

Research Article

New Predictor-Corrector Methods with High Efficiency for Solving Nonlinear Systems

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Received 20 July 2012; Revised 27 August 2012; Accepted 1 September 2012

Academic Editor: Fazlollah Soleymani

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A new set of predictor-corrector iterative methods with increasing order of convergence is proposed in order to estimate the solution of nonlinear systems. Our aim is to achieve high order of convergence with few Jacobian and/or functional evaluations. Moreover, we pay special attention to the number of linear systems to be solved in the process, with different matrices of coefficients. On the other hand, by applying the pseudocomposition technique on each proposed scheme we get to increase their order of convergence, obtaining new efficient high-order methods. We use the classical efficiency index to compare the obtained procedures and make some numerical test, that allow us to confirm the theoretical results.

1. Introduction

Many relationships in nature are inherently nonlinear, which according to these effects are not in direct proportion to their cause. Approximating a solution ξ of a nonlinear system, $F(x) = 0$, is a classical problem that appears in different branches of science and engineering (see, e.g. [1]). In particular, the numerical solution of nonlinear equations and systems is needed in the study of dynamical models of chemical reactors [2] or in radioactive transfer [3]. Moreover, many of numerical applications use high precision in their computations; in [4], high-precision calculations are used to solve interpolation problems in astronomy; in [5] the authors describe the use of arbitrary precision computations to improve the results obtained in climate simulations; the results of these numerical experiments show that the high-order methods associated with a multiprecision arithmetic floating point are very useful, because it yields a clear reduction in iterations. A motivation for an arbitrary precision

in interval methods can be found in [6], in particular for the calculation of zeros of nonlinear functions.

Recently, many robust and efficient methods with high convergence order have been proposed to solve nonlinear equations, but in most of cases the schemes cannot be extended to multivariate problems. Few papers for the multidimensional case introduce methods with high order of convergence. The authors design in [7] a modified Newton-Jarrat scheme of sixth order; in [8] a third-order method is presented for computing real and complex roots of nonlinear systems; Shin et. al. compare in [9] Newton-Krylov methods and Newton-like schemes for solving big-sized nonlinear systems; the authors in [10] and A. Iliev and I. Iliev in [11] show general procedures to design high-order methods by using frozen Jacobian and Taylor expansion, respectively. Special case of sparse Jacobian matrices is studied in [12].

Dayton et al. in [13] formulate the multiplicity for the general nonlinear system at an isolated zero. They present an algorithm for computing the multiplicity structure, propose a depth-deflation method for accurate computation of multiple zeros, and introduce the basic algebraic theory of the multiplicity.

In this paper, we present three new Newton-like schemes, of order of convergence four, six, and eight, respectively. After the analysis of convergence of the new methods, we apply the pseudocomposition technique in order to get higher-order procedures. This technique (see [14]) consists of the following: we consider a method of order of convergence p as a predictor, whose penultimate step is of order q , and then we use a corrector step based on the Gaussian quadrature. So, we obtain a family of iterative schemes whose order of convergence is $\min\{q + p, 3q\}$. This is a general procedure to improve the order of convergence of known methods.

To analyze and compare the efficiency of the proposed methods we use the classic efficiency index $I = p^{1/d}$ due to Ostrowski [15], where p is the order of convergence and d is the number of functional evaluations at each iteration.

The convergence theorem in Section 2 is demonstrated by means of the n -dimensional Taylor expansion of the functions involved. Let $F : D \subseteq R^n \rightarrow R^n$ be sufficiently Frechet differentiable in D . By using the notation introduced in [7], the q th derivative of F at $u \in R^n$, $q \geq 1$, is the q -linear function $F^{(q)}(u) : R^n \times \dots \times R^n \rightarrow R^n$ such that $F^{(q)}(u)(v_1, \dots, v_q) \in R^n$. It is easy to observe that

- (1) $F^{(q)}(u)(v_1, \dots, v_{q-1}, \cdot) \in \mathcal{L}(R^n)$,
- (2) $F^{(q)}(u)(v_{\sigma(1)}, \dots, v_{\sigma(q)}) = F^{(q)}(u)(v_1, \dots, v_q)$, for all permutation σ of $\{1, 2, \dots, q\}$.

So, in the following we will denote:

- (a) $F^{(q)}(u)(v_1, \dots, v_q) = F^{(q)}(u)v_1 \cdots v_q$,
- (b) $F^{(q)}(u)v^{q-1}F^{(p)}v^p = F^{(q)}(u)F^{(p)}(u)v^{q+p-1}$.

It is well known that, for $\xi + h \in R^n$ lying in a neighborhood of a solution ξ of the nonlinear system $F(x) = 0$, Taylor's expansion can be applied (assuming that the jacobian matrix $F'(\xi)$ is nonsingular), and

$$F(\xi + h) = F'(\xi) \left[h + \sum_{q=2}^{p-1} C_q h^q \right] + O[h^p], \quad (1.1)$$

where $C_q = (1/q!)[F'(\xi)]^{-1}F^{(q)}(\xi)$, $q \geq 2$. We observe that $C_q h^q \in R^n$ since $F^{(q)}(\xi) \in \mathcal{L}(R^n \times \dots \times R^n, R^n)$ and $[F'(\xi)]^{-1} \in \mathcal{L}(R^n)$.

In addition, we can express the Jacobian matrix of F, F' as

$$F'(\xi + h) = F'(\xi) \left[I + \sum_{q=2}^{p-1} qC_q h^{q-1} \right] + O[h^p], \quad (1.2)$$

where I is the identity matrix. Therefore, $qC_q h^{q-1} \in \mathcal{L}(R^n)$. From (1.2), we obtain

$$[F'(\xi + h)]^{-1} = \left[I + X_2 h + X_3 h^2 + X_4 h^3 + \dots \right] [F'(\xi)]^{-1} + O[h^p], \quad (1.3)$$

where $X_2 = -2C_2, X_3 = 4C_2^2 - 3C_3, \dots$

We denote $e_k = x^{(k)} - \xi$ the error in the k th iteration. The equation $e_{(k+1)} = L e_k^p + O[e_k^{p+1}]$, where L is a p -linear function $L \in \mathcal{L}(R^n \times \dots \times R^n, R^n)$, is called the *error equation* and p is the *order of convergence*.

