

$$\sup_{\alpha} \mathbb{E} \left[ \int_0^T L(s, X_s^\alpha, \alpha_s) ds + g(X_T^\alpha) \right] \rightarrow \text{subject to constraints.}$$

Assesement:  $\rightarrow$  Exam 50%

$\rightarrow$  Project (group of 2 persons.)  $\rightarrow$  oral presentation. } 50%

$\rightarrow$  short report.

## I. Static Optimization Problem

$$(P): \inf_{x \in K} f(x).$$

where,  $f: \mathbb{R}^n \rightarrow \mathbb{R}$

$$K := \left\{ x \in \mathbb{R}^n : \begin{array}{l} g_i(x) \leq 0, \quad i=1, \dots, m. \\ h_j(x) = 0, \quad j=1, \dots, l. \end{array} \right\}$$

$$\left( \min_{x \in K} f(x) \right)$$

Remark: If  $m=l=0$ , then  $K = \mathbb{R}^n$ .

## 1. Existence of solution.

- Definition: Let  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  be a function, and  $K \subseteq \mathbb{R}^n$  be ~~is~~ non-empty.

We say the infimum of  $f$  on  $K$  is the real value  $l \in \underline{[-\infty, \infty)}$

s.t. ①  $l \leq f(x)$ ,  $\forall x \in K$ .

②  $\exists (x_n)_{n \geq 1} \subseteq K$  s.t.  $\lim_{n \rightarrow \infty} f(x_n) = l$ .

- Remark: ① We denote  $l = \inf_{x \in K} f(x)$ , which is always well defined.

② We call  $(x_n)_{n \geq 1}$  a minimization sequence.

- Definition: If  $\inf_{x \in K} f(x) > -\infty$ , and there exists  $x^* \in K$  s.t.  $f(x^*) = \inf_{x \in K} f(x)$ .

Then we say  $x^* \in K$  is a solution to (P).

In this case, we write.  $f(x^*) = \min_{x \in K} f(x)$ .

Example:  $K = (0, 1)$ ,  $f(x) = x$ . then  $\inf_{x \in K} f(x) = 0$ , but  $x^*$  does not exist in  $K$ .

Remark: If  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  is continuous, and  $K$  is compact,

Then, a minimizer  $x^*$  exists.

Definition: We say a function  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  is coercive.

$$\text{If } \lim_{\|x\| \rightarrow +\infty} f(x) = +\infty$$

Example: ①  $f(x) = x^T A x + b^T x + c$ . where  $A$  is definitely positive matrix  
 $b \in \mathbb{R}^n$ ,  $c \in \mathbb{R}$ .

→ then  $f$  is coercive.

②  $f(x) = \frac{1}{x}$ .  $K = (0, +\infty)$ . then  $f$  is not coercive.



Proposition: If  $f_i: \mathbb{R} \rightarrow \mathbb{R}$  is bounded from below, and coercive for each  $i=1, \dots, n$

→ then  $f(x) := \sum_{i=1}^n f_i(x_i)$  is coercive.  $x = (x_1, \dots, x_n) \in \mathbb{R}^n$ .

Proof: W.l.o.g., we assume that  $f_i \geq 0$ .

Assume that  $(x_m)_{m \geq 1}$  is sequence s.t.  $\|x_m\| \rightarrow +\infty$ , as  $m \rightarrow +\infty$   
 $\hookrightarrow \sqrt{\sum_{i=1}^n x_{m,i}^2}$

$$x_m = (x_{m,1}, \dots, x_{m,n}) \in \mathbb{R}^n.$$

$\leftarrow$  Then, for some  $i \in \{1, \dots, n\}$ , one has.  $|x_{m,i}| \rightarrow +\infty$  as  $m \rightarrow +\infty$

$\leftarrow$  Then:  $f(x_m) \geq f_i(x_{m,i}) \rightarrow +\infty$  as  $m \rightarrow +\infty$ . #

Theorem: Let  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  be continuous and coercive.

$K \subseteq \mathbb{R}^n$  be a closed nonempty subset.

$\leftarrow$  Then: (P) has a solution.

Proof: let  $M := \inf_{x \in K} f(x) < +\infty$ , then there exists  $L > 0$   
s.t.  $f(x) \geq M$  for all  $x \in K$   
 $\otimes \underline{\|x\| \geq L}$ .

Therefore,  $\inf_{x \in K} f(x) = \inf_{x \in K_L} f(x)$ . with  $K_L = \{x \in K : \|x\| \leq L\}$

Since  $K_L$  is bounded and closed, then  $K_L$  is compact.

