

# Final Examination

CS 730 - Spring 2010

Thursday, May 13, 2010, 5:05pm-7:05pm

No electronic computing devices, notes, or books allowed, except that you may bring one standard-size sheet of paper, handwritten on both sides, into the test. **Give reasoning and justify all your answers.**

The number of points is given at the start of each question. There are 75 points in total.

1. (15 points)

- (a) Given any two matrices  $A$  and  $B$  with the same number of columns, show that the following two statements are equivalent:
- I. The rows of  $A$  are linearly independent, and there is a vector  $d \neq 0$  such that  $Ad = 0$  and  $Bd > 0$ .
  - II. There is no vector pair  $(\mu, \lambda)$  (not all zero) such that  $\lambda \geq 0$  and  $A^T\mu + B^T\lambda = 0$ .
- (b) Consider the constraint system

$$c_i(x) \geq 0 \quad (i \in \mathcal{I}), \quad c_i(x) = 0 \quad (i \in \mathcal{E}),$$

at the point  $x^* \in \mathbf{R}^n$ , and let  $\mathcal{A}(x^*)$  denote the active set at  $x^*$ . One way to state the Mangasarian-Fromovitz constraint qualification (MFCQ) is: The gradients  $\nabla c_i(x^*)$ ,  $i \in \mathcal{E}$  are linearly independent and there exists a vector  $d \neq 0$  such that

$$\nabla c_i(x^*)^T d = 0 \quad \text{for all } i \in \mathcal{E}, \quad \nabla c_i(x^*)^T d > 0 \quad \text{for all } i \in \mathcal{A}(x^*) \cap \mathcal{I}.$$

Use part (a) to derive an alternative statement of MFCQ.

2. (20 points)

Consider the following nonlinear program:

$$\min_x f(x) \text{ subject to } c(x) \geq 0,$$

where  $f : \mathbf{R}^n \rightarrow \mathbf{R}$  and  $c : \mathbf{R}^n \rightarrow \mathbf{R}^m$  are smooth functions, and its  $\ell_1$ -penalized counterpart

$$\min_{(x,t)} f(x) + \mu e^T t \text{ subject to } c(x) + t \geq 0, t \geq 0,$$

where  $\mu > 0$  is a penalty parameter and  $e = (1, 1, \dots, 1)^T$ .

- (a) Write down the KKT conditions for both problems.
- (b) Suppose that  $x^*$  is a KKT point for the first problem with optimal multipliers  $\lambda^*$ . Under what condition on  $\mu$  is the same  $x^*$  together with  $t^* = 0$  a KKT point for the penalized formulation? If this condition holds, what are the optimal multipliers for the constraints  $c(x) + t \geq 0$  and  $t \geq 0$ ?
- (c) Suppose in addition to the assumptions of (b) that *strict complementarity* is satisfied for the first problem by  $x^*$  and  $\lambda^*$ . Under what additional condition on  $\mu$  will strict complementarity also be satisfied at the corresponding KKT point for the penalized formulation?

3. (15 points)

Consider the following convex quadratic programming problem:

$$\min \frac{1}{2} x^T D x + c^T x \text{ subject to } x \geq 0, e^T x = 1,$$

where  $D$  is an  $n \times n$  positive definite diagonal matrix and  $e = (1, 1, \dots, 1)^T$ .

- (a) Write down the Wolfe dual of this problem, and eliminate variables as necessary to express it in the following form:

$$\max_{\lambda, \mu} \frac{1}{2} \begin{bmatrix} \lambda \\ \mu \end{bmatrix}^T P \begin{bmatrix} \lambda \\ \mu \end{bmatrix} + t^T \begin{bmatrix} \lambda \\ \mu \end{bmatrix} \text{ subject to } \lambda \geq 0,$$

where  $\lambda \in \mathbf{R}^n$  is the vector of multipliers for the constraints  $x \geq 0$  and  $\mu$  is the (scalar) multiplier for  $e^T x = 1$ . Give explicit formulas for  $P$  and  $t$ .

- (b) Is it always possible to reformulate the dual as a problem involving  $\lambda$  alone (that is, to eliminate  $\mu$ )? Explain your answer.

4. (25 points)

- (a) Solve the following semidefinite program in the symmetric  $2 \times 2$  matrix  $X$ :

$$\min_{X \succeq 0} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \bullet X \quad \text{s.t.} \quad \begin{bmatrix} -1 & 0 \\ 0 & 0 \end{bmatrix} \bullet X = -1, \quad \begin{bmatrix} 0 & 0.5 \\ 0.5 & -1 \end{bmatrix} \bullet X = -1.$$

- (b) Write down the dual of the problem in (a), formulated in terms of two real variables (say  $y_1$  and  $y_2$ ). (You do not need to solve it.)
- (c) For the barrier function  $f : S\mathbf{R}^{n \times n} \rightarrow \mathbf{R}$  defined by  $f(X) = -\ln \det X$ , we know that the second derivative operator  $f''(X)$  is defined by

$$f''(X)UV = (X^{-1}UX^{-1}) \bullet V = \text{trace}(X^{-1}UX^{-1}V),$$

where  $U$  and  $V$  are any two matrices in  $S\mathbf{R}^{n \times n}$ . Show that

$$f''(X)UV = f''(X)VU.$$