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## A GENERAL APPROACH TO METHODS WITH A SPARSE JACOBIAN FOR SOLVING NONLINEAR SYSTEMS OF EQUATIONS

Nikolay Kyurkchiev, Anton Iliev

*Communicated by G. Nikolov*

ABSTRACT. Here we give methodological survey of contemporary methods for solving nonlinear systems of equations in  $R^n$ . The reason of this review is that many authors in present days rediscovered such classical methods. In particular, we consider Newton's-type algorithms with sparse Jacobian. Method for which the inverse matrix of the Jacobian is replaced by the inverse matrix of the Vandermondian is proposed. A number of illustrative numerical examples are displayed. We demonstrate Herzberger's model with fixed-point relations to the some discrete versions of Halley's and Euler-Chebyshev's methods for solving such kind of systems.

**1. Introduction.** The multivariant Newton's method for solving equation

$$(1) \quad f(x) = 0, \quad f = (f_1, \dots, f_n), \quad x = (x_1, \dots, x_n)$$

is described by the iteration formula

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2000 *Mathematics Subject Classification*: 65H10.

*Key words*: Nonlinear systems of equations, numerical solution, Iliev's, Halley's and Euler-Chebyshev's methods, fixed-point relations.

$$(2) \quad x^{k+1} = x^k - (f'(x))^{-1} f(x^k), \quad k = 0, 1, 2, \dots,$$

where

$$f'(x) = J(x) = \left( \frac{\partial f_i(x)}{\partial x_j} \right), \quad i, j = 1, 2, \dots, n,$$

and suppose that  $f'(x)^{-1}$  exists in neighbourhood of solution  $\xi(\xi_1, \xi_2, \dots, \xi_n)$ .

In a number of cases, the calculation of the Jacobian in (2) is quite difficult. Readers interested in some modifications of Newton's-type iterative schemes are referred to the books by Ortega and Rheinboldt [13] and by Petkov and Kyurkchiev [15].

Let  $\theta_i(x_1, \dots, x_{i-1}, x_{i+1}, \dots, x_n)$  solves

$$(3) \quad \begin{aligned} f_i(x_1, \dots, x_{i-1}, \theta_i(x_1, \dots, x_{i-1}, x_{i+1}, \dots, x_n), x_{i+1}, \dots, x_n) = 0, \\ i = 1, \dots, n. \end{aligned}$$

Following Herzberger [5], we present a model for a class of iterative methods for the determination of the roots  $\xi_1, \dots, \xi_n$  of a nonlinear system defined by some based fixed-point relations

$$(4) \quad \varphi_i(x_1, \dots, x_{i-1}, \xi_i, x_{i+1}, \dots, x_n) = \xi_i, \quad i = 1, \dots, n,$$

i.e. the  $i$ -th iteration function is stationary for  $x_i = \xi_i$ .

Our last assumption restrict the methods described here in our model to a class of simultaneous methods by demanding

$$(5) \quad \begin{aligned} \varphi_i(\xi_1, \dots, \xi_{i-1}, \theta_i(x_1, \dots, x_{i-1}, x_{i+1}, \dots, x_n), \xi_{i+1}, \dots, \xi_n) = \theta_i, \\ i = 1, \dots, n. \end{aligned}$$

We demonstrate our model for the method

$$(6) \quad \varphi_i(x_1, \dots, x_n) = x_i - \frac{f_i(x_1, \dots, x_n)}{\frac{\partial f_i(x_1, \dots, x_n)}{\partial x_i}}$$

and define the iteration methods by

$$(7) \quad x_i^{k+1} = \varphi_i(x_1^k, \dots, x_n^k) \quad (\text{total-step method}),$$

$$(8) \quad x_i^{k+1} = \varphi_i(x_1^{k+1}, \dots, x_{i-1}^{k+1}, x_i^k, \dots, x_n^k) \quad (\text{single-step method})$$

for given initial values  $x_1^0, \dots, x_n^0$ .

**2. Main results.** We prove the following theorem:

**Theorem 1.** *For every member of the class of iteration methods defined by properties (4) and (5) the convergence of the total-step iteration (7) is locally linear.*

**Proof.** In order to estimate the order of convergence for the methods defined by the iteration functions  $\varphi_i$ , we derive an error recursion for these functions.

