

midterm exam

1. True or False. You are not required to give any justification of your answer. (2*5=10)

⌈ (1) Let $\Omega \subseteq \mathbb{R}^n$ and \bar{x} is an interior point of Ω . Then any vector $d \in \mathbb{R}^n \setminus \{0\}$ is a feasible direction at $\bar{x} \in \Omega$.

(2) The sequence $\{\log(1 + \frac{1}{k})\}_{k \geq 1}$ exhibits a linear rate of convergence.

⌈ (3) Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be a smooth function with $\bar{x} \in \mathbb{R}^n$. Then $-\nabla f(\bar{x})$ is the max-rate descending direction of f at \bar{x} if \bar{x} is not stationary.

⌈ (4) The quasi-Newton methods avoid calculation of second derivatives of the objective function.

⌈ (5) If a standard-form linear programming problem has a feasible solution, then there exists an extreme point solution.

2. (5+5=10). Answer True or False for each of the following statements and justify your answer.

(a) We say a set $C \subseteq \mathbb{R}^n$ is locally closed around $\bar{x} \in C$ if there is $r > 0$ such that the set $C \cap \mathbb{B}_r(\bar{x})$ is closed, where $\mathbb{B}_r(\bar{x})$ stands for the closed ball with center \bar{x} and radius $r > 0$. Then $\Omega \subset \mathbb{R}^n$ is closed if it is locally closed around any $x \in \Omega$.

(b) Applying the Newton's method to solve $\min_{x \in \mathbb{R}^n} f(x) = \frac{1}{2}x^T Qx - b^T x$, where Q is symmetric and invertible. It takes at least two iterations to obtain the optimal solution.

3. (16) linear programming.

$$\begin{array}{ll} \max & 4x_1 + 3x_2 + 5x_3 \\ \text{s.t.} & 3x_1 + x_2 + 3x_3 \leq 30 \\ & 2x_1 + 2x_2 + 3x_3 \leq 40 \\ & x_1, x_2, x_3 \geq 0 \end{array}$$

4. (6+10=16).

(a) The table below shows the amount of materials A and materials B needed to make each product. The right column shows that the total amount of A and B available to make these productions. Establish integer linear programming model to maximize the total profit

	Production 1	Production 2	Total available
material A	3	5	300
material B	7	5	350
Profit of each production	7.5	10	

(b) 0 – 1 integer programming.

$$\begin{aligned}
\max \quad & z = 26x_1 + 28x_2 + 9x_3 + 10x_4 \\
s.t. \quad & 14x_1 + 17x_2 + 6x_3 + 7x_4 \leq 24 \\
& 11x_1 + 11x_2 + 3x_3 + 4x_4 \leq 14 \\
& 8x_1 + 7x_2 + 2x_3 \leq 8 \\
& x_1 + x_2 \geq 1 \\
& x_1, x_2, x_3, x_4 \in \{0, 1\}
\end{aligned}$$

5. (6+6+6=18). Consider the optimality conditions on minimizing a differentiable function $f : \mathbb{R}^n \rightarrow \mathbb{R}$ over a closed set Ω . For any $x \in \Omega$, the tangent direction of Ω at x is given by

$$T(x) := \left\{ d \in \mathbb{R}^n \mid \exists \{t_k, d^k\}_{k \geq 0} \subset \mathbb{R}_{++} \times \mathbb{R}^n \text{ with } t_k \downarrow 0, d^k \rightarrow d \text{ as } k \rightarrow \infty \text{ and } x + t_k d^k \in \Omega \text{ for all } k \geq 0 \right\}.$$

(a) If \bar{x} is a local minimizer, then $d^T \nabla f(\bar{x}) \geq 0$ for all $d \in T(\bar{x})$.

(b) If \bar{x} is a local minimizer, then $\bar{d} = 0$ is a minimizer of the problem

$$\min_d d^T \nabla f(\bar{x}), \quad \text{s.t. } d \in T(\bar{x}).$$

(c) Let x^* be a feasible point. If there exists $\eta > 0$ such that

$$d^T \nabla f(x^*) \geq \eta \|d\|, \quad \forall d \in T(x^*).$$

Then x^* is a strict local minimizer.

6. (4+4+5+4+5+5+3=30). Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be μ -strongly convex and L -smooth ($0 < \mu < L$).

Consider the following algorithm:

Initialize $y^0 = x^0$.

for $k = 0, 1, 2, \dots$ **do**

$$\begin{aligned}
x^{k+1} &= y^k - \frac{1}{L} \nabla f(y^k) \\
y^{k+1} &= x^{k+1} + \frac{1-\theta}{1+\theta} (x^{k+1} - x^k),
\end{aligned}$$

where $\theta = \sqrt{\frac{\mu}{L}}$.

Moreover define an estimate function sequence $\{\varphi_k : \mathbb{R}^n \rightarrow \mathbb{R}\}_{k \geq 0}$ as follows:

$$\begin{aligned}\varphi_0(x) &= f(x^0) + \frac{\mu}{2} \|x - x^0\|^2; \\ \varphi_{k+1}(x) &= (1 - \theta) \varphi_k(x) + \theta \left[f(y^k) + \langle \nabla f(y^k), x - y^k \rangle + \frac{\mu}{2} \|x - y^k\|^2 \right] \text{ for } k = 0, 1, 2, \dots\end{aligned}$$

Show that

(a) Use the strong convexity of f , we have

$$\varphi_{k+1}(x) \leq (1 - \theta) \varphi_k(x) + \theta f(x).$$

(b) Use (a) by induction, we have

$$\varphi_k(x) \leq \left[1 - (1 - \theta)^k \right] f(x) + (1 - \theta)^k \varphi_0(x).$$

(c) Observe that φ_k has the quadratic form for any $k = 0, 1, 2, \dots$. Define $\varphi_k^* = \min_x \varphi_k(x)$. Then

$$\begin{aligned}\varphi_0^* &= f(x^0), \\ \varphi_{k+1}^* &= (1 - \theta) \varphi_k^* + \theta f(y^k) - \frac{1}{2L} \|\nabla f(y^k)\|^2 + \theta (1 - \theta) \left[\langle \nabla f(y^k), z^k - y^k \rangle + \frac{\mu}{2} \|z^k - y^k\|^2 \right],\end{aligned}$$

where

$$\begin{aligned}z^0 &= x^0; \\ z^{k+1} &= (1 - \theta) z^k + \theta \left(y^k - \frac{1}{\mu} \nabla f(y^k) \right).\end{aligned}$$

Hint: Suppose φ_k has the form $\varphi_k(x) = \varphi_k^* + \frac{\mu}{2} \|x - z^k\|^2$ and by induction.

(d) By the L -smoothness of f , we have

$$f(x^{k+1}) \leq f(y^k) - \frac{1}{2L} \|\nabla f(y^k)\|^2.$$

(e) Use the recursions of z^k in (c), we have

$$z^k = \frac{1 + \theta}{\theta} y^k - \frac{1}{\theta} x^k.$$

(f) Use (c), (d) and (e) by induction, we have $\varphi_k^* \geq f(x^k)$.

(g) Use (b) and (f) to show that

$$f(x^k) - f(x^*) \leq (1 - \theta)^k [\varphi_0(x^*) - f(x^*)].$$