# Anderson-Björck for Linear Sequences* 

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#### Abstract

The proposed one-point method for finding the limit of a slowly converging linear sequence features an Anderson-Björck extrapolation step that had previously been applied to the Regula Falsi problem. Convergence is of order 1.839 as compared to $\sqrt{2}$ for the well-known Aitken-Steffensen $\delta^{2}$-process, and to 1.618 for another one-point extrapolation procedure of King. There are examples for computing a polynomial's mutiple root with Newton's method and for finding a fixed point of a nonlinear function.


1. Introduction. Let us suppose that we have a stationary, one-point generating function $\phi$ for a sequence $x_{0}, x_{1}, \ldots$, but that convergence to the limit $\alpha$ is only linear. That is, starting from a given $x_{0}$,

$$
\begin{equation*}
x_{n+1}=\phi\left(x_{n}\right) \equiv \phi_{n}, \quad n=0,1,2, \ldots, \tag{1.1}
\end{equation*}
$$

and the error $\varepsilon_{n}=x_{n}-\alpha$ satisfies

$$
\begin{equation*}
\varepsilon_{n+1}=K \varepsilon_{n}+L \varepsilon_{n}^{2}+M \varepsilon_{n}^{3}+N \varepsilon_{n}^{4}+\cdots, \tag{1.2}
\end{equation*}
$$

with $0<|K|<1$. The error equation (1.2) for $\phi$ we get by expanding $\phi\left(x_{n}\right)$ in a Taylor series about $\alpha$, noting that $\phi(\alpha)=\alpha$, and setting $K=\phi^{\prime}(\alpha), L=\phi^{\prime \prime}(\alpha) / 2!$, $M=\phi^{\prime \prime \prime}(\alpha) / 3$ !, and $N=\phi^{\mathrm{iv}}(\alpha) / 4$ !. The function $\phi$ is called stationary because it is independent of $n$.

Several extrapolation methods for improving the convergence rate are known. We shall review two such methods, and then develop an even faster extrapolation procedure-one based on the Anderson-Björck step introduced and used in [2] for finding a bracketed simple root of a nonlinear equation.

Each of the extrapolation methods may be derived by applying the secant method to a suitable $\phi$-related function having a simple zero at $\alpha$. Since the secant method is hyperlinear for a simple $\alpha$, so will the resulting extrapolation methods be hyperlinear.

In the classical $\delta^{2}$-process, developed by Aitken [1] and applied to $\phi$ by Steffensen [10] (see [4, pp. 135-139] or [9, Appendix E]), the related function is $g=x-\phi$. The point $\alpha$ is a simple root for $g$ because $g(\alpha)=\alpha-\phi(\alpha)=0$, and $g^{\prime}(\alpha)=1-\phi^{\prime}(\alpha)$ $=1-K \neq 0$. Each step of the $\delta^{2}$-process goes as follows: from $x_{n}$ take two linear-error substeps, $x_{n+1}=\phi\left(x_{n}\right)$ and $x_{n+2}=\phi\left(x_{n+1}\right)$. The extrapolation substep

[^0]then consists of one application of the secant method
\[

$$
\begin{equation*}
\bar{x}_{n+2}=x_{n+1}-\left(x_{n}-x_{n+1}\right) \frac{g_{n+1}}{g_{n}-g_{n+1}} \tag{1.3}
\end{equation*}
$$

\]

to $g$, wherein $g_{n} \equiv g\left(x_{n}\right)=x_{n}-x_{n+1}$ and $g_{n+1} \equiv g\left(x_{n+1}\right)=x_{n+1}-x_{n+2}$. Note that the extrapolation itself does not require any evaluations of $\phi$. The corresponding asymptotic error equation

$$
\begin{equation*}
\bar{\varepsilon}_{n+2}=\frac{\varepsilon_{n} \varepsilon_{n+2}-\varepsilon_{n+1}^{2}}{\varepsilon_{n}-2 \varepsilon_{n+1}+\varepsilon_{n+2}} \cong\left(\frac{-L K}{1-K}\right) \varepsilon_{n}^{2} \tag{1.4}
\end{equation*}
$$

for error $\bar{\varepsilon}_{n+2}=\bar{x}_{n+2}-\alpha$ comes out of (1.2) and (1.3). Next, point $\bar{x}_{n+2}$ becomes a new $x_{n}$ for two more linear-error substeps and another extrapolation, and so on. From (1.4), the whole $\delta^{2}$-process has second-order convergence; its efficiency in the sense of Traub [11, p. 263] is $\sqrt{2}$, because $\phi$ must be evaluated twice per step. Again following Traub [11, p. 8], the $\delta^{2}$-process is a two-point method without memory.

