsets of S that span V, a finite collection since any finite set (in this case S) has only finitely many subsets. Let $\mathcal{N}_S = \{|A| : A \in \mathcal{G}_S\}$, a finite set since \mathcal{G}_S is finite. Since $S \in \mathcal{G}_S$, the set \mathcal{G}_S is nonempty, and therefore so is so is \mathcal{N} . Thus, \mathcal{N} is a nonempty, finite set of integers, and therefore has a smallest element, say n_S , which we will simply call n. (I bothered writing the notation " n_S " just for the sake of a comment later in these notes). Let $\beta \in \mathcal{G}_S$ be a spanning set for V with (exactly) n elements.

If n = 0 then the spanning set β is the empty set, which is linearly independent, hence is also a basis for V (and V must be the trivial vector space $\{0_V\}$). Suppose now that $n \ge 1$, that $v \in \beta$, and that $\beta' := \beta \setminus \{v\}$ spans V. Then $\beta' \in \mathcal{G}_S$, but β' has only n-1 elements, contradicting the definition of n. Hence the finite set β is a minimal spanning set for V, so by Corollary 4, β is a basis of V.

Definition 9 A vector space V is *finite-dimensional* if V has a finite basis. Otherwise we say that V is *infinite-dimensional*.

(Note that we have not yet defined *dimension*, so we have not yet seen the reason for choosing the terminology "finite/infinite-dimensional".)

Remark 10 We proved the first sentence of Proposition 8 as a consequence of the second. An argument that proves just the first sentence without reyling on the second is the following:

Let V be a finitely generated vector space, and let \mathcal{G} be the collection of all finite spanning-sets for V. By definition of "finitely generated" the set \mathcal{G} is nonempty. Let $\mathcal{N} = \{|A| : A \in \mathcal{G}\}$. Since \mathcal{G} is nonempty, so is \mathcal{N} . Thus, \mathcal{N} is a nonempty set of non-negative integers. But every non-empty set of nonnegative integers contains a smallest element.¹ Let n be the smallest element of \mathcal{N} , and let $\beta \in \mathcal{G}$ be a spanning set for V with n elements.

If n = 0 then the spanning set β is the empty set, which is linearly independent, hence is also a basis for V (and V must be the trivial vector space $\{0_V\}$). Suppose now that $n \ge 1$, that $v \in \beta$, and that $\beta' := \beta \setminus \{v\}$ spans V. Then $\beta' \in \mathcal{G}$, but β' has only n - 1 elements, contradicting the definition of n. Hence the finite set β is a minimal spanning set for V, so by Corollary 4, β is a basis of V.

In the proof directly above, the basis β we found was not only a minimal spanning set in the sense of Definition 2, it also was minimal in the sense of having smallest *cardinality* among all spanning sets for V. This proof establishes that *at least one* such set β exists (for a finitely generated vector space), but does not address the question of whether *every* basis of a finitely generated vector space is minimal in this other, cardinality-based sense. Neither this proof nor the one we gave for Proposition 8 addresses the question of whether all bases of a given, finitely generated vector space have the same number of elements (or even whether the cardinality n_S in the proof of Proposition 8 can depend on the spanning set S). We will address that question later in these notes.

Corollary 11 A vector space is finite-dimensional if and only if it is finitely generated.

Proof: Let V be a vector space.

 (\Longrightarrow) : Assume that V is finite-dimensional and let β be a finite basis. Then β is a finite spanning-set for V, so V is finitely generated.

 (\Leftarrow) : Assume that V is finitely generated. Then by Proposition 8, V has a finite basis.

In the above proof, because terminology in Corollary 11 had just been defined, and possibly not absorbed yet by every student, I included extra details that amounted to reminders of the

¹This fact is called the "Well-Ordering Principle", a fundamental property of the integers that is equivalent to the Axiom of Induction and that you may have learned in Sets and Logic. In its usual statement, the Well-Ordering Principle asserts that any non-empty set of *positive* integers has a smallest element. However, an easy corollary is that the same holds for any non-empty set of *non-negative integers*.

relevant definitions. That was just a pedagogical choice on my part. The following would have been a perfectly fine proof:

Let V be a vector space. If V is finite-dimensional then V has a finite basis, which, by definition, spans V. Conversely, if V is finitely generated, then by Proposition 8, V has a finite basis.

Finding a basis (given a finite spanning set)

Proposition 8 assures that if S is a finite spanning set for V, then there *exists* a subset of S that is a basis of V, but the proposition does not tell us how to go about *finding* such a subset. The algorithm given below accomplishes this.

When V is a subspace of \mathbb{R}^n there is a more efficient algorithm that we'll learn later in the semester; we haven't developed the necessary tools yet.

Extracting a basis from a finite spanning set. Suppose that we are given a finite spanning set S for a vector space V. (Possible only if V is finite-dimensional, by Definition 9!) Below is an algorithm for finding a subset of S that is a basis of V.

(*Note*: This is the same algorithm used in the proof of Theorem 1.9 in FIS Section 1.6. I've just worded it differently, expanding on the "Continue, if possible ..." sentence in the first paragraph of the book's proof.)

Informal description of process:

If $S = \emptyset$ or $S = \{0_V\}$, then $V = \{0_V\}$ and \emptyset is a subset of S that is a basis of V, and we're done.

Now assume that S is nonempty and is not $\{0_V\}$. Then we can extract a subset of S that is a basis for V as follows.

Mentally create "keep", "discard", and "candidate" bins to hold certain elements of S. Initially, the "keep" and "discard" bins are empty, and the "candidate" bin holds all the nonzero elements of S, listed in some order. (Remember, we assumed that S is finite and has at least one element.) As we proceed, the sets Cand, Keep, and Disc, consisting of the elements of the "candidate", "keep", and "discard" bins respectively, will change.

Take the first nonzero vector out of the "candidate" bin and put it into the "keep" bin, creating a linearly independent set Keep (initially with just one element). If anything remains in Cand, determine whether the first remaining vector is in span(Keep). If yes, throw that vector into the discard bin. If no, move that vector to the "keep" bin, creating a new, linearly independent Keep set with one more element than before.

