

A MODIFICATION OF AN OPTIMIZATION METHOD

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Abstract. *A modification of the Danilin-Pschenichnij method is presented. Under certain assumptions it is proved that the obtained sequence of points converges to a unique optimal solution to the given problem of unconstrained optimization and that the rate of convergence is superlinear.*

1. Introduction

In this paper we are concerned with the problem of unconstrained optimization

$$(1) \quad \min\{\varphi(x) \mid x \in R^n\},$$

where $\varphi: R^n \rightarrow R$ is a twice continuously differentiable function.

The aim of this paper is to present a modification of the Danilin-Pschenichnij method where the direction vectors are defined by the Danilin-Pschenichnij method (see [1] and [4] and the step α_k is defined by a new step-size algorithm.

2. Preliminaries

We begin first with a few preliminaries.

Suppose a sequence of points $\{x_k\}$ is given. We correspond to $\{x_k\}$ another sequence $\{y_k\}$ in this way:

$$(2) \quad y_k = x_k + r_k,$$

where the vectors $r_k \in R^n$ satisfy the following conditions:

1. If by D_k we denote the matrix whose columns are the vectors $r_k/\|r_k\|, \dots, r_{k-n+1}/\|r_{k-n+1}\|$, then $|\det D_k| \geq \epsilon$ for all $k \geq n-1$, where ϵ is some predetermined small positive number.

2. The vectors r_k are chosen arbitrarily except for $\|r_k\| \rightarrow 0, k \rightarrow \infty$.

We define the matrix A_k by the following system of equations:

$$(3) \quad A_k r_{k-i} = e_{k-i}, \quad i = 0, 1, \dots, n-1; \quad k \geq n-1,$$

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where $e_{k-i} = \nabla\varphi(y_{k-i}) - \nabla\varphi(x_{k-i})$ and r_k, y_k are elements of the sequence (2). The Danilin-Pschenichnij method generates the sequence of points $\{x_k\}$ of the form:

$$x_{k+1} = x_k - \alpha_k p_k, \quad k = 0, 1, \dots,$$

where $p_k = A_k^{-1} \nabla\varphi(x_k)$ for $k \geq n-1$, and the matrix A_k is defined by (3) (for $k = 0, 1, \dots, n-2$ we can put, for example $p_k = \nabla\varphi(x_k)$).

The step-size α_k is computed by the following algorithm:

Step 1. Put $\alpha = 1$ and compute $x = x_k - \alpha p_k$.

Step 2. Compute $\varphi(x) = \varphi(x_k - \alpha p_k)$.

Step 3. Check the inequality

$$\varphi(x_k) - \varphi(x) \geq \epsilon \alpha \langle \nabla\varphi(x_k), p_k \rangle, \quad 0 < \epsilon < \frac{1}{2}.$$

Step 4. If the inequality (4) is satisfied, set $\alpha_k = 1 (= \alpha)$; otherwise reduce α until the condition (4) is satisfied.

In the literature there already exist the particular cases of this method obtained for special choices of the vector r_k given by Bulanić A.P. and Danilin J.M. (see [1] and [2]).

We will also need the following lemmas.

Lemma 1. (see [4]) *Let $\{x_k\}$ be a bounded sequence of points, $\|x_{k+1} - x_k\| \rightarrow 0, k \rightarrow \infty$ and let for every $k \geq n-1$ the matrix A_k be defined by (3). Then $\lim_{k \rightarrow \infty} \|A_k - H(x_k)\| = 0$, where by $H(x)$ we denote the Hessian of the function φ at the point x .*

Lemma 2. (see [4]) *Let $\varphi: R^n \rightarrow R$ be a twice continuously differentiable function such that there exist constants m and $M, 0 < M < \infty$ with the property*

$$(5) \quad m\|y\|^2 \leq \langle y, H(x)y \rangle \leq M\|y\|^2 \quad \text{for any } x, y \in R^n$$

and let $\{x_k\}$ be a sequence such that $\varphi(x_{k+1}) \leq \varphi(x_k)$ and $\langle \nabla\varphi(x_k), x_{k+1} - x_k \rangle \rightarrow 0, k \rightarrow \infty$. Then $\|x_{k+1} - x_k\| \rightarrow 0, k \rightarrow \infty$.

