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A One-Parameter Family of Methods with a Higher Order of Convergence for Equations in a Banach Space

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Abstract: The conventional approach of the local convergence analysis of an iterative method on \mathbb{R}^m , with m a natural number, depends on Taylor series expansion. This technique often requires the calculation of high-order derivatives. However, those derivatives may not be part of the proposed method(s). In this way, the method(s) can face several limitations, particularly the use of higher-order derivatives and a lack of information about a priori computable error bounds on the solution distance or uniqueness. In this paper, we address these drawbacks by conducting the local convergence analysis within the broader framework of a Banach space. We have selected an important family of high convergence order methods to demonstrate our technique as an example. However, due to its generality, our technique can be used on any other iterative method using inverses of linear operators along the same line. Our analysis not only extends in \mathbb{R}^m spaces but also provides convergence conditions based on the operators used in the method, which offer the applicability of the method in a broader area. Additionally, we introduce a novel semilocal convergence analysis not presented before in such studies. Both forms of convergence analysis depend on the concept of generalized continuity and provide a deeper understanding of convergence properties. Our methodology not only enhances the applicability of the suggested method(s) but also provides suitability for applied science problems. The computational results also support the theoretical aspects.

Keywords: higher-order method; convergence; nonlinear systems; Banach space; continuity

MSC: 65G99; 47H17; 49M15



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1. Introduction

Iterative methods stand as a foundation in numerical analysis that can handle the tough challenge of solving nonlinear equations. These equations came from diverse domains ranging from physics and engineering to economics and finance [1–7]. In general, the analytical solutions to such problems are almost nonexistent. They can be transmitted in the form of:

$$\Delta(x) = 0, \quad (1)$$

where $\Delta : \Phi \subseteq \mathbb{M}_1 \rightarrow \mathbb{M}_2$, with \mathbb{M}_1 and \mathbb{M}_2 being Banach spaces.

Iterative methods represent numerical strategies that work with an initial approximation or more than one initial approximation. They enhance the solution until it reaches a predefined level of precision. One such iterative method is the Newton–Raphson method. This is one of the most significant iterative methods, which is given by

$$x_{\sigma+1} = x_{\sigma} - \Delta'(x_{\sigma})^{-1}\Delta(x_{\sigma}),$$

where σ denotes a natural number, including zero. Newton’s method is a well-known and frequently applied iterative procedure for solving nonlinear problems. However, this method encounters various challenges when applied to nonlinear equations, such as slower convergence, divergence when the Jacobian matrix approaches a null matrix, and failure to function when the Jacobian matrix is null. Consequently, researchers have proposed improvements or modifications to this method over time. We have also opted for such an iterative approach.

$$\begin{aligned}
 y_\sigma &= x_\sigma - \alpha \Delta'(x_\sigma)^{-1} \Delta(x_\sigma), \quad A_\sigma = A(\alpha, x_\sigma, y_\sigma), \\
 z_\sigma^{(1)} &= x_\sigma - A_\sigma^{-1} \Delta(x_\sigma), \quad B_\sigma = B(\alpha, x_\sigma, y_\sigma), \\
 &\vdots \\
 z_\sigma^{(k)} &= z_\sigma^{(k-1)} - B_\sigma \Delta(z_\sigma^{(k-1)}), \\
 x_{\sigma+1} &= z_\sigma^{(k+1)} = z_\sigma^{(k)} - B_\sigma \Delta(z_\sigma^{(k)}), \quad \sigma = 0, 1, 2, \dots,
 \end{aligned}
 \tag{2}$$

where $\alpha \in \mathbb{R} - \{0\}$, $k \in \mathbb{N}$, $A_\sigma = \frac{1}{2\alpha} \Delta'(y_\sigma) + \left(1 - \frac{1}{2\alpha}\right) \Delta'(x_\sigma)$, and $B_\sigma = 2A_\sigma^{-1} - \Delta'(x_\sigma)^{-1}$ and $x_0 \in \Phi$.

The convergence order $k + 2$ is shown in [8] using a local Taylor series by assuming the existence and boundedness of $\Delta^{(4)}$ for $\mathbb{M}_1 = \mathbb{M}_2 = \mathbb{R}^m$, where m is a natural number.

There exist the same concerns with the Taylor expansion series technique mostly used to study the convergence of iterative methods.

1.1. Motivation

- (L₁) The convergence is shown by assuming derivatives not in this method, such as Δ'' , Δ''' , and $\Delta^{(4)}$. Let $\Phi_0 = [-1.3, 1.3]$. Define the function $g(\zeta) = s_1 \zeta^3 \log t + s_2 \zeta^5 + s_3 \zeta^4$ for $t \neq 0$ and $g(\zeta) = 0$ for $\zeta = 0$, where $s_1 \neq 0$ and $s_2 + s_3 = 0$. It follows by this definition that the functions g''' and $g^{(4)}$ are not continuous at $\zeta = 0$. Thus, the results in [8] can not guarantee the convergence of the method to the solution $\zeta^* = 1 \in \Phi_0$. But the method converges to ζ^* , say for $k = 1$ and $\zeta_0 = 0.9$. It follows by this academic example that the convergence conditions can be weakened. Clearly, it is preferable to use conditions only on Δ , Δ' that are in the method.
- (L₂) The selection of the initial point constitutes a “shot in the dark”. This is because there is no computable radius of convergence to assist us in choosing possible points x_0 .
- (L₃) A priori and computable estimates on $\|x_\sigma - \zeta^*\|$ are not available. Hence, we do not know in advance how many iterations should be carried out to arrive at a desirable error tolerance.
- (L₄) There is no computable domain about ζ^* that contains no other solution of the equation $\Delta(x) = 0$.
- (L₅) There is no information in [8] about the semilocal analysis of convergence for the method.
- (L₆) This study is restricted on \mathbb{R}^m .

The concerns (L₁) – (L₆) constitute our motivation for this article.

It is worth noticing that these concerns are present in any other studies using the Taylor series on \mathbb{R}^m [4–7,9–29]. Therefore, the handling of these concerns will also extend the applicability and inverse of linear operators of other methods along the same lines. Our technique is demonstrated in method (2) as an example. But if it is so general, it can be used on any other method using inverses along the same lines. In particular, we deal with these concerns as follows:

1.2. Novelty

- (L₁)' The new sufficient convergence conditions involve only the operators in the method. (See the conditions (H₁) – (H₇) or (E₁) – (E₆)).
- (L₂)' A computable radius of convergence is determined. Thus, the selection of initial points that guarantee the convergence of the method becomes possible. (See (3)).
- (L₃)' The number of iterations required to achieve an error tolerance $\epsilon > 0$ is known in advance, because we provide computable upper error bounds on $\|x_\sigma - \zeta^*\|$. (See Theorem 1).
- (L₄)' A neighborhood of ζ^* is found that contains no other solution of the equation $\Delta(x) = 0$. (See the Propositions).
- (L₅)' A semilocal analysis of convergence is developed by relying on majorizing sequences [1–5]. (See Section 3).
- (L₆)' The convergence is established for Banach space-valued operators.

Notice also that the convergence in both cases is based on the generalized continuity [1–3] assumption on the operator Δ' .

The rest of this article is structured as follows: Section 1 contains the local analysis followed by the semilocal analysis in Section 3. The numerical examples appear in Section 4. This article ends with the ending remarks in Section 5.

2. Local Analysis

Let $Q = [0, +\infty)$. The convergence is based on some conditions.

Suppose the following:

- (H₁) There exists a function $\aleph_0 : Q \rightarrow \mathbb{R}_+$, which is continuous as well as nondecreasing, such that $\aleph_0(\xi) - 1 = 0$ admits a smallest solution, which is positive. Denote such a solution with the letter ρ_0 .
Take $Q_0 = [0, \rho_0)$.
- (H₂) The rest exists as a function that is continuous and nondecreasing $\aleph : Q_0 \rightarrow \mathbb{R}_+$ such that for $h_0 : Q_0 \rightarrow \mathbb{R}$, the equation $h_0(\xi) - 1 = 0$ admits a smallest positive solution denoted by r_0 , where

$$h_0(\xi) = \frac{\int_0^1 \aleph((1 - \omega)\xi) d\omega + |1 - \alpha| \left(1 + \int_0^1 \aleph_0(\omega\xi) d\omega\right)}{1 - \aleph_0(\xi)}.$$

- (H₃) If $\tilde{\aleph}(\xi) = \begin{cases} \aleph\left(\left(1 + h_0(\xi)\right)\xi\right) \\ \text{or} \\ \aleph_0(\xi) + \aleph_0\left(h_0(\xi)\xi\right), \end{cases}$ define $P(\xi) = \frac{1}{2|\alpha|} \tilde{\aleph}(\xi) + \aleph_0(\xi)$. The equation

$P(\xi) - 1 = 0$ admits a smallest solution, which is positive, denoted by $\rho_1 \in (0, \rho_0)$.

Set $\rho = \min\{\rho_0, \rho_1\}$ and $Q_1 = [0, \rho)$.