By virtue of (4), (5), we have

$$\begin{aligned}
 \varphi_i(\xi_1, \dots, \xi_n) &= \xi_i, \quad i = 1, \dots, n \\
 (9) \quad \frac{\partial \varphi_i(\xi_1, \dots, \xi_n)}{\partial x_i} &= 0, \quad i = 1, \dots, n \\
 \frac{\partial \varphi_i(\xi_1, \dots, \xi_n)}{\partial x_j} &= \frac{\partial \theta_i(\xi_1, \dots, \xi_{i-1}, \xi_{i+1}, \dots, \xi_n)}{\partial x_j}, \quad j \neq i.
 \end{aligned}$$

We expand  $\varphi_i$  in a Taylor series at the point  $(\xi_1, \dots, \xi_n)$

$$\varphi_i(\xi_1 + \epsilon_1, \dots, \xi_n^* + \epsilon_n) = \varphi_i(\xi_1, \dots, \xi_n) + \sum_{j=1}^n \frac{\partial \varphi_i}{\partial x_j}(\xi_1 + \eta \epsilon_1, \dots, \xi_n + \eta \epsilon_n) \epsilon_j,$$

$0 < \eta < 1.$

On the other hand

$$\begin{aligned}
 \frac{\partial \varphi_i}{\partial x_j}(\xi_1 + \eta \epsilon_1, \dots, \xi_n + \eta \epsilon_n) &= \frac{\partial \varphi_i}{\partial x_j}(\xi_1, \dots, \xi_n) + \\
 &+ \sum_{l=1}^n \frac{\partial^2 \varphi_i}{\partial x_j \partial x_l}(\xi_1 + \sigma \epsilon_1, \dots, \xi_n + \sigma \epsilon_n) \eta \epsilon_l,
 \end{aligned}$$

$0 < \sigma < 1$

and

$$\begin{aligned}
 \varphi_i(\xi_1 + \epsilon_1, \dots, \xi_n + \epsilon_n) &= \varphi_i(\xi_1, \dots, \xi_n) + \frac{\partial \varphi_i}{\partial x_i}(\xi_1, \dots, \xi_n) \epsilon_i + \\
 &+ \sum_{\substack{j=1 \\ j \neq i}}^n \frac{\partial \varphi_i}{\partial x_j}(\xi_1, \dots, \xi_n) \epsilon_j + \\
 &+ \sum_{j=1}^n \sum_{l=1}^n \frac{\partial^2 \varphi_i}{\partial x_j \partial x_l}(\xi_1 + \sigma \epsilon_1, \dots, \xi_n + \sigma \epsilon_n) \eta \epsilon_l \epsilon_j.
 \end{aligned}$$

Setting  $\epsilon_i = x_i^k - \xi_i$ ,  $e_i^k = |x_i^k - \xi_i|$  we get from (9) in view of the total-step method (7) the error recursion:

$$\begin{aligned}
 e_i^{k+1} &= |x_i^{k+1} - \xi_i| = |\varphi_i(x_1^k, \dots, x_n^k) - \xi_i| = \\
 &= |\varphi_i(\xi_1 + \epsilon_1, \dots, \xi_n + \epsilon_n) - \xi_i| \leq \\
 (10) \quad &\leq \sum_{j \neq i}^n \left| \frac{\partial \theta_i}{\partial x_j}(\xi_1, \dots, \xi_{i-1}, \xi_{i+1}, \dots, \xi_n) \right| e_j^k + \\
 &+ \sum_{j=1}^n \sum_{l=1}^n \left| \frac{\partial^2 \varphi_i}{\partial x_j \partial x_l}(\xi_1 + \sigma \epsilon_1, \dots, \xi_n + \sigma \epsilon_n) \right| \eta e_j^k e_l^k.
 \end{aligned}$$

Obviously from this recursion that the convergence is linear.

**Remark 1.** Some methods, known from the literature occur as special cases on the family (4) ((9)).

In [22], Wege proposed method (6). Here the inverse matrix of the Jacobian in well-known Newton's method (2) is replaced by the inverse of a matrix which consists of the main diagonal of the Jacobian.