Lemma 3. (see [3]) *Let $\varphi: R^n \rightarrow R$ be a twice continuously differentiable function such that there exists a constant $m, 0 < m < \infty$ with the property*

$$\langle y, H(x)y \rangle \geq m\|y\|^2 \quad \text{for any } x, y \in R^n,$$

where $H(x)$ is the Hessian of the function φ at the point x . Then for any point $x_0 \in R^n$ the set $L = \{x \in R^n | \varphi(x) \leq \varphi(x_0)\}$ is convex and bounded and

$$m\|y\|^2 \leq \langle y, H(x)y \rangle \leq M\|y\|^2 \quad \text{for any } x \in L, y \in R^n, 0 < m < M < \infty.$$

Proposition 1. (see [4]) *Let the assumptions of Lemma 2 be valid and the matrix A_k be defined by (3). Also suppose the inequality*

$$(6) \quad \langle A_k^{-1} \nabla \varphi(x_k), \nabla \varphi(x_k) \rangle > 0$$

holds and the step α_k is computed by (4). Then, independent of the initial point x_0 the sequence $\{x_k\}$ defined by the iteration

$$x_{k+1} = x_k - \alpha_k A_k^{-1} \nabla \varphi(x_k), \quad \alpha_k > 0$$

has the following properties: 1. $\varphi(x_{k+1}) < \varphi(x_k)$; 2. $\|x_k - x^\| \rightarrow 0, k \rightarrow \infty$; 3. the rate of convergence of the sequence $\{x_k\}$ to the point x^* is superlinear.*

3. The modification of the Danilin-Pschenichnij method

Now we can present the modification of the Danilin-Pschenichnij method. The modified Danilin-Pschenichnij algorithm is an iterative algorithm for finding an optimal solution to the problem (1) generating sequences of points $\{x_k\}$ of the form:

$$(7) \quad x_{k+1} = x_k - \alpha_k p_k, \quad k = 0, 1, 2, \dots,$$

$$(8) \quad p_k = A_k^{-1} \nabla \varphi(x_k), \quad k \geq n-1,$$

where the matrix A_k is defined by (3) (for $k = 0, 1, \dots, n-2$ we can for example put $p_k = \nabla \varphi(x_k)$).

We define the step α_k by the following step-size algorithm:

Step 1. Put $\alpha = \bar{\alpha}_k = \min\left\{\frac{d(\langle \nabla \varphi(x_k), p_k \rangle)}{\|p_k\|^3}, 1\right\}$, where $d: [0, \infty) \rightarrow [0, \infty)$ is a function such that $\delta_1 t \leq d(t) \leq \delta_2 t$ for any $t \geq 0$ and some $0 < \delta_1 < \delta_2$.

Step 2. If the inequality

$$(9) \quad \varphi(x_k) - \varphi(x_k - \alpha p_k) \geq \epsilon \alpha^2 d(\langle \nabla \varphi(x_k), p_k \rangle), \quad 0 < \epsilon < \frac{1}{2}$$

is satisfied for $\alpha = \bar{\alpha}_k$, set $\alpha_k = \bar{\alpha}_k$; otherwise reduce $\bar{\alpha}_k$ until the condition (9) is satisfied.

Theorem. *Let $\varphi: R^n \rightarrow R$ be a twice continuously differentiable function with the property*

$$(10) \quad \langle H(x)y, y \rangle \geq m \|y\|^2 \quad \text{for any } x, y \in R^n, \quad 0 < m < \infty,$$

let the sequence $\{x_k\}$ be defined by the relations (7) and (8). If the condition (6) is satisfied and the step α_k is defined by the above step-size algorithm, then: 1. $x_k \in L = \{x \in R^n | \varphi(x) \leq \varphi(x_0)\}$; 2. $\varphi(x_{k+1}) < \varphi(x_k)$; 3. $\|x_k - x^\| \rightarrow 0, k \rightarrow \infty$, where x^* is the unique optimal solution to the problem (1).*

If, in addition to the above conditions $\delta_1 < \delta_2 \leq \frac{2^{\delta_1}}{2\epsilon\delta_1 + 1}$, $0 < \delta_1 < \frac{1}{2\epsilon}$, then the rate of convergence is superlinear.

PROOF. Firstly we shall prove that the sequence $\{x_k\}$ satisfies the assumptions of Lemma 1. We proceed by induction and assume that $x_k \in L$ and $\nabla\varphi(x_k) \neq 0$.