Hence, there exists  $x^* \in K_L$  s.t.  $f(x^*) = \inf_{x \in K_L} f(x) = \inf_{x \in K} f(x)$ .  $\neq$

Proposition: Let  $K$  be an open and bounded subset in  $\mathbb{R}^n$ .

and  $f$  is continuous on  $\bar{K} := \{x \in \mathbb{R}^n : \exists (x_n) \in K \text{ s.t. } x_n \rightarrow x\}$   
 $\hookrightarrow$  closure of  $K$ .

and there exists  $x_0 \in K$  s.t.  $f(x_0) \leq f(x)$  for all  $x \in \partial K := \bar{K} \setminus K$ .

Then: (P) has a solution  $x^* \in K$ .

Proof: First,  $\bar{K}$  is a compact subset in  $\mathbb{R}^n$ . then  $\inf_{x \in \bar{K}} f(x)$  has a solution  $x^* \in \bar{K}$ .

Next, if  $x^* \in K$ , then  $f(x^*) = \inf_{x \in \bar{K}} f(x) = \inf_{x \in K} f(x)$ .

If  $x^* \in \partial K$ , then  $f(x_0) \leq f(x^*) = \inf_{x \in K} f(x)$ .  
 $\inf_{x \in K} f(x) \leq$

$$\Rightarrow \inf_{x \in K} f(x) = f(x_0) = f(x^*) = \inf_{x \in K} f(x) \quad \text{since } \inf_{x \in K} f(x) = \inf_{x \in K} f(x)$$

$\Rightarrow x_0 \in K$  is solution to (P). #

2. Necessary condition of the optimal solution  $x^*$ .

2.1. Euler condition.

Thm: Let  $K \subseteq \mathbb{R}^n$  be an open set.  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  is in  $C^1$ .

Assume that (P) has a solution  $x^* \in K$ .

Then,  $\nabla f(x^*) = 0$ .

Proof: Let  $e \in \mathbb{R}^n$ ,  $\varepsilon > 0$  small enough so that  $x_\varepsilon := x^* + \varepsilon e \in K$ .  
(since  $K$  is an open set).

$$\Rightarrow f(x_\varepsilon) \geq f(x^*) \quad \forall \varepsilon > 0.$$

$$\Rightarrow \lim_{\varepsilon \downarrow 0} \frac{f(x_\varepsilon) - f(x^*)}{\varepsilon} = \lim_{\varepsilon \downarrow 0} \frac{f(x^* + \varepsilon e) - f(x^*)}{\varepsilon} = \langle \nabla f(x^*), e \rangle$$

$0 \leq$

$$\Rightarrow \langle \nabla f(x^*), e \rangle \geq 0, \text{ for all } e \in \mathbb{R}^n,$$

$$\Rightarrow \nabla f(x^*) = 0. \quad \#$$

2.2. Kuhn-Tucker Theorem.

$$- K := \left\{ x \in \mathbb{R}^n : \begin{array}{l} g_i(x) \leq 0, \quad i=1, \dots, l. \\ h_j(x) = 0, \quad j=1, \dots, m. \end{array} \right\} \quad \begin{array}{l} I = \{1, \dots, l\} \\ J = \{1, \dots, m\}. \end{array}$$

We will assume that  $g_i, h_j \in C^1$ , so that  $K$  is closed.  
 $f \in C^1$ .

$$(\mathcal{P}) : \inf_{x \in K} f(x) \quad \Leftrightarrow \quad \inf_{x \in \mathbb{R}^n} f(x) \quad \text{subject to} \quad \begin{array}{l} g_i(x) \leq 0, \quad i \in I \\ h_j(x) = 0, \quad j \in J. \end{array}$$

Theorem: Let  $x^*$  be a solution to  $(\mathcal{P})$ .

Then, there exists  $\varphi_0 \geq 0$ ,  $\varphi \in \mathbb{R}_+^l$ ,  $q \in \mathbb{R}^m$  s.t.

$$\textcircled{1} \quad (\varphi_0, \varphi, q) \neq 0.$$

$$\textcircled{2} \quad \sum_{i \in I} \varphi_i g_i(x^*) = 0.$$

$$\textcircled{3} \quad \rho_0 \nabla f(x^*) + \sum_{i \in I} \rho_i \nabla g_i(x^*) + \sum_{j \in J} \rho_j \nabla h_j(x^*) = 0$$

Remark: ① If  $\rho_0 > 0$ , then ③ is equivalent to

$$\nabla f(x^*) + \sum_{i \in I} \rho_i \nabla g_i(x^*) + \sum_{j \in J} \rho_j \nabla h_j(x^*) = 0$$

Example: ( $\rho_0$  could be 0).  $\min_{x^2=0} x$        $f(x) = x$   
 $h(x) = x^2$

$$\nabla f(x^*) = 1, \quad \nabla h(x^*) = 2x^* = 0, \quad \Rightarrow K := \{x : x^2 = 0\} = \{0\}$$

$$x^* = 0.$$

$\Rightarrow$  to ensure that  $\rho_0 \nabla f(x^*) + \rho \nabla h(x^*) = 0$ .

