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Lecture 5

January 19

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Recall :

- X is a real Hilbert space with inner product $\langle \cdot, \cdot \rangle$ and induced norm.
- $A: X \rightarrow X$ is mono. if $(\forall (x, x^*) \in \text{gr } A) (\forall (y, y^*) \in \text{gr } A)$
 $\langle x - y, x^* - y^* \rangle \geq 0$.
- Recall

Characterization of Projection

Theorem L4-12

$\emptyset \neq C \subseteq X$ is convex closed. Let $x \in X$, let $p \in X$. Then $p = P_C x$ if and only if

- i $p \in C$
- ii $(\forall c \in C) \quad \langle c - p, x - p \rangle \leq 0$.

Example LS-1

$$C = \text{unit ball} \\ = \{ x \in X \mid \|x\| \leq 1 \}.$$

$$P_C x = \begin{cases} x, & \text{if } x \in C; \\ \frac{x}{\|x\|}, & \text{otherwise} \end{cases}$$

$$= \frac{x}{\max\{1, \|x\|\}}.$$

Proof ..

Let $x \in X$,
set

$$p = \frac{x}{\max\{1, \|x\|\}}.$$

i) $p \in C$??

Observe that $p \in C \Leftrightarrow \|p\| \leq 1$.

$$\text{Indeed, } \|p\| = \left\| \frac{x}{\max\{1, \|x\|\}} \right\| = \frac{\|x\|}{\max\{1, \|x\|\}} \leq 1$$

$$\Rightarrow p \in C.$$

ii) Let $c \in C$. Goal:

$$\langle c-p, x-p \rangle \leq 0.$$

$$p = P_C x \Leftrightarrow \begin{cases} \text{i} & p \in C \\ \text{ii} & (\forall c \in C) \quad \langle c-p, x-p \rangle \leq 0 \end{cases}$$

Case 1: $x \in C$

$$\langle c-p, x-p \rangle = \langle c-x, x-x \rangle = 0 \quad \checkmark$$

Case 2: $x \notin C$.

$$\Rightarrow p = \frac{x}{\|x\|} \text{ and } .$$

$$\langle c-p, x-p \rangle \leq 0$$

$$\Leftrightarrow \langle c - \frac{x}{\|x\|}, x - \frac{x}{\|x\|} \rangle \leq 0$$

$= (1 - \frac{1}{\|x\|}) x$

$$\Leftrightarrow \left(1 - \frac{1}{\|x\|}\right) \langle c - \frac{x}{\|x\|}, x \rangle \leq 0$$

$\|x\| > 1 \Rightarrow > 0$

$$\Leftrightarrow \langle c - \frac{x}{\|x\|}, x \rangle \leq 0.$$

$$\Leftrightarrow \langle c, x \rangle \leq \langle \frac{x}{\|x\|}, x \rangle = \frac{1}{\|x\|} \langle x, x \rangle \leq \|x\| \quad ?$$

Indeed, $c \in C \Rightarrow \|c\| \leq 1$

By Cauchy-Schwarz

$$\langle c, x \rangle \leq \|c\| \|x\| \leq 1 \cdot \|x\| = \|x\| \quad \checkmark$$

Definition L5-2:

A subset $\emptyset \neq K \subseteq X$ is a cone if for
 $(\forall \lambda \geq 0) (\forall k \in K) \quad \lambda k \in K$.

Definition L5-3:

Let $K \subseteq X$ be a cone. The polar cone of
 K , denoted by K^\ominus is

$$K^\ominus = \{ u \in X \mid \sup \langle k, u \rangle \leq 0 \}.$$

Proposition L5-4:

Let K be a nonempty closed convex cone of X ,
 let $x \in X$, let $p \in X$. Then:

$$p = P_K x \iff [p \in K, x-p \perp p, x-p \in K^\ominus].$$

Proof:

$$\begin{array}{l} \text{i} \quad p \in K \\ \text{ii} \quad (\forall k \in K) \end{array} \quad p = P_K x \iff \langle x-p, x-p \rangle \leq 0.$$

Suppose $\square \Rightarrow p \in K$ ① ✓ and

$$(\forall y \in K) \quad \langle y-p, x-p \rangle \leq 0$$

$$(\forall y \in K) \quad \langle y-p, x-p \rangle \leq 0$$

$$\begin{aligned} \text{set } y=0 & \quad \langle -p, x-p \rangle \leq 0 \\ y=2p & \quad \langle p, x-p \rangle \leq 0 \end{aligned}$$

$$\begin{aligned} \Rightarrow & \quad \langle p, x-p \rangle = 0 \\ \Rightarrow & \quad p \perp x-p \quad (2) \quad \checkmark \end{aligned}$$

Let $y \in K$

$$\begin{aligned} \langle y, x-p \rangle &= \langle y, x-p \rangle - \langle p, x-p \rangle \\ &= \langle y-p, x-p \rangle \leq 0 \end{aligned}$$

$$\Rightarrow x-p \in K^\ominus \quad (3) \quad \checkmark$$

Conversely, suppose $[p \in K, x-p \perp p, x-p \in K^\ominus]$.

