

Math 412, Lecture 17

Part 3: Analysis on Vector spaces

For the rest of the course, $F = \mathbb{R}$ or \mathbb{C} .

Norms on Vector spaces Metrics

Def: A **metric space** is a pair (X, d_x) where

$d_x: X \times X \rightarrow \mathbb{R}_{\geq 0}$ s.t.

(1) $d(x, y) = d(y, x)$

(2) $d(x, y) = 0$ iff $x = y$

(3) (**triangle inequality**) $d(x, z) \leq d(x, y) + d(y, z)$

(**pseudometric** if weaken (2) to $d(x, x) = 0$)

Notation: The **ball** of radius r is

$$B_X(x, r) = \{y \in X \mid d(x, y) \leq r\}$$

($B_X^\circ(x, r) = \{y \mid d(x, y) < r\}$ is the open ball)

Def: let (X, d_X) (Y, d_Y) be metric spaces.
Say $f: X \rightarrow Y$ is

(1) **continuous** if $\forall x \forall \epsilon > 0 \exists \delta > 0: \forall y$
 $d_X(x, y) < \delta \Rightarrow d_Y(B(x), f(y)) < \epsilon.$

(2) **uniformly cts** if $\forall \epsilon > 0 \exists \delta > 0 \forall x \forall y$ if $d_X(x, y) < \delta$
then $d_Y(f(x), f(y)) < \epsilon$

(3) **Lipschitz cts** if $\exists L > 0 \forall x \forall y: d_Y(f(x), f(y)) \leq$
 $\leq L \cdot d_X(x, y)$

Clearly (3) \Rightarrow (2) \Rightarrow (1).

Set: $\|f\|_{Lip} = \text{smallest } L = \sup_{x \neq y} \frac{d_Y(f(x), f(y))}{d_X(x, y)}$

Lemma: If f, g composable, both of type (1), (2), or (3), then $f \circ g$ of same type.

$$\|f \circ g\|_{Lip} \leq \|f\|_{Lip} \|g\|_{Lip}.$$

Def: Call (X, d) **complete** if every Cauchy sequence in X has a limit.

Vector spaces: Norms

Fix vsp V .

Def: A **norm** on V is a map $\|\cdot\|: V \rightarrow \mathbb{R}_{\geq 0}$ s.t.

$$(1) \|v\| = 0 \text{ iff } v = \underline{0}$$

$$(2) \|\alpha v\| = |\alpha| \cdot \|v\|$$

$$(3) \|u + v\| \leq \|u\| + \|v\|.$$

Lemma: $d(\underline{u}, \underline{v}) = \|\underline{v} - \underline{u}\|$ is a metric on V

Conversely, if d is a metric on V which is translation-invariant ($d(\underline{x}, \underline{y}) = d(\underline{x} + \underline{u}, \underline{y} + \underline{u})$) and 1-homogeneous ($d(\alpha \underline{x}, \alpha \underline{y}) = |\alpha| d(\underline{x}, \underline{y})$) then d comes from a norm.

Clear: If $W \subset V$ is a subspace, $\|\cdot\|_W$ is a norm on W .

Examples: standard norms on \mathbb{F}^n

(1) supremum norm $\|v\|_\infty = \max\{|v_i|\}_{i=1}^n$.

↳ uniform convergence

(2) Manhattan norm $\|v\|_1 = \sum_{i=1}^n |v_i|$

(3) Euclidean norm $\|v\|_2 = \left(\sum_{i=1}^n |v_i|^2\right)^{1/2}$

(comes from inner prod $\langle u, v \rangle = \sum_{i=1}^n \bar{u}_i v_i$)

(*) l^p : $\|v\|_p = \left(\sum_{i=1}^n |v_i|^p\right)^{1/p}$, $1 \leq p < \infty$

(Ex: $\lim_{p \rightarrow \infty} \|v\|_p = \|v\|_\infty$)

Geometrically, can relate norm to its unit ball

$$\{v \mid \|v\| \leq 1\}.$$

Examples: In infinite dimensions, norm comes first, defines the vsp

Def: $l^\infty(X) = \{f: X \rightarrow \mathbb{F} \mid \sup\{|f(x)| : x \in X\} < \infty\}$

equipped with norm $\|f\|_\infty = \sup_{x \in X} |f(x)|$

Better: Define $\|f\|_\infty$ for all f , maybe with value ∞ , check:
 $\{f \in F^X \mid \|f\| < \infty\}$ is a subspace.