We need to set  $\rho_0 = 0$ .

$$\textcircled{2} \quad \sum_{i \in I} \rho_i g_i(x^*) = 0, \quad \rho_i \geq 0, \quad g_i(x^*) \leq 0, \quad \forall i \in I.$$

$$\leq 0 \quad \Rightarrow \rho_i g_i(x^*) = 0, \quad \forall i \in I.$$



③ Formally, it is a first order necessary condition for  $x^*$ , because:

$$\inf_{\substack{g(x) \leq 0 \\ h(x) = 0}} f(x) \approx \inf_{\substack{g(x) \leq 0 \\ h(x) = 0}} \phi_0 f(x)$$

$$\Leftrightarrow \inf_{x \in \mathbb{R}^n} \sup_{\substack{\phi \geq 0 \\ q \in \mathbb{R}}} \left( \phi_0 f(x) + \phi g(x) + q h(x) \right)$$

penalization.

$$\begin{aligned} \ll \phi_0 f(x) & \quad \text{if } g(x) \leq 0 \text{ and } h(x) = 0 \\ +\infty & \quad \text{if } g(x) > 0 \text{ or } h(x) \neq 0. \end{aligned}$$

$$\langle \approx \rangle \sup_{\substack{\phi \geq 0 \\ q \in \mathbb{R}}} \inf_{x \in \mathbb{R}^n} \left( \phi_0 f(x) + \phi g(x) + q h(x) \right)$$

Euler condition

$$\phi_0 \nabla f(x^*) + \phi \nabla g(x^*) + q \nabla h(x^*) = 0.$$

Proof. ① Let  $\underline{f_N(x)} := \underline{f(x)} + \underline{\|x-x^*\|^2} + \underline{\frac{N}{2} \left( \sum_{i \in I} \max(0, g_i(x))^2 + \sum_{j \in J} h_j(x)^2 \right)}$

It is clear that  $f_N(x) \geq f(x) \quad \forall x \in \mathbb{R}^n$ .

and  $\underline{f_N(x^*)} = \underline{f(x^*)}$ , since  $x^* \in K \Leftrightarrow g_i(x^*) \leq 0, h_j(x^*) = 0$

and  $f_N \in C^1$ , since  $x \mapsto x^2$  is  $C^1$ .

and  $\min_{x \in K} f_N(x)$  has a unique solution  $x^*$ .

② Claim:  $\exists \varepsilon_0 > 0$ , s.t.  $\forall \varepsilon \in (0, \varepsilon_0]$ ,  $\exists N_\varepsilon > 0$ .

$$\text{s.t. } \underline{f_{N_\varepsilon}(x)} > \underline{f_{N_\varepsilon}(x^*)}, \quad \forall \|x-x^*\| = \varepsilon, \quad x \in \mathbb{R}^n.$$

We consider the pb:  $\min_{\|x-x^*\| < \varepsilon} f_{N_\varepsilon}(x)$ , whose solution is not  $x^*$  a priori

which has a solution since  $\{x : \|x-x^*\| < \varepsilon\}$  is open.

and  $\underline{f_{N_\varepsilon}(x^*)} < \underline{f_{N_\varepsilon}(x)}$ ,  $\forall x \in \partial K$ . (claim)

Let  $\underline{x}_\varepsilon^*$  be a solution to  $\min_{\|x-x^*\|<\varepsilon} f_{N_\varepsilon}(x)$ .

By Euler condition,  $\nabla f_{N_\varepsilon}(\underline{x}_\varepsilon^*) = 0$

$$\Leftrightarrow \underline{1} \cdot \nabla f(\underline{x}_\varepsilon^*) + \varepsilon (\underline{x}_\varepsilon^* - x^*) + N_\varepsilon \left( \sum_{i \in I} \max(0, g_i(\underline{x}_\varepsilon^*)) \cdot \nabla g_i(\underline{x}_\varepsilon^*) + \sum_{j \in J} h_j(\underline{x}_\varepsilon^*) \cdot \nabla h_j(\underline{x}_\varepsilon^*) \right) = 0$$

