Convergence analysis for a new faster four steps iterative algorithm with an application

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Abstract: In this paper, we introduce a four step iterative algorithm which converges faster than some leading iterative algorithms in the literature. We show that our new iterative scheme is $T$-stable and data dependent. As an application, we use the new iterative algorithm to find the unique solution of a nonlinear integral equation. Our results are generalizations and improvements of several well known results in the existing literature.

Keywords: Banach space; Stability; Contraction map; Data dependence; Strong convergence; Iterative algorithm; Nonlinear integral equation.

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1. Introduction

Throughout this paper, let $\Gamma$ be a nonempty closed subset of a real Banach space $\Psi$, $\mathbb{N}$ the set of all natural numbers, $\mathbb{R}$ the set of all real numbers and $C([d,e])$ denotes the set of all continuous real-valued functions defined on $[d,e] \subset \mathbb{R}$.

A mapping $T: \Gamma \rightarrow \Gamma$ is called

- **contraction** if there exists a constant $\delta \in [0, 1)$ such that
  \[ \|T\psi - T\zeta\| \leq \delta \|\psi - \zeta\|, \ \forall \psi, \zeta \in \Gamma; \]  \hspace{1cm} (1)

- **nonexpansive** if
  \[ \|T\psi - T\zeta\| \leq \|\psi - \zeta\|, \ \forall \psi, \zeta \in \Gamma. \]  \hspace{1cm} (2)

Clearly, every contraction map is a nonexpansive map for $\delta = 1$.

For some decades now, fixed point theory has been developed into a basic and an essential tool for different branches of both applied and pure mathematics. Particularly, it may be seen as an essential subject of nonlinear functional analysis. Moreover, fixed point theory is one of the useful tools to solve many problems in applied sciences and engineering such as the existence of solutions to integral equations, differential equations, matrix equations, dynamical system, models in economy, game theory, fractals, graph theory, optimization theory, approximation theory, computer science and many other subjects.

It is well known that several mathematical problems are naturally formulated as fixed point problem,

\[ T\psi = \psi, \]  \hspace{1cm} (3)
where $T$ is some suitable mapping, which may be nonlinear. For example, given a mapping $\varphi : [d, e] \subset \mathbb{R} \to \mathbb{R}$ and $k : [d, e] \times [d, e] \times \mathbb{R} \to \mathbb{R}$, then the solution to the following nonlinear integral equation:

$$\psi(c) = \varphi(c) + \int_d^e k(c, \rho, \psi(\rho))d\rho, \quad (4)$$

where $\psi \in C([d, e])$, is tantamount to finding the fixed point of the integral operator $T : C([d, e]) \to C([d, e])$ defined by

$$T(\psi)(c) = \varphi(c) + \int_d^e k(c, \rho, \psi(\rho))d\rho, \quad (5)$$

for all $\psi \in C([d, e])$.

A solution $\psi$ of the problem (3) is called a fixed point of the mapping $T$. We will denote the set of all fixed points of $T$ by $F(T)$, i.e., $F(T) = \{ \psi \in \Gamma : T\psi = \psi \}$.

On the other hand, once the existence of some fixed points of a given mapping is guaranteed, then finding such fixed point of the mapping is cumbersome in some cases. To surmount this difficulty, iterative algorithm should be data dependent (see [1]).

The Picard iterative algorithm

$$\psi_{s+1} = T\psi_s, \forall s \in \mathbb{N} \quad (6)$$

is one of the first iterative algorithms which has been widely used for approximating the fixed points of contraction mappings. However, the success of Picard iterative algorithm has not been carried over to the more general classes of operators such as nonexpansive mappings even when a fixed exists.

For example, the mapping $T : [0, 1] \to [0, 1]$ defined by $T\psi = 1 - \psi$ for all $\psi \in [0, 1]$ is a nonexpansive mapping with a unique fixed point $\psi = \frac{1}{2}$. Notice that for $\psi_0 \in [0, 1], \psi_0 \neq \frac{1}{2}$, the Picard iterative algorithm (6) generates the sequence $\{1 - \psi_0, \psi_0, 1 - \psi_0, \ldots\}$ for which fails to converge to the fixed point $z$ of $T$.

To overcome the failure recorded by Picard iterative algorithm, many researchers in nonlinear analysis got busy with constructing several new iterative algorithms for approximating the fixed points of nonexpansive mappings and other mappings more general than the classes of nonexpansive mappings.

Some notable iterative algorithms in the existing literature includes: Mann [2], Ishikawa [3], Noor [4], Argawal et al., [5], Abbas and Nazir [6], SP [7], S* [8], CR [9], Normal-S [10], Picard-S [11], Thakur et al., [12], M [13], M* [14], Garodia and Uddin [15], Two-Step Mann [16] iterative algorithms and so on.

In 2007, the following iterative algorithm which is known as S iteration was introduced by Argawal et al., [5]:

$$\begin{cases} w_0 \in \Gamma, \\
               m_s = (1 - \mu_s)w_s + \mu_sTw_s, \quad s \in \mathbb{N}, \\
               w_{s+1} = (1 - r_s)Tw_s + r_sTm_s, \quad (7) 
\end{cases}$$

where $\{\mu_s\}$ and $\{r_s\}$ are sequences in $[0, 1]$.

In 2014, the following iterative algorithm known as Picard-S iteration was introduced by Gursoy and Karakaya [11]:

$$\begin{cases} w_0 \in \Gamma, \\
               p_s = (1 - \mu_s)w_s + \mu_sTw_s, \\
               m_s = (1 - r_s)Tw_s + r_sTp_s, \quad s \in \mathbb{N}, \\
               w_{s+1} = Tm_s, \quad (8) 
\end{cases}$$

where $\{\mu_s\}$ and $\{r_s\}$ are sequences in $[0, 1]$. The authors showed with the aid of an example that Picard-S iterative algorithm (8) converges at a rate faster than all of Picard, Mann, Ishikawa, Noor, SP, CR, S, S*, Abbas and Nazir, Normal-S and Two-Step Mann iteration processes for contraction mappings.
In 2016, Thakur et al. [12] introduced the following three-steps iterative algorithm:

\[
\begin{aligned}
& w_0 \in \Gamma, \\
& p_s = (1 - \mu_s)w_s + \mu_sTw_s, \quad s \in \mathbb{N}, \\
& m_s = T((1 - r_s)w_s + r_s p_s), \\
& w_{s+1} = Tm_s,
\end{aligned}
\]

where \( \{\mu_s\} \) and \( \{r_s\} \) are sequences in \([0, 1]\). With the help of numerical example, they proved that (9) is faster than Picard, Mann, Ishikawa, Agarwal, Noor and Abbas iterative algorithm for Suzuki generalized nonexpansive mappings.

In 2018, Ullah and Arshad [13] introduced the M iterative algorithm as follows:

\[
\begin{aligned}
& w_0 \in \Gamma, \\
& p_s = (1 - r_s)w_s + r_s Tw_s, \quad s \in \mathbb{N}, \\
& m_s = Tp_s, \\
& w_{s+1} = Tm_s,
\end{aligned}
\]

where \( \{r_s\} \) is a sequence in \([0, 1]\). Numerically they show that M iterative algorithm (10) converges faster than S iterative algorithm (7) and Picard-S iteration process (8) for Suzuki generalized nonexpansive mapping.

