Mathematical Modeling for Systems of Large Dimension Through a Modification of the Method of Iterative Aggregation

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Abstract

This article provides a new method for accelerating the convergence such that is a synthesis of two methods: Seidel method and an iterative method for solving a linear system of equations. Using a new method accelerating the convergence for various grades of operator equations allows to construct a sequence of approximations that more convergence to solve the operator equations. In this paper, the method allows to significantly speed up the process of convergence to the precise solution. There is an implementation of the resulting method and sufficient conditions of convergence.

Keywords: approximate methods, operator equations, spectral radius, Seidel method, method of iterative aggregation.

1 Introduction

A development of mankind result in speed up the amount of processed information, that's why it is required to solve optimization problems of large dimensions. This is particularly marked in areas such as bioengineering and economy. To meet the challenges of large dimension along with the increased processing power of computers it is necessary to develop new algorithms and methods to effectively use computing resources and to obtain solutions to previously unsolvable tasks.

Many algorithms and methods has been developed for solving various optimization integrated problems. Many practical problems are characterized by high dimensionality and a presence of difficulty formalizable restrictions. Both the formalization as well as numerical solution of problem of decision-making in planning, allocation of resources, optimize complex processes in various application cause the known issues [Arh75], [Dud79], [Gol04].

Significant difficulties, inter alia, connected with the large dimensionality of problems, which is not solved by traditional ways, so develop the new approaches of solution involving powerful computing resources is therefore required.

It is required to find the precise solution of linear and nonlinear algebraic equations systems for solving many problems of analysis and algebra. Typically, the modeling of complex systems confronted with large dimension tasks, a considerable number of internal linkages and different stochastic properties. The modeling process is difficult to implement at a sufficiently large number of unknown parameters. In such case, it is required to use the various methods to find the precise solution by using iterative processes. One of the simplest and well-known approximate method is Seidel method, which is a derivative of the method of successive approximations.

In the papers [Kra69] and [Gro11], authors provided the Seidel method and various modification of it. Consider now a one of the possible modification of Seidel method for solving the linear algebraic equations systems.

Let *x* be an operator equation of the form

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In: S. Hölldobler, A. Malikov, C. Wernhard (eds.): YSIP2 – Proceedings of the Second Young Scientist's International Workshop on Trends in Information Processing, Dombai, Russian Federation, May 16–20, 2017, published at http://ceur-ws.org.

$$x = Ax + f \tag{1}$$

where x be an unknown vector of Banach space E, f be a prescribed vector of E, A be a linear operator in the Banach space E.

Let the operator A be as a sum of operators A_1 and A_2

$$A = A_1 + A_2 \tag{2}$$

If there is an operator inverse to the operator $(I-A_1)$, then the equation (1) will be such that

$$x = (I - A_1)^{-1} \cdot A_2 x + (I - A_1)^{-1} \cdot f$$
(3)

Using the method of successive approximations to (3), we get

$$x_{m+1} = (I - A_1)^{-1} A_2 x_m + (I - A_1)^{-1} f, \qquad (m = 1, 2, ...)$$
(4)

There has been a significant acceleration of convergence to the solution compared with the method of successive approximations when solving the equation (1) by Seidel method [MEU99].

This article provides a modification of method of one-parameter iterative aggregation: the solution for equation (3) will search by using method of one-parameter iterative aggregation and we get a new method of accelerating for convergence in the result.

2 Approach

There were the methods of one-parameter and multiparameter aggregation, and their conditions of convergence in the paper [GRO14].

Consider now the convergence for method of one-parameter aggregation to solve an integral equation (1) with integral operator A

$$Ax(t) = \int_{0}^{1} K(t,s) x(s) \, ds$$

where a kernel K(t,s) be an nonnegative measurable function and f(t) be a nonnegative constant term, when the kernel K(t,s) is enough small.

In paper [GRO14], if the conditions (5) and (6) are true, the method of one-parameter aggregation is just an equation, if "aggregation" functionality is (7).

$$\|A\| \le \sup_{0 \le t \le 1} \int_{0}^{1} K(t,s) \, ds = c < 1 \tag{5}$$

$$\frac{(2-c)c}{(1-c)^2} < 1 \tag{6}$$

$$l_0[x(t)] = \int_0^1 x(s) \, ds \tag{7}$$

In paper [GRO14] the condition of convergence method of one-parameter aggregation is true, if the parameter c in the inequality (5) be

$$c < 1 - \frac{\sqrt{2}}{2} \approx 0,292 \tag{8}$$

Moreover, the method ensures convergence to the precise solution with denominator q < 1 of geometric progression. The condition (8) is a sufficient condition for convergence of method of one-parameter aggregation and the experimental material shows that the convergence of method is relevant at less stringent requirements on the smallness of the kernel.

