

# Empirical study of a Padé type accelerating method of Picard iteration

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## ABSTRACT.

We use a Padé type acceleration technique for the method of successive approximations in [J. Biazar and A. Amirteimoori, *An improvement to the fixed point iterative method*, Applied Mathematics and Computation **182** (2006), 567-571, doi:10.1016/j.amc.2006.04.019] to empirically study the possibility of accelerating Picard iteration for some other known test functions.

## 1. INTRODUCTION

Recently, Biazar and Amirteimoori considered in [9] a Padé-type technique to accelerate Picard iteration method for solving three scalar equations of the form

$$f(x) = 0 \tag{1.1}$$

which were equivalently written as a fixed point problem

$$x = g(x), \tag{1.2}$$

where  $g : [a, b] \rightarrow [a, b]$  is the iteration function.

Under appropriate assumptions on  $f$  (and therefore on  $g$ ), the Picard iteration (or the sequence of successive approximations, as it is generally known), i.e.,

$$x_{n+1} = g(x_n), \quad n \geq 0, \tag{1.3}$$

converges to the (unique) fixed point of  $g$ , say  $\alpha$ , which is the (unique) solution of (1.1) in the interval  $[a, b]$ .

Note that for a certain nonlinear equation (1.1), the fixed point problem (1.2) is not uniquely defined. For example, the equation  $x^3 + 4x^2 - 10 = 0$  can be written under a fixed point form as  $x = \frac{1}{2}\sqrt{10 - x^3}$  or  $x = \sqrt[3]{10 - 4x^2}$ .

As the convergence order of the Picard iteration (1.3) is generally linear (see for example Berinde [6]), the method converges rather slowly to the fixed point  $\alpha$ .

In order to improve the convergence speed of (1.3), the authors in [9] considered the following equivalent fixed point problem

$$x = g_\lambda(x) \tag{1.4}$$

with  $g_\lambda$  of the form

$$g_\lambda(x) = \frac{g(x) + \lambda_1 x + \lambda_2 x^2 + \lambda_3 x^3 + \dots + \lambda_k x^k}{1 + \lambda_1 + \lambda_2 x + \lambda_3 x^2 + \dots + \lambda_k x^{k-1}}, \tag{1.5}$$

where  $k \in \mathbb{N}$ ,  $k \geq 2$  and  $\lambda_1, \lambda_2, \lambda_3, \dots, \lambda_k \in \mathbb{R}$  are parameters that should be determined in such a way that the new iteration function  $g_\lambda$  will yield a faster Picard iteration.

Note that the method of constructing (1.5) is rather similar to the way in which the Padé approximant of order  $(m, n)$ ,  $[m/n]_f(x)$ , is obtained, see for example [5]:

$$[m/n]_f(x) = \frac{p_0 + p_1 x + p_2 x^2 + \dots + p_m x^m}{1 + q_1 x + q_2 x^2 + \dots + q_n x^n}. \tag{1.6}$$

This is the reason we shall name in the following (1.5) as a Padé type transform.

The aim of this paper is twofold: first, to derive the convergence order of the Picard iteration associated to (1.4) and secondly, to perform a similar empirical study of the rate of convergence for other values of  $k$ , in the case of the equations from [9], as well as for other test functions taken from literature. This will allow us to infer which value of  $k$  is optimal for each equation.

## 2. THE PADÉ-TYPE ACCELERATION OF THE PICARD ITERATION

This result is taken from [9].

Based on the fact that the fixed point equation

$$x = g(x)$$

is equivalent to

$$x + \lambda_1 x + \lambda_2 x^2 + \lambda_3 x^3 + \dots + \lambda_k x^k = g(x) + \lambda_1 x + \lambda_2 x^2 + \lambda_3 x^3 + \dots + \lambda_k x^k,$$

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which can be written under the form

$$x = g_\lambda(x) = \frac{g(x) + \lambda_1 x + \lambda_2 x^2 + \lambda_3 x^3 + \dots + \lambda_k x^k}{1 + \lambda_1 + \lambda_2 x + \lambda_3 x^2 + \dots + \lambda_k x^{k-1}} \tag{2.7}$$

we get exactly the fixed point problem (1.4).

