

# Homework 1

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## Q1: Affine images and preimages of convex sets.

Grade:

Let  $A \in \mathbb{R}^{m \times n}$ ,  $b \in \mathbb{R}^m$ ,  $C \subset \mathbb{R}^n$ ,  $D \subset \mathbb{R}^m$  be convex sets. Show that following sets are convex.

- (a) The image of  $C$  under the affine map  $x \mapsto Ax + b$ . That is

$$\{Ax + b \mid x \in C\} \subset \mathbb{R}^m.$$

- (b) The preimage of  $D$  under the affine map  $x \mapsto Ax + b$ . That is

$$\{x \mid Ax + b \in D\} \subset \mathbb{R}^n.$$

### Solution

- (a) *Proof.* Let  $x_1, x_2 \in C$  and  $\lambda \in [0, 1]$ , then we have

$$\begin{aligned} \lambda(Ax_1 + b) + (1 - \lambda)(Ax_2 + b) &= A(\lambda x_1 + (1 - \lambda)x_2) + b \\ &\in A(C) + b. \end{aligned}$$

Thus, the image of  $C$ ,  $A(C) + b$  is convex. □

- (b) *Proof.* Let  $y_1, y_2 \in A^{-1}(D - b)$  so that  $Ay_1 + b \in D$ ,  $Ay_2 + b \in D$  and  $\lambda \in [0, 1]$ , then we have

$$\begin{aligned} A(\lambda y_1 + (1 - \lambda)y_2) + b &= \lambda(Ay_1 + b) + (1 - \lambda)(Ay_2 + b) \\ &\in \lambda D + (1 - \lambda)D \\ &= D. \end{aligned}$$

Thus, The preimages of  $D$ ,  $A^{-1}(D - b)$  is convex. □

## Q2: Affine functions.

Grade:

Suppose that  $f : \mathbb{R}^n \rightarrow \mathbb{R} \setminus \{-\infty, \infty\}$  always obey the convex function relation at equality, that is,

$$f(\lambda x + (1 - \lambda)y) \leq \lambda f(x) + (1 - \lambda)f(y), \quad \forall x, y \in \mathbb{R}^n, \lambda \in [0, 1]. \quad (1)$$

Show that

- (a) If eq. (1) holds as stated for all  $\lambda \in [0, 1]$ , it in fact holds for all  $\lambda \in \mathbb{R}$ .
- (b) Any  $f$  for which eq. (1) holds must be of the form  $f(x) = \langle a, x \rangle + b$  for  $a \in \mathbb{R}^n$ ,  $b \in \mathbb{R}$  (that is,  $f$  is an inner product with some fixed vector plus a constant).
- (c) Any function of this form has the property eq. (1).

*Hint:* given  $f$  satisfying the condition above, show that  $g : x \mapsto f(x) - f(0)$  is linear. You may then use (without proof, although the proof is very easy) that a linear function  $g : \mathbb{R}^n \rightarrow \mathbb{R}$  must be of the form  $x \mapsto \langle a, x \rangle$  for some  $a \in \mathbb{R}^n$ .

### Solution

(a) *Proof.* To extend eq. (1) to  $\lambda \in \mathbb{R}^n$ , we need to show that eq. (1) holds for  $\lambda \in (-\infty, 0) \cup (1, \infty)$ .

First, let  $x, y \in \mathbb{R}^n$ , and for  $\lambda \in (-1, 0)$  let  $\alpha = -\lambda \in (0, 1)$ . Then we have

$$f(\alpha x + (1 - \alpha)y) \leq \alpha f(x) + (1 - \alpha)f(y),$$

which shows convexity for  $\lambda \in (-1, 0)$ . Similarly, for  $\lambda \in (1, \infty)$  let  $\alpha = \frac{1}{\lambda} \in (0, 1)$ , and for  $\lambda \in (-\infty, -1)$  let  $\alpha = -\frac{1}{\lambda} \in (0, 1)$ , we can prove item (a) holds for  $\lambda \in (1, \infty)$  and  $\lambda \in (-\infty, -1)$  respectively.  $\square$

(b) *Proof.* Let's define  $g : \mathbb{R}^n \rightarrow \mathbb{R}$  as  $g(x) = f(x) - f(0)$ , then we have  $g(0) = f(0) - f(0) = 0$ . For any  $x, y \in \mathbb{R}^n$  and  $\lambda \in \mathbb{R}$  (as proved above), we have

$$\begin{aligned} g(\lambda x + (1 - \lambda)y) &= f(\lambda x + (1 - \lambda)y) - f(0) \\ &\leq \lambda f(x) + (1 - \lambda)f(y) - f(0) \\ &= \lambda(f(x) - f(0)) + (1 - \lambda)(f(y) - f(0)) \\ &= \lambda g(x) + (1 - \lambda)g(y). \end{aligned}$$

