

Recall the separating hyperplane theorem from last time.

If $M \cap N = \emptyset$, M, N convex, and $I(M) \neq \emptyset$ or $I(N) \neq \emptyset$. then \exists linear functional f and value c so that

$$f(m) \leq c \quad \forall m \in M, \quad f(n) \geq c \quad \forall n \in N$$

Q: What can we make these into strict inequalities?

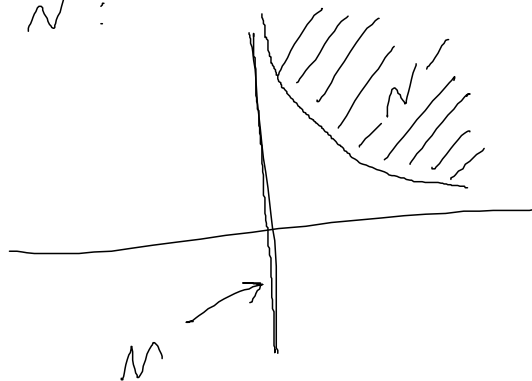
A: if both M and N are open

A: if M is compact and N is closed

ex: take $M = y\text{-axis}$

$$N = \{y \geq \frac{1}{2}x \mid x > 0\}$$

then $M \cap N = \emptyset$, M, N closed, M, N convex, $I(N) \neq \emptyset$. But you cannot strictly separate M from N :



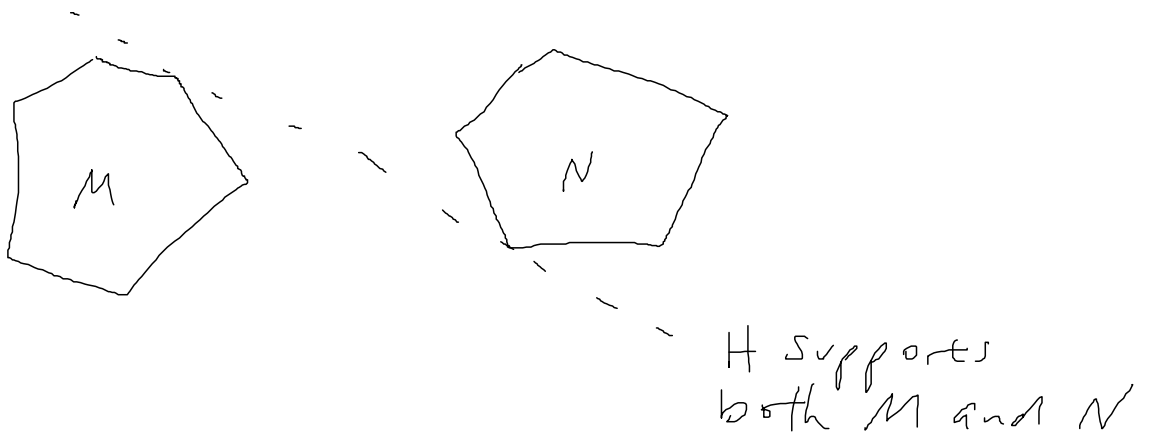
Another concept is a supporting hyperplane i.e.

$H = \{x \mid f(x) = c\}$ is a supporting hyperplane to a convex set M if

$f(x) \leq c \quad \forall x \in M$ and if

$[M] \cap H \neq \emptyset$,

i.e. H touches $[M]$



Q: When is there a supporting hyperplane?

Q: When can you separate two convex sets w/ two supporting hyperplanes?
With one supporting hyperplane?

Note: All of the above referred to open and closed and compact. We need a topology!

Given a vector space L , a norm ρ
 L is a convex functional $\|\cdot\| : L \rightarrow [0, \infty)$
that satisfies

1) $\rho(x) < \infty \forall x, \rho(x+y) \leq \rho(x) + \rho(y)$

2) $\rho(x) = 0 \iff x = \vec{0}$

3) $\rho(\alpha x) = |\alpha| \rho(x) \quad \forall x \in L, \forall \alpha$

We denote ρ by $\|\cdot\|$. Condition 2 is what distinguishes a norm from a convex functional. Also 3, since (3) was only assumed for $\alpha > 0$ for convex functionals.

$(L, \|\cdot\|)$ is a normed vector space. We've already seen a lot of these

$L = C([a, b])$ with $\|f\|_2 = \sqrt{\int_a^b |f(x)|^2 dx}$

$L = C([a, b])$ with $\|f\|_p = \sqrt[p]{\int_a^b |f(x)|^p dx}$

$L =$ polynomials on $[a, b]$ with $\|f\|_\infty = \max_{x \in [a, b]} |f(x)|$

Given $(L, \|\cdot\|)$ you can use $\|\cdot\|$ to define a metric by $\rho(x, y) = \|x - y\|$. Once we have a metric, we can say whether or not (L, ρ) is complete

$L = \text{polynomials}$, $\|f\| = \rho_\infty(f, \vec{0})$
is not complete, since $[\text{polynomials}] = \text{cont. functs.}$

If $(L, \|\cdot\|)$ is complete (w.r.t. the induced metric ρ)
then $(L, \|\cdot\|)$ is a Banach space.

We do have to be a little careful since $(L, \|\cdot\|)$
can play some games on us.

for example

$\text{span}\{x_1, x_2, \dots\}$ might not be closed!