It is tacitly assumed that  $g_\lambda(x)$  is well defined on the interval  $[a, b]$  where the original equation is solved, that is, the equation

$$1 + \lambda_1 + \lambda_2 x + \lambda_3 x^2 + \dots + \lambda_k x^{k-1} = 0$$

has no real roots on  $[a, b]$ .

The main idea of constructing such an accelerated method is to determine the parameters  $\lambda_1, \lambda_2, \dots, \lambda_k$  such that the new iteration function  $g_\lambda$  satisfies

$$g_\lambda^{(i)}(\alpha) = 0, \quad i = 1, 2, \dots, k, \tag{2.8}$$

where  $\alpha$  is the unique solution of (1.1) and (1.2) in the interval  $[a, b]$ .

Using (2.7), the equation (2.8) yields an upper diagonal linear system of equations with the unknowns  $\lambda_1, \lambda_2, \dots, \lambda_k$  which always is uniquely solvable, as in the case of the original Padé transform. Indeed, by (2.7) we have

$$g_\lambda(x)(1 + \lambda_1 + \lambda_2 x + \lambda_3 x^2 + \dots + \lambda_k x^{k-1}) = g(x) + \lambda_1 x + \lambda_2 x^2 + \lambda_3 x^3 + \dots + \lambda_k x^k$$

which, by differentiating with respect to  $x$ , gives

$$\begin{aligned} g'_\lambda(x)(1 + \lambda_1 + \lambda_2 x + \dots + \lambda_k x^{k-1}) + g_\lambda(x)(\lambda_2 + 2\lambda_3 x + \dots + (k-1)\lambda_k x^{k-2}) = \\ = g'(x) + \lambda_1 + 2\lambda_2 x + \dots + k\lambda_k x^{k-1}. \end{aligned} \tag{2.9}$$

If we take  $x = \alpha$  in (2.9) and use the fact that  $g_\lambda(\alpha) = g_\lambda(x) = \alpha$  and  $g'_\lambda(\alpha)$  is required to be zero, we get the linear equation

$$\lambda_1 + 2\lambda_2 \alpha + \dots + k\lambda_k \alpha^{k-1} = -g'(\alpha).$$

Now we differentiate again (2.9) and then, by letting  $x = \alpha$ , we are lead to the linear equation

$$2\lambda_2 + 3\lambda_3 \alpha + \dots + k(k-1)\lambda_k \alpha^{k-1} = -g''(\alpha)$$

and so on. The generic formula for the  $i^{th}$  derivative of  $g_\lambda$  is

$$-g^{(j)}(\alpha) = \sum_{i=j}^k i(i-1)(i-2)\dots(i-j+1)\lambda_i \alpha^{i-j}, \quad j = 1, 2, \dots, k. \tag{2.10}$$

If we rewrite the linear  $k \times k$  system (2.10) in a matrix form we have

$$\begin{pmatrix} 1 & 2\alpha & 3\alpha^2 & \dots & k\alpha^{k-1} \\ 0 & 2 & 6\alpha & \dots & k(k-1)\alpha^{k-2} \\ 0 & 0 & 6 & \dots & k(k-1)(k-2)\alpha^{k-3} \\ \vdots & \vdots & \vdots & \dots & \vdots \\ 0 & 0 & 0 & \dots & k! \end{pmatrix} \begin{pmatrix} \lambda_1 \\ \lambda_2 \\ \lambda_3 \\ \vdots \\ \lambda_k \end{pmatrix} = \begin{pmatrix} -g'(\alpha) \\ -g^{(2)}(\alpha) \\ -g^{(3)}(\alpha) \\ \vdots \\ -g^{(k)}(\alpha) \end{pmatrix}. \tag{2.11}$$

By solving (2.11), we can uniquely find the values of  $\lambda_1, \lambda_2, \lambda_3, \dots, \lambda_k$  and hence get the iteration function of the accelerated process

$$x_{n+1} = g_\lambda(x_n), \quad n \geq 0.$$

We end this section by reminding the concept of convergence order that will be used in the paper.

Let  $\{x_n\} \subset \mathbb{R}$  be a sequence of real numbers convergent to  $\alpha \in \mathbb{R}$  (which is obtained by iterating a fixed point equation)

**Definition 2.1.** [13] Let  $\{x_n\}$  converge to  $\alpha$ . If there exist an integer constant  $p$ , and a real positive constant  $C$  such that

$$\lim_{n \rightarrow \infty} \left| \frac{x_{n+1} - \alpha}{(x_n - \alpha)^p} \right| = C,$$

then  $p$  is called the order and  $C$  the constant of convergence.

