Lecture interp4: Error formula for polynomial interpolation

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Summary: Derivation of error formula and examples of its use.

References: T. Sauer, Numerical Analysis, Section 3.2, first edition, pages 153-159.

Error formula and its derivation

Suppose that we have a data set $D_N = \{(x_i, y_i)\}_{i=1}^N$, where we assume that $y_i = f(x_i)$ for i = 1, ..., N. That is, we assume that the y_i data values stem from evaluation of a known function f(x) on the x_i data points. Further, we assume that f is sufficiently smooth; namely, that its Nth derivative $f^{(N)}(x)$ is itself a continuous function on an open interval which contains $[x_1, x_N]$. For simplification, we also assume $x_1 < x_2 < \cdots < x_{N-1} < x_N$. For many common functions, such as $\sin x$ and e^x , all derivatives are continuous on all of the real line \mathbb{R} . The analysis to follow pertains to such functions, among others.

Theorem 1. Assume the conditions stated in the last paragraph, and let p(x) be the (degree N-1 or less) polynomial which interpolates the data set D_N . Then

$$f(x) - p(x) = (x - x_1)(x - x_2) \cdots (x - x_N) \frac{f^{(N)}(c)}{N!},$$
(1)

where $x \in [x_1, x_N]$ and also $c = c(x) \in [x_1, x_N]$.

Before starting the proof, first notice that the formula "looks right". Indeed, since p(x) interpolates f(x) at the nodal points x_i , we have $f(x_i) - p(x_i) = 0$ for all i = 1, ..., N. The right-hand side of (1) also vanishes at each x_i , since the factor $(x - x_i)$ present in right-hand side will of course vanish when $x = x_i$. The point c = c(x) in general depends on the evaluation point x. Were it a constant (a possibility), then $f(x) = p(x) + (x - x_1)(x - x_2) \cdots (x - x_N) f^{(N)}(c)/N!$ would itself be a polynomial (of degree N or less).

Finally, while (1) is a equality, from it we may derive a more useful inequality. Set

$$M_N = \max_{x_1 \le t \le x_N} |f^{(N)}(t)|.$$

Then, from (1) we obtain

$$|f(x) - p(x)| \le |(x - x_1)(x - x_2) \cdots (x - x_N)| \frac{M_N}{N!}.$$
 (2)

This error estimate is more useful in practice.

Proof. Now turn to the proof of **Theorem 1**. We start by writing down the Newton formula for the interpolating polynomial p(x),

$$p(x) = \sum_{k=1}^{N} f[x_1, \dots, x_k] \underbrace{\prod_{i=1}^{k-1} (x - x_i)}^{\text{basis function } \phi_i(x)}$$

$$= f(x_1) + f[x_1, x_2](x - x_1) + \dots + f[x_1, \dots, x_N](x - x_1) \cdots (x - x_{N-1}),$$
(3)

where, as noted before, the last Newton basis function $\phi_N(x)$ does not depend on the last data point x_N (it only depends on x_1 through x_{N-1}). Here is the trick. We now consider the addition of one more data point (t, f(t)) to D_N , giving a larger data set $D_{N+1} = \{(x_1, f(x_1)), \ldots, (x_N, f(x_N)), (t, f(t))\}$. The trick is that t is essentially arbitrary, except that we assume $t \in [x_1, x_N]$. By the recursive property of the Newton approach, the (degree n or less) polynomial which interpolates this larger data set is

$$q(x) = p(x) + f[x_1, \dots, x_N, t](x - x_1) \cdots (x - x_{N-1})(x - x_N). \tag{4}$$

By construction q(t) = f(t), since q(t) interpolates f at x = t. Whence we see that

$$f(t) = p(t) + f[x_1, \dots, x_N, t](t - x_1) \cdots (t - x_{N-1})(t - x_N).$$

In fact, the t here is a "dummy variable"; we could replace all t factors in the last formula with ξ , or s, or, indeed, x. Therefore, replacement of t with x yields

$$f(x) - p(x) = f[x_1, \dots, x_N, x](x - x_1) \cdots (x - x_{N-1})(x - x_N).$$

This is the result (1), provided that we can show $f[x_1, \ldots, x_N, x] = f^{(N)}(c(x))/N!$.

To finish the proof, rather than the boxed result we will show $f[x_1, \ldots, x_N, t] = f^{(N)}(c(t))/N$ instead, which is of course the same thing. To this end, refer back to (4) and define

$$h(x) := f(x) - q(x),$$

which, by construction, vanishes at x_1, x_2, \ldots, x_N, t . Moreover, the Nth derivative $h^{(N)}$ is certainly a continuous function on $[x_1, x_N]$, since we have assumed f is N times continuously differentiable on an open interval larger than $[x_1, x_N]$ and, as a polynomial, any derivative of q(x) of any order is a continuous function on all of \mathbb{R} (if q is differentiated more than N times, one gets the zero function which is plenty continuous). Now we repeatedly appeal to Rolle's Theorem (a particular instance of the Mean Value Theorem for Derivatives, a basic result from Calculus): If a real-valued function g is continuous on a closed interval [a, b], differentiable on the open interval (a, b), and g(a) = g(b), then there exists at least one ξ in the open interval (a, b) such that $g'(\xi) = 0$. Since h has N + 1 roots and is continuously differentiable, by Rolle's Theorem, its derivative h' will have N roots. Then h'' will have N - 1 roots. By induction, $h^{(N)}$ will have one root, call it c = c(t). That is, $h^{(N)}(c) = 0$, or

