

## ON A SECOND-ORDER STEP-SIZE ALGORITHM\*

Nada I. DJURANOVIĆ-MILIČIĆ

*Department of Mathematics  
Faculty of Technology and Metallurgy  
University of Belgrade, Belgrade, Yugoslavia  
nmilicic@elab.tmf.bg.ac.yu*

**Abstract:** In this paper we present a modification of the second-order step-size algorithm. This modification is based on the so called forcing functions. It is proved that this modified algorithm is well-defined. It is also proved that every point of accumulation of the sequence generated by this algorithm is a second-order point of the nonlinear programming problem. Two different convergence proofs are given having in mind two interpretations of the presented algorithm.

**Keywords:** Forcing function, step-size algorithm, second-order conditions.

### 1. INTRODUCTION

We are concerned with the following problem of the unconstrained optimization:

$$\min\{\varphi(x) \mid x \in D\} \quad (1)$$

where  $\varphi: D \subset R^n \rightarrow R$  is a twicecontinuously differentiable function on an open set  $D$ .

We consider iterative algorithms to find an optimal solution to problem (1) generating sequences of points  $\{x_k\}$  of the following form:

$$x_{k+1} = x_k + \alpha_k s_k + \beta_k d_k, \quad k = 0, 1, \dots, \quad (2)$$

$$s_k, d_k \neq 0, \quad \langle \nabla \varphi(x_k), s_k \rangle \leq 0, \quad (3)$$

and the steps  $\alpha_k$  and  $\beta_k$  are defined by a particular step-size algorithm.

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Before we present the modified algorithm, we shall define the original second-order step-size algorithm.

The original Mc Cormick-Armijo's second order step-size algorithm [4] defines  $\alpha_k$  in the following way:  $\alpha_k > 0$  is a number satisfying

$$\alpha_k = 2^{-i(k)},$$

where  $i(k)$  is the smallest integer from  $i = 0, 1, \dots$ , such that

$$x_{k+1} = x_k + 2^{-i(k)} s_k + 2^{\frac{-i(k)}{2}} d_k \in D$$

and

$$\varphi(x_k) - \varphi(x_{k+1}) \geq \gamma \left[ -\langle \nabla \varphi(x_k), s_k \rangle - \frac{1}{2} \langle H(x_k) d_k, d_k \rangle \right] 2^{-i(k)},$$

where  $0 < \gamma < 1$  is a preassigned constant,  $H(x)$  - the Hessian matrix of the function  $\varphi$  at  $x$ ,  $s_k, d_k$  - direction vectors satisfying relations (3).

We begin with the definition which we need in the following text.

**Definition** (See[5]). A mapping  $\sigma : [0, \infty) \rightarrow [0, \infty)$  is a forcing function if for any sequence  $\{t_k\} \subset [0, \infty)$

$$\lim_{k \rightarrow \infty} \sigma(t_k) = 0 \quad \text{implies} \quad \lim_{k \rightarrow \infty} t_k = 0$$

and  $\sigma(t) > 0$  for  $t > 0$ .

(The concept of the forcing function was introduced first by Elkin in [3].)

## 2. A MODIFICATION OF THE SECOND-ORDER STEP-SIZE ALGORITHM

The modified algorithm defines  $\alpha_k$  in the following way:  $\alpha_k > 0$  is a number satisfying

$$\alpha_k = q^{-i(k)}, \quad q > 1,$$

where  $i(k)$  is the smallest integer from  $i = 0, 1, \dots$ , such that

$$x_{k+1} = x_k + q^{-i(k)} s_k + q^{\frac{-i(k)}{2}} d_k \in D \tag{4}$$

and

$$\varphi(x_k) - \varphi(x_{k+1}) \geq q^{-i(k)} \left[ \sigma_1(-\langle \nabla \varphi(x_k), s_k \rangle) + \sigma_2\left(-\frac{1}{2} \langle H(x_k) d_k, d_k \rangle\right) \right] \tag{5}$$



where  $\sigma_1 : [0, \infty) \rightarrow [0, \infty)$  and  $\sigma_2 : [0, \infty) \rightarrow [0, \infty)$  are the forcing functions such that  $\delta_1 t \leq \sigma_1(t) \leq \bar{\delta}_1 t$ ,  $\delta_2 t \leq \sigma_2(t) \leq \bar{\delta}_2 t$   $0 < \delta_1 < \bar{\delta}_1 < 1$ ,  $0 < \delta_2 < \bar{\delta}_2 < 1$  and  $s_k, d_k$  are the direction vectors satisfying (3) and  $\langle H(x_k)d_k, d_k \rangle \leq 0$ .

