# Numerical Stability of the Halley-Iteration for the Solution of a System of Nonlinear Equations 


#### Abstract

By Annie A. M. Cuyt* Abstract. Let $F: \mathbf{R}^{q} \rightarrow \mathbf{R}^{q}$ and $x^{*}$ a simple root in $\mathbf{R}^{q}$ of the system of nonlinear equations $F(x)=0$.

Abstract Pade approximants (APA) and abstract Rational approximants (ARA) for the operator $F$ have been introduced in [2] and [3]. The adjective "abstract" refers to the use of abstract polynomials [ 5$]$ for the construction of the rational operators.

The APA and ARA have been used for the solution of a system of nonlinear equations in [4]. Of particular interest was the following third order iterative procedure: $$
x_{i+1}=x_{i}+\frac{a_{i}^{2}}{a_{i}+\frac{1}{2} F_{i}^{\prime-1} F_{i}^{\prime \prime} a_{i}^{2}}
$$ with $F_{i}^{\prime}$ the 1st Frechet-derivative of $F$ in $x_{i}, a_{i}=-F_{i}^{\prime-1} F_{i}$ the Newton-correction where $F_{i}=F\left(x_{i}\right), F_{i}^{\prime \prime}$ the 2nd Frechet-derivative of $F$ in $x_{i}$ where $F_{i}^{\prime \prime} a_{i}^{2}$ is the bilinear operator $F_{i}^{\prime \prime}$ evaluated in ( $a_{i}, a_{i}$ ), and componentwise multiplication and division in $\mathbf{R}^{q}$. For $q=1$ this technique is known as the Halley-iteration [6, p. 91]. In this paper the numerical stability [7] of the Halley-iteration for the case $q>1$ is investigated and illustrated by a numerical example.


1. Numerical Stability of Iterations. We consider the numerical solution of the equation

$$
\begin{equation*}
F(x)=0 \tag{1}
\end{equation*}
$$

with $F: \mathbf{R}^{q} \rightarrow \mathbf{R}^{q}: x \rightarrow F(x)$, abstract analytic in 0 [5]. Assume that (1) has a simple root $x^{*}$.
We briefly repeat the definition of condition-number given by Woźniakowski [7]. The condition-number should measure the sensitivity of the solution (output) with respect to changes in the data (input). We assume that $F$ depends parametrically on a vector $d \in \mathbf{R}^{p}$, called data vector

$$
F(x)=F(x ; d)
$$

and instead of the exact value $F(x ; d)$ we only have the computed value $\mathrm{fl}(F(x ; d))$ in $t$ digit floating-point binary arithmetic. At best we can expect that $\mathrm{fl}(F(x ; d))$ is the exact value of a slightly perturbed operator at slightly perturbed data

$$
\begin{equation*}
\mathrm{fl}(F(x ; d))=(I+\Delta F) F(x+\Delta x ; d+\Delta d) \tag{2}
\end{equation*}
$$

where $I$ is the $q \times q$ unit-matrix and

$$
\begin{gathered}
\|\Delta x\| \leqslant C_{1} \rho\|x\|, \quad\|\Delta d\| \leqslant C_{2} \rho\|d\|, \\
\|\Delta F\| \leqslant C_{3} \rho \quad(\Delta F \text { a } q \times q \text { matrix }),
\end{gathered}
$$

[^0]for constants $C_{1}, C_{2}, C_{3}$ (only depending on the dimensions of the problem) and with $\rho=2^{-t}$ the relative computer precision [8]. By introducing the Landau-symbol $O$, we could also write
$$
\Delta x=O(\rho), \quad \Delta d=O(\rho), \quad \Delta F=O(\rho)
$$
where the constants in the Landau-notation depend on $x, d$ and the dimensions. We will always, for a given $F$, define the data vector so that (2) holds and so that the condition number (see Definition 1.1) is minimized. Let $\mathrm{fl}(d)$ denote the $t$ digit binary representation of the vector $d$ in floating-point arithmetic
$$
\|\mathrm{fl}(d)-d\| \leqslant C \rho\|d\|, \quad \text { i.e. } \mathrm{fl}(d)-d=O(\rho)
$$

Since $d$ is represented by $\mathrm{fl}(d)$, we solve in fact $F(x ; \mathrm{fl}(d))=0$ instead of $F(x)=0$, independent of the method used to solve (1). Let $F_{x}^{\prime}$ and $F_{d}^{\prime}$ denote the partial Fréchet-derivatives of $F$, respectively with respect to $x$ and $d$.

Now $F(x ; \mathrm{fl}(d))=0$ has a root $\widetilde{x^{*}}$ in the neighborhood of $x^{*}$ and $\widetilde{x^{*}}-x^{*}=$ $O(\rho)$ if $t$ is sufficiently large; thus,

$$
\begin{aligned}
\widetilde{x^{*}}-x^{*}= & -F_{x}^{\prime}\left(x^{*} ; d\right)^{-1} F_{d}^{\prime}\left(x^{*} ; d\right)(\mathrm{fl}(d)-d) \\
& + \text { higher order terms in } \widetilde{x^{*}}-x^{*} \text { and } \mathrm{fl}(d)-d \\
= & -F_{x}^{\prime}\left(x^{*} ; d\right)^{-1} F_{d}^{\prime}\left(x^{*} ; d\right)(\mathrm{fl}(d)-d)+O\left(\rho^{2}\right),
\end{aligned}
$$

where the constant in the Landau-notation depends on $x^{*}, d$ and $F$. For $x^{*} \neq 0:\left\|\widetilde{x^{*}}-x^{*}\right\| /\left\|x^{*}\right\| \leqslant\left\|F_{x}^{\prime}\left(x^{*} ; d\right)^{-1} F_{d}^{\prime}\left(x^{*} ; d\right)\right\| C \rho\|d\| /\left\|x^{*}\right\|+O\left(\rho^{2}\right)$.