$$f^{(N)}(c) = q^{(N)}(c) = p^{(N)}(c) + N!f[x_1, \dots, x_N, t].$$

Since p is a polynomial of degree $\leq N-1$, $p^{(N)}(x)=0$ and $f^{(N)}(c)=qN!f[x_1,\ldots,x_N,t]$. \square

Examples

Example 1

Consider the data set $\mathcal{D}_3 = \{(0, \frac{1}{4}), (2, \frac{1}{6}), (4, \frac{1}{8})\}$. (a) Construct the divided-difference table associated with \mathcal{D}_3 , and then write down the associated degree-2 Newton interpolating polynomial p(x). (b) Using the formula from (a), estimate the error |f(3) - p(3)| in using p(3) to approximate f(3), where f(x) = 1/(x+4) gives the data \mathcal{D}_3 from (a). Compare with the exact error.

Solution. The divided difference table is as follows.

$$\begin{array}{c|cccc}
0 & \frac{1}{4} & & \\
 & -\frac{1}{24} & \\
2 & \frac{1}{6} & \frac{1}{192} \\
4 & \frac{1}{8} & &
\end{array}$$

From the table the interpolating polynomial is

$$p(x) = \frac{1}{4} - \frac{1}{24}x + \frac{1}{192}x(x-2).$$

Therefore, we see that

$$p(3) = \frac{1}{4} - \frac{3}{24} + \frac{3}{192} = \frac{9}{64} \simeq 0.14062$$

The exact error between p(3) and $f(3) = 1/(3+4) = \frac{1}{7}$ is $f(3) - p(3) = \frac{1}{7} - \frac{9}{64} = \frac{1}{448}$, and so

$$|f(3) - p(3)| = \frac{1}{448} \simeq 0.0022321.$$

The third derivative of f is $f'''(x) = -6/(x+4)^4$, and $M_3 = |f'''(0)| = \frac{3}{128}$ is a bound on |f'''(x)| over [0,4], indeed over $[0,\infty)$. So the estimate (2) becomes

$$|f(3) - p(3)| \le \frac{|(3-0)(3-2)(3-4)|}{3!} \cdot \frac{3}{128} \simeq 0.011719.$$

Note that the estimate for the error is a bit worse than the exact error: 0.011719 > 0.0022321.

Example 2

(Model for the sine function key on a digital calculator). Construct the polynomial p which interpolates $f(x) = \sin x$ at 4 equally spaced points on $[0, \frac{\pi}{2}]$, and then estimate the error in using p(1) as an approximation to $\sin(1)$.

Solution. The data set is $\mathcal{D}_4 = \{(0,0), (\frac{\pi}{6}, \frac{1}{2}), (\frac{\pi}{3}, \frac{\sqrt{3}}{2}), (\frac{\pi}{2}, 1)\}$, with the associated divided difference table as follows.

We then read-off the interpolating polynomial from the top of the divided difference table,

$$p(x) = 0 \cdot 1 + \frac{3}{\pi}(x - 0) + \frac{9(\sqrt{3} - 2)}{\pi^2}(x - 0)(x - \frac{\pi}{6}) + \frac{18(5 - 3\sqrt{3})}{\pi^3}(x - 0)(x - \frac{\pi}{6})(x - \frac{\pi}{3})$$
$$= \frac{3}{\pi}x + \frac{9(\sqrt{3} - 2)}{\pi^2}x(x - \frac{\pi}{6}) + \frac{18(5 - 3\sqrt{3})}{\pi^3}x(x - \frac{\pi}{6})(x - \frac{\pi}{3}).$$

Next, from this formula we calculate

$$p(1) = \frac{3}{\pi} + \frac{9(\sqrt{3} - 2)}{\pi^2} (1 - \frac{\pi}{6}) + \frac{18(5 - 3\sqrt{3})}{\pi^3} (1 - \frac{\pi}{6}) (1 - \frac{\pi}{3}) \simeq 0.841086.$$

Since the 4th derivative of the sine function is again the sine function, the M_n factor in (2) is $M_4 = 1$. Therefore, the error estimate becomes

$$|\sin(1) - p(1)| \le \frac{|(1-0)(1-\frac{\pi}{6})(1-\frac{\pi}{3})(1-\frac{\pi}{2})|}{4!} \simeq 5.3476e-04.$$

So the approximation p(1) is good to better than 3-digits in a relative sense: $|\sin(1) - p(1)|/|\sin(1)| \approx 6.3551e-04$.