The rest of the paper is organized as follows: in the next section, we present the new methods of order four, six, and eight, respectively. Moreover, the convergence order is increased when the pseudocomposition technique is applied. Section 3 is devoted to the comparison of the different methods by means of several numerical tests.

2. Design and Convergence Analysis of the New Methods

Let us introduce now a new Jarratt-type scheme of five steps which we will denote as M8. We will prove that its first three steps define a fourth-order scheme, denoted by M4, and its four first steps become a sixth-order method that will be denoted by M6. The coefficients involved have been obtained optimizing the order of the convergence, and the whole scheme requires three functional evaluations of F and two of F' to attain eighth order of convergence. Let us also note that the linear systems to be solved in first, second, and last step have the same matrix and also have the third and fourth steps, so the number of operations involved is not as high as it can seem.

Theorem 2.1. *Let $F : \Omega \subseteq R^n \rightarrow R^n$ be a sufficiently differentiable in a neighborhood of $\xi \in \Omega$ which is a solution of the nonlinear system $F(x) = 0$, and let $x^{(0)}$ be an initial estimation close enough to the solution ξ . One also supposes that $F'(x)$ is continuous and nonsingular at ξ . Then, the sequence $\{x^{(k)}\}_{k \geq 0}$ obtained by*

$$\begin{aligned} y^{(k)} &= x^{(k)} - \frac{2}{3} [F'(x^{(k)})]^{-1} F(x^{(k)}), \\ z^{(k)} &= y^{(k)} + \frac{1}{6} [F'(x^{(k)})]^{-1} F(x^{(k)}), \\ u^{(k)} &= z^{(k)} + [F'(x^{(k)}) - 3F'(y^{(k)})]^{-1} F(x^{(k)}), \\ v^{(k)} &= z^{(k)} + [F'(x^{(k)}) - 3F'(y^{(k)})]^{-1} [F(x^{(k)}) + 2F(u^{(k)})] \\ x^{(k+1)} &= v^{(k)} - \frac{1}{2} [F'(x^{(k)})]^{-1} [5F'(x^{(k)}) - 3F'(y^{(k)})] [F'(x^{(k)})]^{-1} F(v^{(k)}) \end{aligned} \quad (2.1)$$

converges to ξ with order of convergence eight. The error equation is

$$e_{k+1} = \left(C_2^2 - \frac{1}{2}C_3 \right) \left(2C_2^3 + 2C_3C_2 - 2C_2C_3 - \frac{20}{9}C_4 \right) e_k^8 + O[e_k^9]. \quad (2.2)$$

Proof. From (1.1) and (1.2) we obtain

$$\begin{aligned} F(x^{(k)}) &= F'(\xi) \left[e_k + C_2 e_k^2 + C_3 e_k^3 + C_4 e_k^4 + C_5 e_k^5 + C_6 e_k^6 + C_7 e_k^7 + C_8 e_k^8 \right] + O[e_k^9], \\ F'(x^{(k)}) &= F'(\xi) \left[I + 2C_2 e_k + 3C_3 e_k^2 + 4C_4 e_k^3 + 5C_5 e_k^4 + 6C_6 e_k^5 + 7C_7 e_k^6 + 8C_8 e_k^7 \right] + O[e_k^8]. \end{aligned} \quad (2.3)$$

As $[F'(x^{(k)})]^{-1} F'(x^{(k)}) = I$, we calculate

$$\left[F'(x^{(k)}) \right]^{-1} = \left[I + X_2 e_k + X_3 e_k^2 + X_4 e_k^3 + X_5 e_k^4 + X_6 e_k^5 + X_7 e_k^6 + X_8 e_k^7 \right] [F'(\xi)]^{-1} + O[e_k^8], \quad (2.4)$$

where $X_1 = I$ and $X_s = -\sum_{j=2}^s j X_{s-j+1} C_j$, for $s = 2, 3, \dots$. So,

$$\left[F'(x^{(k)}) \right]^{-1} F(x^{(k)}) = e_k + M_2 e_k^2 + M_3 e_k^3 + M_4 e_k^4 + M_5 e_k^5 + M_6 e_k^6 + M_7 e_k^7 + M_8 e_k^8 + O[e_k^9], \quad (2.5)$$

where $M_2 = C_2 + X_4$ and $M_s = C_s + \sum_{j=3}^s X_{s-j+2} C_{j-1} + X_{s+2}$, $s = 3, 4, \dots$

Then, $y^{(k)} = \xi + (1/3)e_k - (2/3)M$ and $z^{(k)} = \xi + (1/2)e_k - (1/2)M$, where $M = M_2 e_k^2 + M_3 e_k^3 + M_4 e_k^4 + M_5 e_k^5 + M_6 e_k^6 + M_7 e_k^7 + M_8 e_k^8 + O[e_k^9]$.

