

## SEMI-LOCAL CONVERGENCE OF AN ALGORITHM FOR SOLVING NONLINEAR EQUATIONS

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**Abstract.** We propose a general coordinate relaxation algorithm  $x^{k+1} = x^k - \omega_k t_{i_k} e^{i_k}$  for solving the nonlinear equation  $F(x) = 0$ , where  $F : D \subset \mathbb{R}^n \rightarrow \mathbb{R}^n$  is a nonlinear mapping. Under appropriate conditions we prove a semi-local convergence theorem for this algorithm.

**Key Words and Phrases:** Local-semiconvergence, algorithm, nonlinear equation, relaxation method.

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### 1. INTRODUCTION

Consider the system of nonlinear equations

$$F(x) = 0, \quad x \in D \tag{1.1}$$

where  $F : D \subset \mathbb{R}^n \rightarrow \mathbb{R}^n$  is a nonlinear mapping and  $D$  is a nonempty subset of  $\mathbb{R}^n$ . Write

$$F(x) = \begin{bmatrix} f_1(x_1, x_2, \dots, x_n) \\ f_2(x_1, x_2, \dots, x_n) \\ \vdots \\ f_n(x_1, x_2, \dots, x_n) \end{bmatrix}.$$

Equation (1.1) is rewritten as

$$\begin{cases} f_1(x_1, x_2, \dots, x_n) = 0, \\ f_2(x_1, x_2, \dots, x_n) = 0, \\ \vdots \\ f_n(x_1, x_2, \dots, x_n) = 0, \end{cases} \quad x \in D.$$

In this paper we discuss the convergence for Eq. (1.1) of the following general coordinate relaxation method

$$x^{k+1} = x^k - \omega_k t_{i_k} e^{i_k}, \quad k = 0, 1, \dots \quad (1.2)$$

where  $t_{i_k}$  is a step-size of the iteration,  $\omega_k$  is a relaxation parameter, and  $i_k \in \{1, 2, \dots, n\}$ .

If  $F : D \subset \mathbb{R}^n \rightarrow \mathbb{R}^n$  is a gradient mapping (i.e., there is a function  $g : D \subset \mathbb{R}^n \rightarrow \mathbb{R}$  such that  $F(x)^T = g'(x)$  for  $x \in D$ ) and if  $F$  is a continuously differentiable and uniformly monotone or strictly monotone mapping on  $D$ , Schechter [5] and respectively, Brewster-Kannan [2] proved the global convergence of the nonlinear SOR-Newton method

$$x^{k+1} = x^k - \omega_k \frac{f_{i_k}(x^k)}{f_{i_k i_k}(x^k)} e^{i_k}, \quad k = 0, 1, \dots$$

where  $f_{ij}(x)$  denotes the  $(i, j)$ -entry of the Jacobi matrix of  $F'(x)$ .

Consider now the case where  $F$  is a nongradient mapping. In [1], Bers proved that if  $F'(x)$  is uniformly diagonally dominant, then the Gauss-Seidel method

$$\begin{cases} x^{k+1} = x^k - \alpha_k e^{i_k} \\ \text{where } \alpha_k \text{ is such that } f_{i_k}(x^k - \alpha_k e^{i_k}) = 0 \end{cases}$$

has global convergence.

While Gipser [3] proved the global convergence of the following method

$$\begin{cases} x^{k+1} = x^k + t_k e^{i_k} \\ \text{where } t_k \text{ is such that } f_{i_k}(x^{k+1}) = c_k f_{i_k}(x^k), \quad c_k \in [0, 1 - \varepsilon], \quad \varepsilon > 0, \end{cases}$$

where Gipser assumed that there exists a nonsingular diagonal matrix  $S$  such that either  $F'(x)S$  is uniformly row diagonally dominant or  $SF'(x)$  is uniformly column diagonally dominant.