From the condition (6) it follows that

$$(11) \quad \langle \nabla\varphi(x_k), p_k \rangle = \langle \nabla\varphi(x_k), A_k^{-1} \nabla\varphi(x_k) \rangle > 0 \quad \text{for any } k \geq n-1$$

By the Taylor's theorem, we have for $\alpha > 0$:

$$\varphi(x_k) - \varphi(x_k - \alpha p_k) = \alpha \langle \nabla\varphi(x_k), p_k \rangle - \frac{\alpha^2}{2} \langle H(\xi_k) p_k, p_k \rangle =$$

$$(\xi_k = x_k - \theta_k \alpha p_k, \quad \theta_k \in (0, 1))$$

$$\alpha d(\langle \nabla\varphi(x_k), p_k \rangle) \left[\frac{\langle \nabla\varphi(x_k), p_k \rangle}{d(\langle \nabla\varphi(x_k), p_k \rangle)} - \frac{\alpha}{2} \frac{\langle H(\xi_k) p_k, p_k \rangle}{d(\langle \nabla\varphi(x_k), p_k \rangle)} \right].$$

From (11) it follows that $d(\langle \nabla\varphi(x_k), p_k \rangle) > 0$. Consequently, the inequality (9) holds if

$$\frac{\langle \nabla\varphi(x_k), p_k \rangle}{d(\langle \nabla\varphi(x_k), p_k \rangle)} - \frac{\alpha}{2} \frac{\langle H(\xi_k) p_k, p_k \rangle}{d(\langle \nabla\varphi(x_k), p_k \rangle)} \geq \alpha \epsilon$$

that is

$$(12) \quad \frac{1}{\alpha} \frac{\langle \nabla\varphi(x_k), p_k \rangle}{d(\langle \nabla\varphi(x_k), p_k \rangle)} - \frac{1}{2} \frac{\langle H(\xi_k) p_k, p_k \rangle}{d(\langle \nabla\varphi(x_k), p_k \rangle)} \geq \epsilon.$$

Since $\frac{\langle \nabla\varphi(x_k), p_k \rangle}{d(\langle \nabla\varphi(x_k), p_k \rangle)} > 0$ and by the inequality (10) $\frac{\langle H(\xi_k) p_k, p_k \rangle}{d(\langle \nabla\varphi(x_k), p_k \rangle)} > 0$, for some $\alpha = \alpha_k$ the inequality (12) will hold, and consequently the condition (9).

From the relation (9) it follows that $\varphi(x_{k+1}) < \varphi(x_k)$ and $x_{k+1} \in L = \{x \in R^n \mid \varphi(x) \leq \varphi(x_0)\}$, where L (by Lemma 3) is a compact and convex set. Since the function φ is bounded below on the compact and convex set L , it follows that $\varphi(x_{k+1}) - \varphi(x_k) \rightarrow 0$, $k \rightarrow \infty$. Hence by (9) it follows that

$$(13) \quad \alpha_k^2 d(\langle \nabla\varphi(x_k), p_k \rangle) \rightarrow 0, \quad k \rightarrow \infty.$$

Furtheron, since $\alpha_k \leq \bar{\alpha}_k \leq \frac{d(\langle \nabla\varphi(x_k), p_k \rangle)}{\|p_k\|^3}$ we have that $d(\langle \nabla\varphi(x_k), p_k \rangle) \geq \alpha_k \|p_k\|^3$.

From this inequality using (13) we obtain

$$(14) \quad \alpha_k^3 \|p_k\|^3 \leq \alpha_k^2 d(\langle \nabla\varphi(x_k), p_k \rangle) \rightarrow 0, \quad k \rightarrow \infty.$$

From (14) it follows that $\|x_{k+1} - x_k\| = \alpha_k \|p_k\| \rightarrow 0$, $k \rightarrow \infty$. Consequently, all assumptions of Lemma 1 are satisfied. Hence, we have

$$(15) \quad \|A_k - H(x_k)\| \rightarrow 0, \quad k \rightarrow \infty.$$

From (15) and Lemma 3, it follows that for any M_1 such that $M_1 \geq M$ and m satisfying $0 < m_1 \leq m$ there exist an integer k_0 such that for $k \geq k_0$ and any $y \in R^n$

$$(16) \quad m_1 \|y\| \leq \langle A_k y, y \rangle \leq M_1 \|y\|^2$$

holds.

