



- 1 Consider the constrained optimisation problem

$$x + y \rightarrow \min \quad \text{subject to } x^2 + y^2 \leq 1.$$

Formulate a logarithmic barrier method for the solution of this constrained optimisation problem and compute its solution for each parameter $\mu > 0$ in the barrier functional.

- 2 Assume that $A \in \mathbb{R}^{m \times n}$ with $m < n$ is a matrix of full rank and that $b \in \mathbb{R}^m \setminus \{0\}$. Consider the optimisation problem

$$\frac{1}{2} \|x\|^2 \rightarrow \min \quad \text{subject to} \quad Ax = b. \quad (1)$$

(See also problem 3 in exercise set 7.)

- Formulate the augmented Lagrangian \mathcal{L}_A for problem (1), and find for all possible parameters $\lambda \in \mathbb{R}^m$ and $\mu > 0$ a formula for the global solutions of the resulting unconstrained optimisation problem.
- For which parameters $\lambda \in \mathbb{R}^m$ and $\mu > 0$ is the minimiser of the augmented Lagrangian equal to the solution of (1)?
- An iterative algorithm for the solution of (1) using the augmented Lagrangian may have the form

$$\begin{aligned} x^{k+1} &\in \arg \min_x \mathcal{L}_A(x, \lambda^k; \mu), \\ \lambda^{k+1} &= \lambda^k - \mu(Ax^{k+1} - b). \end{aligned}$$

Show that this iteration converges for all initial values $x^0 \in \mathbb{R}^n$, $\lambda^0 \in \mathbb{R}^m$, and all $\mu > 0$ to the unique solution of (1).¹

Hint: Interpret the iteration for the Lagrange parameter λ as a fixed-point iteration—by using the explicit formula for x^{k+1} derived in the first part of this exercise—and then use results from previous numerics courses to show that this fixed-point scheme converges.

Useful matrix formula: if B , C , and $B + C$ are invertible matrices of the same size, then

$$(B^{-1} + C^{-1})^{-1} = B(B + C)^{-1}C = C(B + C)^{-1}B.$$

¹This problem does not completely fall within the curriculum of this optimisation class, but it is still recommended to *try* to solve it.