Recently, Garodia and Uddin [15] introduced the following three-steps iterative algorithm:

\[
\begin{aligned}
& w_0 \in \Gamma, \\
& p_s = Tw_s, \\
& m_s = T((1 - r_s)p_s + r_s Tw_s), \quad s \in \mathbb{N}, \\
& w_{s+1} = Tm_s,
\end{aligned}
\]

where \( \{r_s\} \) is a sequence in \([0, 1]\). The authors showed both analytically and numerically that their iterative algorithm (11) converges faster than M iterative algorithm (10). Also, they showed that the iterative algorithm (11) converges faster than all of S, Abbas and Nazir, Thakur New, M, Noor, Picard-S, Thakur, M* iterative algorithms for contractive-like mappings and Suzuki generalized nonexpansive mappings.

Clearly, from the performance of the iterative algorithm (11), we note that it is one of the leading iterative schemes.

**Problem 1.** Is it possible to construct a four-steps iterative algorithm which have better rate of converges than the three-step iterative algorithm (11) for contraction mappings?

To solve the above problem, we introduce the following iterative algorithm, called AU iterative algorithm:

\[
\begin{aligned}
& \psi_0 \in \Gamma, \\
& g_s = T((1 - r_s)\psi_s + r_s T\psi_s), \\
& k_s = Tg_s, \\
& \eta_s = Tk_s, \\
& \psi_{s+1} = T\eta_s,
\end{aligned}
\]

where \( \{r_s\} \) is a sequence in \([0, 1]\).

The aim of this paper is to prove the strong convergence of AU iterative algorithm (12) to the fixed points of contraction mappings. We will prove analytically that AU iterative algorithm (12) converges faster than the iterative algorithm (11). With the aid of an example, we show numerically that AU iterative algorithm has a better rate of convergence than the iterative algorithm (11) and several other leading iterative algorithms existing in the literature. Also, we will show that AU iterative algorithm is \( T \)-stable and data dependent.

Additionally, we will use AU iterative algorithm (12) to find the unique solutions of a nonlinear integral equation.
2. Preliminaries

The following definitions and lemmas will be useful in proving our main results.

Definition 2. [17] Let \( \{a_s\} \) and \( \{b_s\} \) be two sequences of real numbers that converge to \( a \) and \( b \) respectively and assume that there exists \( \ell = \lim_{s \to \infty} \frac{\|a_s - a\|}{\|b_s - b\|} \).

Then,

\((R_1)\) if \( \ell = 0 \), we say that \( \{a_s\} \) converges faster to \( a \) than \( \{b_s\} \) does to \( b \).

\((R_2)\) If \( 0 < \ell < \infty \), we say that \( \{a_s\} \) and \( \{b_s\} \) have the same rate of convergence.

Definition 3. [17] Let \( \{\omega_s\} \) and \( \{\mu_s\} \) be two fixed point iteration processes that converge to the same point \( z \), the error estimates

\[ \|\omega_s - z\| \leq a_s, \quad s \in \mathbb{N}, \]

\[ \|\mu_s - z\| \leq b_s, \quad s \in \mathbb{N}, \]

are available where \( \{a_s\} \) and \( \{b_s\} \) are two sequences of positive numbers converging to zero. Then we say that \( \{\omega_s\} \) converges faster to \( z \) than \( \{\mu_s\} \) does if \( \{a_s\} \) converges faster than \( \{b_s\} \).

Definition 4. [17] Let \( T, \tilde{T} : \Gamma \to \Gamma \) be two operators. We say that \( \tilde{T} \) is an approximate operator for \( T \) if for some \( \epsilon > 0 \), we have

\[ \|T\psi - \tilde{T}\psi\| \leq \epsilon, \quad \forall \psi \in \Gamma. \]  \hspace{1cm} (13)

Definition 5. [18] Let \( \{\zeta_s\} \) be any sequence in \( \Gamma \). Then, an iterative algorithm \( \psi_{s+1} = f(T, \psi_s) \), which converges to fixed point \( z \), is said to be \( T \)-stable if for \( \varepsilon_s = ||\zeta_{s+1} - f(T, \zeta_s)||, \forall s \in \mathbb{N} \), we have

\[ \lim_{s \to \infty} \varepsilon_s = 0 \Leftrightarrow \lim_{s \to \infty} \zeta_s = z. \]  \hspace{1cm} (14)

Lemma 6. [19] Let \( \{s\} \) and \( \{\lambda_s\} \) be nonnegative real sequences satisfying the following inequalities:

\[ s_{s+1} \leq (1 - \sigma_s)s_s + \lambda_s, \]  \hspace{1cm} (15)

where \( \sigma_s \in (0, 1) \) for all \( s \in \mathbb{N} \), \( \sum_{s=0}^{\infty} \sigma_s = \infty \) and \( \lim_{s \to \infty} \frac{s_s}{s} = 0 \), then \( \lim s_s = 0 \).

Lemma 7. [20] Let \( \{s\} \) be a nonnegative real sequence and there exits an \( s_0 \in \mathbb{N} \) such that for all \( s \geq s_0 \) satisfying the following condition:

\[ s_{s+1} \leq (1 - \sigma_s)s_s + \sigma_s \lambda_s, \]

where \( \sigma_s \in (0, 1) \) for all \( s \in \mathbb{N} \), \( \sum_{s=0}^{\infty} \sigma_s = \infty \) and \( \lambda_s \geq 0 \) for all \( s \in \mathbb{N} \), then

\[ 0 \leq \limsup_{s \to \infty} s_s \leq \limsup_{s \to \infty} \lambda_s. \]

3. Convergence result

In this section, we prove the strong convergence of AU iterative algorithm (12) for contraction mappings.
Theorem 8. Let $\Gamma$ be a nonempty closed convex subset of a real Banach space $\Psi$ and $T : \Gamma \rightarrow \Gamma$ be a contraction mapping such that $F(T) \neq \emptyset$. Let $\{ \psi_s \}$ be the sequence iteratively generated by (12) with a real sequence $\{ r_s \}$ in $[0, 1]$ satisfying $\sum_{s=0}^{\infty} r_s = \infty$. Then $\{ \psi_s \}$ converges strongly to a unique fixed point of $T$.

Proof. Given $z \in F(T)$, then from (12) we have

$$
\| g_s - z \| = \| T((1 - r_s)\psi_s + r_s T\psi_s) - z \|
\leq \delta \| (1 - r_s)\psi_s + r_s T\psi_s - z \|
\leq \delta (1 - r_s) \| \psi_s - z \| + r_s \| T\psi_s - z \|
\leq \delta (1 - r_s) \| \psi_s - z \| + r_s \| \psi_s - z \|
= \delta (1 - (1 - \delta) r_s) \| \psi_s - z \|. \tag{16}
$$

Now, from (12) and (16) we obtain

$$
\| k_s - z \| = \| Tg_s - z \|
\leq \delta \| g_s - z \|
\leq \delta^2 (1 - (1 - \delta) r_s) \| \psi_s - z \|. \tag{17}
$$

Also, from (12) and (17) we get

$$
\| \eta_s - z \| = \| Tk_s - z \|
\leq \delta \| k_s - z \|
\leq \delta^3 (1 - (1 - \delta) r_s) \| \psi_s - z \|. \tag{18}
$$

Finally, from (12) and (18) we have

$$
\| \psi_{s+1} - z \| = \| T\eta_s - z \|
\leq \delta \| \eta_s - z \|
\leq \delta^4 (1 - (1 - \delta) r_s) \| \psi_s - z \|. \tag{19}
$$