In order to have a finite value  $i(k)$ , it is sufficient that  $s_k$  and  $d_k$  satisfy (3) and, in addition, that

$$\langle \nabla \varphi(x_k), s_k \rangle < 0 \quad \text{whenever} \quad \nabla \varphi(x_k) \neq 0 \quad (6A)$$

and

$$\langle H(x_k)d_k, d_k \rangle < 0 \quad \text{whenever} \quad \nabla \varphi(x_k) = 0. \quad (6B)$$

Now we shall prove the first convergence theorem.

**Theorem 1.** Let  $\varphi : D \subset R^n \rightarrow R$  be a twicecontinuously differentiable function on the open set  $D$ . Let the sequence  $\{x_k\}$  be defined by relations (2), (3), (4), (5), (6A) and (6B). Let  $\bar{x}$  be a point of accumulation of  $\{x_k\}$  and  $K_1$  a set of indices such that  $x_k \rightarrow \bar{x}$  for  $k \in K_1$ .

Assume that:

1. the sequences  $\{s_k\}$  and  $\{d_k\}$  are uniformly bounded;
2.  $-\langle \nabla \varphi(x_k), s_k \rangle \geq \mu_k(\|\nabla \varphi(x_k)\|)$ ,  $k \in K_1$ , where  $\mu_k : [0, \infty) \rightarrow [0, \infty)$ ,  $k \in K_1$  are forcing functions;
3. there exists a value  $\beta > 0$  such that

$$-\langle H(x_k)d_k, d_k \rangle \geq \beta \langle H(x_k)e_k^{\min}, e_k^{\min} \rangle,$$

where  $e_k^{\min}$  is an eigenvector of  $H(x_k)$  associated with its minimum eigenvalue.

Then  $\bar{x}$  is a stationary point, that is

$$\nabla \varphi(\bar{x}) = 0$$

and  $H(\bar{x})$  is a positive semidefinite matrix with at least one eigenvalue equal to zero.

**Proof:** There are two cases to consider.

- a) The integers  $\{i(k)\}$  for  $k \in K_1$  are uniformly bounded from above by some value  $I$ .

Because of the descent property it follows that all points of the accumulation have the same function value and

$$\begin{aligned} (0 \geq) \varphi(x_0) - \varphi(\bar{x}) &= \sum_{k \in K_1} [\varphi(x_k) - \varphi(x_{k+1})] \geq \\ &\geq \sum_{k \in K_1} q^{-i(k)} \left[ \sigma_1(-\langle \nabla \varphi(x_k), s_k \rangle) + \sigma_2\left(-\frac{1}{2} \langle H(x_k)d_k, d_k \rangle\right) \right] \geq \\ &\geq q^{-I} \delta \sum_{k \in K_1} \left[ -\langle \nabla \varphi(x_k), s_k \rangle - \frac{1}{2} \langle H(x_k)d_k, d_k \rangle \right], \quad (\delta = \max\{\delta_1, \delta_2\}) \end{aligned}$$



$$\geq q^{-I} \delta \sum_{k \in K_1} \left[ \mu_k(\|\nabla \varphi(x_k)\|) + \frac{1}{2} \beta \langle H(x_k) e_k^{\min}, e_k^{\min} \rangle \right].$$

Since  $\varphi(\bar{x})$  is finite and since each term in the brackets is greater than, or equal to zero for each  $k \in K_1$ , it follows that  $\mu_k(\nabla \varphi(x_k)) \rightarrow 0 \Rightarrow \|\nabla \varphi(x_k)\| \rightarrow 0$  (according to the definition of forcing functions)  $\Rightarrow \nabla \varphi(\bar{x}) = 0$  and that  $\langle H(\bar{x}) \bar{e}_{\min}, \bar{e}_{\min} \rangle = 0$ , where  $\bar{e}_{\min}$  is some accumulation point of  $\{e_k^{\min}\}$  for  $k \in K_1$ .

b) There is a subset  $K_2 \subset K_1$  such that  $\lim_{k \rightarrow \infty} i(k) = \infty$ .