Definition 1.1. $\operatorname{Cond}(F ; d)=\left\|F_{x}^{\prime}\left(x^{*} ; d\right)^{-1} F_{d}^{\prime}\left(x^{*} ; d\right)\right\| \cdot\|d\| /\left\|x^{*}\right\|$ is called the condition number of $F$ with respect to the data vector $d$.

A problem is ill-conditioned if $\operatorname{cond}(F ; d) \gg 1$.
Let us now suppose that $F(x ; d)=0$ is solved by an iterative procedure $\Phi\left(x_{i}, F\right.$ ), where $\Phi$ can use several $F_{i}^{(j)}$, the $j$ th Fréchet-derivative of $F$ at $x_{i}$ (if $j=1$ or 2 , a single or double prime is used instead of the superscript $j$ ). If $\left\{x_{i}\right\}$ is the sequence of successive approximations of $x^{*}$, we can at best expect $x_{i}$ to be the representation of a computed value for $\widetilde{x^{*}}$,

$$
\left\|x_{i}-\widetilde{x^{*}}\right\| \leqslant K \rho\left\|\widetilde{x^{*}}\right\| .
$$

So

$$
\begin{aligned}
\left\|x_{i}-x^{*}\right\| & \leqslant\left\|x_{i}-\widetilde{x^{*}}\right\|+\left\|\widetilde{x^{*}}-x^{*}\right\| \leqslant K \rho\left\|\widetilde{x^{*}}\right\|+C \rho \operatorname{cond}(F ; d) \cdot\left\|x^{*}\right\|+O\left(\rho^{2}\right) \\
& \leqslant K \rho\left(\left\|\widetilde{x^{*}}-x^{*}\right\|+\left\|x^{*}\right\|\right)+C \rho \operatorname{cond}(F ; d) \cdot\left\|x^{*}\right\|+O\left(\rho^{2}\right) \\
& \leqslant[K \rho+C \rho \operatorname{cond}(F ; d)] \cdot\left\|x^{*}\right\|+O\left(\rho^{2}\right) .
\end{aligned}
$$

Definition 1.2. An iteration $\Phi$ is called numerically stable if

$$
\lim _{i \rightarrow \infty}\left\|x_{i}-x^{*}\right\| \leqslant \rho \cdot\left\|x^{*}\right\| \cdot(C \operatorname{cond}(F ; d)+K)+O\left(\rho^{2}\right)
$$ where the constants $C$ and $K$ depend on $x^{*}, d$ and $F$.

[n practice we often want to find an approximation $x_{i}$ such that $\left\|x_{i}-x^{*}\right\| \leqslant \varepsilon$. $\mid x^{*} \|$. This is possible if the problem is sufficiently well-conditioned, i.e., , $\operatorname{cond}(F ; d)=O(\varepsilon)$. In floating-point arithmetic we have

$$
x_{i+1}=\Phi\left(x_{i}, F\right)+\xi_{i}, \quad \text { where } \xi_{i}=\mathrm{fl}\left(\Phi\left(x_{i}, F\right)\right)-\Phi\left(x_{i}, F\right)
$$

Theorem 1.1. A convergent iterative procedure $\Phi\left(x_{i}, F\right)$, i.e.

$$
\lim _{i \rightarrow \infty}\left\|\Phi\left(x_{i}, F\right)-x^{*}\right\|=0,
$$

is numerically stable if $\lim _{i \rightarrow \infty}\left\|\xi_{i}\right\| \leqslant \rho\left\|x^{*}\right\|(C \operatorname{cond}(F ; d)+K)+O\left(\rho^{2}\right)$.
Proof. We simply verify the definition.

$$
\begin{aligned}
\lim _{i \rightarrow \infty}\left\|x_{i}-x^{*}\right\| & \leqslant \lim _{i \rightarrow \infty}\left[\left\|\Phi\left(x_{i-1}, F\right)-x^{*}\right\|+\left\|\xi_{i-1}\right\|\right] \\
& =\lim _{i \rightarrow \infty}\left\|\xi_{i-1}\right\| \leqslant \rho\left\|x^{*}\right\|(C \operatorname{cond}(F ; d)+K)+O\left(\rho^{2}\right)
\end{aligned}
$$

## 2. Abstract Padé Approximants (APA) and Abstract Rational Approximants (ARA)

 for the Solution of a System of Nonlinear Equations. Let $x_{i}$ be the $i$ th approximant of the root $x^{*}$ in the iterative process, $y_{i}=F\left(x_{i}\right)$ and the Newton-correction $a_{i}=-F_{i}^{\prime-1} F_{i}$. Using the Inversion Theorem [1, p. 381] we can see that$$
\begin{equation*}
x^{*}=x_{i}+a_{i}-\frac{1}{2} F_{i}^{\prime-1} F_{i}^{\prime \prime} a_{i}^{2}+O\left(a_{i}^{3}\right), \tag{3}
\end{equation*}
$$

where $F_{i}^{\prime \prime} a_{i}^{2}$ is the bilinear operator $F_{i}^{\prime \prime}$ evaluated on $\left(a_{i}, a_{i}\right)$. The Newton-iteration results from approximating the series in (3) by its first two terms, i.e., the ( 1,0 )-APA [2].

