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Abstract

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MATHEMATICS

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AN ALGORITHM FOR SOLVING THE PROBLEM OF CHEBYSHEV APPROXIMATION IN HILBERT SPACE

(Presented by Academician N. N. Bogolyubov, 28 V 1964)

1. By the problem of Chebyshev approximation in a Hilbert space we shall, following ⁽¹⁻³⁾, understand the following.

Let an operator-function $A(q)$ be defined on a compact set Q , which for each $q \in Q$ is a closed linear operator acting from the Hilbert space H_1 into the Hilbert space H_2 , with a domain of definition D common to all $A(q)$ and dense in H_1 ; moreover, for each fixed $x \in D$ the function $A(q)x$, with values in H_2 , is continuous on Q . Let, further, $f(q)$ be a function continuous on Q with values in H_2 . One seeks a vector $x^* \in D$ such that

$$\max_{q \in Q} \|A(q)x^* - f(q)\| = \inf_{x \in D} \max_{q \in Q} \|A(q)x - f(q)\|.$$

It is clear that, in order to construct a numerical solution of the problem, one may consider it not on the whole compact set Q , but on an ε -net $\{q_1, \dots, q_n\}$ of this compact set, for sufficiently small $\varepsilon > 0$. Then, denoting

$$A(q_i) = A_i, \quad f(q_i) = f_i \quad (i = 1, \dots, n),$$

we arrive at the problem of Chebyshev approximation in a Hilbert space of the system

$$A_i x - f_i \quad (i = 1, \dots, n), \tag{1}$$

i.e., at the problem of finding such a vector $x^* \in D$ (the Chebyshev vector of the system (1)) for which

$$\max_i \|A_i x^* - f_i\| = \inf_{x \in D} \max_i \|A_i x - f_i\|. \quad (2)$$

If we denote by R the subspace of those vectors $x \in D$ for which $A_i x = 0$ for all $i = 1, \dots, n$, and by S the orthogonal complement to R in H_1 , then for every vector $x \in D$ we shall have the representation

$$x = x_R + x_S \quad (x_R \in R; x_S \in S),$$

$$\inf_{x \in D} \max_i \|A_i x - f_i\| = \inf_{x_S \in D \cap S} \max_i \|A_i x_S - f_i\|,$$

so that one may assume $R = 0$, $S = H_1$.

As is known ⁽³⁾, in order that for each system of vectors f_1, \dots, f_n from H_2 there exist a vector $x^* \in D$ (a Chebyshev vector) for which (2) holds, it is necessary and sufficient that the operators A_1, \dots, A_n satisfy the condition

$$\max_i \|A_i x\| \geq m \|x\| \quad \text{for all } x \in D, \quad (3)$$

where $m > 0$ is a constant.

In the case when $n = 1$, the operator A is bounded and the equation $Ax = f$ has an exact solution, a descent algorithm for finding this solution was constructed in ⁽⁴⁾*. Other algorithms for solving a similar problem are given or may be obtained from the methods described in ⁽⁵⁻⁷⁾.

* There the case is also considered where the operator A is unbounded but satisfies certain special conditions.

In the proposed paper we give an algorithm for solving problem (1)–(2) for the case when the operators A_i ($i = 1, \dots, n$) are bounded.

2. As the initial approximation take an arbitrary vector $x^{(0)} \in H_1$ and an arbitrary sufficiently small $\delta_1 > 0$. Let

$$\max_i \|A_i x^{(0)} - f_i\|^2 = \|A_{i_1} x^{(0)} - f_{i_1}\|^2;$$

$$\|A_{i_\nu} x^{(0)} - f_{i_\nu}\|^2 > \|A_{i_1} x^{(0)} - f_{i_1}\|^2 - \delta_1 \quad (\nu = 1, \dots, \nu_1),$$

$$\|A_i x^{(0)} - f_i\|^2 \leq \|A_{i_1} x^{(0)} - f_{i_1}\|^2 - \delta_1 \quad (i \neq i_1, \dots, i_{\nu_1}).$$

The set of indices $\{i_1, i_2, \dots, i_{\nu_1}\}$ for which the deviations $\|A_i x^{(0)} - f_i\|^2$ differ from the maximal ones by less than δ_1 , i.e., are “almost maximal,” will be denoted by $I(x^{(0)}, \delta_1)$.