$p \in K \quad \checkmark$

Now, let $y \in K$

$$\begin{aligned} \langle y-p, x-p \rangle &= \underbrace{\langle y, x-p \rangle}_{\leq 0} - \underbrace{\langle p, x-p \rangle}_{\perp = 0} \\ &\leq 0 + 0 = 0 \end{aligned}$$

Hence,

$$P_K x = p \quad \text{by}$$

Characterization of Projection

Theorem L4-12

$C \neq \emptyset \subseteq X$ is convex closed. Let $x \in X$, let $p \in X$. Then $p = P_C x$ if and only if

$$\begin{aligned} & (i) \quad p \in C \\ & (ii) \quad (\forall c \in C) \quad \langle c-p, x-p \rangle \leq 0. \end{aligned}$$

$$\begin{aligned} \text{i} \quad & x = P_K x + P_{K^c} x \\ \text{ii} \quad & P_K x \perp P_{K^c} x \end{aligned}$$

$$\begin{aligned} \text{iii} \quad \|x\|^2 &\stackrel{\text{i}}{=} \|P_K x + P_{K^c} x\|^2 \\ &= \|P_K x\|^2 + \|P_{K^c} x\|^2 \\ &\quad + 2 \langle \underbrace{P_K x}_{=0}, P_{K^c} x \rangle \\ &\stackrel{\text{ii}}{=} \|P_K x\|^2 + \|P_{K^c} x\|^2 \\ &\stackrel{\text{i}}{=} \|x - P_{K^c} x\|^2 + \|x - P_K x\|^2 \\ &= d_{K^c}^2 x + d_K^2 x \end{aligned}$$



Example L5-6

Suppose $X = S^n$, let $K = S_+^n$ be the closed convex cone of symmetric PSD matrices, let $A \in S^n$. Then there exists a diagonal matrix D and a unitary matrix U such that $A = U D U^T$. Moreover:

$$P_K A = U D_+ U^T, \quad (D_+)_{ii} = \max\{0, d_{ii}\}, \\ (D_+)_{ij} = 0, \quad i \neq j$$

Proof:

$$\text{Set } A_+ = U D_+ U^T, \quad D_- = D - D_+$$

$$P = P_K x \Leftrightarrow [p \in K, x - p \perp p, x - p \in K^\ominus]$$

Clearly, $A_+ \in S_+^n$. $\textcircled{1}$ ✓

$$A - A_+ = U D U^T - U D_+ U^T = U D_- U^T \in -K \stackrel{??}{=} K^\ominus$$

HW 2

Finally:

$$\begin{aligned} \langle A_+, A - A_+ \rangle &= \text{tr}(A_+ (A - A_+)) \\ &= \text{tr}(U D_+ U^T (U D_- U^T)) \\ &= \text{tr}(U D_+ U^T U D_- U^T) \\ &= \text{tr}(U D_+ D_- U^T) = \text{tr}(D_+ D_-) \\ &= 0 \end{aligned}$$

$\textcircled{2}$ ✓ □

Example L5-7

Let $a \in X \setminus \{0\}$, let $\beta \in \mathbb{R}$. The hyperplane.

$H = \{ x \in X \mid \langle a, x \rangle = \beta \}$ has

$$P_H x = x - \frac{\langle a, x \rangle - \beta}{\|a\|^2} a$$

Proof:

Let $x \in X$, set $p = x - \frac{\langle a, x \rangle - \beta}{\|a\|^2} a$

$p \in H$??

$$\begin{aligned} \langle a, p \rangle &= \langle a, x \rangle - \frac{\langle a, x \rangle - \beta}{\|a\|^2} \langle a, a \rangle \\ &= \langle a, x \rangle - (\langle a, x \rangle - \beta) = \beta \quad \checkmark \end{aligned}$$

Let $y \in H$. Examine

$$\begin{aligned} \langle y - p, x - p \rangle &= \left\langle y - x + \frac{\langle a, x \rangle - \beta}{\|a\|^2} a, \frac{\langle a, x \rangle - \beta}{\|a\|^2} a \right\rangle \\ &= \frac{\langle a, x \rangle - \beta}{\|a\|^2} \left[\langle y, a \rangle - \langle x, a \rangle + \frac{\langle a, x \rangle - \beta}{\|a\|^2} \|a\|^2 \right] \\ &= \frac{\langle a, x \rangle - \beta}{\|a\|^2} \left[\beta - \beta \right] = 0 \leq 0 \quad \checkmark \end{aligned}$$

Remember A is max mono

$\overline{\text{Dom } A}$, $\overline{\text{ran } A}$ are convex,
of course $\neq \emptyset$, closed.

Let $\emptyset \neq C \subseteq X$ be closed.

$$\Pi_C(x) = \{ p \in C \mid d_C(x) = \|x - p\| \}.$$

$\Pi_C(x) \neq \emptyset$ because C is closed.

$\Pi_C(x)$ is possibly set valued.