In this case G. Iliev [8] shows that  $\{x_i^k\}$  from (7) converges quadratic, when in (9)

$$\frac{\partial \theta_i(\xi_1, \dots, \xi_{i-1}, \xi_{i+1}, \dots, \xi_n)}{\partial x_j} = 0; \quad i, j = 1, \dots, n; \quad j \neq i.$$

This result follows immediately from our recursion (10).

Other results can be found in the papers by Scheurle [17], Schubert [18] and Taiwo [21].

Extensions and applications in Banach spaces were given by Kantorovič [11], Altman [2], Stein [20], Ben-Israel [3], Yamamoto and Chen [23], Iliev [7] and others.

Note that (6) is a Jacoby-Newton-like iteration, but it possible to use a Gauss-Seidel procedure as follows:

$$\begin{aligned}
 x_i^{k+1} &= \varphi_i(x_1^{k+1}, \dots, x_{i-1}^{k+1}, x_i^k, \dots, x_n^k) \\
 (11) \quad &= x_i^k - \frac{f_i}{\partial f_i / \partial x_i}(x_1^{k+1}, \dots, x_{i-1}^{k+1}, x_i^k, \dots, x_n^k), \\
 & \quad \quad \quad i = 1, 2, \dots, n; \quad k = 0, 1, 2, \dots
 \end{aligned}$$

For other SOR (successive overrelaxation) methods with parameter see the paper by Ishihara, Muroya and Yamamoto [9].

We give some discrete versions of Halley's and Euler-Chebyshev's methods for solving nonlinear systems of equations.

First, we define the following iteration (total-step Halley's method (HM)):

$$(12) \quad x_i^{k+1} = x_i^k - \frac{f_i(x^k)}{\frac{\partial f_i(x^k)}{\partial x_i}} \left( 1 + \frac{1}{2} \frac{f_i(x^k) \frac{\partial^2 f_i(x^k)}{\partial x_i^2}}{\left( \frac{\partial f_i(x^k)}{\partial x_i} \right)^2} \right)^{-1},$$

$$i = 1, \dots, n; \quad k = 0, 1, \dots$$

This method obviously is improvement of the method (6).

We note that, in (12) we use the main diagonal of the Hessian

$$H := \left\| f_i^{mn} = \frac{\partial^2 f_i}{\partial x_m \partial x_n} \right\|.$$

There are many methods based on the fixed-point relations (see, the books by Alefeld and Herzberger [1], Petkovic [16], Sendov, Andreev and Kyurkchiev [19]) and Kyurkchiev [11].

These algorithms can be arranged for solving nonlinear equations in several variables.

For example, we define the following iteration (total-step Euler-Chebyshev's method):

$$(13) \quad x_i^{k+1} = x_i^k - \frac{f_i(x^k)}{\frac{\partial f_i(x^k)}{\partial x_i}} \left( 1 - \frac{1}{2} \frac{f_i(x^k) \frac{\partial^2 f_i(x^k)}{\partial x_i^2}}{\left( \frac{\partial f_i(x^k)}{\partial x_i} \right)^2} \right), i = 1, \dots, n; \quad k = 0, 1, \dots$$

**Numerical example** (G.Iliev [8]). Let

$$\begin{cases} f_1(x_1, x_2) = (x_2 - 0.07)x_1^2 - 0.3x_1x_2 - 0.15x_2^2 = 0 \\ f_2(x_1, x_2) = (x_1 - 0.15)x_2^2 - 0.14x_1x_2 - 0.07x_1^2 = 0 \end{cases}$$

and  $x_1^0 = x_2^0 = 1$ .

The solutions of this system are:

$$\begin{cases} x_1 = 0.504939015319191968 \\ x_2 = 0.344939015319191968. \end{cases}$$

For the numerical determination of these roots, we apply Newton-Raphson method, method (6) and method (12).

The results are shown in Table 1, Table 2 and Table 3.

The presented numerical example was realized in precision arithmetic (about 18 significant digits).