Because of the definition of  $i(k)$ , then either

$$x_k + q^{-i(k)+1} s_k + q^{\frac{-i(k)+1}{2}} d_k \notin D$$

or

$$\begin{aligned} & \varphi(x_k) - \varphi \left( x_k + q^{-i(k)+1} s_k + q^{\frac{-i(k)+1}{2}} d_k \right) < \\ & < q^{-i(k)+1} \left[ \sigma_1(-\langle \nabla \varphi(x_k), s_k \rangle) + \sigma_2 \left( -\frac{1}{2} \langle H(x_k) d_k, d_k \rangle \right) \right]. \end{aligned} \quad (7)$$

If the former condition held infinitely often, then because

$$x_k + q^{-i(k)+1} s_k + q^{\frac{-i(k)+1}{2}} d_k \rightarrow \bar{x}, \quad k \in K_2,$$

it would follow that  $\bar{x}$  is on the boundary of  $D$ . Since  $D$  is an open set,  $\bar{x} \notin D$ , it contradicts the theorem hypothesis. Therefore, without the loss of generality (7) can be considered to hold for all  $k \in K_2$ .

Since  $\varphi \in C^2$ , and since the sequences  $\{s_k\}$  and  $\{d_k\}$  are assumed to be uniformly bounded, the left-hand side of inequality (7) can be written as

$$\begin{aligned} & -q^{-i(k)+1} \langle \nabla \varphi(x_k), s_k \rangle - q^{\frac{-i(k)+1}{2}} \langle \nabla \varphi(x_k), d_k \rangle - \\ & - \frac{1}{2} \left\langle H(x_k) \left( q^{-i(k)+1} s_k + q^{\frac{-i(k)+1}{2}} d_k \right), q^{-i(k)+1} s_k + q^{\frac{-i(k)+1}{2}} d_k \right\rangle - o(q^{-i(k)+1}) < \\ & < q^{-i(k)+1} \left[ \sigma_1(-\langle \nabla \varphi(x_k), s_k \rangle) + \sigma_2 \left( -\frac{1}{2} \langle H(x_k) d_k, d_k \rangle \right) \right] < \\ & < q^{-i(k)+1} \left[ -\bar{\delta}_1 \langle \nabla \varphi(x_k), s_k \rangle - \bar{\delta}_2 \cdot \frac{1}{2} \langle H(x_k) d_k, d_k \rangle \right]. \end{aligned}$$

Combining terms and incorporating a term where appropriate into  $o(q^{-i(k)+1})$  yields (using the fact that  $-\langle \nabla \varphi(x_k), s_k \rangle \geq 0$ ):



$$o(q^{-i(k)+1}) > q^{-i(k)+1} \left[ (-1 + \bar{\delta}_1) \langle \nabla \varphi(x_k), s_k \rangle - (-\bar{\delta}_2 + 1) \frac{1}{2} \langle H(x_k) d_k, d_k \rangle \right].$$

Using the theorem hypothesis 3 we obtain

$$o(q^{-i(k)+1}) > q^{-i(k)+1} \left[ (-1 + \bar{\delta}_1) \langle \nabla \varphi(x_k), s_k \rangle + (-\bar{\delta}_2 + 1) \frac{\beta}{2} \langle H(x_k) e_k^{\min}, e_k^{\min} \rangle \right].$$

Dividing by  $q^{-i(k)+1}$  yields

$$\begin{aligned} \frac{o(q^{-i(k)+1})}{q^{-i(k)+1}} &> (-1 + \bar{\delta}_1) \langle \nabla \varphi(x_k), s_k \rangle + (-\bar{\delta}_2 + 1) \frac{\beta}{2} \langle H(x_k) e_k^{\min}, e_k^{\min} \rangle \geq \\ &\geq (1 - \bar{\delta}_1) \mu_k(\|\nabla \varphi(x_k)\|) + \frac{-\bar{\delta}_2 + 1}{2} \cdot \beta \cdot \langle H(x_k) e_k^{\min}, e_k^{\min} \rangle. \end{aligned}$$

