

**CS726, Fall 2006**  
**Final Examination**

Tuesday, December 19, 2006, 10:05am-12:05pm

Answer all FOUR questions below. One handwritten sheet of notes (written front and back) is allowed. Explain all your answers.

1. The DFP quasi-Newton updating formula for the approximate Hessian  $B_k$  can be written as follows:

$$B_{k+1} = (I - \rho_k y_k s_k^T) B_k (I - \rho_k s_k y_k^T) + \rho_k y_k y_k^T,$$

where

$$\rho_k = \frac{1}{y_k^T s_k}.$$

Show that if  $B_k$  is positive definite and the curvature condition  $y_k^T s_k > 0$  holds, then  $B_{k+1}$  is also positive definite.

**Solution:** Defining  $E_k = I - \rho_k y_k s_k^T$ , the update formula can be rewritten as

$$B_{k+1} = E_k B_k E_k^T + \rho_k y_k y_k^T.$$

For any  $z \in \mathbf{R}^n$  we have

$$z^T B_{k+1} z = z^T E_k B_k E_k^T z + \rho_k z^T y_k y_k^T z = (E_k^T z)^T B_k (E_k^T z) + \rho \|y_k^T z\|_2^2.$$

The first term is nonnegative (by positive definiteness of  $B_k$ ) while the second term is also nonnegative, since  $\rho_k > 0$  by the curvature condition. Hence,  $B_{k+1}$  is at least positive semidefinite.

If  $z^T B_{k+1} z = 0$  for some nonzero  $z$ , we must have  $y_k^T z = 0$  (from the second term). But then  $E_k^T z = z$  so we have  $z^T B_{k+1} z = z^T B_k z > 0$ , a contradiction. Hence  $z^T B_{k+1} z > 0$  for all nonzero  $z$  and so  $B_{k+1}$  is positive definite.

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2. (a) Consider the problem

$$\min_x f(x) \text{ subject to } l \leq x \leq u,$$

where  $x \in \mathbf{R}^n$ ,  $f : \mathbf{R}^n \rightarrow \mathbf{R}$  is continuously differentiable, and the lower- and upper-bound vectors  $l$  and  $u$  are also in  $\mathbf{R}^n$ . Write down the first-order necessary (KKT) conditions for this problem, eliminating Lagrange multipliers to obtain a simplified form.

**Solution:** Introducing Lagrange multipliers  $\lambda$  and  $\mu$  for  $x - l \geq 0$  and  $u - x \geq 0$ , resp., we can write the Lagrangian as

$$\mathcal{L}(x, \lambda, \mu) = f(x) - \lambda^T(x - l) - \mu^T(u - x).$$

The KKT conditions are then:

$$\begin{aligned} \nabla_x \mathcal{L}(x, \lambda, \mu) &= \nabla f(x) - \lambda + \mu = 0, \\ x - l &\geq 0, \quad u - x \geq 0, \\ \lambda &\geq 0, \quad \mu \geq 0, \\ \lambda^T(x - l) &= 0, \quad \mu^T(u - x) = 0. \end{aligned}$$

Assuming that  $l < u$ , we have from the complementarity conditions that at most one of  $\lambda_i$  and  $\mu_i$  is nonzero for each  $i$ . We consider three cases:

- (i)  $l_i < x_i < u_i$ : Then  $\lambda_i = \mu_i = 0$  and  $\frac{\partial f}{\partial x_i} = 0$ ;
- (ii)  $x_i = l_i$ : Then  $\mu_i = 0$  and  $\lambda_i \geq 0$ , and we have by substituting into the first condition that  $\frac{\partial f}{\partial x_i} \geq 0$ ;
- (iii)  $x_i = u_i$ : Then  $\lambda_i = 0$  and  $\mu_i \geq 0$ , and we have by substituting into the first condition that  $\frac{\partial f}{\partial x_i} \leq 0$ .

(b) Consider the problem

$$\begin{aligned} \min_{(x_1, x_2) \in \mathbf{R}^2} & \frac{1}{2}x_1^2 + \frac{1}{2}x_2^2 - x_1x_2 - 2x_2 \\ \text{subject to} & \quad 1 - x_1 - x_2 \geq 0, \\ & \quad 1 + x_1 - x_2 \geq 0. \end{aligned}$$

Show that the KKT conditions are satisfied at  $x^* = (0, 1)$ , and determine the optimal values of the Lagrange multipliers for the two constraints.

**Solution:** We have

$$\nabla f(x) = \begin{bmatrix} x_1 - x_2 \\ x_2 - x_1 - 2 \end{bmatrix}, \quad \nabla c_1(x) = \begin{bmatrix} -1 \\ -1 \end{bmatrix}, \quad \nabla c_2(x) = \begin{bmatrix} 1 \\ -1 \end{bmatrix}.$$

At  $x^* = (0, 1)^T$ , both constraints are active. We need  $\lambda_1^* \geq 0$  and  $\lambda_2^* \geq 0$  such that

$$\nabla f(x^*) = \begin{bmatrix} -1 \\ -1 \end{bmatrix} = \lambda_1^* \begin{bmatrix} -1 \\ -1 \end{bmatrix} + \lambda_2^* \begin{bmatrix} 1 \\ -1 \end{bmatrix},$$

and it is clear by inspection that  $\lambda_1^* = 1$ ,  $\lambda_2^* = 0$  is the unique solution of this system.