This shows  $g$  is a linear function. From the hint, we can represent  $g$  as  $g(x) = \langle a, x \rangle$  for some  $a \in \mathbb{R}^n$ . Thus,  $f(x) = \langle a, x \rangle + b$  where  $b = f(0)$ .  $\square$

(c) *Proof.* If  $f(x) = \langle a, x \rangle + b$ , then for any  $x, y \in \mathbb{R}^n$  and  $\lambda \in [0, 1]$

$$\begin{aligned} f(\lambda x + (1 - \lambda)y) &= \langle a, \lambda x + (1 - \lambda)y \rangle + b \\ &= \lambda \langle a, x \rangle + (1 - \lambda) \langle a, y \rangle + b \\ &= \lambda(\langle a, x \rangle + b) + (1 - \lambda)(\langle a, y \rangle + b) \\ &\leq \lambda f(x) + (1 - \lambda)f(y). \end{aligned}$$

$\square$

### Q3: Convex hulls.

### Grade:

Show that for any set  $X \subseteq \mathbb{R}^n$ , the convex hull  $\text{conv}(X)$  of  $X$  (the intersection of all convex sets containing  $X$ ) is equal to the set of all convex combinations of points in  $X$ .

*Hint:* Define  $Y$  to be the set of all convex combinations of points from  $X$ , that is,

$$Y = \left\{ \sum_{i=1}^m \lambda_i x_i \mid m \geq 1, \lambda_i \geq 0, \sum_{i=1}^m \lambda_i = 1 \right\},$$

and then prove that both  $Y \subseteq \text{conv}(X)$  (which may be accomplished by showing that it is convex and contains  $X$ ), and  $\text{conv}(X) \subseteq Y$  (which may be accomplished by showing that every convex set containing  $X$  also contains  $Y$ ).

### Solution

1.  $Y \subseteq \text{conv}(X)$

*Proof.* Let  $y_1, y_2 \in Y$ . By definition of  $Y$ ,

$$\begin{aligned} y_1 &= \sum_{i=1}^{m_1} \alpha_i x_i, & \sum_{i=1}^{m_1} \alpha_i &= 1, \\ y_2 &= \sum_{j=1}^{m_2} \beta_j x_j, & \sum_{j=1}^{m_2} \beta_j &= 1. \end{aligned}$$

For any  $\lambda \in [0, 1]$ , consider the point  $y = \lambda y_1 + (1 - \lambda)y_2$ . Then

$$\begin{aligned} y &= \lambda \sum_{i=1}^{m_1} \alpha_i x_i + (1 - \lambda) \sum_{j=1}^{m_2} \beta_j x_j \\ &= \sum_{i=1}^{m_1} (\lambda \alpha_i) x_i + \sum_{j=1}^{m_2} ((1 - \lambda) \beta_j) x_j \end{aligned}$$

where

$$\begin{aligned} &\sum_{i=1}^{m_1} \lambda \alpha_i + \sum_{j=1}^{m_2} (1 - \lambda) \beta_j \\ &= \lambda \sum_{i=1}^{m_1} \alpha_i + (1 - \lambda) \sum_{j=1}^{m_2} \beta_j \\ &= \lambda + (1 - \lambda) \\ &= 1. \end{aligned}$$

Clearly,  $y \in Y$ , which shows  $Y$  is convex. Also, every point  $x_i \in X$  is in  $Y$  with  $\lambda_i = 1$ , which shows  $X \subseteq Y$ . Since  $Y$  is convex and contains  $X$ , then it must contain  $\text{conv}(X)$  as  $\text{conv}(X)$  is the intersection of all convex sets containing  $X$ .  $\square$

## 2. $\text{conv}(X) \subseteq Y$

*Proof.* Let  $Z$  be any convex set containing  $X$ . We want to show that  $Z$  also contains  $Y$ . Take any  $y \in Y$ , since  $Z$  is convex and contains  $X$ ,  $Z$  must contain  $y$ , the convex combination of points in  $X$ . Thus,  $Z$  contains  $Y$ . Since arbitrary  $Z$  contains  $Y$ ,  $\text{conv}(X)$  must contain  $Y$  as  $\text{conv}(X)$  is the intersection of all convex sets containing  $X$ . Therefore,  $\text{conv}(X) \subseteq Y$ .  $\square$

## Q4: Affine sets and hulls.

Grade:

The scalars  $\lambda_i$  in this problem may take negative values.