In this paper we prove the semi-local convergence of the algorithm (1.2) in which we consider the following iteration order:

(I)  $i_k$  satisfies the equation

$$\max_{1 \leq j \leq n} |F(x^k)^T F'(x^k) e^j| = |F(x^k)^T F'(x^k) e^{i_k}|. \quad (1.3)$$

Our step-size  $t_{i_k}$  satisfies the condition

$$\begin{aligned} \left( \frac{1-\beta}{1-\delta} \right) \frac{|F(x^k)^T F'(x^k) e^{i_k}|}{r_{i_k}} &\leq t_{i_k} \operatorname{sgn}(F(x^k)^T F'(x^k) e^{i_k}) \\ &\leq \left( \frac{1+\beta}{1+\delta} \right) \frac{|F(x^k)^T F'(x^k) e^{i_k}|}{r_{i_k}} \end{aligned}$$

and the relaxation parameters  $\omega_k$  satisfies the condition

$$1 - \delta < \omega_k < 1 + \delta.$$

## 2. SEMI-LOCAL CONVERGENCE OF THE ALGORITHM

There has been little discussion on the semi-local convergence for the coordinate-relaxation algorithm 1.2. Below we give such a semi-local convergence theorem for the algorithm 1.2 with iteration order (I).

**Theorem 2.1.** *Let  $F : D \subset \mathbb{R}^n \rightarrow \mathbb{R}^n$  be continuously differentiable on the set  $D$ . Assume that  $F'(x)$  is nonsingular on  $D$  and there holds the following condition*

$$|F(x)^T F'(x)^T e^j - F(y)^T F'(y)^T e^j| \leq r_j \|x - y\| \quad (2.1)$$

where  $r_j > 0$ ,  $j = 1, 2, \dots, n$ . Suppose

$$\nu = \max_{x \in D} \max_{1 \leq j \leq n} \|F'(x) e^j\|_2 < \infty$$

and

$$r = \inf_{\|y\|_2=1} \inf_{x \in D} \max_{1 \leq j \leq n} \|y^T F'(x) e^j\|_2 > 0. \quad (2.2)$$

If  $x^0 \in D$  is such that the closed ball  $S = S(x^0, t_0) := \{x : \|x - x^0\|_2 \leq t_0\} \subset D$ , where

$$t_0 = \frac{(1+\beta)\nu}{(1-q)r_{\min}} \|F(x^0)\|_2,$$

with  $\beta$  being a real number such that  $0 < \delta \leq \beta \leq 1$  and  $(1-\beta^2)r^2 < r_{\min}$ , and  $r_{\min} = \min_{1 \leq j \leq N} r_j$ , and if the iteration order  $\{i_k\}$  satisfies (I) with the

step-size  $\{t_{i_k}\}$  satisfying the condition

$$\begin{aligned} \left(\frac{1-\beta}{1-\delta}\right) \frac{|F(x^k)^T F'(x^k)e^{i_k}|}{r_{i_k}} &\leq t_{i_k} \operatorname{sgn}(F(x^k)^T F'(x^k)e^{i_k}) \quad (2.3) \\ &\leq \left(\frac{1+\beta}{1+\delta}\right) \frac{|F(x^k)^T F'(x^k)e^{i_k}|}{r_{i_k}} \end{aligned}$$

where  $\delta$  and  $\beta$  are such that  $0 < \delta \leq \beta < 1$ , then for any  $1 - \delta \leq \omega_k \leq 1 + \delta$ , there hold the following conclusions for the algorithm 1.2:

- (i)  $\{x^k\} \subset S$ ;
- (ii)  $x^k \rightarrow x^* \in S$  as  $k \rightarrow \infty$  and  $F(x^*) = 0$ ; and
- (iii)  $\|x^k - x^*\|_2 \leq t_0 q^k$ .

*Proof.* First we prove the following conclusion by induction: for all  $k \geq 1$ ,

- (a)  $x^k \in S$ ,
- (b)  $\|F(x^k)\|_2 \leq q\|F(x^{k-1})\|_2$ , and
- (c)  $\|x^k - x^{k-1}\|_2 \leq \frac{(1+\beta)\nu}{r_{min}}\|F(x^0)\|_2 q^{k-1}$ .

Indeed, when  $k = 1$ , we have

$$\begin{aligned} \|x^1 - x^0\|_2 &= \omega_0 |t_{i_0}| \\ &\leq (1+\beta) \frac{|F(x^0)^T F'(x^0)e^{i_0}|}{r_{i_0}} \\ &\leq \frac{1+\beta}{r_{min}} \|F(x^0)\|_2 \|F'(x_0)e^{i_0}\|_2 \\ &\leq \frac{(1+\beta)\nu}{r_{min}} \|F(x^0)\|_2 < t_0. \end{aligned}$$

