

**C&O 463/663 Convex Optimization and Analysis (Fall 2007)**  
**Assignment 2**

Due 1PM (before class), on Tuesday, Oct. 23, 2007

## 1 Convex, Gauge and Support Functions

1. Suppose that the function  $f : E \rightarrow (-\infty, +\infty]$  is essentially strictly convex. Prove that all distinct points  $x, y$  in  $E$  satisfy  $\partial f(x) \cap \partial f(y) = \emptyset$ . Deduce that  $f$  has at most one minimizer.
2. For real  $p > 1$ , define  $q$  by  $\frac{1}{p} + \frac{1}{q} = 1$  and for  $x \in \mathfrak{R}^n$  define  $\|x\|_p = (\sum_1^n |x_i|^p)^{1/p}$ .
  - (a) Prove  $\frac{1}{p}\|x\|_p^p$  is a convex function and deduce the set  $B_p = \{x : \|x\|_p \leq 1\}$  is convex.
  - (b) Prove the gauge function  $\gamma_{B_p}(\cdot)$  is exactly  $\|\cdot\|_p$ , and deduce  $\|\cdot\|_p$  is convex.
  - (c) Use the Fenchel-Young inequality to prove that any vectors  $x, \phi$  in  $\mathfrak{R}^n$  satisfy the inequality

$$\frac{1}{p}\|x\|_p^p + \frac{1}{q}\|\phi\|_q^q \geq \langle \phi, x \rangle.$$

3. Suppose that  $C, D$  are closed convex sets in  $E$ . Show that  $C = D$  if and only if their support functions are equal,  $\sigma_C = \sigma_D$ .
4. Show that:  $\sigma_B = \sigma_{\text{conv } B}$ ;  $\sigma_{A+B} = \sigma_A + \sigma_B$ ;  $\sigma_{A \cup B} = \max\{\sigma_A, \sigma_B\}$ .

## 2 Recession and Normal Cones

1. Suppose that  $S \subset E$  is nonempty, closed, and convex. Define the *recession cone of S* to be

$$0^+S := \{d \in E : x + td \in S, \forall x \in S, \forall t \geq 0\}$$

- (a) Prove that  $d \in 0^+S$  iff  $\exists x \in S$  s.t.  $x + td \in S, \forall t \geq 0$ .
  - (b) Prove that  $0^+\text{relint}(S) = 0^+\text{cl}(S)$ , i.e. the recession cones of the relative interior and the closure are equal.
2. (a) Prove that the normal cone is a closed convex cone.  
(b) Compute the normal cone  $N_C(\bar{x})$  for points  $\bar{x}$  in the sets:  $C = B$  the unit ball;  $C$  a closed halfspace.