Remark: A vsp V with basis B , can be embedded in $\ell^\infty(B)$ via coeff.

Example: $\ell^p(\mathbb{N}) = \{ \underline{a} \in F^{\mathbb{N}} : \sum_{i=1}^{\infty} |a_i|^p < \infty \}$

$$\| \underline{a} \|_p = \left(\sum_{i=1}^{\infty} |a_i|^p \right)^{1/p}.$$

check: $\ell^p(\mathbb{N}) \subseteq \ell^q(\mathbb{N})$ iff $1 \leq p \leq q \leq \infty$.

Example: $L^p(\mathbb{R}) = \{ f : \mathbb{R} \rightarrow F \text{ measurable} \mid$

$$\int_{\mathbb{R}} |f(x)|^p dx < \infty \} / \{ f \mid f=0 \text{ a.e.} \}$$

Theorem: (Elliptic regularity) Let $\Omega \subset \mathbb{R}^2$ be open,
let $f \in L^2(\Omega)$ satisfy $\Delta f = \lambda f$ distributionally:

$$\forall g \in C_c^\infty(\Omega), \int_{\Omega} f \cdot \Delta g = \lambda \int_{\Omega} fg$$

Then there is a smooth function \tilde{f} s.t. $f = \tilde{f}$ a.e.
Then $\Delta \tilde{f} = \lambda \tilde{f}$ pointwise.

□

Def: Say $\underline{V}_n \rightarrow \underline{V}$ if $\|\underline{V}_n - \underline{V}\| \xrightarrow{n \rightarrow \infty} 0$

Fact: on \mathbb{R}^n all norms define same notion of convergence.

Lemma: All norms on \mathbb{R}^n are cts (wrt $\|\cdot\|_1$)

pf:

$$\|\underline{x}\| = \left\| \sum_{i=1}^n x_i \underline{e}_i \right\| \leq \sum_{i=1}^n |x_i| \cdot \|\underline{e}_i\| \leq M \cdot \|\underline{x}\|_1,$$

std basis of \mathbb{R}^n if $M = \max_i \|\underline{e}_i\|$

so $|\|\underline{x}\| - \|\underline{y}\|| \leq \|\underline{x} - \underline{y}\| \leq M \|\underline{x} - \underline{y}\|_1,$

□

Def: Call $\|\cdot\|, \|\cdot\|'$ **equivalent** if $\exists \alpha, \beta > 0$ s.t.

$$\alpha \|x\| \leq \|x\|' \leq \beta \|x\|$$

Ex: This is an equivalence relation. Two norms equiv iff agree on set of sequences s.t.

Theorem: All norms on \mathbb{R}^n are equivalent. $x_n \rightarrow \underline{0}$.

Pf: let $\|\cdot\|$ be any norm. Then $\|x\| \leq M \cdot \|x\|_1$.

let $S = \{x : \|x\|_1 = 1\}$ ("1' sphere")

S is closed and bounded so compact:

$\Rightarrow m = \min \{ \|x\| : x \in S \}$ exists

$m > 0$: $\underline{0} \notin S$ so no $x \in S$ has $\|x\| = 0$.

now if $x \in \mathbb{R}^n$, non zero, $\left\| \frac{x}{\|x\|_1} \right\|_1 = 1$

$\Rightarrow \left\| \frac{x}{\|x\|_1} \right\| \geq m \Rightarrow \|x\| \geq \|x\|_1 \cdot m$.

$$\Rightarrow m \|x\| \leq \|Ax\| \leq M \|x\|,$$

□

Norm on linear maps: operator norm

Def: Let U, V be normed vector spaces,
A linear map $T \in \text{Hom}_F(U, V)$ is **bounded** if
 $\exists M \geq 0$ s.t.

$$\forall u \in U: \|Tu\|_V \leq M \cdot \|u\|_U.$$

The **operator norm** $\|T\| = \|T\|_{U \rightarrow V}$ is the
least such M

Example: Say U is the space of initial conditions
for some evolution. Let V be the space of possible
states at time t . Say $T: U \rightarrow V$ is time evolution

Then T being bounded reflects stability
of PDE.