Table 1. Solution by Newton's method

$k$	$x_1^k$	$x_2^k$
0	1.	1.
1	0.770320656226696495	0.695749440715883669
2	0.626370174040210892	0.505673164801260020
3	0.545377788002368922	0.398768875062025654
4	0.511666539615088034	0.353997741004263732
5	0.505175082853654779	0.345262185297826159
6	0.504939323674820933	0.34493944574494562
7	0.504939015319723450	0.344939015319950318
8	0.504939015319191968	0.344939015319950318

Table 2. Solution by method (6)

$k$	$x_1^k$	$x_2^k$
0	1.	1.
1	0.692307692307692308	0.589743589743589744
2	0.555123173090900958	0.409350643541659678
3	0.510754813273147686	0.352251613156086770
4	0.505035199913144336	0.345057597334594785
5	0.504939042018383390	0.344939047675290015
6	0.504939015319193995	0.344939015319194388
7	0.504939015319191968	0.344939015319191968

In [6], Ibidapo-Obe, Asaolu and Badiri developed a new technique for solving nonlinear system of equations (1). This is achieved by the iterative solution of a parametric linear system coupled with a nonlinear single variable equation.

Let  $\mathbf{x}_0$  be an approximate solution to (1). The authors construct a new vector  $\mathbf{y} = (y_1, \dots, y_{n-1}, t)$  such that  $y_j, j = 1, \dots, n-1$  are distinct elements of

Table 3. Solution by method (12)

$k$	$x_1^k$	$x_2^k$
0	1.	1.
1	0.623188405797101449	0.471634208298052498
2	0.515759913152825672	0.352702914879984042
3	0.504992271181411001	0.344986324556096995
4	0.504939017246729918	0.344939016459300214
5	0.504939015319191969	0.344939015319191969
6	0.504939015319191968	0.344939015319191968

$\mathbf{x}_0$  (though  $y_j$  does not necessarily map into  $x_j$ );  $t$  is variable correspond to the  $\mathbf{x}_0$  and renumber the system of equations in (1) such that  $y_j$  corresponds to  $x_j$ . Assuming  $\mathbf{y}_0$  is an approximate solution but with  $t$  fixed and unknown implies (1) becomes

$$\mathbf{f}_j(\mathbf{y}_0) + \sum_{k=1}^{n-1} \frac{\partial \mathbf{f}_j(\mathbf{y}_0)}{\partial y_k} \xi_k = 0, \quad j = 1, 2, \dots, n; \quad (\xi_k = dy_k).$$

The updated solution to (1) becomes  $\mathbf{y}_1 = (y_{01} + \xi_1, y_{02} + \xi_2, \dots, t)$ . The proposed method [6] is a convergent method for a good initial guess.

Evidently, this procedure yields the exact solution  $\mathbf{x}$ , when applied to a linear system since the Jacobian then becomes the constant coefficient matrix.

**Remark 2.** Note that the iteration  $\mathbf{y}_j, j = 0, 1, 2, \dots$  has been mentioned by N. Obreshkoff in [12].

**Remark 3.** The convergence order can be increased by the following way.

Let

$$\begin{aligned} \mathbf{f}_j(\mathbf{x}) &= \mathbf{f}_j(\mathbf{y}_0 + \xi) = \\ (14) \quad &= \mathbf{f}_j(\mathbf{y}_0) + \sum_{k=1}^{n-1} \frac{\partial \mathbf{f}_j(\mathbf{y}_0)}{\partial y_k} \xi_k + \frac{1}{2} \sum_{l=1}^{n-1} \sum_{k=1}^{n-1} \frac{\partial^2 \mathbf{f}_j(\mathbf{y}_0)}{\partial y_l \partial y_k} \xi_l \xi_k = 0, \\ & \quad j = 1, \dots, n. \end{aligned}$$

The simultaneous system of equations given by (14) is quadratic in the  $\xi'_k$ s.

Convergence to the solution is attained when  $\mathbf{f}(y_{l+1}) \approx 0$  and  $\|y_{l+1} - y_l\| < \eta$ , for an arbitrary chosen  $\eta$ , depending on the accuracy desired ( $y_{l+1} = (y_{l1} + \xi_1, y_{l2} + \xi_2, \dots, t)$ ).



The new procedure (based on (14)) yields the exact solution  $\mathbf{x}$ , when applied to quadratic nonlinear system.