To improve the approximation we shall move (descend) from the point $x^{(0)}$ in some direction $g^{(1)} \in H_1$, i.e., we shall increase $t > 0$ in the expression $x^{(0)} + tg^{(1)}$, choosing the descent direction $g^{(1)}$ so that all the “almost maximal” deviations decrease:

$$\|A_i x^{(0)} + tg^{(1)} - f_i\|^2, \quad i \in I(x^{(0)}, \delta_1),$$

i.e., so that the derivatives be negative,

$$\begin{aligned} \frac{1}{2} \frac{d}{dt} \|A_i(x^{(0)} + tg^{(1)}) - f_i\|^2 \Big|_{t=0} &= (A_i g^{(1)}, A_i x^{(0)} - f_i) = \\ &= (g^{(1)}, A_i^*(A_i x^{(0)} - f_i)) = (g^{(1)}, g_i(x^{(0)})), \quad i \in I(x^{(0)}, \delta_1), \end{aligned}$$

and so that this decrease (descent) be as steep as possible. Taking into account the negativity of the derivatives, this means that the normalized direction $g^{(1)}$ must satisfy the relation

$$\max_{i \in I(x^{(0)}, \delta_1)} (g^{(1)}, g_i(x^{(0)})) = \min_{g \in H_1} \max_{i \in I(x^{(0)}, \delta_1)} (g, g_i(x^{(0)})).$$

Since the orthogonal complement to the subspace spanned by the vectors $g_i(x^{(0)})$, $i \in I(x^{(0)}, \delta_1)$, evidently plays no role in the problem under consideration, it may be assumed that

$$g = \sum_{j \in I(x^{(0)}, \delta_1)} \xi_j g_j(x^{(0)}).$$

Then the problem of finding the direction of steepest descent from $x^{(0)}$ is reduced to finding

$$\min_{\xi} \max_{i \in I(x^{(0)}, \delta_1)} \sum_{j \in I(x^{(0)}, \delta_1)} (g_j, g_i) \xi_j = \min_{\xi} \max_{i \in I(x^{(0)}, \delta_1)} \sum_{j \in I(x^{(0)}, \delta_1)} a_{ij} \xi_j \quad (4)$$

under the normalization, for example, $|\xi_j| \leq 1$, $j \in I(x^{(0)}, \delta_1)$.

The last problem is the usual linear programming problem: minimize

$$u = \xi_{n+1} \quad (5)$$

subject to the constraints

$$\sum_{j \in I(x^{(0)}, \delta_1)} a_{ij} \xi_j \leq \xi_{n+1}, \quad i \in I(x^{(0)}, \delta_1), \quad |\xi_j| \leq 1, \quad j \in I(x^{(0)}, \delta_1). \quad (6)$$

With the normalization

$$\|g\|^2 = \left\| \sum_{j \in I(x^{(0)}, \delta_1)} \xi_j g_j(x^{(0)}) \right\|^2 \leq 1,$$

the determination of the descent direction is reduced to solving the problem of minimizing a linear form

$$u = \xi_{n+1} \quad (5')$$

under the constraints

$$\sum_{j \in I(x^{(0)}, \delta_1)} a_{ij} \xi_j \leq \xi_{n+1}, \quad i \in I(x^{(0)}, \delta_1), \quad \left\| \sum_{j \in I(x^{(0)}, \delta_1)} \xi_j g_j(x^{(0)}) \right\|^2 \leq 1. \quad (6')$$

Denote by u_1 the minimum u under the constraints (6'), and suppose that $u_1 < -\delta_1$. For definiteness choose one of the normalizations, for example the second one.

Having thus constructed a descent direction from $x^{(0)}$, we proceed to determine the approximation step t , i.e., to find the boundary of the admissible increase of t in the formula $x = x^{(0)} + tg^{(1)}$. To this end, let us first observe that problem (1)–(2) is equivalent to the problem of minimizing the function $w = \xi$ under the constraints

$$\|A_i x - f_i\|^2 - \xi \leq 0 \quad (i = 1, \dots, n). \quad (7)$$

In the descent direction $(g^{(1)}; \eta_1)$ from the point $(x^{(0)}; \mu_1^2)$ (where $\mu_1 = \|A_{i_1} x^{(0)} - f_{i_1}\|$) we use the vector $g^{(1)}$ found above, and determine η_1 from the condition that, along the direction $(g^{(1)}; \eta_1)$, the maximum of the derivatives of the functions $w = \xi, \|A_i x - f_i\|^2 - \xi, i \in I(x^{(0)}, \delta_1)$, at the point $(x^{(0)}; \mu_1^2)$ be negative and as small as possible, i.e., that the relation

$$\max \left\{ \eta_1, \max_{i \in I(x^{(0)}, \delta_1)} [2(A_i g^{(1)}, A_i x^{(0)} - f_i) - \eta_1] \right\} =$$

$$\begin{aligned}
 &= \min_{\eta} \max \left\{ \eta, \max_{i \in I(x^{(0)}, \delta_1)} [2(A_i g^{(1)}, A_i x^{(0)} - f_i) - \eta] \right\} = \\
 &= \min_{\eta} \max \{ \eta, 2u_1 - \eta \}.
 \end{aligned}$$

