

最优化作业二

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1 Problem 1

1.1 a

解:

$$f(x) = -\sum_{i=1}^n \log x_i$$

$$f'(x) = -\sum_{i=1}^n \frac{1}{x_i}$$

$$f''(x) = \sum_{i=1}^n \frac{1}{x_i^2}$$

$$f''(x) > 0$$

所以 $f(x)$ 为严格凸函数.

1.2 b

解:

如果 f 二阶可微且为凸函数, 则:

$$f(y) \geq f(x) + \nabla f(x)^T (y - x)$$

$$f(x) \geq f(y) + \nabla f(y)^T (x - y)$$

$$\therefore (\nabla f(x) - \nabla f(y))^T (x - y) > 0$$

如果 ∇f 是单调的, 那么

令

$$\begin{aligned}
 g(t) &= f(x+t(y-x)), t \in [0, 1] \\
 g'(t) &= \nabla f(x+t(y-x))^T (y-x) \\
 g'(1) - g'(0) &> 0 \\
 \therefore g'(t) - g'(0) &\leq 0 \\
 \therefore f(y) = g(1) &= g(0) + \int_0^1 g'(x) dx \geq g(0) + g'(0) = f(x) + \nabla f(x)^T (y-x)
 \end{aligned}$$

因此 f 为凸函数

1.3 c

证明: 即需要证明: $\text{dom}(g)$ 是凸集, 且 $\forall (x, t), (y, s) \in \text{dom}(g), 0 < \theta < 1$, 有 $g(\theta x + (1-\theta)y, \theta t + (1-\theta)s) \leq \theta g(x, t) + (1-\theta)g(y, s)$

$$\begin{aligned}
 g(x, t) &= tf\left(\frac{x}{t}\right), f(x) \text{ 是凸函数} \\
 g(\theta x + (1-\theta)y, \theta t + (1-\theta)s) &= (\theta t + (1-\theta)s)f\left(\frac{\theta x + (1-\theta)y}{\theta t + (1-\theta)s}\right) \\
 \therefore g(x, t) &= tf\left(\frac{x}{t}\right), g(y, s) = sf\left(\frac{y}{s}\right) \\
 \therefore f\left(\frac{\theta x + (1-\theta)y}{\theta t + (1-\theta)s}\right) &= \frac{f(\theta t(\frac{x}{t}) + (1-\theta)s(\frac{y}{s}))}{\theta t + (1-\theta)s} \\
 \therefore g(\theta x + (1-\theta)y, \theta t + (1-\theta)s) &= f(\theta t(\frac{x}{t}) + (1-\theta)s(\frac{y}{s})) \\
 &\leq \theta tf\left(\frac{x}{t}\right) + (1-\theta)sf\left(\frac{y}{s}\right) = \theta g(x, t) + (1-\theta)g(y, s)
 \end{aligned}$$

又因定义域 $\text{dom}(g)$ 是 $\text{dom}(f)$ 在透视函数 $P: R^{n+1} \rightarrow R^n, P(x, t) = x/t (t > 0)$ 下的逆象, $\text{dom}(f)$ 为凸集, 故 $\text{dom}(g)$ 是凸集, g 是凸函数。

2 Problem 2

$$\begin{aligned}
\frac{\partial f(x)}{\partial x_i} &= \left(\sum_{i=1}^n x_i^p \right)^{\frac{1-p}{p}} x_i^{p-1} = \left(\frac{f(x)}{x_i} \right)^{1-p} \\
\frac{\partial^2 f(x)}{\partial x_i \partial x_j} &= \frac{1-p}{x_i} \left(\frac{f(x)}{x_i} \right)^{-p} \left(\frac{f(x)}{x_j} \right)^{1-p} = \frac{1-p}{f(x)} \left(\frac{f^2(x)}{x_i x_j} \right)^{1-p} \\
\text{while } i \neq j, \frac{\partial^2 f(x)}{\partial x_i^2} &= \frac{1-p}{f(x)} \left(\frac{f(x)^2}{x_i} \right)^{1-p} - \frac{1-p}{x_i} \left(\frac{f(x)}{x_i} \right)^{1-p} \\
\therefore \nabla^2 f(x) &= \frac{1-p}{f(x)} \left(\left(\sum_{i=1}^n \frac{f(x)^{1-p}}{x_i^{1-p}} \right)^2 - \sum_{i=1}^n \frac{f(x)^{2-p}}{x_i^{2-p}} \right) \\
\left(\sum_{i=1}^n \frac{f(x)^{1-p}}{x_i^{1-p}} \right)^2 &= \left(\sum_{i=1}^n \left(\frac{f(x)}{x_i} \right)^{-p/2} \left(\frac{f(x)}{x_i} \right)^{1-p/2} \right)^2 \\
&\leq \sum_{i=1}^n \frac{f(x)^{-p}}{x_i} \sum_{i=1}^n \frac{f(x)^{2-p}}{x_i} \\
&= \sum_{i=1}^n \left(\frac{f(x)}{x_i} \right)^{2-p} \\
\therefore \nabla^2 f(x) &\leq 0 \\
\therefore f(x) &\text{ 是凹函数}
\end{aligned}$$