Hence  $x^1 \in S$ . Put

$$\varphi(x) = \frac{1}{2} \|F(x)\|_2^2.$$

Then

$$\varphi(x^0) - \varphi(x^1) \geq \frac{1-\beta^2}{2r_{i_0}} [F(x^0)^T F'(x^0)e^{i_0}]^2. \quad (2.4)$$

Notice that

$$|F(x^0)^T F'(x^0)e^{i_0}| = \max_{1 \leq j \leq n} |F(x^0)^T F'(x^0)e^j|.$$

If  $F(x^0) \neq 0$ , then using  $\gamma > 0$  we infer that

$$\max_{1 \leq j \leq n} |F(x^0)^T F'(x^0)e^j| \geq \gamma \|F(x^0)\|_2. \quad (2.5)$$

From (2.4) and (2.5) it follows that

$$\|F(x^0)\|_2^2 - \|F(x^1)\|_2^2 \geq \frac{(1 - \beta^2)r^2}{r_{max}} \|F(x^0)\|_2^2$$

and

$$\|F(x^1)\|_2 \leq q \|F(x^0)\|_2.$$

Consequently (a)-(c) hold for  $k = 1$ .

Assume that for some  $k > 1$ , (a)-(c) hold for all  $1 \leq j \leq k$  and prove that (a)-(c) remain true for  $k + 1$ . Indeed we have

$$\begin{aligned} \|x^{k+1} - x^k\|_2 &= \omega_k |t_{i_k}| \\ &\leq (1 + \beta) \frac{|F(x^k)F'(x^k)e^{i_k}|}{r_{i_k}} \\ &\leq \frac{1 + \beta}{r_{min}} \|F(x^k)F'(x^k)e^{i_k}\|_2 \\ &\leq \frac{(1 + \beta)\nu}{r_{min}} \|F(x^k)\|_2. \end{aligned}$$

By the induction assumption we get

$$\|x^{k+1} - x^k\|_2 \leq \frac{(1 + \beta)\nu}{r_{min}} \|F(x^0)\|_2 q^k.$$

In the meanwhile we have

$$\begin{aligned} \|x^{k+1} - x^0\|_2 &\leq \sum_{i=0}^k \|x^{i+1} - x^i\|_2 \\ &\leq \frac{(1 + \beta)\nu}{r_{min}} \|F(x^0)\|_2 \sum_{i=0}^k q^i < t_0. \end{aligned}$$

We therefore have that  $x^{k+1} \in S$ . Since  $\gamma > 0$  and

$$\varphi(x^k) - \varphi(x^{k+1}) \geq \frac{1 - \beta^2}{2r_{i_k}} |F(x^k)^T F'(x^k)e^{i_k}|^2, \quad (2.6)$$

we obtain

$$|F(x^k)^T F'(x^k)e^{i_k}| = \max_{1 \leq j \leq n} |F(x^j)^T F'(x^j)e^j| \geq \gamma \|F(x^k)\|_2. \quad (2.7)$$

By (2.6) and (2.7) we get

$$\|F(x^k)\|_2^2 - \|F(x^{k+1})\|_2^2 \geq \frac{(1 - \beta^2)\gamma^2}{r_{max}} \|F(x^k)\|_2^2.$$

Hence

$$\|F(x^k)\|_2^2 \leq q\|F(x^k)\|^2$$

and we have proved that (a)-(c) hold for all  $k \geq 1$ . Thus (i) holds.

For all integers  $k, m \geq 1$ , we have

$$\begin{aligned} \|x^{k+m} - x^k\|_2 &\leq \sum_{j=0}^{m-1} \|x^{k+j+1} - x^{k+j}\|_2 \\ &\leq \frac{(1+\beta)\nu}{r_{min}} \|F(x^0)\|_2 \sum_{j=0}^{m-1} q^{k+j} < t_0 q^k. \end{aligned} \quad (2.8)$$

It follows that  $\{x^k\}$  is Cauchy and hence convergent to  $x^*$  (say) in  $S$ . By (a)-(c) we see that

$$\|F(x^k)\|_2 \leq q^k \|F(x^0)\|_2, \quad 0 < q < 1.$$

Taking the limit as  $k \rightarrow \infty$  to obtain that  $F(x^*) = 0$ . This verifies (ii). To see (iii) we let  $m \rightarrow \infty$  in (2.8) to get  $\|x^k - x^*\|_2 \leq t_0 q^k$ .  $\square$

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