Repeat this process until the candidate bin is empty, which must eventually happen since the initial "Cand" set had only finitely many elements. At each iteration, the new vector added to the linearly independent then-current set Keep is not in the span of what's already there, so the updated "Keep" set is again linearly independent (by Proposition 0(d)). When we're done, every vector we discarded was in the span of previously-kept vectors, so would not have contributed to span(Keep) had we kept it (by Proposition 0(a), applied successively to each discarded vector). Hence, once the "candidate" bin is empty, the final Keep set (still linearly independent) has the same span as the original set *S*—namely, all of *V*—and is therefore a basis of *V*.

In case there's anything in the above description that's not clear, here is a formal

description of the process:

"Reduction Algorithm" (name just for these notes!)

1. Let $S_1 = \{v \in S : v \neq 0_V\} \subseteq S$. (In other words, if S contains the zero vector, remove 0_V from S to get S_1 . If S does not contain the zero vector, let $S_1 = S$.) Note that if $0_V \in S$ then for any $v \in \text{span}(S)$ we have

$$v = v' + c0_V = v$$

for some $v' \in \operatorname{span}(S_1)$. It follows that $\operatorname{span}(S_1) = \operatorname{span}(S)$.

- 2. If $S_1 = \emptyset$, then $V = \text{span}(S_1) = \{0_V\}$, and $\emptyset \subseteq S$ is a basis of V. In this case, stop; we are done.
- 3. In this and the remaining steps, we assume $S_1 \neq \emptyset$. Note that, by definition of S_1 , all elements of S_1 are nonzero.

Let $n = |S_1|$ and enumerate the elements of S_1 as v_1, \ldots, v_n . Let $T_1 = \{v_1\}$. Since $v_1 \neq 0_V$, the set T_1 is linearly independent. If n = 1 then T_1 is a basis of V, and we are done.

Assume now that n > 1. Next, successively check v_2, \ldots, v_n until we either (i) find a j such that v_j is not a scalar multiple of v_1 (equivalently, $v_j \notin \text{span}\{v_1\}$), or (ii) run out of vectors (i.e, we find that v_j is a multiple of v_1 for every $j \in \{2, \ldots, n\}$). In case (ii), it is easily seen that $\text{span}(S_1) = \text{span}(T_1)$ (using Proposition 0(a) repeatedly), in which case $T_1 \subseteq S$ is a linearly independent spanning set for V—i.e. a basis of V—and we are done.

Suppose now that we are in case (i). Let $j_2 \in \{2, \ldots, v_n\}$ be the smallest j for which $v_j \notin \operatorname{span}\{v_1\} = \operatorname{span}(T_1)$ (mentally throwing into the "discard" bin each vector v_k with $j_1 < k < j_2$). Let $T_2 = \{v_1, v_{j_2}\}$, a linearly independent set (by Proposition 0(d)) and let $S_2 = T_2 \bigcup \{v_j : j_2 < j \leq n\}$ (i.e. S_2 is S_1 with the "discarded" vectors v_j , $1 < j < j_2$, removed). Since each of the discarded vectors lies in $\operatorname{span}\{v_1\} \subseteq \operatorname{span}(S_2)$, it follows from Proposition 0(a) that $\operatorname{span}(S_1) = \operatorname{span}(S_2)$, hence that $\operatorname{span}(S_2) = V$.

4. Now continue recursively. To have a consistent notational pattern, define $j_1 = 1$. Suppose that $k \geq 2$ and that we have found j_1, \ldots, j_k , satisfying $1 = j_1 < j_2 < j_3 < \cdots < j_k \leq n$, (with j_3 present only if $k \geq 3$) such that $T_k := \{v_{j_1}, v_{j_2}, \ldots, v_{j_k}\}$ is linearly independent and $V = \operatorname{span}(S_k)$, where $S_k = T_k \bigcup \{v_j : j_k < j \leq n\}$.

If $j_k = n$ then $S_k = T_k$, and $T_k \subseteq S$ is both linearly independent and a spanning set for V— i.e. a basis of V—and we are done.

Assume now that $j_k < n$, and successively check v_{j_k+1}, \ldots, v_n until we either (i) find a $j > j_k$ such that $v_j \notin \operatorname{span}(T_k)$, or (ii) run out of vectors (i.e, we find that

 $v_j \in \operatorname{span}(T_k)$ for every $j \in \{j_k + 1, \ldots, n\}$). (Note: as a practical matter, for each j this check may require its own algorithm², not discussed in this set of notes! However, it is *logically* indisputable that, for each j, either $v_j \in \operatorname{span}(T_k)$ or $v_j \notin \operatorname{span}(T_k)$.) In case (ii), $V = \operatorname{span}(S_k) = \operatorname{span}(T_k)$ (again using Proposition 0(a) repeatedly). Thus, in this case, $T_k \subseteq S$ is a linearly independent spanning set for V, i.e. a basis of V, and we are done.

Suppose now that we are in case (i), and let $j_{k+1} \in \{j_k+1, \ldots, v_n\}$ be the smallest j for which $v_j \notin \operatorname{span}(T_k)$. Let $T_{k+1} = T_k \cup \{v_{j_{k+1}}\} = \{v_{j_1}, v_{j_2}, \ldots, v_{j_{k+1}}\}$, a linearly independent set (by Proposition 0(d)) and let $S_{k+1} = T_{k+1} \cup \{v_j : j_{k+1} < j \leq n\}$ (i.e. S_{k+1} is S_k with the newly "discarded" vectors—those v_j for which $j_k < j < j_{k+1}$ —removed). Since each of the discarded vectors lies in $\operatorname{span}(T_k) \subseteq \operatorname{span}(S_k)$, it follows from Proposition 0(a) that $\operatorname{span}(S_k) = \operatorname{span}(S_{k+1})$, hence that $\operatorname{span}(S_{k+1}) = V$.

Since $j_{k+1} \ge j_k + 1$ and *n* is finite, eventually this process must terminate. I.e. we will eventually find *k* for which either $j_k = n$, or for which $j_k < n$ but $v_j \in \text{span}(T_k)$ for every $j \in \{j_k + 1, \ldots, n\}$ ("case (ii)" above). In either of these terminating situations, the set $T_k \subseteq S$ is a basis of *V*.