In [7] Woźniakowski proves numerical stability of the Newton-iteration under a natural assumption on the computed evaluation of $F$.

Theorem 2.1. If
(a) $\mathrm{fl}\left(F\left(x_{i} ; d\right)\right)=\left(I+\Delta F_{i}\right) F\left(x_{i}+\Delta x_{i} ; d+\Delta d_{i}\right)=F\left(x_{i}\right)+\delta F_{i}$, with

$$
\delta F_{i}=\Delta F_{i} F\left(x_{i}\right)+F_{x}^{\prime}\left(x_{i}\right) \Delta x_{i}+F_{d}^{\prime}\left(x_{i}\right) \Delta d_{i}+O\left(\rho^{2}\right),
$$

(b) $\mathrm{fl}\left(F^{\prime}\left(x_{i} ; d\right)\right)=F^{\prime}\left(x_{i}\right)+\delta F_{i}^{\prime}$, with $\delta F_{i}^{\prime}=O(\rho)$,
(c) the computed correction $\mathrm{fl}\left(a_{i}\right)$ is the exact solution of a perturbed linear system

$$
\left(F^{\prime}\left(x_{i}\right)+\delta F_{i}^{\prime}+E_{i}\right) \mathrm{fl}\left(a_{i}\right)=-F\left(x_{i}\right)-\delta F_{i} \quad \text { with } E_{i}=O(\rho)
$$

then the Newton-iteration is numerically stable.

Proof. In [7].
Another way to approximate $x^{*}$ is to use the (1, 1)-ARA [2] for the power series (3), i.e.

$$
\begin{equation*}
x_{i+1}=x_{i}+\frac{a_{i}^{2}}{a_{i}+\frac{1}{2} F_{i}^{\prime-1} F_{i}^{\prime \prime} a_{i}^{2}}, \tag{4}
\end{equation*}
$$

where multiplication and division of the vectors in $\mathbf{R}^{q}$ in the numerator and denominator of (4) are componentwise. For $q=1$ the iteration (4) is the wellknown Halley-iteration. We will also use the name Halley-iteration for the case $q \geqslant 1$. We will now prove numerical stability of this iteration under assumptions similar to the assumptions for the Newton-iteration. We will also assume that the divisions in (4) are such that

$$
\begin{equation*}
\left(\frac{1}{a_{i}+\frac{1}{2} F_{i}^{\prime-1} F_{i}^{\prime \prime} a_{i}^{2}}\right)^{j} O\left(\left\|a_{i}\right\|^{j-k} \rho^{k+l}\right)=O\left(\rho^{l}\right) \tag{5}
\end{equation*}
$$

Condition (5) takes care of the fact that the denominator of the correction-term in
(4) does not become too small in comparison with $O\left(\left\|a_{i}\right\|^{j-k} \rho^{k}\right)$.

The assumption of (5) is a natural generalization of the following relations:

$$
\text { for } q=1, \lim _{i \rightarrow \infty} \frac{a_{i}}{a_{i}+\frac{1}{2} F_{i}^{\prime-1} F_{i}^{\prime \prime} a_{i}^{2}}=1,
$$

$$
\begin{align*}
& \text { and so } \exists L \in \mathbf{N} \supset-\forall i \geqslant L:\left|\frac{a_{i}}{a_{i}+\frac{1}{2} F_{i}^{\prime-1} F_{i}^{\prime \prime} a_{i}^{2}}\right| \leqslant 1+D  \tag{5a}\\
& \text { (case } j=1, k=0, l=0 \text { ) with } D \in \mathbf{R}_{0}^{+}
\end{align*}
$$

in a convergent process (4): $\lim _{i \rightarrow \infty}\left\|x^{*}-x_{i}\right\|=0$, and thus

$$
\lim _{i \rightarrow \infty} a_{i}=0, \quad \text { i.e. } \exists M \in \mathbf{N} \supset \quad \forall i \geqslant M: a_{i}=O(\rho)
$$

$$
\text { and so } \forall i \geqslant M: a_{i}^{2}=O\left(\left\|a_{i}\right\| \rho\right) \text {; also }
$$

$$
\lim _{i \rightarrow \infty} \frac{a_{i}^{2}}{a_{i}+\frac{1}{2} F_{i}^{\prime-1} F_{i}^{\prime \prime} a_{i}^{2}}=0, \text { i.e. }
$$

$$
\begin{equation*}
\exists N \in \mathbf{N} \supset \quad \forall i \geqslant N: \frac{a_{i}^{2}}{a_{i}+\frac{1}{2} F_{i}^{\prime-1} F_{i}^{\prime \prime} a_{i}^{2}}=O(\rho), \tag{5b}
\end{equation*}
$$

$$
\text { and so } \forall i \geqslant \max (N, M): \frac{a_{i}^{2}}{a_{i}+\frac{1}{2} F_{i}^{\prime-1} F_{i}^{\prime \prime} a_{i}^{2}}
$$

$$
=\frac{1}{a_{i}+\frac{1}{2} F_{i}^{\prime-1} F_{i}^{\prime \prime} a_{i}^{2}} O\left(\left\|a_{i}\right\| \rho\right)
$$

$$
=O(\rho)
$$

$$
(\text { case } j=1, k=0, l=1)
$$

Theorem 2.2. If
(a) $\mathrm{fl}\left(F\left(x_{i} ; d\right)\right)=\left(I+\Delta F_{i}\right) F\left(x_{i}+\Delta x_{i} ; d+\Delta d_{i}\right)=F\left(x_{i}\right)+\delta F_{i}$ with

$$
\delta F_{i}=\Delta F_{i} F\left(x_{i}\right)+F_{x}^{\prime}\left(x_{i}\right) \Delta x_{i}+F_{d}^{\prime}\left(x_{i}\right) \Delta d_{i}+O\left(\rho^{2}\right)
$$