This relation implies

$$(17) \quad m_1 \|p_k\|^2 \leq \langle A_k p_k, p_k \rangle = \langle \nabla \varphi(x_k), p_k \rangle \leq M_1 \|p_k\|^2, \quad k \geq k_0.$$

By (16) we have that for sufficiently large k

$$(18) \quad \frac{d(\langle \nabla \varphi(x_k), p_k \rangle)}{\|p_k\|^3} \geq \frac{\delta_1 \langle \nabla \varphi(x_k), p_k \rangle}{\|p_k\|^3} \geq \frac{\delta_1 m_1}{\|p_k\|}.$$

Since $\nabla \varphi$ is bounded on L , applying (16) we get

$$\|p_k\| = \|A_k^{-1} \nabla \varphi(x_k)\| \leq \frac{1}{m_1} K \quad \text{for any } k \geq n-1.$$

From (17) and (18) we obtain

$$\frac{d(\langle \nabla \varphi(x_k), p_k \rangle)}{\|p_k\|^3} \geq \frac{\delta_1 m_1^2}{K} = \bar{\alpha} > 0.$$

Since $\bar{\alpha}_k = \min\left\{\frac{d(\langle \nabla \varphi(x_k), p_k \rangle)}{\|p_k\|^3}, 1\right\}$, we have that for sufficiently large k $\bar{\alpha}_k \geq \bar{\alpha} > 0$. Furtheron, since, $\delta_1 t \leq d(t) \leq \delta_2 t$, by Lemma 3

$$m \|y\|^2 \leq \langle H(x)y, y \rangle \leq M \|y\|^2 \quad \text{for any } x \in L, y \in R^n$$

It is evident that the condition (12) is satisfied if the following is valid:

$$(19) \quad \begin{aligned} & \frac{1}{\alpha} \frac{\langle \nabla \varphi(x_k), p_k \rangle}{d(\langle \nabla \varphi(x_k), p_k \rangle)} - \frac{1}{2} \frac{\langle H(\xi_k) p_k, p_k \rangle}{d(\langle \nabla \varphi(x_k), p_k \rangle)} \geq \\ & \frac{1}{\alpha} \frac{\langle \nabla \varphi(x_k), p_k \rangle}{\delta_2 \langle \nabla \varphi(x_k), p_k \rangle} - \frac{1}{2} \frac{M \|p_k\|^2}{\delta_1 \langle \nabla \varphi(x_k), p_k \rangle} \geq \\ & \geq \frac{1}{\alpha \delta_2} - \frac{M}{2 \delta_1 m_1 \|p_k\|^2} = \frac{1}{\alpha \delta_2} - \frac{M}{2 m_1 \delta_1} \geq \epsilon. \end{aligned}$$

From (19) it follows that

$$(20) \quad 0 < \alpha \leq \frac{2\delta_1 m_1}{\delta_2(2\delta_1 m_1 \epsilon + M)}.$$

This relation implies that there exists a constant α^* , $0 < \alpha^* < \frac{2\delta_1 m_1}{\delta_2(2\delta_1 m_1 \epsilon + M)}$ such that the inequality (19) is satisfied for any k if $\alpha^* \leq \alpha_k \leq \frac{2\delta_1 m_1}{\delta_2(2\delta_1 m_1 \epsilon + M)}$.

Now, since $\alpha_k \geq \alpha^*$ and $\bar{\alpha}_k \geq \bar{\alpha}$ for sufficiently large k it follows that there exist a constant $C > 0$, such that $\alpha_k \geq C > 0$. Hence,

$$(21) \quad \|p_k\| = \frac{1}{\alpha_k} \|x_{k+1} - x_k\| \leq \frac{\|x_{k+1} - x_k\|}{C} \rightarrow 0, \quad k \rightarrow \infty.$$

From (161) and (21) we get

$$(22) \quad \|\nabla\varphi(x_k)\| = \|A_k p_k\| \leq M_1 \|p_k\| \rightarrow 0, \quad k \rightarrow \infty.$$

Since the function φ is strictly convex and $\{x_k\} \subset L$, where L is a compact convex set, applying (22) we have that the sequence $\{x_k\}$ converges to a unique optimal solution x^* to the problem (1).

Now we shall show that $\alpha_k = 1$ for sufficiently large k . The relations (18) and (21) imply the following:

$$\frac{d(\langle \nabla\varphi(x_k), p_k \rangle)}{\|p_k\|^3} \geq \frac{\delta_1 m_1}{\|p_k\|^3} \rightarrow \infty \quad k \rightarrow \infty.$$

From this relation and the definition of $\bar{\alpha}_k$ it follows that, for sufficiently large k , $\bar{\alpha}_k = 1$.

