## Anderson-Björck for Linear Sequences\*

## By Richard F. King

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  $x_0$ ,

(1.1) 
$$x_{n+1} = \phi(x_n) \equiv \phi_n, \quad n = 0, 1, 2, \dots,$$

and the error  $\varepsilon_n = x_n - \alpha$  satisfies

(1.2) 
$$\varepsilon_{n+1} = K\varepsilon_n + L\varepsilon_n^2 + M\varepsilon_n^3 + N\varepsilon_n^4 + \cdots,$$

with 0 < |K| < 1. The error equation (1.2) for  $\phi$  we get by expanding  $\phi(x_n)$  in a Taylor series about  $\alpha$ , noting that  $\phi(\alpha) = \alpha$ , and setting  $K = \phi'(\alpha)$ ,  $L = \phi''(\alpha)/2!$ ,  $M = \phi'''(\alpha)/3!$ , and  $N = \phi^{iv}(\alpha)/4!$ . 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  $g'(\alpha) = 1 - \phi'(\alpha) = 1 - K \neq 0$ . Each step of the  $\delta^2$ -process goes as follows: from  $x_n$  take two linear-error substeps,  $x_{n+1} = \phi(x_n)$  and  $x_{n+2} = \phi(x_{n+1})$ . The extrapolation substep

Received June 24, 1982.

<sup>1980</sup> Mathematics Subject Classification. Primary 65B99, 65H05.

Key words and phrases. Linear convergence, extrapolation, Aitken's  $\delta^2$ -process, Steffensen, Anderson-Björck, efficiency, nonlinear equation, order of convergence, Regula Falsi, linear sequence.

<sup>\*</sup>Work performed under the auspices of the U. S. Department of Energy.

then consists of one application of the secant method

(1.3) 
$$\bar{x}_{n+2} = x_{n+1} - (x_n - x_{n+1}) \frac{g_{n+1}}{g_n - g_{n+1}}$$

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

(1.4) 
$$\bar{\varepsilon}_{n+2} = \frac{\varepsilon_n \varepsilon_{n+2} - \varepsilon_{n+1}^2}{\varepsilon_n - 2\varepsilon_{n+1} + \varepsilon_{n+2}} \cong \left(\frac{-LK}{1-K}\right) \varepsilon_n^2$$

for error  $\bar{\epsilon}_{n+2} = \bar{x}_{n+2} - \alpha$  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  $\phi_n = \phi(x_n)$ , 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

(1.5) 
$$x_{n+2} = x_{n+1} - (x_n - x_{n+1}) \frac{g_{n+1}}{g_n - g_{n+1}},$$

starting at  $g_0 \equiv g(x_0) = x_0 - \phi_0$  and  $g_1 \equiv g(x_1) = x_1 - \phi_1$ . Since ultimately  $|\varepsilon_{n+1}| \ll |\varepsilon_n|$ , 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  $\phi$ ):

(1.6) 
$$\varepsilon_{n+2} \cong \left(\frac{-LK}{1-K}\right) \varepsilon_n \varepsilon_{n+1}.$$

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  $(1 + \sqrt{5})/2 = 1.618$ . 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  $x - (x - \phi)/(1 - \phi')$ .

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  $(x_n, g_n)$ ,  $(x_{n+1}, g_{n+1})$ ,  $(x_{n+2}, g_{n+2})$  of a function g whose zero is sought, (ii) to compute the derivative  $\tilde{g}'_{n+2}$  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  $\tilde{g}'_{n+2}$ :

(2.1) 
$$x_{n+3} = x_{n+2} - \frac{g_{n+2}}{\tilde{g}'_{n+2}}.$$

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

(2.2) 
$$\tilde{g}'_{n+2} = g[x_{n+2}, x_{n+1}] + g[x_{n+2}, x_n] - g[x_n, x_{n+1}],$$

where  $g[x_j, x_k] = (g_k - g_j)/(x_k - x_j)$ .

In our procedure we take g to be the  $\phi$ -related function  $g = x - \phi$ . Starting from a given  $x_0$ , we (1) compute  $\phi(x_0) \equiv \phi_0$  and set  $g_0 = x_0 - \phi_0$ . Now (2) take  $x_1 = \phi_0$ , compute  $\phi(x_1) \equiv \phi_1$ , 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  $P_0 = (x_0, g_0)$  and  $P_1 = (x_1, g_1)$ . 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  $(x_0, g_0)$  and  $(x_1, g_1)$ , compute  $\phi(x_2) \equiv \phi_2$ , 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)  $(x_2, g_2) = (x_1, g_1), (x_1, g_1) = (x_0, g_0), (x_0, g_0) = (2x_1 - x_2, 2g_1 - g_2)$ , 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  $g'(\alpha) = 1 - \phi'(\alpha) = 1 - K \neq 0$ ). Furthermore,  $g''(\alpha) = -2L$ ,  $g'''(\alpha) = -6M$ , and  $g^{iv}(\alpha) = -24N$ . It can then readily be shown from (1.2), (2.1), and (2.2) that

(2.3) 
$$\varepsilon_{n+3} = \frac{M}{1-K}\varepsilon_n\varepsilon_{n+1}\varepsilon_{n+2} - \frac{L}{1-K}\varepsilon_{n+2}^2 + \frac{N}{1-K}\varepsilon_n^2\varepsilon_{n+1}\varepsilon_{n+2} + \cdots$$

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

(2.4) 
$$\varepsilon_{n+3} \cong \frac{M}{1-K} \varepsilon_n \varepsilon_{n+1} \varepsilon_{n+2}.$$