From (19), the following inequality is obtained:

$$
\| \psi_{s+1} - z \| \leq \delta^4 (1 - (1 - \delta) r_s) \| \psi_s - z \|
\| \psi_s - z \| \leq \delta^4 (1 - (1 - \delta) r_{s-1}) \| \psi_{s-1} - z \|
\| \psi_{s-1} - z \| \leq \delta^4 (1 - (1 - \delta) r_{s-2}) \| \psi_{s-2} - z \|
\vdots
\| \psi_1 - z \| \leq \delta^4 (1 - (1 - \delta) r_0) \| \psi_0 - z \|. \tag{20}
$$

From (20), we have

$$
\| \psi_{s+1} - z \| \leq \delta^4 (s+1) \prod_{n=0}^{s} (1 - (1 - \delta) r_n) \| \psi_0 - z \|. \tag{21}
$$

Since for all $s \in \mathbb{N}$, $\{ r_s \} \in [0, 1]$ and $\delta \in (0, 1)$, it follows that $(1 - (1 - \delta) r_s) < 1$. From classical analysis we know that $1 - \psi < \exp^{-\psi}$ for all $\psi \in [0, 1]$. Then from (21), we have

$$
\| \psi_{s+1} - z \| \leq \frac{\delta^4 (s+1) \| \psi_0 - z \|}{(1 - \delta) \sum_{i=0}^{s} r_i \exp}, \tag{22}
$$
If we take the limits of both sides of (22), we get \( \lim_{s \to \infty} \| \psi_s - z \| = 0 \). Hence, \( \{ \psi_s \} \) converges strongly to the fixed point of \( T \) as required. \( \square \)

4. Stability result

**Theorem 9.** Let \( \Gamma \) be a nonempty closed convex subset of a Banach space \( \Psi \) and \( T : \Gamma \to \Gamma \) be a contraction mapping. Let \( \{ \psi_s \} \) be an iterative algorithm defined by (12) with a real sequence \( \{ r_s \} \) in \([0,1]\) satisfying \( \sum_{s=0}^{\infty} r_s = \infty \). Then the iterative algorithm (12) is T-stable.

**Proof.** Let \( \{ \xi_s \} \subset \Psi \) be an arbitrary sequence in \( \Gamma \) and suppose that the sequence iteratively generated by (12) is \( \psi_{s+1} = f(T, \psi_s) \), which converges to a unique point \( z \) and that \( \epsilon_s = \| \xi_{s+1} - f(T, \xi_s) \| \). To prove that (12) is T-stable, we have to show that \( \lim_{s \to \infty} \epsilon_s = 0 \) if \( \lim_{s \to \infty} \xi_s \to z \).

Let \( \lim_{s \to \infty} \epsilon_s = 0 \). Then from (12) and the demonstration above, we have

\[
\begin{align*}
\| \xi_{s+1} - z \| &= \| \xi_{s+1} - f(T, \xi_s) + f(T, \xi_s) - z \| \\
&\leq \| \xi_{s+1} - f(T, \xi_s) \| + \| f(T, \xi_s) - z \| \\
&= \epsilon_s + \| f(T, \xi_s) - z \| \\
&= \epsilon_s + \| T \xi_s - z \| \\
&\leq \epsilon_s + \delta \| \xi_s - z \|. \tag{23}
\end{align*}
\]

Putting (24) into (23) we obtain

\[
\| \xi_{s+1} - z \| = \epsilon_s + \delta^2 \| k_s - z \|. \tag{25}
\]

Substituting (26) into (25), we obtain

\[
\| \xi_{s+1} - z \| = \epsilon_s + \delta^3 \| g_s - z \|. \tag{27}
\]

For all \( s \in \mathbb{N} \), put

\[
\begin{align*}
\theta_s &= \| \xi_s - z \|, \\
\sigma_s &= (1 - \delta) r_s \in (0,1), \\
\lambda_s &= \epsilon_s.
\end{align*}
\]
Since $\lim_{s \to \infty} \varepsilon_s = 0$, this implies that $\frac{\Delta s}{\varepsilon_s} = \frac{\varepsilon_s}{(1-\delta)s} \to 0$ as $s \to \infty$. Apparently, all the conditions of Lemma 6 are fulfilled. Hence, we have $\lim_{s \to \infty} \zeta_s = z$.

Conversely, let $\lim_{s \to \infty} \zeta_s = z$. The we have

$$\varepsilon_s = \|\zeta_{s+1} - f(T, \zeta_s)\| = \|\zeta_{s+1} - z + z - f(T, \zeta_s)\| \leq \|\zeta_{s+1} - z\| + \|f(T, \zeta_s) - z\| \leq \|\zeta_{s+1} - z\| + \delta^4(1 - (1 - \delta)\varepsilon)\|\zeta_s - z\|. \quad (30)$$

From (30), it follows that $\lim_{s \to \infty} \varepsilon_s = 0$. Hence, our new iterative algorithm (12) is stable with respect to $T$. □

5. Rate of convergence

In this section, we show that AU iterative algorithm (12) converges faster than Garodia and Uddin iterative algorithm (11) for contraction mappings.

**Theorem 10.** Let $\Gamma$ be a nonempty closed convex subset of a Banach space and $T : \Gamma \to \Gamma$ be a contraction mapping with fixed point $z$. For any $w_0 = \psi_0 \in \Gamma$, let $\{w_s\}$ and $\{\psi_s\}$ be two sequences iteratively generated by (11) and (12) respectively, with real sequence $\{r_s\} \in [0,1]$ such that $r \leq r_s < 1$, for some $r > 0$ and $s \in \mathbb{N}$. Then $\{\psi_s\}$ converges faster to $z$ than $\{w_s\}$ does.

**Proof.** Recalling the inequality (21), we have

$$\|\psi_{s+1} - z\| \leq \delta^4(s+1) \prod_{n=0}^{s} (1 - (1 - \delta)\varepsilon)\|\psi_0 - z\|. \quad (31)$$

Following our assumption that $r \leq r_s < 1$, for some $r > 0$ and together with $s \in \mathbb{N}$, then from (31) we obtain

$$\|\psi_{s+1} - z\| \leq \delta^4(s+1) \prod_{n=0}^{s} (1 - (1 - \delta)\varepsilon)\|\psi_0 - z\| = \delta^4(s+1) (1 - (1 - \delta)r)^s+1 \|\psi_0 - z\|. \quad (32)$$

Let

$$a_s = \delta^4(s+1) (1 - (1 - \delta)r)^s+1 \|\psi_0 - z\|. \quad (33)$$

Similarly, for any $z \in F(T)$, it follows from (11) that

$$\|p_s - z\| = \|Tw_s - z\| \leq \delta\|w_s - z\|. \quad (34)$$

Using (11) and (34), we obtain

$$\|m_s - z\| = \|T((1 - r_s)p_s + r_sTp_s) - z\| \leq \delta\|p_s - z\| + r_s\|Tp_s - z\| \leq \delta((1 - r_s)\|p_s - z\| + r_s\|Tp_s - z\|) \leq \delta(1 - r_s)\|p_s - z\| + r_s\|p_s - z\| = \delta(1 - (1 - \delta)r_s)\|p_s - z\| \leq \delta^2(1 - (1 - \delta)r_s)\|w_0 - z\|. \quad (35)$$
Again, from (11) and (35) we get
\[ \|w_{k+1} - z\| = \|Tm_k - z\| \leq \delta \|m_k - z\| \leq \delta^3 (1 - (1 - \delta) r_s) \|w_k - z\|. \] (36)