Remark 12 Since the span of every subset of a vector space V is itself a vector space, the same algorithm can be used to find a basis of span(S) when we are given a finite subset $S \subseteq V$.

The description of our algorithm above can easily be rewritten as another proof of Proposition 8 (though much a much longer proof than our original one).

²You've already done this type of check in homework exercises. It often involves solving systems of simultaneous linear equations. Eventually, we will have more efficient ways of doing this than we've seen so far.

More relations among linearly independent sets, spanning sets, and bases

So far, we still have not addressed the question of whether two bases of a finitedimensional vector space can have different cardinalities. Nothing we've done (so far) rules out the possibility that, starting with two different finite spanning sets, or with the same finite spanning set ordered in two different ways, our "reduction algorithm" might produce two different bases with different numbers of elements. After proving the next theorem below—the "Replacement Theorem"—we'll be able to show that this can't happen.

(Note: I am *deliberately* wording this theorem differently from the corresponding theorem in FIS [Theorem 1.10 in Section 1.6], to help you understand the *concepts* by seeing them expressed in more than one way. In accordance with the name chosen for this theorem, my wording introduces notation for the subset of G that's actually being *replaced*, my set R. The set $G \setminus R$ in the wording below is the set H in the book's wording, the subset of G that's *left over* when R is removed from G.)

(Note: The proof of this theorem will be inductive. If you are used to thinking that the terms "base case" and "inductive step" need to appear in an inductive proof, then before reading the proof of the theorem below, see the handout "Inductive proofs: some common mistakes and misconceptions" [particularly, the first full paragraph on p. 2] on the Miscellaneous Handouts page.)

Theorem 13 ("Replacement Theorem") Let V be a vector space with a finite spanning set G. If $L \subseteq V$ is a finite linearly independent subset of V, then

- (a) $|L| \leq |G|$ (recall Notation 7), and
- (b) there exists a subset $R \subseteq G$, having exactly as many elements as L, such that the set $L \cup (G \setminus R)$ spans V. Equivalently, there exists a subset $H \subseteq G$, having exactly |G| |L| elements, such that $L \cup H$ spans V.

Said yet another way: there is some subset R of G, having exactly as many elements as L, such that if we remove the elements of R from G, and replace them with the elements of L, then the new set $G' := L \cup \underbrace{(G \setminus R)}_{V}$ we obtain still spans V:

$$V = \operatorname{span}(G) = \operatorname{span}(G').$$

Below, we will give the same proof of this theorem twice: first, a heavily commented version with a diagram that may be helpful (the comments and diagram are not actually part of the proof); second, the same proof with all the comments and diagram removed.

Proof: [The idea behind this proof is essentially to find elements of G, one at a time, that we can replace by elements of L without changing the span; i.e. with the modified version of G still

spanning V. The cleanest way to write the argument, however, is to structure it as an inductive proof, proceeding by induction on the cardinality of L, rather than by running one at a time through the elements of a single, fixed (but arbitrary) linearly independent set L. The set R in part (b) is a "discard" set (whose elements we're replacing by elements of L) that's not really important to keep track of; what matters more at the end is the set $H = G \setminus R$ of un-replaced elements.]

We will proceed by induction on the cardinality of L. First suppose that |L| = 0. Then $L = \emptyset$, and H = G itself is a subset of G of cardinality |G| - |L| = |G|. Trivially, $L \cup H = \emptyset \cup G = G$, so $\operatorname{span}(L \cup H) = \operatorname{span}(G) = V$. Just as trivially, $0 = |L| \le |G|$ (even if G is empty!). This establishes (a) and (b) in the case |L| = 0.

Suppose now that m is a non-negative integer such that statements (a) and (b) are true whenever $L \subseteq V$ is linearly independent set of cardinality $|L| \leq m$. (Above, we showed that 0 is such an m.) Assume now that |L| = m + 1 and let v_1, \ldots, v_{m+1} be the elements of L. Let n = |G|. Then $L_1 := L \setminus \{v_{m+1}\}$ is an m-element linearly independent subset of V (by Proposition 0(b)), so by the inductive hypothesis, $m \leq n$ ($|L_1| \leq |G|$) and we may select a subset $H_1 \subseteq G$ of cardinality n - m such that $G_1 := L_1 \cup H_1$ spans V. Let u_1, \ldots, u_{n-m} be the elements of H_1 .

(This gives us line (1) in the diagram below; the other lines will be explained later. The vertical arrows in the diagram are informal notation, not mathematical arrows; they just indicate certain steps we'll be taking. In line (2), we start examining the (potentially) larger set $G \cup \{v_{m+1}\}$ for a reason that becomes clear only near the end of the proof.

The diagram, like these blue comments, is not part of the proof; I've included it as a guide to make the proof easier to follow.)

$$G_1 = \underbrace{\{v_1, \dots, v_m\}}_{L_1} \cup \underbrace{\{u_1, u_2, \dots, u_{n-m}\}}_{H_1}$$
(1)
$$\downarrow$$

$$G_1 \cup \{v_{m+1}\} = \underbrace{\{v_1, \dots, v_m, v_{m+1}\}}_{L} \cup \underbrace{\{u_1, u_2, \dots, u_{n-m}\}}_{H_1 \text{ (still)}}$$
(2)

$$= \{v_1, \dots, v_m, v_{m+1}\} \cup \underbrace{\{u_1, u_2, \dots, u_{\text{whatever}}, \dots, u_{n-m}\}}_{H_1 \text{ (still)}} (3)$$

$$= \{v_1, \dots, v_m, v_{m+1}\} \cup \{\underbrace{u_1, \dots, u_{n-m}}_{\text{previous renumbered from}}\}$$
(4)
$$\downarrow$$

$$(G_1 \cup \{v_{m+1}\}) \setminus \{u_1\} = \underbrace{\{v_1, \dots, v_m, v_{m+1}\}}_{L \text{ (still)}} \qquad \cup \underbrace{\{u_2, \dots, u_{n-m}\}}_{H}$$
 (5)

Since $L_1 \cup H_1$ spans V (part of our inductive hypothesis), in particular $v_{m+1} \in \operatorname{span}(G_1)$, so there exist $x \in \operatorname{span}(L_1)$ and $y \in \operatorname{span}(H_1)$ such that $v_{m+1} = x + y$ (by Proposition 0(e)). Select such x and y. Note that if $y = 0_V$ then $v_{m+1} \in \operatorname{span}(L)$, contradicting the assumed linear independence of L (by Proposition 0(c)). Hence $y \neq 0_V$, implying that $H_1 \neq \emptyset$. Thus n - m > 0, so $n - m \ge 1$. Equivalently, $m + 1 \le n$, which is exactly the statement that $|L| \le |G|$. This establishes assertion (a) of the theorem for our given set L.