(b) $\mathrm{fl}\left(F^{\prime}\left(x_{i} ; d\right)\right)=F^{\prime}\left(x_{i}\right)+\delta F_{i}^{\prime}$ with $\delta F_{i}^{\prime}=O(\rho)$,
(c) $\mathrm{fl}\left(F^{\prime \prime}\left(x_{i} ; d\right)\right)=F^{\prime \prime}\left(x_{i}\right)+\delta F_{i}^{\prime \prime}$ with $\delta F_{i}^{\prime \prime}=O(\rho)$,
(d) the computed correction $\mathrm{fl}\left(a_{i}\right)$ is the exact solution of a perturbed linear system

$$
\left(F^{\prime}\left(x_{i}\right)+\delta F_{i}^{\prime}+E_{i, 1}\right) \mathrm{fl}\left(a_{i}\right)=-F\left(x_{i}\right)-\delta F_{i} \quad \text { with } E_{i, 1}=O(\rho)
$$

(e) analogously,

$$
\begin{aligned}
\left(F^{\prime}\left(x_{i}\right)+\delta F_{i}^{\prime}+E_{i, 2}\right) \mathrm{fl}\left(b_{i}\right)= & \left(F^{\prime \prime}\left(x_{i}\right)+\delta F_{i}^{\prime \prime}\right) \mathrm{fl}\left(a_{i}\right)^{2} \\
& \text { with } E_{i, 2}=O(\rho) \text { and } b_{i}=F_{i}^{\prime-1} F_{i}^{\prime \prime} a_{i}^{2}
\end{aligned}
$$

and (5) holds, then the iteration (4) is numerically stable.
Proof. Let $F^{\prime}\left(x_{i}\right)+\delta F_{i}^{\prime}+E_{i, 1}=F^{\prime}\left(x_{i}\right)\left(I+H_{i, 1}\right)$, where

$$
H_{i, 1}=F^{\prime}\left(x_{i}^{*}\right)^{-1}\left\{\delta F_{i}^{\prime}+E_{i, 1}\right\}=O(\rho)
$$

because of (b) and (d). So for small $\rho$,

$$
\left(I+H_{i, 1}\right)^{-1}=I-H_{i, 1}+O\left(\rho^{2}\right)
$$

Thus
(6)

$$
\mathrm{fl}\left(a_{i}\right)=\left(I-H_{i, 1}\right) F_{i}^{\prime-1}\left(-F_{i}-\delta F_{i}\right) .
$$

Analogously

$$
\mathrm{fl}\left(b_{i}\right)=\left(I-H_{i, 2}\right) F_{i}^{\prime-1}\left(F_{i}^{\prime \prime}+\delta F_{i}^{\prime \prime}\right) \mathrm{fl}\left(a_{i}\right)^{2} \text { with } H_{i, 2}=O(\rho) .
$$

Now

$$
\begin{aligned}
\left(F_{i}^{\prime \prime}+\right. & \left.\delta F_{i}^{\prime \prime}\right) \mathrm{fl}\left(a_{i}\right)^{2}=\left(F_{i}^{\prime \prime}+\delta F_{i}^{\prime \prime}\right)\left[\left(I-H_{i, 1}\right) F_{i}^{\prime-1}\left(-F_{i}-\delta F_{i}\right)\right]^{2} \\
& =\left(F_{i}^{\prime \prime}+\delta F_{i}^{\prime \prime}\right) a_{i}^{2}+2\left(F_{i}^{\prime \prime}+\delta F_{i}^{\prime \prime}\right)\left(F_{i}^{\prime-1} F_{i}, F_{i}^{\prime-1} \delta F_{i}-H_{i, 1} F_{i}^{\prime-1} F_{i}\right)+O\left(\rho^{2}\right) \\
& =\left(F_{i}^{\prime \prime}+\delta F_{i}^{\prime \prime}\right) a_{i}^{2}-2 F_{i}^{\prime \prime}\left(a_{i}, F_{i}^{\prime-1} \delta F_{i}-H_{i, 1} F_{i}^{\prime-1} F_{i}\right)+O\left(\rho^{2}\right)
\end{aligned}
$$

Thus

$$
\begin{aligned}
\mathrm{fl}\left(b_{i}\right)= & F_{i}^{\prime-1}\left(F_{i}^{\prime \prime}+\delta F_{i}^{\prime \prime}\right) a_{i}^{2}-2 F_{i}^{\prime-1} F_{i}^{\prime \prime}\left(a_{i}, F_{i}^{\prime-1} \delta F_{i}-H_{i, 1} F_{i}^{\prime-1} F_{i}\right) \\
& -H_{i, 2} F_{i}^{\prime-1} F_{i}^{\prime \prime} a_{i}^{2}+O\left(\rho^{2}\right) .
\end{aligned}
$$

