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ON AN ITERATIVE METHOD FOR UNCONSTRAINED OPTIMIZATION

Abstract. We present a local and a semi-local convergence analysis of an iterative method for approximating zeros of derivatives for solving univariate and unconstrained optimization problems. In the local case, the radius of convergence is obtained, whereas in the semi-local case, sufficient convergence criteria are presented. Numerical examples are also provided.

1. Introduction. In this study we are concerned with the problem of approximating a solution x^* of the equation

$$(1.1) \quad F'(x) = 0,$$

where F is a differentiable function defined on a convex subset D of S where S is \mathbb{R} or \mathbb{C} . Many problems in computational sciences such as univariate minimization problems and unconstrained optimization problems, and in other disciplines, can be brought in a form like equation (1.1) using mathematical modeling [1, 2, 5, 16]. The solutions of equation (1.1) can be found in closed form only in special cases. That is why most methods of solving these equations are usually iterative. In particular, the practice of numerical functional analysis for finding such solutions involves variants of Newton-like methods. The study of convergence of iterative procedures is usually based of two types: semi-local and local convergence analysis. The semi-local convergence analysis is based on the information around an initial point, to give criteria ensuring the convergence of the method, while the local analysis is based on the information around a solution, to find estimates of the radii of convergence balls. There are many methods for solving equation (1.1)

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[12, 13, 14, 16, 18, 19]. It is well known that higher dimensional problems involve steps of searching for extrema along certain derivatives in \mathbb{R}^m . So, finding a step size μ_k along the directional vector v_i involves solving the subproblem

$$(1.2) \quad \min F(x_{n+1}) = F(x_n + \mu_n d_n),$$

which is a unidimensional problem. Consequently, the unidimensional search methods are important for any iterative procedure in higher dimensions.

Newton’s method for solving nonlinear, univariate and unconstrained optimization problems is defined by

$$(1.3) \quad x_{n+1} = x_n - \frac{F'(x_n)}{F''(x_n)} \quad \text{for } n = 0, 1, 2, \dots,$$

where $x_0 \in D$ is an initial point. With suitable assumptions, this method converges to x^* with an order of convergence at least two [2, 5, 16, 17]. In particular, its variants may achieve a third order of convergence [11, 10, 15]. In this paper we focus on a class of second derivative free methods. Notice that the evaluation of the second derivatives can be avoided if replaced by

$$(1.4) \quad F''(x_{n+1}) \approx \frac{F'(x_n) - F'(x_{n-1})}{x_n - x_{n-1}},$$

leading to the secant-method defined by

$$(1.5) \quad x_{n+1} = x_n - \frac{(x_n - x_{n-1})F'(x_n)}{F'(x_n) - F'(x_{n-1})} \quad \text{for } n = 0, 1, 2, \dots,$$

where $x_{-1}, x_0 \in D$ are initial points. The secant-method is Q -superlinearly convergent with order $p = \frac{1}{2}(\sqrt{5} + 1) \approx 1.618$ [2, 5, 16].

A more efficient replacement of the second derivative was given by Kahya and Chen [14]:

$$(1.6) \quad F''(x_{n+1}) \approx A_n = \frac{2\{F'(x_n) - \frac{F(x_n) - F(x_{n-1})}{x_n - x_{n-1}}\}}{x_n - x_{n-1}},$$

leading to the method

$$(1.7) \quad x_{n+1} = x_n - A_n^{-1}F'(x_n) \quad \text{for } n = 0, 1, 2, \dots$$

Suppose that $F'(x^*) = 0$, $F''(x^*) \neq 0$, $F'''(x^*) \neq 0$. Then method (1.7) converges with the same order of convergence as the secant method (1.5), but the leading error is $-\frac{1}{3}F'''(x_n)(x_n - x_{n-1})$, which is better approximation than $-\frac{1}{2}F'''(x_n)(x_n - x_{n-1})$ given for the secant method [14]. Although the convergence order of method (1.7) was established in [14], we have no information about the radius of convergence in the local case, and we do not have sufficient semi-local convergence criteria. The contribution of this prepare is twofold: First we provide a local convergence analysis and secondly

we present a semi-local convergence analysis. It is also worth noticing that we make no assumptions on the third derivative.