$$\Leftrightarrow \left( \phi_0^\varepsilon \right) \nabla f(\underline{x}_\varepsilon^*) + \left( \varepsilon \phi_0^\varepsilon (\underline{x}_\varepsilon^* - x^*) \right) + \sum \left( \phi_i^\varepsilon \right) \nabla g_i(\underline{x}_\varepsilon^*) + \sum \left( \rho_j^\varepsilon \right) \nabla h_j(\underline{x}_\varepsilon^*) = 0$$

where,  $\rho^\varepsilon := \sqrt{1 + N_\varepsilon^2 \varepsilon^2 \max(0, g_i(\underline{x}_\varepsilon^*))^2 + N_\varepsilon^2 \varepsilon^2 h_j(\underline{x}_\varepsilon^*)^2}$

$$\phi_0^\varepsilon := \frac{1}{\rho^\varepsilon} > 0, \quad \phi_i^\varepsilon := \frac{N_\varepsilon \cdot \max(0, g_i(\underline{x}_\varepsilon^*))}{\rho^\varepsilon} \geq 0, \quad \rho_j^\varepsilon := \frac{N_\varepsilon h_j(\underline{x}_\varepsilon^*)}{\rho^\varepsilon}$$

So that  $\left( \phi_0^\varepsilon \right)^2 + \sum_{i \in I} \left( \phi_i^\varepsilon \right)^2 + \sum_{j \in J} \left( \rho_j^\varepsilon \right)^2 = 1$

Let  $\varepsilon \downarrow 0$ . then:  $\|x_\varepsilon^* - x^*\| \leq \varepsilon \rightarrow 0 \Leftrightarrow x_\varepsilon^* \rightarrow x^*$ .

along a subsequence,  $(\phi_0^\varepsilon, \phi^\varepsilon, q^\varepsilon) \longrightarrow (\phi_0, \phi, q) \in \mathbb{R}_+ \times \mathbb{R}_+^l \times \mathbb{R}^m$

$$\Rightarrow \phi_0 \nabla f(x^*) + \sum_{i \in I} \phi_i \nabla g_i(x^*) + \sum_{j \in J} q_j \nabla h_j(x^*) = 0 \quad (3)$$

and  $\|(\phi_0, \phi, q)\| = 1 \Rightarrow (\phi_0, \phi, q) \neq 0 \quad (1)$

Besides: if  $g_i(x^*) < 0$ ,  $\Rightarrow \phi_i^\varepsilon := N_\varepsilon \frac{\max(0, g_i(x_\varepsilon^*))}{\rho_\varepsilon} = 0$   
for  $\varepsilon > 0$  small enough.

$$\Rightarrow \phi_i = 0.$$

Then:  $\phi_i g_i(x^*) = 0, \forall i \in I \Rightarrow \sum_{i \in I} \phi_i g_i(x^*) = 0 \quad (2)$

(3) Let us finally prove the claim, at the beginning of (2).

Assume that it is not true, then for  $\varepsilon > 0$ , for all  $N > 0$ ,

$$\exists x_N \text{ s.t. } \|x_N - x^*\| = \varepsilon \text{ and } f_N(x_N) \leq f(x^*)$$

$\Rightarrow \{x_N\}_{N \geq 1}$  is a sequence in  $\{x : \|x - x^*\| = \varepsilon\}$ .

$\Rightarrow$  Along a subsequence,  $x_N \rightarrow \bar{x}$

Besides, 
$$\lim_{N \rightarrow +\infty} \left( \frac{N}{2} \left( \sum_{i \in I} \max(0, g_i(x_N))^2 + \sum_{j \in J} (h_j(x_N))^2 \right) \right)$$

$$\lim_{N \rightarrow +\infty} f_N(x_N) \leq f(x^*) < +\infty$$

$$\Rightarrow \lim_{N \rightarrow +\infty} \left( \sum_{i \in I} \max(0, g_i(x_N))^2 + \sum_{j \in J} (h_j(x_N))^2 \right) = 0$$

$$\Rightarrow \sum \max(0, g_i(\bar{x}))^2 + \sum (h_j(\bar{x}))^2 = 0. \Rightarrow g_i(\bar{x}) \leq 0, h_j(\bar{x}) = 0$$

$\Rightarrow \bar{x} \in K.$

$$\Rightarrow \underline{f(\bar{x})} + \underline{\|\bar{x} - x^*\|^2} \leq \lim_{N \rightarrow \infty} f_N(x_N) \leq \underline{f(x^*)}$$

which is impossible since,  $\|\bar{x} - x^*\| = \varepsilon.$

and  $f(x^*) \leq f(\bar{x})$  as  $\bar{x} \in K. \#$