A computed approximation $x_{i+1}$ satisfies

$$
x_{i+1}=\left(I+\delta I_{i, 1}\right)\left(x_{i}+\left(I+\delta I_{i, 2}\right) \frac{\mathrm{fl}\left(a_{i}\right)^{2}}{\mathrm{fl}\left(a_{i}\right)+\frac{1}{2} \mathrm{fl}\left(b_{i}\right)}\right)
$$

where $\delta I_{i, 1}$ and $\delta I_{i, 2}$ are diagonal matrices and $\delta I_{i, 1}=O(\rho)$ and $\delta I_{i, 2}=O(\rho)$. So

$$
x_{i+1}=\left(I+\delta I_{i, 1}\right)\left(x_{i}+\left(I+\delta I_{i, 2}\right) \frac{a_{i}^{2}-2 a_{i} \cdot\left(F_{i}^{\prime-1} \delta F_{i}+H_{i, 1} a_{i}\right)+O\left(\rho^{2}\right)}{a_{i}+\frac{1}{2} b_{i}-\delta a_{i}+O\left(\rho^{2}\right)}\right)
$$

where

$$
\begin{aligned}
\delta a_{i}= & F_{i}^{\prime-1} \delta F_{i}+H_{i, 1} a_{i}-\frac{1}{2} F_{i}^{\prime-1} \delta F_{i}^{\prime \prime} a_{i}^{2} \\
& +\frac{1}{2} H_{i, 2} F_{i}^{\prime-1} F_{i}^{\prime \prime} a_{i}^{2}+F_{i}^{\prime-1} F_{i}^{\prime \prime}\left(a_{i}, F_{i}^{\prime-1} \delta F_{i}-H_{i, 1} F_{i}^{\prime-1} F_{i}\right)
\end{aligned}
$$

Using (6), we find

$$
\mathrm{fl}\left(a_{i}\right)-a_{i}+H_{i, 1} a_{i}-H_{i, 1} F_{i}^{\prime-1} \delta F_{i}=-F_{i}^{\prime-1} \delta F_{i},
$$

and thus, for positive constants $D_{1}$ and $D_{2}$,

$$
\left\|F_{i}^{\prime-1} \delta F_{i}\right\| \leqslant D_{2} \rho\left\|a_{i}\right\| \quad \text { since }\left\|\mathrm{fl}\left(a_{i}\right)-a_{i}\right\| \leqslant D_{1} \rho\left\|a_{i}\right\|
$$

and

$$
\left\|F_{i}^{\prime-1}\right\| \cdot\left\|F_{i}\right\| \leqslant\left\|F_{i}^{\prime-1}\right\| \cdot\left\|F_{i}^{\prime}\right\| \cdot\left\|a_{i}\right\|
$$

Thus

$$
x_{i+1}=\left(I+\delta I_{i, 1}\right)\left(x_{i}+\frac{a_{i}^{2}-2 a_{i}\left(F_{i}^{\prime-1} \delta F_{i}+H_{i, 1} a_{i}\right)+\delta I_{i, 2} a_{i}^{2}+O\left(\rho^{2}\left\|a_{i}\right\|^{2}\right)}{a_{i}+\frac{1}{2} b_{i}-\delta a_{i}+O\left(\rho^{2}\right)}\right),
$$

where $\delta I_{i, 2} a_{i}^{2}$ is the linear operator $\delta I_{i, 2}$ evaluated in $a_{i}^{2}$ (componentwise square of the vector $a_{i}$ ). So

$$
x_{i+1}=\left(I+\delta I_{i, 1}\right)\left(x_{i}+\frac{a_{i}^{2}-2 a_{i}\left(F_{i}^{\prime-1} \delta F_{i}+H_{i, 1} a_{i}\right)+\delta I_{i, 2} a_{i}^{2}+O\left(\rho^{2}\left\|a_{i}\right\|^{2}\right)}{a_{i}+\frac{1}{2} b_{i}} \cdot c_{i}\right]
$$

with

$$
c_{i}=1+\frac{1}{a_{i}+\frac{1}{2} b_{i}}\left(\delta a_{i}+O\left(\rho^{2}\right)\right)+\left(\frac{1}{a_{i}+\frac{1}{2} b_{i}}\right)^{2} O\left(\left\|a_{i}\right\|^{2-k} \rho^{k+2}, k=0,1,2\right)
$$

since $\delta a_{i}=O\left(\rho\left\|a_{i}\right\|\right)$; in $c_{i}$ we have used the notation 1 for the unit vector ( $1, \ldots, 1$ ).

Using (5), we conclude

$$
\left(\frac{1}{a_{i}+\frac{1}{2} b_{i}}\right)^{2} O\left(\left\|a_{i}\right\|^{2-k} \rho^{k+2}, k=0,1,2\right)=O\left(\rho^{2}\right)
$$

For $\xi_{i}=x_{i+1}-\Phi\left(x_{i}, F\right)$, we have

$$
\begin{aligned}
\xi_{i}= & \delta I_{i, 1} x_{i}+\frac{a_{i}^{2}}{a_{i}+\frac{1}{2} b_{i}}\left(c_{i}-1\right) \\
& +\frac{-2 a_{i}\left(F_{i}^{\prime-1} \delta F_{i}+H_{i, 1} a_{i}\right)+\delta I_{i, 2} a_{i}^{2}+O\left(\rho^{2}\left\|a_{i}\right\|^{2}\right)}{a_{i}+\frac{1}{2} b_{i}} \cdot c_{i} \\
& +\delta I_{i, 1} \frac{a_{i}^{2}}{a_{i}+\frac{1}{2} b_{i}} \cdot c_{i}+O\left(\rho^{2}\right)
\end{aligned}
$$