The concept of rate of convergence given by Definition 2.1 is also known as the  $Q$ -order of convergence, see the monographs by Mărușter [13] and Ortega and Rheinboldt [14].

The next theorem shows how the fixed point iteration defined by the function  $g_\lambda$  accelerates the fixed point iteration defined by  $g$ .

**Theorem 2.1.** Let  $g \in C^{k+1}[a, b]$  such that the associated iteration function  $g_\lambda$  satisfy (2.8), where  $\alpha$  is the unique solution in  $[a, b]$  of (1.2). Then the accelerated Picard iteration

$$x_{n+1}^\lambda = g(x_n^\lambda), \quad n \geq 0$$

has  $Q$ -order of convergence  $k$ .

*Proof.* By the Taylor expansion of  $g_\lambda$  at  $x$  we find

$$g_\lambda(x_n) = g_\lambda(x) + \frac{g'_\lambda(x)}{1!}(x_n - x) + \dots + \frac{g_\lambda^{(k)}(x)}{k!}(x_n - x)^k + \dots$$

which yields, in view of  $g_\lambda(\alpha) = \alpha$  and (2.8)

$$g_\lambda(x_n) - \alpha = \frac{g_\lambda^{(k+1)}(x)}{(k+1)!}(x_n - \alpha)^{k+1} + \dots$$

that is

$$\lim_{n \rightarrow \infty} \frac{|x_{n+1} - \alpha|}{|x_n - \alpha|^{k+1}} = \frac{g_\lambda^{(k+1)}(\alpha)}{(k+1)!},$$

which completes the proof.  $\square$

**Remark 2.1.** Note that, generally,  $g'(\alpha) \neq 0$ , so  $(x_n)$  has the  $Q$ -order of convergence equal to 1, see the Examples in the next section.

### 3. SOME USEFUL FIXED POINT THEOREMS

In this section we present three known results in fixed point theory, taken from [6], that ensure, under various assumptions, the existence and uniqueness of a fixed point of a mapping  $g$  as well as the convergence of the Picard iteration to that fixed point. For two of them, the rate of convergence is also given.

**Theorem 3.2 (Contraction Mapping Principle).** Let  $(X, d)$  be a complete metric space and  $T : X \rightarrow X$  a map satisfying

$$d(Tx, Ty) \leq a d(x, y), \quad \text{for all } x, y \in X, \quad (3.12)$$

where  $0 \leq a < 1$  is constant. Then:

- (p1)  $T$  has a unique fixed point  $x^*$  in  $X$ ;
- (p2) The Picard iteration  $\{x_n\}_{n=0}^\infty$  defined by

$$x_{n+1} = Tx_n, \quad n = 0, 1, 2, \dots \quad (3.13)$$

converges to  $x^*$ , for any  $x_0 \in X$ .

- (p3) The following estimate holds:

$$d(x_{n+i-1}, x^*) \leq \frac{a^i}{1-a} d(x_n, x_{n-1}), \quad n = 0, 1, 2, \dots; i = 1, 2, \dots \quad (3.14)$$

- (p4) The rate of convergence of Picard iteration is given by

$$d(x_n, x^*) \leq a d(x_{n-1}, x^*), \quad n = 1, 2, \dots \quad (3.15)$$

**Theorem 3.3 (Zamfirescu's Mapping Principle).** Let  $(X, d)$  be a complete metric space and let  $T : X \rightarrow X$  be a mapping for which there exist  $a \in [0, 1)$ ,  $b, c \in [0, \frac{1}{2})$  such that for all  $x, y \in X$ , at least one of the following conditions is true:

- (z<sub>1</sub>)  $d(Tx, Ty) \leq a d(x, y)$ ;
- (z<sub>2</sub>)  $d(Tx, Ty) \leq b[d(x, Tx) + d(y, Ty)]$ ;
- (z<sub>3</sub>)  $d(Tx, Ty) \leq c[d(x, Ty) + d(y, Tx)]$ .