**Numerical example** (Paterson [14], [6]). Consider the following system

$$(A) \quad \begin{cases} 4x_1 + x_2^2 + x_3 - 11 = 0, \\ x_1 + 4x_2 + x_3^2 - 18 = 0, \\ x_1^2 + x_2 + 4x_3 - 15 = 0 \end{cases}$$

with  $\mathbf{x}_0 = (1, 1, 1)^T$ . The system (A) has the different roots:  $\mathbf{x} = (1, 2, 3)^T$ . The method by Ibidapo-Obe, Asaolu, Badiru [6], with  $\mathbf{y}_0 = (t, x_{20}, x_{30})^T$  gives

$$(B) \quad \begin{cases} 4t + x_{20}^2 + x_{30} - 11 + 2x_{20}\xi_1 + \xi_2 = 0, \\ t + 4x_{20} + x_{30}^2 - 18 + 4\xi_1 + 2x_{30}\xi_2 = 0, \\ t^2 + x_{20} + 4x_{30} - 15 + \xi_1 + 4\xi_2 = 0. \end{cases}$$

In particular, with  $y_0 = (t, 1, 1)^T$ , we have

$$(C) \quad \begin{cases} 4t - 9 + 2\xi_1 + \xi_2 = 0, \\ t - 13 + 4\xi_1 + 2\xi_2 = 0, \\ t^2 - 10 + \xi_1 + 4\xi_2 = 0. \end{cases}$$

From the last two equations in (C) we find

$$\xi_1 = \frac{32 - 4t + 2t^2}{14}, \quad \xi_2 = \frac{-4t^2 + t + 27}{14}.$$

Substituting in the first equation (C) yields

$$49t - 35 = 0, \quad \text{or } t = 0.7145; \quad \xi_1 = 2.1543; \quad \xi_2 = 1.834,$$

i.e.  $\mathbf{y}_1 = (0.7145, 3.1543, 2.834)^T = (t, 1 + \xi_1, 1 + \xi_2)^T$ .

Finally

$$\mathbf{x} = \mathbf{y}_5 = \mathbf{y}_6 = (1.0000, 2.0000, 3.0000)^T.$$

The new method (based on (14), with  $\mathbf{y}_0 = (t, 1, 1)^T$  gives

$$(B') \quad \begin{cases} 4t + x_{20}^2 + x_{30} - 11 + 2x_{20}\xi_1 + \xi_2 + \xi_1^2 = 0, \\ t + 4x_{20} + x_{30}^2 - 18 + 4\xi_1 + 2x_{30}\xi_2 + \xi_2^2 = 0, \\ t^2 + x_{20} + 4x_{30} - 15 + \xi_1 + 4\xi_2 = 0 \end{cases}$$

and

$$(C') \quad \begin{cases} 4t - 9 + 2\xi_1 + \xi_2 + \xi_1^2 = 0, \\ t - 13 + 4\xi_1 + 2\xi_2 + \xi_2^2 = 0, \\ t^2 - 10 + \xi_1 + 4\xi_2 = 0. \end{cases}$$

From (C'), we have

$$\xi_2 = \frac{321 + 12t - 42t^2 - t^4}{137 + 8t^2}; \quad \xi_1 = 10 - t^2 - 4\xi_2.$$

Substituting in the second equation (C') yields

$$(D) \quad t - 13 + 4\xi_1(t) + 2\xi_2(t) + \xi_2^2(t) = 0.$$

Evidently, for  $t = 1$ , we have  $\xi_2 = 2, \xi_1 = 1$ , i.e.  $y_1 = (1, 2, 3)^T$ .

Here we give a method for which the inverse matrix of the Jacobian  $J(x)$  is replaced by the inverse matrix of the Vandermondian  $V(x)$ . The notation  $V(x_1^k, \dots, x_n^k)$  will denote a Vandermonde matrix:

$$V(x_1^k, \dots, x_n^k) = \begin{pmatrix} (x_1^k)^{n-1} & (x_1^k)^{n-2} & \dots & x_1^k & 1 \\ (x_2^k)^{n-1} & (x_2^k)^{n-2} & \dots & x_2^k & 1 \\ \dots & \dots & \dots & \dots & \dots \\ (x_n^k)^{n-1} & (x_n^k)^{n-2} & \dots & x_n^k & 1 \end{pmatrix}$$

of dimension  $n$ , and  $|V(x_1^k, \dots, x_n^k)|$  is its determinant and is equal to  $\prod_{i < j} (x_i^k - x_j^k) \neq 0$  ( $x_i^k \neq x_j^k, i \neq j, i, j = 1, 2, \dots, n$ ). Let

$$V^{-1} = \begin{pmatrix} c_{11} & c_{12} & \dots & c_{1n} \\ c_{21} & c_{22} & \dots & c_{2n} \\ \dots & \dots & \dots & \dots \\ c_{n1} & c_{n2} & \dots & c_{nn} \end{pmatrix} = \frac{1}{|V(x_1^k, x_2^k, \dots, x_n^k)|} \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \dots & \dots & \dots & \dots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{pmatrix}.$$