- (H₄) The equations $h_1(\xi) - 1 = 0$ and $h_i(\xi) - 1 = 0, i = 2, 3, \dots, k + 1$ admit a smallest positive solution denoted by $r_1, r_i \in (0, \rho)$, respectively, where

$$h_1(\xi) = \frac{\int_0^1 \aleph((1 - \omega)\xi) d\omega}{1 - \aleph_0(\xi)} + \frac{\tilde{\aleph}(\xi) \left(1 + \int_0^1 \aleph_0(\omega\xi) d\omega\right)}{2|\alpha|(1 - \aleph_0(\xi))(1 - P(\xi))}$$

and for

$$\tilde{w}_{i-1}(\xi) = \begin{cases} \aleph\left(\left(1 + h_{i-1}(\xi)\right)\xi\right) \\ \aleph_0(\xi) + \aleph_0\left(h_{i-1}(\xi)\xi\right), \end{cases}$$

$$h_i(\xi) = \left[\frac{\int_0^1 \aleph((1-\omega)h_{i-1}(\xi)\xi) d\omega}{1 - \aleph_0(h_{i-1}(\xi)\xi)} + \left(\frac{\frac{1}{2^{|\alpha|}} \tilde{\aleph}(\xi) + \tilde{w}_{i-1}(\xi)}{(1 - \aleph_0(h_{i-1}(\xi)\xi))(1 - P(\xi))} + \frac{\tilde{\aleph}(\xi)}{(1 - \aleph_0(\xi))(1 - P(\xi))} \right) \left(1 + \int_0^1 \aleph_0(\omega h_{i-1}(\xi)\xi) d\omega \right) \right] h_{i-1}(\xi)$$

Take

$$r = \min\{r_j\}, j = 0, 1, 2, \dots, k + 1, \tag{3}$$

and set $Q^* = [0, r)$.

The definitions imply that for each $\xi \in Q^*$,

$$0 \leq \aleph_0(\xi) < 1, \tag{4}$$

$$0 \leq P(\xi) < 1, \tag{5}$$

$$0 \leq \aleph_0(h_{i-1}(\xi)\xi) < 1, \tag{6}$$

$$0 \leq h_j(\xi) < 1. \tag{7}$$

The developed scalar functions \aleph_0 and \aleph are associated to the operator Δ' in the method.

(H5) There exists an invertible operator $\gamma \in \ell(\mathbb{M}_1, \mathbb{M}_2)$ such that for each $x \in \Phi$,

$$\|\gamma^{-1}(\Delta'(x) - \gamma)\| \leq \aleph_0(\|x - \xi^*\|).$$

Take $\mathbb{T}_0 = S(\xi^*, \rho_0) \cap \Phi$.

(H6) $\|\gamma^{-1}(\Delta(y) - \Delta(x))\| \leq \aleph(\|y - x\|)$ for each $x, y \in \mathbb{T}_0$

and

(H7) $S[\xi^*, r] \subset \Phi$.

Remark 1.

- (i) A possible and popular choice for $\gamma = I$ or $\gamma = \Delta'(\xi^*)$. In the latter choice, ξ^* is a simple solution of the equation $\Delta(x) = 0$ according to the condition (H5). In our analysis, we do not necessarily assume that ξ^* is simple. Thus, the method can also be used to find solutions of ξ^* of multiplicity greater than one. Another selection is $\gamma = I$. Other choices can be considered as (H5) and (H6) hold.
- (ii) We shall choose the smallest version of the functions $\tilde{\aleph}$ in the examples to obtain tighter error bounds on $\|x_\sigma - \xi^*\|$ as well as a larger r .

Next, the local analysis follows for the method (2).

Theorem 1. Suppose that the conditions (H1) – (H7) are satisfied. Then, if $x_0 \in S_0 = S(\xi^*, r) - \{\xi^*\}$, the sequence $\{x_\sigma\}$ is convergent to ξ^* .

Proof. The following items are established by induction:

$$\|y_\sigma - \xi^*\| \leq h_0(\|x_\sigma - \xi^*\|)\|x_\sigma - \xi^*\| \leq \|x_\sigma - \xi^*\| < r, \tag{8}$$

$$\|z_\sigma^{(1)} - \xi^*\| \leq h_1(\|x_\sigma - \xi^*\|)\|x_\sigma - \xi^*\| \leq \|x_\sigma - \xi^*\|, \tag{9}$$

$$\|z_\sigma^{(i)} - \xi^*\| \leq h_i(\|x_\sigma - \xi^*\|)\|x_\sigma - \xi^*\| \leq \|x_\sigma - \xi^*\|, i = 2, 3, \dots, k + 1, \tag{10}$$

where the radius r is determined by the formula (3), and all the scalar functions g_m , $m = 0, 1, 2, \dots, k + 1$ are as given previously. The condition (H_5) , (3), and the definition of ρ_0 and r give in turn

$$\left\| \gamma^{-1} \left(\Delta'(x_0) - \gamma \right) \right\| \leq \aleph_0(\|x_0 - \zeta^*\|) \leq \aleph_0(r) < 1. \tag{11}$$

So, the linear operator $\Delta'(x_0)$ is invertible by the standard perturbation Lemma on operators that are invertible due to Banach [1–5]. We also have

$$\left\| \Delta'(x_0)^{-1} \gamma \right\| \leq \frac{1}{1 - \aleph_0(\|x_0 - \zeta^*\|)}. \tag{12}$$

Consequently, the iterate x_0 exists by the first substep of the method (2), and

$$\begin{aligned} y_0 - \zeta^* &= x_0 - \zeta^* - \Delta'(x_0)^{-1} \Delta(x_0) + (1 - \alpha) \Delta'(x_0)^{-1} \Delta(x_0) \\ &= \int_0^1 \Delta'(x_0)^{-1} (\Delta'(\zeta^* + \omega(x_0 - \zeta^*)) - \Delta'(x_0)) d\omega(x_0 - \zeta^*) \\ &\quad + (1 - \alpha) \Delta'(x_0)^{-1} S'_0(\Delta'(\zeta^* + \omega(x_0 - \zeta^*)) - \Delta'(x_0) + \Delta'(x_0)) d\omega(x_0 - \zeta^*). \end{aligned} \tag{13}$$

By applying the condition (H_6) , and using (7) (for $j = 0$), (12), and (13), we obtain in turn that

$$\begin{aligned} \|y_0 - \zeta^*\| &\leq \frac{\left[\int_0^1 \aleph((1 - \omega)\|x_0 - \zeta^*\|) d\omega + |1 - \alpha|(1 + \int_0^1 \aleph_0(\omega\|x_0 - \zeta^*\|) d\omega) \right] \|x_0 - \zeta^*\|}{1 - \aleph_0(\|x_0 - \zeta^*\|)} \\ &\leq h_0(\|x_0 - \zeta^*\|) \|x_0 - \zeta^*\| \leq \|x_0 - \zeta^*\| < r. \end{aligned} \tag{14}$$

Thus, the iterate $y_0 \in S(\zeta^*, r)$, and the item (8) is satisfied for $\sigma = 0$. We need the estimate

$$\begin{aligned} \left\| \gamma^{-1} (A_0 - \alpha) \right\| &\leq \frac{1}{2|\alpha|} \left\| \gamma^{-1} (\Delta'(y_0) - \Delta'(x_0)) \right\| + \left\| \gamma^{-1} (\Delta'(x_0) - \gamma) \right\| \\ &\leq \frac{1}{2|\alpha|} \tilde{\aleph}_0(\|y_0 - \zeta^*\|) + \aleph_0(\|x_0 - \zeta^*\|) = \rho_0 < 1, \end{aligned} \tag{15}$$

(by (5)), where the first norm in (15) can be calculated by two different ways:

$$\begin{aligned} \left\| \gamma^{-1} (\Delta'(y_0) - \Delta'(x_0)) \right\| &\leq \aleph(\|y_0 - \zeta^*\| + \|x_0 - \zeta^*\|) \\ &\leq \aleph((1 + h_0(\|x_0 - \zeta^*\|)) \|x_0 - \zeta^*\|) \\ &\leq \tilde{\aleph}_0(\|x_0 - \zeta^*\|) \end{aligned}$$

or

$$\begin{aligned} \left\| \gamma^{-1} (\Delta'(y_0) - \Delta'(x_0)) \right\| &\leq \left\| \gamma^{-1} (\Delta'(x_0) - \gamma) \right\| + \left\| \gamma^{-1} (\Delta'(y_0) - \gamma) \right\| \\ &\leq \aleph_0(\|x_0 - \zeta^*\|) + \aleph_0(\|y_0 - \zeta^*\|) \\ &\leq \aleph_0(\|x_0 - \zeta^*\|) + \aleph_0(h_0(\|x_0 - \zeta^*\|) \|x_0 - \zeta^*\|) \\ &\leq \tilde{\aleph}_0(\|x_0 - \zeta^*\|). \end{aligned}$$

So, the linear operator A_0 is invertible and

$$\left\| A_0^{-1} \gamma \right\| \leq \frac{1}{1 - P(\|x_0 - \zeta^*\|)}. \tag{16}$$

