

### Numerical optimization, Problem sheet 3

1. Midpoint convexity. A set  $C$  is midpoint convex if whenever two points  $a, b$  are in  $C$ , the midpoint  $(a + b)/2$  is in  $C$ . Check that convex set is midpoint convex and that set of rational numbers  $\mathbb{Q}$  is midpoint convex but not convex. However, prove that if  $C$  is closed and midpoint convex, then  $C$  is convex.
2. Which of the following sets are convex?
- (a) A slab, i.e., a set of the form  $\{x \in \mathbb{R}^n : \alpha \leq \langle a, x \rangle \leq \beta\}$ .
  - (b) A wedge, i.e.,  $\{x \in \mathbb{R}^n : \langle a_1, x \rangle \leq b_1, \langle a_2, x \rangle \leq b_2\}$ .
  - (c) The set of points closer to a given point than a given set, i.e.,

$$\{x : \|x - x_0\|_2 \leq \|x - y\|_2 \text{ for all } y \in S\}$$

where  $S \subset \mathbb{R}^n$ .

- (d) The set of points closer to one set than another, i.e.,  $\{x : \text{dist}(x, S) \leq \text{dist}(x, T)\}$ , where  $S, T \subset \mathbb{R}^n$ , and  $\text{dist}(x, S) = \inf_{z \in S} \|xz\|_2$ .
- (e) The set  $\{x : x + S_1 \subset S_2\}$ , where  $S_1, S_2 \subset \mathbb{R}^n$  with  $S_2$  convex.
- (f) The set of points whose distance to  $a$  does not exceed a fixed fraction  $\theta$  of the distance to  $b$ , i.e., the set  $\{x : \|x - a\|_2 \leq \theta \|x - b\|_2\}$ . You can assume  $a \neq b$  and  $0 \leq \theta \leq 1$ .

3. Support function. The support function of a set  $C \subset \mathbb{R}^n$  is defined as

$$S_C(y) = \sup_{x \in C} \langle x, y \rangle.$$

(We allow  $S_C(y)$  to take on the value  $\infty$ ). Suppose that  $C$  and  $D$  are closed convex sets in  $\mathbb{R}^n$ . Show that  $C = D$  if and only if their support functions are equal. Hint: use separating hyperplane.

4. For each of the following functions determine whether it is convex or concave.

- (a)  $f(x) = x \exp(x)$  on  $(0, \infty)$
- (b)  $f(x_1, x_2) = x_1 x_2$  on  $(0, \infty)^2$ .
- (c)  $f(x_1, x_2) = 1/(x_1 x_2)$  on  $(0, \infty)^2$ .
- (d)  $f(x_1, x_2) = x_1/x_2$  on  $(0, \infty)^2$ .
- (e)  $f(x_1, x_2) = x_1^2/x_2$  on  $\mathbb{R} \times (0, \infty)$ .
- (f)  $f(x_1, x_2) = x_1^\alpha x_2^{1-\alpha}$ , where  $0 \leq \alpha \leq 1$ , on  $(0, \infty)^2$ .

5. Convex hull or envelope of a function. The convex hull or convex envelope of a function  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  is defined as

$$g(x) = \inf\{t : (x, t) \in \text{conv}(\text{epi}(f))\}.$$

Geometrically, the epigraph of  $g$  is the convex hull of the epigraph of  $f$ . Show that  $g$  is the largest convex underestimator of  $f$ . In other words, show that if  $h$  is convex and satisfies  $h(x) \leq f(x)$  for all  $x$ , then  $h(x) \leq g(x)$  for all  $x$ .