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ON THE MINIMAX PROBLEM

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Abstract

Full Text

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MATHEMATICS

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ON THE MINIMAX PROBLEM

(Presented by Academician L. V. Kantorovich, 16 XII 1968)

§ 1. Let the function $f(X, Y)$ be continuously differentiable with respect to X on $E_n \times \Omega$, where $\Omega \subset E_m$, and let the intersection of any compact set with Ω be closed. Put

$$\varphi(X) = \sup_{Y \in \Omega} f(X, Y), \quad R_\varepsilon(X) = \{Y \mid \varphi(X) - f(X, Y) \leq \varepsilon\}.$$

Suppose that, for some set $\Omega_1 \subset E_n$, there exists $\bar{\varepsilon} > 0$ such that the set $\bigcup_{X \in \Omega_1} R_{\bar{\varepsilon}}(X)$ is bounded (and then it is also closed). In this case, for every $X \in \Omega_1$ there exists a bounded closed set $R(X) = R_0(X) = \{Y \mid \varphi(X) = f(X, Y)\}$. Then

$$\varphi(X) = \max_{Y \in \Omega} f(X, Y).$$

As shown in (1-3), the function φ is differentiable in any direction g ($\|g\| < \infty$) at every interior point of Ω_1 , and

$$\frac{\partial \varphi(X)}{\partial g} = \lim_{\alpha \rightarrow 0+} \frac{\varphi(X + \alpha g) - \varphi(X)}{\alpha} = \max_{Y \in R(X)} \left(\frac{\partial f(X, Y)}{\partial X}, g \right). \quad (1)$$

From (1) it immediately follows that

Theorem 1. *In order that at a point X (let $X \in \Omega_1$) the function φ attain its minimum value on E_n , it is necessary (and, in the case of convexity of φ , also sufficient) that*

$$\Psi(X) = \min_{g \in G} \partial \varphi(X) / \partial g = 0, \quad (2)$$

where $G \subset E_n$ is a compact set containing the origin of coordinates as an interior point.

In (2), the unit sphere is taken as G .

A point satisfying (2) is called a stationary point of the function φ on E_n . If for some G at the point X it turns out that $\Psi(X) = 0$, then this will also be so for any other G .

Since the function $\Psi(X)$, generally speaking, is discontinuous, condition (2) cannot be used directly for constructing algorithms for minimizing the function φ . For example, the method of steepest descent, constructed using the necessary condition (2), does not, in general, lead to a stationary point. Therefore let us consider the function

$$\Psi_\varepsilon(X) = \min_{\|g\| \leq 1} \max_{Y \in R_\varepsilon(X)} (\partial f(X, Y) / \partial X, g) \quad (3)$$

(we shall consider only $\varepsilon \in [0, \bar{\varepsilon}]$).

Theorem 2. *In order that at a point $X \in \Omega_1$ the function φ attain its minimum value on E_n , it is necessary (and, in the case of convexity of φ , also sufficient) that*

$$D(X) = \inf_{\varepsilon \in [0, \bar{\varepsilon}]} h(\varepsilon) \Psi_\varepsilon(X), \quad (4)$$

where $h(\varepsilon)$ is any continuous strictly increasing function on $[0, \bar{\varepsilon}]$, with $h(0) = 0$. Conditions (2) and (4) are equivalent: (4) follows from (2), and (2) follows from (4).

The function $D(X)$ is continuous; therefore it can be used to construct methods of successive approximations for minimizing φ .

Consider the following method of successive approximations. As the first approximation choose any point $X_1 \in E_n$ such that the set $M(X_1) = \{X / \varphi(X) \leq \varphi(X_1)\} \subset \Omega_1$ is bounded. Suppose X_k has already been found and $X_k \in M(X_1)$. If $D(X_k) = 0$, then the point X_k is stationary, and the process of searching for a minimum stops. If, however, $D(X_k) < 0$ (positive D cannot occur), then find any $\varepsilon_k \in [0, \bar{\varepsilon}]$ such that

$$h(\varepsilon_k) \Psi_{\varepsilon_k}(X_k) \leq \frac{1}{s_1} D(X_k),$$

where $s_1 > 1$ is independent of k .