It follows that the iterate $z_0^{(1)}$ exists by the second substep of the method (2), and

$$\begin{aligned} z_0^{(1)} - \zeta^* &= x_0 - \zeta^* - \Delta'(x_0)^{-1} \Delta(x_0) + (\Delta'(x_0)^{-1} - A_0^{-1}) \Delta(x_0) \\ &= x_0 - \zeta^* - \Delta'(x_0)^{-1} \Delta(x_0) - A_0^{-1} (\Delta'(x_0) - A_0) \Delta'(x_0)^{-1} \Delta(x_0). \end{aligned} \tag{17}$$

Hence, we obtain by (3), (7) (for $j = 2$), (16), and (17) that

$$\begin{aligned} \|z_0^{(1)} - \zeta^*\| &\leq \left[\frac{\int_0^1 \aleph((1-\omega)\|x_0 - \zeta^*\|)d\omega}{1 - \aleph_0(\|x_0 - \zeta^*\|)} \right. \\ &\quad \left. + \frac{\tilde{\aleph}_0(\|x_0 - \zeta^*\|)(1 + \int_0^1 \aleph_0(\omega\|x_0 - \zeta^*\|)d\omega)}{2|\alpha|(1 - \aleph_0(\|x_0 - \zeta^*\|))(1 - P(\|x_0 - \zeta^*\|))} \right] \|x_0 - \zeta^*\| \\ &\leq h_1(\|x_0 - \zeta^*\|)\|x_0 - \zeta^*\| \leq \|x_0 - \zeta^*\|. \end{aligned} \tag{18}$$

It follows by (18) that the iterate $z_0^{(1)} \in S(\zeta^*, r)$, and the item (9) holds if $\sigma = 0$. Moreover, the iterate $z_0^{(2)}$ is well defined by the third substep of the method (2), and

$$\begin{aligned} z_0^{(2)} - \zeta^* &= z_0^{(1)} - \zeta^* - \Delta'(z_0^{(1)})^{-1}\Delta(z_0^{(1)}) + (\Delta'(z_0^{(1)})^{-1} + \Delta'(x_0)^{-1} - 2A_0^{-1})\Delta(z_0^{(1)}) \\ &= z_0^{(1)} - \zeta^* - \Delta'(z_0^{(1)})^{-1}\Delta(z_0^{(1)}) + \Delta'(z_0^{(1)})^{-1}(A_0 - \Delta'(z_0^{(1)}))A_0^{-1}\Delta(z_0^{(1)}) \\ &\quad + \Delta'(x_0)^{-1}(A_0 - \Delta'(x_0))A_0^{-1}\Delta(z_0^{(1)}), \end{aligned} \tag{19}$$

leading to

$$\begin{aligned} \|z_0^{(2)} - \zeta^*\| &\leq \left[\frac{\int_0^1 \aleph((1-\omega)\|z_0^{(1)} - \zeta^*\|)d\omega}{1 - \aleph_0(\|z_0^{(1)} - \zeta^*\|)} + \left(\frac{\frac{1}{2|\gamma|}\tilde{\aleph}_0(\|x_0 - \zeta^*\|) + \aleph(\|z_0^{(1)} - \zeta^*\|)}{(1 - \aleph_0(\|z_0^{(1)} - \zeta^*\|))(1 - P(\|x_0 - \zeta^*\|))} \right. \right. \\ &\quad \left. \left. + \frac{\tilde{\aleph}_0(\|x_0 - \zeta^*\|)}{(1 - \aleph_0(\|x_0 - \zeta^*\|))(1 - P(\|x_0 - \zeta^*\|))} \right) \left(1 + \int_0^1 \aleph_0(\omega\|z_0^{(1)} - \zeta^*\|)d\omega \right) \right] \|z_0^{(1)} - \zeta^*\| \\ &\leq h_2(\|x_0 - \zeta^*\|)\|x_0 - \zeta^*\| \leq \|x_0 - \zeta^*\|. \end{aligned} \tag{20}$$

Thus, the iterate $z_0^{(2)} \in S(\zeta^*, r)$, and the item (10) is satisfied if $i = 2$. But the computations for the derivation of (20) can be repeated if $z_0^{(3)}, z_0^{(4)}, \dots, z_0^{(k+1)}$ replace $z_0^{(1)}$, respectively, in the preceding estimations, which terminates the induction for (10). Hence, the iterate $x_{k+1} \in S(\zeta^*, r)$ and

$$\|x_{\sigma+1} - \zeta^*\| = \|z_{\sigma}^{k+1} - \zeta^*\| \leq c\|x_{\sigma} - \zeta^*\| < r, \tag{21}$$

where $c = h_{k+1}(\|x_0 - \zeta^*\|) \in [0, 1)$. Therefore, we conclude that all the iterates $\{x_{\sigma}\}$ stay in $S(\zeta^*, r)$ and $\lim_{\sigma \rightarrow \infty} x_{\sigma} = \zeta^*$. \square

Next, a neighborhood of ζ^* is determined containing no other solution.

Proposition 1. *Suppose the following:*

The solution $\zeta^ \in S(\zeta^*, \rho_2)$ for some $\rho_2 > 0$.*

The condition (H₅) is satisfied in the ball $S(\zeta^, \rho_2)$, and there exists $\rho_3 \geq \rho_2$ such that*

$$\int_0^1 \aleph_0(\omega\rho_3)d\omega < 1. \tag{22}$$

Define $D_1 = \Phi \cap S[\zeta^, \rho_3]$.*

Then, the only solution of the equation $\Delta(x) = 0$ in the set D_1 is ζ^ .*

Proof. Let $\tilde{x} \in D_1$ be such that $\Delta(\tilde{x}) = 0$.

Consider the linear operator $T = \int_0^1 \Delta'(\zeta^* + \omega(\tilde{x} - \zeta^*))d\omega$. Then, by applying the condition (H₅) and using (22), we obtain in turn that

$$\|\gamma^{-1}(T - \gamma)\| \leq \int_0^1 v_0(\omega\|\tilde{x} - \zeta^*\|)d\omega \leq \int_0^1 v_0(\omega\rho_3)d\omega < 1. \tag{23}$$

It follows by (23) that the operator T is invertible. Finally, by the identity, we obtain

$$\tilde{x} - \zeta^* = T^{-1}(\Delta(\tilde{x}) - \Delta(\zeta^*)) = T^{-1}(0) = 0.$$

Thus, we deduce that $\tilde{x} = \zeta^*$. \square

Remark 2. Clearly, we can take $\rho_2 = r$ in Proposition 1.

In the next section, similar estimates are used to show the semilocal analysis of convergence for the method (2). But the role of the functions \aleph_0, \aleph and the solution ζ^* is exchanged by the functions v_0, v and the initial point x_0 , respectively.

3. Semilocal Analysis

Suppose the following:

- (E₁) There exists a function that is continuous and nondecreasing $v_0 : Q \rightarrow \mathbb{R}_+$ such that the equation $v_0(\xi) - 1 = 0$ has a smallest solution, which is positive, denoted by ρ_4 . Set $Q_2 = [0, \rho_4)$.
- (E₂) There exists a function that is continuous and nondecreasing $v : Q_2 \rightarrow \mathbb{R}_+$.

Define the scalar sequence $\{a_\sigma^{(m)}\}$ for $a_0^{(0)} = 0$, some $a_1^{(0)} \in [0, \rho_4)$, and each $\sigma = 0, 1, 2, \dots, m = 0, 1, 2, \dots, k$ by

$$\begin{aligned} \tilde{v}_\sigma &= \begin{cases} v(a_\sigma^{(1)} - a_\sigma^{(0)}) \\ \text{or} \\ v_0(a_\sigma^{(0)}) + v_0(a_\sigma^{(1)}), \end{cases} \\ \lambda_\sigma &= \frac{1}{2|\alpha|} \tilde{v}_\sigma + (1 + v_0(a_\sigma^{(0)})), \\ \mu_\sigma^{(m)} &= \left(1 + \int_0^1 v_0(a_\sigma^{(0)} + \omega(a_\sigma^{(m+1)} - a_\sigma^{(0)}))d\omega\right)(a_\sigma^{(m+1)} - a_\sigma^{(0)}) \\ &\quad + \frac{1}{|\alpha|} (1 + v_0(a_\sigma^{(0)}))(a_\sigma^{(1)} - a_\sigma^{(0)}), \\ q_\sigma &= \frac{1}{2|\alpha|} \tilde{v}_\sigma + v_0(a_\sigma^{(0)}), \\ a_\sigma^{(2)} &= a_\sigma^{(1)} + \frac{|1 - \alpha| \left(1 + v_0(a_\sigma^{(0)}) + \frac{1}{2} \tilde{v}_\sigma\right) (a_\sigma^{(1)} - a_\sigma^{(0)})}{|\alpha|(1 - q_\sigma)}, \\ a_\sigma^{(m+2)} &= a_\sigma^{(m+1)} + \frac{\lambda_\sigma \mu_\sigma^{(m)}}{(1 - v_0(a_\sigma^{(0)}))(1 - q_\sigma)}, \\ b_{\sigma+1} &= \int_0^1 v((1 - \omega)(a_{\sigma+1}^{(0)} - a_\sigma^{(0)}))d\omega (a_{\sigma+1}^{(0)} - a_\sigma^{(0)}) \\ &\quad + \frac{1}{|\alpha|} (1 + v_0(a_\sigma^{(0)}))(a_\sigma^{(1)} - a_\sigma^{(0)}), \end{aligned} \tag{24}$$

and

$$a_{\sigma+1}^{(1)} = a_{\sigma+1}^{(0)} + |\alpha| \frac{b_{\sigma+1}}{1 - v_0(a_{\sigma+1}^{(0)})}.$$

This sequence is shown to be majorizing for the method (2) in Theorem 2. However, let us first give a convergence condition for it.

