



Conditions for an Extremum in Metric Spaces*

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Abstract. General necessary and sufficient conditions of the k -th order (where $k \geq 0$) for an extremum of an arbitrary function defined on an arbitrary metric space are stated. Examples illustrating the theory are described.

Key words: Metric space; Local (global) minimum and maximum; k -th order necessary optimality condition; k -th order sufficient condition; k -th order rate of steepest descent and ascent

1. Introduction

Let X be a metric space with the metric ρ and let us assume that a functional f is defined on X and takes values from $\overline{\mathbb{R}} = [-\infty, +\infty]$. Our aim is to describe k -th order necessary and sufficient conditions for a maximum and a minimum of f on X . It turns out that such conditions can be formulated in a very general form by means of the k -th order rates of steepest descent and ascent introduced in the paper. Being applied to specific spaces (normed spaces, for example) the obtained conditions generate, among others, some well-known optimality conditions (see, e.g., [1–6]).

In Section 2 the first-order conditions are proved. In Section 3 the k -th order conditions are described.

2. First-order optimality conditions

Let X be a metric space with the metric ρ and let us assume that a functional f is defined on X and takes values from $\overline{\mathbb{R}} = \mathbb{R} \cup \{+\infty, -\infty\} = [-\infty, +\infty]$.

Put

$$\text{dom } f = \{x \in X \mid f(x) \in \mathbb{R}\}$$

and assume that

$$\text{dom } f \neq \emptyset. \tag{2.1}$$

Let $x \in \text{dom } f$. Denote

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$$f^{\downarrow}(x) = \liminf_{\substack{y \in X \\ y \rightarrow x}} \frac{f(y) - f(x)}{\rho(x, y)}. \quad (2.2)$$

If there exists no sequence $\{y_k\}$, such that

$$y_k \in X, y_k \neq x \forall k, \quad y_k \rightarrow x,$$

then by definition $f^{\downarrow}(x) = +\infty$. Since $x \in \text{dom } f$, then the limit in (2.2) always exists though it may be equal to $+\infty$ or $-\infty$.

The quantity $f^{\downarrow}(x)$ is called the *rate of steepest descent* of the function f at the point x .

(2.2) implies the expansion

$$f(y) = f(x) + \rho(x, y)f^{\downarrow}(x) + \underline{o}(\rho(x, y)),$$

where

$$\liminf_{y \rightarrow x} \frac{\underline{o}(\rho(x, y))}{\rho(x, y)} = 0. \quad (2.3)$$

Analogously, for $x \in \text{dom } f$ one can define the quantity

$$f^{\uparrow}(x) = \limsup_{\substack{y \in X \\ y \rightarrow x}} \frac{f(y) - f(x)}{\rho(x, y)}. \quad (2.4)$$

If there exists no sequence $\{y_k\}$, such that

$$y_k \in X, y_k \neq x \forall k, \quad y_k \rightarrow x,$$

then by definition $f^{\uparrow}(x) = -\infty$. Since $x \in \text{dom } f$, then the limit in (2.4) always exists though it may be equal to $+\infty$ or $-\infty$.

(2.4) implies the expansion

$$f(y) = f(x) + \rho(x, y)f^{\uparrow}(x) + \bar{o}(\rho(x, y)),$$

where

$$\limsup_{y \rightarrow x} \frac{\bar{o}(\rho(x, y))}{\rho(x, y)} = 0. \quad (2.5)$$

The quantity $f^{\uparrow}(x)$ is called the *rate of steepest ascent* of the function f at the point x .

Put

$$f_* = \inf_{x \in X} f(x), \quad f^* = \sup_{x \in X} f(x).$$

It follows from (2.1) that

$$f_* < +\infty, \quad f^* > -\infty.$$

If for some point $x_* \in X$ it holds that $f(x_*) = f_*$, then the point x_* is called a

minimum point (or a *global minimum point*, or a *global minimizer*) of the function f on X . Of course, it may happen that such a point x_* doesn't exist.

Put

$$A_* = \arg \min_{x \in X} f = \{x \in X \mid f(x) = f_*\}.$$

If

$$x_0 \notin \text{dom } f, \quad f(x_0) = -\infty,$$

then

$$f_* = f(x_0) = -\infty, \quad x_0 \in A_*.$$

If for points $x_1 \in X$ and $x_2 \in X$ it turns out that

$$f(x_1) = -\infty, \quad f(x_2) = -\infty,$$

then we shall assume that

$$f(x_1) = f(x_2) = f_*.$$

Analogously, if for some point $x^* \in X$ we have $f(x^*) = f^*$, then the point x^* is called a *maximum point* (or a *global maximum point*, or a *global maximizer*) of the function f on X . Of course, it may happen that such a point x^* doesn't exist.