But we can attain a higher convergence rate merely by retaining, at each step, information held over from the previous step. This idea is incorporated (with a change of notation) into the one-point extrapolation method with memory described in [7], to wit: define $\phi_{n}=\phi\left(x_{n}\right)$, and get started from $x_{0}$ by taking $x_{1}=\phi_{0}$. For each and every subsequent step, apply a secant-method step to the related function $g=x-\phi$. Thus the method is

$$
\begin{equation*}
x_{n+2}=x_{n+1}-\left(x_{n}-x_{n+1}\right) \frac{g_{n+1}}{g_{n}-g_{n+1}}, \tag{1.5}
\end{equation*}
$$

starting at $g_{0} \equiv g\left(x_{0}\right)=x_{0}-\phi_{0}$ and $g_{1} \equiv g\left(x_{1}\right)=x_{1}-\phi_{1}$. Since ultimately $\left|\varepsilon_{n+1}\right| \ll\left|\varepsilon_{n}\right|$, the error equation for the one-point method of [7] turns out to be the following (from (1.5) and the error expression (1.2) for $\phi$ ):

$$
\begin{equation*}
\varepsilon_{n+2} \cong\left(\frac{-L K}{1-K}\right) \varepsilon_{n} \varepsilon_{n+1} \tag{1.6}
\end{equation*}
$$

That is to say, the error equation is of the same form as that of the usual secant method for a simple root. Thus the method has a convergence rate of $(1+\sqrt{5}) / 2 \doteq$ 1.618. Furthermore the scheme also has an efficiency of 1.618 because only one new $\phi$-value needs to be calculated each step. This compares with 1.414 for the $\delta^{2}$-process.

In passing, we mention several other procedures that, while not as efficient as the method to be developed, are nevertheless hyperlinear: the extrapolation methods of Van de Vel [12] (efficiency 1.414) and of King [6] (1.587), in which the linear error term for $\phi$ is effectively subtracted out; the nonextrapolation methods of Esser [3] (1.587) and of King [5] (1.618), in which $\phi$ is replaced by a hyperlinear generating function using divided differences and analogous to $x-(x-\phi) /\left(1-\phi^{\prime}\right)$.

But now, on to the proposed extrapolation procedure.
2. A New Iteration Procedure Based on the Anderson-Björck Extrapolation Step. The idea in an Anderson-Björck step is (i) to fit a parabola $\tilde{g}_{n+2}$ to three points $\left(x_{n}, g_{n}\right),\left(x_{n+1}, g_{n+1}\right),\left(x_{n+2}, g_{n+2}\right)$ of a function $g$ whose zero is sought, (ii) to compute the derivative $\tilde{g}_{n+2}^{\prime}$ to $\tilde{g}_{n+2}$ at $x_{n+2}$, and then (iii) to take a Newton-like
step from $x_{n+2}$ to a new $x_{n+3}$, using $g_{n+2} \equiv \tilde{g}_{n+2}$ and $\tilde{g}_{n+2}^{\prime}$ :

$$
\begin{equation*}
x_{n+3}=x_{n+2}-\frac{g_{n+2}}{\tilde{g}_{n+2}^{\prime}} . \tag{2.1}
\end{equation*}
$$

It turns out that $\tilde{g}_{n+2}^{\prime}$ can be written (and computed) in terms of divided differences of $g$ as

$$
\begin{equation*}
\tilde{g}_{n+2}^{\prime}=g\left[x_{n+2}, x_{n+1}\right]+g\left[x_{n+2}, x_{n}\right]-g\left[x_{n}, x_{n+1}\right] \tag{2.2}
\end{equation*}
$$

where $g\left[x_{j}, x_{k}\right]=\left(g_{k}-g_{j}\right) /\left(x_{k}-x_{j}\right)$.
In our procedure we take $g$ to be the $\phi$-related function $g=x-\phi$. Starting from a given $x_{0}$, we (1) compute $\phi\left(x_{0}\right) \equiv \phi_{0}$ and set $g_{0}=x_{0}-\phi_{0}$. Now (2) take $x_{1}=\phi_{0}$, compute $\phi\left(x_{1}\right) \equiv \phi_{1}$, and set $g_{1}=x_{1}-\phi_{1}$. At this point we could have chosen to repeat the cycle, i.e., ( $3^{*}$ ) take $x_{2}=\phi_{1}$, compute $\phi_{2}$, and set $g_{2}=x_{2}-\phi_{2}$. Finally (4) find $x_{3}$ and succeeding points by means of the Anderson-Björck step (2.1), and at each point set the new $g_{n+3}=x_{n+3}-\phi_{n+3}$.