3 Problem 3

3.1 a

$$\Delta \psi(x, y) = \psi(x) - \psi(y) - \langle \nabla \psi(y), x - y \rangle = \psi(x) - \psi(y) - \nabla \psi(y)^T (x - y)$$

因为 ψ 严格凸，连续可微所以有：

$$\psi(x) \geq \psi(y) + \nabla \psi(y)^T (x - y)$$

$$\Delta \psi(x, y) \geq 0$$

当 $x = y$ 时， $\psi(x) = \psi(y) = 0, x = y$, 即 $\Delta \psi(x, y) = 0$

3.2 b

$$f(x) = L(x) + \Delta_{\psi}(x, x_0)$$

$$\nabla f(x) = \nabla L(x) + \nabla \psi(x) - \nabla \psi(x_0)$$

$$\therefore \nabla f(x^*) = 0$$

$$g(y) = f(y) - f(x^*) - \Delta_{\psi}(y, x^*)$$

$$\nabla g(y) = \nabla f(y) - \nabla \psi(x^*) - \nabla \psi(x^*) = \nabla L(y) + \nabla \psi(x^*) - \nabla \psi(x_0)$$

$$\nabla^2 g(y) = \nabla^2 L(y) \geq 0$$

$$g(x^*) = \nabla g(x^*) = 0$$

因此函数单调递增，原式得证。

4 Problem 4**4.1 a**

证明：

$$\begin{aligned}
& \langle \prod_X(x) - \prod_X(y), x - y \rangle - \|\prod_X(x) - \prod_X(y)\| \\
&= \langle \prod_X(x) - \prod_X(y), (x - \prod_X(x)) + (\prod_X(y) - y) \rangle \\
& \quad x^* = \operatorname{argmin} \frac{1}{2} \|x - y\|_2^2 \\
& \therefore \langle x - x^*, \prod_X(x) - \prod_X(y) \rangle = 0 \\
& \text{if } x^* \in X, \langle \prod_X(x) - \prod_X(y), x - \prod_X(x) \rangle = 0 \\
& \text{else } \langle \prod_X(x) - \prod_X(y), x - \prod_X(x) \rangle \geq 0 \\
& \therefore \langle \prod_X(x) - \prod_X(y), x - \prod_X(x) \rangle \geq 0 \\
& \therefore \langle \prod_X(x) - \prod_X(y), \prod_X(y) - y \rangle \geq 0 \\
& \therefore \langle \prod_X(x) - \prod_X(y), (x - \prod_X(x)) + (\prod_X(y) - y) \rangle \geq 0
\end{aligned}$$

得证

4.2 b

$$\begin{aligned}
& \because \langle \prod_X(x) - \prod_X(y), x - \prod_X(x) \rangle \geq 0 \\
& \quad \langle \prod_X(x) - \prod_X(y), \prod_X(y) - y \rangle \geq 0
\end{aligned}$$

x, y 在以 $\prod_X(x) - \prod_X(y)$ 为法向量, $\prod_X(x)$ 为交点的超平面两侧, 同理 x, y 在以 $\prod_X(x) - \prod_X(y)$ 为法向量, $\prod_X(y)$ 为交点的超平面两侧。

$$\therefore \|\prod_X(x) - \prod_X(y)\|_2 \leq \|x - y\|_2$$

5 Problem 5

5.1 a

$$f^*(y) = \sup(yx - \max\{0, 1 - x\})$$

$$\text{if } x < 1, \frac{\partial f^*}{\partial x} = 1 + y$$

$$\text{else } \frac{\partial f}{\partial x} = y$$

为了保证有上界 $y \in [-1, 0]$

$$\therefore f^*(y) = y, y \in [-1, 0]$$

5.2 b

$$f^*(y) = \sup(yx - \ln(1 + e^{-x}))$$

$$\frac{\partial f^*}{\partial x} = y + \frac{1}{1 + e^x}$$

$$\frac{1}{1 + e^x} \in (0, 1)$$

为保证有上界, 则 $y \in (-1, 0)$

$$\therefore f^*(y) = (y + 1) \ln(y + 1) - y \ln(-y)$$