The close condition to the condition (8) on the smallness of the rules of the matrix operator A be obtained as a sufficient condition for convergence of method of one-parameter iterative aggregation in solving linear system of algebraic equations. Consider now an algorithm of method for solving linear system of algebraic equations.

Suppose that A be a nonnegative matrix and it is required to solve the following set of equations:

$$x_{i} = \sum_{j=1}^{n} a_{ij} x_{j} + f_{i} \qquad (n = 1, n)$$
⁽⁹⁾

Let $x=x_1$ for solution x^* of equation (9) and $l_0(x)$ be a functionality:

$$l_0(x) = \sum_{i=1}^n x_i \qquad (x = (x_1, x_2, ..., x_n))$$

Using (1), we get the equation (10) with an unknown scalar *t*. Let the condition (11) will be fulfilled, if for any $\alpha \in [0, 1)$ a condition (12) is true where A^* be a transpose for matrix *A*:

$$t l_0(x_1) = t l_0(Ax_1) + l_0(f)$$
⁽¹⁰⁾

$$l_0(x_1) - l_0(Ax_1) \neq 0 \tag{11}$$

$$A^* l_0 \le \alpha l_0 \qquad \qquad \alpha < 1 \tag{12}$$

If the condition (12) is true then the equation (10) has a unique solution (13) $t=t(x_1)$ and this solution will be positive, if the vector f is positive $f: f \ge \theta$.

$$t(x_1) = \frac{l_0(f)}{l_0(x_1) - l_0(Ax_1)} = \frac{l_0(f)}{(l_0 - A^* l_0)(x_1)}$$
(13)

When the value $t=t(x_1)$ is founded, let find an element x_2 by using the formula:

$$x_2 = t(x_1) A x_1 + f \tag{14}$$

Using induction, we have the assertion (15):

$$x_{m+1} = t(x_m) A x_m + f$$
 (m=1,2,...) (15)

This means that the method of one-parameter iterative aggregation is to construct a sequence $\{x_m\}$.

In papers [GRO14] and [Gro11], author's computational experiments indicate, that the method of one-parameter iterative aggregation, firstly, isn't depend from functionality, secondly, the method converges fast enough for relatively large values $\rho(A)$. It is interesting, that the method converges for $\rho(A)$ be close to the unit and greater than the unit as opposed to the method of successive approximations, which isn't convergence for $\rho(A) \ge 1$. Therefore, let interpret Seidel method to use the aggregation coefficient to the both part of equation (7):

$$tl_0(x_1) = tl_0[(I - A_1)^{-1} \cdot A_2 x_1] + l_0[(I - A_1)^{-1} \cdot f]$$
(16)

where *t* be an unknown scalar.

Let we have the condition (17)

$$l_0(x_1) - l_0 \left[\left(\left(I - A_1 \right)^{-1} A_2 x_1 \right) \right] \neq 0$$
(17)

If the condition (17) is true, that the equation (16) has a unique solution $t=t(x_1)$ and $t(x_1)$ be

$$t(x_1) = \frac{l_0 \left[\left(I - A_1 \right)^{-1} \cdot f \right]}{l_0 (x_1) - l_0 \left[\left(I - A_1 \right)^{-1} A_2 x_1 \right]}$$
(18)

Finally, let find the element *x*₂:

$$x_2 = t(x_1) A x_1 + f$$
 (19)

Using induction, we have the assertion (20):

$$x_{m+1} = t(x_m) A x_m + f$$
 (m=1,2,...) (20)

These computational experiments indicate that the proposed modification method of one-parameter iterative aggregation has the higher speed of convergence than Seidel method for the large values of spectral radius $\rho(A)$ and the faster than the method of successive approximations.

3 Usage Examples

For the examples from 1 to 3, an A be a prescribed nonnegative matrix of order n, f be prescribed vector and x_1 be a start value.

3.1 The first example

Let consider the system of linear algebraic equations

$$\begin{cases} x = 0.15 \cdot x + 0.2 \cdot y + 0.31 \cdot z + 0.4 \\ y = 0.2 \cdot x + 0.33 \cdot y + 0.4 \cdot z + 0.1 \\ z = 0.3 \cdot x + 0.2 \cdot y + 0.14 \cdot z + 0.1 \end{cases}$$

Table 1: Start parameters

Matrix A	$\rho(A)$	Free vector f	Precise solution x^*	Initial value <i>x</i> ₁
$\begin{pmatrix} 0.15 & 0.2 & 0.31 \\ 0.2 & 0.03 & 0.4 \\ 0.3 & 0.2 & 0.14 \end{pmatrix}$	0.644	$f = \begin{pmatrix} 0.4\\ 0.1\\ 0.1 \end{pmatrix}$	$\begin{pmatrix} 0.756 \\ 0.460 \\ 0.487 \end{pmatrix}$	$\begin{pmatrix} 1\\1\\1 \end{pmatrix}$

Compare the results of the given system on method of successive approximations, Seidel method and modification of method of one-parameter iterative aggregation. We get the following results.