But after step (2) we already have enough information to apply the $\delta^{2}$-process to points $P_{0}=\left(x_{0}, g_{0}\right)$ and $P_{1}=\left(x_{1}, g_{1}\right)$. This we choose to do because it accelerates the convergence process. Consequently, instead of (3*) in our procedure we substitute the following: Step (3) take $x_{2}$ to be the $\delta^{2}$-extrapolant of $\left(x_{0}, g_{0}\right)$ and $\left(x_{1}, g_{1}\right)$, compute $\phi\left(x_{2}\right) \equiv \phi_{2}$, and set $g_{2}=x_{2}-\phi_{2}$. It is easy to show that we can accomplish the programming of Step (3) quite simply by setting (in sequence) $\left(x_{2}, g_{2}\right)=\left(x_{1}, g_{1}\right),\left(x_{1}, g_{1}\right)=\left(x_{0}, g_{0}\right),\left(x_{0}, g_{0}\right)=\left(2 x_{1}-x_{2}, 2 g_{1}-g_{2}\right)$, and then using the general step (2.1). (In this case the parabola $\tilde{g}_{2}$ degenerates into the straight line through $P_{0}$ and $P_{1}$.)

We know that $g$ has a simple root at $\alpha$ (because $g(\alpha)=\alpha-\phi(\alpha)=0$ and $\left.g^{\prime}(\alpha)=1-\phi^{\prime}(\alpha)=1-K \neq 0\right)$. Furthermore, $g^{\prime \prime}(\alpha)=-2 L, g^{\prime \prime \prime}(\alpha)=-6 M$, and $g^{\mathrm{iv}}(\alpha)=-24 N$. It can then readily be shown from (1.2), (2.1), and (2.2) that

$$
\begin{equation*}
\varepsilon_{n+3}=\frac{M}{1-K} \varepsilon_{n} \varepsilon_{n+1} \varepsilon_{n+2}-\frac{L}{1-K} \varepsilon_{n+2}^{2}+\frac{N}{1-K} \varepsilon_{n}^{2} \varepsilon_{n+1} \varepsilon_{n+2}+\cdots . \tag{2.3}
\end{equation*}
$$

Asymptotically the second and third terms are negligible, so the error equation for our procedure is

$$
\begin{equation*}
\varepsilon_{n+3} \cong \frac{M}{1-K} \varepsilon_{n} \varepsilon_{n+1} \varepsilon_{n+2} . \tag{2.4}
\end{equation*}
$$

Both the rate of convergence and the efficiency of the method, therefore, are 1.839 (see Muller [8, p. 212] for a derivation of convergence rate for a method with an error equation of the form (2.4)). In the terminology of Traub, the procedure may be classified as a one-point method with memory.

After starting, of course, we need not compute later terms of the original sequence (1.1). Instead, $g$ is computed from $\phi$ as applied to successively more and more accurate estimates of the root $\alpha$, and $g$ converges rapidly to zero.

We can get an estimate for the coefficient $K$ in the linear error term of (1.2) by forming the ratio

$$
\begin{equation*}
K_{n+4}=\frac{\phi_{n+3}-\phi_{n+2}}{x_{n+3}-x_{n+2}} \tag{2.5}
\end{equation*}
$$

To determine $K_{n+4}$ in terms of $\varepsilon_{n+2}$ and $\varepsilon_{n+3}$, apply (1.2) to $\phi_{n+3}-\phi_{n+2}$, thus obtaining

$$
\begin{equation*}
K_{n+4}=\frac{K\left(\varepsilon_{n+3}-\varepsilon_{n+2}\right)+L\left(\varepsilon_{n+3}^{2}-\varepsilon_{n+2}^{2}\right)+\cdots}{\varepsilon_{n+3}-\varepsilon_{n+2}} \doteq K+L\left(\varepsilon_{n+2}+\varepsilon_{n+3}\right) . \tag{2.6}
\end{equation*}
$$

The estimate $K_{n+4}$ may be computed and displayed each step along with the current iterate $x_{n+3}$.