So

$$
\begin{aligned}
\xi_{i}= & \delta I_{i, 1} x_{i}+\left(\frac{1}{a_{i}+\frac{1}{2} b_{i}}\right)^{2} O\left(\rho\left\|a_{i}\right\|^{3}, \rho^{2}\left\|a_{i}\right\|^{2}\right)+\frac{1}{a_{i}+\frac{1}{2} b_{i}} O\left(\rho^{2}\left\|a_{i}\right\|^{2}\right) \\
& +\frac{1}{a_{i}+\frac{1}{2} b_{i}}\left(-2 a_{i} F_{i}^{\prime-1} \delta F_{i}+O\left(\rho\left\|a_{i}\right\|^{2}, \rho^{2}\left\|a_{i}\right\|^{2}\right)\right) \cdot(1+O(\rho)) \\
& +O\left(\rho^{2}\right) .
\end{aligned}
$$

Thus

$$
\left\|\xi_{i}\right\| \leqslant k_{1} \rho\left\|x_{i}\right\|+k_{2} \rho\left\|a_{i}\right\|+\left\|\frac{-2 a_{i}}{a_{i}+\frac{1}{2} b_{i}} F_{i}^{\prime-1} \delta F_{i}\right\|+O\left(\rho^{2}\right),
$$

and since

$$
\begin{aligned}
\frac{-2 a_{i}}{a_{i}+\frac{1}{2} b_{i}} F_{i}^{\prime-1} \delta F_{i}= & \frac{-2 a_{i}}{a_{i}+\frac{1}{2} b_{i}} F_{i}^{\prime-1}\left(\Delta F_{i} F\left(x_{i}\right)+F_{i}^{\prime} \Delta x_{i}+F_{d}^{\prime} \Delta d_{i}+O\left(\rho^{2}\right)\right) \\
= & \frac{1}{a_{i}+\frac{1}{2} b_{i}} O\left(\rho\left\|a_{i}\right\|\right) F\left(x_{i}\right)-\frac{2 a_{i}}{a_{i}+\frac{1}{2} b_{i}} \Delta x_{i} \\
& -\frac{2 a_{i}}{a_{i}+\frac{1}{2} b_{i}} F_{i}^{\prime-1} F_{d}^{\prime} \Delta d_{i}+\frac{1}{a_{i}+\frac{1}{2} b_{i}} O\left(\rho^{2}\left\|a_{i}\right\|\right)
\end{aligned}
$$

we find that

$$
\lim _{i \rightarrow \infty}\left\|\xi_{i}\right\| \leqslant \rho\left\|x^{*}\right\|(K+C \operatorname{cond}(F ; d))+O\left(\rho^{2}\right)
$$

for $\lim _{i \rightarrow \infty} a_{i}=0=\lim _{i \rightarrow \infty} F\left(x_{i}\right)$ in a convergent process and $a_{i} \Delta x_{i}=O\left(\rho\left\|a_{i}\right\|\right)$ and $a_{i} F_{i}^{\prime-1} F_{d}^{\prime} \Delta d_{i}=O\left(\rho\left\|a_{i}\right\|\right)$.
3. Numerical Example. Consider the following operator:

$$
F: \mathbf{R}^{2} \rightarrow \mathbf{R}^{2}:(x, y) \rightarrow\binom{e^{-x+y}-d_{1}}{e^{-x-y}-d_{2}} \quad \text { with } d_{1}>0 \text { and } d_{2}>0
$$

The operator $F$ has a simple root $x^{*}=\left(-\frac{1}{2} \ln \left(d_{1} d_{2}\right), \frac{1}{2} \ln \left(d_{1} / d_{2}\right)\right)$. Clearly

$$
d=\left(d_{1}, d_{2}\right)
$$

is the data vector. Now

$$
\mathrm{fl}(F(x, y ; d))=\left(\begin{array}{l}
{\left[\left(1+\varepsilon_{1}\right) e^{\left(-x-\Delta^{\prime} x+y+\Delta^{\prime} y\right)\left(1+\theta_{1}\right)}-\left(d_{1}+\Delta_{1}^{\prime} d\right)\right]\left(1+\kappa_{1}\right)} \\
{\left[\left(1+\varepsilon_{2}\right) e^{\left(-x-\Delta^{\prime} x-y-\Delta^{\prime} y\right)\left(1+\theta_{2}\right)}-\left(d_{2}+\Delta_{2}^{\prime} d\right)\right]\left(1+\kappa_{2}\right)}
\end{array}\right]
$$

where $\mathrm{fl}(x)=x+\Delta^{\prime} x, \mathrm{fl}(y)=y+\Delta^{\prime} y, \mathrm{fl}\left(d_{1}\right)=d_{1}+\Delta_{1}^{\prime} d, \mathrm{fl}\left(d_{2}\right)=d_{2}+\Delta_{2}^{\prime} d, \theta_{1}$ is caused by $-\mathrm{fl}(x)+\mathrm{fl}(y), \theta_{2}$ is caused by $-\mathrm{fl}(x)-\mathrm{fl}(y), \varepsilon_{i}$ are caused by the exponential evaluations $(i=1,2), \kappa_{i}$ are caused by the subtraction of $\mathrm{fl}\left(d_{i}\right)$ ( $i=1,2$ ).