The Taylor expansion of $F'(y^{(k)})$ is

$$F'(y^{(k)}) = F'(\xi) \left[I + Q_1 e_k + Q_2 e_k^2 + Q_3 e_k^3 + Q_4 e_k^4 + Q_5 e_k^5 + Q_6 e_k^6 + Q_7 e_k^7 + Q_8 e_k^8 \right] + O[e_k^9], \quad (2.6)$$

where

$$\begin{aligned}
Q_1 &= \frac{2}{3}C_2, \\
Q_2 &= \frac{1}{3}C_3 - \frac{4}{3}C_2M_2, \\
Q_3 &= \frac{4}{27}C_4 - \frac{4}{3}C_3M_2 - \frac{4}{3}C_2M_3, \\
Q_4 &= \frac{5}{81}C_5 - \frac{8}{9}C_4M_2 + \frac{4}{3}C_3(M_2^2 - 2M_3) - \frac{4}{3}C_2M_4, \\
Q_5 &= \frac{2}{81}C_6 - \frac{40}{81}C_5M_2 + \frac{4}{9}C_4(M_2^2 - M_3) + \frac{4}{3}C_3(M_2M_3 + M_3M_2 - M_4) - \frac{4}{3}C_2M_5, \\
Q_6 &= \frac{7}{729}C_7 - \frac{20}{81}C_6M_2 + \frac{40}{81}C_5(3M_2^2 - M_3) + \frac{8}{27}C_4(6M_2M_3 + 6M_3M_2 - 3M_4 - 4M_2^3) - C_2M_6, \\
Q_7 &= \frac{4}{3}C_2M_7 + \frac{4}{3}C_3(M_2M_5 + M_3M_4 + M_4M_3 + M_5M_2 + M_6M_2 - M_7) \\
&\quad + \frac{8}{27}C_4(6M_2M_4 + 6M_3^2 + 6M_4M_2 - 3M_5 - 4M_2^2M_3 - 4M_2M_3M_2 - 4M_3M_2^2) \\
&\quad + \frac{40}{81}C_5(3M_2M_3 + 3M_3M_2 - M_4 - 4M_2^3) + \frac{20}{81}C_6(4M_2^2 - M_3) - \frac{28}{243}C_7M_2 + \frac{8}{2187}C_8, \\
Q_8 &= -\frac{4}{3}C_2M_8 + \frac{4}{3}C_3(M_2M_6 + M_3M_5 + M_4^2 + M_5M_3 + M_6M_2 - M_7) \\
&\quad + \frac{16}{9}C_4(M_2M_5 + M_3M_4 + M_4M_3 + M_5M_2 - M_6) \\
&\quad + \frac{32}{27}C_4(M_2^2M_4 + M_2M_3^2M_4M_2 + M_3M_2M_3 + M_3^2M_2 + M_4M_2^2) \\
&\quad + \frac{40}{81}C_5(-M_5 - 4M_2^2M_3 - 4M_2M_3M_2 - 4M_3M_2^2 + 3M_2M_4 + 3M_3^2 + 3M_4M_2 + 2M_4) \\
&\quad + \frac{20}{41}C_6(4M_2M_3 + 4M_3M_2 - M_4 - 8M_2^3) \\
&\quad + \frac{28}{243}C_7(5M_2^2 - M_3) - \frac{112}{2187}C_8M_2 + \frac{1}{729}C_9.
\end{aligned} \tag{2.7}$$

We also obtain the Taylor expansion of $F'(x^{(k)}) - 3F'(y^{(k)})$:

$$\begin{aligned}
F'(x^{(k)}) - 3F'(y^{(k)}) &= F'(\xi) \left[-2I + A_1e_k + A_2e_k^2 + A_3e_k^3 + A_4e_k^4 + A_5e_k^5 + A_6e_k^6 + A_7e_k^7 + A_8e_k^8 \right] \\
&\quad + O[e_k^9],
\end{aligned} \tag{2.8}$$

where $A_s = (s+1)C_{s+1} - 3Q_s, s = 1, 2, \dots$. As $[F'(x^{(k)}) - 3F'(y^{(k)})]^{-1}[F'(x^{(k)}) - 3F'(y^{(k)})] = I$, we obtain

$$\begin{aligned} [F'(x^{(k)}) - 3F'(y^{(k)})]^{-1} &= \left[-\frac{1}{2}I + Y_2e_k + Y_3e_k^2 + Y_4e_k^3 + Y_5e_k^4 + Y_6e_k^5 + Y_7e_k^6 + Y_8e_k^7 \right] [F'(\xi)]^{-1} \\ &\quad + O[e_k^8], \end{aligned} \quad (2.9)$$

where $Y_2 = 0$ and $Y_s = (1/2) \sum_{j=3}^s Y_{s-j+2}A_{j-2} - (1/4)A_{s-1}, s = 3, 4, \dots$
So,

$$\begin{aligned} [F'(x^{(k)}) - 3F'(y^{(k)})]^{-1} F(x^{(k)}) &= -\frac{1}{2}e_k + R_2e_k^2 + R_3e_k^3 + R_4e_k^4 + R_5e_k^5 + R_6e_k^6 + R_7e_k^7 + R_8e_k^8 \\ &\quad + O[e_k^9], \end{aligned} \quad (2.10)$$

where $R_2 = Y_2 - (1/2)C_2$ and $R_s = Y_s + \sum_{j=3}^s Y_{s-j+2}C_{j-1} - (1/2)C_s, s = 3, 4, \dots$

We now calculate $u^{(k)} = z^{(k)} + [F'(x^{(k)}) - 3F'(y^{(k)})]^{-1}F(x^{(k)})$, and the error equation of the method at this step is

$$\begin{aligned} e_{u^{(k)}} &= \frac{1}{2}e_k - \frac{1}{2} \left[M_2e_k^2 + M_3e_k^3 + M_4e_k^4 + M_5e_k^5 + M_6e_k^6 + M_7e_k^7 + M_8e_k^8 \right] \\ &\quad - \frac{1}{2}e_k + R_2e_k^2 + R_3e_k^3 + R_4e_k^4 + R_5e_k^5 + R_6e_k^6 + R_7e_k^7 + R_8e_k^8 + O[e_k^9] \\ &= P_4e_k^4 + P_5e_k^5 + P_6e_k^6 + P_7e_k^7 + P_8e_k^8 + O[e_k^9], \end{aligned} \quad (2.11)$$

where $P_s = -(1/2)M_s + R_s, s = 4, 5, \dots$. Then the first three steps define a fourth-order procedure, and

$$\begin{aligned} F(u^{(k)}) &= F'(\xi) \left[P_4e_k^4 + P_5e_k^5 + P_6e_k^6 + P_7e_k^7 + (P_8 + P_4^2)e_k^8 \right] + O[e_k^9], \\ F(x^{(k)}) + 2F(u^{(k)}) &= F'(\xi) \left[e_k + C_2e_k^2 + C_3e_k^3 + (C_4 + 2P_4)e_k^4 + (C_5 + 2P_5)e_k^5 \right. \\ &\quad \left. + (C_6 + 2P_6)e_k^6 + (C_7 + 2P_7)e_k^7 + (C_8 + 2P_8)e_k^8 \right] + O[e_k^9]. \end{aligned} \quad (2.12)$$