In all three of the methods outlined, the slowly converging generating function $\phi$ is replaced by a function $g=x-\phi$ with a simple zero at $\alpha$. It should be emphasized that with such a transformation most of the many well-known hyperlinear methods for solving $g(x)=0$ may be utilized to find $\alpha$.
3. Examples. All three of the examples are taken from [7], with computations done in quadruple precision on an IBM 370.
a. The first two examples are for Newton's method,

$$
\begin{equation*}
x_{n+1}=\phi\left(x_{n}\right)=x_{n}-\frac{f_{n}}{f_{n}^{\prime}}, \tag{3.1}
\end{equation*}
$$

applied to a nonlinear function $f$ with a multiple root at $\alpha$ of multiplicity $m \neq 1$. Tables 1 and 2 are for functions $f=(x-1)^{2} \tan (\pi x / 4)$ and $f=x \sin \left([x-1]^{4}\right)$, respectively. For the proposed method, the actual error $\varepsilon_{n}=x_{n}-\alpha$ and the final estimate of $\varepsilon_{n}$ from the first three error terms (2.3) are given. We can approximate the multiplicity by $m_{n}=1 /\left(1-K_{n}\right)$ because we know $K_{n}$ from (2.5) and because (see [6]) $m=1 /(1-K)$. Both $K_{n}$ and $m_{n}$ are given in the tables. Results for the two methods reviewed in Section 1 -the $\delta^{2}$-process and the scheme of [7]-are also included for comparison. Extrapolation substeps have a $P$ (for prime) following the step number $n$ in the first column.

Table 1


Table 2

| f |  |  | $\alpha$ | m | K | L | M | N |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| $x \sin \left([x-1]^{4}\right)$ |  |  | 1 | 4 | 3/4 | 1/ | -5/64 | 25/256 |
|  | $\delta^{2}$-process | Ref.[7] | Proposed Method |  |  |  |  |  |
| n | $\varepsilon_{\mathrm{n}}$ | $\varepsilon_{n}$ | $\varepsilon_{\mathrm{n}}$ |  | $\begin{gathered} \text { Error } \\ \text { Estimate (2.3) } \end{gathered}$ |  | $\mathrm{K}_{\mathrm{n}}$ | $\mathrm{m}_{\mathrm{n}}$ |
| 0 | -. 500000 | -. 500000 | -. 500000 |  |  |  |  |  |
| 1 | -. 333043 | -. 333043 | -. 3 |  |  |  |  |  |
| 2 | -. 237900 |  |  |  |  |  |  |  |
| 2P | -. 111849 | -. 111849 | -. 1 |  |  |  |  |  |
| 3 | -. 829778(-1) |  |  |  |  |  |  |  |
| 3P |  | -. $154860(-1)$ |  | -1) |  |  |  |  |
| 4 | -. 617533(-1) |  |  |  |  |  | . 700391 | 3.33769 |
| 4P | -. $284075(-2)$ | -. $494979(-3)$ | -. 5 | -3) |  |  |  |  |
| 5 | -. $213006(-2)$ |  |  |  |  |  | . 743037 | 3.89160 |
| 5P |  | -. 194741(-5) |  | -6) |  |  |  |  |
| 6 | -. 159726(-2) |  |  |  |  |  | . 750969 | 4.01556 |
| 6 P | -. 152028(-5) | -. $241102(-9)$ | -. 1 | -11) | -. 1123 |  |  |  |

Table 3


Since Newton's method requires the calculation of both $f$ and $f^{\prime}$ to get $\phi$, the effective efficiency for the $\delta^{2}$-process is really only $2^{1 / 4} \doteq 1.189$, and for Method [7] is $(1.618)^{1 / 2} \doteq 1.272$. Similarly the proposed scheme with Newton's method has an effective efficiency of $(1.839)^{1 / 2} \doteq 1.356$.
b. The third example is for a direct application of the iteration $x_{n+1}=\phi\left(x_{n}\right)$ of (1.1) to finding fixed point $\alpha=1$ of

$$
\begin{equation*}
\phi=\frac{e^{x-1}+1}{2} . \tag{3.2}
\end{equation*}
$$

We can think of the problem as that of finding a root of the nonlinear function $f=x-\phi$.

Table 3 shows results for the proposed method: the actual error $\varepsilon_{n}$, the final error estimate from (2.3), and the approximation $K_{n}$ to $K$ ( $m_{n}$ has no apparent significance in this case). Again, corresponding calculations for the $\delta^{2}$-process and for the method of [7] are listed for comparison.
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[^0]:    Received June 24, 1982.
    1980 Mathematics Subject Classification. Primary 65B99, 65H05.
    Key words and phrases. Linear convergence, extrapolation, Aitken's $\delta^{2}$-process, Steffensen, AndersonBjörck, efficiency, nonlinear equation, order of convergence, Regula Falsi, linear sequence.
    *Work performed under the auspices of the U. S. Department of Energy.