Proposition L5-8:

$C \neq \emptyset$ and closed $\Rightarrow \Pi_C$ is monotone.

Proof.

Let $\{(x, x^*), (y, y^*)\} \subseteq \text{gr } \Pi_C$

$$\Leftrightarrow x^* \in \Pi_C(x)$$

$$y^* \in \Pi_C(y)$$

def

\Leftrightarrow

\Rightarrow

$$\|x - x^*\| = d_C(x) \leq \|x - y^*\|$$

$$\|y - y^*\| = d_C(y) \leq \|y - x^*\|$$

That is

$$\begin{array}{l} \|x - x^*\|^2 \leq \|x - y^*\|^2 \\ \|y - y^*\|^2 \leq \|y - x^*\|^2 \end{array}$$

$$\begin{aligned} \|x - x^*\|^2 &\leq \|x - y^*\|^2 \\ \|y - y^*\|^2 &\leq \|y - x^*\|^2 \end{aligned}$$

Adding yields

$$\|x - x^*\|^2 + \|y - y^*\|^2 \leq \|x - y^*\|^2 + \|y - x^*\|^2$$

Expanding yields

$$\cancel{\|x\|^2} + \cancel{\|x^*\|^2} - 2\langle x, x^* \rangle + \underbrace{\|y\|^2} + \underbrace{\|y^*\|^2} - 2\langle y, y^* \rangle$$

$$\leq \underbrace{\|x\|^2} + \underbrace{\|y^*\|^2} - 2\langle x, y^* \rangle + \underbrace{\|y\|^2} + \cancel{\|x^*\|^2} - 2\langle x^*, y \rangle$$

$$\Leftrightarrow \langle x, x^* \rangle + \langle y, y^* \rangle - \langle x, y^* \rangle - \langle x^*, y \rangle \geq 0$$

$$\Leftrightarrow \langle x - y, x^* - y^* \rangle \geq 0$$

Remember

$$\text{Let } \{(x, x^*), (y, y^*)\} \subset \text{gr } T_C$$

The Chebyshev problem:

Suppose $C \neq \emptyset$, closed such that
 $(\forall x \in X) \Pi_C(x)$ is single-valued.

Must C be convex?

Bunt-Motzkin (1930's) L5-10

Yes! when X is finite-dimensional.

Proof:

Recall:

Theorem L2-9:

Let $A: X \rightarrow X$ be:

① monotone

③ single-valued

be:

② $\text{dom } A = X$

④ A continuous

Then A is maximally monotone.

We proved Π_C is mono,

By assumption $\text{dom } \Pi_C = X$
 Π_C is single-valued.

If we can show that Π_C is continuous
 then Π_C is max. mono.

Done $\ddot{\cup}$ because: $\overline{\text{ran } \Pi_C}$

is convex

• But: $\overline{\text{ran } \Pi_C} = \overline{C} = C$ ✓

Suppose $C \neq \emptyset$, closed such that
 $(\forall x \in X) \Pi_C(x)$ is single-valued.

We now show that $\Pi_C(x)$ is continuous.
Write $\Pi_C = P$.

P is mono, single-valued, $\text{dom } P = X$

$$d_C : X \rightarrow \mathbb{R} : x \rightarrow \inf_{c \in C} \|x - c\|$$

is continuous (See HW 2).

Let $x \in X$, $x_n \rightarrow x$.

$$\|x_n - P_C x_n\| = d_C(x_n) \longrightarrow d_C(x)$$

$(x_n)_{n \in \mathbb{N}}$ converges $\Rightarrow (x_n)_{n \in \mathbb{N}}$ is bounded.
 $(d_C x_n)_{n \in \mathbb{N}}$ " $\Rightarrow (d_C x_n)_{n \in \mathbb{N}}$ is

$\Rightarrow (P_C x_n)_{n \in \mathbb{N}}$ must be bounded.

Done if: we can show we have a unique cluster point.

Let \bar{c} be a cluster point of $P_C x_n$
 $\exists k_n \quad P_C x_{k_n} \rightarrow \bar{c} \in C$

C closed, $(P_C x_n)_{n \in \mathbb{N}}$ lies in C .

$$\|x_n - P_C x_n\| = d_C(x_n) \longrightarrow d_C(x)$$

$$P_C x_{k_n} \longrightarrow c \in C$$

$$\begin{aligned} \Rightarrow \|x_{k_n} - P_C x_{k_n}\| &= d_C(x_{k_n}) \longrightarrow d_C(x) \\ \xrightarrow[\text{limit } n \rightarrow \infty]{\text{taking the}} & \|x - \bar{c}\| = d_C(x) \end{aligned}$$

By assumption, C is Chebyshev set, i.e., the projection is unique $\Rightarrow \bar{c} = P_C x$

That is all the cluster points of $P_C x_n$ are equal to $P_C x$.

Therefore $P_C x_n \longrightarrow P_C x$. Hence

P_C is continuous, and mono by

single valued, full domain by assumption
 $\Rightarrow P_C$ is max mono

$$C = \bar{C} = \overline{\text{ran } P_C} \text{ is convex!}$$

□