

Coursework (3) for *Introductory Lectures on Optimization*

Hanxuan Li
3220106039

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Exercise 1. Proof that if $f_i(\mathbf{x})$, $i \in I$, are convex, then

$$g(\mathbf{x}) = \max_{i \in I} f_i(\mathbf{x})$$

is also convex.

Proof of Exercise 1: Because $\forall i \in I$, $f_i(\mathbf{x})$ are convex, we have:

$$\forall i \in I, f_i(\alpha \mathbf{x} + (1 - \alpha) \mathbf{y}) \leq \alpha f_i(\mathbf{x}) + (1 - \alpha) f_i(\mathbf{y}) \leq \alpha g(\mathbf{x}) + (1 - \alpha) g(\mathbf{y}) \quad (1)$$

where $\alpha \in [0, 1]$. So we have:

$$g(\alpha \mathbf{x} + (1 - \alpha) \mathbf{y}) = \max_{i \in I} \{f_i(\alpha \mathbf{x} + (1 - \alpha) \mathbf{y})\} \leq \max \{\alpha g(\mathbf{x}) + (1 - \alpha) g(\mathbf{y})\} = \alpha g(\mathbf{x}) + (1 - \alpha) g(\mathbf{y}) \quad (2)$$

where $\alpha \in [0, 1]$. So $g(\mathbf{x})$ is also convex. \square

Exercise 2. Proof that

1. if f is a convex function on \mathbb{R}^n and $F(\cdot)$ is a convex and non-decreasing function on \mathbb{R} , then $g(\mathbf{x}) = F(f(\mathbf{x}))$ is convex.
2. If $f_i, i = 1, \dots, m$ are convex functions on \mathbb{R}^n and $F(\mathbf{y}_1, \dots, \mathbf{y}_m)$ is convex and non-decreasing (component-wise) in each argument, then

$$g(\mathbf{x}) = F(f_1(\mathbf{x}), \dots, f_m(\mathbf{x}))$$

is convex.

Proof of Exercise 2:

1. Because f is convex, so $\forall \alpha \in [0, 1]$, we have

$$f(\alpha \mathbf{x} + (1 - \alpha) \mathbf{y}) \leq \alpha f(\mathbf{x}) + (1 - \alpha) f(\mathbf{y}) \quad (3)$$

So we have $\forall \alpha \in [0, 1]$:

$$g(\alpha \mathbf{x} + (1 - \alpha) \mathbf{y}) = F(f(\alpha \mathbf{x} + (1 - \alpha) \mathbf{y})) \quad (4)$$

$$\leq F(\alpha f(\mathbf{x}) + (1 - \alpha) f(\mathbf{y})) \quad (5)$$

$$\leq \alpha F(f(\mathbf{x})) + (1 - \alpha) F(f(\mathbf{y})) \quad (6)$$

$$= \alpha g(\mathbf{x}) + (1 - \alpha) g(\mathbf{y}) \quad (7)$$

We obtain (5) because F is non-decreasing and (3), and we obtain (6) because F is convex. Thus from (7) we obtain that $g(\mathbf{x})$ is convex.

2. Because f is convex, so $\forall \alpha \in [0, 1]$, we have

$$f(\alpha \mathbf{x} + (1 - \alpha) \mathbf{y}) \leq \alpha f(\mathbf{x}) + (1 - \alpha) f(\mathbf{y}) \quad (8)$$

So we have $\forall \alpha \in [0, 1]$:

$$g(\alpha \mathbf{x} + (1 - \alpha) \mathbf{y}) = F(f_1(\alpha \mathbf{x} + (1 - \alpha) \mathbf{y}), \dots, f_m(\alpha \mathbf{x} + (1 - \alpha) \mathbf{y})) \quad (9)$$

$$\leq F(\alpha f_1(\mathbf{x}) + (1 - \alpha) f_1(\mathbf{y}), \dots, \alpha f_m(\mathbf{x}) + (1 - \alpha) f_m(\mathbf{y})) \quad (10)$$

$$\leq \alpha F(f_1(\mathbf{x}), \dots, f_m(\mathbf{x})) + (1 - \alpha) F(f_1(\mathbf{y}), \dots, f_m(\mathbf{y})) \quad (11)$$

$$= \alpha g(\mathbf{x}) + (1 - \alpha) g(\mathbf{y}) \quad (12)$$

We obtain (10) because F is non-decreasing and (8), and we obtain (11) because F is convex. Thus from (12) we obtain that $g(\mathbf{x})$ is convex. □

