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ON SOME MODIFICATIONS OF THE NEKRASSOV METHOD FOR NUMERICAL SOLUTION OF LINEAR SYSTEMS OF EQUATIONS*

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ABSTRACT. A modification of the Nekrassov method for finding a solution of a linear system of algebraic equations is given and a numerical example is shown.

1. Introduction. Let us consider the linear system $Ax - b = 0$ or

$$(1) \quad \begin{aligned} a_{i1}x_1 + a_{i2}x_2 + \cdots + a_{ii}x_i + \cdots + a_{in}x_n - b_i &= 0 = f_i(x_1, x_2, \dots, x_n), \\ i &= 1, 2, \dots, n. \end{aligned}$$

Suppose that the matrix A is diagonally dominant and $a_{ii} > 0$, $i = 1, \dots, n$.

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One of the more effective iteration methods for solving the system (1) is the Jacobi procedure (his method is also known as the *method of simultaneous displacements*):

$$\begin{aligned}
 x_i^{k+1} &= -\sum_{j \neq i}^n \frac{a_{ij}}{a_{ii}} x_j^k + \frac{b_i}{a_{ii}} \\
 (2) \quad &= x_i^k - \frac{1}{a_{ii}} f_i(x_1^k, \dots, x_n^k) \\
 &= x_i^k - \frac{f_i(x_1^k, \dots, x_n^k)}{\partial f_i / \partial x_i^k}, \\
 & \quad i = 1, 2, \dots, n; \quad k = 0, 1, 2, \dots,
 \end{aligned}$$

i.e., (2) is the *Newton scheme applied for the equation* $f_i = 0$.

A more powerful class of methods can be described by the recursion (*Richardson iteration*):

$$(3) \quad x^{k+1} = x^k - \alpha_k (Ax^k - b),$$

where α_i , $i = 1, \dots, k$ are damping factors.

For instance, the Richardson iteration (3) with the application of *Chebyshev acceleration factors* is defined by

$$\begin{aligned}
 \alpha_i &= 2 \left(a + b - (b - a) \cos \frac{(2i + 1)\pi}{2(k + 1)} \right)^{-1}, \\
 & \quad i = 0, 1, \dots, k
 \end{aligned}$$

$a \leq \lambda_i \leq b$, $i = 1, \dots, n$ (λ_i are the eigenvalues of matrix A).

In [8] we give the following modification of the Richardson method:

$$\begin{aligned}
 (4) \quad x_i^{k+1} &= x_i^k - \frac{1}{M_i^k} \left(\sum_{j=1}^n a_{ij} x_j^k - b_i \right), \\
 & \quad i = 1, 2, \dots, n; \quad k = 0, 1, 2, \dots,
 \end{aligned}$$

where

$$M_i^k = \prod_{j \neq i}^n |x_i^k - x_j^k|, \quad i = 1, 2, \dots, n; \quad k = 0, 1, \dots$$

For other contributions see Saad and van der Vorst [14], Freund, Golub and Nachtigal [6], Ishihara, Muroya and Yamamoto [7], Maleev [10], Stork [17], Zawilski [18].

One geometric interpretation of method (4) is also given in [8].

In a similar manner other iterations can be obtained which are modifications of algorithms which have been explored in details in books by Björck [2], Fadeev, D. and Fadeev, V. [4] and Barrett, R., M. Berry and others [1].

As an example a scheme of the Gauss–Seidel or the Nekrassov method (see Nekrassov [13], Mehmke [11] and Nekrassov and Mehmke [12]) look thus:

$$(5) \quad x_i^{k+1} = -\sum_{j=1}^{i-1} \frac{a_{ij}}{a_{ii}} x_j^{k+1} - \sum_{j=i+1}^n \frac{a_{ij}}{a_{ii}} x_j^k + \frac{b_i}{a_{ii}},$$

$$i = 1, 2, \dots, n; \quad k = 0, 1, 2, \dots$$

2. Main results. Let us explore the following modification of the Nekrassov method (assume that $x_i \neq x_j$ and $x_i^0 \neq x_j^0$ for $i \neq j$):

$$(6) \quad x_i^{k+1} = x_i^k - \frac{1}{N_i^k} \left(\sum_{j=1}^{i-1} a_{ij} x_j^{k+1} + a_{ii} x_i^k + \sum_{j=i+1}^n a_{ij} x_j^k - b_i \right),$$

$$i = 1, 2, \dots, n; \quad k = 0, 1, 2, \dots,$$

where

$$N_i^k = \prod_{j=1}^{i-1} |x_i^k - x_j^{k+1}| \prod_{j=i+1}^n |x_i^k - x_j^k|, \quad i = 1, 2, \dots, n; \quad k = 0, 1, \dots$$