Let us consider the following method

$$(15) \quad x_i^{k+1} = x_i^k - \sum_{j=1}^n c_{ij} f_j(x_1^k, x_2^k, \dots, x_n^k), \quad i = 1, 2, \dots, n; \quad k = 0, 1, 2, \dots,$$

or

$$(16) \quad x^{k+1} = x^k - V^{-1}(x^k) f(x^k), \quad k = 0, 1, 2, \dots$$

Let  $x_i^k = x_i + \epsilon_i, i = 1, 2, \dots, n$  and  $f$  has a continuous derivative in the neighborhood of the root  $x$ . Then

$$f_i(x_1^k, x_2^k, \dots, x_n^k) = f_i(x_1 + \epsilon_1, x_2 + \epsilon_2, \dots, x_n + \epsilon_n) =$$

$$\begin{aligned}
&= f_i(x_1, x_2, \dots, x_n) + \sum_{j=1}^n \frac{\partial f_i(x_1, x_2, \dots, x_n)}{\partial x_j} \epsilon_j + o(\|\epsilon\|), \\
x_i^{k+1} - x_i &= x_i^k - x_i - \sum_{j=1}^n c_{ij} \left( f_j(x_1^k, x_2^k, \dots, x_n^k) - f_j(x_1, x_2, \dots, x_n) \right) = \\
&= x_i^k - x_i - \sum_{j=1}^n c_{ij} \left( \sum_{l=1}^n \frac{\partial f_j(x_1, x_2, \dots, x_n)}{\partial x_l} (x_l^k - x_l) + o(\|\epsilon\|) \right).
\end{aligned}$$

Let  $\|x - x^k\| \rightarrow 0$ , then

$$x^{k+1} - x = (x^k - x) \left( I - V^{-1}(x^k) \frac{\partial f(p^k)}{\partial x} \right),$$

with  $p^k$  on the line  $l_{x^k x} = 0$ , where  $l_{x^k x}$  denotes the equation of the straight-line through the points  $x^k = (x_1^k, x_2^k, \dots, x_n^k)$  and  $x = (x_1, x_2, \dots, x_n)$ . Let  $\epsilon, \delta < 1$  be positive constants such that for all  $k = 0, 1, 2, \dots$

$$\left\| I - V^{-1}(x^k) \frac{\partial f(p^k)}{\partial x} \right\| \leq \delta < 1,$$

$$\rho(x^k, x) = \|x^k - x\| \leq \epsilon.$$

Then the sequence  $\{x_i^{k+1}\}_{i=1}^n$  converges to a solution of  $f(x) = 0$  and

$$(17) \quad \rho(x^{k+1}, x) < \epsilon.$$

For  $k = 0$ , we have

$$\rho(x^1, x) = \|x^1 - x\| = \left\| I - V^{-1}(x^0) \frac{\partial f(p^0)}{\partial x} \right\| \|x^0 - x\| \leq \delta \epsilon < \epsilon.$$

Assuming (17) it is true for all subscripts  $\leq k$ , we prove (17) for  $k + 1$ . Indeed,

$$\rho(x^{k+1}, x) = \|x^{k+1} - x\| \leq \delta \|x^k - x\| \leq \delta \epsilon < \epsilon,$$

$$(18) \quad \rho(x^{k+1}, x) \leq \delta^{k+1} \rho(x^0, x).$$

But  $\delta < 1$ , the right-hand side of (18) tends to zero as  $k \rightarrow \infty$ . So that we now frame a theorem

**Theorem 2.** Let  $f$  be defined in the closed ball  $B(\|x^0 - x\| \leq \epsilon)$  and have continuous Frechét derivative. Moreover, suppose that

$$\left\| I - V^{-1}(x^0) \frac{\partial f(x^0)}{\partial x} \right\| \leq \delta < 1,$$

$$\rho(x^0 - x) = \|x^0 - x\| \leq \epsilon, \quad \epsilon > 0.$$

Then the sequence  $\{x^k\}$  generated by (16), starting with  $x^0$  converges to a solution of the equation  $f(x) = 0$ .

