# Problem Set 3 for EE227C (Spring 2018): Convex Optimization and Approximation

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In this homework we will prove the claims necessary for the convergence bound of the barrier method we saw in Lecture 25.

**Note:** This assignment might seem long, but that's due to lots of hints. So, don't be discouraged.

## Background

We collect some useful background material below. Feel free to consult additional resources on this topic.

**Barrier function.** Recall that we defined the modified objective  $f_{\epsilon} \colon \mathbb{R}^n \to \mathbb{R}$  as

$$f_{\epsilon}(x) = c^{\top} x - \epsilon \sum_{j=1}^{m} \ln(A_j^{\top} x - b_j).$$

Here,  $A_j \in \mathbb{R}^n$  denotes the *j*-th row of A in column form. Let  $s_j(x) = A_j^\top x - b_j$  and define the  $m \times m$  diagonal matrix  $S(x) = \operatorname{diag}(s_1(x), \dots, s_m(x))$ .

Denoting by 1 the all ones vector, we have

$$\nabla f_{\epsilon}(x) = c - \epsilon S(x)^{-1} A^{\top} \mathbf{1} = c - \epsilon \sum_{j} \frac{A_{j}}{S_{j}(x)}$$

$$\nabla^{2} f_{\epsilon}(x), = \epsilon A^{\top} S(x)^{-2} A = \epsilon \sum_{j} \frac{A_{j} A_{j}^{\top}}{S_{j}(x)^{2}},$$

We will assume that the system  $Ax \ge b$  is feasible, and that the rows of A span  $\mathbb{R}^n$ . The latter implies that  $f_{\epsilon}$  is strongly convex. We will denote by

$$x_{\epsilon}^{\star} = \arg\min_{x} f_{\epsilon}(x)$$
,

the unique minimizer for  $f_{\epsilon}$ . Further, let  $x^*$  denote a minimizer of the original constrained problem  $\min\{c^{\top}x \mid Ax \ge b\}$ .

**Newton decrement.** The central quantity  $q(x, \epsilon)$  we needed for the analysis is defined as

$$q(x,\epsilon)^2 = \nabla f_{\epsilon}(x)^{\top} \nabla^2 f_{\epsilon}(x)^{-1} \nabla f_{\epsilon}(x)$$
.

This is sometimes called *Newton decrement*.

For the remainder of this problem set, we fix  $\epsilon > 0$  and assume that x is a point satisfying Ax > b, i.e., it is in the interior of the polytope  $\{x \colon Ax \geqslant b\}$ . Further we let  $\bar{x}$  denote the outcome of taking a single Newton step starting from x,

$$\bar{x} = x - \nabla^2 f_{\epsilon}(x)^{-1} \nabla f(x)$$
.

We can also express  $q(x, \epsilon)$  as a "local norm" defined by the Hessian applied to the Newton step  $\delta = \nabla^2 f_{\epsilon}(x)^{-1} \nabla f(x)$ . That is,

$$q(x,\epsilon) = \|\delta\|_{\nabla^2 f_{\epsilon}(x)} = \sqrt{\delta^{\top} \nabla^2 f_{\epsilon}(x) \delta}.$$

## 1 Problem 1 (Optional)

In class, we stated the following proposition.

**Proposition 1.** Assume Ax > b and  $q(x, \epsilon) \le 1/2$ . Then,  $c^{\top}x - c^{\top}x^{\star} \le O(\epsilon m)$ .

(A) Prove that

$$c^{\top} x_{\epsilon}^{\star} - c^{\top} x^{\star} \leqslant \epsilon m$$
.

This claim does not need the assumptions on x.

Hint: Use the fact that the gradient of  $f_{\epsilon}$  vanishes at  $x_{\epsilon}^{\star}$  in order to get the upper bound  $\sum_{i}(s_{i}(x_{\epsilon}^{\star})-s_{i}(x^{\star}))/s_{i}(x_{\epsilon}^{\star})$ .

**(B)** Prove that

$$c^{\top}x - c^{\top}x_{\epsilon}^{\star} \leqslant O(\epsilon m)$$
.

### Problem 2

In this problem you will show that Newton's method has quadratic convergence provided that  $q(x, \epsilon)$  is small enough.

**Proposition 2.** Assume Ax > b and  $q(x, \epsilon) \le 1/2$ . Then, the Newton iterate  $\bar{x}$  satisfies  $A\bar{x} > b$  and  $q(\bar{x}, \epsilon) \le q(x, \epsilon)^2$ .

(A) Prove that

$$\nabla f_{\epsilon}(\bar{x}) = -\epsilon \sum_{j=1}^{m} \frac{A_{j}(A_{j}^{\top}(\bar{x}-x))^{2}}{s_{j}(x)^{2}s_{j}(\bar{x})}.$$

Use the fact that  $\bar{x}$  minimizes the second order approximation of  $f_{\varepsilon}$  at x and hence  $\nabla f_{\varepsilon}(x) + \nabla^2 f_{\varepsilon}(x)(\bar{x} - x) = 0$ . Write out what this condition means and use it to derive the expression for the gradient at  $\bar{x}$ .

**(B)** Show that

$$q(x,\epsilon)^2 = \epsilon \sum_{j=1}^m \frac{(A_j^\top (\bar{x} - x))^2}{s_j(x)^2}$$

**(C)** Find a norm  $\|\cdot\|$  of the form  $\|x\| = \|Mx\|_2$ , for some matrix M, for which you can show for every vector z,

$$z^{\top} \nabla f_{\epsilon}(\bar{x}) \leqslant ||z|| \cdot q(x, \epsilon)^{2}.$$

Use the previous steps and Cauchy-Schwartz.

**(D)** Complete the proof by relating  $q(\bar{x}, \epsilon)$  and  $\sup_{z} \frac{z^{\top} \nabla f_{\epsilon}(\bar{x})}{\|z\|}$ .

## Problem 3

Your goal is to prove the following proposition.

**Proposition 3.** Assume Ax > b and  $q(x, \epsilon) \le 1/2$ . Then, for  $\bar{\epsilon} = \epsilon/(1+\delta)$  with  $\delta = \frac{1}{4}n^{-1/2}$ , we have  $q(\bar{x}, \bar{\epsilon}) \le 1/2$ .

(A) Show that

$$\nabla^2 f_{\bar{\epsilon}}(\bar{x})^{-1} \nabla f_{\bar{\epsilon}}(\bar{x}) = \nabla^2 f_{\epsilon}(\bar{x})^{-1} \nabla f_{\epsilon}(\bar{x}) + \delta \nabla^2 f_{\epsilon}(\bar{x})^{-1} c.$$

**(B)** From the previous step, conclude that

$$q(\bar{x},\bar{\epsilon}) \leq q(\bar{x},\epsilon) + \delta \|\nabla^2 f_{\epsilon}(\bar{x})^{-1} c\|_{\nabla^2 f_{\epsilon}(\bar{x})}$$

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**(C)** Show that

$$\|\nabla^2 f_{\epsilon}(\bar{x})^{-1} c\|_{\nabla^2 f_{\epsilon}(\bar{x})} \leqslant \sqrt{m}.$$

Hint: Relate the expression in the left hand side to a projection operator applied to the all ones vector.

**(D)** Complete the proof by putting together the previous steps and applying Proposition 2.