It is clear that $\eta_1 = u_1$. To determine the approximation step, we increase t in the expression $(x^{(0)} + tg^{(1)}; \mu_1^2 + u_1 t) = (x; \xi)$ until we reach the boundary of the domain defined by the inequalities (7), i.e., the approximation step t_1 is the least positive root of the equations

$$\|A_i(x^{(0)} + tg^{(1)}) - f_i\|^2 - \mu_1^2 - u_1 t = 0 \quad (i = 1, \dots, n).$$

As the new approximation we take $x^{(1)} = x^{(0)} + t_1 g^{(1)}$. We regard the obtained point $x^{(1)}$ as the initial one, set $\delta_2 = \delta_1$, determine a descent direction from the point $x^{(1)}$, and so on, until for some $x^{(k)}$ and the corresponding δ_{k+1} the corresponding problem of type (5')–(6') leads to $u_{k+1} \geq -\delta_{k+1}$. Then we set $\delta_{k+2} = \delta_{k+1}/2$, and in the case $u_{k+1} < 0$ find t_{k+1} and $x^{(k+1)}$ and continue the computations, taking $x^{(k+1)}$ as the initial point and δ_{k+2} as the parameter value. In the case $u_{k+1} = 0$, in determining the descent direction we leave among the constraints (6') only those which correspond to strictly maximal deviations. If it again turns out that $\min u = u_{k+1} = 0$, then the point $x^{(k)}$ is a solution of problem (1)–(2). If, however, $u_{k+1} < 0$, then we find t_{k+1} and $x^{(k+1)}$ and continue the computations, taking $x^{(k+1)}$ as the initial point.

3. We outline a proof of convergence of the algorithm described. First note that $\delta_k \rightarrow \delta = 0$, since otherwise, for $\delta > 0$, we would have $u_k < -\delta$ beginning with some k , which is impossible, since then it would turn out that, on the one hand, $t_k \rightarrow 0$, and on the other hand, $t_k > t_0 > 0$. We now verify that

$$\inf_x \max_{1 \leq i \leq n} \|A_i x - f_i\| = \lim_{k \rightarrow \infty} \max_{1 \leq i \leq n} \|A_i x^{(k)} - f_i\| = \mu^*.$$

Indeed, let i_k be the number of the step at which the parameter δ changes. From the bounded sequence $\{x^{(i_k-1)}\}$ one can extract a weakly convergent ...

a convergent subsequence. We may assume that $x^{(i_k-1)} \xrightarrow{\text{weakly}} \tilde{x}$ and that, at the steps i_k , the descent direction is described by one and the same set I of deviation indices. The set of indices

$$I_0 = \{i \mid \|A_i \tilde{x} - f_i\| = \mu^*\} \cap I$$

is nonempty.

Suppose that the vector \tilde{x} is not a solution of problem (1)–(2), so that from the point \tilde{x} there exists a descent direction \tilde{g} . Then, as $i_k \rightarrow \infty$ and for sufficiently small $\varepsilon > 0$, we obtain

$$(A_i(\tilde{x} - x^{(i_k-1)} + \varepsilon\tilde{g}), A_i x^{(i_k-1)} - f_i) \rightarrow \varepsilon(A_i\tilde{g}, A_i x^{(0)} - f_i) \leq 0, \quad i \in I_0,$$

$$(A_i(\tilde{x} - x^{(i_k-1)} + \varepsilon\tilde{g}), A_i x^{(i_k-1)} - f_i) \rightarrow \|A_i\tilde{x} - f_i\|^2 - (\mu^*)^2 + \varepsilon(A_i\tilde{g}, A_i\tilde{x} - f_i) < 0, \quad i \in I \setminus I_0,$$

i.e., for sufficiently large k , the direction $\xi^{(k)} = \tilde{x} - x^{(i_k-1)} + \varepsilon\tilde{g}$ is a descent direction from the point $x^{(i_k-1)}$. Therefore a number $\alpha < 0$ will be found such that $u_{i_k} < \alpha$. But this is impossible, since from $\delta_{i_k} \rightarrow 0$ it follows that $u_{i_k} \rightarrow 0$.

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Note: Figure translations are in progress. See original paper for figures.

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