

Math homework 4 - OSM Bootcamp 2018

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2018.07.15

Exercise 6.6

The FOC are:

$$\begin{aligned}f_x &= 6xy + 4y^2 = 0 \\f_y &= 3x^2 + 8xy + x = 0\end{aligned}$$

First, we need to go through some cases to check whether x or y can be 0.

If $x = 0$ then $4y^2 + y = 0 \Rightarrow y$ is imaginary

If $x \neq 0$ then $3x + 8y + 1 = 0$

Then if $y = 0, x \neq 0$ then $3x + 1 = 0 \Rightarrow x = -\frac{1}{3}$

If $x \neq 0, x \neq 0$ then:

$$\begin{aligned}f_x &= 6x + 4y + 1 = 0 \\f_y &= 3x + 8y + 1 = 0 \\(-16y - 1) + (4y + 1) &= 0 \\-12y - 1 &= 0 \\\Rightarrow y &= -\frac{1}{12} \\\Rightarrow 3x - \frac{8}{12} + \frac{12}{12} &= 0 \\x &= \frac{-4}{36}\end{aligned}$$

And finally our critical points are $(0, 0)$, $(-\frac{1}{3}, 0)$, $(-\frac{4}{36}, -\frac{1}{12})$

Exercise 6.7

(i) Denote $A = [a_{i,j}]$

Then $A^T = [a_{j,i}]$ and $Q = [a_{j,i} + a_{i,j}] = [a_{i,j} + a_{j,i}] = Q^T$

And $x^T Q x = x^T (A^T + A)x = (x^T A^T + x^T A)x = x^T A^T x + x^T A x = 2x^T A x \square$

(ii) $f(x) = \frac{1}{2}x^T Q x - b^T x + c$

$$\begin{aligned}f'(x) &= Q^T x - b = 0 \\\Rightarrow Q^T x^* &= b\end{aligned}$$

(iii) NEED TO DO THIS QUESTION

Exercise 6.11

$$x_1 = x_0 - f'(x_0)/f''(x_0)$$

$$x_1 = \frac{-b}{2ax_0} \Rightarrow f'(x_0) = 2ax_0 + b = -b + b = 0$$

And also, $f''(x_0) = 2a > 0$

Exercise 7.1

Let $x, y \in \text{conv}(S)$

Then x and y are of the form:

$$a_1x_1 + \dots + a_nx_n = x \text{ and } b_1y_1 + \dots + b_my_m = y$$

Where $\sum a_i = \sum b_i = 1$ and $x_i, y_i \in S$

Then if $\lambda \in [0, 1]$ then $\lambda x + (1 - \lambda)y = \sum_{i=1}^n \lambda a_i x_i + \sum_{i=1}^m (1 - \lambda)b_i y_i \in \text{conv}(S)$

Note, the last part holds because $\sum \lambda a_i + \sum (1 - \lambda)b_i = 1$

Exercise 7.2

(i) Let $P = \{x \in V | \langle a, x \rangle = b\}$

Then let $x, y \in P$ so $\langle a, x \rangle = \langle a, y \rangle = b$

Let $\lambda \in [0, 1]$ then $\langle a, \lambda x + (1 - \lambda)y \rangle =$

$$= \lambda \langle a, x \rangle + (1 - \lambda) \langle a, y \rangle = b \square$$

(ii) Let $x, y \in H = \{x \in V | \langle a, x \rangle \leq b\}$ and $\lambda \in [0, 1]$.

Then $\langle a, \lambda x + (1 - \lambda)y \rangle =$

$$= \lambda \langle a, x \rangle + (1 - \lambda) \langle a, y \rangle \leq b \square$$

Exercise 7.4

(i) $\|x - y\|^2 = \|x - p + p - y\|^2 = \langle x - p + p - y, x - p + p - y \rangle$

$$= \langle x - p, x - p \rangle + \langle x - p, p - y \rangle + \langle p - y, x - p \rangle + \langle p - y, p - y \rangle$$

$$= \|x - p\|^2 + \|p - y\|^2 + 2\langle x - p, p - y \rangle$$

(ii) By (i) we have $\|x - y\|^2 = \|x - p\|^2 + \|p - y\|^2 + 2\langle x - p, p - y \rangle$

And by assumption we have $\langle x - p, p - y \rangle \geq 0$

And $\|p - y\|^2 > 0$ if $y \neq p$

Thus, together $\|x - y\|^2 > \|x - p\|^2 \Rightarrow \|x - y\| > \|x - p\|$, where the inequality is consistent across the square root due to the positivity of the norm.