**Remark 4.** The standard implementation of the method (16) is as follows:

$$(19) \quad x^{k+1} = x^k - \frac{\begin{pmatrix} a_{11} & \dots & a_{1n} \\ \vdots & & \vdots \\ a_{n1} & \dots & a_{nn} \end{pmatrix} \begin{pmatrix} f_1 \\ \vdots \\ f_n \end{pmatrix} \Big|_{x=x^k}}{|V(x_i^k, x_1^k, \dots, x_{i-1}^k, x_{i+1}^k, \dots, x_n^k)|}, \quad k = 0, 1, 2, \dots$$

**Remark 5.** For each  $1 \leq j \leq n$ , we have

$$(20) \quad \begin{cases} c_{1j} = (-1)^{j+1} \frac{|V(x_1^k, \dots, x_{j-1}^k, x_{j+1}^k, \dots, x_n^k)|}{|V(x_1^k, \dots, x_n^k)|}, \\ c_{2j} = (-1)^{j+2} \frac{|V(x_1^k, \dots, x_{j-1}^k, x_{j+1}^k, \dots, x_n^k)|}{|V(x_1^k, \dots, x_n^k)|} \sum_{i \neq j}^n x_i^k, \\ \dots \\ c_{nj} = (-1)^{j+n} \frac{|V(x_1^k, \dots, x_{j-1}^k, x_{j+1}^k, \dots, x_n^k)|}{|V(x_1^k, \dots, x_n^k)|} \prod_{i \neq j}^n x_i^k. \end{cases}$$

**Remark 6.** The convergence order can be increased by calculating the new approximations  $x_i^{k+1}$ , in (19) using the already calculated approximations  $x_1^{k+1}, \dots, x_{i-1}^{k+1}$  (the so called Gauss-Seidel approach).

**Remark 7.** An interesting modification of (6) is to use so-called Weierstrass' correction:

$$(21) \quad x_i^{k+1} = x_i^k - \frac{f_i(x_1^k, \dots, x_n^k)}{\prod_{j \neq i} (x_i^k - x_j^k)}, \quad i = 1, 2, \dots, n; \quad k = 0, 1, 2, \dots$$

**Remark 8.** Wide area of problems and practical tasks in electro-dynamics, and physics are reduced to the problem of solving a system of nonlinear algebraic equations with some constraint conditions. For example, from Kirchoff law, we have system of nonlinear algebraic equations for determining the bus voltages (see, Cai and Chen [4]).

In this case, equations (15) can be rewritten in the following form:

$$\begin{aligned}
 x_1^{k+1} &= x_1^k - \sum_{j=1}^n \frac{f_j(x_1^k, \dots, x_n^k)}{\prod_{s \neq j} (x_j^k - x_s^k)}, \\
 x_2^{k+1} &= x_2^k - \sum_{j=1}^n \frac{\sum_{s \neq j} x_s^k f_j(x_1^k, \dots, x_n^k)}{\prod_{s \neq j} (x_j^k - x_s^k)}, \\
 &\dots \\
 x_n^{k+1} &= x_n^k - \sum_{j=1}^n \frac{\prod_{s \neq j} x_s^k f_j(x_1^k, \dots, x_n^k)}{\prod_{s \neq j} (x_j^k - x_s^k)}
 \end{aligned}$$

and iteration algorithm can be refined using Euler’s formula:

$$\sum_{i=1}^n \frac{(x_i^k)^t}{\prod_{j \neq i} (x_i^k - x_j^k)} = \begin{cases} \sum_{i=1}^n x_i^k, & t = n \\ 1, & t = n - 1 \\ 0, & 0 \leq t \leq n - 2. \end{cases}$$

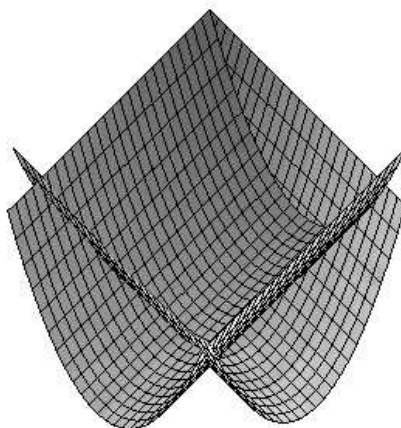
**3. Numerical examples.** To illustrate the method (19) we present the following simple example (see Ortega and Rheinboldt [13]):