So,

$$\begin{aligned} [F'(x^{(k)}) - 3F'(y^{(k)})]^{-1} [F(x^{(k)}) + 2F(u^{(k)})] &= -\frac{1}{2}e_k + L_2e_k^2 + L_3e_k^3 + L_4e_k^4 + L_5e_k^5 + L_6e_k^6 \\ &\quad + L_7e_k^7 + L_8e_k^8 + O[e_k^9], \end{aligned} \quad (2.13)$$

where $L_2 = Y_2 - (1/2)(C_2 + P_2)$ and $L_s = Y_s + \sum_{j=3}^s Y_{s-j+2}C_{j-1} - (1/2)(C_s + P_s)$, $s = 3, 4, \dots$

At the fourth step, we calculate

$$e_{v^{(k)}} = z_k - \xi + \left[F'(x^{(k)}) - 3F'(y^{(k)}) \right]^{-1} \left[F(x^{(k)}) + 2F(u^{(k)}) \right] = N_6 e_k^6 + N_7 e_k^7 + N_8 e_k^8 + O[e_k^9], \quad (2.14)$$

where $N_s = L_s - (1/2)M_s$ and $s = 6, 7, \dots$. This shows that the first four steps of the method define a sixth-order scheme. Indeed,

$$F(v^{(k)}) = F'(\xi) \left[N_6 e_k^6 + N_7 e_k^7 + N_8 e_k^8 \right] + O[e_k^9], \quad (2.15)$$

$$\begin{aligned} 5F'(x^{(k)}) - 3F'(y^{(k)}) &= F'(\xi) \left[2I + B_1 e_k + B_2 e_k^2 + B_3 e_k^3 + B_4 e_k^4 + B_5 e_k^5 + B_6 e_k^6 + B_7 e_k^7 + B_8 e_k^8 \right] \\ &\quad + O[e_k^9], \end{aligned} \quad (2.16)$$

where $B_s = 5(s+1)C_{s+1} - 3Q_s$ and $s = 1, 2, \dots$. Then,

$$\begin{aligned} \left[F'(x^{(k)}) \right]^{-1} \left[5F'(x^{(k)}) - 3F'(y^{(k)}) \right] &= 2I + S_1 e_k + S_2 e_k^2 + S_3 e_k^3 + S_4 e_k^4 + S_5 e_k^5 + S_6 e_k^6 \\ &\quad + S_7 e_k^7 + S_8 e_k^8 + O[e_k^9], \end{aligned} \quad (2.17)$$

$$\left[F'(x^{(k)}) \right]^{-1} F(v^{(k)}) = 2N_6 e_k^6 + (N_7 + X_2 N_6) e_k^7 + (N_8 + X_2 N_7 + X_3 N_6) e_k^8 + O[e_k^9], \quad (2.18)$$

where $S_1 = B_1 + 2X_2$ and $S_s = B_s + \sum_{j=2}^s X_{s-j+2} B_{j-1} + 2X_{s+1}$, $s = 2, 3, \dots$. By multiplying (2.17) and (2.18),

$$\left[F'(x^{(k)}) \right]^{-1} \left[5F'(x^{(k)}) - 3F'(y^{(k)}) \right] \left[F'(x^{(k)}) \right]^{-1} F(v^{(k)}) = 2N_6 e_k^6 + K_7 e_k^7 + K_8 e_k^8 + O[e_k^9], \quad (2.19)$$

where $K_7 = 2(N_7 + X_2 N_6) + S_1 N_6$ and $K_8 = 2(N_8 + X_2 N_7 + X_3 N_6) + S_1(N_7 + X_2 N_6) + S_2 N_6$.

Finally,

$$x^{(k+1)} = \xi + \left[L_8 - \frac{1}{2}M_8 - \frac{1}{2}(2(N_8 + X_2 N_7 + X_3 N_6) + S_1(N_7 + X_2 N_6) + S_2 N_6) \right] e_k^8 + O[e_k^9], \quad (2.20)$$

and the error equation is

$$e_{k+1} = \left(C_2^2 - \frac{1}{2}C_3 \right) \left(2C_2^3 + 2C_3 C_2 - 2C_2 C_3 - \frac{20}{9}C_4 \right) e_k^8 + O[e_k^9], \quad (2.21)$$

proving that the order of convergence of the analyzed method is eight. \square

Table 1: Values of the parameters for the different quadratures used.

Number of nodes	Quadratures							
	Chebyshev		Legendre		Lobatto		Radau	
	σ	σ_1	σ	σ_1	σ	σ_1	σ	σ_1
1	π	0	2	0	2	0	2	-1
2	π	0	2	0	2	0	2	0
3	π	0	2	0	2	0	2	0

In [14] the authors presented a new procedure to design higher-order schemes. This technique, called pseudocomposition, uses the two last steps of the predictor method to obtain a corrected scheme with higher order of convergence.

