- Newton's method.
- Newton step.
- Newton decrement.
- Damped Newton phase.
- Quadratically convergent phase.
- Self-concordant functions.
- Conjugate gradient method.
- Projected gradient methods.
- Penalty methods.

## 49.14 Problems

**Problem 49.1.** Consider the function  $J: \mathbb{R}^n \to \mathbb{R}$  given by

$$J(v) = \frac{1}{2} \langle Av, v \rangle - \langle b, v \rangle + g(v),$$

where A is a real  $n \times n$  symmetric positive definite matrix,  $b \in \mathbb{R}^n$ , and  $g \colon \mathbb{R}^n \to \mathbb{R}$  is a continuous (not necessarily differentiable) convex function such that  $g(v) \geq 0$  for all  $v \in \mathbb{R}^n$ , and let U be a nonempty, bounded, closed, convex subset of  $\mathbb{R}^n$ .

(1) Prove that there is a unique element  $u \in U$  such that

$$J(u) = \inf_{v \in U} J(v).$$

*Hint*. Prove that J is strictly convex on  $\mathbb{R}^n$ .

(2) Check that

$$J(v) - J(u) = \langle Au - b, v - u \rangle + g(v) - g(u) + \frac{1}{2} \langle A(v - u), v - u \rangle.$$

Prove that an element  $u \in U$  minimizes J in U iff

$$\langle Au - b, v - u \rangle + g(v) - g(u) \ge 0$$
 for all  $v \in U$ .

**Problem 49.2.** Consider n piecewise  $C^1$  functions  $\varphi_i : [0,1] \to \mathbb{R}$  and assume that these functions are linearly independent and that

$$\sum_{i=1}^{n} \varphi_i(x) = 1 \quad \text{for all } x \in [0, 1].$$