From (36), we obtain the following inequalities:
\[ \begin{align*}
\|w_{k+1} - z\| & \leq \delta^3 (1 - (1 - \delta) r_s) \|w_s - z\| \\
\|w_k - z\| & \leq \delta^3 (1 - (1 - \delta) r_{k-1}) \|w_{k-1} - z\| \\
\|w_{k-1} - z\| & \leq \delta^3 (1 - (1 - \delta) r_{k-2}) \|w_{k-2} - z\| \\
& \vdots \\
\|w_1 - z\| & \leq \delta^3 (1 - (1 - \delta) r_0) \|w_0 - z\|. \end{align*} \] (37)

From (37), we have
\[ \|w_{k+1} - z\| \leq \delta^{3(k+1)} \prod_{n=0}^{k} (1 - (1 - \delta) r_n) \|w_0 - z\|. \] (38)

Then from (38) we obtain that
\[ \begin{align*}
\|w_{k+1} - z\| & \leq \delta^{3(k+1)} \prod_{n=0}^{k} (1 - (1 - \delta) r_n) \|w_0 - z\| \\
& = \delta^{3(k+1)} (1 - (1 - \delta) r)^{k+1} \|w_0 - z\|. \end{align*} \] (39)

Put
\[ b_s = \delta^{3(s+1)} (1 - (1 - \delta) r)^{s+1} \|w_0 - z\|. \] (40)

Now, from (33) and (40) we have that
\[ \theta_s = \frac{d_s}{B_s} = \frac{\delta^{4(s+1)} (1 - (1 - \delta) r)^{s+1} \|\psi_0 - z\|}{\delta^{3(s+1)} (1 - (1 - \delta) r)^{s+1} \|w_0 - z\|} = \delta^{s+1}. \] (41)

Since \( \lim_{s \to \infty} \frac{\theta_s}{\theta_s} = \lim_{s \to \infty} \frac{\theta_{s+2}}{\theta_s} = \delta < 1 \), so from ratio test we know that \( \sum_{s=0}^{\infty} \theta_s < \infty \). Hence, from (41) we have
\[ \lim_{s \to \infty} \frac{\|\psi_{s+1} - z\|}{\|w_{s+1} - z\|} = \lim_{s \to \infty} \frac{d_s}{B_s} = \lim_{s \to \infty} \theta_s = 0. \] (42)

From the above demonstrations, it implies that \( \{\psi_s\} \) converges at a rate faster than \( \{w_s\} \). Hence, our new iterative algorithm (12) converges faster than Garodia and Uddin iterative algorithm (11). To support the analytical proof of Theorem 10 and to illustrate the efficiency of AU iterative algorithm (12), we will consider the following numerical example.

**Example 1.** Let \( \Psi = \mathbb{R} \) and \( \Gamma = [1, 50] \). Let \( T : \Gamma \to \Gamma \) be a mapping defined by \( T\psi = \sqrt{2\psi + 4} \) for all \( \psi \in \Gamma \). Clearly, \( T \) is contraction and \( z = 2 \) is a fixed point of \( T \). Take \( \mu_s = r_s = \frac{1}{2} \), with an initial value of 30.

By using the above example, we will show that AU iterative algorithm (12) converges faster a number of leading iterative algorithms in existing literature.
Table 1. A comparison of the different iterative algorithm.

| Step | S     | MANN  | M     | NOOR  | AU     |
|------|-------|-------|-------|-------|--------|
| 1    | 30.000000000 | 30.000000000 | 30.000000000 | 30.000000000 | 30.000000000 |
| 2    | 3.6809877034  | 17.0000000000 | 2.2052188845 | 16.671526557  | 2.0055900713  |
| 3    | 2.1987214827  | 10.180987703  | 2.0032795388 | 9.769074673   | 2.000025151    |
| 4    | 2.0258396187  | 6.5399580891  | 2.000531287  | 6.1528499922  | 2.000000011    |
| 5    | 0.00034028155 | 4.5576312697  | 0.00000008609| 4.2363659316  | 2.0000000000   |
| 6    | 0.0004488682  | 3.4579473856  | 0.0000000139 | 3.216323854   | 2.0000000000   |
| 7    | 0.0001010310  | 2.8111112127  | 0.0000000002 | 2.75629271    | 2.0000000000   |
| 8    | 0.000001360   | 2.1634532374  | 0.0000000000 | 2.1063366837  | 2.0000000000   |
| 9    | 0.000000179   | 0.951662880   | 0.0000000000 | 0.9508056416  | 2.0000000000   |
| 10   | 0.000000024   | 0.85455593    | 0.0000000000 | 0.8316457914  | 2.0000000000   |
| 11   | 0.000000003   | 0.032325723   | 0.0000000000 | 0.0172673331  | 2.0000000000   |
| 12   | 2.0000000000  | 0.0188493597  | 0.0000000000 | 0.0094223922  | 2.0000000000   |
| 13   | 2.0000000000  | 0.0109929989  | 0.0000000000 | 0.0051417581  | 2.0000000000   |
| 14   | 2.0000000000  | 0.0064117448  | 0.0000000000 | 0.0031645791  | 2.0000000000   |
| 15   | 2.0000000000  | 0.0032795388  | 0.0000000000 | 0.0017267331  | 2.0000000000   |
| 16   | 2.0000000000  | 0.0010310310  | 0.0000000000 | 0.0005141758  | 2.0000000000   |

Table 2. A comparison of the different Iterative methods.

| Step | ISHIKAWA  | GARODIA  | ABBASS  | THAKUR  | AU     |
|------|-----------|----------|---------|---------|--------|
| 1    | 30.000000000 | 30.000000000 | 30.000000000 | 30.000000000 | 30.000000000 |
| 2    | 16.680987703 | 2.0287550076 | 2.8050437150 | 2.2052183845 | 2.0055900713 |
| 3    | 9.7832248288 | 2.0000774455 | 2.0456798729 | 2.0032795388 | 2.000025151   |
| 4    | 6.1683753023 | 2.0000002091 | 2.0027147499 | 2.0000531287 | 2.000000011   |
| 5    | 4.250917634  | 2.0000000006 | 2.001617881  | 2.00008609   | 2.0000000000 |
| 6    | 3.228294683  | 2.0000000000 | 2.000096435  | 2.00000139   | 2.0000000000 |
| 7    | 2.6669712534 | 2.0000000000 | 2.00005748   | 2.0000000002 | 2.0000000000 |
| 8    | 2.3646881630 | 2.0000000000 | 2.00000343   | 2.0000000000 | 2.0000000000 |
| 9    | 2.196958627  | 2.0000000000 | 2.0000000020 | 2.0000000000 | 2.0000000000 |
| 10   | 2.1094405974 | 2.0000000000 | 2.0000000001 | 2.0000000000 | 2.0000000000 |
| 11   | 2.060005172  | 2.0000000000 | 2.0000000000 | 2.0000000000 | 2.0000000000 |
| 12   | 2.0320901715 | 2.0000000000 | 2.0000000000 | 2.0000000000 | 2.0000000000 |
| 13   | 2.018051628  | 2.0000000000 | 2.0000000000 | 2.0000000000 | 2.0000000000 |
| 14   | 2.0099011119 | 2.0000000000 | 2.0000000000 | 2.0000000000 | 2.0000000000 |
| 15   | 2.0054321205 | 2.0000000000 | 2.0000000000 | 2.0000000000 | 2.0000000000 |
| 16   | 2.002980349  | 2.0000000000 | 2.0000000000 | 2.0000000000 | 2.0000000000 |