- (a) The textbook defines a set  $X \subseteq \mathbb{R}^n$  as being affine if it is of the form  $S + x = \{s + x \mid s \in S\}$  for some  $x \in \mathbb{R}^n$  and linear subspace  $S$  of  $\mathbb{R}^n$ . Show that  $X$  is affine according to this definition if and only if  $X$  is

$$\left. \begin{array}{l} x_1, x_2, \dots, x_m \in X \\ \lambda_1, \lambda_2, \dots, \lambda_m \in \mathbb{R} \\ \sum_{i=1}^m \lambda_i = 1 \end{array} \right\} \Rightarrow \sum_{i=1}^m \lambda_i x_i \in X$$

*Hint:* For the “if”, take any  $x \in X$  and show that the set  $S = X - x = \{x' - x \mid x' \in X\}$  is a linear subspace of  $\mathbb{R}^n$ .

- (b) In the text, the *affine hull*  $\text{aff}(Y)$  of a set  $Y$  is defined to be the intersection of all affine sets containing

Y. Show that

$$\text{aff}(Y) = \left\{ \sum_{i=1}^m \lambda_i y_i \mid m \geq 1, \lambda_i \in \mathbb{R}, \sum_{i=1}^m \lambda_i = 1, y_i \in Y \right\},$$

that is, the affine hull of  $Y$  is the set of all affine combinations of points in  $Y$ .

### Solution

(a) • “if”

*Proof.* Assume that the condition holds for  $X$ . We want to show that  $X$  is affine by showing  $S$  is a linear subspace of  $\mathbb{R}^n$ . Take any  $x \in X$  and let  $S = X - x$ , we have

(1)  $0 \in S$  because  $x' - x = 0$  for  $x' = x$  and  $x' \in X$ .

(2) For  $s_1, s_2, \dots, s_m \in S$ ,  $\sum_{i=1}^m \lambda_i s_i = \sum_{i=1}^m \lambda_i (x_i - x) = (\sum_{i=1}^m \lambda_i x_i) - x \in S$  for  $\sum_{i=1}^m \lambda_i = 1$ .

(3) For any  $s \in S$  and any scalar  $\lambda$ ,  $\lambda s = \lambda(x' - x) = [\lambda(x') + (1 - \lambda)x] - x \in S$ .

Thus,  $S$  is a linear subspace of  $\mathbb{R}^n$ , which shows  $X$  is affine.  $\square$

• “only if”

*Proof.* Suppose  $X$  is affine as  $X = S + x$  for some  $x \in \mathbb{R}^n$  and some linear subspace  $S$  of  $\mathbb{R}^n$ , we want to show that when the conditions hold,  $\sum_{i=1}^m \lambda_i x_i \in X$ . Since  $X = S + x$ , for any point  $x_i = s_i + x$ , we have

$$\sum_{i=1}^m \lambda_i x_i = \sum_{i=1}^m \lambda_i (s_i + x) = \sum_{i=1}^m \lambda_i s_i + \sum_{i=1}^m \lambda_i x = \sum_{i=1}^m \lambda_i s_i + x \in X.$$

$\square$

(b) The proof of this is similar to Q3 without condition  $\sum_{i=1}^m \lambda_i = 1$ .

### Q5: Arithmetic-Geometric Mean Inequality.

Grade:

Show that if  $\lambda_1, \lambda_2, \dots, \lambda_m$  are positive scalars with  $\sum_{i=1}^m \lambda_i = 1$ , then for every set of positive scalars  $x_1, x_2, \dots, x_m$ , we have

$$\prod_{i=1}^m x_i^{\lambda_i} \leq \sum_{i=1}^m \lambda_i x_i,$$

with equality if and only if  $x_1 = x_2 = \dots = x_m$ .

*Hint:* Show that  $-\ln x$  is a strictly convex function on  $(0, \infty)$ .

### Solution

Consider the function  $f(x) = -\ln x$ , then  $f''(x) = \frac{1}{x^2} > 0$  for  $x > 0$ . Thus,  $f(x)$  is strictly convex on  $(0, \infty)$ . By Jensen's inequality, we have

$$\begin{aligned} -\ln\left(\sum_{i=1}^m \lambda_i x_i\right) &= f\left(\sum_{i=1}^m \lambda_i x_i\right) \\ &\leq \sum_{i=1}^m \lambda_i f(x_i) \\ &= \sum_{i=1}^m \lambda_i (-\ln x_i) \end{aligned}$$

$$\begin{aligned} &= -\ln\left(\prod_{i=1}^m x_i^{\lambda_i}\right) \\ \Rightarrow \exp\left\{-\ln\left(\sum_{i=1}^m \lambda_i x_i\right)\right\} &\leq \exp\left\{-\ln\left(\prod_{i=1}^m x_i^{\lambda_i}\right)\right\} \\ \Rightarrow \sum_{i=1}^m \lambda_i x_i &\geq \prod_{i=1}^m x_i^{\lambda_i}. \end{aligned}$$

Since  $f(x)$  is strictly convex, the equality holds if and only if  $x_1 = x_2 = \cdots = x_m$ .