(*Note*: In our diagram, the elements v_1, \ldots, v_m are present only if $m \neq 0$. [This is I defined L_1 by saying " $L_1 = L \setminus \{v_{m+1}\}$ " rather than " $L = \{v_1, \ldots, v_m\}$ ".] Since $n - m \geq 1$, there is always a u_1 present. The elements u_2, \ldots, u_{n-m} are present only if $n - m \geq 2$. The notation " $H_1 \setminus \{u_1\}$ " would be more precise notation for " $\{u_2, \ldots, u_{n-m}\}$ ", being valid even if n - m = 1.

Also note that, by definition, all the v's are distinct from each other, and all the u's are distinct from each other. We do not know, or care, whether $v_i = u_j$ for some i, j. However, because it is possible that v_{m+1} is u_j for some j, it is possible that $G_1 \cup \{v_{m+1}\} = G_1$. I.e. the set $G_1 \cup \{v_{m+1}\}$ might not actually have one more element than G_1 does; both sets could be the same. This makes absolutely no difference to our proof; I am simply pointing it out so that

you don't think that $G_1 \cup \{v_{m+1}\}$ is *automatically* larger than G_1 . That misimpression could confuse you if, for example, you wonder how this proof could possibly handle the case L = G.)

 \in Since y $\operatorname{span}(H_1),$ there exist scalars b_1, \ldots, b_{n-m} such that $y = b_1 u_1 + \cdots + b_{n-m} u_{n-m}$; furthermore, since $y \neq 0_V$, at least one of these scalars must be nonzero. ((1) The inference that "at least one of these scalars must be nonzero" is crucial; this is what makes the whole argument work!! (2) Selecting an *i* for which $b_i \neq 0$, the corresponding element of H_1 is what we've called " u_{whatever} " in line (3) of the diagram.) Without loss of generality we may assume $b_1 \neq 0$. (This amounts to relabeling the elements of H_1 so that " u_{whatever} " in line (3) of the diagram becomes u_1 in line (4).) Let $H = H_1 \setminus \{u_1\}$. Then $y = b_1 u_1 + y'$ for some $y' \in \text{span}(H)$. (Specifically, $y' = b_2 u_2 + \cdots + b_{n-m} u_{n-m}$ if $n-m \ge 2$, and $y' = 0_V$ if n-m = 1.) Therefore $v_{m+1} = x + y = x + b_1 u_1 + y'$, which implies $b_1u_1 = v_{m+1} - x - y'$, and since $b_1 \neq 0$, it follows that

$$u_1 = \underbrace{\left(\frac{1}{b_1}v_{m+1} - \frac{1}{b_1}x\right)}_{\in \operatorname{span}(L)} + \underbrace{\frac{-1}{b_1}y'}_{\in \operatorname{span}(H)},$$

the sum of an element of span(L) and an element of span(H). Thus $u_1 \in \text{span}(L \cup H)$, implying that $\text{span}(L \cup H \cup \{u_1\}) = \text{span}(L \cup H)$ (by Proposition 0(a)). But $L \cup H \cup \{u_1\} = G_1 \cup \{v_{m+1}\}$. (Compare lines (2) and (5) of the diagram.) Since $\text{span}(G_1)$ is already all of V, so is the span of the more inclusive set $G_1 \cup \{v_{m+1}\}$. Hence

$$\operatorname{span}(L \cup H) = \operatorname{span}(L \cup H \cup \{u_1\}) = G_1 \cup \{v_{m+1}\} = \operatorname{span}(G_1) = V_1$$

(This string of equalities is the reason we "augmented" the set G_1 on line (1) of the diagram to the set $G_1 \cup \{v_{m+1}\}$ on line (2). We needed to show that inserting the element v_{m+1} would allow us to delete one of the elements of H_1 from $G_1 \cup \{v_{m+1}\}$ without "losing span".) This establishes assertion (b) of the theorem for our given set L.

Hence assertions (a) and (b) of the theorem are true whenever |L| = m + 1. By induction, these assertions hold for linearly independent subsets $L \subseteq V$ of any finite cardinality.

Here is the same proof of the "Replacement Theorem" with all the blue comments and the diagram removed (as well as some simple steps, also in blue previously, that would be okay to treat as obvious to any competent reader without the writer's help).

Proof: We will proceed by induction on the cardinality of L. First suppose that |L| = 0. Then $L = \emptyset$, and H = G itself is a subset of G of cardinality |G| - |L| = |G|. Trivially, $L \cup H = \emptyset \cup G = G$, so $\operatorname{span}(L \cup H) = \operatorname{span}(G) = V$. Just as trivially, $0 = |L| \le |G|$. This establishes (a) and (b) in the case |L| = 0.

Suppose now that m is a non-negative integer such that statements (a) and (b) are true whenever $L \subseteq V$ is linearly independent set of cardinality $|L| \leq m$. Assume now that |L| = m + 1 and let v_1, \ldots, v_{m+1} be the elements of L. Let n = |G|. Then $L_1 := L \setminus \{v_{m+1}\}$ is an m-element linearly independent subset of V, so by the inductive hypothesis, $m \leq n$ ($|L_1| \leq |G|$) and we may select a subset $H_1 \subseteq G$ of cardinality n - m such that $G_1 := L_1 \cup H_1$ spans V. Let u_1, \ldots, u_{n-m} be the elements of H_1 .