Figure 1. Graph corresponding to Table 1.
Table 3. A comparison of the different Iterative methods.

| Step | SP     | CR         | AU         |
|------|--------|------------|------------|
| 1    | 30.0000000000 | 30.0000000000 | 30.0000000000 |
| 2    | 6.5399580891  | 2.9645498633 | 2.0058900073 |
| 3    | 2.8381224183  | 2.0693639599 | 2.0000025151 |
| 4    | 2.1634532374  | 2.0053107041 | 2.0000000011 |
| 5    | 2.0323255723  | 2.0004085862 | 2.0000000000 |
| 6    | 2.0064117448  | 2.0000314469 | 2.0000000000 |
| 7    | 2.0012725149  | 2.000024204  | 2.0000000000 |
| 8    | 2.0002528810  | 2.000001863  | 2.0000000000 |
| 9    | 2.0000501359  | 2.00000143   | 2.0000000000 |
| 10   | 2.0000099517  | 2.000000011  | 2.0000000000 |
| 11   | 2.0000019754  | 2.000000001  | 2.0000000000 |
| 12   | 2.0000003921  | 2.000000000  | 2.0000000000 |
| 13   | 2.0000000778  | 2.000000000  | 2.0000000000 |
| 14   | 2.0000000154  | 2.000000000  | 2.0000000000 |
| 15   | 2.0000000031  | 2.000000000  | 2.0000000000 |
| 16   | 2.0000000006  | 2.000000000  | 2.0000000000 |
| 17   | 2.0000000001  | 2.000000000  | 2.0000000000 |
| 18   | 2.0000000000  | 2.000000000  | 2.0000000000 |

Figure 2. Graph corresponding to Table 2.

Figure 3. Graph corresponding to Table 3.

From the above Tables and Figures, it is clear that AU iterative algorithm converges faster than a number of existing iterative algorithms.
6. Data dependence result

In this section, our focus is on the prove of data dependence result for fixed points of contraction mappings by utilizing AU iterative algorithm (12).

**Theorem 11.** Let \( \bar{T} \) be an approximate solution of a contraction mapping \( T \). Let \( \{ \psi_s \} \) be a sequence iteratively generated by AU iterative algorithm (12) and define an iterative algorithm \( \{ \hat{\psi}_s \} \) as follows:

\[
\begin{cases}
\hat{\psi}_0 \in \Gamma, \\
\hat{g}_s = \bar{T}((1 - r_s)\psi_s + r_s \bar{T}\psi_s), \\
\hat{k}_s = T\hat{g}_s, \\
\hat{\eta}_s = \bar{T}\hat{k}_s, \\
\hat{\psi}_{s+1} = \bar{T}\hat{\eta}_s,
\end{cases} \tag{43}
\]

where \( \{ r_s \} \) is a sequence in \([0, 1]\) satisfying the following conditions:

(i) \( \frac{1}{\delta} \leq r_s, \quad s \in \mathbb{N}, \)

(ii) \( \sum_{s=0}^{\infty} r_s = \infty. \)

If \( Tz = z \) and \( \bar{T}z = \hat{z} \) such that \( \lim_{s \to \infty} \hat{\psi}_s = \hat{z} \), we have

\[
\|z - \hat{z}\| \leq \frac{9\epsilon}{1 - \delta}, \tag{44}
\]

where \( \epsilon > 0 \) is a fixed number.

**Proof.** From (12) and (43), we have

\[
\|g_n - \hat{g}_s\| = \|T((1 - r_s)\psi_s + r_s T\psi_s) - \bar{T}((1 - r_s)\psi_s + r_s \bar{T}\psi_s)\| \\
\leq \|T((1 - r_s)\psi_s + r_s T\psi_s) - T((1 - r_s)\psi_s + r_s \bar{T}\psi_s)\| + \|T((1 - r_s)\psi_s + r_s \bar{T}\psi_s) - \bar{T}((1 - r_s)\psi_s + r_s \bar{T}\psi_s)\| \\
\leq \delta \|\psi_s - \psi_s\| + \delta r_s \|T\psi_s - \bar{T}\psi_s\| + \epsilon \\
\leq \delta (1 - r_s) \|\psi_s - \psi_s\| + \delta r_s \|T\psi_s - \bar{T}\psi_s\| + \delta r_s \|\psi_s - \psi_s\| + \delta r_s \epsilon + \epsilon \\
= \delta (1 - (1 - \delta) r_s) \|\psi_s - \psi_s\| + \delta r_s \epsilon + \epsilon. \tag{45}
\]

\[
\|k_s - \hat{k}_s\| = \|Tg_s - \bar{T}\hat{g}_s\| \\
= \|Tg_s - g_s + g_s - \bar{T}\hat{g}_s\| \\
\leq \|Tg_s - g_s\| + \|g_s - \bar{T}\hat{g}_s\| \\
\leq \delta \|g_s - \hat{g}_s\| + \epsilon. \tag{46}
\]

Putting (45) into (46), we obtain

\[
\|k_s - \hat{k}_s\| \leq \delta^2 (1 - (1 - \delta) r_s) \|\psi_s - \psi_s\| + \delta^2 r_s \epsilon + \delta \epsilon + \epsilon. \tag{47}
\]

\[
\|\eta_s - \hat{\eta}_s\| = \|Tk_s - \bar{T}\hat{k}_s\| \\
= \|Tk_s - Tk_s + Tk_s - \bar{T}\hat{k}_s\| \\
\leq \|Tk_s - Tk_s\| + \|Tk_s - \bar{T}\hat{k}_s\| \\
\leq \delta \|k_s - \hat{k}_s\| + \epsilon. \tag{48}
\]

Substituting (47) into (48), we have
\[ \|\eta_s - \bar{\eta}_s\| \leq \delta^3(1 - (1 - \delta)\rho_s)\|\psi_s - \tilde{\psi}_s\| + \delta^3\rho_s\epsilon + \delta^2\epsilon + \delta\epsilon + \epsilon. \]  
(49)

\[
\|\psi_{s+1} - \bar{\eta}_{s+1}\| = \|T\eta_s - T\bar{\eta}_s\| \\
= \|T\eta_s - T\bar{\eta}_s + T\bar{\eta}_s - T\bar{\eta}_s\| \\
\leq \|T\eta_s - T\bar{\eta}_s\| + \|T\bar{\eta}_s - T\bar{\eta}_s\| \\
\leq \delta\|\eta_s - \bar{\eta}_s\| + \epsilon.
\]  
(50)

Putting (49) into (50), we obtain
\[ \|\psi_{s+1} - \tilde{\psi}_{s+1}\| \leq \delta^4(1 - (1 - \delta)\rho_s)\|\psi_s - \tilde{\psi}_s\| + \delta^4\rho_s\epsilon + \delta^3\epsilon + \delta^2\epsilon + \delta\epsilon + \epsilon. \]

