

Numerical Optimization - Sheet 12

If you are a student in mathematics please solve the exercises with no tag and the ones with the tag **Mathematics**. If you are a data science student please solve the problems with no tag and those with the tag **Data Science**. Submissions with tags other than your subject count as bonus points. The tag **Programming** marks programming exercises.

Ex 1 Programming

(5 Points)

Consider the problem

$$\begin{aligned} \min & x_1^2 + 4x_1x_2 + 5x_2^2 - 10x_1 - 20x_2 \\ \text{s.t.} & 2 - x_1 - x_2 = 0. \end{aligned}$$

- (i) Implement the quadratic penalty algorithm as described in Section 6.1.1 in the script. Use `scipy.optimize.minimize` to solve the unconstrained sub-problems.
- (ii) Test it on the above problem with `tol` = 10^{-7} , $\mu = (10, 10^3, 10^5, 10^7)$.
 - Plot the error of the solution against the number of steps (i.e. the index k of μ_k) in the penalty method.
 - Plot the condition of the inverse hessian in the solution against μ_k . You can use `np.linalg.cond`, and you might want to use a log-log-scaling.

Hint: The result of `scipy.optimize.minimize` contains the solution and the hessian inverse (or its approximation) which you can use to obtain a condition number. You find the details in the documentation.

Ex 2

(3 Points)

Let $z \in \mathbb{R}^2$, and $\lambda \in \mathbb{R}$. Consider the problem

$$\begin{bmatrix} 1 & 0 & 1 \\ 0 & 0 & -1 \\ 1 & -1 & 0 \end{bmatrix} \begin{pmatrix} z_1 \\ z_2 \\ \lambda \end{pmatrix} = \begin{pmatrix} 1 - x_1 \\ 1 \\ 0 \end{pmatrix}.$$

Recover the quadratic problem which is solved by the linear system.

Solution 2:

$$\begin{aligned} \min_{z \in \mathbb{R}^2} & \frac{1}{2} z^\top \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} z - \begin{pmatrix} 1 - x_1 \\ 1 \end{pmatrix}^\top z \\ \text{s.t.} & [1, -1] z = 0 \end{aligned}$$

Ex 3

(2 Points)

Assume $f, g: \mathbb{R}^n \rightarrow \mathbb{R}$ are sufficiently smooth functions, and $A, B \in \mathbb{R}^{m \times n}$. Construct the KKT-system of the quadratic subproblem of the SQP-method corresponding to the following nonlinear problems.

(i)

$$\begin{aligned} \min_{\theta, z} \quad & \frac{1}{2} z^2 \\ \text{s.t.} \quad & \\ & z = g(\theta) \end{aligned}$$

(ii)

$$\begin{aligned} \min_{\alpha, \beta} \quad & f(\alpha) + g(\beta) \\ \text{s.t.} \quad & \\ & A\alpha + B\beta = 0 \end{aligned}$$

Hint: The relevant material is presented in lecture 20 and 21.

Solution 3:

The KKT-System is given by

$$\begin{bmatrix} \text{Hess} f(x^k) & C(x^k)^\top \\ C(x^k) & 0 \end{bmatrix} \begin{pmatrix} \Delta x^k \\ \lambda^{k+1} \end{pmatrix} = \begin{pmatrix} -\nabla f(x^k) \\ -c(x^k) \end{pmatrix}$$

(i)

$$\mathcal{L}(z, \theta, \lambda) = \frac{1}{2} z^2 - \lambda(z - g(\theta))$$

$$\begin{bmatrix} 1 & 0 & 1 \\ 0 & \lambda \text{Hess} g(\theta^k) & -\nabla(g(\theta^k)) \\ 1 & -\nabla(g(\theta^k))^\top & 0 \end{bmatrix} \begin{pmatrix} \Delta z^k \\ \Delta \theta^k \\ \lambda^{k+1} \end{pmatrix} = \begin{pmatrix} -z^k \\ 0 \\ -z^k + g(\theta^k) \end{pmatrix}$$

(ii)

$$\mathcal{L}(z, \theta, \lambda) = f(\alpha) + g(\beta) - \lambda^\top (A\alpha + B\beta)$$

$$\begin{bmatrix} \text{Hess} f(\alpha^k) & 0 & A \\ 0 & \text{Hess} g(\beta^k) & B \\ A & B & 0 \end{bmatrix} \begin{pmatrix} \Delta \alpha^k \\ \Delta \beta^k \\ \lambda^{k+1} \end{pmatrix} = \begin{pmatrix} -\nabla f(\alpha^k) \\ -\nabla g(\beta^k) \\ -A\alpha^k - B\beta^k \end{pmatrix}$$