## 对偶原理

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$$(P) p^* = \min f_0(x) \qquad L(x, \lambda, v)$$

$$5t. f(x) \le 0 \leftarrow \lambda \qquad = f_0(x) + \lambda^T f(x) + v^T h(x)$$

$$h(x) = 0 \leftarrow v$$

$$f: R^n \mapsto R^m, h: R^n \mapsto R^p$$

$$f: \mathbb{R}^n \mapsto \mathbb{R}^m, h: \mathbb{R}^n \mapsto \mathbb{R}^p$$

$$\frac{g(\lambda, v) = \inf_{x} L(x, \lambda, v) \leq p^* (\lambda \geq 0)}{\sum_{x} Dual}$$

$$(D): d^* = \sup \{(\lambda, v)\} \Rightarrow d^* \leq p^*$$
s.t.  $\lambda \geq 0$ 

$$S = C - B^T A^{-1} B$$

3) 
$$X > 0 \iff A > 0$$
,  $(I - AA^{\dagger})B = 0$ ,  $S = C - B^{\dagger}A^{\dagger}B \ge 0$ 

$$\begin{array}{c|c}
(I - AA^{+}) & B & V = 0, \forall V \\
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BV & & & \\
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BV & & & \\
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Eg: 
$$p^* = \min_{(x,y)} e^{-x} s.t. \quad x^2/y \le 0, \quad D = \{(x,y) \mid y > 0\}$$

$$p^* = \min_{(x,y)} e^{-x}$$
 s.t.  $x = 0 = 1$ 

$$L(xy\lambda) = e^{-x} + \frac{x^2}{y}\lambda \Rightarrow g(\lambda) = \inf_{x \to y>0} \left\{ e^{-x} + \frac{\lambda x^2}{y} \right\}$$

$$d^*=0$$
 but  $p^*-d^*=1>0$ 

$$f_{o}(x,y) = e^{-x}$$

$$f_{o}(x,y) = x^{2}/y$$

$$A = \{(u, t) \mid f_{o}(x,y) \leq t \mid f_{o}(x,y) \leq u\}$$

$$t = \{(x, y) \in D\}$$

$$u > 0$$

$$x^{2} \leq u$$

$$(o,1)$$

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d^{*}=0 \text{ but } p^{*}-d^{*}=1>0
(SCQ): \exists (\widehat{x}, \widehat{y}) \in D \text{ s.t. } \widehat{x}^{2}/\widehat{y} < 0 \quad (x)
= \{(x,y)|y>0\}
   \min_{x} f_{o}(x), s.t. f(x) \leq 0, h(x) = 0
    G = \{ (f_{i}(x), h(x), f_{0}(x)) \in \mathbb{R}^{m} \times \mathbb{R}^{p} \times \mathbb{R} \mid x \in \mathbb{D} \}
    g(\lambda,\mu) = \inf_{x \in \mathcal{X}} L(x,\lambda,\mu) = f_0(x) + \lambda^T f(x) + \mu^T f_0(x)
                                       = (\lambda, \mu, 1)^{T} (u, v, t), (u, v, t) \in G
  \Rightarrow (\lambda,\mu,i)^{T}(u,v,t) \geq g(\lambda,\mu), \forall (u,v,t) \in G_{i,j}(\lambda \geq 0)
          => supporting hyperplane ( non-verticle).
      P^* = \inf \{ \{ \{ (u, v, t) \in G, u \le 0, v = 0 \} \}
Weak \begin{cases} (\lambda, \mu, 1)^T (u, v, t) \mid (u, v, t) \in G, u \leq 0, v = 0 \end{cases}
duality) = \inf \{(\lambda, \mu, 1)^{T}(\lambda, \nu, t) \mid (\lambda, \nu, t) \in G\} = g(\lambda, \mu)
    A = G + (R_{+}^{m} \times \{o\} \times R_{+}) \Rightarrow Convex \text{ if } f, h, fo convex.}
         = \left\{ (u, v, t) \mid \exists x \in D \quad \text{s.t. } f(x) \leq u, h(x) = v, f_{o}(x) \leq t \right\}
                                                          U>+10 to+10
    P^* = \{ t \mid (0,0,t) \in A \}
   (\lambda,\mu,1)^{T}(u,v,t) = \lambda^{T}f(x) + \mu^{T}h(x) + tf_{0}(x) + \lambda^{T}u+ts : (u,s) \geq 0
     (\lambda,\mu,I)^{T}(u,V,t) \geq g(\lambda,\mu), \forall (u,V,t) \in A.
              Supporting hyperplane (non-verticle)
    Since (0,0,p*) & bdA => I supporting hyperplane.
                                                     (if nonvertide => strong duality)
                                                                    min f_o(x)
    Proof of (SCO), O & Gint (D)
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分区 凸优化课程笔记 的第2页