Excercise 3. Proof that if $f(\mathbf{x}, \mathbf{y})$ is convex in $(\mathbf{x}, \mathbf{y}) \in \mathbb{R}^n$ and Y is a convex set, then

$$g(\mathbf{x}) = \inf_{\mathbf{y} \in Y} f(\mathbf{x}, \mathbf{y})$$

is convex.

Proof of Excercise 3: For given $\mathbf{x}_1, \mathbf{x}_2$, we define $\{\mathbf{y}_n^{(1)}\}, \{\mathbf{y}_n^{(2)}\}$ as follows:

$$\inf_{\mathbf{y} \in Y} f(\mathbf{x}_1, \mathbf{y}) = \lim_{n \rightarrow \infty} f(\mathbf{x}_1, \mathbf{y}_n^{(1)}) \quad (13)$$

$$\inf_{\mathbf{y} \in Y} f(\mathbf{x}_2, \mathbf{y}) = \lim_{n \rightarrow \infty} f(\mathbf{x}_2, \mathbf{y}_n^{(2)}) \quad (14)$$

So we have:

$$\alpha g(\mathbf{x}_1) + (1 - \alpha) g(\mathbf{x}_2) = \alpha \inf_{\mathbf{y} \in Y} f(\mathbf{x}_1, \mathbf{y}) + (1 - \alpha) \inf_{\mathbf{y} \in Y} f(\mathbf{x}_2, \mathbf{y}) \quad (15)$$

$$= \alpha \lim_{n \rightarrow \infty} f(\mathbf{x}_1, \mathbf{y}_n^{(1)}) + (1 - \alpha) \lim_{n \rightarrow \infty} f(\mathbf{x}_2, \mathbf{y}_n^{(2)}) \quad (16)$$

$$= \lim_{n \rightarrow \infty} (\alpha f(\mathbf{x}_1, \mathbf{y}_n^{(1)}) + (1 - \alpha) f(\mathbf{x}_2, \mathbf{y}_n^{(2)})) \quad (17)$$

$$\geq \lim_{n \rightarrow \infty} f(\alpha \mathbf{x}_1 + (1 - \alpha) \mathbf{x}_2, \alpha \mathbf{y}_n^{(1)} + (1 - \alpha) \mathbf{y}_n^{(2)}) \quad (18)$$

$$\geq \inf_{\mathbf{y} \in Y} f(\alpha \mathbf{x}_1 + (1 - \alpha) \mathbf{x}_2, \mathbf{y}) \quad (19)$$

$$= g(\alpha \mathbf{x}_1 + (1 - \alpha) \mathbf{x}_2) \quad (20)$$

We obtain (18) because $f(\mathbf{x}, \mathbf{y})$ is convex, and we obtain (19) since Y is a convex set. So from (20) we have proved that $g(\mathbf{x})$ is also convex. □

Excercise 4. Proof that the following univariate functions are in the set of $\mathcal{F}^1(\mathbb{R})$

$$f(x) = e^x,$$

$$f(x) = |x|^p, \quad p > 1,$$

$$f(x) = \frac{x^2}{1 + |x|},$$

$$f(x) = |x| - \ln(1 + |x|).$$

Proof of Exercise 4: From lecture, we have a continuously differentiable function f belongs to $\mathcal{F}^1(\mathbb{R}^n)$ iff for any $x, y \in \mathbb{R}^n$ we have:

$$\langle \nabla f(x) - \nabla f(y), x - y \rangle \geq 0 \quad (21)$$

1. $f(x) = e^x$ Obviously, we have:

$$\langle \nabla f(x) - \nabla f(y), x - y \rangle = (e^x - e^y)(x - y) \geq 0 \quad (22)$$