Then the Picard iteration  $\{x_n\}$  defined by (3.13) and starting from  $x_0 \in X$  converges to the unique fixed point  $x^*$  of  $T$  with the following error estimate

$$d(x_{n+i-1}, x^*) \leq \frac{\delta^i}{1-\delta} d(x_n, x_{n-1}), \quad n = 0, 1, 2, \dots; i = 1, 2, \dots$$

where  $\delta = \max \left\{ a, \frac{b}{1-b}, \frac{c}{1-c} \right\}$ .

Moreover, the convergence rate of the Picard iteration is given by

$$d(x_n, x^*) \leq \delta \cdot d(x_{n-1}, x^*), \quad n = 1, 2, \dots \quad (3.16)$$

**Theorem 3.4 (Almost Contraction Mapping Principle).** Let  $(X, d)$  be a complete metric space and  $T : X \rightarrow X$  an almost contraction, that is, a mapping for which there exist a constant  $\delta \in (0, 1)$  and some  $L \geq 0$  such that

$$d(Tx, Ty) \leq \delta \cdot d(x, y) + Ld(y, Tx), \quad \text{for all } x, y \in X. \quad (3.17)$$

Then

- 1)  $F(T) = \{x \in X : Tx = x\} \neq \emptyset$ ;
- 2) For any  $x_0 \in X$ , the Picard iteration  $\{x_n\}_{n=0}^\infty$  given by (1.2) converges to some  $x^* \in F(T)$ ;
- 3) The following estimate holds

$$d(x_{n+i-1}, x^*) \leq \frac{\delta^i}{1-\delta} d(x_n, x_{n-1}), \quad n = 0, 1, 2, \dots; i = 1, 2, \dots \quad (3.18)$$

4. NUMERICAL EXAMPLES

**Example 4.1.** [9]Test function:  $f(x) = x^3 + 4x^2 - 10$ , which has a unique root in the interval (1,2). We use an approximate value for  $\alpha$ ,  $\alpha \cong 1.5$  and  $g(x) = \frac{1}{2}\sqrt{10 - x^3}$ . The values of the parameters  $\lambda_i$  involved in (2.7) are For  $k = 2$ :

$$\lambda_1 = -1.15660903, \lambda_2 = 1.20815133.$$

For  $k = 3$ :

$$\lambda_1 = 1.57623135, \lambda_2 = -2.43563586, \lambda_3 = 1.21459573.$$

For  $k = 4$ :

$$\lambda_1 = -3.122090855, \lambda_2 = 6.961008590, \lambda_3 = -5.049833910, \lambda_4 = 1.392095477.$$

and for  $k = 5$ :

$$\lambda_1 = 6.176012965, \lambda_2 = -17.83393493, \lambda_3 = 19.74510962, \lambda_4 = -9.627879427,$$

$\lambda_5 = 1.836662484$ . The results for the three fastest methods used in Example 4.1 are listed in Table 1.

Table 1

n	k = 2	k = 3	k = 4	k = 5	$x_{n+1} = g(x_n)$
	$x_{n+1} = g_\lambda(x_n)$	$x_{n+1} = g_\lambda(x_n)$	$x_{n+1} = g_\lambda(x_n)$	$x_{n+1} = g_\lambda(x_n)$	
0	1.5	1.5	1.5	1.5	1.5
1	1.37131921	1.37131920	1.37131921	1.37131923	1.28695377
2	1.36517040	1.36525174	1.36523987	1.36524189	1.40254080
3	1.36523078	1.36523005	1.36523000	1.36523001	1.34545838
4	1.36523000	1.36523001	1.36523001		1.37517025
5	1.36523001				1.36009419
6					1.36784697
⋮					⋮
25					1.36523001

For Example 4.1 we observe that for  $k = 5$  we have the best rate of convergence.

**Example 4.2.** [9]Test function  $f(x) = x - \tan x = 0$ . This equation has a root which lies near  $\frac{3\pi}{2}$ . Let  $g(x) = \tan x$ , then  $g'(x) = 1 + \tan^2 x \geq 1$ , which is not a suitable  $g(x)$ . Let  $\alpha \cong 4.5$  and  $g(x) = \tan x$ . We show that the new technique works even in this case. The values of the parameters  $\lambda_i$  involved in (2.7) are

For  $k = 3$ :

$$\lambda_1 = -28939.740120, \lambda_2 = 13060.829480, \lambda_3 = -1474.394932.$$