Furtheron, since the step-size α_k must satisfy the condition (12), by the assumptions of Theorem we have

$$(23) \quad \begin{aligned} & \frac{1}{\alpha_k} \frac{\langle \nabla\varphi(x_k), p_k \rangle}{d(\langle \nabla\varphi(x_k), p_k \rangle)} - \frac{1}{2} \frac{\langle H(\xi_k)p_k, p_k \rangle}{d(\langle \nabla\varphi(x_k), p_k \rangle)} \geq \dots \\ & \geq \frac{1}{\alpha_k \delta_2} - \frac{1}{2\delta_1} \frac{\langle H(\xi_k)p_k, p_k \rangle}{\langle \nabla\varphi(x_k), p_k \rangle} \geq \epsilon, \end{aligned}$$

$$\xi_k = x_k + \theta_k(x_{k+1} - x_k), \quad \theta_k \in (0, 1).$$

Since

$$\frac{\langle H(\xi_k)p_k, p_k \rangle}{\langle \nabla\varphi(x_k), p_k \rangle} = \frac{\langle [H(\xi_k) - H(x_k)]p_k, p_k \rangle}{\langle A_k p_k, p_k \rangle} + \frac{\langle H(x_k)p_k, p_k \rangle}{\langle A_k p_k, p_k \rangle},$$

by (15), (21) and uniform continuity of H on L , if $k \rightarrow \infty$ then $\frac{\langle H(\xi_k)p_k, p_k \rangle}{\langle \nabla\varphi(x_k), p_k \rangle} \rightarrow 1$.

Consequently, by (23), for sufficiently large k we obtain:

$$\begin{aligned} \frac{1}{\alpha_k} \frac{\langle \nabla \varphi(x_k), p_k \rangle}{d(\langle \nabla \varphi(x_k), p_k \rangle)} - \frac{1}{2} \frac{\langle H(\xi_k) p_k, p_k \rangle}{d(\langle \nabla \varphi(x_k), p_k \rangle)} &\geq \dots \\ &\geq \frac{1}{\alpha_k \delta_2} - \frac{1}{2\delta_1} \geq \epsilon. \end{aligned}$$

It is easy to show that the last inequality is satisfied if $\delta_1 < \delta_2 \leq \frac{2\delta_1}{1+2\epsilon\delta_1}$, $0 < \delta_1 < \frac{1}{2\epsilon}$, $\alpha_k = 1$.

It means that for sufficiently large k the inequality (9) is also satisfied for $\alpha_k = 1$. Now we can estimate the rate of convergence. We have:

$$\|x_{k+1} - x_k\|^2 = \langle x_{k+1} - x^*, x_{k+1} - x^* \rangle = \langle x_k - x^* - A_k^{-1} \nabla \varphi(x_k), x_{k+1} - x^* \rangle.$$

By the Mean Value Theorem we obtain:

$$\begin{aligned} \langle A_k^{-1} \nabla \varphi(x_k), x_{k+1} - x^* \rangle &= \langle A_k^{-1} [\nabla \varphi(x_k) - \nabla \varphi(x^*)], x_{k+1} - x^* \rangle = \\ &= \langle A_k^{-1} H(\eta_k)(x_k - x^*), x_{k+1} - x^* \rangle, \quad \eta_k = x_k + \nu_k(x_k - x^*), \nu_k \in (0, 1). \end{aligned}$$

The last two inequalities imply

$$\begin{aligned} \|x_{k+1} - x^*\|^2 &= \langle (I - A_k^{-1} H(\eta_k))(x_k - x^*), (x_{k+1} - x^*) \rangle = \\ &= \langle (A_k^{-1} (A_k - H(\eta_k)))(x_k - x^*), (x_{k+1} - x^*) \rangle \leq \\ &\leq \|A_k^{-1}\| \|A_k - H(\eta_k)\| \|x_k - x^*\| \|x_{k+1} - x^*\| \leq \\ &\leq \frac{1}{m_1} \|A_k - H(\eta_k)\| \|x_k - x^*\| \|x_{k+1} - x^*\|. \end{aligned}$$

Hence, $\|x_{k+1} - x^*\| \leq \gamma_k \|x_k - x^*\|$, where $\gamma_k = \frac{1}{m_1} \|A_k - H(\eta_k)\|$.

By Lemma 1 and uniform continuity of H on L it follows that

$$\|A_k - H(\eta_k)\| \leq \|A_k - H(x_k)\| + \|H(x_k) - H(\eta_k)\| \rightarrow 0 \quad k \rightarrow \infty.$$

Consequently, $\gamma_k = \frac{1}{m_1} \|A_k - H(\eta_k)\| \rightarrow 0$, $k \rightarrow \infty$.

The last relation implies that the sequence $\{x_k\}$ converges to a unique optimal solution x^* superlinearly.

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