So $f(x) = e^x$ is in the set of $\mathcal{F}^1(\mathbb{R})$

2. $f(x) = |x|^p, p > 1$ we have:

$$f'(x) = \begin{cases} p|x|^{(p-1)} & x > 0 \\ 0 & x = 0 \\ -p|x|^{(p-1)} & x < 0 \end{cases} \quad (23)$$

Now divide the problem into four cases:

(a) $x \geq 0, y \geq 0$: we have $\langle \nabla f(x) - \nabla f(y), x - y \rangle = p(x^{p-1} - y^{p-1})(x - y) \geq 0$

(b) $x \geq 0, y < 0$ we have $\langle \nabla f(x) - \nabla f(y), x - y \rangle = p(x^{p-1} + |y|^{p-1})(x - y) \geq 0$

(c) $x < 0, y \geq 0$ we have $\langle \nabla f(x) - \nabla f(y), x - y \rangle = p(-|x|^{p-1} - y^{p-1})(x - y)$

since $x - y < 0$ and $-|x|^{p-1} - y^{p-1} < 0$, we have $\langle \nabla f(x) - \nabla f(y), x - y \rangle \geq 0$

(d) $x < 0, y < 0$ we have $\langle \nabla f(x) - \nabla f(y), x - y \rangle = p(-|x|^{p-1} + |y|^{p-1})(x - y) \geq 0$

Above all, we have $\langle \nabla f(x) - \nabla f(y), x - y \rangle \geq 0$, So end of the proof.

3. $f(x) = \frac{x^2}{1+|x|}$ Firstly, we have

$$f'(x) = \begin{cases} 1 - \frac{1}{(1+|x|)^2} & x > 0 \\ 0 & x = 0 \\ \frac{1}{(1+|x|)^2} - 1 & x < 0 \end{cases} \quad (24)$$

Then:

(a) $x \geq 0, y \geq 0$ $\langle \nabla f(x) - \nabla f(y), x - y \rangle = (\frac{1}{(1+|x|)^2} - \frac{1}{(1+|y|)^2})(x - y) \geq 0$

(b) $x \geq 0, y < 0$ $\langle \nabla f(x) - \nabla f(y), x - y \rangle = (2 - \frac{1}{(1+|x|)^2} - \frac{1}{(1+|y|)^2})(x - y)$ Since $x > y$ and $f(x) = \frac{1}{(1+|x|)^2} \leq 1$, so $2 - \frac{1}{(1+|x|)^2} - \frac{1}{(1+|y|)^2} \geq 0$, thus $\langle \nabla f(x) - \nabla f(y), x - y \rangle \geq 0$

(c) $y \geq 0, x < 0$ simliar as case (b), just switching x, y .

(d) $x < 0, y < 0$ $\langle \nabla f(x) - \nabla f(y), x - y \rangle = (\frac{1}{(1+|y|)^2} - \frac{1}{(1+|x|)^2})(x - y) \geq 0$

Above all, $\langle \nabla f(x) - \nabla f(y), x - y \rangle \geq 0$, so end of the proof.

4. $f(x) = |x| - \ln(1 + |x|)$ Firstly, we have:

$$f'(x) = \begin{cases} 1 - \frac{1}{1+|x|} & x > 0 \\ 0 & x = 0 \\ \frac{1}{1+|x|} - 1 & x < 0 \end{cases} \quad (25)$$