Theorem 2.2 (see [14]). *Let $F : \Omega \subseteq R^n \rightarrow R^n$ be differentiable enough Ω , let $\xi \in \Omega$ be a solution of the nonlinear system $F(x) = 0$, and let $x^{(0)}$ be an initial estimation close enough to the solution ξ . We suppose that $F'(x)$ is continuous and nonsingular at ξ . Let $y^{(k)}$ and $z^{(k)}$ be the penultimate and final steps of orders q and p , respectively, of a certain iterative method. Taking this scheme as a predictor we get a new approximation $x^{(k+1)}$ of ξ given by*

$$x^{(k+1)} = y^{(k)} - 2 \left[\sum_{i=1}^m \omega_i F'(\eta_i^{(k)}) \right]^{-1} F(y^{(k)}), \quad (2.22)$$

where $\eta_i^{(k)} = (1/2)[(1 + \tau_i)z^{(k)} + (1 - \tau_i)y^{(k)}]$ and $\tau_i, \omega_i, i = 1, 2, \dots, m$ are the nodes and weights of the orthogonal polynomial corresponding to the Gaussian quadrature used. Then,

- (1) the obtained set of families will have an order of convergence at least q ;
- (2) if $\sigma = 2$ is satisfied, then the order of convergence will be at least $2q$;
- (3) if, also, $\sigma_1 = 0$, the order of convergence will be $\min\{p + q, 3q\}$,

where $\sum_{i=1}^n \omega_i = \sigma$ and $\sum_{i=1}^n \omega_i \tau_i^j / \sigma = \sigma_j$ with $j = 1, 2$.

Depending on the orthogonal polynomial corresponding to the Gaussian quadrature used in the corrector step, this procedure will determine a family of schemes. Furthermore, it is possible to obtain different methods in these families by using distinct number of nodes corresponding to the orthogonal polynomial used (see Table 1). However, according to the proof of Theorem 2.2 the order of convergence of the obtained methods does not depend on the number of nodes used.

Let us note that these methods, obtained by means of Gaussian quadratures, seem to be known interpolation quadrature schemes such as midpoint, trapezoidal, or Simpson's method (see [16]). It is only a similitude, as they are not applied on the last iteration $x^{(k)}$, and the last step of the predictor $z^{(k)}$, but on the two last steps of the predictor. In the following, we will use a midpoint-like as a corrector step, which corresponds to a Gauss-Legendre quadrature with one node; for this scheme the order of convergence will be at least $\min\{q + p, 3q\}$, by applying Theorem 2.2.

The pseudocomposition can be applied to the proposed scheme M8 with iterative expression (2.1), but also to M6. By pseudocomposing on M6 and M8 there can be obtained two procedures of order of convergence 10 and 14 (denoted by PsM10 and PsM14),

Table 2: Numerical results for functions F_1 to F_4 .

Function	Method	Iter	Sol	$\ x^{(k)} - x^{(k-1)}\ $	$\ F(x^{(k)})\ $	ρ	e -time (sec)
F_1 $x^{(0)} = (0.8, \dots, 0.8)$	NC	8	ξ_1	$1.43e - 121$	$2.06e - 243$	2.0000	8.6407
	JT	4	ξ_1	$1.69e - 60$	$2.06e - 243$	4.0000	3.9347
	M4	4	ξ_1	$1.69e - 60$	$2.06e - 243$	4.0000	3.7813
	M6	4	ξ_1	$6.94e - 193$	$4.33e - 1160$	6.0000	5.3911
	M8	3	ξ_1	$9.40e - 50$	$3.51e - 4011$	8.0913	5.0065
	P sM10	3	ξ_1	$1.28e - 91$	$9.54e - 921$	10.0545	4.9061
	P sM14	3	ξ_1	$4.65e - 164$	0	14.0702	6.1018
F_1 $x^{(0)} = (0.0015, \dots, 0.0015)$	NC	17	ξ_1	$3.37e - 340$	$1.14e - 340$	—	9.2128
	JT	9	ξ_1	$8.18e - 085$	$1.14e - 340$	4.0000	10.1416
	M4	9	ξ_1	$8.18e - 085$	$1.14e - 340$	4.0000	10.9104
	M6	7	ξ_1	$1.40e - 035$	$9.46e - 216$	—	12.3266
	M8	19	ξ_1	$9.50e - 030$	$1.29e - 240$	—	59.4832
	P sM10	6	ξ_1	$3.02e - 102$	$5.23e - 1027$	—	17.9957
	P sM14	5	ξ_1	$1.84e - 162$	0	—	22.6130
F_2 $x^{(0)} = (-0.5, -0.5)$	NC	9	ξ_1	$2.45e - 181$	$5.92e - 362$	2.0148	0.2395
	JT	5	ξ_1	$9.48e - 189$	$8.13e - 754$	4.0279	0.3250
	M4	5	ξ_1	$9.48e - 189$	$8.13e - 754$	4.0279	0.1841
	M6	4	ξ_1	$1.34e - 146$	$2.14e - 878$	5.9048	0.2744
	M8	3	ξ_1	$1.90e - 038$	$1.23e - 302$	7.8530	0.3718
	P sM10	3	ξ_1	$6.72e - 72$	$2.68e - 714$	9.9092	0.4674
	P sM14	3	ξ_1	$2.13e - 122$	$1.95e - 1706$	13.9829	0.3187
F_2 $x^{(0)} = (-5, -3)$	NC	13	ξ_1	$2.20e - 182$	$2.73e - 374$	1.9917	0.3713
	JT	7	ξ_1	$2.10e - 179$	$4.51e - 716$	3.9925	0.4001
	M4	7	ξ_1	$2.10e - 179$	$4.51e - 716$	3.9925	0.7535
	M6	8	ξ_1	$2.55e - 036$	$5.81e - 216$	—	0.9382
	M8	>5000					
	P sM10	4	ξ_1	$2.59e - 021$	$3.51e - 208$	—	0.4363
	P sM14	29	ξ_2	$9.45e - 020$	$5.05e - 273$	—	7.8090
F_3 $x^{(0)} = (2, -3)$	NC	10	ξ_1	$1.65e - 190$	$4.61e - 380$	2.0000	1.4675
	JT	5	ξ_1	$8.03e - 113$	$7.59e - 450$	3.9995	0.3151
	M4	5	ξ_1	$8.03e - 113$	$7.59e - 450$	3.9995	0.3034
	M6	4	ξ_1	$1.25e - 082$	$2.83e - 493$	6.0015	0.3696
	M8	4	ξ_1	$1.54e - 162$	$3.16e - 1296$	7.9993	0.4463
	P sM10	3	ξ_1	$5.59e - 044$	$1.40e - 436$	9.4708	0.4682
	P sM14	3	ξ_1	$3.46e - 068$	$3.45e - 948$	13.1659	0.5925
F_3 $x^{(0)} = (0.2, 0.1)$	NC	35	ξ_1	$3.71e - 177$	$2.33e - 253$	—	1.4828
	JT	11	ξ_1	$3.29e - 143$	$1.67e - 574$	—	0.7781
	M4	11	ξ_1	$3.29e - 143$	$1.67e - 574$	—	0.7535
	M6	9	ξ_1	$1.31e - 064$	$3.61e - 385$	—	0.8001
	M8	n.c.	ξ_1				
	P sM10	5	ξ_1	$6.85e - 156$	$1.06e - 1555$	—	0.6352
	P sM14	8	ξ_2	$7.87e - 155$	0	—	1.1870