Table 2: Method of Successive Approximations

Values\Number of iteration	n=1	n=10	n=20	n=25	n=30
$x_1^{(n)}$	1.06	0.762	0.757	0.757	0.756
$x_{2}^{(n)}$	0.73	0.465	0.46	0.46	0.46
$x_{3}^{(n)}$	0.74	0.492	0.487	0.487	0.487

Table 3: Seidel Method

Values\Number of					
iteration	<i>n</i> =1	<i>n</i> =10	<i>n</i> =15	<i>n</i> =20	n=21
$x_1^{(n)}$	1.06	0.757	0.757	0.757	0.756
$x_2^{(n)}$	0.742	0.461	0.46	0.46	0.46
$x_{3}^{(n)}$	0.706	0.488	0.487	0.487	0.487

Let consider the modification of method of iterative aggregation - «mixed method».

Table 4: Mixed Method

Values\Number of iteration	n=1	n=2	n=4	n=5
$x_1^{(n)}$	0.816	0.749	0.757	0.756
$x_2^{(n)}$	0.497	0.451	0.46	0.46
$x_3^{(n)}$	0.503	0.483	0.487	0.487

We have the precise solution on the fifth iteration.

As can be seen, the mixed method provides the convergence better in comparison with considerable methods.

3.2 The Second Example

Let consider the system of forth equations with forth unknown variables.

 $\begin{cases} x_1 = 0.1 \cdot x_1 + 0.2 \cdot x_2 + 0.3 \cdot x_3 + 0.2 \cdot x_4 + 0.1 \\ x_2 = 0.2 \cdot x_1 + 0.3 \cdot x_2 + 0.4 \cdot x_3 + 0.2 \cdot x_4 + 0.1 \\ x_3 = 0.3 \cdot x_1 + 0.2 \cdot x_2 + 0.1 \cdot x_3 + 0.2 \cdot x_4 + 0.1 \\ x_4 = 0.1 \cdot x_1 + 0.1 \cdot x_2 + 0.1 \cdot x_3 + 0.1 \cdot x_4 + 0.1 \end{cases}$

Let $f = \begin{pmatrix} 0.1 \\ 0.1 \\ 0.1 \\ 0.1 \end{pmatrix}$ be a free vector, $x_1 = \begin{pmatrix} 0.1 \\ 0.1 \\ 0.1 \\ 0.1 \end{pmatrix}$ be a start value. We get the results:

Table 5: Start Parameters

Matrix A	$\rho(A)$	Free vector f	The precise	Initial
			solution x^*	value x_1
$\begin{pmatrix} 0.1 & 0.2 & 0.3 & 0.2 \\ 0.2 & 0.3 & 0.4 & 0.2 \\ 0.3 & 0.2 & 0.1 & 0.2 \\ 0.1 & 0.1 & 0.1 & 0.1 \end{pmatrix}$	0.787	$f = \begin{pmatrix} 0.1 \\ 0.1 \\ 0.1 \\ 0.1 \end{pmatrix}$	$\begin{pmatrix} 0.471\\ 0.629\\ 0.471\\ 0.286 \end{pmatrix}$	$\begin{pmatrix} 0.1\\ 0.1\\ 0.1\\ 0.1 \end{pmatrix}$

Table 6: Method of Successive Approximations

Values\Number of iteration	<i>n</i> =1	n=10	n=20	n=30	n=33
$x_1^{(n)}$	0.18	0.437	0.468	0.471	0.471

x ₂ ⁽ⁿ⁾	0.21	0.58	0.624	0.628	0.629
x ₃ ⁽ⁿ⁾	0.18	0.438	0.468	0.471	0.471
$x_4^{(n)}$	0.14	0.269	0.284	0.286	0.286

Table 7: Seidel method

Values\Number of					
iteration	<i>n</i> =1	<i>n</i> =10	<i>n</i> =15	n=16	<i>n</i> =17
$x_1^{(n)}$	0.296	0.461	0.47	0.47	0.471
$x_2^{(n)}$	0.196	0.615	0.626	0.627	0.629
$x_{3}^{(n)}$	0.178	0.463	0.47	0.471	0.471
$x_{4}^{(n)}$	0.150	0.282	0.285	0.285	0.286

Table 8: Mixed method

Values\Number of			
iteration	<i>n</i> =1	<i>n</i> =2	<i>n=3</i>
$x_1^{(n)}$	0.498	0.471	0.471
$x_{2}^{(n)}$	0.648	0.628	0.629
$x_3^{(n)}$	0.498	0.471	0.471
$x_4^{(n)}$	0.299	0.285	0.286

In this case, the mixed method has a priority of thirty iteration order as compared with the method of successive approximations and a priority of fourteen iterations as compared with Seidel method.