Then:

$$(a) \ x \geq 0, y \geq 0 \ \langle \nabla f(x) - \nabla f(y), x - y \rangle = (\frac{1}{1+|x|} - \frac{1}{1+|y|})(x - y) \geq 0$$

$$(b) \ x \geq 0, y < 0 \ \langle \nabla f(x) - \nabla f(y), x - y \rangle = (2 - \frac{1}{1+|y|} - \frac{1}{1+|x|})(x - y) \text{ Also, we have } f(x) = \frac{1}{1+|x|} \leq 1, \text{ So } 2 - \frac{1}{1+|y|} - \frac{1}{1+|x|} \geq 0, \text{ thus } \langle \nabla f(x) - \nabla f(y), x - y \rangle \geq 0$$

$$(c) \ x < 0, y \geq 0: \text{ simliar as case (b), just switching } x, y.$$

$$(d) \ x < 0, y < 0: \langle \nabla f(x) - \nabla f(y), x - y \rangle = (\frac{1}{1+|y|} - \frac{1}{1+|x|})(x - y) \geq 0$$

Above all, we have $\langle \nabla f(x) - \nabla f(y), x - y \rangle \geq 0$, so end of the proof. \square

Excercise 5. For $f \in \mathcal{F}_L^{1,1}(\mathbb{R}^n)$ and function $\phi(\mathbf{y}) = f(\mathbf{y}) - \langle \nabla f(\mathbf{x}_0), \mathbf{y} \rangle$, prove that $\phi \in \mathcal{F}_L^{1,1}(\mathbb{R}^n)$, and its optimal point is $\mathbf{y}^* = \mathbf{x}_0$.

Proof of Excercise 5:

1. $\phi(y)$ is continuously differentiable: Obviously we have:

$$\nabla \phi(y) = \nabla f(y) - \nabla f(x_0) \quad (26)$$

Since $f \in \mathcal{F}_L^{1,1}(\mathbb{R}^n)$, so $f(x)$ is continuously differentiable. So $\phi(y)$ is continuously differentiable.

2. $\phi(y)$ is convex: Since $f \in \mathcal{F}_L^{1,1}(\mathbb{R}^n)$, so $f(x)$ is convex, which means $\langle \nabla f(x) - \nabla f(y), x - y \rangle \geq 0$
So

$$\langle \nabla \phi(x) - \nabla \phi(y), x - y \rangle = \langle \nabla f(x) - \nabla f(y), x - y \rangle \geq 0 \quad (27)$$

which means $\phi(x)$ is convex.

3. $\phi(y)$ satisfies the Lipschitz continuous with constant L . Since $f \in \mathcal{F}_L^{1,1}(\mathbb{R}^n)$, so $|\nabla f(x) - \nabla f(y)| \leq L\|x - y\|$ we have

$$|\nabla \phi(x) - \nabla \phi(y)| = |\nabla f(x) - \nabla f(y)| \leq L\|x - y\| \quad (28)$$

Above all, $\phi(y) \in \mathcal{F}_L^{1,1}(\mathbb{R}^n)$ We can easily find that $\nabla \phi(\mathbf{x}_0) = 0$.

Because of the properties that the convex function satisfies, $\mathbf{y}^* = \mathbf{x}_0$ must be the global minimum, i.e, optimal point. \square

Excercise 6. Proof that, for $f : \mathbb{R}^n \rightarrow \mathbb{R}$ and α from $[0, 1]$, if

$$\begin{aligned} \alpha f(\mathbf{x}) + (1 - \alpha)f(\mathbf{y}) &\geq f(\alpha\mathbf{x} + (1 - \alpha)\mathbf{y}) \\ &\quad + \frac{\alpha(1 - \alpha)}{2L} \|\nabla f(\mathbf{x}) - \nabla f(\mathbf{y})\|^2, \end{aligned}$$

then $f \in \mathcal{F}_L^{1,1}(\mathbb{R}^n)$.

Proof of Excercise 6:

$$\begin{aligned} \alpha f(\mathbf{x}) + (1 - \alpha)f(\mathbf{y}) &\geq f(\alpha\mathbf{x} + (1 - \alpha)\mathbf{y}) + \frac{\alpha(1 - \alpha)}{2L} \|\nabla f(\mathbf{x}) - \nabla f(\mathbf{y})\|^2 \\ &\geq f(\alpha\mathbf{x} + (1 - \alpha)\mathbf{y}) \end{aligned} \quad (29)$$

So $f(x)$ is convex.