Table 2: Continued.

Function	Method	Iter	Sol	$\ x^{(k)} - x^{(k-1)}\ $	$\ F(x^{(k)})\ $	ρ	e -time (sec)
F_4 $x^{(0)} = (1, -1.5, -0.5)$	NC	10	ξ_1	$1.03e - 135$	$1.55e - 270$	1.9995	2.3263
	JT	5	ξ_1	$9.94e - 073$	$2.09e - 289$	4.0066	0.5296
	M4	5	ξ_1	$9.94e - 073$	$2.09e - 289$	4.0066	0.6340
	M6	4	ξ_1	$9.31e - 057$	$4.86e - 338$	5.9750	0.7443
	M8	4	ξ_1	$4.43e - 046$	$1.08e - 364$	—	0.8282
	Psm10	3	ξ_1	$1.43e - 031$	$1.04e - 311$	9.6674	0.8100
	Psm14	3	ξ_1	$1.91e - 033$	$4.05e - 462$	13.9954	1.0465
F_4 $x^{(0)} = (7, -5, -5)$	NC	12	ξ_1	$1.08e - 192$	$1.55e - 384$	1.9996	2.7271
	JT	6	ξ_1	$2.31e - 103$	$7.97e - 412$	4.0090	0.7761
	M4	6	ξ_1	$2.31e - 103$	$7.97e - 412$	4.0090	1.0301
	M6	5	ξ_1	$2.99e - 086$	$4.69e - 515$	—	1.0090
	M8	15	ξ_3	$1.77e - 071$	$1.48e - 568$	—	3.4007
	Psm10	4	ξ_1	$6.86e - 067$	$1.25e - 666$	—	1.0245
	Psm14	7	ξ_2	$1.09e - 130$	$9.15e - 1825$	—	1.8179

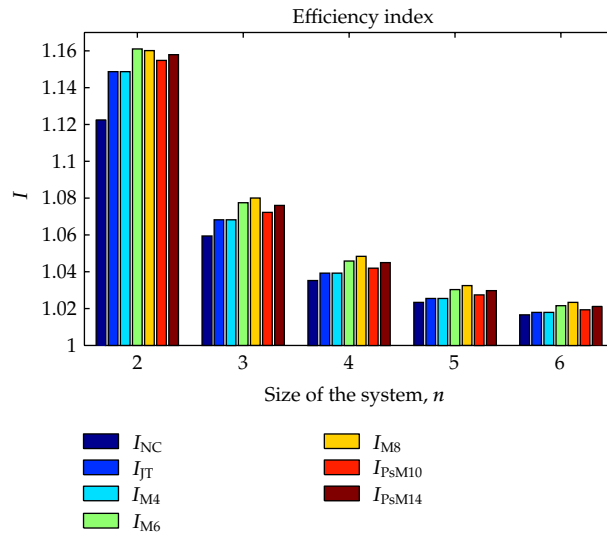


Figure 1: Efficiency index of the different methods for different sizes of the system.

respectively. Let us note that it is also possible to pseudocompose on M4, but the resulting scheme would be of third order of convergence, which is worse than the original M4, so it will not be considered.

Following the notation used in (2.1), the last step of Psm10 is

$$x^{(k+1)} = u^{(k)} - 2 \left[F' \left(\frac{v^{(k)} + u^{(k)}}{2} \right) \right]^{-1} F(u^{(k)}), \tag{2.23}$$

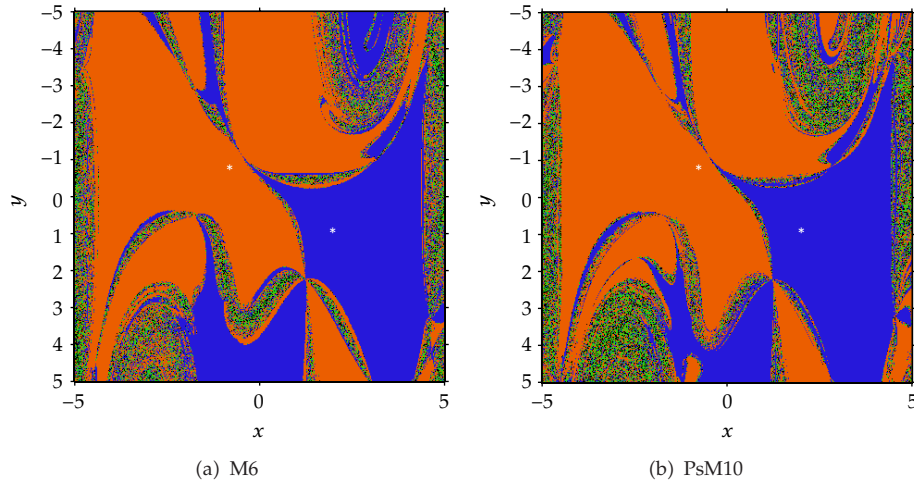


Figure 2: Real dynamical planes for system $F_2(x) = 0$ and methods M6 and PsM10.