- (E₃) There exists $\rho_5 \in [0, \rho_4)$ such that for each $\sigma = 0, 1, 2, \dots, m = 1, 2, \dots, k$,

$$v_0(a_\sigma^{(0)}) < 1, q_\sigma < 1, \text{ and } a_\sigma^{(m)} < \rho_5.$$

It follows by the formula (24), (E₁), (E₂), and this condition that

$$0 \leq a_\sigma^{(m)} \leq a_{\sigma+1}^{(m)} \leq \rho_5$$

and this sequence is convergent to some $a^* \in [0, \rho_5]$.

As in the local analysis, the functions v_0 and v relate to Δ' as follows:

(E₄) There exist $x_0 \in \Phi$ and an invertible operator γ such that for each $x \in \Phi$,

$$\|\gamma^{-1}(\Delta'(x) - \gamma)\| \leq v_0(\|x - x_0\|).$$

Notice that

$$\|\gamma^{-1}(\Delta'(x_0) - \gamma)\| \leq v_0(0) < 1.$$

Consequently, the inverse of the linear operator $\Delta'(x_0)$ exists, and we can choose

$$\|\Delta'(x_0)^{-1}\Delta(x_0)\| \leq a_1^{(0)}.$$

Set $D_2 = \Phi \cap S(x_0, \rho_4)$.

(E₅) $\|\gamma^{-1}(\Delta'(y) - \Delta'(x))\| \leq v(\|y - x\|)$ for each $x, y \in D_2$.

(E₆) $S[x_0, a^*] \subset \Phi$.

The main semilocal analysis of convergence follows under these conditions in the next result.

Theorem 2. *Suppose that the conditions (E₁) – (E₆) are satisfied. Then, there exists a solution $\xi^* \in S[x_0, a^*]$ of the equation $\Delta(x) = 0$ such that for each $\sigma = 0, 1, 2, \dots$,*

$$\|\xi^* - x_\sigma\| \leq a^* - a_\sigma^{(k)}. \tag{25}$$

Proof. As the local case but using the conditions E instead of H, we have the series of calculations

$$\begin{aligned} \Delta'(x_\sigma) - \alpha A_\sigma &= \Delta'(x_\sigma) - \frac{1}{2}\Delta'(y_\sigma) - \alpha\Delta'(x_\sigma) + \frac{1}{2}\Delta'(x_\sigma) \\ &= (1 - \alpha)\Delta'(x_\sigma) + \frac{1}{2}(\Delta'(x_\sigma) - \Delta'(y_\sigma)), \\ \|\gamma^{-1}(\Delta'(x_\sigma) - \alpha A_\sigma)\| &\leq |1 - \alpha| \left(1 + v_0(a_\sigma^{(0)})\right) + \frac{1}{2}\tilde{v}_\sigma. \end{aligned} \tag{26}$$

So, from the first two substeps, we have

$$\begin{aligned} z_\sigma^{(1)} - y_\sigma &= [\alpha\Delta'(x_\sigma)^{-1} - A_\sigma^{-1}]\Delta(x_\sigma) \\ &= -(A_\sigma^{-1} - \alpha\Delta'(x_\sigma)^{-1})\Delta(x_\sigma) \\ &= -A_\sigma^{-1}(\Delta'(x_\sigma) - \alpha A_\sigma)\Delta'(x_\sigma)^{-1}\Delta(x_\sigma) \\ &= \frac{1}{\alpha}A_\sigma^{-1}(\Delta'(x_\sigma) - \alpha A_\sigma)(y_\sigma - x_\sigma). \end{aligned} \tag{27}$$

Thus, we obtain from (26) and (27),

$$\begin{aligned} \|z_\sigma^{(1)} - y_\sigma\| &\leq \frac{1}{|\alpha|} \|A_\sigma^{-1}\gamma\| \|\gamma^{-1}(\Delta'(x_\sigma) - \alpha A_\sigma)\| \|y_\sigma - x_\sigma\| \\ &\leq \frac{\left(|1 - \alpha|(1 + v_0(a_\sigma^{(0)})) + \frac{1}{2}\tilde{v}_\sigma\right) (a_\sigma^{(1)} - a_\sigma^{(0)})}{|\alpha|(1 - q_\sigma)} = a_\sigma^{(2)} - a_\sigma^{(1)}, \end{aligned} \tag{28}$$

and

$$\begin{aligned} \|z_\sigma^{(1)} - x_0\| &\leq \|z_\sigma^{(1)} - y_\sigma\| + \|y_\sigma - x_0\| \\ &\leq a_\sigma^{(2)} - a_\sigma^{(1)} + a_\sigma^{(1)} - a_0^{(0)} = a_\sigma^{(2)} < a^*, \end{aligned}$$

hence, the iterate $z_\sigma^{(1)} \in S[x_0, a^*]$.

We also need the estimate for the definition of B_σ that

$$\begin{aligned} 2\Delta'(x_\sigma) - A_\sigma &= 2\Delta'(x_\sigma) - \frac{1}{2\alpha} (\Delta'(y_\sigma) - \Delta'(x_\sigma)) - \Delta'(x_\sigma), \\ &= \Delta'(x_\sigma) - \frac{1}{2\alpha} (\Delta'(y_\sigma) - \Delta'(x_\sigma)). \end{aligned}$$

So, we have

$$\begin{aligned} \|\gamma^{-1}(2\Delta'(x_\sigma) - A_\sigma)\| &\leq \frac{1}{2|\alpha|} \bar{v}_\sigma + (1 + v_0(a_\sigma^{(0)})) = \lambda_\sigma, \\ \|B_\sigma \gamma\| &\leq \|A_\sigma^{-1} \gamma\| \|\gamma^{-1}(2\Delta'(x_\sigma) - A_\sigma)\| \|\Delta'(x_\sigma)^{-1} \gamma\| \\ &\leq \frac{\lambda_\sigma}{(1 - v_0(a_\sigma^{(0)}))(1 - q_\sigma)}, \\ \Delta(z_\sigma^{(1)}) &= \Delta(z_\sigma^{(1)}) - \Delta(x_\sigma) - \frac{1}{\alpha} \Delta'(x_\sigma)(y_\sigma - x_\sigma), \end{aligned} \tag{29}$$

and

$$\begin{aligned} \|\gamma^{-1} \Delta(z_\sigma^{(1)})\| &\leq \left(1 + \int_0^1 v_0(a_\sigma^{(0)} + \omega(a_\sigma^{(2)} - a_\sigma^{(0)})) d\omega\right) (a_\sigma^{(2)} - a_\sigma^{(0)}) \\ &\quad + \frac{1}{|\alpha|} (1 + v_0(a_\sigma^{(0)}))(a_\sigma^{(1)} - a_\sigma^{(0)}) \\ &= \mu_\sigma^{(1)}. \end{aligned} \tag{30}$$

Thus, (29) and (30) lead to

$$\begin{aligned} \|z_\sigma^{(2)} - z_\sigma^{(1)}\| &\leq \|B_\sigma^{-1} \gamma\| \|\gamma^{-1} \Delta(z_\sigma^{(1)})\| \\ &\leq \frac{\lambda_\sigma \mu_\sigma^{(1)}}{(1 - v_0(a_\sigma^{(0)}))(1 - q_\sigma)} = a_\sigma^{(3)} - a_\sigma^{(2)}, \end{aligned} \tag{31}$$

and

$$\begin{aligned} \|z_\sigma^{(2)} - x_0\| &\leq \|z_\sigma^{(2)} - z_\sigma^{(1)}\| + \|z_\sigma^{(1)} - x_0\| \\ &\leq a_\sigma^{(3)} - a_\sigma^{(2)} + a_\sigma^{(2)} - a_\sigma^{(0)} = a_\sigma^{(3)} < a^*. \end{aligned}$$

Hence, the iterate $z_\sigma^{(2)} \in S(x_0, a^*)$. Similarly, we have from

$$z_\sigma^{(m+1)} - z_\sigma^{(m)} = -B_\sigma \Delta(z_\sigma^{(m)})$$

that

$$\|z_\sigma^{(m+1)} - z_\sigma^{(m)}\| \leq \frac{\lambda_\sigma \mu_\sigma^{(m)}}{(1 - v_0(a_\sigma^{(0)}))(1 - q_\sigma)} = a_\sigma^{(m+2)} - a_\sigma^{(m+1)} \tag{32}$$

and

$$\begin{aligned} \|z_\sigma^{(m+1)} - x_0\| &\leq \|z_\sigma^{(m+1)} - z_\sigma^{(m)}\| + \|z_\sigma^{(m)} - x_0\| \\ &\leq a_\sigma^{(m+2)} - a_\sigma^{(m+1)} + a_\sigma^{(m+1)} - a_\sigma^{(0)} = a_\sigma^{(m+2)} < a^*, \end{aligned}$$

so the iterate $z_\sigma^{(m+1)} \in S(x_0, a^*)$.