Let the vector $g_k \in G$ be such that

$$\max_{Y \in R_{\varepsilon_k}(X_k)} \left(\frac{\partial f(X_k, Y)}{\partial X}, g_k \right) \leq \frac{1}{s_2} \Psi_{\varepsilon_k}(X_k),$$

where $s_2 > 1$ is also independent of k . Clearly, $\|g_k\| > 0$. Now consider the ray

$$X_{k\alpha} = X_k - \alpha g_k, \quad \alpha \geq 0,$$

and find $\alpha_k \in [0, \infty)$ such that

$$\varphi(X_k) - \varphi(X_{k\alpha_k}) \geq [\varphi(X_k) - a_k]/s_3,$$

where

$$a_k = \min_{\alpha \in [0, \infty)} \varphi(X_{k\alpha}), \quad s_3 > 1$$

and s_3 is independent of k . Put $X_{k+1} = X_{k\alpha_k}$. It is clear that if $D(X_k) < 0$, then

$$\varphi(X_{k+1}) < \varphi(X_k), \quad X_{k+1} \in M(X_1).$$

Thus we construct a sequence $\{X_k\}$. If this sequence consists of a finite number of points, then its rightmost endpoint is a stationary point. If, on the other hand, this sequence consists of an infinite number of terms, then the following holds.

Theorem 3. *Every limit point of the sequence $\{X_k\}$ is a stationary point of the function φ on E_n .*

Remark 1. Applying the method set out above to the problem of mathematical programming and choosing the corresponding function $h(\varepsilon)$ and set G , one can obtain algorithms close to the algorithms presented in ^(4,5).

§ 2. Let the function $f(X, Y)$ be continuously differentiable with respect to X on $\Omega_1 \times \Omega_2$, where Ω_1 is a convex compact subset of E_n , and Ω_2 is a compact subset of the space E_m . The function φ and the set $R_\varepsilon(X)$ are defined as above. It is required to find

$$\min_{X \in \Omega_1} \varphi(X).$$

Theorem 4. *In order that the function φ attain at a point $X \in \Omega_1$ its minimum value on Ω_1 , it is necessary (and, in the case of convexity of φ , also sufficient) that*

$$\Psi_1(X) = \min_{Z \in \Omega_1} \max_{Y \in R(X)} (\partial f(X, Y)/\partial X, Z - X) = 0. \quad (5)$$

A point $X \in \Omega_1$ satisfying (5) will be called a **stationary point** of the function φ on the set Ω_1 .

Since the function $\Psi_1(X)$ is not continuous, let us introduce into consideration the function

$$\Psi_{1\varepsilon}(X) = \min_{Z \in \Omega_1} \max_{Y \in R_\varepsilon(X)} (\partial f(X, Y) / \partial X, Z - X), \quad \varepsilon > 0$$

and formulate the necessary condition in the following form.

Theorem 5. In order that at the point $X \in \Omega_1$ the function φ attain its minimum on φ_1 , it is necessary (and, in the case of convexity of φ , also sufficient) that

$$D_1(X) = \inf_{\varepsilon \in [0, \bar{\varepsilon}]} h(\varepsilon) \Psi_{1\varepsilon}(X) = 0. \quad (6)$$

The function $h(\varepsilon)$ in (6) satisfies the conditions formulated in Theorem 2, and $\bar{\varepsilon} > 0$ is any fixed number. Conditions (5) and (6) are equivalent. The function $D_1(X)$ is continuous on Ω_1 and therefore can be used for the development of successive-approximation methods. Let us describe one such possible algorithm.

As a first approximation an arbitrary $X_1 \in \Omega_1$ is chosen. Suppose that $X_k \in \Omega_1$ has already been found. If $D_1(X_k) = 0$, then the point X_k is stationary, and the process stops. If, however, $D_1(X_k) < 0$, then we proceed as follows.

Find $\varepsilon_k \in [0, \bar{\varepsilon}]$ such that

$$h(\varepsilon_k) \Psi_{1\varepsilon_k}(X_k) \leq \frac{1}{s_1} D_1(X_k)$$

and a point $z_k \in \Omega_1$

$$\max_{Y \in R_{\varepsilon_k}(X_k)} (\partial f(X_k, Y) / \partial X, Z_k - Y_k) \leq \frac{1}{s_2} \Psi_{1\varepsilon_k}(X_k).$$

Now consider the segment

$$X_{k\alpha} = \alpha X_k + (1 - \alpha) Z_k; \quad \alpha \in [0, 1]; \quad X_{k\alpha} \in \Omega_1$$

and find $\alpha_k \in [0, 1]$ such that

$$\varphi(X_k) - \varphi(X_{k\alpha_k}) \geq [\varphi(X_k) - a_k] / s_3, \quad a_k = \min_{\alpha \in [0, 1]} \varphi(X_{k\alpha}).$$

Above, s_1, s_2, s_3 are fixed, independent of k , and $s_i > 1$ ($i = 1, 2, 3$). Put $X_{k+1} = X_{k\alpha_k}$. Clearly, $X_{k+1} \in \Omega_1$ and $\varphi(X_{k+1}) < \varphi(X_k)$ if $D_1(X_k) < 0$. We then continue analogously. If the sequence $\{X_k\}$ constructed in this way consists

of a finite number of points, then its rightmost point is stationary; otherwise the following is true.