The paper is organized as follows: In Section 2 we present the local and semi-local convergence analysis for method (1.7). The numerical examples are presented in Section 3.

2. Convergence analysis. In this section we present first the local convergence and then the semi-local convergence analysis of method (1.7). Let $U(w, \rho)$, $\bar{U}(w, \rho)$ denote the open and closed intervals in S with center $w \in S$ and of radius $\rho > 0$.

THEOREM 2.1. *Let $F : D \subseteq S \rightarrow S$ be a twice differentiable function. Suppose that there exist $x^* \in D$ and $\ell > 0$ such that*

$$(2.1) \quad F'(x^*) = 0, \quad F''(x^*) \neq 0,$$

$$(2.2) \quad |F''(x^*)^{-1}(F''(x) - F''(x^*))| \leq \ell|x - x^*| \quad \text{for each } x \in D,$$

$$(2.3) \quad \bar{U}(x^*, r) \subseteq D,$$

where

$$(2.4) \quad r = \frac{6}{23\ell}.$$

Then the sequence $\{x_n\}$ generated by method (1.7) for $x_{-1}, x_0 \in U(x^*, r)$ is well defined, remains in $U(x^*, r)$ for each $n = -1, 0, 1, 2, \dots$ and converges to x^* . Moreover,

$$(2.5) \quad |x_{n+1} - x^*| \leq \frac{\alpha_n}{\beta_n} |x_n - x^*| < |x_n - x^*| < r,$$

where

$$(2.6) \quad \alpha_n = \int_0^1 \int_0^1 |x_{n-1} + \theta(x_n - x_{n-1}) - x^* + s(1 - \theta)(x_n - x_{n-1})|(1 - \theta) d\theta ds + \frac{1}{2}|x_n - x^*|\ell,$$

$$(2.7) \quad \beta_n = 1 - 2\ell \int_0^1 \int_0^1 |x_{n-1} + \theta(x_n - x_{n-1}) - x^* + s(1 - \theta)(x_n - x_{n-1})|(1 - \theta) d\theta ds.$$

Furthermore, if there exists $T \in [r, 2/L)$ such that $\bar{U}(x^*, T) \subseteq D$, then the limit point x^* is the only solution of the equation $F'(x) = 0$ in $\bar{U}(x^*, T)$.

Proof. By hypothesis $x_{-1}, x_0 \in U(x^*, r)$. Suppose that $x_{k-1}, x_k \in U(x^*, r)$ for all $k \leq n$, where n is an integer and $x_{k-1} \neq x_k$. We shall show that A_k is invertible. Using the definition of A_k , conditions (2.1)–(2.4), and (2.7), we get

$$\begin{aligned}
 & |F''(x^*)^{-1}(A_k - F''(x^*))| \\
 &= 2 \left| F''(x^*)^{-1} \left[\frac{F'(x_k) - \frac{F(x_k) - F(x_{k-1})}{x_k - x_{k-1}}}{x_k - x_{k-1}} - \frac{1}{2} F''(x^*) \right] \right| \\
 &= \frac{2 |F''(x^*)^{-1} [F'(x_k)(x_k - x_{k-1}) - (F(x_k) - F(x_{k-1})) - \frac{1}{2} F''(x^*)(x_k - x_{k-1})^2]|}{|x_k - x_{k-1}|^2} \\
 &= \frac{2 |F''(x^*)^{-1} [\int_0^1 (F'(x_k) - F'(x_k(\theta)))(x_k - x_{k-1}) d\theta - \frac{1}{2} F''(x^*)(x_k - x_{k-1})^2]|}{|x_k - x_{k-1}|^2} \\
 &= \frac{2 |\int_0^1 \int_0^1 F''(x^*)^{-1} [F''(x_k(\theta) + s(1-\theta)(x_k - x_{k-1})) - F''(x^*)(1-\theta)(x_k - x_{k-1})^2] d\theta ds|}{|x_k - x_{k-1}|^2} \\
 &\leq \frac{2\ell |x_k - x_{k-1}|^2 \int_0^1 \int_0^1 |x_{k-1} + (\theta(x_k - x_{k-1}) - x^* + s(1-\theta)(x_k - x_{k-1}))| (1-\theta) d\theta ds}{|x_k - x_{k-1}|^2} \\
 &\leq 2\ell \int_0^1 \int_0^1 [(1-\theta)|x_{k-1} - x^*| + \theta|x_k - x^*| + s(1-\theta)(|x_{k-1} - x^*| + |x_k - x^*|)] \\
 &\hspace{25em} \times (1-\theta) d\theta ds \\
 &< 2\ell \int_0^1 \int_0^1 [r + 2s(1-\theta)r](1-\theta) d\theta ds = \frac{5}{3}\ell r < 1.
 \end{aligned}$$

