

10. Chapter 3: Vector spaces

\mathbb{R}^n is a (real) vector space (over \mathbb{R}) *or:* *there are others; general def' after midterm*
 \mathbb{C}^n (complex) \mathbb{C}
 \mathbb{Q}^n (rational) \mathbb{Q}
 \mathbb{F}_2 (binary) \mathbb{F}_2

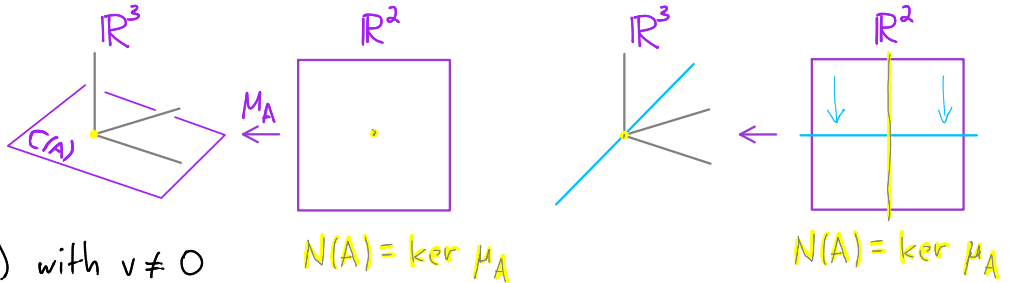
These are the "flat things" we've been talking about all along, though only the ones through 0.

Def: A subset $V \subseteq \mathbb{R}^n$ is a subspace if 1. $V \neq \emptyset$ and 2. $v, w \in V$ and $c \in \mathbb{R} \Rightarrow v + cw \in V$.

Remark 1. can use instead: $0 \in V$ ($v=w; c=-1$) *but it's not easier to check*
 2. $\Rightarrow V$ closed under arbitrary linear combinations (Pf: induction.)

Examples

- $V = \{0\}$
- $V = \mathbb{R}^n$



3. line $l = \text{span}(v)$ with $v \neq 0$

4. hyperplane $H = \{x \in \mathbb{R}^n \mid ax = 0\}$ for some fixed $a \in \mathbb{R}^{\text{row}}$

5. the nullspace of any matrix $A \in \mathbb{R}^{m \times n}$ Pf: $A0 = 0$ and μ_A is linear!

$$N(A) = \{x \in \mathbb{R}^n \mid Ax = 0\} (= \text{kernel of } \mu_A) = \text{sols } [A|0]$$

6. $\text{span}(v_1, \dots, v_k)$ for any $\begin{bmatrix} v_1, \dots, v_k \\ 1 \end{bmatrix} \in \mathbb{R}^n$ Pf: 1. $0 = 0v_1$.
 $= C(A)$ for $A = \begin{bmatrix} 1 & \dots & 1 \end{bmatrix} \in \mathbb{R}^{n \times k}$

2. $v = c_1v_1 + \dots + c_kv_k$ and $w = d_1v_1 + \dots + d_kv_k$
 $\Rightarrow v + cw = (c_1 + cd_1)v_1 + \dots + (c_k + cd_k)v_k \in V$.

The point is that none of this is new; you already know what subspaces look like and how to describe all of them — yes, all:

Thm: Every subspace of \mathbb{R}^n has the form of #5. #6.

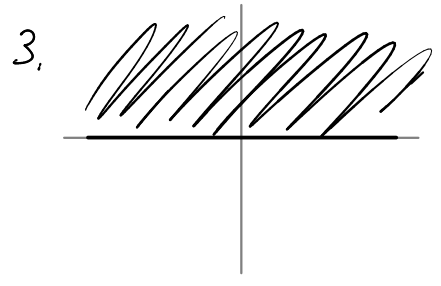
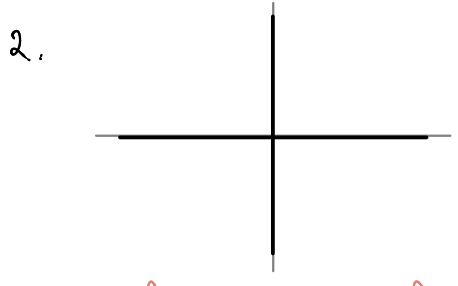
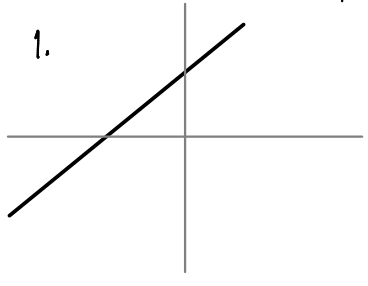
Pf: For #6, keep finding "independent" vectors that increase dimension

until you can't anymore: if $\exists v_1 \neq 0$ in V , then $\text{span}(v_1) \subseteq V$.
 $v_2 \notin \text{span}(v_1)$ $(v_1, v_2) \in V$
 \vdots
 stops at v_k for some $k \leq n$

because $\text{span}(v_1, \dots, v_k) = C(A)$ for $A = \begin{bmatrix} v_1 & \dots & v_k \\ 1 \end{bmatrix}$ is spanned by r vectors, where $r = \text{rank } A \leq n$.

