July 8 (Lecture 15)

Overview: We'll start today with dimension (finally). After seeing that definition and some examples, we'll extend the concept of linear maps to linear maps between abstract vector spaces.

Learning Goals:

- Precisely define and calculate dimension of a vector space.
- Explain (using theorems) how dimension relates to spanning sets, linearly independent sets, and bases.
- Define and work with linear maps between abstract vector spaces.

As you're getting settled: • Reflection 10 available after class!

- Homework 8 is due tomorrow (Friday, July 9), at 11:30 pm.
- Homework 9 to come out tomorrow, due Tuesday July 20th at 11:30 pm.
- Test 2 is next Thursday! July 15. It will be available on Crowdmark 10:00 am to 8:00 pm Pacific time but you'll only get 1.5 hours to do it, same as last time. I'll try to post some targetted practice problems that I think might be helpful.
- Watch for HW or Test instructions that say "You may use a computer or calculator to perform [specific computations]"!

(not §93)

(4.3.2) If T is a spanning set of K vectors for V, then some subset of T is a boois for V.

(4.3.5) If S is a LI set in V w) fever then dim(V) vectors, the s can be extended to a basis for V.

(4.3.6) If U is a subspace of V_1 then $\dim(U) \leq \dim(V)$.

"Goldilock Thorem"

p. 259 Theorem (4.3.7). Let V be a vector space with $\dim(V) = n \ge 1$.

1. If SEV has more thin in Vectors, the S is linearly dependent.

2. If SEV has feaser than n vectors, the s cannot spain V.

3. If BCV has exactly in vectors than B is linearly independent if and only if B spains V. (44).

Example. Let $A = \begin{bmatrix} 1 & 0 \\ 1 & 2 \end{bmatrix}$ and $B = \begin{bmatrix} 0 & 2 \\ -2 & 0 \end{bmatrix}$. Let $W = \text{span} \{A, A^T, B, B^T, A + B\}$. Find dim(W).

Solution. Observe: by part 1 of 4.3.7, since $5 \times \dim(M_{2,a}(\mathbb{R})) = 4$, thus spanning set is not LI. <u>Note</u>: A+B is in span $\xi A_1B_3^2$, $B^T = -B$, and $A - A^T = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix} = \frac{-1}{2}B_1$ so $A^T = A + \frac{1}{2}B \in \text{span} \xi A_1B_3^2$. $=> U = \text{span} \xi A_1B_3^2$. Ue can show that $\xi A_1B_3^2$ is LI, so it is also a bois for U! $=> \dim(U) = 3$. • If we wanted to extend $\xi A_1B_3^2$ into a bois for $(M_{2,a}(\mathbb{R}))$, then by 4.3.7(3), we simply need to add two more vectors to get a LI set.

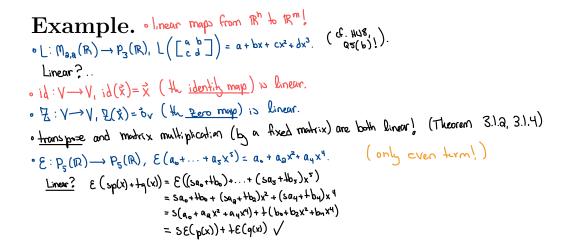
(try [':"] and [";"])

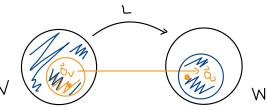
Section 4.5

General Linear Maps

p. 273

Definition. Let V and W be vector spaces. A map $L: V \to W$ is a *linear* map (or transformation) when for all $\hat{\mathbf{x}}, \hat{\mathbf{y}} \in V$, $s, t \in \mathbb{R}$, ue have $L(s\hat{\mathbf{x}}+t\hat{\mathbf{y}})=sL(\hat{\mathbf{x}})+JL(\hat{\mathbf{y}})$.