Here,

$$(2.8) \quad x_k(\theta) = x_{k-1} + \theta(x_k - x_{k-1}).$$

It follows from (2.8) and the Banach lemma on invertible functions that A_k is invertible and

$$(2.9) \quad |A_k^{-1}F''(x^*)| \leq \frac{1}{\beta_k} < \frac{1}{1 - \frac{5}{3}\ell r}.$$

Hence, x_{k+1} is well defined by method (1.7).

Next, we need an estimate on $A_k(x_k - x^*) - F'(x_k)$. In view of the definition of A_k , we have

$$\begin{aligned}
 (2.10) \quad & A_k(x_k - x^*) - F'(x_k) \\
 &= 2 \left[\frac{F'(x_k) - \frac{F(x_k) - F(x_{k-1})}{x_k - x_{k-1}}}{x_k - x_{k-1}} (x_k - x^*) - \frac{1}{2} F'(x_k) \right] \\
 &= 2 \left[\frac{(F'(x_k) - \int_0^1 F'(x_k(\theta))) (x_k - x^*) d\theta}{x_k - x_{k-1}} - \frac{1}{2} F'(x_k) \right] \\
 &= \frac{2}{x_k - x_{k-1}} \left[\int_0^1 (F'(x_k) - F'(x_k(\theta)))(x_k - x^*) d\theta - \frac{1}{2} F'(x_k)(x_k - x_{k-1}) \right]
 \end{aligned}$$

$$\begin{aligned}
 &= \frac{2 \left[\int_0^1 \int_0^1 (F''(x_k(\theta) + s(1-\theta)(x_k - x_{k-1})) - F''(x^*)) (1-\theta)(x_k - x_{k-1})(x_k - x^*) d\theta ds \right]}{x_k - x_{k-1}} \\
 &\quad + \frac{1}{x_k - x_{k-1}} \left[F''(x^*)(x_k - x_{k-1})(x_k - x^*) - F'(x_k)(x_k - x_{k-1}) \right] \\
 &= \frac{2 \left[\int_0^1 \int_0^1 (F''(x_k(\theta) + s(1-\theta)(x_k - x_{k-1})) - F''(x^*)) (1-\theta)(x_k - x_{k-1})(x_k - x^*) d\theta ds \right]}{x_k - x_{k-1}} \\
 &\quad + \frac{F''(x^*)(x_k - x_{k-1})(x_k - x^*) - \int_0^1 F''(x^* + \theta(x_k - x^*)) (x_k - x^*)(x_k - x_{k-1}) d\theta}{x_k - x_{k-1}} \\
 &= 2 \int_0^1 \int_0^1 (F''(x_k(\theta) + s(1-\theta)(x_k - x_{k-1})) - F''(x^*)) (1-\theta)(x_k - x^*) d\theta ds \\
 &\quad + \int_0^1 (F''(x^*) - F''(x^* + \theta(x_k - x^*))) (x_k - x^*) d\theta.
 \end{aligned}$$

Using (2.1)–(2.5), and (2.10), we obtain

$$\begin{aligned}
 (2.11) \quad &|F''(x^*)^{-1}(x^*)[A_k(x_k - x^*) - F'(x_k)]| \\
 &\leq \ell \left[2 \int_0^1 \int_0^1 |x_k(\theta) - x^* + s(1-\theta)(x_k - x_{k-1})| (1-\theta) d\theta ds + \frac{1}{2} |x_k - x^*| \right] |x_k - x^*| \\
 &\leq \alpha_k |x_k - x^*| < \frac{13\ell r}{6} |x_k - x^*|.
 \end{aligned}$$