Since \(\rho_s \in [0, 1]\) and \(\delta \in [0, 1]\), it implies that
\[
\begin{cases}
(1 - (1 - \delta)\rho_s) < 1, \\
\delta^4, \delta^3, \delta^2 < 1,
\end{cases}
\]  
(51)

and using our assumption (i), we have
\[ 1 - \rho_s \leq \rho_s \Rightarrow 1 = 1 - \rho_s + \rho_s \leq \rho_s + \rho_s = 2\rho_s. \]

From (51), we have
\[
\|\psi_{s+1} - \tilde{\psi}_{s+1}\| \leq (1 - (1 - \delta)\rho_s)\|\psi_s - \tilde{\psi}_s\| + \rho_s\epsilon + 4\epsilon \\
= (1 - (1 - \delta)\rho_s)\|\psi_s - \tilde{\psi}_s\| + \rho_s\epsilon + 4(1 - \rho_s)\epsilon \\
\leq (1 - (1 - \delta)\rho_s)\|\psi_s - \tilde{\psi}_s\| + 9\rho_s\epsilon \\
= (1 - (1 - \delta)\rho_s)\|\psi_s - \tilde{\psi}_s\| + \rho_s(1 - \delta)\frac{9\epsilon}{(1 - \delta)}. \]  
(52)

Let \(\theta_s = \|\psi_s - \tilde{\psi}_s\|,\) \(c_s = (1 - \delta)\rho_s,\) \(\lambda_s = \frac{9\epsilon}{(1 - \delta)},\) then from Lemma 7 and (52), we obtain
\[ 0 \leq \limsup_{s \to \infty} \|\psi_s - \tilde{\psi}_s\| \leq \limsup_{s \to \infty} \frac{9\epsilon}{(1 - \delta)}. \]  
(53)

Recalling Theorem 8, we have \(\lim_{s \to \infty} \psi_s = z\) and from the assumption that \(\lim_{s \to \infty} \tilde{\psi}_s = \bar{z}\) together with (53) we have
\[ \|z - \bar{z}\| \leq \frac{9\epsilon}{(1 - \delta)}. \]  
(54)

Hence, AU iterative algorithm (12) is data dependent. This completes the proof of Theorem 6.1.

7. Application to a Volterra-Fredholm functional integral equation

In this section, we will use AU iterative algorithm (12) to find the solutions of a nonlinear integral equation.

Many problems of mathematical physics, applied sciences and engineering are reduced to Volterra-Fredholm integral equations (see for example, [21, 22] and the references therein).

In 2011, Cracium and Serbian [23] considered and studied the following mixed-type Volterra-Fredholm functional nonlinear integral equation:
\[
\psi(t) = F \left( t, \psi(t), \int_{t_1}^{t_2} K(t, \rho, \psi(\rho))d\rho, \int_{t_1}^{t_2} \cdots \int_{t_1}^{t_n} H(t, \rho, \psi(\rho))d\rho \right),
\]  
(55)
where \([u_1; v_1] \times \cdots \times [u_m; v_m]\) is an interval in \(\mathbb{R}^m\), \(K, H : [u_1; v_1] \times \cdots \times [u_m; v_m] \times [u_1; v_1] \times \cdots \times [u_m; v_m] \times \mathbb{R} \rightarrow \mathbb{R}\) continuous functions and \(F : [u_1; v_1] \times \cdots \times [u_m; v_m] \times \mathbb{R}^3 \rightarrow \mathbb{R}\).

Recently, many authors in nonlinear analysis have constructed some iterative algorithms for approximating the unique solution of the mixed-type Volterra–Fredholm functional nonlinear integral equation (55) in Banach spaces (see for example, [24–26] and the references therein).

In this paper, we will prove the strong convergence of AU iterative algorithm (12) to the unique solution of the problem (55). The following theorem which was given by Cracium and Serbian [23] will be of great importance in proving our main results.

**Theorem 12.** [23] We assume that the following conditions are satisfied:

\((B_1)\) \(K, H \in \Gamma([u_1; v_1] \times \cdots \times [u_m; v_m] \times [u_1; v_1] \times \cdots \times [u_m; v_m] \times \mathbb{R})\);

\((B_2)\) \(F \in ([u_1; v_1] \times \cdots \times [u_m; v_m] \times \mathbb{R}^3)\);

\((B_3)\) there exists nonnegative constants \(\alpha, \beta, \gamma\) such that

\[|F(t, f_1, \xi_1, h_1) - F(t, f_2, \xi_2, h_2)| \leq \alpha|f_1 - f_2| + \beta|\xi_1 - \xi_2| + \gamma|h_1 - h_2|,\]

for all \(t \in [u_1; v_1] \times \cdots \times [u_m; v_m], f_1, \xi_1, h_1, f_2, \xi_2, h_2 \in \mathbb{R}\);

\((B_4)\) there exist nonnegative constants \(L_K\) and \(L_H\) such that

\[|K(t, \rho, f) - K(t, \rho, \xi)| \leq L_K|f - \xi|,\]

\[|H(t, \rho, f) - H(t, \rho, \xi)| \leq L_H|f - \xi|,\]

for all \(t, \rho \in [u_1; v_1] \times \cdots \times [u_m; v_m], f, \xi \in \mathbb{R}\);

\((B_5)\) \(\alpha + (\beta L_K + \gamma L_H)(v_m - v_1) \cdots (v_m - u_m) < 1\). Then, the nonlinear integral equation (55) has a unique solution \(z \in C([u_1; v_1] \times \cdots \times [u_m; v_m]).\)

We are now ready to prove our main result.

**Theorem 13.** Assume that all the conditions \((B_1)-(B_5)\) in Theorem 12 are satisfied. Let \(\{\psi_n\}\) be defined by AU iterative algorithm (12) with real sequence \(r_s \in [0, 1]\), satisfying \(\sum_{s=1}^{\infty} r_s = \infty\). Then (55) has a unique solution and the AU iterative algorithm (12) converges strongly to the unique solution of the mixed type Volterra–Fredholm functional nonlinear integral equation (55), say \(z \in C([u_1; v_1] \times \cdots \times [u_m; v_m]).\)

**Proof.** We now consider the Banach space \(\Psi = C([u_1; v_1] \times \cdots \times [u_m; v_m], \|\cdot\|_C)\), where \(\|\cdot\|_C\) is the Chebyshev’s norm. Let \(\{\psi_n\}\) be the iterative sequence generated by AU iterative algorithm (12) for the operator \(A : \Psi \rightarrow \Psi\) defined by

\[
A(\psi)(t) = F \left( t, \psi(t), \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} K(t, \rho, \psi(\rho))d\rho, \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} H(t, \rho, \psi(\rho))d\rho \right). (56)
\]

Our intention now is to prove that \(\psi_s \rightarrow z\) as \(s \rightarrow \infty\). Now, by using (12), (55), (56) and the assumptions \((B_1)-(B_5)\), we have that

\[
\|\psi_s - z\| = \|A((1 - r_s)\psi_s + r_s A\psi_s) - z\| = |A\left((1 - r_s)\psi_s + r_s A\psi_s\right)(t) - A(z)(t)|
\]

\[
= |F(t, (1 - r_s)\psi_s + r_s A\psi_s)(t) - F(t, z)(t)|
\]

\[
= \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} H(t, \rho, (1 - r_s)\psi_s + r_s A\psi_s(\rho))d\rho d\rho d\rho d\rho d\rho d\rho.
\]

\[
- F(t, z(t), \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} K(t, \rho, z(\rho))d\rho, \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} H(t, \rho, z(\rho))d\rho).
\]
Putting (59) into (58), we obtain

\[
\begin{align*}
\|g\| &\leq \|g\| - z + \|\psi_s - z\| + r_s A \psi_s - z,
\end{align*}
\]

