

# CO 466/666: Continuous Optimization

## Problem Set 5

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### Contents

1 Lagrange Multiplier Theorem	1
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### 1 Lagrange Multiplier Theorem

Consider the abstract convex program discussed in class, i.e.

$$\begin{aligned} p^* := \inf & \quad f(x) \\ \text{s.t.} & \quad g(x) \preceq_K 0 \\ & \quad x \in \Omega. \end{aligned} \tag{CP}$$

Here:

- $X, Y$  are Hilbert spaces (wlog e.g.,  $\mathbb{R}^n, \mathbb{R}^m$ , respectively) with inner-products denoted by  $\langle \cdot, \cdot \rangle$ ;
- $\Omega \subseteq X$  is convex;  
 $K \subseteq Y$  is a closed convex cone with non-empty interior;
- $f : X \rightarrow \mathbb{R}$  is a convex real valued function;
- $g : X \rightarrow Y$  is a  $K$ -convex function, i.e.,  
 $g(\lambda x + (1 - \lambda)y) \preceq_K \lambda g(x) + (1 - \lambda)g(y), \forall x, y \in X, \forall 0 \leq \lambda \leq 1$ ;
- The optimal value  $p^*$  is finite valued.

Show the following:

1.  $g(x)$  is a  $K$ -convex function if, and only if,  $x \mapsto \langle \phi, g(x) \rangle$  is a convex function for all  $\phi \in K^+$ , where  $K^+$  is the (nonnegative) polar cone of  $K$ .

2. Show that Strong duality holds for (CP) under the Slater constraint qualification:  $\exists \hat{x} \in \Omega$  such that  $g(\hat{x}) \prec_K 0^1$ , i.e., show that there exists an optimal Lagrange multiplier  $\lambda^* \in K^+$  such that

$$p^* = \inf_{x \in \Omega} \{f(x) + \langle \lambda^*, g(x) \rangle\}. \quad (1)$$

(Use the hyperplane separation argument outlined in class.)

3. Moreover, suppose that we have attainment in (CP), i.e., there exists  $x^* \in \Omega$  such that  $g(x^*) \preceq_K 0$  and  $f(x^*) = p^*$ . Show that complementary slackness holds, i.e., that  $\langle \lambda^*, g(x^*) \rangle = 0$ .
4. Use the above Lagrange multiplier theorem to prove strong duality for (CP), i.e. show that

$$p^* = \min_{x \in \Omega} \max_{\lambda \in K^+} \{f(x) + \langle \lambda, g(x) \rangle\} = d^* := \max_{\lambda \in K^+} \min_{x \in \Omega} \{f(x) + \langle \lambda, g(x) \rangle\}.$$

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<sup>1</sup> equivalently  $g(\hat{x}) \in \text{int}(-K)$