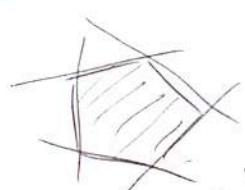


LECTURE 15

Geometric view: half-spaces ~~are~~ in X are sets of the form $\{f \leq a\}$, $a \in \mathbb{R}$, $f \in X^*$.

- Cor. says that ~~the~~ convex symmetric convex set B_X ~~is the intersection of the half~~ can be represented as intersection of ~~half~~ half-spaces (which contain B_X). 
- Well known for ~~polytopes~~ polyhedra in \mathbb{R}^n .

Cor (X^* separates points of X).

For every points $x_1 \neq x_2$ in X , there exists $f \in X^*$ such that $f(x_1) \neq f(x_2)$.

Proof Apply the previous Cor to $x = x_1 - x_2$. QED.

Second Dual

- Functionals $f \in X^*$ act on vectors $x \in X$ as ~~as~~. $x \mapsto f(x)$.
- Vice versa, vectors $x \in X$ act on functionals $f \in X^*$ as ~~as~~. $f \mapsto f(x)$.
- In other words, vectors x may be considered as linear functionals on X^* , i.e. $x \in X^*$ (second dual). The inequality $|f(x)| \leq \|x\| \|f\|$ shows that these linear functionals are continuous, i.e. $x \in X^*$, ~~and~~ ~~continuous~~ their norm as functionals is $\leq \|x\|$.
- Considering the supporting functional f of x , we see that the norm of x as a functional is actually $= \|x\|$.
- Notation: $x^* \in X^*$, $\langle x^*, x \rangle = x^*(x)$.

THM (Second dual).

Let X be a normed space. Then X can be considered as a subspace of X^{**} (via the action $f \mapsto f(x)$). ~~isomorphic~~

Remark • X^{**} may be strictly larger than X ,

i.e. ~~If~~ $C_0^* = l_1$, $l_1^* = l_\infty \Rightarrow C_0^{**} = l_\infty$. ~~is~~ $\neq C_0$
(but $C_0 \subseteq l_\infty$ canonically).

• Spaces for which $X = X^{**}$ are called reflexive.

Examples are: ~~If~~ l_p, L_p for $1 < p < \infty$.

• ~~By~~ Cor. on supporting functionals,
every $f \in X^*$ on ~~a~~ a reflexive X attains its norm.
By a result of James, this characterizes reflexive spaces.

Separation of convex sets.

Remark: Not all norm axioms were used in the proof of H.-B. Thm.

Just these two:

$$(i) \| \lambda x \| = \lambda \| x \| \quad \text{for } \lambda \geq 0 \quad (\text{positive homogeneity})$$

$$(ii) \| x+y \| \leq \| x+y \| \quad (\text{triangle inequality}).$$

Functions $\| \cdot \| : X \rightarrow \mathbb{R}_+$ that satisfy (i), (ii) are called sublinear functionals.

Hahn-Banach Theorem holds for them:

$$\{ f_0(x) \leq \| x \| \text{ for all } x \in X_0 \}$$

then f_0 can be extended to ~~X_0~~ to X while refining

} (check!)

Sublinear functionals offer more flexibility than norms in geometric applications. They arise as Minkowski functionals of arbitrary convex sets K ; ~~Similarly to $\| \cdot \|$ (not necessarily convex as in the case of norms).~~

$$\| x \|_K := \inf \{ t \geq 0 : x/t \in K \}.$$

Proposition (Minkowski functional). ~~(i)~~

Let K be a convex set in a linear vector space X ,

Then $\| \cdot \|_K$ is a sublinear functional on X ; ~~The level set $\{ \| x \|_K = 1 \}$~~ .

$x \in K$.

~~(ii) $\| \cdot \|_K$ is a sublinear functional on X .~~

~~Then $\| \cdot \|_K$~~ (see HW 09/18 #3)

Separation of convex sets.

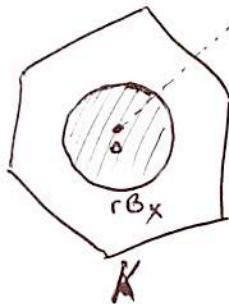
THM (Separating a set from a point)

Let K be an open convex subset of a normed space X ;

let $x_0 \in X \setminus K$. Then $\exists f \in X^*$, $f \neq 0$ such that

$$f(x) \leq f(x_0) \quad \text{for all } x \in K.$$

Proof



w.l.o.g. $0 \in K$ (by translation).

Since A is open, it contains some ball of the form $B_X(0, r) = r \cdot B_X$.
Then $r \cdot B_X$ is the largest ball. Then we can consider the Minkowski functional $\| \cdot \|_K$ of K , which by Prop

By Prop. p. 60, the Minkowski functional $\| \cdot \|_K$ of K is a sublinear functional.

K is open, so it contains a ball $B_X(0, r) = r \cdot B_X$ for some $r > 0$.

The inclusion of the sets implies the inequality for their Minkowski functionals:

$$\|r \cdot B_X\|_K \leq K \quad \text{implies} \Rightarrow \frac{1}{r} \|x\|_K \geq \|x\|_K, x \in K.$$

Similarly to the proof of Cor. on supporting functionals (p. 57), we consider the one-dimensional subspace

$$X_0 = \text{span}(x_0); \quad \text{on } X_0$$

define a linear functional f_0 by

$$f_0(\lambda x_0) = \lambda \|x_0\|_K. \quad (\text{It is bounded because } \|x_0\|_K \text{ is finite})$$

~~f_0 is a bounded linear functional~~

Then f is dominated by $\|\cdot\|_K$ on X_0 :

$$f_0(\lambda x_0) = \|\lambda x_0\|_K \quad \text{for } \lambda \geq 0;$$

$$f_0(-\lambda x_0) = -\lambda f(x_0) = -\lambda \|x_0\|_K \leq 0 \leq \|\lambda x_0\|_K.$$

By Hahn-Banach Thm, f_0 can be extended to a linear functional f on X while retaining the domination:

~~$$f(x) \leq \|x\|_K, \quad x \in X.$$~~

- $f \in X^*$ (bounds):

$$f(x) \leq \|x\|_K \leq \frac{1}{r} \|x\|, \quad x \in X.$$

- $f(x) \leq f(x_0)$ (separation):

~~$f(x) \in \|\cdot\|_K$~~ because $x \notin K$, $x_0 \notin K$

$$f(x) \leq \|x\|_K \leq L \leq \|x_0\| \quad (\text{because } x \in K^c, x_0 \in K)$$

~~$$= f(x_0).$$~~

QED.

THM (Hahn-Banach, geometric form). Let A, B be disjoint convex subsets

of a normed space X , and assume that A is open.

Then $\exists f^* \in X^*, f \neq 0, \exists C \in \mathbb{R}$:

$$f(a) < C \leq f(b) \quad \text{for all } a \in A, b \in B.$$

Proof Consider $K := A - B$; then K is open and convex (check!).

Since $A \cap B = \emptyset$, we have $0 \notin K$. Using Thm p.61, we get

$\exists f \in X^* \in X^*, f \neq 0$, such that

$$f(a - b) \leq 0 \quad \text{for all } a \in A, b \in B.$$

Hence $\sup_{a \in A} f(a) =: C \leq \inf_{b \in B} f(b).$

It follows that $f(a) \leq C \leq f(b)$. The strict inequality in the RHS follows because A is open (check!). QED

Remark: Openness, convexity are needed