(58)

\[
\begin{align*}
\|g\| &\leq \|g\| - z + \|\psi_s(t) - z(t)\| + r_s|A(\psi_s(t) - A(p))| \\
&= (1 - r_s)|\psi_s(t) - z(t)| \\
&\quad + r_s \left| F(t, \psi_s(t), \int_{u_1}^{u_2} \cdots \int_{u_m}^{|\psi_s|} K(t, \rho, \psi_s(\rho))d\rho, \int_{u_1}^{u_2} \cdots \int_{u_m}^{|\psi_s|} H(t, \rho, \psi_s(\rho))d\rho \right| \\
&\quad + F(t, z(t), \int_{u_1}^{u_2} \cdots \int_{u_m}^{|z|} K(t, \rho, z(\rho))d\rho, \int_{u_1}^{u_2} \cdots \int_{u_m}^{|z|} H(t, \rho, z(\rho))d\rho) \\
&\quad + r_s \left| F(t, z(t), \int_{u_1}^{u_2} \cdots \int_{u_m}^{|z|} K(t, \rho, z(\rho))d\rho, \int_{u_1}^{u_2} \cdots \int_{u_m}^{|z|} H(t, \rho, z(\rho))d\rho) \\
&\quad + r_s \left| F(t, z(t), \int_{u_1}^{u_2} \cdots \int_{u_m}^{|z|} K(t, \rho, z(\rho))d\rho, \int_{u_1}^{u_2} \cdots \int_{u_m}^{|z|} H(t, \rho, z(\rho))d\rho) \\
&\quad \leq \{1 - r_s(1 - [a + (BLK + \gamma LH)] \prod_{i=1}^{m} (v_i - u_i))\} ||\psi_s - z||. \\
\end{align*}
\]

(59)

Putting (59) into (58), we obtain

\[
\begin{align*}
\|g - z\| &\leq [a + (BLK + \gamma LH)] \prod_{i=1}^{m} (v_i - u_i)) \{1 - r_s(1 - [a + (BLK + \gamma LH)] \prod_{i=1}^{m} (v_i - u_i))\} \|\psi_s - z\|. \\
\end{align*}
\]

(60)

From (12) and (60), we have

\[
\begin{align*}
\|k - z\| &= \|A g - z\| = |A(\psi_s(t) - A(z))| \\
&= \left| F(t, \psi_s(t), \int_{u_1}^{u_2} \cdots \int_{u_m}^{|\psi_s|} K(t, \rho, \psi_s(\rho))d\rho, \int_{u_1}^{u_2} \cdots \int_{u_m}^{|\psi_s|} H(t, \rho, \psi_s(\rho))d\rho \right| \\
&\quad - F(t, z(t), \int_{u_1}^{u_2} \cdots \int_{u_m}^{|z|} K(t, \rho, z(\rho))d\rho, \int_{u_1}^{u_2} \cdots \int_{u_m}^{|z|} H(t, \rho, z(\rho))d\rho) \\
&\quad + r_s \left| F(t, z(t), \int_{u_1}^{u_2} \cdots \int_{u_m}^{|z|} K(t, \rho, z(\rho))d\rho, \int_{u_1}^{u_2} \cdots \int_{u_m}^{|z|} H(t, \rho, z(\rho))d\rho) \\
&\quad + r_s \left| F(t, z(t), \int_{u_1}^{u_2} \cdots \int_{u_m}^{|z|} K(t, \rho, z(\rho))d\rho, \int_{u_1}^{u_2} \cdots \int_{u_m}^{|z|} H(t, \rho, z(\rho))d\rho) \\
&\quad \leq \{1 - r_s(1 - [a + (BLK + \gamma LH)] \prod_{i=1}^{m} (v_i - u_i))\} \|\psi_s - z\|. \\
\end{align*}
\]
\[
\begin{align*}
\leq \alpha |g_s(t) - z(t)| + \beta \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} K(t, \rho, g_s(\rho)) d\rho - \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} K(t, \rho, z(\rho)) d\rho \\
+ \gamma \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} K(t, \rho, g_s(\rho)) d\rho - \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} K(t, \rho, z(\rho)) d\rho \\
\leq \alpha |g_s(t) - z(t)| + \beta \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} |K(t, \rho, g_s(\rho)) - K(t, \rho, z(\rho))| d\rho \\
+ \gamma \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} |H(t, \rho, g_s(\rho)) - H(t, \rho, z(\rho))| d\rho \\
\leq \alpha |g_s(t) - z(t)| + \beta \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} L_K |g_s(\rho) - z(\rho)| d\rho + \gamma \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} L_K |z(\rho)| d\rho \\
\leq \alpha \|g_s - z\| + \beta \prod_{i=1}^{m} (v_i - u_i) \|K\| \|g_s - z\| + \gamma \prod_{i=1}^{m} (v_i - u_i) \|z\| \\
= [\alpha + (\beta L_K + \gamma L) \prod_{i=1}^{m} (v_i - u_i)] \|K\| \|g_s - z\| \\
\leq (\alpha + (\beta L_K + \gamma L)) \prod_{i=1}^{m} (v_i - u_i))^2 \{1 - r_s(1 - [\alpha + (\beta L_K + \gamma L) \prod_{i=1}^{m} (v_i - u_i)])\} \|\psi_s - z\|. \quad (61)
\end{align*}
\]

From (12) and (61), we have

\[
\| \eta_s - z \| = \| A k_s - z \| \\
= |A(k_s)(t) - A(z)(t)| \\
= | F(t, k_s(t), \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} K(t, \rho, k_s(\rho)) d\rho, \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} H(t, \rho, k_s(\rho)) d\rho) \\
- F(t, z(t), \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} K(t, \rho, z(\rho)) d\rho, \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} H(t, \rho, z(\rho)) d\rho)| \\
\leq \alpha |k_s(t) - z(t)| + \beta \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} |K(t, \rho, k_s(\rho)) - K(t, \rho, z(\rho))| d\rho \\
+ \gamma \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} |H(t, \rho, k_s(\rho)) - H(t, \rho, z(\rho))| d\rho \\
\leq \alpha |k_s(t) - z(t)| + \beta \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} |K(t, \rho, k_s(\rho)) - K(t, \rho, z(\rho))| d\rho \\
+ \gamma \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} |H(t, \rho, k_s(\rho)) - H(t, \rho, z(\rho))| d\rho \\
\leq \alpha |k_s(t) - z(t)| + \beta \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} L_K |k_s(\rho) - z(\rho)| d\rho + \gamma \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} L_K |z(\rho)| d\rho \\
\leq \alpha \|k_s - z\| + \beta \prod_{i=1}^{m} (v_i - u_i) \|K\| \|k_s - z\| + \gamma \prod_{i=1}^{m} (v_i - u_i) \|z\| \\
= [\alpha + (\beta L_K + \gamma L) \prod_{i=1}^{m} (v_i - u_i)] \|k_s - z\| \\
\leq (\alpha + (\beta L_K + \gamma L) \prod_{i=1}^{m} (v_i - u_i))^3 \{1 - r_s(1 - [\alpha + (\beta L_K + \gamma L) \prod_{i=1}^{m} (v_i - u_i)])\} \|\psi_s - z\|. \quad (62)
\]