Since $L_1 \cup H_1$ spans V (part of our inductive hypothesis), in particular $v_{m+1} \in \text{span}(G_1)$, so there exist $x \in \text{span}(L_1)$ and $y \in \text{span}(H_1)$ such that $v_{m+1} = x + y$. Select such x and y. Note that if $y = 0_V$ then $v_{m+1} \in \text{span}(L)$, contradicting the assumed linear independence of L. Hence $y \neq 0_V$, implying that $H_1 \neq \emptyset$. Thus n - m > 0, so $n - m \ge 1$. Equivalently, $m + 1 \le n$, which is exactly the statement that $|L| \le |G|$. This establishes assertion (a) of the theorem for our given set L.

Since $y \in \operatorname{span}(H_1)$, there exist scalars b_1, \ldots, b_{n-m} such that $y = b_1u_1 + \cdots + b_{n-m}u_{n-m}$; furthermore, since $y \neq 0_V$, at least one of these scalars must be nonzero. we may assume $b_1 \neq 0$. Let $H = H_1 \setminus \{u_1\}$. Then $y = b_1u_1 + y'$ for some $y' \in \operatorname{span}(H)$. Therefore $v_{m+1} = x + y = x + b_1u_1 + y'$, and it follows that $u_1 = \left(\frac{1}{b_1}v_{m+1} - \frac{1}{b_1}x\right) + \frac{-1}{b_1}y'$, the sum of an element of $\operatorname{span}(L)$ and an element of $\operatorname{span}(H)$. Thus $u_1 \in \operatorname{span}(L \cup H)$, implying that $\operatorname{span}(L \cup H \cup \{u_1\}) = \operatorname{span}(L \cup H)$. But $L \cup H \cup \{u_1\} = G_1 \cup \{v_{m+1}\}$. Since $\operatorname{span}(G_1)$ is already all of V, so is the span of the more inclusive set $G_1 \cup \{v_{m+1}\}$. Hence

$$\operatorname{span}(L \cup H) = \operatorname{span}(L \cup H \cup \{u_1\}) = G_1 \cup \{v_{m+1}\} = \operatorname{span}(G_1) = V.$$

This establishes assertion (b) of the theorem for our given set L.

Hence assertions (a) and (b) of the theorem are true whenever |L| = m + 1. By induction, these assertions hold for linearly independent subsets $L \subseteq V$ of any finite cardinality.

Corollary 14

(a) In a finite-dimensional vector space, a linearly independent set can never have <u>more</u> elements than any given, finite, spanning set. (Thus the "Replacement Theorem" remains true even if we delete the hypothesis that L is finite; that hypothesis is unnecessary.)

(b) A finite-dimensional vector space cannot contain an infinite linearly independent set.

Proof: Let V be a finite-dimensional vector space and let G be a finite spanning set for V. Let $L \subseteq V$ be a linearly independent set.

First suppose that the set L is infinite. Then L has finite subsets of arbitrarily large cardinality. Let $L' \subseteq L$ be a subset of cardinality greater than |G|. Since any subset of a linearly independent set is linearly independent (Proposition 0(b)), L' is a finite linearly independent set in V whose cardinality exceeds |G|, contradicting part (a) of the "Replacement Theorem". Hence L must be finite.

This establishes statement (b). Statement (a) then follows from the Replacement Theorem's part (a). \blacksquare

Remark 15 Although the blue comment I inserted into Corollary 14 may seem to follow "obviously" from statement (a) of the corollary, technically the comment is premature, because (i) Notation 7 does not tell us what "|L|" means if L is an infinite set; (ii) although there *is* a definition of what *cardinality* means for infinite sets, we have not given that definition in this course; and (iii) even had we given that definition, we'd also have to define what " \leq " means when comparing an infinite cardinality (the cardinality of an infinite set) to a finite cardinality (the cardinality of a finite set). However, as we would intuitively expect, it *is* true that when these definitions are given properly, and we use the notation "|A|" for the cardinality of an *arbitrary* set A, then $|A| \not\leq |B|$ if A is an infinite set and B is a finite set.³

Corollary 14 is one of many important corollaries of the "Replacement Theorem". The *most* important of these corollaries is this:

Theorem 16 Let V be a finite-dimensional vector space. Then all bases of V are finite and have the same number of elements.

Proof: Let β and β' be bases of V. Since bases are linearly independent sets, and V is finite-dimensional, Corollary 14(b) guarantees us that β and β' are finite sets.

Since β is linearly independent and β' spans V, Theorem 13(a) (the "Replacement Theorem", part (a)) implies that $|\beta| \leq |\beta'|$. Similarly, since β' is linearly independent and β spans V, Theorem 13(a) implies that $|\beta'| \leq |\beta|$. Hence $|\beta'| = |\beta|$.

 $^{{}^{3}}$ " $|A| = \infty$ " is not a mathematically meaningful equation for a general infinite set A. However, for purposes of reminding ourselves of what Theorem 13(a) would say if we removed the redundant "L is finite" hypothesis, no great harm is done if we mentally substitute the symbol " ∞ " for any infinite cardinality, and agree to regard " $\infty > n$ " (or " $\infty \leq n$ ") as a true statement for any integer n.

Dimension

Theorem 16 is crucial to the notion of *dimension*:

Definition 17 Let V be a finite-dimensional vector space. The *dimension* of V, written $\dim(V)$, is the cardinality of any basis of V.

Note that this definition makes sense only because of two of our previous results: first, that *some* basis of V exists (by Proposition 8); and second, Theorem 16's guarantee that whatever basis of V we choose to count the elements of, we will always get the same number.

Examples. In class we exhibited bases for the vector spaces $\{0\}, \mathbf{R}^n \ (n \geq 1), M_{m \times n}(\mathbf{R}) \ (m, n \geq 1), \text{ and } P_n(\mathbf{R}) \ (n \geq 0).$ Counting the elements in those bases, we see that $\dim(\{0\}) = 0$; $\dim(\mathbf{R}^n) = n$; $\dim(M_{m \times n}(\mathbf{R})) = mn$; and $\dim(P_n(\mathbf{R})) = n + 1$. We also saw that the space $P(\mathbf{R})$ does not have a finite spanning set; hence $P(\mathbf{R})$ is infinite-dimensional.

Knowing the dimension of a given finite-dimensional vector space V (e.g., knowing that $\dim(\mathbf{R}^n) = n$) can greatly simplify the task of determining whether certain subsets of V, or lists in V, can possibly be linearly independent or can possibly span V.

Proposition 18 Let V be a finite-dimensional vector space and let $n = \dim(V)$.

