

## Homework 5

Please answer the following questions in complete sentences in a typed manuscript and submit the solution to me in class on February 21th, 2012.

The previous homework was focused heavily on implementations and studying these methods in a practical setting. In this homework, we will return our focus back to theory.

### Problem 0: List your collaborators.

Please identify anyone, whether or not they are in the class, with whom you discussed your homework.

### Problem 1: Solving the trust region problem (Griva, Nash, and Sofer, 11.6.6)

In class, your professor mentioned that any solution to the trust region subproblem,

$$\begin{aligned} \text{minimize}_{\mathbf{p}} \quad & f(\mathbf{x}_k) + g(\mathbf{x}_k)^T \mathbf{p} + 1/2 \mathbf{p}^T \mathbf{H}(\mathbf{x}_k) \mathbf{p} = m_k(\mathbf{p}) \\ \text{subject to} \quad & \|\mathbf{p}\| \leq \Delta, \end{aligned}$$

could be characterized as follows:

- $(\mathbf{H}(\mathbf{x}_k) + \lambda \mathbf{I}) \mathbf{p}^* = -\mathbf{g}$
- $\lambda(\|\mathbf{p}^*\| - \Delta) = 0$
- $\mathbf{H}(\mathbf{x}_k) + \lambda \mathbf{I} \succeq 0$  (it's positive semi-definite).

Use these conditions to show that, if  $\lambda \neq 0$ , then  $\mathbf{p}$  solves:

$$\begin{aligned} \text{minimize}_{\mathbf{p}} \quad & f(\mathbf{x}_k) + g(\mathbf{x}_k)^T \mathbf{p} + 1/2 \mathbf{p}^T \mathbf{H}(\mathbf{x}_k) \mathbf{p} \\ \text{subject to} \quad & \|\mathbf{p}\| = \Delta. \end{aligned}$$

*Hint* First show that  $m_k(\mathbf{p}^*) \leq m_k(\mathbf{p}) + 1/2\lambda(\mathbf{p}^T \mathbf{p} - \mathbf{p}^{*T} \mathbf{p}^*)$  for any  $\mathbf{p}$ .

### Problem 2: Advantages and disadvantages

Is a dog-leg method or a subspace minimization method better when using a trust region method on a convex optimization problem? Discuss and explain why.

### Problem 3: Read and prove

Read the proof of Lemma 4.2. There is one missing step. Let  $\mathbf{B}$  be symmetric positive definite. Prove the inequality:

$$\frac{(\mathbf{g}^T \mathbf{g})^2}{\mathbf{g}^T \mathbf{B} \mathbf{g} \mathbf{g}^T \mathbf{B}^{-1} \mathbf{g}} \leq 1.$$

(Hint: Use Cauchy-Schwarz!) Can you remove any of the conditions from this proof? When do you have equality?