

# COMPREHENSIVE EXAMINATION

Math 650 – Optimization

August 1998

You must show all your work for full credit!

## INSTRUCTIONS:

You must do problem 1 (35 points). Do either 2 or 3 (30 points), and either 4 or 5 (35 points), for a total of 3 problems and 100 points. You may do either 6 or 7 for extra credit (5 points).

**Q1.** Consider the optimization problem

$$\begin{array}{ll} \min & r \\ \text{subject to} & x_1^2 + x_2^2 - r \leq 0 \\ & (x_1 - 3)^2 + x_2^2 - r \leq 0 \\ & (x_1 - 2)^2 + (x_2 - 1)^2 - r \leq 0. \end{array}$$

- (a) Show that the above problem is exactly the problem of finding the smallest circle that contains the points  $(0, 0)$ ,  $(3, 0)$ , and  $(2, 1)$  in the plane (or the triangle determined by these points). Obviously  $r$  is the square of the radius of the circle. What do the variables  $(x_1, x_2)$  correspond to?
- (b) Write the Lagrangian for the problem using the multiplier  $\lambda_i$  for the  $i$ th constraint,  $i = 1, 2, 3$ . Explain why  $\lambda_0$  can be assumed to be 1 in the Fritz John conditions. Then write down the KKT conditions, including the complementarity conditions. Use the KKT conditions to verify that  $\lambda_i$ 's add up to one, express  $x_1, x_2$  in terms of the multipliers  $\lambda_i$ . Finally, show that the optimal  $x_i^*$ 's are non-negative.
- (c) Show that at least one multiplier must be zero by demonstrating that assumption that they are all non-zero leads to the contradiction that one of the variables  $x_i^* < 0$ .
- (d) Show that assumptions  $\lambda_2 = 0$ , and  $\lambda_1 \neq 0, \lambda_3 \neq 0$  lead to contradiction.
- (e) Similarly, one can show that the assumptions  $\lambda_1 = 0, \lambda_2 \neq 0, \lambda_3 \neq 0$  lead to a contradiction (do not show this). Use all the information so far to find the optimal solution(s)  $(x_1^*, x_2^*, r^*, \lambda_1^*, \lambda_2^*, \lambda_3^*)$  to the optimization problem.
- (f) Give a theoretical reason as to why the optimal solution you found above is the global one(s).

**Q2.** Let  $C \subseteq \mathbb{R}^n$  be a closed, convex set and  $x \in \mathbb{R}^n$  a point such that  $x \notin C$ . Recall that the projection  $\pi_C(x)$  is the (unique) point in  $C$  that is closest to  $x$ , i.e.,  $\|x - \pi_C(x)\| \leq \|x - z\|$  for all  $z \in C$ . Also recall that the distance from  $x$  to  $C$  is  $d_C(x) := \|x - \pi_C(x)\|$ .

- (a) Write the *variational inequality* which gives a characterization of  $\pi_C(x)$ .
- (b) Show that if  $D \supseteq C$  is a closed convex set, then  $d_D(x) \leq d_C(x)$ . Conclude that if the halfspace

$$H_{d,\alpha}^- := \{z \in \mathbb{R}^n : \langle d, z \rangle \leq \alpha\}, \quad d \neq 0,$$

contains  $C$ , then  $d_{H_{d,\alpha}^-}(x) \leq d_C(x)$ .

- (c) Show that the variational inequality in (a) implies

$$(i) C \subseteq H_{d,\alpha}^-, \quad (ii) d_C(x) = d_{H_{d,\alpha}^-}(x),$$

where  $d = x - \pi_C(x)$  and  $\alpha = \langle d, \pi_C(x) \rangle$ .

- (d) Show that all these lead to the following, geometrically appealing “duality” result: *the shortest distance from a point  $x$  to a convex set  $C$  not containing  $x$  is equal to the maximum among the distances from  $x$  to half spaces containing  $C$ .*

**Q3.** Recall that the polar of a convex set  $C \subseteq \mathbb{R}^n$  is the set

$$C^* = \{y \in \mathbb{R}^n : \langle y, x \rangle \leq 1, \quad \forall x \in C\}.$$

- (a) Show that if  $C$  is a convex body (a compact convex set) containing zero in its interior ( $0 \in C^\circ$ ), then so is  $C^*$ .
- (b) Show that if  $C$  is convex body containing zero in its interior, then  $C^{**} \subseteq C$ . (Since  $C \subseteq C^{**}$  (do not show this), this implies  $C^{**} = C$ .) *Hint:* use a separation argument.
- (c) Show that if  $P := \{x : Ax \leq b\}$  is a polyhedron, then  $P^*$  is also one, and describe  $P^*$ . *Hint:* use affine Farkas Lemma.

**Q4.** In the quadratic program

$$\min\left\{\frac{1}{2}x^T Qx + c^T x : Ax \leq b\right\},$$

$Q$  is an  $n \times n$  symmetric, positive definite matrix. Write the Lagrangian function and use it to determine the dual program. In particular, show that the dual program can be written in the form

$$\max\left\{\frac{1}{2}y^T Ry + d^T y : y \geq 0\right\},$$

where  $R$  is an  $n \times n$  symmetric matrix and  $d \in \mathbb{R}^n$ .

**Q5.** Solve, that is, find the optimal solution(s) to the constrained minimization problem

$$\begin{array}{ll} \min & \frac{x_1}{x_2} + \frac{x_2}{x_3} + \frac{x_3}{x_1} \\ \text{subject to} & x_1 \geq 0, x_2 \geq 0, x_3 \geq 0 \end{array}$$

- (a) By forming the Lagrangian and setting up the Karush–Kuhn–Tucker (KKT) conditions,
- (b) By making the substitutions  $x_i = e^{t_i}$  and then solving the unconstrained minimization problem in the variables  $t_1, t_2, t_3$ ,
- (c) By applying the arithmetic–geometric mean (AGM) inequality to the objective function in either (a) or (b). State the AGM inequality first!
- (d) The original problem is not a convex program. Yet, the optimal solution(s) you found is (are) global one(s). Why is this so, give a reason.

**Q6.** Let  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  be a continuously differentiable function. It is well-known that  $f$  is a convex function if and only if

$$f(y) \geq f(x) + \langle \nabla f(x), y - x \rangle, \quad \forall x, y \in \mathbb{R}^n. \quad (1)$$

Show that (1) is equivalent to

$$\langle \nabla f(y) - \nabla f(x), y - x \rangle \geq 0, \quad \forall x, y \in \mathbb{R}^n. \quad (2)$$

*Hint:* To prove (2) implies (1), use the mean value theorem to  $f$ , then apply (2).

**Q7.** Let  $P := \{x : Ax \leq b\}$  be a polytope (bounded polyhedron) where  $A$  is an  $m \times n$  matrix. Show that the rows  $\{a_1, \dots, a_m\}$  of  $A$  span  $\mathbb{R}^n$ . *Hint:* pick a basis  $\{c_1, \dots, c_n\}$  for  $\mathbb{R}^n$ . Formulate the duals of the linear programs  $\{\max c_i^T x : Ax \leq b\}$  and invoke the LP duality.