

# 1 BASICS OF SET CONSTRAINED AND UNCONSTRAINED OPTIMIZATION

## 6.1 Consider the problem

$$\begin{aligned} &\text{minimize } f(\mathbf{x}) \\ &\text{subject to } \mathbf{x} \in \Omega, \end{aligned}$$

where  $f \in C^2$ . For each of the following specifications for  $\Omega$ ,  $\mathbf{x}^*$  and  $f$ , determine if the given point  $\mathbf{x}^*$  is: (i) definitely a local minimizer; (ii) definitely not a local minimizer; or (iii) possibly a local minimizer.

- $f: \mathbb{R}^2 \rightarrow \mathbb{R}$ ,  $\Omega = \{\mathbf{x} = [x_1, x_2]^\top : x_1 \geq 1\}$ ,  $\mathbf{x}^* = [1, 2]^\top$ , and gradient  $\nabla f(\mathbf{x}^*) = [1, 1]^\top$ .
- $f: \mathbb{R}^2 \rightarrow \mathbb{R}$ ,  $\Omega = \{\mathbf{x} = [x_1, x_2]^\top : x_1 \geq 1, x_2 \geq 2\}$ ,  $\mathbf{x}^* = [1, 2]^\top$ , and gradient  $\nabla f(\mathbf{x}^*) = [1, 0]^\top$ .
- $f: \mathbb{R}^2 \rightarrow \mathbb{R}$ ,  $\Omega = \{\mathbf{x} = [x_1, x_2]^\top : x_1 \geq 0, x_2 \geq 0\}$ ,  $\mathbf{x}^* = [1, 2]^\top$ , gradient  $\nabla f(\mathbf{x}^*) = [0, 0]^\top$ , and Hessian  $\mathbf{F}(\mathbf{x}^*) = \mathbf{I}$ .
- $f: \mathbb{R}^2 \rightarrow \mathbb{R}$ ,  $\Omega = \{\mathbf{x} = [x_1, x_2]^\top : x_1 \geq 1, x_2 \geq 2\}$ ,  $\mathbf{x}^* = [1, 2]^\top$ , gradient  $\nabla f(\mathbf{x}^*) = [1, 0]^\top$ , and Hessian  $\mathbf{F}(\mathbf{x}^*) = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$ .

## 6.9 Consider the following function:

$$f(x_1, x_2) = x_1^2 x_2 + x_2^3 x_1.$$

- In what direction does the function  $f$  decrease most rapidly at the point  $\mathbf{x}^{(0)} = [2, 1]^\top$ ?
- What is the rate of increase of  $f$  at the point  $\mathbf{x}^{(0)}$  in the direction of maximum decrease of  $f$ ?
- Find the rate of increase of  $f$  at the point  $\mathbf{x}^{(0)}$  in the direction  $\mathbf{d} = [3, 4]^\top$ .

## 6.11 Consider the problem

$$\begin{aligned} &\text{minimize } -x_2^2 \\ &\text{subject to } |x_2| \leq x_1^2 \\ &\quad x_1 \geq 0, \end{aligned}$$

where  $x_1, x_2 \in \mathbb{R}$ .

- Does the point  $[x_1, x_2]^\top = \mathbf{0}$  satisfy the first-order necessary condition for a minimizer? That is, if  $f$  is the objective function, is it true that  $\mathbf{d}^\top \nabla f(\mathbf{0}) \geq 0$  for all feasible directions  $\mathbf{d}$  at  $\mathbf{0}$ ?
- Is the point  $[x_1, x_2]^\top = \mathbf{0}$  a local minimizer, a strict local minimizer, a local maximizer, a strict local maximizer, or none of the above?

## 6.23 Let $f: \mathbb{R}^2 \rightarrow \mathbb{R}$ be defined by

$$f(\mathbf{x}) = (x_1 - x_2)^4 + x_1^2 - x_2^2 - 2x_1 + 2x_2 + 1,$$

where  $\mathbf{x} = [x_1, x_2]^\top$ . Suppose that we wish to minimize  $f$  over  $\mathbb{R}^2$ . Find all points satisfying the FONC. Do these points satisfy the SONC?