Let

$$\delta_i^k = \frac{a_{ii}}{N_i^k}, \quad i = 1, 2, \dots, n; \quad k = 0, 1, 2, \dots$$

The iteration procedure (6) (*successive overrelaxation procedure*) can be rewritten as

$$(7) \quad x_i^{k+1} = x_i^k - \frac{a_{ii}}{N_i^k} \left(\sum_{j=1}^{i-1} \frac{a_{ij}}{a_{ii}} x_j^{k+1} + x_i^k + \sum_{j=i+1}^n \frac{a_{ij}}{a_{ii}} x_j^k - \frac{b_i}{a_{ii}} \right)$$

$$= x_i^k (1 - \delta_i^k) - \delta_i^k \left(\sum_{j=1}^{i-1} \frac{a_{ij}}{a_{ii}} x_j^{k+1} + \sum_{j=i+1}^n \frac{a_{ij}}{a_{ii}} x_j^k - \frac{b_i}{a_{ii}} \right).$$

1. When $\delta_i^k = 1$ from (7) we obtain the Nekrassov method.
2. One geometric interpretation of method (7) is the following:
Let

$$F_{k,i} = (x - x_1^{k+1}) \dots (x - x_{i-1}^{k+1})(x - x_{i+1}^k) \dots (x - x_n^k).$$

Then

$$F'_{k,i}(x_i^k) = \prod_{j=1}^{i-1} (x_i^k - x_j^{k+1}) \prod_{j=i+1}^n (x_i^k - x_j^k)$$

and the previous expression can be used for approximation of a_{ii} in the Nekrassov procedure.

We give a convergence theorem for the relaxation method (7).

Theorem 1. *Let*

$$\beta_i = \sum_{j=1}^{i-1} \frac{|a_{ij}|}{a_{ii}}, \quad \gamma_i = \sum_{j=i+1}^n \frac{|a_{ij}|}{a_{ii}}, \quad \delta_i^k \in (1, 2),$$

(8)

$$\beta_i + \gamma_i \in \left(0, \frac{1 - |1 - \delta_i^k|}{\delta_i^k}\right) \subset (0, 1), \quad i = 1, 2, \dots, n; \quad k = 0, 1, 2, \dots$$

Then the iteration procedure (7) converges to the unique solution x_i , $i = 1, 2, \dots, n$ of the system (1).

Proof. For the error $x_i^{k+1} - x_i$, we have

$$\begin{aligned} x_i^{k+1} - x_i &= x_i^k(1 - \delta_i^k) - x_i \\ &- \delta_i^k \left(\sum_{j=1}^{i-1} \frac{a_{ij}}{a_{ii}} x_j^{k+1} + \sum_{j=i+1}^n \frac{a_{ij}}{a_{ii}} x_j^k - \sum_{j=1}^{i-1} \frac{a_{ij}}{a_{ii}} x_j - \sum_{j=i+1}^n \frac{a_{ij}}{a_{ii}} x_j - x_i \right) \\ &= (x_i - x_i^k)(\delta_i^k - 1) + \delta_i^k \sum_{j=1}^{i-1} \frac{a_{ij}}{a_{ii}} (x_j - x_j^{k+1}) + \delta_i^k \sum_{j=i+1}^n \frac{a_{ij}}{a_{ii}} (x_j - x_j^k) \end{aligned}$$

and

(10)

$$\begin{aligned} |x_i^{k+1} - x_i| &\leq |\delta_i^k - 1| |x_i^k - x_i| + \delta_i^k \sum_{j=1}^{i-1} \frac{|a_{ij}|}{a_{ii}} |x_j - x_j^{k+1}| + \delta_i^k \sum_{j=i+1}^n \frac{|a_{ij}|}{a_{ii}} |x_j - x_j^k| \\ &\leq |\delta_i^k - 1| \|x - x^k\|_1 + \delta_i^k \beta_i \|x - x^{k+1}\|_1 + \delta_i^k \gamma_i \|x - x^k\|_1 \\ &= (|\delta_i^k - 1| + \gamma_i \delta_i^k) \|x - x^k\|_1 + \delta_i^k \beta_i \|x - x^{k+1}\|_1. \end{aligned}$$

