

Advanced Optimization Homework 3

Autumn 1400 Due date: Dey 3th

Analytical Questions

- 1. Show by example that in the Manifold Suboptimization, if several constraints $\mathbf{a}_{\mathbf{j}}^{\mathsf{T}}\mathbf{x} \leq \mathbf{b}_{\mathbf{j}}$ with negative values $\mu_{\mathbf{j}}$ are simultaneously dropped, then the corresponding vector $\bar{\mathbf{p}}_{\mathbf{k}}$ need not to be a feasible direction.
- 2. Let $\mathbf{H}^{\mathbf{k}}$ be a positive definite matrix and define $\bar{\mathbf{x}}^{\mathbf{k}}$ by

$$\mathbf{\bar{x}^k} = \arg\min_{\mathbf{x} \in \mathbf{X}} \bigg\{ \nabla \mathbf{f} \left(\mathbf{x^k} \right)^\intercal \left(\mathbf{x} - \mathbf{x^k} \right) + \frac{1}{2\mathbf{s^k}} \left(\mathbf{x} - \mathbf{x^k} \right)^\intercal \mathbf{H^k} \left(\mathbf{x} - \mathbf{x^k} \right) \bigg\}.$$

Show that $\bar{\mathbf{x}}^{\mathbf{k}}$ solves the problem

$$\begin{aligned} & \min \quad \frac{1}{2} \left\| \mathbf{x} - \left(\mathbf{x}^k - \mathbf{s}^k \left(\mathbf{H}^k \right)^{-1} \nabla f \left(\mathbf{x}^k \right) \right) \right\|_{\mathbf{H}^k}^2 \\ & \text{s.t.} \quad \mathbf{x} \in \mathbf{X}. \end{aligned}$$

where $\|.\|_{\mathbf{H}^{\mathbf{k}}}$ is the norm defined by $\|\mathbf{z}\|_{\mathbf{H}^{\mathbf{k}}}^{2} = \mathbf{z}^{\intercal}\mathbf{H}^{\mathbf{k}}\mathbf{z}$.

Computer Questions

1. Solve the below 3-dimensional quadratic problem, using Manifold Suboptimization method for the constraint sets a and b, and once again by using two-metric projection method for the last constraint set c. Start from the initial point $\mathbf{x}^0 = [0 \ 0 \ 1]^\intercal$.

$$\min \ x_1^2 + 2x_2^2 + 3x_3^2$$

- (a) $x_1 \ge 0$, $x_2 \ge 0$, $x_3 \ge 0$, $x_1 + x_2 + x_3 \ge 1$
- $\mathrm{(b)} \ \ x_1 \geq 0, \ x_2 \geq 0, \ x_3 \geq 0, \ x_1 + x_2 + x_3 \geq 1, \ x_1 \leq 0.5$
- (c) $\mathbf{x_1} \ge \mathbf{0}, \ \mathbf{x_2} \ge \mathbf{0}, \ \mathbf{x_3} \ge \mathbf{0}$
- 2. Minimize the following function,

$$f\left(x_{1}, x_{2}\right) = \left(x_{2} - 0.129x_{1}^{2} + 1.6x_{1} - 6\right)^{2} + 6.07\cos\left(x_{1}\right) + 10,$$

with the Trust-Region method. Use $\mathbf{x^0} = [6\ 14]^\intercal$ as starting point and $\Delta_0 = \mathbf{2},\ \hat{\Delta} = \mathbf{5}$ as Trust-Region radius constants. Solve the sub-problems using Cauchy point and Dog-Leg methods then compare the difference in performance. Report the results thoroughly.

3. As you know from the previous homework, the auto-encoder network consists of an encoder and a decoder with the following cost function:

$$\mathbf{f}\left(\theta\right) = \sum_{\mathbf{q}=1}^{\mathbf{Q}} \|\mathbf{x}_{\mathbf{q}} - \mathbf{D}\left(\mathbf{E}(\mathbf{x}_{\mathbf{q}})\right)\|^{2}$$

where

$$\mathbf{E}(\mathbf{x}) = \frac{1}{1 + \exp(-\mathbf{w^T}\mathbf{x} - \mathbf{b_e})} \text{ and } \mathbf{D}(\mathbf{x}) = \mathbf{w^T}\mathbf{E}(\mathbf{x}) + \mathbf{b_d}$$

Implement the Levenberg-Marquardt method to minimize the cost function of the auto-encoder network for the MNIST database considering what follows:

- Use the Marquardt suggestion for adaptive value of λ .
- ullet To reduce the amount of computation required, use only the train and test data associated with the digits of ullet and ullet.
- As implied from the equations, parameters of the problem are $\theta = \{\mathbf{w}, \mathbf{b_e}, \mathbf{b_d}\}$, with the corresponding dimensions $\theta = \{\mathbf{w}: \mathbf{1} \times \mathbf{784}, \mathbf{b_e}: \mathbf{1} \times \mathbf{1}, \mathbf{b_d}: \mathbf{784} \times \mathbf{1}\}$.
- Consider zero as the initial value for all parameters.
- The CCR criteria for test data is shown below:

$$\mathbf{y}_{\mathbf{q}}^{'} = \begin{cases} +1 & \mathbf{w}^{\mathbf{T}}\mathbf{x}_{\mathbf{q}} + \mathbf{b}_{\mathbf{e}} \geq \mathbf{0} \\ -1 & \mathbf{w}^{\mathbf{T}}\mathbf{x}_{\mathbf{q}} + \mathbf{b}_{\mathbf{e}} < \mathbf{0} \end{cases}$$

$$\mathbf{CCR} = \frac{\sum_{\mathbf{q}=\mathbf{1}}^{\mathbf{Q}} \|\mathbf{y}_{\mathbf{q}} - \mathbf{y}_{\mathbf{q}}^{'}\|}{\mathbf{2Q}}$$

- The report must contain the following elements:
 - A brief report of the modeling procedure.
 - Diagram of cost function value per iteration for the train data.
 - Diagram of λ values per iteration
 - Value of $\it CCR$ criteria on test data.