Put

$$A^* = \arg \max_{x \in X} f = \{x \in X \mid f(x) = f^*\}.$$

If

$$x_0 \notin \text{dom } f, \quad f(x_0) = +\infty,$$

then

$$f^* = f(x_0) = +\infty, \quad x_0 \in A^*.$$

If for points $x_1, x_2 \in X$ it turns out that

$$f(x_1) = +\infty, \quad f(x_2) = +\infty,$$

then we assume that $f(x_1) = f(x_2) = f^*$.

In other words, a point $x_* \in X$ is a global minimum point of f on X , if

$$f(x_*) \leq f(x) \quad \forall x \in X, \tag{2.6}$$

and a point $x^* \in X$ is a global maximum point of f on X , if

$$f(x^*) \geq f(x) \quad \forall x \in X. \tag{2.7}$$

A point $x_* \in X$ is called a *strict global minimum point* or a *strict global minimizer* of the function f on X , if

$$f(x_*) < f(x) \quad \forall x \in X, \quad x \neq x_*. \tag{2.8}$$

A point $x^* \in X$ is called a *strict global maximum point* or a *strict global maximizer* of the function f on X , if

$$f(x^*) > f(x) \quad \forall x \in X, \quad x \neq x^*. \quad (2.9)$$

A point $x_* \in X$ is called a *local minimum point* or a *local minimizer* of f on X , if there exists a $\delta > 0$, such that

$$f(x_*) \leq f(x) \quad \forall x \in X : \rho(x, x_*) < \delta. \quad (2.10)$$

If $\delta = +\infty$ then the point x_* is a *global minimum point*. A point $x_* \in X$ is called a *strict local minimum point* or a *strict local minimizer* if there exists a $\delta > 0$, such that

$$f(x_*) < f(x) \quad \forall x \in X : x \neq x_*, \quad \rho(x, x_*) < \delta. \quad (2.11)$$

A point $x^* \in X$ is called a *local maximum point* or a *local maximizer* of f on X , if there exists a $\delta > 0$, such that

$$f(x^*) \geq f(x) \quad \forall x \in X : \rho(x, x^*) < \delta. \quad (2.12)$$

If $\delta = \infty$ then the point x^* is a *global maximum point*. A point $x^* \in X$ is called a *strict local maximum point* or a *strict local maximizer* if there exists a $\delta > 0$, such that

$$f(x^*) > f(x) \quad \forall x \in X : x \neq x^*, \quad \rho(x, x^*) < \delta. \quad (2.13)$$

If for some point $\bar{x} \in X$ we have $f(\bar{x}) = +\infty$, then by definition \bar{x} is a global maximum point of the function f on X ; and if $f(\bar{x}) = -\infty$, then by definition \bar{x} is a global minimum point of f on X .

THEOREM 2.1. *For a point $x_* \in \text{dom } f$ to be a global or local minimizer of the function f on X it is necessary that*

$$f^\downarrow(x_*) \geq 0. \quad (2.14)$$

If

$$f^\downarrow(x_*) > 0, \quad (2.15)$$

then the point x_* is a *strict local minimizer* of f on X .

Proof. **Necessity** follows directly from the definition. Indeed, let x_* be a local or global minimizer. Then (2.10) holds, therefore

$$f^\downarrow(x_*) = \liminf_{\substack{x \in X \\ x \rightarrow x_*}} \frac{f(x) - f(x_*)}{\rho(x, x_*)} \geq 0.$$

Sufficiency. Let condition (2.15) be satisfied at the point x_* . We have to show that a $\delta > 0$ exists such that (2.11) holds. Assume the contrary. Let us choose a sequence $\{\delta_k\}$ such that $\delta_k \downarrow 0$. By assumption, the point x_* is not a strict local minimizer, therefore there exists an $x_k \in X$, such that

$$f(x_k) \leq f(x_*), \quad \rho(x_k, x_*) \leq \delta_k.$$

Hence,

$$f^\downarrow(x_k) = \liminf_{\substack{x \in X \\ x \rightarrow x_*}} \frac{f(x) - f(x_*)}{\rho(x, x_*)} \leq \liminf_{k \rightarrow \infty} \frac{f(x_k) - f(x_*)}{\rho(x_k, x_*)} \leq 0,$$

which contradicts (2.15). The sufficiency is proved. \square

THEOREM 2.2. For a point $x^* \in \text{dom } f$ to be a global or local maximizer of the function f on X , it is necessary that

$$f^\uparrow(x^*) \leq 0. \tag{2.16}$$

If

$$f^\uparrow(x^*) < 0. \tag{2.17}$$

then the point x^* is a strict local maximizer of f on X .