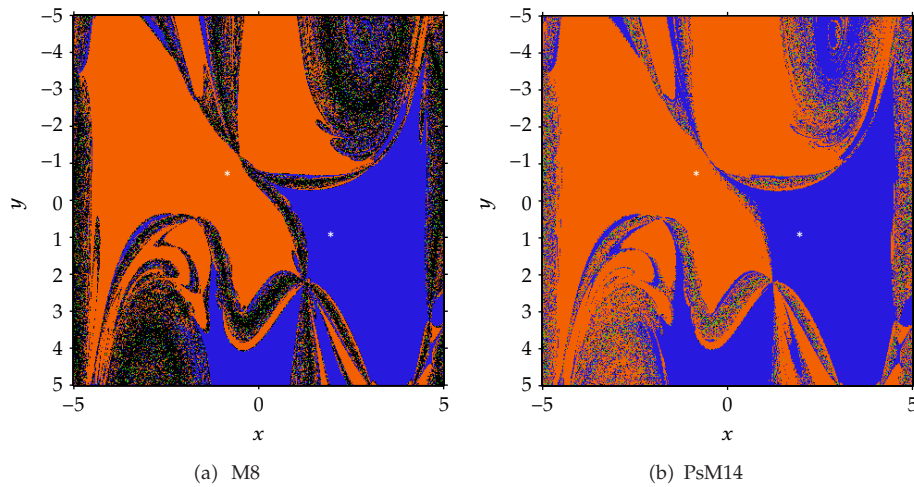


Figure 3: Real dynamical planes for system $F_2(x) = 0$ and methods M8 and PsM14.

and the last three steps of psM14 can be expressed as

$$\begin{aligned}
 v^{(k)} &= z^{(k)} + \left[F'(x^{(k)}) - 3F'(y^{(k)}) \right]^{-1} \left[F(x^{(k)}) + 2F(u^{(k)}) \right], \\
 w^{(k+1)} &= v^{(k)} - \frac{1}{2} \left[F'(x^{(k)}) \right]^{-1} \left[5F'(x^{(k)}) - 3F'(y^{(k)}) \right] \left[F'(x^{(k)}) \right]^{-1} F(v^{(k)}), \\
 x^{(k+1)} &= v^{(k)} - 2 \left[F' \left(\frac{w^{(k)} + v^{(k)}}{2} \right) \right]^{-1} F(v^{(k)}).
 \end{aligned} \tag{2.24}$$

In Figure 1, we analyze the efficiency indices of the proposed methods, compared with Newton and Jarrat's schemes and between themselves. There can be deduced the

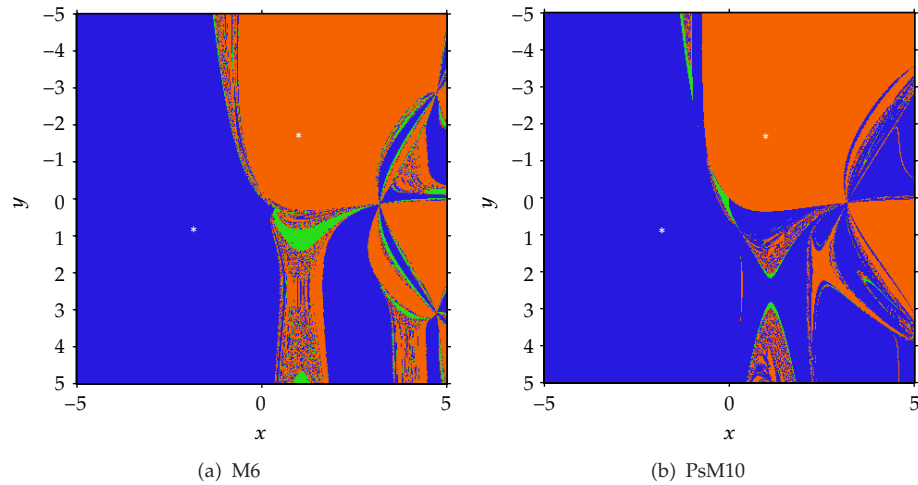


Figure 4: Real dynamical planes for system $F_3(x) = 0$ and methods M6 and PsM10.

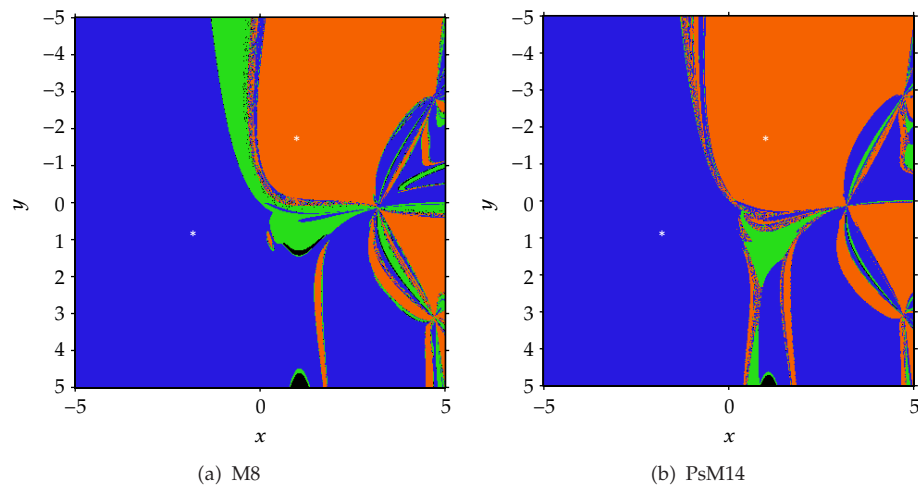


Figure 5: Real dynamical planes for system $F_3(x) = 0$ and methods M8 and PsM14.

following conclusions: the new methods M4, M6, and M8 (and also the pseudocomposed PsM10 and PsM14) improve Newton and Jarratt's schemes (in fact, the indices of M4 and Jarratt's are equal). Indeed, for $n \geq 3$ the best index is that of M8. Nevertheless, none of the pseudocomposed methods improve the efficiency index of their original partners. Nevertheless, as we will see in the following section, the pseudocomposed schemes will show a very stable behavior that makes them worth.