One can rewrite $\mathrm{fl}(F(x, y ; d))=(I+\Delta F) F(x+\Delta x, y+\Delta y ; d+\Delta d)$ with

$$
\begin{gathered}
\Delta x=x \theta_{1}+\Delta^{\prime} x\left(1+\theta_{1}\right), \quad \Delta y=y \theta_{1}+\Delta^{\prime} y\left(1+\theta_{1}\right), \quad \Delta d=\left(\Delta_{1} d, \Delta_{2} d\right), \\
\Delta_{1} d=\frac{\Delta_{1}^{\prime} d-\varepsilon_{1} d_{1}}{1+\varepsilon_{1}}, \\
\Delta_{2} d=\frac{\Delta_{2}^{\prime} d-\varepsilon_{2} d_{2}}{1+\varepsilon_{2}}+\frac{d_{2}+\Delta_{2}^{\prime} d}{1+\varepsilon_{2}}\left(e^{\left(x+\Delta^{\prime} x+y+\Delta^{\prime} y\right)\left(\theta_{2}-\theta_{1}\right)}-1\right) \\
\Delta F=\left(\begin{array}{cc}
\left(1+\varepsilon_{1}\right)\left(1+\kappa_{1}\right)-1 & 0 \\
0 & \left(1+\varepsilon_{2}\right)\left(1+\kappa_{2}\right) e^{\left(x+\Delta^{\prime} x+y+\Delta^{\prime} y\right)\left(\theta_{1}-\theta_{2}\right)}-1
\end{array}\right)
\end{gathered}
$$

The inverse of the Jacobian matrix in the root $x^{*}$ is

$$
\frac{1}{2\left(d_{1} \cdot d_{2}\right)}\left(\begin{array}{cc}
-d_{2} & -d_{1} \\
d_{2} & -d_{1}
\end{array}\right) \quad \text { and } \quad F_{d}^{\prime}=\left(\begin{array}{cc}
-1 & 0 \\
0 & -1
\end{array}\right) .
$$

The condition number of $F$ with respect to the data vector $d$ is

$$
\left\|F_{x}^{\prime}\left(x^{*} ; d\right)^{-1}\right\| \cdot \frac{\left\|\left(d_{1}, d_{2}\right)\right\|}{\left\|x^{*}\right\|}
$$

Using the Schur-norm $\|A\|=\sqrt{\sum_{i, j} a_{i j}^{2}}$ of a matrix $A=\left(a_{i j}\right)$ and the $l_{2}$-norm $\|a\|=\sqrt{\sum_{i} a_{i}^{2}}$ of a vector $a=\left(a_{i}\right)$, the condition number is

$$
\frac{d_{1}^{2}+d_{2}^{2}}{\sqrt{2} d_{1} \cdot d_{2} \cdot\left\|x^{*}\right\|}
$$

Putting $d_{1}=d=d_{2}$, the root $x^{*}=(-\ln d, 0)$ and the condition number is $\sqrt{2} /|\ln d|$. The problem is extremely well-conditioned if $\operatorname{cond}(F ; d) \leqslant 1$, i.e.,

$$
\left.d \in]-\infty, e^{-\sqrt{2}}\right] \cup\left[e^{\sqrt{2}},+\infty[\right.
$$

The problem is very ill-conditioned if $d=e^{\varepsilon}$ with $\varepsilon$ very small. We will now check some of the conditions of Theorem 2.2. We already know $\mathrm{fl}(F(x, y ; d))=$ $(I+\Delta F) F(x+\Delta x, y+\Delta y ; d+\Delta d)$.

Now

$$
\mathrm{fl}\left(F^{\prime}(x, y ; d)\right)=\mathrm{fl}\left(\begin{array}{cc}
-e^{-x+y} & e^{-x+y} \\
-e^{-x-y} & -e^{-x-y}
\end{array}\right)
$$

where

$$
\begin{aligned}
\mathrm{fl}\left(e^{-x+y}\right)= & \left(1+\varepsilon_{1}\right) e^{\left(-x-\Delta^{\prime} x+y+\Delta^{\prime} y\right)\left(1+\theta_{1}\right)}=\left(1+\varepsilon_{1}\right) e^{-x+y} e^{-\Delta x+\Delta y} \\
= & e^{-x+y}\left[1+\varepsilon_{1}+\left(1+\varepsilon_{1}\right)\left(e^{-\Delta x+\Delta y}-1\right)\right], \\
\mathrm{fl}\left(e^{-x-y}\right)= & \left(1+\varepsilon_{2}\right) e^{\left(-x-\Delta^{\prime} x-y-\Delta^{\prime} y\right)\left(1+\theta_{2}\right)} \\
= & \left(1+\varepsilon_{2}\right) e^{-x-y} e^{-\Delta x-\Delta y} e^{\left(x+\Delta^{\prime} x+y+\Delta^{\prime} y\right)\left(\theta_{1}-\theta_{2}\right)} \\
= & e^{-x-y}\left[1+\varepsilon_{2}+\left(1+\varepsilon_{2}\right)\left(e^{-\Delta x-\Delta y} e^{\left(x+\Delta^{\prime} x+y+\Delta^{\prime} y\right)\left(\theta_{1}-\theta_{2}\right)}-1\right)\right] .
\end{aligned}
$$