Proof is similar to that of Theorem 2.1. \square

DEFINITION 2.1. A point $x_* \in X$, satisfying condition (2.14), is called an *inf-stationary point* of the function f on X . A point $x^* \in X$, satisfying condition (2.16), is called a *sup-stationary point* of f on X .

DEFINITION 2.2 A sequence $\{x_k\}$, such that

$$x_k \in X, \quad f(x_k) \rightarrow f_* = \inf_{x \in X} f(x),$$

is called a *minimizing sequence* (for the function f on X).

A sequence $\{x_k\}$, such that

$$x_k \in X, \quad f(x_k) \rightarrow f^* = \sup_{x \in X} f(x),$$

is called a *maximizing sequence* (for the function f on X).

3. k -th order conditions

Let $x \in \text{dom } f$, $k \in 0 : \infty$. Put

$$f_k^\downarrow(x) = \liminf_{\substack{y \in X \\ y \rightarrow x}} \frac{f(y) - f(x)}{\rho^k(x, y)}. \tag{3.1}$$

The quantity $f_k^\downarrow(x)$ is called the *k -th order rate of steepest descent*.

Analogously, for $x \in \text{dom } f$ we define the quantity

$$f_k^\uparrow(x) = \limsup_{\substack{y \in X \\ y \rightarrow x}} \frac{f(y) - f(x)}{\rho^k(x, y)}. \tag{3.2}$$

The quantity $f_k^\uparrow(x)$, is called the k -th order rate of steepest ascent. It is clear, that

$$f^\uparrow(x) = f_1^\downarrow(x), \quad f^\downarrow(x) = f_1^\uparrow(x).$$

If $f_k^\downarrow(x) \in \mathbb{R}$ (i.e. it is finite), then (3.1) yields the expansion

$$f(y) = f(x) + \rho^k(x, y)f^\downarrow(x) + \underline{a}(\rho(x, y)), \quad (3.3)$$

where

$$\liminf_{y \rightarrow x} \frac{\bar{a}(\rho(x, y))}{\rho^k(x, y)} = 0. \quad (3.4)$$

Analogously, if $f_k^\uparrow(x) \in \mathbb{R}$, then (3.2) implies the expansion

$$f(y) = f(x) + \rho^k(x, y)f^\uparrow(x) + \bar{a}(\rho(x, y)), \quad (3.5)$$

where

$$\limsup_{y \rightarrow x} \frac{\bar{a}(\rho(x, y))}{\rho^k(x, y)} = 0. \quad (3.6)$$

The following conditions for an extremum hold.

THEOREM 3.1. *For a point $x_* \in \text{dom } f$ to be a global or local minimizer of the function f on X it is necessary that*

$$f_k^\downarrow(x_*) \geq 0 \forall k \in 0 : \infty. \quad (3.7)$$

If for some $k \in 0 : \infty$ it turns out that

$$f_k^\downarrow(x_*) > 0, \quad (3.8)$$

then x_ is a strict local minimizer of f on X .*

Proof is similar to that of Theorem 2.1. □

THEOREM 3.2. *For a point $x^* \in \text{dom } f$ to be a global or local maximizer of the function f on X it is necessary that*

$$f_k^\uparrow(x^*) \leq 0 \forall k \in 0 : \infty. \quad (3.9)$$

If for some $k \in 0 : \infty$ it turns out that

$$f_k^\uparrow(x^*) < 0, \quad (3.10)$$

then x^ is a strict local maximizer of f on X .*

Proof is similar to that of Theorem 2.2. □

DEFINITION 3.1. We say that a point x_0 is an *inf-stationary point of the k -th order* of the function f , if

$$f_i^\downarrow(x_0) = 0 \forall i \in 0 : k .$$

A function f is called *lower semicontinuous* at a point x_0 , if

$$\liminf_{x \rightarrow x_0} f(x) = f(x_0) .$$

Clearly, if a function f is lower semicontinuous at a point x_0 , then the point x_0 is an inf-stationary point of the zero order.