Then, from the identity

$$\Delta(x_{\sigma+1}) = \Delta(x_{\sigma+1}) - \Delta(x_\sigma) - \frac{\Delta'(x_\sigma)}{\alpha} (x_{\sigma+1} - x_\sigma) + \Delta'(x_\sigma)(x_{\sigma+1} - y_\sigma),$$

we obtain

$$\begin{aligned} \|\gamma^{-1}\Delta(x_{\sigma+1})\| &\leq \int_0^1 v((1-\omega)(a_{\sigma+1}^{(0)} - a_{\sigma}^{(0)}))d\omega(a_{\sigma+1}^{(0)} - a_{\sigma}^{(0)}) + \frac{1}{|\alpha|}(1 + v_0(a_{\sigma}^{(0)}))(a_{\sigma}^{(1)} - a_{\sigma}^{(0)}) \\ &= b_{\sigma+1}, \end{aligned} \tag{33}$$

leading to

$$\begin{aligned} \|y_{\sigma+1} - x_{\sigma+1}\| &= |\alpha| \left\| \Delta'(x_{\sigma+1})^{-1}\gamma \right\| \left\| \gamma^{-1}\Delta(x_{\sigma+1}) \right\| \\ &\leq |\alpha| \frac{b_{\sigma+1}}{1 - v_0(a_{\sigma+1}^{(0)})} = a_{\sigma+1}^{(1)} - a_{\sigma+1}^{(0)}, \end{aligned} \tag{34}$$

and

$$\begin{aligned} \|y_{\sigma+1} - x_0\| &\leq \|y_{\sigma+1} - x_{\sigma+1}\| + \|x_{\sigma+1} - x_0\| \\ &\leq a_{\sigma+1}^{(1)} - a_{\sigma+1}^{(0)} + a_{\sigma+1}^{(0)} - a_0^{(0)} = a_{\sigma+1}^{(1)} < a^*, \end{aligned}$$

so the iterate $y_{\sigma+1} \in S(x_0, a^*)$.

Hence, we showed that all the iterates $\{x_{\sigma}\} \subset S(x_0, a^*)$ and the items (31), (32), and (34) are satisfied. By the condition (E₃), the majorizing sequence $\{a_{\sigma}^{(m)}\}$ is Cauchy and convergent. Thus, the sequence $\{x_{\sigma}\}$ is also Cauchy in the Banach space by (31), (32), and (34). Consequently, it is convergent to some $\xi^* \in S[x_0, a^*]$.

By letting $\sigma \rightarrow +\infty$ in (33), and using the continuity of the operator Δ , we deduce that $\Delta(\xi^*) = 0$. Then, in the estimation $\|x_{\sigma+i}^{(k)} - x_{\sigma}^{(k)}\| \leq a_{\sigma+i}^{(k)} - a_{\sigma}^{(k)}$. Let $i \rightarrow +\infty$ to conclude that (25) is satisfied. \square

Next, as in the local case, the uniqueness of the solution is established for the equation $\Delta(x) = 0$ in a neighborhood of the initial point x_0 .

Proposition 2. *Suppose the following:*

There exists a solution $\tilde{x} \in S(x_0, \rho_6)$ for some $\rho_6 > 0$.

The condition (E₄) is satisfied in the ball $S(x_0, \rho_6)$, and there exists $\rho_7 \geq \rho_6$ such that

$$\int_0^1 v_0(\omega\rho_6 + (1-\omega)\rho_7)d\omega < 1. \tag{35}$$

Take $D_3 = \Phi \cap S[x_0, \rho_7]$.

Then, the only solution of the equation $\Delta(x) = 0$ in the set D_3 is \tilde{x} .

Proof. Let $u \in D_3$ such that $\Delta(u) = 0$. Define the linear operator $T_1 = \int_0^1 \Delta'(\tilde{x} + \omega(u - \tilde{x}))d\omega$. It follows by the condition (E₄) and (35) that

$$\begin{aligned} \left\| \gamma^{-1}(T_1 - \gamma) \right\| &\leq \int_0^1 v_0(\omega\|\tilde{x} - \tilde{x}_0\| + (1-\omega)\|u - x_0\|)d\omega \\ &\leq \int_0^1 v_0(\omega\rho_6 + (1-\omega)\rho_7)d\omega < 1, \end{aligned} \tag{36}$$

so the operator T_1 is invertible. Then, from the identity

$$u - \tilde{x} = T_1^{-1}(\Delta(u) - \Delta(\tilde{x})) = T_1^{-1}(0) = 0,$$

we conclude that $u = \tilde{x}$. \square

Remark 3.

- (i) *A popular choice for $\gamma = \Delta'(x_0)$ or $\gamma = I$. Other selections are possible as long as the conditions (E₄) and (E₅) hold.*
- (ii) *The limit point a^* can be replaced by ρ_4 in the condition (E₆).*
- (iii) *Under all the conditions (E₁) – (E₆) in Proposition 2, one can take $\tilde{x} = \xi^*$ and $\rho_6 = a^*$.*

4. Numerical Illustrations

We illustrate numerically the theoretical consequences proposed in the previous sections. We have illustrated the applicability in real-life and standard nonlinear problems along with initial guesses that are depicted in Examples 1–5. These examples involve a good mixture of scalar and nonlinear systems. We display not only the radius of convergence in the Tables 1–5 but also the least number of iterations corresponding to the solutions of $P(x)$, absolute residual error at the corresponding iteration, and (COC). Additionally, we obtain the COC approximated by means of

$$\chi = \frac{\ln \frac{\|x_{\sigma+1} - \xi^*\|}{\|x_{\sigma} - \xi^*\|}}{\ln \frac{\|x_{\sigma} - \xi^*\|}{\|x_{\sigma-1} - \xi^*\|}}, \quad \text{for } \sigma = 1, 2, \dots \tag{37}$$

or ACOC [7,15–17] by:

$$\chi^* = \frac{\ln \frac{\|x_{\sigma+1} - x_{\sigma}\|}{\|x_{\sigma} - x_{\sigma-1}\|}}{\ln \frac{\|x_{\sigma} - x_{\sigma-1}\|}{\|x_{\sigma-1} - x_{\sigma-2}\|}}, \quad \text{for } \sigma = 2, 3, \dots \tag{38}$$

We adopt $\epsilon = 10^{-100}$ as a sanction error. Ceasing criteria are adopted for computer programming to solve the nonlinear system: (i) $\|x_{\sigma+1} - x_{\sigma}\| < \epsilon$ and (ii) $\|\Delta(x_{\sigma})\| < \epsilon$.

The computational work is implemented with the software *Mathematica 9* by adopting higher-precision arithmetic. In addition, we have chosen $\alpha = 1$ and $\gamma = \Delta'(\xi^*)$ in the next two examples.

Example 1. We selected a well-known mathematical issue involving a first-order nonlinear integral equation and a Hammerstein operator that is used in many fields, including physics and engineering. Both integral and nonlinear components are present in this problem. These problems have either no analytical answers at all or very complex ones. To provide an example, let us take $\phi = S[0, 1]$ and $\mathbb{M}_1 = \mathbb{M}_2 = C[0, 1]$. The Hammerstein operator Δ [1–4,6] is involved in this first-kind nonlinear integral equation, which is defined by

$$\Delta(v)(x) = v(x) - 8 \int_0^1 x\omega v(\omega)^3 d\omega.$$

The derivative of operator Δ is given below:

$$\Delta'(v(q))(x) = q(x) - 24 \int_0^1 x\omega v(\omega)^2 q(\omega) d\omega,$$

for $q \in C[0, 1]$. The values of the operator Δ' satisfy the hypotheses of Theorem 1, if we choose for $\xi^* = 0$

$$\aleph(\xi) = 24\xi, \aleph_0(\xi) = 12\xi, k = 2, \tilde{\aleph}(\xi) = \aleph\left(\left(1 + h_0(\xi)\right)\xi\right), \text{ and } \tilde{\aleph}_1(\xi) = \aleph\left(\left(1 + h_1(\xi)\right)\xi\right).$$

Hence, we present the convergence radii of illustration (Example 1) in Table 1.

Table 1. Radii of illustration (Example 1) of method (2).