 $min f_o(x)$ Proof of (SCQ): D X & int (D) St. f(x) ≤0 3 rank (A) = p Ax = b, A & R PXM  $f(\widehat{x}) < 0$ B= { (0,0,5) & R M X R P X R S < p } Comex B={(0,5)|SCA\*} and  $A \cap B = \emptyset$ I (2, 2, M) +0 and &, s,t.  $(u, v, t) \in A \Rightarrow \tilde{\chi}^T u + \tilde{v}^T v + \mu t \geq \alpha \quad (*)$  $(u,v,t) \in B \Rightarrow \widehat{\chi}^T u + \widehat{v}^T v + \mu t \leq \lambda. \quad (**)$ 由 (\*)  $\Rightarrow$   $\stackrel{\sim}{\sim}$   $\stackrel{\sim}{\sim}$  0 (otherwise  $\stackrel{\sim}{\sim}$   $\stackrel{\sim}{\sim}$  ) M ≥0 (otherwise M+ → -∞)  $\Phi(\$\$) \Rightarrow M + \leq Q + \forall t \neq p^* \Rightarrow M + p^* \leq Q + (\$\$\$)$ 将(對對)代入(對) 可知  $\widetilde{\chi}^{T} f(x) + \widetilde{\chi}^{T} (Ax-b) + M f_{o}(x) \geq d \geq M p^{*}$ Case I: M > 0  $P^* \leq \frac{\widetilde{N}}{M} f(x) + \frac{\widetilde{N}^T}{M} (Ax - b) + f_0(x)$  $=\inf_{x \in \mathbb{R}} (x, \widehat{x}/\mu, \widehat{x}/\mu) = g(\widehat{x}/\mu, \widehat{x}/\mu) \leq d^*$ ( ⇒ p\* = d\*.) Case M=0.  $\widetilde{\mathcal{N}}^T f(x) + \widetilde{\mathcal{V}}^T (Ax-b) \ge 0$ ,  $\forall x \in \mathbb{R}$  $\Rightarrow \widehat{\chi}^{T} f(\widehat{\chi}) + \widehat{\chi}^{T} (A\widehat{\chi} - b) \geq 0$ 由 $SCQ: A\tilde{x}-b=0$ ,  $f(\tilde{x})<0 \Rightarrow \widehat{\mathcal{T}}f(\tilde{x})\geq 0$ 国时分20 多分二0

Since  $(\widetilde{\lambda}, \widetilde{\lambda}, M) \neq 0 \Rightarrow \widetilde{\lambda} \neq 0 \Rightarrow \widetilde{\lambda}^{T}(Ax - b) \geq 0$  $\Rightarrow \widetilde{\lambda}^{T}(Ax - Ax) \geq 0, \forall x \in D$   $\Rightarrow \widetilde{\lambda}^{T}A(x - Ax) \geq 0, \forall x \in D$   $\Rightarrow \widetilde{\lambda}^{T}A(x - Ax) \geq 0, \forall x \in D$   $\Rightarrow \widetilde{\lambda}^{T}A(x - x) \geq 0, \forall x \in D$   $\Rightarrow \widetilde{\lambda}^{T}A(x - x) \geq 0, \forall x \in D$   $\Rightarrow \widetilde{\lambda}^{T}A(x - x) \geq 0, \forall x \in D$   $\Rightarrow \widetilde{\lambda}^{T}A(x - b) \geq 0$   $\Rightarrow \widetilde{\lambda}^{T}A(x - b) = 0$   $\Rightarrow \widetilde{\lambda}$