Theorem 6. Any limit point of the sequence $\{X_k\}$ constructed by the method described above is a stationary point of the function φ on the set Ω_1 .

Remark 2. The algorithms described above may be applied to solve a number of practically important problems, for example, optimal-control problems in the presence of constraints on phase coordinates.

§ 3. The algorithms constructed above are first-order methods, since they use only the concept of the first directional derivative of the function φ . It may be useful to find the second directional derivative of φ in the direction g .

Let the function $f(X, Y)$ be given and twice continuously differentiable with respect to X and Y on the open bounded set $S_1 \times S_2$, where $S_1 \subset E_n$, $S_2 \subset E_m$. Form the function

$$\varphi(X) = \max_{Y \in \Omega} f(X, Y),$$

where Ω is a closed bounded set, $\Omega \subset S_2$. The function φ is defined on S_1 . Fix $g \in E_n$. Introduce into consideration the sets $R(X)$ and

$$R_2(X, g) = \{Y \mid Y \in R(X), (\partial f(X, Y)/\partial X, g) = \max_{Z \in R(X)} (\partial f(X, Z)/\partial X, g)\}. \quad (7)$$

We shall call a vector $V \in E_m$ an admissible direction in the broad sense at the point $Y \in \Omega$ if, for every $\varepsilon > 0$, there exist a point $V_\varepsilon \in S_\varepsilon(V)$ and a number $\alpha_\varepsilon \in (0, \varepsilon]$ such that $Y + \alpha_\varepsilon V_\varepsilon \in \Omega$. Here $S_\varepsilon(V) = \{Z \mid Z \in E_m, \|Z - V\| \leq \varepsilon\}$. We denote by $M(Y)$ the cone of admissible directions in the broad sense. It is closed.

Suppose that, for all $Y \in R(X)$,

$$\max_{V \in Q(X, Y)} \left(\frac{\partial^2 f(X, Y)}{\partial Y^2} V, V \right) \leq -m(X) \|V\|^2, \quad (8)$$

where $m(X) > 0$, $Q(X, Y) = \{V \mid V \in M(Y), (V, \partial f(X, Y)/\partial Y) = 0\}$.

Then the following is true.

Theorem 7. The function φ is twice differentiable with respect to the direction g at the point X , and

$$\partial^2 \varphi(X)/\partial g^2 = \max_{Y \in R_2(X, g)} \max_{V \in Q(X, Y)} \left\{ \left(\frac{\partial^2 f(X, Y)}{\partial X^2} g, g \right) + \right.$$

$$+ 2 \left(\frac{\partial^2 f(X, Y)}{\partial Y \partial X} g, V \right) + \left(\frac{\partial^2 f(X, Y)}{\partial Y^2} V, V \right) \}. \quad (9)$$

Thus,

$$\varphi(X + \alpha g) = \varphi(X) + \alpha \partial \varphi(X) / \partial g + \frac{1}{2} \alpha^2 \partial^2 \varphi(X) / \partial g^2 + O_g(\alpha^2). \quad (10)$$

In particular, if the point $Y \in R(X)$ is an interior point of Ω , then $Q(X, Y) = E_m$, and condition (8) means strict negative definiteness of the matrix $A(X, Y) = \partial^2 f(X, Y) / \partial Y^2$. In this case there exists an inverse matrix $A^{-1}(X, Y)$, and for this Y the inner maximum in (9) is attained at

$$V = - \left(\frac{\partial^2 f(X, Y)}{\partial Y^2} \right)^{-1} \left(\frac{\partial^2 f(X, Y)}{\partial Y, \partial X} g \right).$$

If, on the other hand, $Y \in R(X)$ is an isolated point of the set Ω , then $Q(X, Y) = \{0\}$.

Remark 3. Condition (8) is not very restrictive, since, by virtue of the fact that $Y \in R(X)$ (i.e., $f(X, Y) = \max_{Z \in \Omega} f(X, Z)$), one must have

$$\max_{V \in Q(X, Y)} (A(X, Y)V, V) \leq 0.$$

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REFERENCES

1. V. F. Dem'yanov, *Vestn. Leningrad. Univ.*, No. 7, 21 (1966).
2. V. F. Dem'yanov, *Kibernetika*, No. 6, 58 (1966); No. 3, 62 (1967).
3. B. N. Pshenichnyi, *Kibernetika*, No. 6, 54 (1967).
4. G. Zoutendijk, *Methods of Feasible Directions*, IL, 1963.
5. S. I. Zukhovitskii, R. A. Polyak, M. E. Primak, *DAN*, 163, No. 2, 282 (1965).

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