α	ρ_0	ρ_1	ρ	r_0	r_1	r_2	r
1	0.083333	0.036537	0.036537	0.04167	0.018854	0.013528	0.013528
$\frac{1}{2}$	0.083333	0.017014	0.017014	0.018518	0.0091788	0.0072646	0.0072646
$\frac{1}{4}$	0.083333	0.0083754	0.0083754	0.0087719	0.0045910	0.0039538	0.0039538
$\frac{2}{3}$	0.083333	0.023074	0.023074	0.025641	0.012293	0.0093465	0.0093465

Example 2. A three-by-three nonlinear system is a complicated math problem found in science and engineering. It becomes more complicated when it mixes together polynomials and exponential terms. That is why we are picking one of these systems to work on in order to show the applicability of our methods. Let $\mathbb{M}_1 = \mathbb{M}_2 = \mathbb{R}^3$ and $\phi = \mathbb{S}(0, 1)$, where $u_* = (0, 0, 0)^T$. Consider P on ϕ by means of

$$\Delta(u) = \Delta(u_1, u_2, u_3) = \left(e^{u_1} - 1, \frac{e-1}{2}u_2^2 + u_2, u_3 \right)^T, \tag{39}$$

where $u = (u_1, u_2, u_3)^T$. It has the succeeding Fréchet-derivative defined by

$$\Delta'(u) = \begin{bmatrix} e^{u_1} & 0 & 0 \\ 0 & 1 + u_2(e-1) & 0 \\ 0 & 0 & 1 \end{bmatrix}.$$

Then, $\Delta'(\zeta^*) = \Delta'(\bar{\zeta}^*)^{-1} = \text{diag}\{1, 1, 1\}$; so, we have

$$\aleph(\zeta) = e^{\frac{1}{e-1}\zeta}, \aleph_0(\zeta) = (e-1)\zeta, \quad k=2, \tilde{\aleph}(\zeta) = \aleph\left(\left(1+h_0(\zeta)\right)\zeta\right),$$

$$\text{and } \tilde{\aleph}_1(\zeta) = \aleph\left(\left(1+h_1(\zeta)\right)\zeta\right).$$

Hence, we present the convergence radii of illustration (Example 2) in Table 2.

Table 2. Radii of illustration (Example 1) of method (2).

α	ρ_0	ρ_1	ρ	r_0	r_1	r_2	r
1	0.581977	0.380074	0.380074	0.382692	0.20423	0.15438	0.15438
$\frac{1}{2}$	0.581977	0.18349	0.18349	0.16433	0.103706	0.084580	0.084580
$\frac{1}{4}$	0.581977	0.092482	0.092482	0.076748	0.053632	0.046965	0.046965
$\frac{2}{3}$	0.581977	0.24531	0.24531	0.22993	0.13639	0.10772	0.10772

Examples for SLAC

We illustrate the theoretical results of the semilocal convergence on three different problems, namely, (3)–(5). In addition, we select four cases from expression (2): the fifth-order method with ($k = 2$, & $\alpha = 1$), and ($k = 2$, & $\alpha = \frac{2}{3}$), denoted as Case-1 and Case-2, respectively; and the seventh-order method with ($k = 3$, & $\alpha = 1$), and ($k = 3$, & $\alpha = \frac{2}{3}$), denoted by Case-3 and Case-4, respectively. The considered three examples, (3)–(5), are renowned 2D Bratu, BVP, and Fisher problems, which are applied science concerns that comprise three nonlinear systems of order 100×100 , 90×90 , and 100×100 , respectively.

Example 3. Bratu 2D Problem:

Defined in [18], there is the well-known 2D Bratu problem $\phi = \{(x, t) : 0 \leq x \leq 1, 0 \leq t \leq 1\}$, which is given below:

$$u_{xx} + u_{tt} + Ce^u = 0, \text{ on} \tag{40}$$

with boundary conditions $u = 0$ on $\partial\phi$.

To have a nonlinear system, we use the finite difference discretization procedure for the expression (40). The required answer at a mesh's grid points is $u_{i,j} = u(u_i, t_j)$. In addition, the step numbers x and t correspond to the corresponding directions M and N . Moreover, the step sizes in the

corresponding M and N are h and k . Applying the central differences to the PDE (40) (u_{xx} and u_{tt}) mentioned before, we obtain

$$u_{xx}(u_i, t_j) = (u_{i+1,j} - 2u_{i,j} + u_{i-1,j})/h^2, C = 0.1, t \in [0, 1]$$

As an example of a nonlinear system of size 100×100 , we picked $M = 11$ and $N = 11$. It converges to the following column vector, which is not a matrix,

$$u_* = \begin{pmatrix} 0.00111 \dots, 0.00181 \dots, 0.00226 \dots, 0.00253 \dots, 0.00266 \dots, 0.00266 \dots, 0.00253 \dots, 0.00226 \dots, \\ 0.00181 \dots, 0.00111 \dots, 0.00181 \dots, 0.00305 \dots, 0.00387 \dots, 0.00437 \dots, 0.00461 \dots, 0.00461 \dots, \\ 0.00437 \dots, 0.00387 \dots, 0.00305 \dots, 0.00181 \dots, 0.00226 \dots, 0.00387 \dots, 0.00497 \dots, 0.00565 \dots, \\ 0.00598 \dots, 0.00598 \dots, 0.00565 \dots, 0.00497 \dots, 0.00387 \dots, 0.00226 \dots, 0.00253 \dots, 0.00437 \dots, \\ 0.00565 \dots, 0.00645 \dots, 0.00684 \dots, 0.00684 \dots, 0.00645 \dots, 0.00565 \dots, 0.00437 \dots, 0.00253 \dots, \\ 0.00266 \dots, 0.00461 \dots, 0.00598 \dots, 0.00684 \dots, 0.00726 \dots, 0.00726 \dots, 0.00684 \dots, 0.00598 \dots, \\ 0.00461 \dots, 0.00266 \dots, 0.00266 \dots, 0.00461 \dots, 0.00598 \dots, 0.00684 \dots, 0.00726 \dots, 0.00726 \dots, \\ 0.00684 \dots, 0.00598 \dots, 0.00461 \dots, 0.00266 \dots, 0.00253 \dots, 0.00437 \dots, 0.00565 \dots, 0.00645 \dots, \\ 0.00684 \dots, 0.00684 \dots, 0.00645 \dots, 0.00565 \dots, 0.00437 \dots, 0.00253 \dots, 0.00226 \dots, 0.00387 \dots, \\ 0.00497 \dots, 0.00565 \dots, 0.00598 \dots, 0.00598 \dots, 0.00565 \dots, 0.00497 \dots, 0.00387 \dots, 0.00226 \dots, \\ 0.00181 \dots, 0.00305 \dots, 0.00387 \dots, 0.00437 \dots, 0.00461 \dots, 0.00461 \dots, 0.00437 \dots, 0.00387 \dots, \\ 0.00305 \dots, 0.00181 \dots, 0.00111 \dots, 0.00181 \dots, 0.00226 \dots, 0.00253 \dots, 0.00266 \dots, 0.00266 \dots, \\ 0.00253 \dots, 0.00226 \dots, 0.00181 \dots, 0.00111 \dots, \end{pmatrix}^T.$$

In Table 3, we mention the computational results based on Example 3.

Table 3. Radii of illustration (Example 3).

Method (4)	x_0	NIT	$\ x_{\sigma+1} - x_{\sigma}\ $	$\ \Delta(x_{\sigma})\ $	COC	CPU
Case-1	$\begin{bmatrix} 0.1 \sin(i\pi h) \sin(j\pi h) \\ \vdots \\ 0.1 \sin(i\pi h) \sin(j\pi h) \end{bmatrix}$	3	3.8×10^{-390}	6.7×10^{-391}	4.9999	42.3046
Case-2	$\begin{bmatrix} 0.1 \sin(i\pi h) \sin(j\pi h) \\ \vdots \\ 0.1 \sin(i\pi h) \sin(j\pi h) \end{bmatrix}$	3	2.4×10^{-402}	4.6×10^{-403}	4.9989	43.9509
Case-3	$\begin{bmatrix} 0.1 \sin(i\pi h) \sin(j\pi h) \\ \vdots \\ 0.1 \sin(i\pi h) \sin(j\pi h) \end{bmatrix}$	3	3.8×10^{-1109}	6.5×10^{-1110}	6.9999	44.0704
Case-4	$\begin{bmatrix} 0.1 \sin(i\pi h) \sin(j\pi h) \\ \vdots \\ 0.1 \sin(i\pi h) \sin(j\pi h) \end{bmatrix}$	3	2.5×10^{-1043}	4.9×10^{-1044}	6.9997	43.1974

($i, j = 1, 2, 3, \dots, 10$.)

