An Error Formula for Numerical Differentiation

By

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1. Introduction

Suppose the function f(x) is approximated by the Lagrange interpolating polynomial $P_n(x)$ of degree *n* relative to the tabular points x_0, x_1, \ldots, x_n (arranged in increasing order). Assuming that f(x) is n+1 times continuously differentiable, it is well known that the error

(1.1)
$$e(x) = f(x) - P_n(x)$$

may be written in the form

(1.2)
$$e(x) = \frac{f^{(n+1)}(\xi)}{(n+1)!} w(x),$$

where ξ is some point (varying with x) which belongs to the smallest interval I containing x and all the tabular points x_i , while

(1.3)
$$w(x) = \prod_{i=0}^{n} (x - x_i)$$

is a polynomial of degree n+1 with roots at the tabular points. Since $P_n^{(k)}(x)$ is often used as an approximation to the k-th derivative $(1 \le k \le n)$ of f, it would be useful to have an equally simple formula for the corresponding error $e^{(k)}(x)$.

The earliest, and perhaps the closest, approach to this goal is found in STEFFENSEN [1], who shows that

(1.4)
$$e^{(k)}(x) = \frac{f^{(n+1)}(\xi)}{(n+1)!} w^{(k)}(x)$$

provided x lies outside the open interval (x_0, x_n) . For x in (x_0, x_n) , HILDE-BRAND [2] gives an expression for $e^{(k)}(x)$ involving derivatives of f up to the order n+1+k. An alternative derivation of this expression was recently indicated by RALSTON [3].

BRODSKII [4] produced a formula containing only (n + 1)-th derivatives and from it obtained sharp bounds on $|e^{(k)}|$ in terms of the maximum value M of $|f^{(n+1)}(x)|$ in I. However, his formula is far from simple. It involves a sum of n + 1 integrals whose integrands and intervals of integration all vary with x. BRODSKII also showed that Steffensen's formula (1.4) cannot hold for arbitrary x and f, and, more generally, that no formula can exist of the type

(1.5)
$$e^{(k)}(x) = f^{(n+1)}(\xi) \varphi(x),$$

where $\varphi(x)$ is for fixed x_0, \ldots, x_n a well-determined function of x.

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It is the purpose of this paper to present a simple formula for $e^{(k)}(x)$ which is "almost" of the forbidden type (1.5). The formula (Equation (2.1)) is stated and proved in section 2, while section 3 contains a discussion of some possible applications. In most cases, our formula leads to upper estimates on $|e^{(k)}|$ which are somewhat weaker, but far easier to obtain, than those derivable from BRODSKII's paper [4]. On the other hand, it provides better insight into the functional dependence of $e^{(k)}$ on x, and can in certain special situations produce smaller upper bounds than [4].

2. The Error Formula

Theorem. There exist n+1-k constants y_k , y_{k+1} , ..., y_n depending on the function f but not on x such that

(2.1)
$$e^{(k)}(x) = \frac{f^{(n+1)}(\xi)}{(n+1-k)!} Q(x),$$

where $\xi = \xi(x)$ belongs to I and Q(x) denotes the polynomial

(2.2)
$$Q(x) = \prod_{i=k}^{n} (x - y_i).$$

The constants y_i satisfy the inequalities

$$(2.3) x_{i-k} < y_i < x_i.$$

Proof. From (1.1) and the definition of the Lagrange interpolation polynomial, we have e(x) = 0 at the n + 1 points x_0, \ldots, x_n . Applying Rolle's theorem k times, we find that $e^{(k)}(x)$ must have at least n + 1 - k roots in (x_0, x_n) , and specifically one root y_i in each subinterval (x_{i-k}, x_i) , $i = k, \ldots, n$, with the y_i all distinct. Defining Q(x) by (2.2), we now set

(2.4)
$$g(x) = e^{(k)}(x) - \lambda Q(x)$$
,

with λ a constant to be determined. For $x = y_i$, (2.1) is trivially satisfied: both sides are zero. Suppose $x = \overline{x}$ is not equal to any of the y_i ; we wish to establish (2.1) for \overline{x} . Since $Q(\overline{x}) \neq 0$, we can find a λ such that $g(\overline{x}) = 0$. But also $g(y_i) = 0$ for $i = k, \ldots, n$. Hence g(x) has n + 2 - k distinct zeros in the interval *I*. Again by Rolle's theorem, $g^{(n+1-k)}(x)$ has a zero in *I*, say at ξ . But from (2.4) and (1.1),

$$0 = g^{(n+1-k)}(\xi) = f^{(n+1)}(\xi) - (n+1-k)! \lambda.$$

Solving for λ and substituting its value in the relation $g(\bar{x}) = 0$, we find $e^{(k)}(\bar{x}) = f^{(n+1)}(\xi) Q(\bar{x})/(n+1-k)!$, which is relation (2.1) for \bar{x} . Since \bar{x} is arbitrary, this concludes the proof.

3. Remarks and Applications

At first sight, formula (2.1) seems to belong to the class (1.5) with $\varphi(x) = Q(x)/(n+1-k)!$. This does not contradict Brodskii's conclusion that (1.5) is impossible, since the function Q(x) depends on the partially undetermined constants y_i . Despite this indeterminacy, the bounds (2.3) on the y_i permit the convenient use of (2.1) in obtaining upper bounds on the absolute error $|e^{(k)}|$.