Let

$$\max_i |x_i^{k+1} - x_i| = |x_{i_0}^{k+1} - x_{i_0}|.$$

Then from (10) we get

$$\begin{aligned} \|x - x^{k+1}\|_1 &= \max_i |x_i - x_i^{k+1}| = |x_{i_0}^{k+1} - x_{i_0}| \\ &\leq (|\delta_{i_0}^k - 1| + \gamma_{i_0} \delta_{i_0}^k) \|x - x^k\|_1 + \delta_{i_0}^k \beta_{i_0} \|x - x^{k+1}\|_1 \end{aligned}$$

and

$$(11) \quad \|x - x^{k+1}\|_1 \leq \frac{|\delta_{i_0}^k - 1| + \gamma_{i_0} \delta_{i_0}^k}{1 - \delta_{i_0}^k \beta_{i_0}} \|x - x^k\|_1 = K_{i_0} \|x - x^k\|_1.$$

Evidently from (8) we have

$$K_{i_0} = \frac{|\delta_{i_0}^k - 1| + \gamma_{i_0} \delta_{i_0}^k}{1 - \delta_{i_0}^k \beta_{i_0}} \leq \frac{|\delta_{i_0}^k - 1| + \delta_{i_0}^k \left(\frac{1 - |\delta_{i_0}^k - 1|}{\delta_{i_0}^k} - \beta_{i_0} \right)}{1 - \delta_{i_0}^k \beta_{i_0}} = 1.$$

This proves Theorem 1. \square

Let

$$L = \begin{pmatrix} 0 & 0 & \cdots & 0 \\ a_{21} & 0 & \cdots & 0 \\ \vdots & \vdots & & \vdots \\ a_{n1} & a_{n2} & \cdots & 0 \end{pmatrix}, \quad R = \begin{pmatrix} 0 & a_{12} & \cdots & a_{1n} \\ 0 & 0 & \cdots & a_{2n} \\ \vdots & \vdots & & \vdots \\ 0 & 0 & \cdots & 0 \end{pmatrix}, \quad X^k = \begin{pmatrix} x_1^k \\ x_2^k \\ \vdots \\ x_n^k \end{pmatrix},$$

$$P = \begin{pmatrix} a_{11} & 0 & \cdots & 0 \\ 0 & a_{22} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & a_{nn} \end{pmatrix}, \quad \delta^k = \begin{pmatrix} \delta_{11}^k & 0 & \cdots & 0 \\ 0 & \delta_{22}^k & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \delta_{nn}^k \end{pmatrix}.$$

Theorem 2. *The iteration (6) (or (7)) is convergent when all roots (eigenvalues) of the equation*

$$(12) \quad \left| A\delta^k - (P + \delta^k L) + t(P + \delta^k L) \right| = 0$$

are $|t_i| < 1$, $i = 1, \dots, n$.

Proof. In matrix terms the *successive overrelaxation procedure* (7) can be written as follows:

$$(13) \quad X^{k+1} = (P + \delta^k L)^{-1} \left((I - \delta^k)P - \delta^k R \right) X^k + (P + \delta^k L)^{-1} \delta^k b,$$

i.e.

$$X^{k+1} = BX^k + c.$$

Evidently, $|B - tI| = 0$ can be represented as

$$|B - tI| = \left| (P + \delta^k L)^{-1} \left| A\delta^k - (P + \delta^k L) + t(P + \delta^k L) \right| \right| = 0,$$

and the statement of Theorem 2 follows from the standard iteration theory. \square

3. In a number of cases the success of the procedures of type (5) depends on the proper ordering of the equations (and x_i , $i = 1, \dots, n$) in system (1).

In spite of this fact the following variant of the Nekrassov method is known [4]:

$$(14) \quad x_i^{k+1} = - \sum_{j=1}^{i-1} \frac{a_{ij}}{a_{ii}} x_j^k - \sum_{j=i+1}^n \frac{a_{ij}}{a_{ii}} x_j^{k+1} + \frac{b_i}{a_{ii}}.$$

Further, we are interested in the *successive overrelaxation procedure* (14) based on the method (7):

$$(15) \quad x_i^{k+1} = x_i^k (1 - \delta_i^k) - \delta_i^k \left(\sum_{j=1}^{i-1} \frac{a_{ij}}{a_{ii}} x_j^k + \sum_{j=i+1}^n \frac{a_{ij}}{a_{ii}} x_j^{k+1} - \frac{b_i}{a_{ii}} \right).$$

