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# Chapter 1

## Smooth Convex Optimization

### 1.1 Minimization of Smooth Functions

#### 1.1.1 Smooth Convex Functions

In this section, we consider the unconstrained minimization problem

$$\min_{\mathbf{x} \in \mathbb{R}^n} f(\mathbf{x}), \tag{1.1.1}$$

where the objective function  $f(\cdot)$  is smooth enough.  $\mathcal{F}$  represents *differentiable functions*.

**Assumption 1.1.1**

For any  $f \in \mathcal{F}$ , the first-order optimality condition is sufficient for a point to be a global solution to 1.1.1

**Assumption 1.1.2**

If  $f_1, f_2 \in \mathcal{F}$  and  $\alpha, \beta \geq 0$ , then  $\alpha f_1 + \beta f_2 \in \mathcal{F}$ .

**Assumption 1.1.3**

Any linear function  $l(\mathbf{x}) = \alpha + \langle \cdot, \mathbf{x} \rangle$