In view of the identity

$$\begin{aligned}
 (2.12) \quad x_{k+1} - x^* &= x_k - x^* - A_k^{-1}F'(x_k) \\
 &= [A_k^{-1}F''(x^*)][F''(x^*)^{-1}(A_k(x_k - x^*) - F'(x_k))],
 \end{aligned}$$

(2.9) and (2.11), we have

$$\begin{aligned}
 |x_{k+1} - x^*| &\leq |A_k^{-1}F''(x^*)| |F''(x^*)^{-1}(A_k(x_k - x^*) - F'(x_k))| \\
 &\leq \frac{\alpha_k}{\beta_k} |x_k - x^*| < \frac{13\ell r}{6(1-5/3)} |x_k - x^*| = |x_k - x^*| < r,
 \end{aligned}$$

so (2.5) holds for all k and $x_{k+1} \in U(x^*, r)$. Moreover, from the estimate

$$|x_{k+1} - x^*| < |x_k - x^*|,$$

we deduce that $\lim_{k \rightarrow \infty} x_k = x^*$.

Finally, to show the uniqueness part, let $y^* \in \bar{U}(x^*, T)$ with $F'(y^*) = 0$. Define $T = |x^* - y^*|$ and $Q = \int_0^1 F''(y^* + \theta(x^* - y^*)) d\theta$. Using (2.2) we get

$$|F''(x^*)^{-1}(Q - F'(x^*))| \leq \int_0^1 \ell |y^* + \theta(x^* - y^*)| d\theta \leq \frac{\ell}{2} |x^* - y^*| = \frac{\ell}{2} T < 1.$$

It follows that Q is invertible. Then, from the identity $0 = F'(x^*) - F'(y^*) = Q(x^* - y^*)$, we deduce that $x^* = y^*$. ■

THEOREM 2.2. *Let $F : D \subseteq S \rightarrow S$ be a twice differentiable function. Suppose that there exist $x_{-1}, x_0 \in D$, $L_0, L > 0$, $b, c \geq 0$, and $q \in [0, 1)$ such that*

$$(2.13) \quad F''(x_0) \neq 0, A_0 \text{ is invertible,}$$

$$(2.14) \quad |x_{-1} - x_0| \leq b,$$

$$(2.15) \quad |A_0^{-1}F'(x_0)| \leq c,$$

$$(2.16) \quad |F''(x_0)^{-1}(F''(x) - F''(x_0))| \leq L_0|x - x_0| \quad \text{for all } x \in D,$$

$$(2.17) \quad |F''(x_0)^{-1}(F''(x) - F''(y))| \leq L|x - y| \quad \text{for all } x, y \in D,$$

$$(2.18) \quad g(q) \leq 0,$$

$$(2.19) \quad \bar{U}(x_0, R) \subseteq D,$$

$$(2.20) \quad R < \frac{3}{5L_0},$$

where $g(t) = (5 - 3L_0b)t^2 - (3(1 + 3Lb) - 5L_0(b + c))t + 9L(b + c)$ and

$$(2.21) \quad R = \frac{3q}{9L + 5L_0q}.$$

Then the sequence $\{x_n\}$ generated by method (1.7) is well defined, remains in $\bar{U}(x_0, R)$ for all $n = 0, 1, 2, \dots$, and converges to a solution $x^* \in \bar{U}(x_0, R)$ of equation (1.1). Moreover, for each $n = 0, 1, 2, \dots$,

$$(2.22) \quad |x_{n+2} - x_{n+1}| \leq \frac{\gamma_n}{\delta_n},$$

where

$$(2.23) \quad \gamma_n = L\left(\frac{1}{2}|x_{n+1} - x_n| + |x_n - x_{n-1}|\right)|x_{n+1} - x_n|$$

and

$$(2.24) \quad \delta_n = \int_0^1 \int_0^1 |x_{n-1} + \theta(x_n - x_{n-1}) - x_0 + s(1 - \theta)(x_n - x_{n-1})|(1 - \theta) d\theta ds.$$

Furthermore, if there exists $R_1 \in [R, 2/L_0 - R)$ such that $\bar{U}(x_0, R_1) \subseteq D$, then the limit point x^* is the only solution of the equation $F'(x) = 0$ in $\bar{U}(x_0, R_1)$.