- (a) No subset of V with <u>more</u> than n elements can be linearly independent.
- (a') No list of vectors in V with more than n terms can be linearly independent.
- (b) No subset of V with fewer than n elements can span V.
- (b') No list of vectors in V with fewer than n terms can span V.
- (c) For any subset S of V with exactly n elements, the following are equivalent:
 - (i) S is linearly independent;
 - (ii) S spans V;
 - (iii) S is a basis of V.
- (c') For any list L of vectors in V with exactly n terms, the following are equivalent:
 - (i) L is linearly independent.
 - (ii) L spans V.

(Thus, given a vector space V that we **already** know has dimension n [e.g. \mathbb{R}^{n}], and a specific set S of exactly n vectors in V if we wish to check whether S is a basis of V it suffices to check <u>either</u> that S is linearly independent <u>or</u> that S spans V; we do not have to check *both* of these properties of a basis. The same is true for the set S_L of terms of a list in V with exactly n terms.)

Reminder: In a *list* of vectors, the terms do not have to be distinct; some term(s) can duplicate other term(s). Thus, for the set called S_L above—the set of terms of a list L—if L is an *m*-term list, then $|S_L| \leq m$, with equality holding if and only if the terms of L are distinct.

Note: "the set of terms of L" and "the set of *distinct* terms of L" mean exactly the same thing. Sometimes I insert the word "distinct" into "the set of terms of L" just as a reminder of what the elements of "the set of terms of L" are.

Taken together, parts (a'), (b'), and (c') of Proposition 18 constitute Proposition 13 of the "Lists, linear combinations, and linear independence" handout. (The proof given in the latter handout—originally written for my Fall 2023 class, later in the semester—relied on a theorem about linear transformations, which, in Fall 2024, we had not even defined when that handout was posted. The proof given below does not presume any knowledge of linear transformations; it uses nothing not yet covered at the time it's being posted in Fall 2024.)

Proof: Let β be a basis of V (thus $|\beta| = n$).

For parts (a), (b), and (c) let $S \subseteq V$. For parts (a'), (b'), and (c') let $m \ge 1$, let L be an *m*-term list in V, and let $S_L \subseteq V$ be the set of (distinct) terms of L. (Thus $|S_L| \le m$, with equality holding if and only if the terms of L are distinct.)

(a) Since β spans V and $|\beta| = n$, Corollary 14(a) implies that if |S| > n then S is not linearly independent.

(a') Assume that m > n. If the terms of L are all distinct, then $|S_L| = m > n$, and part (a) above implies that S_L is linearly dependent. If the terms of L are not all distinct, then by Proposition 3 of the "Lists, linear combinations, and linear independence" handout (henceforth "the 'Lists ...' handout"), S_L again is linearly dependent.

Thus in either case, the set S_L is linearly dependent. By Proposition 8(b) of the "Lists ..." handout, it follows that the list L is linearly dependent.

(b) Since β is linearly independent and $|\beta| = n$, Theorem 13(a) implies that if |S| < n then S does not span V.

(b') Assume that m < n. Since $|S_L| \le m$, it follows that $|S_L| < n$. Hence part (b) above implies that $\operatorname{span}(S_L) \ne V$. But by Proposition 8(a) of the "Lists ..." handout, $\operatorname{span}(S_L) = \operatorname{span}(L)$, and thus L does not span V.

(c) Assume that |S| = n. By definition of *basis*, (iii) is equivalent to "(i) and (ii)".

Hence it suffices to show "(i) \iff (ii)" (since then if either condition (i) or (ii) holds, they both do).

 $((i) \implies (ii))$ Assume that S is linearly independent. Suppose that S does not span V. Let $v \in V$ be a vector not in span(S), and let $S' = S \cup \{v\}$. Then S' is linearly independent (by Proposition 0(d)) but has cardinality n + 1 > n, contradicting part (a) above.

 $((ii) \Longrightarrow (i))$ Assume that S spans V. Suppose that S is linearly <u>dependent</u>. Then $S \neq \emptyset$, and there exists $v \in S$ such that $v \in \text{span}(S \setminus \{v\})$ (by Proposition 0(c)). Select such v and let $S' = S \setminus \{v\}$. Then $v \in \text{span}(S')$ and $S' \cup \{v\} = S$, so (using Proposition 0(c)) $\text{span}(S') = \text{span}(S' \cup \{v\}) = \text{span}(S) = V$. But then S' is a spanning set for V with n-1 elements, contradicting part (b) above.

(c') We will again use Propositions 3, 8(a), and 8(b) of the "Lists ..." handout several times (the same ways that they are used earlier in this proof), but without mentioning each use explicitly.

 $((i) \implies (ii))$ Assume that L is linearly independent. Then the terms of L are distinct, so $|S_L| = m = n$. Since L is linearly independent, so is S_L . By part (c) above, S_L spans V. But span $(S_L) = \text{span}(L)$, so L spans V.

 $((ii) \implies (i))$ Assume that L spans V. Then S_L spans V, so by part (b) above, $|S_L| \ge n$. But $|S_L| \le ($ number of terms of L) = n, so $|S_L| = n = ($ number of terms of L). Hence the terms of L are distinct. Since the *n*-element set S_L spans V, part (c) above impliest that S_L is linearly independent. Since the terms of L are distinct, Proposition 8(c) of the "Lists ..." handout shows that the the linear independence of S_L implies that L is linearly independent.

Remark 19 Since \emptyset is a basis of the trivial vector space $\{0\}$, it follows that $\dim(\{0\}) = 0$. Conversely, the trivial vector space is the *only* vector space of dimension zero. (If $V \neq \{0\}$ then V contains a nonzero vector, hence a linearly independent set with one element. If V is finite-dimensional, Proposition 18(a) shows that $\dim(V) > 0$. If V is infinite-dimensional, then we currently have no definition of what $\dim(V)$ is, but no harm is done if we simply agree that $\dim(V)$ is *not* zero in this case. If we were digress in order to give a proper definition of $\dim(V)$ for an infinite-dimensional vector space V, we would see that, indeed, $\dim(V)$ is not 0 or any other real number.)