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To begin with the crudest estimation procedure, note that x and all the y_i belong to the interval I. Hence $|Q(x)| \leq L^{n+1-k}$, where L denotes the length of I. [In the usual situation $x \in (a_0, a_n)$, we have $L = a_n - a_0$.] Denoting by M the maximum value of $|f^{(n+1)}(x)|$ in I, we obtain from (2.1) the uniform estimate

(3.1)
$$|e^{(k)}(x)| \leq M L^{n+1-k}/(n+1-k)!$$

This is exactly the same estimate that would be obtained by a similarly crude treatment of the Steffensen formula (1.4) if the latter were valid for $x \in (a_0, a_n)$. The derivation of (3.1) is the shortest means known to the author of proving that $e^{(k)} = O(ML^{n+1-k})$; its crudity lies only in the unnecessarily large constant factor.

We can considerably improve this constant by making more explicit use of inequality (2.3). We find

$$|x - y_i| \le \max(|x - x_{i-k}|, |x - x_i|)$$

and therefore $|Q(x)| \leq R(x)$, where

(3.2)
$$R(x) = \left| \frac{w(x)}{(x-z_1)\dots(x-z_k)} \right|.$$

Here w(x) is defined by (1.3), while z_1, \ldots, z_k denote those k among the n+1 tabular points x_0, \ldots, x_n which are nearest to x. (Note that R(x) is "piecewise" a polynomial.) Hence

(3.3)
$$|e^{(k)}(x)| \leq \frac{MR(x)}{(n+1-k)!}$$

We illustrate this formula by two examples involving the first derivative (k=1). For example A, we take n=1, $x_0=-L/2$, and $x_1=+L/2$. Then $R(x)=L/2+|x| \leq L$ for $x \in (x_0, x_1)$ and we find the uniform estimate $|e'| \leq ML$, coinciding with (3.1), and the particular estimates $|e'(0)| \leq (1/2) ML$ and $|e'(\pm L/4)| \leq (3/4) ML$. For example B, we choose n=2, $x_0=-L/2$, $x_1=0$, and $x_2=L/2$. Then

$$R(x) = \begin{cases} L^2/4 - x^2 & \text{for} \quad |x| \le L/4\\ L|x|/2 + x^2 & \text{for} \quad L/4 \le |x| \le L/2 \end{cases}$$

leading to the uniform estimate $|e'| \leq (1/4) ML^2$, half the size of (3.1), and the particular estimate $|e'(\pm L/4)| \leq (3/32) ML^2$. Though these estimates are not sharp, they are surprisingly close to the sharp bounds of BRODSKII [4], who gives the bound $|e'(x)| \leq M(L/4 + x^2/L)$ in example A and whose estimate leads in example B to the uniform bound $|e'| \leq (1/4) ML^2$ and the point estimate $|e'(\pm L/4)| \leq (1/96) ML^2$.

In addition to upper estimates, formula (2.1) yields useful information on the form of $e^{(k)}(x)$ as a function of x. For example, the error must be precisely zero at some n + 1 - k points of I, and, if $f^{(n+1)}$ does not change sign, the error will have exactly n + 1 - k changes of sign. If somehow, perhaps by a previous calculation or a fortuitous special circumstance, the points of zero error can be determined, (2.1) becomes almost an exact formula for $e^{(k)}$. Even if just one point x of small error (small compared to mL^{n+1-k} , where m is the minimum of $|f^{(n+1)}|$) is known, (2.1) can be used to pinpoint more closely the position of a neighboring y_i . Several separated points of small error will, of course, result in close fixes on several y_i , which in turn will allow a more accurate determination of $e^{(k)}(x)$ throughout $x_0 \leq x \leq x_n$.

Finally, (2.1) yields bounds sharper than any previously known on the minimum of $|e^{(k)}(x)|$ over two or more suitably chosen points x. We give two illustrations.

In example A, described above, consider $x = \pm L/4$. For each of these two points x separately, we can only conclude $|Q(x)| = |\pm L/4 - y_1| \le 3L/4$ and therefore, as above, $|e'| \le (3/4) ML$. But |Q(x)| cannot be as great as 3L/4 at both points, since y_1 , being a definite point for a fixed *f*, cannot be simultaneously both positive and negative. In particular, we must have |Q(x)| < L/4 for at least one of the points, whence (2.1) yields $|e'(x)| \le (1/4) ML$ at that point. This is smaller than Brodskii's bound for $|e'(\pm L/4)|$, which is (5/16) ML.

A similar procedure can be used in example *B* for the points x=1/4, 3/4 (where we have fixed L=2 for simplicity of notation). Here a division is made into the cases $0 < y_2 \le 1/2$ and $1/2 \le y_2 < 1$. In the first case, we find $|e'(1/4)| \le (5/32)M$, while in the second case $|e'(3/4)| \le (7/32)M$. The corresponding Brodskii bounds are 69/256 at x=1/4 and 143/256 at x=3/4. Naturally, an even smaller error value can be obtained by including the points -1/4 and -3/4 and distinguishing in addition the cases $y_1 \ge -1/2$.

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(Received August 6, 1963)