(iii) By (i), $\|x - y\|^2 = \|x - p\|^2 + \langle x - p, p - z \rangle + \|p - z\|^2$ Then if $z = \lambda y + (1 - \lambda)p$ we have

$$\|x - z\|^2 = \|x - p\|^2 + 2\langle x - p, p - \lambda y - (1 - \lambda)p \rangle + \|p - \lambda y - (1 - \lambda)p\|^2$$

$$= \|x - p\|^2 + 2\lambda \langle x - p, p - y \rangle + \lambda^2 \|y - p\|^2$$

(iv) **NEED TO DO THIS QUESTION**

Exercise 7.8

f is convex so $f(\lambda x + (1 - \lambda)y) \leq \lambda f(x) + (1 - \lambda)f(y)$

I think there's a typo in the question because the matrix/vector operations in the argument are not compatible. If $A \in M_{m \times n}$ then $Ax \in \mathbb{R}^n$ but $b \in \mathbb{R}^m$ so the two can't be added?

So I'll assume $A \in M_{m \times m}$ for compatibility.

Let $x, y \in \mathbb{R}^m$ then $x' = Ax + b$ and $y' = Ay + b$ are also in \mathbb{R}^m Thus, $f(\lambda x' + (1 - \lambda)y') \leq \lambda f(x') + (1 - \lambda)f(y')$

and therefore $g(\lambda x + (1 - \lambda)y) \leq g(x) + (1 - \lambda)g(y)$

Exercise 7.12

(i) Let $A, B \in PD_n(\mathbb{R})$

Then $\lambda A + (1 - \lambda)B = C$

And $v^T C v = \lambda v^T A v + (1 - \lambda)v^T B v \geq 0$, so we have $C \in PD_n(\mathbb{R})$.

(ii) **NEED TO DO THIS QUESTION**

Exercise 7.13

Suppose f is not constant. Then there exists some x, y such that $f(x) > f(y)$

By convexity we have $f(\lambda x + (1 - \lambda)y) \leq \lambda f(x) + (1 - \lambda)f(y)$

$$\Rightarrow \frac{\lambda f(x) + (1 - \lambda)f(y)}{\lambda} = \frac{f(x) - f(y)}{\lambda} \leq f\left(\frac{\lambda x + (1 - \lambda)y}{\lambda}\right)$$

And $f(x) > f(y) \Rightarrow \frac{f(x) - f(y)}{\lambda}$ which goes to infinity as λ goes to zero, which would imply the function is not bounded, which is a contradiction.

Exercise 7.20

By the convexity of both f and $-f$ we have the following inequalities:

$$f(tx + (1 - t)y) \leq tf(x) + (1 - t)f(y)$$

$$-f(tx + (1 - t)y) \leq -tf(x) - (1 - t)f(y)$$

We can multiply the second equation by negative 1, and then see that for both to hold simultaneously we have equality. Thus,

$$f(tx + (1 - t)y) = tf(x) + (1 - t)f(y)$$

And we have our result.

Exercise 7.21

Let x^* be a local min. Then $f(x) \geq f(x^*)$ for all x in some epsilon neighborhood of x^*

And because ϕ is strictly increasing we have $\phi(f(x)) \geq \phi(f(x^*))$ for that same neighborhood so x^* is still a minimizer.

Similarly, going the other way, if $\phi(f(x)) \leq \phi(f(x^*))$ and because ϕ is monotone then $f(x^*) \leq f(x)$ for the neighborhood.