Let

$$\begin{cases} f_1(x_1, x_2) = x_1^2 - x_2 - 1 = 0 \\ f_2(x_1, x_2) = -x_1 + x_2^2 - 1 = 0 \end{cases}$$

The solutions of this system are:  $(x_1 = 0, x_2 = -1)$ ;  $(x_1 = -1, x_2 = 0)$ ;  $\left(x_1 = x_2 = \frac{1 + \sqrt{5}}{2}\right)$ ;  $\left(x_1 = x_2 = \frac{1 - \sqrt{5}}{2}\right)$ .

Compute simultaneously, the solution  $(0, -1)$ .



The choice of initial guess  $(x_1^0 = -0.2, x_2^0 = -1.25)$  is critical for the Newton's method.

Indeed,

$$J(x) = \begin{pmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} \end{pmatrix} = \begin{pmatrix} 2x_1 & -1 \\ -1 & 2x_2 \end{pmatrix}, \quad \det J(x^0) = 4x_1^0 x_2^0 - 1 = 0!$$

We illustrate the behavior of Newton's method (2), starting from initial guess:  $x_1^0 = -0.19, x_2^0 = -1.24$ .

Table 4. Solution by Newton's method

$k$	$x_1^k$	$x_2^k$
0	-0.19	-1.24
1	0.5543055555555555	-1.2467361111111111
2	0.187353268299377751	-1.0995527339800818
3	0.036885904446349925	-1.02127985763805421
4	0.002021592710723367	-1.0012114333957109
5	0.000006662067951401	-1.00000405990111203
6	0.000000000072281936	-1.00000000004438219
7	0	-1

The components  $x_1^7$  and  $x_2^7$  tend to solution  $(0, -1)$ . For numerical determination of  $x_i^{k+1}$ ,  $i = 1, 2$ , we apply method (19), with  $x_1^0 = -0.2$  and

$$x_2^0 = -1.25:$$

$$\begin{cases} x_1^{k+1} = x_1^k - \frac{f_1(x_1^k, x_2^k) - f_2(x_1^k, x_2^k)}{x_1^k - x_2^k} \\ x_1^{k+1} = x_1^k - \frac{x_1^k f_2(x_1^k, x_2^k) - x_2^k f_1(x_1^k, x_2^k)}{x_2^k - x_1^k} \end{cases}$$

Table 5. Solution by method (19)

$k$	$x_1^k$	$x_2^k$
0	-0.2	-1.25
1	0.25	-1.05
2	0.05	-0.9875
3	-0.0125	-1
4	0	-1

The components  $x_1^4$  and  $x_2^4$  tend to exact solution  $(0, -1)$ . The method (21) leads to

Table 6. Solution by method (21)

$k$	$x_1^k$	$x_2^k$
0	-0.19	-1.24
1	-0.452952380952380952	-0.547047619047619048
2	2.180406400616926930	-3.18040640061692694
3	0.88683791957015072	-1.88683791957015072
4	0.283552059144475374	-1.28355205914447537
5	0.051305953003849593	-1.051305953003849590
6	0.002387332115035988	-1.00238733211503599
7	0.000005672271435954	-1.00000567227143595
8	0.000000000032174298	-1.0000000000321743
9	0	-1

We receive the exact solution with accuracy of 18 decimal digits after 9 iterations.

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Nikolay Kyurkchiev  
Institute of Mathematics  
and Informatics  
Bulgarian Academy of Sciences  
Acad. G. Bonchev Str., Bl. 8  
1113 Sofia, Bulgaria  
e-mail: nkyurk@math.bas.bg

Anton Iliev  
Faculty of Mathematics and Informatics  
University of Plovdiv  
24, Tsar Assen Str.  
4000 Plovdiv, Bulgaria  
e-mail: aii@uni-plovdiv.bg

Institute of Mathematics and Informatics  
Bulgarian Academy of Sciences  
Acad. G. Bonchev Str., Bl. 8  
1113 Sofia, Bulgaria

Received May 29, 2007