Since each term is, according to the assumptions, greater than or equal to zero, taking the limit as  $k \rightarrow \infty$  for  $k \in K_2$  yields

$$\mu_k(\|\nabla \varphi(x_k)\|) \rightarrow 0 \Rightarrow \|\nabla \varphi(x_k)\| \rightarrow 0 \Rightarrow \nabla \varphi(\bar{x}) = 0$$

and

$$\langle H(x_k) e_k^{\min}, e_k^{\min} \rangle \rightarrow \langle H(\bar{x}) \bar{e}_{\min}, \bar{e}_{\min} \rangle = 0.$$

To prove the second convergence theorem we shall follow Y. Amaya [1]. Namely, we are going to show that the trajectory

$$f(t, x_k) = x_k + t^2 s_k + t d_k \quad (8)$$

proposed by the presented algorithm (i.e. satisfying the relations (2), (3), (4), (5), (6A) and (6B)) and

$$\begin{aligned} \langle \nabla \varphi(x_k), s_k \rangle &< 0 \\ \langle \nabla \varphi(x_k), d_k \rangle &\leq 0 \end{aligned} \quad (9)$$

and

$$\langle H(x_k) d_k, d_k \rangle = 0$$

if  $H(x_k)$  is positive semidefinite, and

$$\begin{aligned} \langle \nabla \varphi(x_k), s_k \rangle &\leq 0 \\ \langle \nabla \varphi(x_k), d_k \rangle &\leq 0 \end{aligned} \quad (10)$$

and

$$\langle H(x_k) d_k, d_k \rangle < 0$$

if  $H(x_k)$  is not positive semidefinite, has the properties set out in Amaya's paper.

Firstly, we shall briefly present Amaya's algorithm [1].



Let  $\varphi : D \subset R^n \rightarrow R$  be a twicecontinuously differentiable function on the open set  $D$  (i.e.  $\varphi \in C^2$ ) which we want to minimize, and  $h : R^+ \times D \rightarrow R^n$  is a function such that, for all  $x \in D$ ,  $h(0, x) = x$ . We suppose that for every  $x \in D$ ,  $h(t, x)$  is  $C^2$  for  $t \geq 0$ .

Given  $x \in D$ , the function  $h(t, x)$  describes a trajectory in  $D \subset R^n$  originating at  $x$ . The minimizing algorithm defines a sequence  $\{x_k\}$  in the following way:

$$x_{k+1} = \begin{cases} x_k & \text{if } x_k \in M, \\ h(t_k, x_k) & \text{if } x_k \notin M, \end{cases} \quad (11)$$

where  $M = \{x \in D \mid \nabla \varphi(x) = 0 \text{ and } \langle H(x)p, p \rangle \geq 0, p \in R^n\}$ .

For  $x \in D$ , we define the  $C^2$  - class function  $f_x : R^+ \rightarrow R^n$  by

$$f_x(t) = \varphi[h(t, x)], \quad t \in R^+.$$

This function is shown to satisfy

$$\begin{aligned} f'_{x_k}(0) &= \langle \nabla \varphi(x_k), \dot{h}(0, x_k) \rangle \quad \text{and} \\ f''_{x_k}(0) &= \langle H(x_k) \dot{h}(0, x_k), \dot{h}(0, x_k) + \langle \nabla \varphi(x_k), \ddot{h}(0, x_k) \rangle \rangle, \end{aligned}$$

where  $\dot{h}$  and  $\ddot{h}$  denote respectively the first and second derivatives of  $h$  with respect to  $t$ .

The following assumptions are made:

- A1.**  $L = \{x \in D \mid \varphi(x) \leq \varphi(x_0)\}$  is bounded;
- A2.**  $f'_x(0) \leq 0$  for all  $x \notin M$ ;
- A3.** if  $x \notin M$  and  $f'_x(0) = 0$ , then  $f''_x(0) < 0$ .