For  $k = 4$ :

$$\lambda_1 = 814467.2540, \lambda_2 = -54910.4993, \lambda_3 = 123474.7892, \lambda_4 = -9255.495122.$$

For  $k = 5$ :

$$\lambda_1 = -2.152270898 \cdot 10^7, \lambda_2 = 1.930605732 \cdot 10^7, \lambda_3 = -6.494947820 \cdot 10^6,$$

$$\lambda_4 = 9.712515583 \cdot 10^5, \lambda_5 = -54472.61408.$$

and for  $k = 6$ :

$$\lambda_1 = 5.464009450 \cdot 10^8, \lambda_2 = -6.117202249 \cdot 10^8, \lambda_3 = 2.739611774 \cdot 10^8,$$

$$\lambda_4 = -6.135233180 \cdot 10^7, \lambda_5 = 6.870369979 \cdot 10^6, \lambda_6 = -3.077707818 \cdot 10^5.$$

The results for the three fastest methods used in Example 4.2 are listed in Table 2.

Table 2

n	k = 3	k = 4	k = 5	k = 6	$x_{n+1} = g(x_n)$
	$x_{n+1} = g_\lambda(x_n)$	$x_{n+1} = g_\lambda(x_n)$	$x_{n+1} = g_\lambda(x_n)$	$x_{n+1} = g_\lambda(x_n)$	
0	4.5	4.5	4.5	4.5	4.5
1	4.493616	4.493280632	4.488372093	5.444444444	4.637332
2	4.493410	4.493716711	4.487779511	5.444305527	13.298192
3		4.493170168	4.479895561	5.444414683	0.898203
4		4.493888939	4.487352445	5.444477321	1.255520
5		4.493311705	4.495007882	5.444416274	
⋮		⋮	⋮	⋮	⋮
10					-0.076296
⋮		⋮	⋮	⋮	⋮
25		4.493647770	4.504339881		

For Example 4.2 we observe that for  $k = 3$  we have the best rate of convergence.

**Example 4.3.** [9]

Test function  $f(x) = x - 3^{-x} = 0$ .  $f(x)$  is continuous on  $[\frac{1}{3}, 1]$  and  $f(\frac{1}{3}) \cdot f(1) < 0$ . By Weierstrass' theorem,  $\alpha$ , the root of  $f(x)$ , lies in  $(\frac{1}{3}, 1)$ . Let  $\alpha \cong 0.6 \in (\frac{1}{3}, 1)$  and  $g(x) = 3^{-x}$ . The values of the parameters  $\lambda_i$  involved in (2.7) are

For  $k = 5$ :  
 $\lambda_1 = 1.0979516, \lambda_2 = -1.2013119, \lambda_3 = 0.6435174, \lambda_4 = -0.2083743, \lambda_5 = 0.0344936$ .

For  $k = 6$ :

$\lambda_1 = 1.0985408, \lambda_2 = -1.2062230, \lambda_3 = 0.6598879, \lambda_4 = -0.2356586, \lambda_5 = 0.0572306,$   
 $\lambda_6 = -0.00755790$ .

For  $k = 7$ :

$\lambda_1 = 1.0986056, \lambda_2 = -1.2068705, \lambda_3 = 0.6625857, \lambda_4 = -0.2416536, \lambda_5 = 0.0647243,$   
 $\lambda_6 = -0.0125748, \lambda_7 = 0.0013877$ .

and for  $k = 8$ :

$\lambda_1 = 1.0986117, \lambda_2 = -1.2069416, \lambda_3 = 0.6629413, \lambda_4 = -0.2426415, \lambda_5 = 0.0663709,$   
 $\lambda_6 = -0.0142213, \lambda_7 = 0.0023024, \lambda_8 = 0.0002177$ .

The results for the three fastest methods used in Example 4.3 are listed in Table 3.