- \mathfrak{p} . $\mathfrak{A74}$ **Definition.** Let $L: V \to W$ be a linear map between vector spaces V and W.
 - The range of L, denoted $\operatorname{Range}(L)$, is $\{ L(x) : x \in V \} \subseteq U$.
 - The null space of L, denoted Null(L), is $\xi \times \epsilon \vee : L(x) = \delta_u \xi \vee v_x$

Example.
$$\xi: P_{5}(\mathbb{R}) \longrightarrow P_{5}(\mathbb{R}), \xi(a_{0}+...+a_{5}x^{5}) = a_{0}+a_{2}x^{2}+a_{4}x^{4}.$$

When is $\xi(p(x)) = 0$? \Longrightarrow $a_{0} = a_{2} = a_{4} = 0$. No reductions on $a_{1}/a_{3}, a_{5}.$
 $a_{0}+a_{3}x^{2}+a_{4}x^{4}$
 \Longrightarrow Null(ϵ) = $\xi a_{1}x + a_{3}x^{3}+a_{5}x^{5}: a_{1}, a_{3}, a_{5} \in \mathbb{R}$ $\xi = span \{x, x^{3}, x^{7}\}$.
For Range (ϵ), let $b_{0}+b_{2}x^{2}+b_{4}x^{4} \in P_{5}(\mathbb{R})$. Every $p(x) \leftarrow P_{5}(\mathbb{R})$ such that $a_{0}=b_{0}, a_{2}=b_{2}, a_{4}=b_{4}$ so the basis of $\epsilon(p(x)) = b_{0}+b_{2}x^{2}+b_{4}x^{4}$.
 \Longrightarrow Range (ϵ): span $\xi |_{1}x^{4}, x^{4}\}$.
Define the trace of $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ to be $t_{1}(A) = a + b$ (can generalize b nixn).
The trace is a linear map from $M_{2xz}(\mathbb{R})$ to $\mathbb{R}^{(1)}$.
Null(t_{1}): if $t_{1}(A) = 0$, then $a_{1}d = 0$, and no other restrictions.
 $\Longrightarrow \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{bmatrix} -d & b \\ c & d \end{bmatrix} = >$ Null(t_{1}) = span $\xi \begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0$

The

=> Range (tr) = R.

Definition. Let $L: V \to W$ be a linear map between vector spaces p.276 V and W.

• The rank of L, denoted $\operatorname{rank}(L)$, is dim (Range (L)).

• the *nullity* of L, denoted nullity(L), is ∂_{m} (Null(L)). For L: $\mathbb{R}^n \to \mathbb{R}^m$, rank(L) = dim(Range(L)) = dim(col([L])) = rank ([L]).

(df. Theorem 3.4.9).

p. 277 Theorem (4.5.2, Rank-Nullity). Let L:V→W be a linear map between abstract vector spaces v/ dim(V)=n. Then, rank(L)+ nullity(L)=n=dim(V).

Proof.

First, We can find a bood $\{\vec{v}_{1},...,\vec{v}_{k}\}$ for Null(L) (so nullity(L) = k), which we can extend to a bood $\{\vec{v}_{11},...,\vec{v}_{k},\vec{w}_{k+11},...,\vec{w}_{n}\}$ for V. (Theorem 4.3.5). Now, if $\vec{w} \in Range(L)$, then by definition we have $\vec{w} = L(\vec{v}) = L(a_{n}\vec{x}_{1} + ... + a_{n}\vec{x}_{k} + a_{k+1}\vec{w}_{k+1} + ... + a_{n}\vec{u}_{n})$ $= a_{k+1}L(\vec{w}_{k+1}) + ... + a_{n}L(\vec{w}_{n})$, since L is linear and the $\vec{v}_{i} \in Null(U)$. => Range(L)= span $\{L(\vec{w}_{k+1}+... + a_{n}L(\vec{w}_{n})\} = L(a_{k+1}\vec{w}_{k+1} + ... + a_{n}\vec{u}_{n}),$ then $a_{k}\vec{u}_{k+1} + ... + a_{n}L(\vec{w}_{n}) = L(a_{k+1}\vec{w}_{k+1} + ... + a_{n}\vec{u}_{n}),$ then $a_{k}\vec{u}_{k+1} + ... + a_{n}L(\vec{w}_{n}) = L(a_{k+1}\vec{w}_{k+1} + ... + a_{n}\vec{u}_{n}),$ => Sio LI, the a basis for Range(L), and theo rank(L) + nullity(L) = n-k+k = n.