Before, when we didn't have a norm, we
defined a subspace to be the smallest
subspace that contained x_1, \dots

Now, when we talk about a subspace, we mean
the smallest closed subspace containing x_1, \dots

So if $\{x_\alpha\} = \text{polynomials}$ then

this is a subspace of the vector space of
continuous functions. But! If we give the
continuous functions the norm $\|f\| = \sup_{x \in [a, b]} |f(x)|$
then the subspace that contains $\{x_\alpha\}$ is $C[a, b]$

In a metric space, we defined
A bounded if

$$d(A) = \sup_{x,y \in A} \rho(x,y) < \infty, \quad \textcircled{*}$$

In $(L, \|\cdot\|)$ we say A is bounded if

$$| \sup_{x \in A} \|x\| < \infty, \quad \textcircled{*}$$

If we start with $(L, \|\cdot\|)$ and let $\rho(x,y) = \|x-y\|$
then A is bounded in sense $\textcircled{*}$ \Leftrightarrow A is bounded
in sense $\textcircled{*}$. And if $B_n \subset (L, \|\cdot\|)$ is
a nested family of closed balls, then $\bigcap B_n \neq \emptyset$.

But! \exists nested families of bounded closed
sets so that $B_1 \supset B_2 \supset \dots$ and $B_n = \emptyset$.

trouble... closed = contains limit points.
open = trickier.

ex: Let $L = \{ \text{bounded infinite sequences} \}$ with
 $\|x\| = \sup_{i \in \mathbb{N}} |x_i|$

$$B_1 = \{ x \mid x_1 = 0, \|x\|_\infty = 1 \}$$

$$B_2 = \{ x \mid x_1 = x_2 = 0, \|x\|_\infty = 1 \}$$

etc.

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So we have

$$B_1 \supset B_2 \supset \dots \supset B_n$$

where each B_i is closed and bounded (convex even!)

$$\text{but } \bigcap_n B_n = \emptyset.$$

What about open sets?

We had the concept of interior before.

Q: if $x \in I(M)$, can you find $\varepsilon > 0$ so that $S(x, \varepsilon) \subset M$?

A: Intuitively, you should run into trouble in infinite dimensions.

$$x \in I(M) \Rightarrow \text{given } y \in L \exists \varepsilon_y > 0 \text{ so that } x + ty \in M \quad |t| < \varepsilon_y.$$

What if $\varepsilon_y \downarrow 0$?

ex: Let $L = \{ \text{bounded sequences w/ only finitely many nonzero components} \}$
with $\|x\| = \sqrt{\sum |x_i|^2}$

$$\text{let } M = \{x \in L \mid \|x\| \leq 1\}.$$

claim: $\vec{0} \in I(M)$ but you cannot put an ε -sphere around $\vec{0}$ that stays in M .

Why? false

$y \in L$, then let

$N = \#$ of nonzero components of y
and $|y_i| \leq C \quad \forall i$.

then $\|\vec{0} + ty\|$

$$= |t| \sqrt{\sum |y_i|^2}$$

$$< |t| C \sqrt{N} \leq 1 \quad \text{if} \quad |t| < \frac{1}{C\sqrt{N}}$$

\Rightarrow given y , $\vec{0} + ty \in M$ if t small enough.

But clearly we have $\sum y_i \downarrow 0$ as $N \uparrow \infty$ so

there's no single ϵ that will work.

\Rightarrow cannot put an ϵ -ball around $\vec{0}$
that will stay in M !!



Given $(L, \|\cdot\|)$ we can use $\|\cdot\|$ to generate a metric ρ .

Given (X, ρ) we can't necessarily put a norm
on the space. (Need X to be a
vector space, for one
thing!)

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Stronger than the normed vector spaces are
the inner product spaces

i.e. L with \langle, \rangle where

$\langle, \rangle : L \times L \rightarrow \mathbb{R}$ satisfies

1) $\langle x, x \rangle \geq 0 \quad \forall x, \quad \langle x, x \rangle = 0 \Leftrightarrow \vec{x} = \vec{0}$

2) $\langle x, y \rangle = \langle y, x \rangle$

3) $\langle \lambda x, y \rangle = \lambda \langle x, y \rangle$

4) $\langle x, y+z \rangle = \langle x, y \rangle + \langle x, z \rangle$

Claim:

$$|\langle x, y \rangle| \leq \|x\| \|y\| \quad \text{where}$$

$$\|x\| := \sqrt{\langle x, x \rangle}, \quad \|y\| := \sqrt{\langle y, y \rangle}$$

proof:

$$\text{let } p(\lambda) = \langle \lambda x + y, \lambda x + y \rangle$$

$$= \lambda^2 \langle x, x \rangle + 2\lambda \langle x, y \rangle + \langle y, y \rangle$$

$$= \lambda^2 \|x\|^2 + 2\lambda \langle x, y \rangle + \|y\|^2 \geq 0 \quad \forall \lambda$$

$$p(\lambda) \text{ minimized at } \lambda = -\langle x, y \rangle / \|x\|^2$$

$$\text{and } p_{\min} = \|y\|^2 - \langle x, y \rangle^2 / \|x\|^2 \geq 0$$

$$\Rightarrow \langle x, y \rangle^2 \leq \|x\|^2 \|y\|^2 \quad \checkmark$$