Example 4. Boundary value problems (BVPs) are important parts of mathematics and physics. They work with differential equations that have several, frequently boundary-based conditions. The modeling of practical phenomena such as heat transfer, fluid flow, and quantum mechanics relies heavily on BVPs, which provide valuable insights into physical systems. Therefore, we chose the following BVP (see [14]):

$$u'' + a^2 u'^2 + 1 = 0 \tag{41}$$

with $u(0) = u(1) = 0$. The partition of the interval $[0, 1]$ into ℓ pieces provides us with the following:

$$\gamma_0 = 0 < \gamma_1 < \gamma_2 < \dots < \gamma_{\ell-1} < \gamma_\ell, \quad \gamma_{\ell+1} = \gamma_\ell + h, h = \frac{1}{\ell}.$$

The $u_0, u_1, u_2, \dots, u_\ell$ stands for $u(\gamma_0), u(\gamma_1), u(\gamma_2), \dots, u(\gamma_\ell)$, respectively. In addition, $u_0 = 0$, and $u_\ell = 1$. By using them in the expression (41), we have

$$u'_\theta = \frac{u_{\theta+1} - u_{\theta-1}}{2h}, \quad u''_\theta = \frac{u_{\theta-1} - 2u_\theta + u_{\theta+1}}{h^2}, \quad \theta = 1, 2, 3, \dots, \ell - 1,$$

with the help of the discretization technique. In this way, we obtain the proceeding nonlinear system of equations of order $(\theta - 1) \times (\theta - 1)$

$$u_{\theta-1} - 2u_\theta + u_{\theta+1} + \frac{\mu^2}{4}(u_{\theta+1} - u_{\theta-1})^2 + h^2 = 0. \tag{42}$$

We picked $\ell = 91$ and $\mu = \frac{1}{2}$ in order to obtain a larger nonlinear system of order 90×90 . In Table 4, we depicted the number of iterations, residual errors, COC, CPU timing, and error differences between two iterations for Example (4). The above system (42) converges to the following estimated zero:

$$\xi_* = \begin{pmatrix} 0.01853 \dots, 0.03685 \dots, 0.05497 \dots, 0.07289 \dots, 0.09061 \dots, 0.1081 \dots, 0.1254 \dots, 0.1426 \dots, \\ 0.1595 \dots, 0.1763 \dots, 0.1928 \dots, 0.2092 \dots, 0.2253 \dots, 0.2413 \dots, 0.2571 \dots, 0.2728 \dots, \\ 0.2882 \dots, 0.3035 \dots, 0.3186 \dots, 0.3335 \dots, 0.3482 \dots, 0.3628 \dots, 0.3772 \dots, 0.3914 \dots, \\ 0.4054 \dots, 0.4193 \dots, 0.4330 \dots, 0.4465 \dots, 0.4599 \dots, 0.4731 \dots, 0.4862 \dots, 0.4990 \dots, \\ 0.5118 \dots, 0.5243 \dots, 0.5367 \dots, 0.5489 \dots, 0.5610 \dots, 0.5730 \dots, 0.5847 \dots, 0.5963 \dots, \\ 0.6078 \dots, 0.6191 \dots, 0.6303 \dots, 0.6413 \dots, 0.6521 \dots, 0.6628 \dots, 0.6734 \dots, 0.6838 \dots, \\ 0.6940 \dots, 0.7042 \dots, 0.7141 \dots, 0.7239 \dots, 0.7336 \dots, 0.7432 \dots, 0.7525 \dots, 0.7618 \dots, \\ 0.7709 \dots, 0.7799 \dots, 0.7887 \dots, 0.7974 \dots, 0.8059 \dots, 0.8143 \dots, 0.8226 \dots, 0.8307 \dots, \\ 0.8387 \dots, 0.8466 \dots, 0.8543 \dots, 0.8619 \dots, 0.8693 \dots, 0.8766 \dots, 0.8838 \dots, 0.8909 \dots, \\ 0.8978 \dots, 0.9046 \dots, 0.9112 \dots, 0.9177 \dots, 0.9241 \dots, 0.9304 \dots, 0.9365 \dots, 0.9425 \dots, \\ 0.9484 \dots, 0.9541 \dots, 0.9597 \dots, 0.9652 \dots, 0.9706 \dots, 0.9758 \dots, 0.9809 \dots, 0.9858 \dots, \\ 0.9907 \dots, 0.9954 \dots \end{pmatrix}^T.$$

Table 4. Radii of illustration (Example 4).

Method (4)	x_0	NIT	$\ x_{\sigma+1} - x_\sigma\ $	$\ \Delta(x_\sigma)\ $	COC	CPU
Case-1	$\begin{bmatrix} 1.002 \\ \vdots \\ 1.002 \end{bmatrix}$	4	3.7×10^{-640}	5.4×10^{-641}	4.9964	26.7025
Case-2	$\begin{bmatrix} 1.002 \\ \vdots \\ 1.002 \end{bmatrix}$	4	3.7×10^{-640}	5.4×10^{-641}	4.9964	26.2708
Case-3	$\begin{bmatrix} 1.002 \\ \vdots \\ 1.002 \end{bmatrix}$	3	1.5×10^{-332}	2.1×10^{-333}	6.9971	17.8274

Table 4. Cont.

Method (4)	x_0	NIT	$\ x_{\sigma+1} - x_{\sigma}\ $	$\ \Delta(x_{\sigma})\ $	COC	CPU
Case-4	$\begin{bmatrix} 1.002 \\ \vdots \\ 1.002 \end{bmatrix}$	3	1.5×10^{-332}	2.1×10^{-333}	6.9971	17.7859

Example 5. We adopt a renowned Fisher’s equation [22]

$$\begin{aligned}
 \eta_t &= D\eta_{xx} + \eta(1 - \eta) = 0, \\
 &\text{along with homogeneous Neumann’s boundary conditions} \\
 \eta(x, 0) &= 1.5 + 0.5\cos(\pi x), 0 \leq x \leq 1, \\
 \eta_x(0, t) &= 0, \forall t \geq 0, \\
 \eta_x(1, t) &= 0, \forall t \geq 0.
 \end{aligned} \tag{43}$$

We refer to the D as a diffusion parameter. To obtain a nonlinear system, we use the finite difference discretization procedure for the expression (43). The required solution at a mesh’s grid points is $w_{i,j} = \eta(u_i, t_j)$. In addition, the step numbers x and t correspond to the corresponding directions M and N , respectively. Moreover, the step sizes in the corresponding M and N are h and l , respectively. Utilizing the center, backward, and forward distinctions that follow, we obtain

$$\begin{aligned}
 \eta_{xx}(u_i, t_j) &= (w_{i+1,j} - 2w_{i,j} + w_{i-1,j})/h^2, \\
 \eta_t(u_i, t_j) &= (w_{i,j} - w_{i,j-1})/l, \\
 &\text{and} \\
 \eta_x(u_i, t_j) &= (w_{i+1,j} - w_{i,j})/h, \quad t \in [0, 1],
 \end{aligned}$$

leading to

$$\frac{w_{1,j} - w_{i,j-1}}{l} - w_{i,j}(1 - w_{i,j}) - D \frac{w_{i+1,j} - 2w_{i,j} + w_{i-1,j}}{h^2}, \quad i = 1, 2, 3, \dots, M, j = 1, 2, 3, \dots, N. \tag{44}$$

Here, $h = \frac{1}{M}, l = \frac{1}{N}$ are used. The system of nonlinear equations of dimension 100×100 is obtained by choosing $M = N = 10, h = l = 0.1$, and $D = 1$. The above nonlinear system converges to the following column vector solution (not a matrix):

$$u_* = u(u_i, t_j) = \begin{pmatrix} 1.6248 \dots, 1.4519 \dots, 1.3549 \dots, 1.2941 \dots, 1.2520 \dots, 1.2201 \dots, 1.1945 \dots, 1.1731 \dots, \\ 1.1547 \dots, 1.1386 \dots, 1.5999 \dots, 1.4411 \dots, 1.3500 \dots, 1.2919 \dots, 1.2509 \dots, 1.2196 \dots, \\ 1.1943 \dots, 1.1730 \dots, 1.1546 \dots, 1.1386 \dots, 1.5541 \dots, 1.4209 \dots, 1.3407 \dots, 1.2876 \dots, \\ 1.2489 \dots, 1.2186 \dots, 1.1938 \dots, 1.1727 \dots, 1.1545 \dots, 1.1386 \dots, 1.4929 \dots, 1.3933 \dots, \\ 1.3280 \dots, 1.2816 \dots, 1.2461 \dots, 1.2173 \dots, 1.1932 \dots, 1.1725 \dots, 1.1544 \dots, 1.1385 \dots, \\ 1.4230 \dots, 1.3612 \dots, 1.3131 \dots, 1.2747 \dots, 1.2428 \dots, 1.2158 \dots, 1.1925 \dots, 1.1721 \dots, \\ 1.1542 \dots, 1.1384 \dots, 1.3514 \dots, 1.3279 \dots, 1.2976 \dots, 1.2674 \dots, 1.2394 \dots, 1.2142 \dots, \\ 1.1917 \dots, 1.1717 \dots, 1.1540 \dots, 1.1383 \dots, 1.2851 \dots, 1.2965 \dots, 1.2828 \dots, 1.2604 \dots, \\ 1.2361 \dots, 1.2126 \dots, 1.1910 \dots, 1.1714 \dots, 1.1539 \dots, 1.1383 \dots, 1.2304 \dots, 1.2702 \dots, \\ 1.2703 \dots, 1.2546 \dots, 1.2334 \dots, 1.2113 \dots, 1.1904 \dots, 1.1711 \dots, 1.1537 \dots, 1.1382 \dots, \\ 1.1921 \dots, 1.2513 \dots, 1.2613 \dots, 1.2503 \dots, 1.2313 \dots, 1.2104 \dots, 1.1899 \dots, 1.1709 \dots, \\ 1.1536 \dots, 1.1381 \dots, 1.1728 \dots, 1.2414 \dots, 1.2566 \dots, 1.2480 \dots, 1.2303 \dots, 1.2099 \dots, \\ 1.1897 \dots, 1.1708 \dots, 1.1536 \dots, 1.1381 \dots \end{pmatrix}^T.$$

We illustrate the numerical results in the Table 5.