In matrix terms the *successive overrelaxation procedure* (15) can be written as follows:

$$(16) \quad X^{k+1} = (P + \delta^k R)^{-1} \left((I - \delta^k)P - \delta^k L \right) X^k + (P + \delta^k R)^{-1} \delta^k b.$$

The pseudocode for the modification of Nekrassov method (6) is given in Figure 1.

```

Choose an initial guess  $x^0$  for the solution  $x$ .
for  $k = 1, 2, \dots$ ,
    for  $i = 1, 2, \dots, n$ 
         $x_i = a_{ii}x_i^{k-1}$ 
         $N_i^{k-1} = 1$ 
        for  $j = 1, 2, \dots, i - 1$ 
             $N_i^{k-1} = N_i^{k-1}|x_i^{k-1} - x_j^k|$ 
             $x_i = x_i + a_{ij}x_j^k$ 
        end
        for  $j = i + 1, \dots, n$ 
             $N_i^{k-1} = N_i^{k-1}|x_i^{k-1} - x_j^{k-1}|$ 
             $x_i = x_i + a_{ij}x_j^{k-1}$ 
        end
         $x_i = (x_i - b_i)/N_i^{k-1}$ 
    end
     $x^k = x^{k-1} - x$ 
    check convergence; continue if necessary
end

```

Fig. 1. The modification of the Nekrassov method (6)

3. Numerical example. As an example we will consider the system:

$$\begin{cases} x_1 + 3x_2 - 2x_3 = 5 \\ 3x_1 + 5x_2 + 6x_3 = 7 \\ 2x_1 + 4x_2 + 3x_3 = 8 \end{cases}$$

The exact solution of the system is $x(-15, 8, 2)$.

For an initial approximation we choose $x^0(-15.02, 8.02, 2.02)$.

We give the results of numerical experiments (8 iterations) for each of methods (5) and (6).

In Table 1 the following notations are used:

- in the first column the serial number of the iteration is given;
- using the modified scheme (6) in the second column the obtained results are given (array $x[[]]$);
- using the classical Nekrassov scheme (5) in the third column the obtained results are given (array $y[[]]$).

Table 1

1	$X[1] = -15.02000000000000$ $X[2] = 8.01884259259259$ $X[3] = 2.01906701123844$	$Y[1] = -15.02000000000000$ $Y[2] = 7.98800000000000$ $Y[3] = 2.02933333333333$
2	$X[1] = -15.01999590828629$ $X[2] = 8.01776735852021$ $X[3] = 2.01820333133504$	$Y[1] = -14.90533333333333$ $Y[2] = 7.90800000000000$ $Y[3] = 2.05955555555556$
3	$X[1] = -15.01998800937772$ $X[2] = 8.01676825375272$ $X[3] = 2.01740388676229$	$Y[1] = -14.60488888888888$ $Y[2] = 7.69146666666666$ $Y[3] = 2.14797037037037$
4	$X[1] = -15.01997656688334$ $X[2] = 8.01583967575863$ $X[3] = 2.01666397500912$	$Y[1] = -13.77845925925925$ $Y[2] = 7.08951111111111$ $Y[3] = 2.39962469135803$
5	$X[1] = -15.01996182501415$ $X[2] = 8.01497643146312$ $X[3] = 2.01597923762914$	$Y[1] = -11.46928395061725$ $Y[2] = 5.40202074074072$ $Y[3] = 3.11016164609054$
6	$X[1] = -15.01994400998709$ $X[2] = 8.01417370750044$ $X[3] = 2.01534563522253$	$Y[1] = -4.98573893004107$ $Y[2] = 0.65924938271599$ $Y[3] = 5.11149344307273$
7	$X[1] = -15.01992333132901$ $X[2] = 8.01342704260065$ $X[3] = 2.01475942421623$	$Y[1] = 13.24523873799748$ $Y[2] = -12.68093537448576$ $Y[3] = 10.74442134064936$
8	$X[1] = -15.01989998308720$ $X[2] = 8.01273230196133$ $X[3] = 2.01421713531614$	$Y[1] = 64.53164880475601$ $Y[2] = -50.21229489163284$ $Y[3] = 26.59529398567311$

4. A wide area of problems and practical tasks in tomography and image processing are reduced to the problem of solving a system of algebraic equations with constraint conditions for the initial approximations x_i^0 , $i = 1, \dots, n$ (see Björck [2], A. van der Sluis and H. van der Vorst [16], A. Louis and F. Natterer [9] and R. Santos and A. de Pierro [15]).

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