Now we should prove that f satisfies the Lipschitz condition. We have:

$$\begin{aligned}\alpha f(\mathbf{x}) + (1 - \alpha)f(\mathbf{y}) &\geq f(\alpha\mathbf{x} + (1 - \alpha)\mathbf{y}) + \frac{\alpha(1 - \alpha)}{2L} \|\nabla f(\mathbf{x}) - \nabla f(\mathbf{y})\|^2 \\ f(\mathbf{y}) &\geq \frac{f(\alpha\mathbf{x} + (1 - \alpha)\mathbf{y}) - \alpha f(\mathbf{x})}{1 - \alpha} + \frac{\alpha}{2L} \|\nabla f(\mathbf{x}) - \nabla f(\mathbf{y})\|^2\end{aligned}\quad (30)$$

When $\alpha \rightarrow 1$,

$$f(\mathbf{y}) \geq f(\mathbf{x}) + \langle \nabla f(\mathbf{x}), \mathbf{y} - \mathbf{x} \rangle + \frac{1}{2L} \|\nabla f(\mathbf{x}) - \nabla f(\mathbf{y})\|^2 \quad (31)$$

switching \mathbf{x}, \mathbf{y} in (27), and add them together, we obtain:

$$\frac{1}{L} \|\nabla f(\mathbf{x}) - \nabla f(\mathbf{y})\|^2 \leq \langle \nabla f(\mathbf{y}) - \nabla f(\mathbf{x}), \mathbf{y} - \mathbf{x} \rangle \quad (32)$$

Applying Cauchy-Schwarz inequality,

$$\frac{1}{L} \|\nabla f(\mathbf{x}) - \nabla f(\mathbf{y})\|^2 \leq \langle \nabla f(\mathbf{y}) - \nabla f(\mathbf{x}), \mathbf{y} - \mathbf{x} \rangle \quad (33)$$

$$\leq \|\nabla f(\mathbf{y}) - \nabla f(\mathbf{x})\| \|\mathbf{y} - \mathbf{x}\| \quad (34)$$

That is: $\|\nabla f(\mathbf{x}) - \nabla f(\mathbf{y})\| \leq L \|\mathbf{y} - \mathbf{x}\|$.

So, $f \in \mathcal{F}_L^{1,1}(\mathbb{R}^n)$

□

Exercise 7. Proof that, for $f : \mathbb{R}^n \rightarrow \mathbb{R}$ and α from $[0, 1]$, if

$$\begin{aligned}0 &\leq \alpha f(\mathbf{x}) + (1 - \alpha)f(\mathbf{y}) - f(\alpha\mathbf{x} + (1 - \alpha)\mathbf{y}) \\ &\leq \alpha(1 - \alpha) \frac{L}{2} \|\mathbf{x} - \mathbf{y}\|^2,\end{aligned}$$

then $f \in \mathcal{F}_L^{1,1}(\mathbb{R}^n)$.

Proof of Exercise 7: From the inequality given above, we can obtain:

$$f(\mathbf{y}) \leq \frac{f(\alpha\mathbf{x} + (1 - \alpha)\mathbf{y}) - \alpha f(\mathbf{x})}{1 - \alpha} + \frac{\alpha L}{2} \|\mathbf{y} - \mathbf{x}\|^2 \quad (35)$$

When $\alpha \rightarrow 1$, we have

$$f(\mathbf{y}) \leq f(\mathbf{x}) + \langle \nabla f(\mathbf{x}), \mathbf{y} - \mathbf{x} \rangle + \frac{L}{2} \|\mathbf{y} - \mathbf{x}\|^2 \quad (36)$$

And it is equivalent to the conclusion that $f \in \mathcal{F}_L^{1,1}(\mathbb{R}^n)$. So end of the proof.

□