For #5: find constraint equations, given v_1, \dots, v_k . \square

Q. Is this a subspace?



4. $x_0 + \text{span}(v_1, \dots, v_k)$ yes, if $x_0 \in \text{span}$; no if not

5. $U+V = \{u+v \mid u \in U \text{ and } v \in V\}$ for subspaces U and V

Yes: $1. 0+0 \in U+V; \quad 2. \left. \begin{matrix} u+v \in U+V \\ u'+v' \in U+V \\ c \in \mathbb{R} \end{matrix} \right\} \Rightarrow u+v+c(u'+v') = (u+cu') + (v+cv') \in U+V.$

Note: $U+V$ is the smallest subspace containing U and V .

E.g. $x\text{-axis} + y\text{-axis} = xy\text{-plane}.$

Def: For any subset $C \subseteq \mathbb{R}^n_{\text{col}}$ set $C^\perp = \{u \in \mathbb{R}^n_{\text{row}} \mid ux = 0 \forall x \in C\}$

$R \subseteq \mathbb{R}^n_{\text{row}}$ set $R^\perp = \{x \in \mathbb{R}^n_{\text{col}} \mid ux = 0 \forall u \in R\}$

WARNING: The book calls C^\perp what I call $(C^\perp)^\top$; book thinks every vector is a column! C^\perp in book: collection of columns
 will make even more sense when we get to row space $R(A)$ C^\perp for us: collection of rows

Prop: C^\perp is a subspace of $\mathbb{R}^n_{\text{row}}$

R^\perp " " " " $\mathbb{R}^n_{\text{col}}$

Pf: 1. $0x = 0. \checkmark$

2. $vx = 0$ and $wx = 0 \Rightarrow (v+cw)x = vx+cwx = 0+c0 = 0 \forall c \in \mathbb{R}.$

For R^\perp , transpose the argument. \square


Lemma: $V^\top \subseteq \mathbb{R}^n_{\text{row}}$ is a subspace $\Leftrightarrow V \subseteq \mathbb{R}^n$ is a subspace.

Pf: Transpose is linear. \square check! $(A+cB)^\top = A^\top + cB^\top \forall A, B \in \mathbb{R}^{m \times n}, c \in \mathbb{R}$

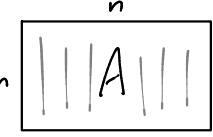
Def: $V \subseteq \mathbb{R}^n_{\text{col}}$ and $W \subseteq \mathbb{R}^n_{\text{row}}$ are orthogonal, written $V \perp W$,

- if $WV = 0$. Equivalently:
- $wv = 0 \forall w \in W \text{ and } v \in V$
 - $W \subseteq V^\perp$
 - $V \subseteq W^\perp$

Also, $V \in \mathbb{R}_{col}^m$ and $W \in \mathbb{R}_{col}^n$: $V \perp W$ if $\cdot W^T V = 0$ $W^T \perp V$
 $\cdot V^T W = 0$ $V^T \perp W$
 $\cdot V \cdot W = 0$



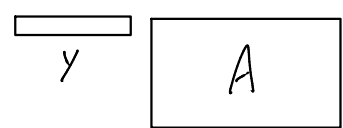
E.g. $C(A)$ = column space of A
 = image of M_A subspace as in #6.



$L(A)$ = left nullspace of $A = \{y \in \mathbb{R}_{row}^m \mid yA = 0\}$
 = kernel of $p_A: \mathbb{R}_{row}^m \rightarrow \mathbb{R}_{row}^n$ y times any column of A is 0.

book omits this
 $= N(A^T)^\perp$

$y \mapsto yA$
 right multiplication by A



WARNING: In the book's notation, $L(A) = N(A^T)^\perp$

Prop: $L(A) = C(A)^\perp$.

CAN SKIP:
 $N(A) = R(A)^\perp$
 PROVED IN NEXT LECTURE

Pf: $y \in L(A) \Rightarrow yA = 0 \Rightarrow y a_j = 0 \forall j = 1, \dots, n$?
 $\Rightarrow c_1 y a_1 + \dots + c_n y a_n = 0 \forall c_1, \dots, c_n \in \mathbb{R}$ so
 $\Rightarrow y(c_1 a_1 + \dots + c_n a_n) = 0$ "
 $\Rightarrow y \in C(A)^\perp$. Hence $L(A) \subseteq C(A)^\perp$.

But $y \in C(A)^\perp \Rightarrow y a_j = 0 \forall j$ by def. (since $a_j \in C(A)$)
 so $yA = 0$; hence $C(A)^\perp \subseteq L(A)$. \square

Similarly:

Def: $R(A)$ = row space of $A = \text{span}(A_1, \dots, A_m) \subseteq \mathbb{R}_{row}^n$

the rows of A

Cor: $N(A) = R(A)^\perp$.

$$\begin{matrix} N(A) & R(A)^\perp \\ \parallel & \parallel \\ L(A^T)^\perp & = (C(A^T)^\perp)^\perp \end{matrix}$$

Pf: Transpose previous Prop. \square

WARNING: The book calls $R(A)^T$ what I call $R(A)$; book thinks every vector is a column!