Amaya in Theorem 3.1 in [1] proves the convergence of a subsequence of points of  $\{x_k\}$  defined by (11) to  $\bar{x} \in M$ , provided that  $\varphi \in C^2$  and that assumptions A1, A2, A3 hold.

Now we can present the second convergence theorem for the modified McCormick-Armijo's algorithm.

**Theorem 2.** Under assumptions A1, A2 and A3 every point of accumulation  $\bar{x}$  of the sequence  $\{x_k\}$  generated by the modified McCormick-Armijo's algorithm and additionally, satisfying (9) and (10) belongs to  $M$ , that is, the second-order necessary conditions are satisfied at  $\bar{x}$ .

**Proof:** Let us suppose that  $x_k \notin M$  for  $k = 0, 1, 2, \dots$ . From the choice of  $t_k = \alpha_k$  by relations (2), (3), (4), (5), (6A) and (6B) we have that  $f_{x_k}(t_k) \leq f_{x_k}(0)$ , i.e. the sequence  $\{\varphi(x_k)\}$  is decreasing; hence  $\{x_k\} \subset L$ . Due to the assumption A1, the sequence  $\{x_k\}$  has a point of accumulation  $\bar{x}$ .



For the trajectory (8) we have:

$$\begin{aligned} f'_{x_k}(0) &= \langle \nabla \varphi(x_k), \dot{h}(0, x_k) \rangle, \quad \dot{h}(0, x_k) = d_k, \\ f''_{x_k}(0) &= \langle H(x_k) \dot{h}(0, x_k), \dot{h}(0, x_k) + \langle \nabla \varphi(x_k), \ddot{h}(0, x_k) \rangle \rangle, \quad \ddot{h}(0, x_k) = s_k, \quad \text{i.e.} \\ f'_{x_k}(0) &= \langle \nabla \varphi(x_k), d_k \rangle, \\ f''_{x_k}(0) &= \langle H(x_k) d_k, d_k \rangle + \langle \nabla \varphi(x_k), s_k \rangle. \end{aligned}$$

From (6A) it follows that the assumption A2 holds. Let us examine the assumption A3. Assuming  $f'_{x_k}(0) = 0$ , we have two cases:

- a) if  $H(x_k)$  is positive semidefinite, by applying (9) to the relation (11), we obtain  $f''_{x_k}(0) < 0$ ;
- b) if  $H(x_k)$  is not positive semidefinite, by applying (10) to the relation (11), we obtain

$$f''_{x_k}(0) < 0.$$

Following Amaya's proof of theorem 3.1 in [1] we conclude that  $\bar{x} \in M$ .

### 3. CONCLUSION

Because of general assumptions on the objective function  $\varphi$ , the modified algorithm can be used for solving a wide class of unconstrained optimization problems. Also, the choice of forcing functions  $\sigma_1(t)$  and  $\sigma_2(t)$ , with the property  $\delta_1 t \leq \sigma_1(t) \leq \bar{\delta}_1 t$ ,  $\delta_2 t \leq \sigma_2(t) \leq \bar{\delta}_2 t$ ,  $0 < \delta_1 < \bar{\delta}_1 < 1$ ,  $0 < \delta_2 < \bar{\delta}_2 < 1$  is wide.

Finally, this modified algorithm can be used for solving constrained optimization problems (see [2]) when constraints are adequately considered.

### REFERENCES

- [1] Amaya, J., "Convergence of curvilinear search algorithms to second order points", *Revista de Matematicas Aplicadas*, 10 (1989) 71-79.
- [2] Djuranovic-Miličić, N., "An algorithm in constrained optimization", in: M. Thoma and A. Wyner (eds.), *Lecture Notes in Control and Information Sciences*, Springer-Verlag, Berlin, 1986, 203-208.
- [3] Elkin, R., "Convergence theorems for Gauss-Siedel and other minimization algorithms", Doctoral Thesis, University of Maryland, College Park, 1968.
- [4] Mc Cormick, G.P., *Nonlinear Programming, Theory, Algorithms and Applications*, Wiley, New York, 1983.
- [5] Ortega, J., and Rheinboldt, W., *Iterative Solution of Nonlinear Equations in Several Variables*, Academic Press, New York, 1970.