## Due Date: Feb 14, 2018, Beginning of the class

## How to submit: Hard copy in the Class

- 1.1 Let  $C \in \mathbf{R}^n$  be a convex set, with  $x_1, \dots, x_k \in C$ , and let  $\theta_1, \dots, \theta_k \in \mathbf{R}$  satisfy  $\theta_i \geq 0$ ,  $\theta_1 + \dots + \theta_k = 1$ . Show that  $\theta_1 x_1 + \dots + \theta_k x_k \in C$ . (The definition of convexity is that this holds for k = 2; you must show it for arbitrary k.) Hint. Use induction on k.
- 1.2 Show that the convex hull of a set S is the intersection of all convex sets that contain S.
- 1.3 Let a and b be distinct points in  $\mathbf{R}^n$ . Show that the set of all points that are closer (in Euclidean norm) to a than b, i.e.,  $\{x \mid \|x a\|_2 \le \|x b\|_2\}$ , is a halfspace. Describe it explicitly as an inequality of the form  $c^T x \le d$ . Draw a picture.

## 1.4 Show that:

- (a) The intersection  $\bigcap_{i \in I} C_i$  of a collection  $\{C_i | i \in I\}$  of cones is a cone.
- (b) The Cartesian product  $\mathcal{C}_1 \times \mathcal{C}_2$  of two cones  $\mathcal{C}_1$  and  $\mathcal{C}_2$  is a cone.
- (c) The vector sum  $C_1 + C_2$  of two cones  $C_1$  and  $C_2$  is a cone.
- (d) The image and the inverse image of a cone under a linear transformation is a cone.
- (e) A subset C is a convex cone if and only if it is closed under addition and positive scalar multiplication, i.e.,  $C + C \subset C$ , and  $\gamma C \subset C$  for all  $\gamma > 0$ .
- 1.5 Is the set  $\{a \in \mathbf{R}^n | p(0) = 1, |p(t)| \le 1 \text{ for } \alpha \le t \le \beta\}$ , where  $p(t) = a_1 + a_2t + \dots + a_kt^{k-1}$ , convex? Give the details of your conclusion.
- 1.6 Let  $C \in \mathbf{R}^n$  be the solution set of a quadratic inequality,  $C = \{x \in \mathbf{R}^n | x^T A x + b^T x + c \le 0\}$  with  $A \in \mathbf{S}^n$ , be  $\mathbf{R}^n$ , and  $c \in \mathbf{R}$ . Show that C is convex if  $A \ge 0$ .
- 1.7 Which of the following sets are convex?
- (a) A slab, i.e., a set of the form  $\{x \in \mathbf{R}^n | \alpha \le \alpha^T x \le \beta\}$
- (b) A rectangle, i.e., a set of the form  $\{x \in \mathbf{R}^n | \alpha_i \le x_i \le \beta_i, i = 1, ..., n\}$
- (c) A wedge, i.e.,  $\{x \in \mathbf{R}^n | a_1^T x \le b_1, a_2^T x \le b_2\}$
- (d) The set of points closer to a given point than a given set, i.e.,  $\{x | \|x x_0\|_2 \le \|x y\|_2$ , for all  $y \in S\}$ , where  $S \subseteq \mathbf{R}^n$ .
- (e) The set of points closer to one set than another, i.e.,  $\{x | \mathbf{dist}(x, S) \leq \mathbf{dist}(x, T)\}$ , where  $S, T \subseteq \mathbf{R}^n$  and  $\mathbf{dist}(x, S) = \inf\{\|x z\|_2 | z \in S\}$ .
- (f) The set  $\{x \mid x + S_2 \subseteq S_1\}$ , where  $S_1, S_2 \subseteq \mathbb{R}^n$ , with  $S_1$  convex
- (g) The set of points whose distance to a does not exceed a fixed fraction  $\theta$  of the distance to b, i.e., the set  $\{x | \|x a\|_2 \le \theta \|x b\|_2\}$ . You can assume  $a \ne b$  and  $0 \le \theta \le 1$ .