Table 3

n	k = 5		k = 6		k = 7		k = 8	
	$x_{n+1} = g\lambda(x_n)$	$x_{n+1} = g\lambda(x_n)$	$x_{n+1} = g\lambda(x_n)$	$x_{n+1} = g\lambda(x_n)$	$x_{n+1} = g\lambda(x_n)$	$x_{n+1} = g\lambda(x_n)$	$x_{n+1} = g\lambda(x_n)$	$x_{n+1} = g(x_n)$
0	0.33333	0.33333	0.33333	0.33333	0.33333	0.33333	0.33333	0.33333
1	0.53769	0.5376929282	0.5376928610	0.5376928610	0.5376928590	0.5376928590	0.5376928590	0.69336
2	0.54779	0.5477874354	0.5477874346	0.5477874346	0.5477874349	0.5477874349	0.5477874349	0.46686
3	0.54781	0.5478086216	0.5478086216	0.5478086216	0.5478086223	0.5478086223	0.5478086223	0.59876
4	0.54781	0.5478086219	0.5478086213	0.5478086213	0.5478086215	0.5478086215	0.5478086215	0.51799
5		0.5478086217	0.5478086215	0.5478086215	0.5478086219	0.5478086219	0.5478086219	
6			0.5478086213	0.5478086213	0.5478086214	0.5478086214	0.5478086214	
7			0.5478086215	0.5478086215	0.5478086216	0.5478086216	0.5478086216	
8					0.5478086218	0.5478086218	0.5478086218	
9					0.5478086218	0.5478086218	0.5478086218	
⋮		⋮	⋮	⋮	⋮	⋮	⋮	⋮
21								0.54781

For the Example 4.3 we observe that for  $k = 5$  we have the best rate of convergence.

**Example 4.4.** [11]Test function:  $f(x) = (x - 1)^3 - 1 = 0$ . We observe that  $x = 2$  is a root of  $f(x)$ . We use an approximative value for  $\alpha, \alpha \cong 1.7$  and  $g(x) = \sqrt[3]{3x^2 - 3x + 2}$ . Note that  $g$  is a contraction on  $\mathbb{R}$ . The values of the parameters  $\lambda_i$  involved in (2.7) are

For  $k = 2$ :

$\lambda_1 = -0.8007005397, \lambda_2 = 0.0217115888$ .

For  $k = 3$ :

$\lambda_1 = -0.4719973830, \lambda_2 = -0.3649980073, \lambda_3 = 0.1137381165$ .

For  $k = 4$ :

$\lambda_1 = 0.2083667801, \lambda_2 = -1.565640648, \lambda_3 = 0.8199984934, \lambda_4 = -0.1384824268$ .

For  $k = 5$ :

$\lambda_1 = 1.180921712, \lambda_2 = -3.854005194, \lambda_3 = 2.839143681, \lambda_4 = -0.9303040692,$   
 $\lambda_5 = 0.1164443592$ .

and for  $k = 6$ :

$\lambda_1 = 2.293100612, \lambda_2 = -7.125119605, \lambda_3 = 6.687513576, \lambda_4 = -3.194051066,$   
 $\lambda_5 = 0.7822522995, \lambda_6 = -0.07833034592$ .

The results for the five fastest methods used in the Example 4.4 are listed in Table 4.

Table 4

For the Example 4.4 we observe that for  $k = 5$  we have the best rate of convergence.

**Example 4.5.** [11]Test function  $f(x) = \cos x - x = 0$ , which has a unique root in the interval  $(0, 1)$ . We use an approximative value for  $\alpha, \alpha \cong 0.5$  and  $g(x) = \cos x$ . Note that  $g$  is a contraction on  $[0, 1]$ . The values of the parameters  $\lambda_i$  involved in (2.7) are

For  $k = 2$ :

$\lambda_1 = 0.04063425765, \lambda_2 = 0.8775825619$ .

n	k = 2	k = 3	k = 4	k = 5	k = 6	$x_{n+1} = g(x_n)$
	$x_{n+1} = g_\lambda(x_n)$	$x_{n+1} = g_\lambda(x_n)$	$x_{n+1} = g_\lambda(x_n)$	$x_{n+1} = g_\lambda(x_n)$	$x_{n+1} = g_\lambda(x_n)$	
0	1.7	1.7	1.7	1.7	1.7	1.7
1	2.007486966	2.007486965	2.007486965	2.007486949	2.007487047	1.772631238
2	1.999773454	2.000100489	1.999981347	2.000012547	2.00006093	1.828035437
3	2.000006791	2.000001178	2.000000064	2.000000040	1.999999912	1.870174554
4	1.999999796	2.000000017	1.999999996	1.999999980	2.000000124	1.902133792
5	2.000000006	1.999999998	2.000000008	2.000000000	1.999999912	1.926313267
6	1.999999999	2.000000004	2.000000004	2.000000000	2.000000124	1.944570353
7	1.999999997	2.000000000	1.999999996			1.958333871
8			2.000000008			1.968697038
9			2.000000004			1.976492529
10			1.999999996			1.982352284
11			2.000000008			1.986754546
⋮	⋮		2.000000008	⋮	⋮	⋮
32						1.999976320