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

(2.5) 
$$K_{n+4} = \frac{\phi_{n+3} - \phi_{n+2}}{x_{n+3} - x_{n+2}}.$$

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

(2.6) 
$$K_{n+4} = \frac{K(\varepsilon_{n+3} - \varepsilon_{n+2}) + L(\varepsilon_{n+3}^2 - \varepsilon_{n+2}^2) + \cdots}{\varepsilon_{n+3} - \varepsilon_{n+2}} \doteq K + L(\varepsilon_{n+2} + \varepsilon_{n+3}).$$

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,

(3.1) 
$$x_{n+1} = \phi(x_n) = x_n - \frac{f_n}{f'_n},$$

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  $f = x \sin([x - 1]^4)$ , 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  $m_n = 1/(1 - K_n)$  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.

t			α	m	ĸ	լ ե	M	N		
	$(x-1)^{2} \tan(\pi x/4)$			2	1/2	π/8	$-\pi^2/32$	3π <sup>3</sup> /128		
	$\delta^2$ -process Ref.[7]			Proposed Method						
n	<sup>e</sup> n	<sup>e</sup> n	ε <sub>n</sub>		Error Estimate (2.3)		K n	m n		
0	500000	500000	500000							
1	.622531(-1)	.622531(-1)	.622531(-1)							
2 2P	.325841(-1) .340712(-1)	.340712(-1)	.340712(-1)							
3 3P	.174802(-1)	168097(-2)	.562	214(-3)						
4 4P	.885852(-2) 468967(-3)	.450433(-4)	816331(-6)				.535944	2.15491		
5 5P	234397(-3)	.594677(-7)	.838	173(-11)			.513264	2.05450		
6	117177(-3)	210270( 11)		((2( 20)	221557	( 20)	.500220	2.00088		
6P	863344(-7)	210378(-11)	.231	462(-20)	.231557	(-20)				

TABLE 1

T

T

Т

V I

Ł

f

f			α	m	К	L	M	N	
	x sin([x-1] <sup>4</sup> )			4	3/4	1/16	-5/64	25/256	
$\delta^2$ -process Ref.[7]			Proposed Method						
n	ε <sub>n</sub>	εn	ε <sub>n</sub>		Error Estimate	(2.3)	к <sub>n</sub>	m n	
0	500000	500000	500000						
1 2	333043 237900	333043	3330	943					
2P	111849	111849	1118	349					
3 3P	829778(-1)	154860(-1)	.1633	380(-1)					
4	617533(-1)		11				.700391	3.33769	
4P 5	284075(-2) 213006(-2)	494979(-3)	5201	23(-3)			.743037	3.89160	
5P	1213000( 2)	194741(-5)	4153	324(-6)				5009100	
6	159726(-2)				110051		.750969	4.01556	
6P	152028(-5)	241102(-9)	1118	322(-11)	112351	(-11)			

TABLE 2

TABLE 3

φ		f	a	К	L	M	N			
[exp(x-1)+1]/2		x-\$(x)	1 1/2		1/4	1/12	1/48			
	$\delta^2$ -process Ref.[7]			Proposed Method						
n	en e	Ref.[7] <sup>E</sup> n	ε <sub>n</sub>		Erron Estimate		Kn			
0 1 2	500000 196735 892957(-1)	500000 196735	5000 196							
2P 3	303500(-1) 149470(-1)	303500(-1)		500(-1)						
3P 4	741794(-2)	250417(-2)		119(-3)			.446848			
4P 5	218535(-3) 109255(-3)	369864(-4)		383(-6)			.492303			
5P 6 6P	546247(-4) 119348(-7)	462123(-7) 854588(-12)		159(-11) 176(-21)	499178	3(-21)	.499813			
51		1054500(12)	• • • • • •	1,0(-21)		Λ 21)				

Since Newton's method requires the calculation of both f and f' 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  $x_{n+1} = \phi(x_n)$  of (1.1) to finding fixed point  $\alpha = 1$  of

(3.2) 
$$\phi = \frac{e^{x-1}+1}{2}$$

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

## RICHARD F. KING

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.

Department of Energy Argonne National Laboratory Argonne, Illinois 60439

1. A. C. AITKEN, "On Bernoulli's numerical solution of algebraic equations," Proc. Roy. Soc. Edinburgh, v. 46, 1926, pp. 289-305.

2. N. ANDERSON & Å. BJÖRCK, "A new high order method of Regula Falsi type for computing a root of an equation," *BIT*, v. 13, 1973, pp. 253-264.

3. H. ESSER, "Eine stets quadratisch konvergente Modifikation des Steffensen-Verfahrens," Computing, v. 14, 1975, pp. 367–369.

4. A. S. HOUSEHOLDER, The Numerical Treatment of a Single Nonlinear Equation, McGraw-Hill, New York, 1970.

5. R. F. KING, "A secant method for multiple roots," BIT, v. 17, 1977, pp. 321-328.

6. R. F. KING, "An extrapolation method of order four for linear sequences," SIAM J. Numer. Anal., v. 16, 1979, pp. 719–725.

7. R. F. KING, "An efficient one-point extrapolation method for linear convergence," Math. Comp., v. 35, 1980, pp. 1285-1290.

8. D. E. MULLER, "A method for solving algebraic equations using an automatic computer," Math. Comp., v. 10, 1956, pp. 208-215.

9. A. M. OSTROWSKI, Solution of Equations and Systems of Equations, 2nd ed., Academic Press, New York, 1966.

10. J. F. STEFFENSEN, "Remarks on iteration," Skandinavisk Aktuarietidskrift, v. 16, 1933, pp. 64-72.

11. J. F. TRAUB, Iterative Methods for the Solution of Equations, Prentice-Hall, Englewood Cliffs, N. J., 1964.

12. H. VAN DE VEL, "A method for computing a root of a single nonlinear equation, including its multiplicity," *Computing*, v. 14, 1975, pp. 167–171.