3. Numerical Results

In order to illustrate the effectiveness of the proposed methods, we will compare them with other known schemes. Numerical computations have been performed in MATLAB

R2011a by using variable-precision arithmetic, which uses floating-point representation of 2000 decimal digits of mantissa. The computer specifications are Intel(R) Core(TM) i5-2500 CPU @ 3.30 GHz with 16.00 GB of RAM. Each iteration is obtained from the former by means of an iterative expression $x^{(k+1)} = x^{(k)} - A^{-1}b$, where $x^{(k)} \in \mathbb{R}^n$, A is a real matrix $n \times n$ and $b \in \mathbb{R}^n$. The matrix A and vector b are different according to the method used, but in any case, we calculate $A^{-1}b$ as the solution of the linear system $Ay = b$, with Gaussian elimination with partial pivoting. The stopping criterion used is $\|x^{(k+1)} - x^{(k)}\| < 10^{-200}$ or $\|F(x^{(k)})\| < 10^{-200}$.

Firstly, let us consider the following nonlinear systems of different sizes:

- (1) $F_1 = (f_1(x), f_2(x), \dots, f_n(x))$, where $x = (x_1, x_2, \dots, x_n)^T$ and $f_i : \mathbb{R}^n \rightarrow \mathbb{R}, i = 1, 2, \dots, n$, such that

$$\begin{aligned} f_i(x) &= x_i x_{i+1} - 1, \quad i = 1, 2, \dots, n-1, \\ f_n(x) &= x_n x_1 - 1. \end{aligned} \tag{3.1}$$

When n is odd, the exact zeros of $F_1(x)$ are: $\xi_1 = (1, 1, \dots, 1)^T$ and $\xi_2 = (-1, -1, \dots, -1)^T$.

- (2) $F_2(x_1, x_2) = (x_1^2 - x_1 - x_2^2 - 1, -\sin(x_1) + x_2)$ and the solutions are $\xi_1 \approx (-0.845257, -0.748141)^T$ and $\xi_2 \approx (1.952913, 0.927877)^T$.
- (3) $F_3(x_1, x_2) = (x_1^2 + x_2^2 - 4, -\exp(x_1) + x_2 - 1)$, being the solutions $\xi_1 \approx (1.004168, -1.729637)^T$ and $\xi_2 \approx (-1.816264, 0.837368)^T$.
- (4) $F_4(x_1, x_2, x_3) = (x_1^2 + x_2^2 + x_3^2 - 9, x_1 x_2 x_3 - 1, x_1 + x_2 - x_3^2)$ with roots $\xi_1 \approx (2.14025, -2.09029, -0.223525)^T$, $\xi_2 \approx (2.491376, 0.242746, 1.653518)^T$ and $\xi_3 \approx (0.242746, 2.491376, 1.653518)^T$.

Table 2 presents results showing the following information: the different iterative methods employed (Newton (NC), Jarratt (JT), the new methods M4, M6, and M8, and the pseudocomposed PsM10 and PsM14), the number of iterations *Iter* needed to converge to the solution Sol, the value of the stopping factors at the last step, and the computational order of convergence ρ (see [17]) approximated by the formula:

$$\rho \approx \frac{\ln(\|x^{(k+1)} - x^{(k)}\|) / (\|x^{(k)} - x^{(k-1)}\|)}{\ln(\|x^{(k)} - x^{(k-1)}\|) / (\|x^{(k-1)} - x^{(k-2)}\|)}. \tag{3.2}$$

The value of ρ which appears in Table 2 is the last coordinate of the vector ρ when the variation between their coordinates is small. Also the elapsed time, in seconds, appears in Table 2, being the mean execution time for 100 performances of the method (the command `cputime` of MATLAB has been used).

We observe from Table 2 that not only the order of convergence and the number of new functional evaluations and operations are important in order to obtain new efficient iterative methods to solve nonlinear systems of equations. A key factor is the range of applicability of the methods. Although they are slower than the original methods when the initial estimation is quite good, when we are far from the solution or inside a region of instability, the original schemes do not converge or do it more slowly, the corresponding pseudocomposed procedures usually still converge or do it faster.

The advantage of pseudocomposition can be observed in Figures 2(a) and 2(b) (methods M6 and PsM10) and Figures 3(a) and 3(b) (methods M8 and PsM14) where the dynamical plane on R^2 is represented: we consider the system of two equations and two unknowns $F_2(x) = 0$, for any initial estimation in R^2 represented by its position in the plane, a different color (blue or orange, as there exist only two solutions in this region) is used for the different solutions found (marked by a white point in the figure). Black color represents an initial point in which the method converges to infinity, and the green one means that no convergence is found (usually because any linear system cannot be solved). It is clear that when many initial estimations tend to infinity (see Figure 3(a)), the pseudocomposition “cleans” the dynamical plane, making the method more stable as it can find one of the solutions by using starting points that do not allow convergence with the original scheme (see Figure 3(b)).

If an analogous study is made on system $F_3(x) = 0$, similar conclusions can be obtained, as the effect of smoothness is clear when the real dynamical plane of a method and its pseudocomposed partner are compared. So, in Figure 4 the amount of points in the lower half of the plane that converge to one of the roots is higher after the pseudocomposition, and, in Figure 5, there is a big green region of no convergence for method M8 that shows to be convergent when pseudocomposition is applied in PsM14.

We conclude that the presented schemes M4, M6, and M8 show to be excellent, in terms of order of convergence and efficiency, but also that the pseudocomposition technique achieves to transform them in competent and more robust new schemes.

Acknowledgments

The authors would like to thank the referees for the valuable comments and for the suggestions to improve the readability of the paper. This research was supported by Ministerio de Ciencia y Tecnología MTM2011-28636-C02-02 and FONDOCYT República Dominicana.

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