3. Consider the determination of a quadratic function of two variables using function value information. That is, we seek the values of the scalars  $a_{11}$ ,  $a_{12}$ ,  $a_{22}$ ,  $b_1$ ,  $b_2$ , and  $c$  such that for the model function  $m(x)$  defined by

$$m(x) = \frac{1}{2} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}^T \begin{bmatrix} a_{11} & a_{12} \\ a_{12} & a_{22} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}^T \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + c,$$

we have  $m(y^i) = f(y^i)$ , for the chosen values  $y^i$ ,  $i = 1, 2, 3, 4, 5, 6$  and the given function  $f$ . Show that if the points  $y^i$  all lie on an ellipse satisfying

$$(y^i)^T W y^i = \gamma, \quad i = 1, 2, 3, 4, 5, 6,$$

where  $W$  is a  $2 \times 2$  positive definite matrix and  $\gamma$  is a positive scalar, then the quadratic  $m$  is not in general determined by the six points  $y^i$ ,  $i = 1, 2, 3, 4, 5, 6$ .

**Solution:**

From the equations  $m(y^i) = f(y^i)$  we obtain a  $6 \times 6$  linear system in which the unknown vector is

$$(a_{11}, a_{12}, a_{22}, b_1, b_2, c)^T$$

and the  $i$ th row of the coefficient matrix is

$$\left( \frac{1}{2}(y_1^i)^2, y_1^i y_2^i, \frac{1}{2}(y_2^i)^2, y_1^i, y_2^i, 1 \right).$$

Since the  $y^i$  all lie on the ellipse indicated, there are scalars  $w_{11}$ ,  $w_{12}$ ,  $w_{22}$ , and  $\gamma$  such that

$$w_{11}(y_1^i)^2 + 2w_{12}y_1^i y_2^i + w_{22}(y_2^i)^2 - \gamma = 0.$$

Hence, if we multiply the coefficient matrix by the vector

$$(2w_{11}, 2w_{12}, 2w_{22}, 0, 0, -\gamma),$$

we get zero. Hence, this  $6 \times 6$  matrix is nonsingular, so  $m$  is not determined in general.

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4. (a) Suppose that an algorithm for minimizing the continuously differentiable function  $f$  generates a sequence  $\{x_k\}$  lying in a bounded set  $\mathcal{B}$ , such that the sequence of gradient norms  $\{\|\nabla f(x_k)\|\}$  has an accumulation point at zero. Show that there exists an accumulation point  $x_\infty$  of  $\{x_k\}$  such that  $\nabla f(x_\infty) = 0$ .

**Solution:**

Let  $x_{k_j}, j = 1, 2, \dots$  be the subsequence for which  $\lim_{j \rightarrow \infty} \nabla f(x_{k_j}) = 0$ . Since  $x_{k_j}$  is in the bounded set  $\mathcal{B}$  for all  $j$ , it has an accumulation point. We have by taking a further subsequence if necessary that  $\lim_{j \rightarrow \infty} x_{k_j} = x_\infty$  for some  $x_\infty$  (which lies in the closure of  $\mathcal{B}$ , incidentally). By continuity of  $\nabla f$  we have that  $\nabla f(x_\infty) = 0$ .

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- (b) Suppose that an algorithm for minimizing the twice continuously differentiable function  $f$  generates a sequence  $\{x_k\}$  for which

$$\lim_{k \rightarrow \infty} \nabla f(x_k) = 0$$

and

$\nabla^2 f(x_k)$  are positive definite for all  $k$ .

Show that all accumulation points of  $\{x_k\}$  satisfy second-order necessary conditions to be a minimizer of  $f$ .

**Solution:** If  $x_\infty$  is any accumulation point, there is a subsequence  $\{x_{k_j}\}_{j=1,2,3,\dots}$  such that

$$\lim_{j \rightarrow \infty} x_{k_j} = x_\infty.$$

By the smoothness assumption we have  $\nabla f(x_\infty) = 0$  and  $\nabla^2 f(x_\infty)$  positive semidefinite, so the second-order necessary conditions for  $x_\infty$  to be a minimizer of  $f$  are satisfied.

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- (c) Can we claim that all accumulation points for the sequence in part (b) satisfy second-order *sufficient* conditions? Why or why not?

**Solution:** We cannot argue in part (b) that the limit of a sequence of positive definite matrices is positive definite — only positive semidefinite, as some of the eigenvalues may be approaching zero.

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