3.3 The Third Example

Let consider the system of fifth equations with five unknown variables.

	$x_1 = 0.1 \cdot x_1 + x_1 + x_2 + x_2 + x_3 + x_4 + x_4 + x_5 $	$0.2 \cdot x_2 + $	$0.1 \cdot x_3 + $	$0.2 \cdot x_4 + $	$0.1 \cdot x_5 + 0.1$
	$x_2 = 0.2 \cdot x_1 + $	$0.1 \cdot x_2 + $	$0.2 \cdot x_3 +$	$0.1 \cdot x_4 + $	$0.2 \cdot x_5 + 0.1$
•	$x_3 = 0.1 \cdot x_1 + $	$0.1 \cdot x_2 + $	$0.2 \cdot x_3 +$	$0.2 \cdot x_4 +$	$0.1 \cdot x_5 + 0.1$
	$x_4 = 0.2 \cdot x_1 +$	$0.2 \cdot x_2 +$	$0.1 \cdot x_3 +$	$0.1 \cdot x_4 +$	$0.2 \cdot x_5 + 0.1$
	$x_5 = 0.1 \cdot x_1 + 1$	$0.1 \cdot x_2 + $	$0.1 \cdot x_3 +$	$0.1 \cdot x_4 +$	$0.1 \cdot x_5 + 0.1$

Table 9: Start Parameters

Matrix A	$\rho(A)$	Free vector f	The precise solution	Initial value x_1
			x^*	
$\begin{pmatrix} 0.1 & 0.2 & 0.1 & 0.2 & 0.1 \\ 0.2 & 0.1 & 0.2 & 0.1 & 0.2 \\ 0.1 & 0.1 & 0.2 & 0.2 & 0.1 \\ 0.2 & 0.2 & 0.1 & 0.1 & 0.2 \\ 0.1 & 0.1 & 0.1 & 0.1 & 0.1 \end{pmatrix}$	0.7	$f = \begin{pmatrix} 0.1\\ 0.1\\ 0.1\\ 0.1\\ 0.1 \end{pmatrix}$	$\begin{pmatrix} 0.339\\ 0.361\\ 0.337\\ 0.363\\ 0.267 \end{pmatrix}$	$\begin{pmatrix} 0.1 \\ 0.1 \\ 0.1 \\ 0.1 \\ 0.1 \\ 0.1 \end{pmatrix}$

We get the following results:

Table 10: Method of Successive Approximations

Values\Number of					
iteration	<i>n</i> =1	<i>n</i> =5	<i>n</i> =10	<i>n</i> =20	<i>n</i> =24
$x_1^{(n)}$	0.8	0.455	0.359	0.34	0.339
$x_{2}^{(n)}$	0.9	0.486	0.382	0.361	0.361
$x_{3}^{(n)}$	0.8	0.452	0.356	0.337	0.337
$x_{4}^{(n)}$	0.9	0.49	0.384	0.364	0.363
$x_{5}^{(n)}$	0.6	0.347	0.280	0.267	0.267

Table 11: Seidel method

Values\Number of					
iteration	<i>n</i> =1	<i>n</i> =5	n=10	n=14	n=15
$x_1^{(n)}$	0.776	0.345	0.342	0.339	0.339
$x_{2}^{(n)}$	0.81	0.366	0.364	0.361	0.361
$x_{3}^{(n)}$	0.721	0.341	0.339	0.337	0.337
$x_4^{(n)}$	0.809	0.367	0.366	0.364	0.363
$x_{5}^{(n)}$	0.534	0.269	0.268	0.267	0.267

Values\Number of				
iteration	n=1	n=2	<i>n=3</i>	n=4
$x_1^{(n)}$	0.34	0.341	0.339	0.339
$x_{2}^{(n)}$	0.374	0.361	0.361	0.361
$x_{3}^{(n)}$	0.34	0.337	0.336	0.337
$x_4^{(n)}$	0.374	0.364	0.363	0.363
$x_{5}^{(n)}$	0.271	0.267	0.267	0.267

As can be seen, four iterations of method of one-parameter iterative aggregation result in precise solution to within 10^{-4} , while the method of successive approximations convergences on twenty forth iteration and Seidel method on fifteens iteration.

4 Conclusion

These computational experiments indicate that the proposed modification method of one-parameter iterative aggregation has the higher speed of convergence than Seidel method for the large values of spectral radius $\rho(A)$ and the faster than the method of successive approximations.

The above mentioned method can significantly speed up the convergence to the precise solution, that's a big advantage.

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