Previously we saw that a basis of any vector space is, simultaneously, a maximal linearly independent set and a minimal spanning set (Proposition 4). In a finite-dimensional vector space, Proposition 18 yields the following even stronger statement:

Corollary 20 Let V be a finite-dimensional vector space and let β be a basis of V. Then

- (a) β has maximal cardinality among all linearly independent subsets of V, and
- (b) β has minimal cardinality among all spanning subsets of V.

Proof: Left to student.

The student should also think about why the maximality/minimality assertions in Corollary 20 are *stronger* than what Proposition 4, by itself, yields.

Next:

Proposition 21 Let V be a finite-dimensional vector space and let $S \subseteq V$.

- (a) If S is linearly independent, then S can be "extended" to a basis of V; i.e. there exists some subset $T \subseteq V$, disjoint from S, such that $S \cup T$ is a basis of V.
- (b) If S spans V, then S can be "shrunk" to a basis of V; i.e. some subset of S is a basis of V.

Note that in part (a), we did not say that T is nonempty! It's possible that S is already a basis of V, in which case we're going to take $T = \emptyset$.

Proof: Let $n = \dim(V)$.

(a) Assume that S is linearly independent. Then by Proposition 18(a), S is finite and $|S| \leq n$. Let m = |S|; thus $m \leq n$. If m = n then Proposition 18(c) implies that $S = S \cup \emptyset$ is already a basis of V.

Assume now that m < n. By Proposition 18(b), S does not span V. Let $v_1 \in V$, with $v_1 \notin \operatorname{span}(S)$. Then $S_1 := S \cup \{v_1\}$ is linearly independent and has cardinality m+1. If m+1 = n, then, by Proposition 18(c), S_1 is a basis of V. If m+1 < n, then, by the same argument, we may select $v_2 \in V$ such that $S_2 := S_1 \cup \{v_2\} = S \cup \{v_1, v_2\}$ is linearly independent. If m+2 = n then S_2 is a basis of V. Otherwise, continuing in this fashion, we produce vectors $v_1, \ldots, v_{n-m} \in V$, none of which lies in S, for which $S \cup \{v_1, \ldots, v_{n-m}\}$ is a basis of V.

(b) If S is finite, then the conclusion follows from Proposition 8 and Corollary 11, so assume S is infinite.

Let \mathcal{F} be the collection of all finite, linearly independent subsets of S and let $\mathcal{N} = \{|A| : A \in \mathcal{F}\}$. Note that $\emptyset \in \mathcal{F}$, so $0 \in \mathcal{N}$; thus \mathcal{N} is nonempty. By Proposition 18(a), $|A| \leq n$ for every $A \in \mathcal{F}$. Hence every $j \in \mathcal{N}$ satisfies $0 \leq j \leq n$. Thus the set $\mathcal{N} \subseteq \mathbb{Z}$ is finite and nonempty, and therefore contains a largest element m. Since $m \in \mathcal{N}$, we know that $0 \leq m \leq n$.

Suppose that m < n. By definition of \mathcal{N} , there is some *m*-element linearly independent subset T of S. By Proposition 18(b), T does not span V, since |T| = m < n.

Since $S = T \cup (S \setminus T)$, Proposition 0(e) tells us that $\operatorname{span}(S) = \operatorname{span}(T) + \operatorname{span}(S \setminus T)$; hence $V = \operatorname{span}(T) + \operatorname{span}(S \setminus T)$ (since $\operatorname{span}(S) = V$ by hypothesis).

If every element of $S \setminus T$ lies in $\operatorname{span}(S)$, then so does every linear combination of elements of $S \setminus T$; i.e. $\operatorname{span}(S \setminus T) \subseteq \operatorname{span}(S)$. But then

$$V = \operatorname{span}(S) + \operatorname{span}(S \setminus T) = \operatorname{span}(S),$$

a contradiction. (Here we have used the fact that if X and Y are subspaces of V, and $Y \subseteq X$, then X + Y = X. The proof of this fact is an easy exercise, left to the student.) Hence there exists some $v \in S \setminus T$ that is not in span(T). For any such v, the set $T \cup \{v\}$ is linearly independent (by Proposition 0(d)) and is a finite subset of S; i.e. $T \cup \{v\} \in \mathcal{F}$, and thus $|T \cup \{v\}| \in \mathcal{N}$. But $|T \cup \{v\}| = m + 1 > m$, a contradiction (since m, by definition, is the largest element of \mathcal{N}).

Hence our assumption that m < n was false, and we conclude that m = n. Thus, since $n = m \in \mathcal{N}$, there is a linearly independent subset $T \subseteq S$ of cardinality n. But by Proposition 18(c), any such T is a basis of V.

Dimension of subspaces

Proposition 22 Let V be a finite dimensional vector space and let $n = \dim(V)$.

- (a) Let $W \subseteq V$ be a subspace. Then:
 - (i) W is finite-dimensional and $0 \leq \dim(W) \leq n$.
 - (ii) dim(W) = 0 if and only if $W = \{0_V\}$.
 - (iii) $\dim(W) = n$ if and only if W = V.
- (b) Let m be an integer for which $0 \le m \le n$. Then V has a subspace of dimension m.

Proof: (a) (i) If $W = \{0_V\}$ then W is finite-dimensional and $\dim(W) = 0$. Assume now that $W \neq \{0_V\}$, and let w_1 be a nonzero vector in W. If $\operatorname{span}\{w_1\} \neq W$, let $w_2 \in W$ be an element not in $\operatorname{span}\{w_1\}$. Then $\{w_1, w_2\}$ is a linearly independent subset of W (by Proposition 0(d)). Continuing in this fashion, suppose we have produced vectors $w_1, \ldots, w_k \in W$ such that $\{w_1, \ldots, w_k\}$ is linearly independent. If $\operatorname{span}\{w_1, \ldots, w_k\} \neq W$, then the same argument shows that there exists $w_{k+1} \in W$ such that $\{w_1, \ldots, w_{k+1}\}$ is linearly independent. But then $\{w_1, \ldots, w_{k+1}\}$ is also a (k+1)-element linearly independent subset of V, so by Proposition 18(a), we must have $k + 1 \leq n$; we cannot keep finding vectors in W indefinitely that are not in the span of the already-found $\{w_1, \ldots, w_k\}$. Thus, there must be some $m \leq n$ for which our linearly independent set $\beta_W := \{w_1, \ldots, w_m\}$ is a basis of W. Hence $m \leq n$.