DEFINITION 3.2. We say that a point x_0 is a *sup-stationary point of the k -th order* of the function f , if

$$f_i^\uparrow(x_0) = 0 \forall i \in 0 : k .$$

A function f is called *upper semicontinuous* at a point x_0 , if

$$\limsup_{x \rightarrow x_0} f(x) = f(x_0) .$$

Clearly, if a function f is upper semicontinuous at a point x_0 , then the point x_0 is a sup-stationary point of the zero order.

REMARK 3.1. It is not difficult to see that the following property holds.

For a function f to be continuous at a point x_0 , it is necessary and sufficient that

$$f_0^\downarrow(x_0) = f_0^\uparrow(x_0) = 0 .$$

In other words, *a function f is continuous at a point x_0 if and only if it is both upper and lower semicontinuous at this point.*

REMARK 3.2. Theorems 3.1 and 3.2 imply the following property: At any point $x \in \text{dom } f$ either

$$f_k^\downarrow(x) \geq 0 \forall k \in 0 : \infty ,$$

or

$$f_k^\downarrow(x) \leq 0 \forall k \in 0 : \infty .$$

If for some $k \in 0 : \infty$ we have $f_k^\downarrow(x) > 0$, then the point x is a strict local minimizer.

If for some $k \in 0 : \infty$ it turns out that $f_k^\downarrow(x) < 0$, then the point x is not a local minimizer.

Analogously, at any point $x \in \text{dom } f$ either

$$f_k^\uparrow(x) \leq 0 \forall k \in 0 : \infty ,$$

or

$$f_k^\uparrow(x) \geq 0 \forall k \in 0 : \infty .$$

If for some $k \in 0 : \infty$ it turns out that $f_k^\uparrow(x) < 0$, then the point x is a strict local maximizer.

If for some $k \in 0 : \infty$ we have $f_k^\uparrow(x) > 0$, then the point x is not a local maximizer. And only in the case where

$$f_k^\downarrow(x) = 0 \forall k \in 0 : \infty$$

or

$$f_k^\uparrow(x) = 0 \forall k \in 0 : \infty,$$

we are unable to make any conclusion whether the point x is an extremum point, or not. The following examples demonstrate that in such cases any situation is possible.

In the examples below $X = \mathbb{R}$, $\rho(x, y) = |x - y|$, $x_0 = 0$.

EXAMPLE 1. Let

$$f(x) = \begin{cases} x^2, & x \neq 0, \\ 1, & x = 0 \end{cases}$$

It is clear that

$$f_0^\downarrow(x_0) = -1, \quad f_0^\uparrow(x_0) = -1, \quad f_k^\downarrow(x_0) = f_k^\uparrow(x_0) = -\infty \forall k \in 1 : \infty.$$

Thus, the sufficient condition for a maximum holds at the point $x_0 = 0$.

EXAMPLE 2. Let

$$f(x) = \begin{cases} e^{-1/|x|}, & x \neq 0, \\ 0, & x = 0. \end{cases}$$

It is easy to find that

$$f_k^\downarrow(x_0) = f_k^\uparrow(x_0) = 0 \forall k \in 0 : \infty.$$

This is just the case where we are unable to get any conclusion on the extremality of the point x_0 (though in fact this point is a minimizer).

EXAMPLE 3. Let

$$f(x) = \begin{cases} -e^{-1/|x|}, & x \neq 0, \\ 0, & x = 0. \end{cases}$$

Like in Example 2, it is easy to find that

$$f_k^\downarrow(x_0) = f_k^\uparrow(x_0) = 0 \forall k \in 0 : \infty.$$

And again by means of Theorems 3.1 and 3.2 we are unable to make any conclusion on the extremality of the point x_0 (though in fact this point is a local maximizer).

EXAMPLE 4. Let

$$f(x) = \begin{cases} e^{-1/x} & x > 0, \\ -e^{1/x}, & x < 0, \\ 0, & x = 0. \end{cases}$$

We have

$$f_k^\downarrow(x_0) = f_k^\uparrow(x_0) = 0 \forall k \in 0 : \infty.$$

Like in Examples 2 and 3, by means of Theorems 3.1 and 3.2 we are unable to make any conclusion on the extremality of the point x_0 (though in fact this point is neither a local maximizer, nor a local minimizer).

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