

Sample Theory Questions for the 408 Final

I'm still working on this list of questions so more will be added at a later date. But I thought it important to get them to you early so that you can get started on them.

- (1) State and prove the *Weak Duality Theorem* for linear programming.
 (2) Let $A \in \mathbb{R}^{m \times n}$, $B \in \mathbb{R}^{k \times n}$ and $c \in \mathbb{R}^n$. Use linear programming duality to show that either

(A) there exists $x \in \mathbb{R}^n$ such that
 $Ax \leq 0$, $Bx = 0$, and $c^T x < 0$

or

(B) there exists $u \in \mathbb{R}^m$, $v \in \mathbb{R}^k$, such that
 $A^T u + B^T v = -c$ and $0 \leq u$,

but not both (A) and (B).

- (3) Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be continuously differentiable. Show that for any $x, d \in \mathbb{R}^n$,

$$f'(x; d) = \lim_{\substack{s \rightarrow d \\ t \downarrow 0}} \frac{f(x + ts) - f(x)}{t}.$$

- (4) Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be continuously differentiable. If \bar{x} solves $\min_{x \in \mathbb{R}^n} f(x)$, show that $\nabla f(\bar{x}) = 0$.
 (5) Given $\Omega \subset \mathbb{R}^n$ and $x \in \Omega$, show that every feasible direction for Ω at x is also a tangent direction to Ω at x .
 (6) Let $f_i : \mathbb{R}^n \rightarrow \mathbb{R}, i = 1, \dots, m$ be continuously differentiable and consider the set

$$\Omega = \{x \mid f_i(x) \leq 0, i = 1, \dots, s, f_i(x) = 0, i = s + 1, \dots, m\}.$$

Show that for every $x \in \Omega$ it is always the case that

$$T_\Omega(x) \subset \{d \mid \nabla f_i(x)^T d \leq 0, i \in I(x), \nabla f_i(x)^T d = 0, i = s + 1, \dots, m\},$$

where $I(x) = \{i \mid i \in \{1, \dots, s\}, f_i(x) = 0\}$.

- (7) Let $C_i \subset \mathbb{R}^n, i = 1, \dots, N$ be convex sets. Show that the set $C = \bigcap_{j=1}^N C_j$ is also a convex set.
 (8) Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be convex function, show that for any $\alpha \in \mathbb{R}$ the set $\{x \mid f(x) \leq \alpha\}$ is a convex set.
 (9) Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be convex function, and let $\phi : \mathbb{R} \rightarrow \mathbb{R}$ be a non-decreasing convex function. Show that the function $h : \mathbb{R}^n \rightarrow \mathbb{R}$ given by $h(x) = \phi(f(x))$ is also a convex function.
 (10) Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be convex function, $A \in \mathbb{R}^{n \times k}$, and $b \in \mathbb{R}^n$. Define $h : \mathbb{R}^k \rightarrow \mathbb{R}$ by $h(y) = f(Ay + b)$. Show that h is also a convex function.
 (11) Let $\ell \in \mathbb{R}^n$ and $u \in \mathbb{R}^n$ and consider the *box*

$$B = \{x \in \mathbb{R}^n : \ell_i \leq x_i \leq u_i, \text{ for } i = 1, 2, \dots, n\}.$$

- (a) Show that the vector $d \in \mathbb{R}^n$ is an element of the tangent cone to B at the point $x \in B$ if and only if

$$\begin{aligned} d_i &\geq 0 \text{ if } x_i = \ell_i, \\ d_i &\leq 0 \text{ if } x_i = u_i, \text{ and} \\ d_i &\text{ is free to be any real number if } \ell_i < x_i < u_i. \end{aligned}$$

(b) Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be continuously differentiable. Show that if $\bar{x} \in B$ solves the problem $\min\{f(x) : x \in B\}$, then

$$\begin{aligned} (\nabla f(\bar{x}))_i &\geq 0 && \text{if } x_i = \ell_i, \\ (\nabla f(\bar{x}))_i &\leq 0 && \text{if } x_i = u_i, \\ (\nabla f(\bar{x}))_i &= 0 && \text{otherwise.} \end{aligned}$$

- (12) Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be convex function, and let $\Omega \subset \mathbb{R}^n$ be a convex set. If \bar{x} is a local solution to the problem $\min_{x \in \Omega} f(x)$, show that \bar{x} is a global solution to this problem.
- (13) Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be a differentiable convex function, and let $\Omega \subset \mathbb{R}^n$ be a convex set. Show that \bar{x} solves the problem $\min_{x \in \Omega} f(x)$ if and only if $\nabla f(\bar{x})^T(x - \bar{x}) \geq 0$ for all $x \in \Omega$. (Hint: Use the the subdifferential inequality.)
- (14) Let $f_i : \mathbb{R}^n \rightarrow \mathbb{R}, i = 0, 1, \dots, s$ be convex differentiable functions and let $f_i : \mathbb{R}^n \rightarrow \mathbb{R}, i = s + 1, \dots, m$ be affine (i.e. $f_i(x) = a_i^T x + \alpha_i, i = s + 1, \dots, m$). and consider the convex set

$$\Omega = \{x \mid f_i(x) \leq 0, i = 1, \dots, s, f_i(x) = 0, i = s + 1, \dots, m\}.$$

Show that if $\bar{x} \in \mathbb{R}^n$ is a KKT point for the convex program $\min_{x \in \Omega} f_0(x)$, then \bar{x} must be a global solution to this problem.

- (15) Let $0 \leq \delta, Q \in \mathbb{R}^{n \times n}$ be symmetric and positive definite, and $m \in \mathbb{R}^n$ be such that each component of m is positive and the vectors m and e are linearly independent. Define the vector $e \in \mathbb{R}^n$ to be the vector each of whose components is the number 1. Consider the convex quadratic program

$$\begin{aligned} \mathcal{M} \quad &\text{mimize} \quad \frac{1}{2}x^T Q x \\ &\text{subject to} \quad e^T x = 1, m^T x \geq \delta. \end{aligned}$$

Use the KKT conditions for this problem to show that either

$$\frac{m^T Q^{-1} e}{e^T Q^{-1} e} \geq \delta$$

in which case

$$x_{\text{mv}} = \frac{1}{e^T Q^{-1} e} Q^{-1} e$$

solves \mathcal{M} , or the solution to \mathcal{M} is given by

$$\bar{x} = \alpha Q^{-1} e + \beta Q^{-1} m,$$

where α and β is the unique solution to the 2×2 system

$$\begin{bmatrix} e^T Q^{-1} e & e^T Q^{-1} m \\ e^T Q^{-1} m & m^T Q^{-1} m \end{bmatrix} \begin{pmatrix} \alpha \\ \beta \end{pmatrix} = \begin{pmatrix} 1 \\ \delta \end{pmatrix}.$$