(ii) Shown in Remark 19.

(iii) Clearly if W = V then $\dim(W) = n$. Conversely, suppose that $\dim(W) = n$ and let β_W be a basis of W. Then β_W is an *n*-element linearly independent subset of V, so by Proposition 18(c), β_W is a basis of V. Hence $W = \operatorname{span}(\beta_W) = V$.

(b) If n = 0 then $V = \{0_V\}$, m = 0, and V is an m-dimensional subspace of itself.

Suppose now that n > 0 and let $\beta = \{v_1, \ldots, v_n\}$ be a basis of V. Let $\beta_m = \{v_i : 1 \leq i \leq m\} = \{v_1, \ldots, v_m\}$. Then β_m is a subset of the linearly independent set β , hence is linearly independent. Let $W = \operatorname{span}(\beta_m)$. Then β_m spans W and is linearly independent, so β_m is a basis of W. Hence W is an m-dimensional subspace of V.

Remark 23 If V is a finite-dimensional vector space and the strict inequalities $0 < m < n = \dim(V)$ hold, then V has *infinitely many* subspaces of dimension m. The student should be able to show this easily if m = 1, and a little less easily (but not with great difficulty) for larger m.

The student should also be able to show that if V is infinite-dimensional, then V has an *m*-dimensional subspace for every m > 0.

Remark 24 It is important to keep in mind that, given a finite-dimensional vector space V, the dimension of V is **NOT** the number of elements of V; the dimension of V is the number of elements in a basis of V. A basis of \mathbb{R}^2 has more elements than a basis of \mathbb{R} (these bases have two elements and one element, respectively), but the vector space \mathbb{R}^2 itself does not have "more" elements than the vector space \mathbb{R} . In fact, with some cleverness and work, one can write down a bijection $f : \mathbb{R}^2 \to \mathbb{R}$ (a function that is one-to-one and onto; hence it is invertible and the inverse function is, in particular, a surjective [onto] function $f^{-1} : \mathbb{R} \to \mathbb{R}^2$. More generally, for any m, n > 0, a bijection from \mathbb{R}^n to \mathbb{R}^m exists.⁴ (Exhibiting such bijections is outside the scope of this course, though it is not above the *level* of this course.)

⁴Once cardinality of a general, possibly infinite, set is defined, this fact says that \mathbf{R}^n has the same cardinality as \mathbf{R}^m .

Unique expansion in terms of a basis

We conclude these notes with the following **extremely** important property of bases. (In Fall 2024 we proved this in class before covering the other material in these notes. I've left it for last in these notes just to avoid interrupting the logical flow of the other material.)

Proposition 25 ("Unique expansion in terms of a basis")

Let V be a finite-dimensional vector space of dimension n > 0 and let $\{v_1, \ldots, v_n\}$ be a basis of V. Then for each $v \in V$, there exists a <u>unique</u> n-tuple (c_1, \ldots, c_n) of scalars such that

$$v = c_1 v_1 + \dots + c_n v_n . ag{6}$$

Proof: Let $v \in V$. The basis $\{v_1, \ldots, v_n\}$ spans V, so an *n*-tuple (c_1, \ldots, c_n) of scalars exists such that the equality (6) holds. Let (c_1, \ldots, c_n) be such an *n*-tuple, and suppose that (b_1, \ldots, b_n) is any *n*-tuple of scalars for which $v = b_1v_1 + \cdots + b_nv_n$. Then

$$b_1v_1 + \dots + b_nv_n = c_1v_1 + \dots + c_nv_n,$$

$$\implies \qquad (b_1 - c_1)v_1 + \dots + (b_n - c_n)v_n = 0_V,$$

$$\implies \qquad \qquad b_i - c_i = 0, \quad 1 \le i \le n,$$

since, being a basis, the set $\{v_1, \ldots, v_n\}$ is linearly independent. Hence $(b_1, \ldots, b_n) = (c_1, \ldots, c_n)$. Thus the *n*-tuple (c_1, \ldots, c_n) for which (6) holds is unique.

[Content of notes ends here, but there's a summary on next page.]

Summary of some highlights

Below, V is always a vector space, and n and m are non-negative integers.

- V is called *finite-dimensional* if V has a finite basis, and *infinite-dimensional* otherwise.
- If V is finite-dimensional, then all bases of V are finite and have the same cardinality. This cardinality is called the *dimension* of V and written $\dim(V)$.
- Dimensions of some common finite-dimensional vector spaces:
 - $-\dim(\mathbf{R}^n) = n \text{ (for } n > 0).$
 - $-\dim(\{\theta\}) = 0.$
 - $-\dim(M_{m\times n}(\mathbf{R})) = mn \text{ (for } m, n > 0).$
 - $-\dim(P_n(\mathbf{R})) = n+1.$
- Suppose V is finite-dimensional and let $n = \dim(V)$. Then:
 - No subset of V with <u>more</u> than n elements can be linearly independent.
 - No subset of V with <u>fewer</u> than n elements can span V.

(The previous two facts provide an easy way of remembering that no linearly independent subset of V can have <u>more</u> elements than a spanning set has.)

- If S is a subset of V with *exactly* n elements, then S is linearly independent if and only if S spans V.

Hence (under the assumption that S has exactly n elements), the following are equivalent:

- (i) S is linearly independent.
- (ii) S spans V.
- (iii) S is a basis of V.

Thus, given a vector space V that we **already** know has dimension n (e.g. \mathbf{R}^n), and a specific set S of exactly n vectors in V, if we wish to check whether S is a basis of V, it suffices to check <u>either</u> that S is linearly independent <u>or</u> that S spans V; we do not have to check *both* of these properties of a basis.

- If V is finite-dimensional, then every *linearly independent* subset $S \subseteq V$ can be *extended* to a basis (or already is one). (I.e. V has a basis that contains the set S. The sense of "extend[ing]" here means "throwing in additional elements of V.")
- If V is finite-dimensional, then every spanning subset $S \subseteq V$ contains a basis. (I.e. some subset of S is a basis of V.)