Proof. By the definition of R , (2.18), (2.19) and (2.21), we have

$$b + \frac{c}{1 - q} \leq \frac{3q}{9L + 5L_0q} = R.$$

It then follows that $x_{-1}, x_0 \in \overline{U}(x_0, R)$, and x_1 is well defined, since A_0 is invertible. Suppose that $x_{k-1}, x_k \in \overline{U}(x_0, R)$ for all $k \leq n$, where n is an integer and $x_{k-1} \neq x_k$. As in (2.8), (2.10) but replacing x_0, β_n by x^*, δ_n , respectively and using (2.16) instead of (2.2), we find that A_{k+1} is invertible and

$$(2.25) \quad |A_{k+1}^{-1}F''(x_0)| \leq \frac{1}{\delta_{k+1}}.$$

Hence, x_{k+1} is well defined by method (1.7). Using (1.7) we obtain

$$\begin{aligned} (2.26) \quad & F'(x_{k+1}) = F'(x_{k+1}) - F'(x_k) - A_k(x_{k+1} - x_k) \\ & = [F'(x_{k+1}) - F'(x_k) - F''(x_k)(x_{k+1} - x_k)] + (F''(x_k) - A_k)(x_{k+1} - x_k) \\ & = \int_0^1 (F''(x_k + \theta(x_{k+1} - x_k)) - F''(x_k))(x_{k+1} - x_k) d\theta \\ & \quad + \left(F''(x_k) - \frac{2(F'(x_k)(x_k - x_{k-1}) - (F(x_k) - F(x_{k-1})))}{(x_k - x_{k-1})^2} \right) (x_{k+1} - x_k) \\ & = \int_0^1 (F''(x_k + \theta(x_{k+1} - x_k)) - F''(x_k))(x_{k+1} - x_k) d\theta \\ & \quad - 2 \left(\frac{F'(x_k) - \int_0^1 F'(x_{k-1} + \theta(x_k - x_{k-1})) d\theta}{x_k - x_{k-1}} - \frac{F''(x_k)}{2} \right) (x_{k+1} - x_k) \\ & = \int_0^1 (F''(x_k + \theta(x_{k+1} - x_k)) - F''(x_k))(x_{k+1} - x_k) d\theta \\ & \quad - 2 \int_0^1 \int_0^1 (F''(x_k(\theta) + s(1-\theta)(x_k - x_{k-1})) - F''(x_k))(1-\theta)(x_{k+1} - x_k) d\theta ds, \end{aligned}$$

where $x_k(\theta)$ in the last equality is defined by (2.8).

Using (2.17) and (2.26), we get

$$(2.27) \quad |F''(x_0)^{-1}F'(x_{k+1})| \leq \frac{L}{2}|x_{k+1} - x_k|^2 + 2L \int_0^1 \int_0^1 (1+s)(1-\theta) d\theta ds |x_k - x_{k-1}| |x_{k+1} - x_k| = \gamma_{k+1}.$$

Then, using method (1.7), (2.25) and (2.27), we get

$$|x_{k+2} - x_{k+1}| \leq |A_{k+1}^{-1}F''(x_0)| |F''(x_0)^{-1}F(x_{k+1})| \leq \frac{\gamma_{k+1}}{\delta_{k+1}},$$

which completes the induction for (2.22).

Using (2.22) and the definition of R , we have

$$\begin{aligned}
 & |x_{k+2} - x_{k+1}| \\
 & \leq \frac{L(\frac{1}{2}(|x_{k+1}-x_0|+|x_k-x_0|)+|x_k-x_0|+|x_{k-1}-x_0|)|x_{k+1}-x_k|}{1-2L_0\int_0^1\int_0^1((1-\theta)|x_{k-1}-x_0|+\theta|x_k-x_0|+s(1-\theta)(|x_k-x_0|+|x_{k-1}-x_0|))(1-\theta)d\theta ds} \\
 & \leq \frac{L(\frac{1}{2}(2R)+2R)}{1-\frac{5}{3}L_0R}|x_{k+1}-x_k| = \frac{3LR}{1-\frac{5}{3}L_0R}|x_{k+1}-x_k| \\
 & \leq q|x_{k+1}-x_k| \leq q^{k+1}|x_1-x_0| \leq q^{k+1}c,
 \end{aligned}$$