For  $k = 3$ :

$$\lambda_1 = -0.01929393468, \lambda_2 = 1.117295331, \lambda_3 = -0.2397127693.$$

For  $k = 4$ :

$$\lambda_1 = -0.001010964635, \lambda_2 = 1.007597511, \lambda_3 = -0.02031712882, \lambda_4 = -0.1462637603.$$

For  $k = 5$ :

$$\lambda_1 = 0.0002375393714, \lambda_2 = 0.9976094789, \lambda_3 = 0.00964697338, \lambda_4 = -0.1862158885,$$

$$\lambda_5 = 0.01997606411.$$

and for  $k = 6$ :

$$\lambda_1 = 0.0000090022458559, \lambda_2 = 0.9998948502, \lambda_3 = 0.0005054823177,$$

$$\lambda_4 = -0.1679329185, \lambda_5 = 0.001693094069, \lambda_6 = 0.007313188016.$$

The results for the five fastest methods used in the Example 4.5 are listed in Table 5.

Table 5

n	k = 2	k = 3	k = 4	k = 5	k = 6	$x_{n+1} = g(x_n)$
	$x_{n+1} = g_\lambda(x_n)$	$x_{n+1} = g_\lambda(x_n)$	$x_{n+1} = g_\lambda(x_n)$	$x_{n+1} = g_\lambda(x_n)$	$x_{n+1} = g_\lambda(x_n)$	
0	0.5	0.5	0.5	0.5	0.5	0.5
1	0.7552224168	0.7552224168	0.7552224168	0.7552224173	0.7552224180	0.8775825619
2	0.7393111553	0.7391639529	0.7391407837	0.73914159935	0.7391416688	0.6390124942
3	0.7390872396	0.739085247	0.7390851314	0.7390851334	0.7390851349	0.8026851007
4	0.7390851525	0.7390851333	0.7390851332	0.7390851335	0.7390851334	0.6947780268
5	0.7390851331	0.7390851327	0.7390851332	0.7390851341	0.7390851334	0.7681958313
6		0.7390851332		0.7390851335		0.719165449
7		0.7390851332		0.7390851341		
⋮				⋮		⋮
52						0.7390852281

For the Example 4.5 we observe that for  $k = 2$  we have the best rate of convergence.

**Example 4.6.** [11] Test function  $f(x) = (\sin x)^2 - x^2 + 1 = 0$ .  $f(x)$  is continuous on  $[1, 2]$  and  $f(1) \cdot f(2) < 0$ . By Weierstrass theorem,  $\alpha$ , the root of  $f(x)$ , lies in  $(1, 2)$ . Let  $g(x) = \sqrt{1 + (\sin x)^2}$  and  $\alpha \cong 1.5$ . The values of the parameters  $\lambda_i$  involved in (2.7) are

For  $k = 2$ :

$$\lambda_1 = -1.103967914, \lambda_2 = 0.7026746168.$$

For  $k = 3$ :

$$\lambda_1 = -0.9630432881, \lambda_2 = 0.5147751162, \lambda_3 = 0.06263316685.$$

For  $k = 4$ :

$$\lambda_1 = 0.03406382402, \lambda_2 = -1.479439108, \lambda_3 = 1.392109316, \lambda_4 = -0.2954391443.$$

For  $k = 5$ :

$$\lambda_1 = -0.4994602957, \lambda_2 = -0.4317081221, \lambda_3 = 0.7193783305, \lambda_4 = -0.1631142617,$$

$$\lambda_5 = 0.005723630675.$$

and for  $k = 6$  we have the solutions

$$\lambda_1 = -0.3772313996, \lambda_2 = -0.8391377758, \lambda_3 = 1.262617869, \lambda_4 = -0.5252739538,$$

$$\lambda_5 = 0.1264435280, \lambda_6 = -0.01609598632.$$

The results for the five fastest methods used in the Example 4.6 are listed in Table 6.

