## MATH 450 - 558: Smooth and Nonsmooth Optimization

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Fall Semester 2015 Midterm Exam Friday November 6th, 2015 Duration: 150 minutes

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- You may use the lecture notes or your own notes, but textbooks are not allowed.

**Question 1** This question concerns a quasi-Newton method based on rank two symmetric inverse Hessian updates of the form

$$H_{k+1} = H_k + us_k^T + s_k u^T.$$
(1)

Above,  $H_{k+1}$  is an approximation for the inverse Hessian  $\left[\nabla^2 f(x^{(k+1)})\right]^{-1}$  and  $s_k := x^{(k+1)} - x^{(k)}$ .

(a) (10 points) Derive an update rule of the form (1) such that  $H_{k+1}$  satisfies the secant equation.

Choose an estimate  $x^{(0)}$  for a local minimizer. Calculate  $\nabla f(x^{(0)})$ , and choose a positive definite  $H_0$ . for k = 0, 1, 2, 3, ... do (1)  $p_k \leftarrow -H_k \cdot \nabla f(x^{(k)})$ (2) Perform a line search in the direction  $p_k$  to determine a step-length  $\alpha_k$  such that  $x^{(k)} + \alpha_k p_k$  satisfies the Wolfe conditions. (3)  $x^{(k+1)} \leftarrow x^{(k)} + \alpha_k p_k$ , and calculate  $\nabla f(x^{(k+1)})$ . (4) Form  $H_{k+1}$  from  $H_k$ ,  $s_k = x^{(k+1)} - x^{(k)}$  and  $y_k = \nabla f(x^{(k+1)}) - \nabla f(x^{(k)})$ using the update rule (1). end for

algorithm guarantee the generation of descent search directions  $p_k$ , i.e., does  $p_k$  necessarily satisfy  $f(x^{(k)} + \alpha p_k) < f(x^{(k)})$  for all  $\alpha > 0$  small enough? Explain.

**Question 2** (25 points) Suppose that  $f : \mathbb{R}^n \to \mathbb{R}$  is a twice continuously differentiable function, and  $x_*$  is a local minimizer of f such that  $\nabla^2 f(x_*)$ is *not invertible*. Furthermore, suppose that there exist a ball  $B(x_*, \delta)$  with positive radius  $\delta > 0$  such that  $\nabla^2 f(x)$  is invertible at all  $x \in B(x_*, \delta) \setminus \{x_*\}$ , and that the limit

$$\lim_{x \to x_*} \left\| \left[ \nabla^2 f(x) \right]^{-1} \right\|_2 \|x - x_*\|$$

exists and is finite.

Prove that there exists a ball  $B(x_*, \eta)$  (with positive radius  $\eta > 0$ ) such that, for all  $x^{(0)} \in B(x_*, \eta)$ , pure Newton's method for unconstrained optimization generates a sequence  $\{x^{(k)}\}$  satisfying

$$\lim_{k \to \infty} x^{(k)} = x_*.$$

Question 3 Consider the inequality constrained optimization problem

minimize<sub>$$x \in \mathbb{R}^2$$</sub>  $\frac{1}{2}(x_1 - 1)^2 - x_1 - x_2$   
subject  $x_1 - x_2^2 \ge 0$   
 $2 - x_1^2 - x_2^2 \ge 0,$  (2)

and the points  $\bar{x} = (0, 0)$  and  $\hat{x} = (1, 1)$ .

(a) (5 points) Write down a feasible sequence  $\{z^{(k)}\}$  leading to  $\hat{x}$ . Write down also a limiting direction associated with this feasible sequence.

(b) (5 points) Find the tangent cone at  $\hat{x}$ .

(c) (20 points) Show, for each one of  $\bar{x}$  and  $\hat{x}$ , either the point is a local minimizer of (2) or not a local minimizer of (2).

**Question 4** Let us focus on the following linearly constrained nonlinear program.

$$\begin{array}{ll} \text{minimize}_{x \in \mathbb{R}^n} & f(x) \\ \text{subject} & Ax = b \\ & x \ge 0 \end{array}$$
(3)

Above,  $A \in \mathbb{R}^{m \times n}$  is a given matrix,  $b \in \mathbb{R}^m$  is a given vector, and the objective function f(x) is twice continuously differentiable.

(a) (15 points) Suppose that  $\nabla^2 f(x)$  is *positive definite* at all x, and let  $x_*$  be a point such that the following hold for some  $\lambda_*$  and  $s_* \ge 0$ :

$$Ax_* = b, \ x_* \ge 0, \ x_*^T s_* = 0 \ \text{and} \ \nabla f(x_*) = A^T \lambda_* + s_*.$$

Prove that  $x_*$  is a global minimizer of (3).

(b) (10 points) Suppose that  $\text{Null}(A) \neq \{0\}$ , and that Z is a matrix whose columns form an orthonormal basis for Null(A). Furthermore, suppose  $Z^T \nabla^2 f(x) Z$  is not positive semi-definite at all x. Let  $x_*$  be a point such that the following hold for some  $\lambda_*$ :

$$Ax_* = b, \ x_* > 0 \ \text{and} \ \nabla f(x_*) = A^T \lambda_*.$$

Prove that  $x_*$  is not a local minimizer of (3).