so

$$\begin{aligned}
 |x_{k+2} - x_0| & \leq |x_{k+2} - x_{k+1}| + |x_{k+1} - x_k| + \dots + |x_1 - x_0| + |x_0 - x_{-1}| \\
 & \leq b + c + cq + \dots + cq^{k+1} \\
 & = b + c\frac{1 - q^{k+2}}{1 - q} < b + \frac{c}{1 - q} \leq \frac{3q}{9L + 5L_0q} = R,
 \end{aligned}$$

which implies that $x_{k+2} \in \bar{U}(x_0, R)$.

Let $m \geq 0$. Then

$$\begin{aligned}
 (2.28) \quad & |x_{m+k} - x_k| \\
 & \leq |x_{m+k} - x_{m+k-1}| + |x_{m+k-1} - x_{m+k-2}| + \dots + |x_{k+1} - x_k| \\
 & \leq (q^{m+k-1} + q^{m+k-2} + \dots + q^k)c = \frac{1 - q^m}{1 - q}q^k c.
 \end{aligned}$$

It follows that $\{x_k\}$ is a Cauchy sequence in S , and as such it converges to some $x^* \in \bar{U}(x_0, R)$ (since $\bar{U}(x_0, R)$ is a closed interval). By letting $k \rightarrow \infty$ in (2.27), (2.28), we get $F'(x^*) = 0$ and

$$(2.29) \quad |x^* - x_k| \leq \frac{q^k c}{1 - q} < R.$$

In particular, if $k = 0$, we infer from (2.29) that $x^* \in \bar{U}(x_0, R)$.

For uniqueness, as in the case of Theorem 2.1 but using (2.16) instead of (2.2) for $y^* \in \bar{U}(x_0, R_1)$ with $F'(y^*) = 0$, we get

$$\begin{aligned}
 |F''(x_0)^{-1}(Q - F'(x_0))| & \leq \int_0^1 L_0|y^* + \theta(x^* - y^*) - x_0| d\theta \\
 & \leq L_0 \int_0^1 [\theta|x^* - x_0| + (1 - \theta)|y^* - x_0|] d\theta \\
 & \leq \frac{L_0}{2}(R + R_1) < 1.
 \end{aligned}$$

It follows again that Q is invertible and $x^* = y^*$. ■

REMARK 2.3. (a) It follows from (2.22) that there exist constants $c_1, c_2 \geq 0$ such that

$$(2.30) \quad |x_{n+2} - x_{n+1}| \leq c_1|x_{n+1} - x_n|^2 + c_2|x_{n+1} - x_n||x_n - x_{n-1}|.$$

It is well known that (2.30) implies that the convergence order of method (1.7) is $p = \frac{1}{2}(\sqrt{5}+1) \approx 1.618$, which is the same order of convergence as that of the secant method (1.5). However, as already noted in the Introduction, the leading error is a better approximation than the one obtained in [14].

(b) Notice that $L_0 \leq L$ and L/L_0 can be arbitrarily large [3, 4, 7, 8, 9].

3. Numerical examples. In this section we present two numerical examples.

EXAMPLE 3.1. Let $S = \mathbb{R}$, $D = \bar{U}(0, 1)$ and define a function F on D by

$$(3.1) \quad F(x) = \exp(x) - x.$$

Then, using (3.2), we see that conditions (2.1)–(2.3) are satisfied if $x^* = 0$, $\ell = e - 1$ and

$$r = \frac{6}{23(e - 1)} = 0.15192001.$$

Then, according to Theorem 2.2, method (1.7) converges to the root 0 if $x_1, x_0 \in U(0, 0.15192001)$. Concerning the uniqueness ball, since

$$T \in [r, 2/L) = [0.15192001, 1.163953414), \quad \bar{U}(0, T) \subseteq D,$$

we can choose $T = 1$.