Again, from (12) and (62), we obtain

\[
\| \eta_s - z \| = \| A k_s - z \| = |A(k_s)(t) - A(z)(t)| \\
= | F(t, k_s(t), \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} K(t, \rho, k_s(\rho)) d\rho, \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} H(t, \rho, k_s(\rho)) d\rho) \\
- F(t, z(t), \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} K(t, \rho, z(\rho)) d\rho, \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} H(t, \rho, z(\rho)) d\rho)| \\
\leq \alpha |k_s(t) - z(t)| + \beta \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} |K(t, \rho, k_s(\rho)) - K(t, \rho, z(\rho))| d\rho \\
+ \gamma \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} |H(t, \rho, k_s(\rho)) - H(t, \rho, z(\rho))| d\rho \\
\leq \alpha |k_s(t) - z(t)| + \beta \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} L_K |k_s(\rho) - z(\rho)| d\rho + \gamma \int_{u_1}^{q_1} \cdots \int_{u_m}^{q_m} L_K |z(\rho)| d\rho \\
\leq \alpha \|k_s - z\| + \beta \prod_{i=1}^{m} (v_i - u_i) \|K\| \|k_s - z\| + \gamma \prod_{i=1}^{m} (v_i - u_i) \|z\| \\
= [\alpha + (\beta L_K + \gamma L) \prod_{i=1}^{m} (v_i - u_i)] \|k_s - z\| \\
\leq (\alpha + (\beta L_K + \gamma L) \prod_{i=1}^{m} (v_i - u_i))^3 \{1 - r_s(1 - [\alpha + (\beta L_K + \gamma L) \prod_{i=1}^{m} (v_i - u_i)])\} \|\psi_s - z\|. \quad (63)
\]
From (65), we have the following inequalities:

\[
\| \psi_{s+1} - z \| = \| A \eta_z - z \|
\]

\[
= |A(\eta_z) (t) - A(z) (t)|
\]

\[
= \left| F \left( t, \eta_z (t), \int_{u_1}^{q_1} \int_{u_m}^{q_m} K(t, \rho, \eta_z (\rho)) d\rho, \int_{u_1}^{q_1} \int_{u_m}^{q_m} H(t, \rho, \eta_z (\rho)) d\rho \right) - F \left( t, z (t), \int_{u_1}^{q_1} \int_{u_m}^{q_m} K(t, \rho, z (\rho)) d\rho, \int_{u_1}^{q_1} \int_{u_m}^{q_m} H(t, \rho, z (\rho)) d\rho \right) \right|
\]

\[
\leq a | \eta_z (t) - z (t) | + \beta \int_{u_1}^{q_1} \int_{u_m}^{q_m} |K(t, \rho, \eta_z (\rho))| d\rho - \int_{u_1}^{q_1} \int_{u_m}^{q_m} |K(t, \rho, z (\rho))| d\rho
\]

\[
+ \gamma \int_{u_1}^{q_1} \int_{u_m}^{q_m} |H(t, \rho, \eta_z (\rho)) - H(t, \rho, z (\rho))| d\rho
\]

\[
\leq a | \eta_z (t) - z (t) | + \beta \int_{u_1}^{q_1} \int_{u_m}^{q_m} |K(t, \rho, \eta_z (\rho))| d\rho - \int_{u_1}^{q_1} \int_{u_m}^{q_m} |K(t, \rho, z (\rho))| d\rho
\]

\[
+ \gamma \int_{u_1}^{q_1} \int_{u_m}^{q_m} |H(t, \rho, \eta_z (\rho)) - H(t, \rho, z (\rho))| d\rho
\]

Finally, from (12) and (63), we obtain

\[
\left( [a + (\beta L_K + \gamma L_H) \prod_{i=1}^{m} (v_i - u_i)] \right)^4 \left\{ 1 - r_s (1 - [a + (\beta L_K + \gamma L_H) \prod_{i=1}^{m} (v_i - u_i)] \right\} \| \psi_s - z \|. \tag{63}
\]

Since from condition (B5) we have \([a + (\beta L_K + \gamma L_H) \prod_{i=1}^{m} (v_i - u_i)] < 1\), it follows that \(( [a + (\beta L_K + \gamma L_H) \prod_{i=1}^{m} (v_i - u_i)] \)^4 < 1. Thus, (64) reduces to

\[
\| \psi_{s+1} - z \| \leq \left\{ 1 - r_s (1 - [a + (\beta L_K + \gamma L_H) \prod_{i=1}^{m} (v_i - u_i)] \right\} \| \psi_s - z \|. \tag{65}
\]

From (65), we have the following inequalities:

\[
\| \psi_{s+1} - z \| \leq \left\{ 1 - r_s (1 - [a + (\beta L_K + \gamma L_H) \prod_{i=1}^{m} (v_i - u_i)] \right\} \| \psi_s - z \|
\]

\[
\| \psi - z \| \leq \left\{ 1 - r_{s-1} (1 - [a + (\beta L_K + \gamma L_H) \prod_{i=1}^{m} (v_i - u_i)] \right\} \| \psi_{s-1} - z \|
\]
\begin{align}
\|\Psi_{n+1} - p\| & \leq \|\Psi_0 - p\| \prod_{n=0}^{s} \left\{ 1 - r_n \left( 1 - [\alpha + (\beta L_K + \gamma L_H)] m \prod_{i=1}^{m} (v_i - u_i) \right) \right\}.
\end{align}

From (66), we have
\begin{align}
\|\Psi_{n+1} - p\| & \leq \|\Psi_0 - p\| \prod_{n=0}^{s} \left\{ 1 - r_n \left( 1 - [\alpha + (\beta L_K + \gamma L_H)] m \prod_{i=1}^{m} (v_i - u_i) \right) \right\}.
\end{align}

Since \(r_n \in [0, 1]\) for all \(n \in \mathbb{N}\) and recalling from assumption \((B_5)\) that \([\alpha + (\beta L_K + \gamma L_H)] m \prod_{i=1}^{m} (v_i - u_i) < 1\), then we have
\begin{align}
1 - r_n (1 - [\alpha + (\beta L_K + \gamma L_H)] m \prod_{i=1}^{m} (v_i - u_i) ) < 1.
\end{align}

From classical analysis, it is that \(1 - \psi \leq e^{-\psi}\) for all \(\psi \in [0, 1]\), thus from (67), we have
\begin{align}
\|\Psi_{n+1} - z\| & \leq \|z_0 - z\| e^{-1 - [\alpha + (\beta L_K + \gamma L_H)] m \prod_{i=1}^{m} (v_i - u_i) )} \cdot \sum_{n=0}^{s} r_n.
\end{align}

Taking the limit of both sides of the above inequalities, we have \(\lim_{s \to \infty} \|\Psi_s - z\| = 0\). Hence, (12) converges strongly to the unique solution of the mixed type Volterra-Fredholm functional nonlinear integral equation (55). \(\square\)

8. Conclusion

In this paper, we have proved that our new iterative algorithm (12) outperforms several well known iterative algorithms in the literature in terms of rate of convergence. The stability result of AU iterative algorithm has also been obtained. We have also shown that AU iterative algorithm (12) is data dependent. Finally, to illustrate the efficiency of AU iterative algorithm (12), we have proved approximated the unique solution of a nonlinear integral equation. Hence, our results are generalization and improvements of several well known results in the existing literature.

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