Table 6

n	k = 2	k = 3	k = 4	k = 5	k = 6	x <sub>n+1</sub> = g(x <sub>n</sub> )
	x <sub>n+1</sub> = g <sub>λ</sub> (x <sub>n</sub> )	x <sub>n+1</sub> = g <sub>λ</sub> (x <sub>n</sub> )	x <sub>n+1</sub> = g <sub>λ</sub> (x <sub>n</sub> )	x <sub>n+1</sub> = g <sub>λ</sub> (x <sub>n</sub> )	x <sub>n+1</sub> = g <sub>λ</sub> (x <sub>n</sub> )	
0	1.5	1.5	1.5	1.5	1.5	1.5
1	1.407839387	1.407839386	1.407839387	1.407839385	1.407839388	1.412443361
2	1.404493085	1.404495094	1.404495963	1.404495477	1.404495477	1.405394334
3	1.404491648	1.404491648	1.404491651	1.404491648	1.404491650	1.404496296
4	1.404491648		1.404491647		1.404491648	1.404493062
5			1.404491649			1.404491813
6			1.404491646			
			1.404491646			
⋮			⋮			⋮
9						1.404491648

For the Example 4.6 we observe that for k = 2, 3, 5 we have the best rate of convergence.

**Example 4.7.** [11] Test function  $f(x) = e^{x^2+7x-30} - 1 = 0$ . We observe that  $x = 3$  is a root for  $f(x)$ . Let  $g(x) = \sqrt{30 - 7x}$  and an approximative value of  $\alpha$ ,  $\alpha \cong 2.5$ . The values of the parameters  $\lambda_i$  involved in (2.7) are

For  $k = 2$ :

$$\lambda_1 = 0.2969848483, \lambda_2 = 0.2771858582.$$

For  $k = 3$ :

$$\lambda_1 = 1.024597726, \lambda_2 = -0.3049044440, \lambda_3 = 0.1164180604.$$

For  $k = 4$ :

$$\lambda_1 = 0.1757160352, \lambda_2 = 0.7137535850, \lambda_3 = -0.2910451512, \lambda_4 = 0.05432842822.$$

For  $k = 5$ :

$$\lambda_1 = 1.215596106, \lambda_2 = -0.9500545285, \lambda_3 = 0.7072397170, \lambda_4 = -0.2118808700, \lambda_5 = 0.02662092982.$$

and for  $k = 6$ :

$$\lambda_1 = -0.09465278297, \lambda_2 = 1.670443250, \lambda_3 = -1.389158506, \lambda_4 = 0.6266784191, \lambda_5 = -0.1410909280, \lambda_6 = 0.01341694862.$$

The results for the five fastest methods used in the Example 4.6 are listed in Table 7.

Table 7

n	k = 2	k = 3	k = 4	k = 5	k = 6	x <sub>n+1</sub> = g(x <sub>n</sub> )
	x <sub>n+1</sub> = g <sub>λ</sub> (x <sub>n</sub> )	x <sub>n+1</sub> = g <sub>λ</sub> (x <sub>n</sub> )	x <sub>n+1</sub> = g <sub>λ</sub> (x <sub>n</sub> )	x <sub>n+1</sub> = g <sub>λ</sub> (x <sub>n</sub> )	x <sub>n+1</sub> = g <sub>λ</sub> (x <sub>n</sub> )	
0	2.5	2.5	2.5	2.5	2.5	2.5
1	3.020382004	3.020382004	3.020382005	3.020382007	3.020382003	3.535533906
2	2.999645349	2.999947206	3.00009189	3.000037463	3.000042237	2.291563366
3	3.000006356	3.000000021	2.999999980	2.999999986	2.999999988	3.736182067
4	2.999999888	2.999999999	2.999999989	2.999999997	2.999999973	1.961307097
5	3.000000002	3.000000000	3.000000002	3.000000001	3.000000005	4.033714209
6	3.000000001		3.000000001	3.000000005	2.999999993	1.328156821
7	3.000000000		2.999999999	3.000000005	2.999999989	4.550044203
8			3.000000002		2.999999994	.
9			3.000000001		2.999999982	.
10			3.000000001		2.999999991	.
11					2.999999986	.
12					3.000000007	.
13					3.000000007	.

For Example 4.7 we observe that for k = 3 we have the best rate of convergence.

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