EXAMPLE 3.2. Let $S = \mathbb{R}$, $x_0 = 0.578$, $R_0 = 0.03$, $D = \bar{U}(x_0, R) = [0.548, 0.608]$ and define a function F on D by

$$(3.2) \quad F(x) = x^3 - x.$$

Then by (2.17), (2.18) and (3.2) we get $L_0 = L = 1.7301038606$. Let $x_{-1} = 0.573$. Then

$$A_0 = 2(2x_0 + x_{-1}) = 3.458, \quad A_0^{-1} = 0.2891845,$$

$$b = 0.005, \quad c = 0.000651243,$$

$$g(t) = 4.974048443t^2 - 3.028968486t + 0.050861191,$$

$$q = 0.017282051, \quad g(q) = 0, \quad R = 0.027583573 < \frac{3}{5L_0} = 0.3468$$

and $\bar{U}(x_0, R) \subset \bar{U}(x_0, R_0)$. Hence, all hypotheses of Theorem 2.2 are satisfied. Therefore, method (1.7) converges to $x^* = \sqrt{1/3} = 0.577350268 \in \bar{U}(x_0, R)$.

Finally, concerning the uniqueness ball, since

$$R_1 \in [R, 2/L_0 - R) = [0.027583573, 1.128416427)$$

and $\bar{U}(x_0, R_1) \subseteq D = \bar{U}(x_0, 1 - \alpha)$, we can choose $R_1 = 0.03$.

References

- [1] S. Amat, S. Busquier, and J. M. Gutiérrez, *Geometric constructions of iterative functions to solve nonlinear equations*, J. Comput. Appl. Math. 157 (2003), 197–205.
- [2] I. K. Argyros, *Convergence and Applications of Newton-Type Iterations*, Springer, New York, 2008.
- [3] I. K. Argyros and J. Chen, *Improved results on estimating and extending the radius of an attraction ball*, Appl. Math. Lett. 23 (2010), 404–408.
- [4] I. K. Argyros and J. Chen, *On local convergence of a Newton-type method in Banach space*, Int. J. Comput. Math. 86 (2009), 1366–1374.
- [5] I. K. Argyros and S. Hilout, *Computational Methods in Nonlinear Analysis*, World Sci., 2013.
- [6] J. Chen, *Some new iterative methods with three-order convergence*, Appl. Math. Comput. 181 (2006), 1519–1522.
- [7] J. Chen and W. Li, *Convergence behaviour of inexact Newton methods under weak Lipschitz condition*, J. Comput. Appl. Math. 191 (2006), 143–164.
- [8] J. Chen and Z. Shen, *Convergence analysis of the secant type methods*, Appl. Math. Comput. 188 (2007), 514–524.
- [9] J. Chen and Q. Sun, *The convergence ball of Newton-like methods in Banach space and applications*, Taiwanese J. Math. 11 (2007), 383–397.
- [10] M. Frontini and E. Sormani, *Some variant of Newton's method with third-order convergence*, Appl. Math. Comput. 140 (2003), 419–426.
- [11] J. M. Gutiérrez and M. A. Hernández, *Third-order iterative methods for operators with bounded second derivative*, J. Comput. Appl. Math. 82 (1997), 171–183.
- [12] E. Kahya, *A new unidimensional search method for optimization: Linear interpolation method*, Appl. Math. Comput. 171 (2005), 912–926.
- [13] E. Kahya, *Modified secant-type methods for unconstrained optimization*, Appl. Math. Comput. 181 (2006), 1349–1356.
- [14] E. Kahya and J. Chen, *A modified secant method for unconstrained optimization*, Appl. Math. Comput. 186 (2007), 1000–1004.
- [15] A. Y. Özban, *Some new variants of Newton's method*, Appl. Math. Lett. 17 (2004), 677–682.
- [16] J. F. Traub, *Iterative Methods for the Solution of Equations*, Prentice-Hall, Englewood Cliffs, NJ, 1964.
- [17] J. F. Traub and H. Woźniakowski, *Convergence and complexity of Newton iteration*, J. Assoc. Comput. Mach. 29 (1979), 250–258.
- [18] M. Rao and N. D. Bhat, *A new unidimensional search scheme for optimization*, Comput. Chem. Engrg. 15 (1991), 671–674.
- [19] C. Tseng, *A Newton-type univariate optimization algorithm for locating the nearest extremum*, Eur. J. Oper. Res. 105 (1998), 236–246.

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