Example: $\text{Id}: U \rightarrow U$ has $\|\text{Id}\|_{U \rightarrow U} = 1$

Example: $A = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$ acting on \mathbb{R}^2 .

$$\|A \begin{pmatrix} x \\ y \end{pmatrix}\|_1 = \left\| \begin{pmatrix} x+y \\ y \end{pmatrix} \right\|_1 = |x+y| + |y| \leq 2(|x| + |y|) \\ = 2 \left\| \begin{pmatrix} x \\ y \end{pmatrix} \right\|_1,$$

so $\|A\|_1 = 2$. (equality if $x=y$)

$$\|A \begin{pmatrix} x \\ y \end{pmatrix}\|_2^2 = \left\| \begin{pmatrix} x+y \\ y \end{pmatrix} \right\|_2^2 = (x+y)^2 + y^2 = x^2 + 2xy + 2y^2$$

$$\leq \frac{3+\sqrt{5}}{2} (x^2 + y^2)$$

$$\Rightarrow \|A\|_2 = \sqrt{\frac{3+\sqrt{5}}{2}}$$

equality if $x=y=1$

$$\|A \begin{pmatrix} x \\ y \end{pmatrix}\|_\infty = \max \{ |x+y|, |y| \} \leq 2 \max \{ |x|, |y| \} \\ = 2 \left\| \begin{pmatrix} x \\ y \end{pmatrix} \right\|_\infty$$

$$\text{so } \|A\|_\infty = 2$$

(Can also try $\|A\|_{l^1 \rightarrow l^3}$, $\|A\|_{l^\infty \rightarrow l^2}$, ...)

Example: $D_x : C^\infty(\mathbb{R}) \rightarrow C^\infty(\mathbb{R})$

never bounded:

$$D_x (e^{\lambda x}) = \lambda e^{\lambda x}$$

so $\|D_x\| \geq |\lambda|$ for any norm on $C^\infty(\mathbb{R})$

so D_x is unbounded

Lemma: Any map on f.d. space is bounded.

PF: Identify U with \mathbb{R}^n , then $\|\cdot\|_u$ is equivalent to $\|\cdot\|_2$, so $\|\underline{u}\|_2 \leq A \cdot \|\underline{u}\|_u$.

$$\text{Now } \|\tau \underline{u}\|_v = \left\| \tau \left(\sum_{i=1}^n u_i \underline{e}_i \right) \right\|_v$$

$$\leq \sum_{i=1}^n |u_i| \|\tau \underline{e}_i\|_v$$

$$B = \max_i \|\tau \underline{e}_i\|_v$$

$$\leq B \sum_{i=1}^n |u_i| \leq B \cdot \|\underline{u}\|_1 \leq BA \|\underline{u}\|_u$$

$$\Rightarrow \|\tau \underline{u}\|_v \leq (BA) \|\underline{u}\|_u$$

QED

Lemma: Let S, T be composable, bounded.
 Then ST is bounded, $\|ST\| \leq \|S\| \cdot \|T\|$.

Pf: Say $T: U \rightarrow V$, $S: V \rightarrow W$.

$$\begin{aligned} \forall u: \quad \| (ST)u \|_W &= \| S(Tu) \|_W \\ &\leq \|S\|_{V \rightarrow W} \cdot \|Tu\|_V \\ &\leq \|S\|_{V \rightarrow W} \|T\|_{U \rightarrow V} \cdot \|u\|_U \end{aligned}$$

Prop: The operator norm is a norm SA
 on $\text{Hom}_B(U, V) = \{ \text{bounded linear maps} \}$.

Pf: Let $S, T \in \text{Hom}_B(U, V)$. Then

$$\begin{aligned} \| (\alpha S + T)u \|_V &= \| \alpha Su + Tu \|_V \\ &\leq |\alpha| \|Su\|_V + \|Tu\|_V \leq (|\alpha| \cdot \|S\| + \|T\|) \|u\|_U \end{aligned}$$

$\Rightarrow \alpha S + T$ is lld, and $\|\alpha S + T\| \leq |\alpha| \|S\| + \|T\|$

So $\|S + T\| \leq \|S\| + \|T\|$, $\|\alpha S\| \leq |\alpha| \|S\|$

Also get $\|S\| = \|\frac{1}{\alpha} \cdot \alpha S\| \leq \frac{1}{|\alpha|} \|\alpha S\|$

$\Rightarrow \|\alpha S\| \geq |\alpha| \|S\|$ so $\|\alpha S\| = |\alpha| \|S\|$

($0 \in \text{Hom}_b(U, V)$) and if $T \neq 0$, $\exists u: Tu \neq 0$

Then $\|Tu\| \neq 0$ so $\frac{\|Tu\|_V}{\|u\|_U} > 0$

so $\|T\| \geq \frac{\|Tu\|_V}{\|u\|_U} > 0$.

□