So $\mathrm{fl}\left(F^{\prime}(x, y ; d)\right)=F^{\prime}(x, y ; d)+\delta F^{\prime}(x, y ; d)$ with

$$
\begin{aligned}
& \delta F^{\prime}(x, y ; d) \\
& =\left(\begin{array}{cc}
\varepsilon_{1}+\left(1+\varepsilon_{1}\right)\left(e^{-\Delta x+\Delta y}-1\right) & 0 \\
0 & \varepsilon_{2}+\left(1+\varepsilon_{2}\right)\left(e^{-\Delta x-\Delta y} e^{\left(x+\Delta^{\prime} x+y+\Delta^{\prime} y\right)\left(\theta_{1}-\theta_{2}\right)}-1\right)
\end{array}\right) \\
& \quad \cdot F^{\prime}(x, y ; d)=O(\rho) .
\end{aligned}
$$

We can write down an analogous formula for $F^{\prime \prime}(x, y ; d)$.

| $k$ | $x_{6}$ | $y_{6}$ | $\ell$ | $c o n d\left(F ; e^{\left.10^{-k}\right)}\right.$ |
| :---: | :--- | :--- | :--- | :--- |
| 0 | $-0.1000000000000000(01)$ | $0.3597855161523896(-18)$ | 16 | $\sqrt{2}$ |
| 1 | $-0.100000000000000(00)$ | $-0.2376055789464463(-17)$ | 16 | $10 \sqrt{2}$ |
| 2 | $-0.10000000000 \cdot 001(-01)$ | $-0.6397150159689099(-17)$ | 15 | $10^{2 \sqrt{2}}$ |
| 3 | $-0.0999999999999997(-02)$ | $0.5077502606368951(-17)$ | 15 | $10^{3} \sqrt{2}$ |
| 4 | $-0.0999999999999844(-03)$ | $0.3913464269882279(-17)$ | 13 | $10^{4} \sqrt{2}$ |
| 5 | $-0.0999999999997470(-04)$ | $-0.3905797959965137(-17)$ | 12 | $10^{5} \sqrt{2}$ |
| 6 | $-0.099999999986935(-05)$ | $0.5633677343553680(-17)$ | 11 | $10^{6} \sqrt{2}$ |
| 7 | $-0.100000000174599(-06)$ | $-0.1058449777227516(-16)$ | 10 | $10^{7} \sqrt{2}$ |
| 8 | $-0.100000000015281(-07)$ | $0.4124494865312562(-17)$ | 11 | $10^{8} \sqrt{2}$ |
| 9 | $-0.100000007452433(-08)$ | $-0.2449359520991520(-17)$ | 9 | $10^{9} \sqrt{2}$ |
| 10 | $-0.09999999143145861-09)$ | $0.4265833288825851(-17)$ | 8 | $10^{10 \sqrt{2}}$ |
| 11 | $-0.1000000261210709(-10)$ | $-0.6446772724219823(-17)$ | 7 | $10^{11 \sqrt{2}}$ |
| 12 | $-0.0999980430668081(-11)$ | $0.3302303528672576(-17)$ | 5 | $10^{12 \sqrt{2}}$ |
| 13 | $-0.0999761308551817(-12)$ | $0.1322187990417560(-16)$ | 4 | $10^{13 \sqrt{2}}$ |
| 14 | $-0.1000372750236664(-13)$ | $-0.1182870095748150(-16)$ | 4 | $10^{14} \sqrt{2}$ |
| 15 | $-0.0963108239652912(-14)$ | $0.1398012990192197(-17)$ | 2 | $10^{15 \sqrt{2}}$ |
| 16 | $-0.0868560967896870(-15)$ | $0.3349523961106902(-17)$ | 1 | $10^{16 \sqrt{2}}$ |

We remark that the algorithm even behaves considerably well for a condition number of the order of $10^{3}$ or $10^{4}$.

The two linear systems of equations are well-conditioned since the condition number of the linear systems in $x^{*}=\lim _{i \rightarrow \infty} x_{i}$ is

$$
\left\|F_{x}^{\prime}\left(x^{*} ; d\right)^{-1}\right\| \cdot\left\|F_{x}^{\prime}\left(x^{*} ; d\right)\right\|=2 .
$$

One can prove that the use of Gaussian elimination with row pivoting for this example satisfies the conditions (d) and (e) of Theorem 2.2. So we can expect to get a reasonable approximation of the solution of $F(x, y ; d)=0$ using the numerically stable iterative method (4); the numerical results illustrate this. Let us at the same time follow the loss of significant digits in the root $x^{*}$ as the problem becomes worse-conditioned. The calculations are performed in double precision $(t=56)$ on the PDP $11 / 45$ of the University of Antwerp. We will solve the nonlinear system
$F(x, y ; d)=0$ for $d=e^{10^{-k}}, k=0, \ldots, 16$. The root $x^{*}=\left(-10^{-k}, 0\right)$. For each $d$ we give the 6th iteration-step $\left(x_{6}, y_{6}\right)$ in the procedure (4) starting from $\left(x_{0}, y_{0}\right)=$ $(2,2)$, the number $l$ of significant digits in $x_{6}$, and the condition number cond $\left(F ; e^{10^{-k}}\right)$. It is also important to know that the iterative procedure stops at the 6th iteration-step, except for $k=7,13$, and 14 where, respectively, $l=11,5$, and 3 in the last iteration-step $\left(x_{7}, y_{7}\right)$. We have used the stop-criterion

$$
\max \left(\left|x_{i+1}-x_{i}\right|,\left|y_{i+1}-y_{i}\right|\right) \leqslant 10^{-15} \max \left(\left|x_{i+1}\right|,\left|y_{i+1}\right|\right) .
$$

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