Table 5. Radii of illustration (Example 5).

Method (4)	x_0	NIT	$\ x_{\sigma+1} - x_\sigma\ $	$\ \Delta(x_\sigma)\ $	COC	CPU
Case-1	$\begin{bmatrix} 1 + \frac{i}{100} \\ \vdots \\ 1 + \frac{i}{100} \end{bmatrix}$	4	4.6×10^{-508}	5.4×10^{-507}	12.8955	4.9998
Case-2	$\begin{bmatrix} 1 + \frac{i}{100} \\ \vdots \\ 1 + \frac{i}{100} \end{bmatrix}$	4	4.6×10^{-508}	5.4×10^{-507}	12.8211	4.9998
Case-3	$\begin{bmatrix} 1 + \frac{i}{100} \\ \vdots \\ 1 + \frac{i}{100} \end{bmatrix}$	4	1.4×10^{-1875}	1.7×10^{-1874}	32.0008	7.0000
Case-4	$\begin{bmatrix} 1 + \frac{i}{100} \\ \vdots \\ 1 + \frac{i}{100} \end{bmatrix}$	4	1.4×10^{-1875}	1.7×10^{-1874}	32.3537	7.0000

($i = 1, 2, 3, \dots, 100$.)

5. Conclusions

This paper concludes by highlighting the nature of a convergence analysis in iterative approaches, especially when there are no explicit assurances about convergence. It is shown that the investigation of convergence behaviors is different from conventional assumptions by taking into account derivatives such as Δ'' , Δ''' , and $\Delta^{(4)}$ with the help of Taylor series expansion. In the previous work, there was no computable radius of convergence, and there was missing information about the choice of the initial point. However, we made a significant contribution by offering computable bounds on $\|x_\sigma - \xi^*\|$, which allows one to make well-informed decisions about the number of iterations needed to achieve acceptable error tolerances. Furthermore, the computation of a convergence radius makes the process of choosing an initial location more confident. This study’s relevance to a wide range of issues is highlighted by its emphasis on Banach space operators. In addition, we also introduce a semilocal analysis and careful consideration of generalized continuity assumptions, which significantly advance our understanding of convergence behaviors in iterative methods. We have checked the semilocal convergence by offering practical examples for real-world applications. In particular, the drawbacks of the Taylor expansions series approach listed in $(L_1) - (L_6)$ of the introduction have all been positively addressed by $(L_1)' - (L_6)'$. The technique developed in this paper can also be used with the same benefits on other iterative methods requiring inverses of linear operators or not in an analogous fashion [5–14,23]. This will be the direction of our future research, including the further weakening of the sufficient convergence conditions presented in this article.

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References

1. Argyros, I.K.; Magreñán, A.A. *A Contemporary Study of Iterative Methods*; Academic Press: Cambridge, MA, USA; Elsevier: Amsterdam, The Netherlands, 2018.
2. Argyros, I.K.; George, S. *Mathematical Modeling for the Solutions with Application*; Nova Publisher: New York, NY, USA, 2019; Volume III.
3. Argyros, I.K. Unified Convergence Criteria for iterative Banach space valued methods with applications. *Mathematics* **2021**, *9*, 1942. [[CrossRef](#)]
4. Ortega, J.M.; Rheinboldt, W.C. *Iterative Solution of Nonlinear Equations in Several Variables*; Academic Press: New York, NY, USA, 1970.
5. Ostrowski, A.M. *Solution of Equations and Systems of Equations*; Academic Press: New York, NY, USA, 1966.
6. Gutiérrez, J.M.; Hernández, M.A. A family of Chebyshev-Halley type methods in Banach spaces. *Bull. Aust. Math. Soc.* **1997**, *55*, 113–130. [[CrossRef](#)]
7. Simpson, R.B. A method for the numerical determination of bifurcation states of nonlinear systems of equations. *SIAM J. Numer. Anal.* **1975**, *12*, 439–451. [[CrossRef](#)]
8. Xiao, X.; Yin, H. A new class of methods with high-order of convergence for solving systems of nonlinear equations. *Appl. Math. Comput.* **2015**, *264*, 300–309.
9. Behl, R.; Sarria, F.; González, R.; Magreñán, Á.A. Highly efficient family of iterative methods for solving nonlinear methods for solving nonlinear models. *J. Comput. Appl. Math.* **2016**, *346*, 110–132. [[CrossRef](#)]
10. Cordero, A.; Torregrosa, J.R. Variants of Newton's method using fifth-order quadrature formulas. *Appl. Math. Comput.* **2007**, *190*, 686–698. [[CrossRef](#)]
11. Cordero, A.; Hueso, J.L.; Martínez, E.; Torregrosa, J.R. Increasing the convergence order of an iterative method for nonlinear systems. *Appl. Math. Lett.* **2012**, *25*, 2369–2374. [[CrossRef](#)]
12. Ezquerro, J.A.; Grau-Sánchez, M.; Grau, Á.; Hernández, M.A.; Noguera, M.; Romero, N. On iterative methods with accelerated convergence for solving systems of nonlinear equations. *J. Optim. Theory Appl. Math. Comput.* **2011**, *151*, 163–174. [[CrossRef](#)]
13. Frontini, M.; Sormani, E. Some variant of Newton's method with third-order convergence. *Appl. Math. Comput.* **2003**, *140*, 419–426. [[CrossRef](#)]
14. Frontini, M.; Sormani, E. Third-order methods from quadrature formulae for solving systems of nonlinear equations. *Appl. Math. Comput.* **2004**, *149*, 771–782. [[CrossRef](#)]
15. Grau-Sánchez, M.; Grau, Á.; Noguera, M. Ostrowski type methods for solving systems of nonlinear equations. *Appl. Math. Comput.* **2011**, *218*, 2377–2385. [[CrossRef](#)]
16. Grau-Sánchez, M.; Grau, Á.; Noguera, M. On the computational efficiency index and some iterative methods for solving systems of nonlinear equations. *J. Comput. Appl. Math.* **2011**, *236*, 1259–1266. [[CrossRef](#)]
17. Grau-Sánchez, M.; Grau, Á.; Noguera, M. Frozen divided difference scheme for solving systems of nonlinear equations. *J. Comput. Appl. Math.* **2011**, *235*, 1739–1743. [[CrossRef](#)]
18. Homeier, H.H.H. A modified Newton method with cubic convergence: The multivariate case. *J. Comput. Appl. Math.* **2004**, *169*, 161–169. [[CrossRef](#)]
19. Kapania, R.K. A pseudo-spectral solution of 2-parameter Bratu's equation. *Comput. Mech.* **1990**, *6*, 55–63. [[CrossRef](#)]
20. Noor, M.A.; Wasseem, M. Some iterative methods for solving a system of nonlinear equations. *Appl. Math. Comput.* **2009**, *57*, 101–106. [[CrossRef](#)]
21. Sharma, J.R.; Guha, R.K.; Sharma, R. An efficient fourth order weighted-Newton method for systems of nonlinear equations. *Numer. Algor.* **2013**, *62*, 307–323. [[CrossRef](#)]
22. Sharma, J.R.; Gupta, P. An efficient fifth order method for solving systems of nonlinear equations. *Comput. Math. Appl.* **2014**, *67*, 591–601. [[CrossRef](#)]
23. Shakhno, S.M. On a Kurchatov's method of linear interpolation for solving nonlinear equations. *PAMM Proc. Appl. Math. Mech.* **2004**, *4*, 650–651. [[CrossRef](#)]
24. Weerakoon, S.; Fernando, T.G.I. A variant of Newton's method with accelerated third-order convergence. *Appl. Math. Lett.* **2000**, *13*, 87–93. [[CrossRef](#)]
25. Dennis, E., Jr.; John, R.; Moré, B. *Numerical Methods for Unconstrained Optimization and Nonlinear Equations*; Society for Industrial and Applied Mathematics: Philadelphia, PA, USA, 1996.
26. Kantorovich, L.V.; Akilov, G.P. *Functional Analysis*; Elsevier: Nauka, Moscow, 1984. (In Russian)

27. Potra, F.A. On an iterative algorithm of order 1.839... for solving nonlinear operator equations. *Numer. Funct. Anal. Optim.* **1985**, *7*, 75–106. [[CrossRef](#)]
28. Fernando, T.G.L.; Weerakoon, S. Improved Newton's method for finding roots of a nonlinear equation. In Proceedings of the 53rd Annual Sessions of Sri Lanka Association for the Advancement of Science (SLAA), Matara, Sri Lanka, 8–12 December 1997; pp. E1–E22.
29. Darvishi, M.T.; Barati, A. A fourth-order method from quadrature formulae to solve systems of nonlinear equations. *Appl. Math. Comput.* **2007**, *188*, 257–261. [[CrossRef](#)]

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