

Barycentric Lagrange Interpolation

As discussed by Jean-Paul Berrut and Lloyd N. Trefethen (2004)



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Lagrange interpolation

Given a set D_n of $n + 1$ nodes x_n with corresponding values f_n , we aim to construct the polynomial that satisfies

$$p(x_j) = f_j, \quad j = 0, \dots, n$$

This data set can be interpolated by the Lagrange form of the interpolation polynomial [3]

$$p_{01\dots n}(x) = \sum_{j=0}^n l_j(x) f_j(x_j), \quad (1)$$

where

$$l_j(x) = \frac{\prod_{k=0, k \neq j}^n (x - x_k)}{\prod_{k=0, k \neq j}^n (x_j - x_k)} \quad (2)$$

One must note that there are several issues with Lagrange's formula:

1. Each evaluation of $p(x)$ requires $O(n^2)$ additions and multiplications.
2. Adding a new data pair (x_{n+1}, f_{n+1}) requires an entirely new computation of every l_j .
3. The computation can be numerically unstable.

This poster will focus on points one and two.

Improved Lagrange Formula

Note that the numerator of (2) can be written as

$$l(x) = (x - x_0)(x - x_1) \dots (x - x_n)$$

Define the *barycentric weights* by

$$w_j = \frac{1}{\prod_{k \neq j} (x_j - x_k)}, \quad j = 0, \dots, n \quad (3)$$

Now, l_j can be written as

$$l_j(x) = l(x) \sum_{j=0}^n \frac{w_j}{x - x_j} f_j$$

which yields the new form of (1):

$$p(x) = l(x) \sum_{j=0}^n \frac{w_j}{x - x_j} f_j \quad (4)$$

This improved formula now requires $O(n^2)$ floating point operations (flops) to calculate quantities independent of x . To evaluate p , only $O(n)$ flops are required.

In addition, this formula can now easily be updated with a new Data set (x_{n+1}, f_{n+1}) just by dividing each w_j from (3) by $(x_j - x_{n+1})$, and then computing w_j using (3). The addition of a new Data set therefore requires a mere $O(n)$ flops, instead of an entire recalculation of every l_j .

The Barycentric Formula

Equation (4) can still be written in an even nicer form. The interpolant of the constant function 1 is itself. Plugging 1 into (4) yields

$$1 = \sum_{j=0}^n l_j(x) = l(x) \sum_{j=0}^n \frac{w_j}{x - x_j}$$

Dividing (4) by the interpolation of 1, $l(x)$ cancels and gives the so-called *barycentric formula* for p :

$$p(x) = \frac{\sum_{j=0}^n \frac{w_j}{x - x_j} f_j}{\sum_{j=0}^n \frac{w_j}{x - x_j}} \quad (5)$$

The barycentric formula is still a Lagrange formula where the weights w_j can be updated with a new data pair (x_{n+1}, f_{n+1}) using $O(n)$ flops. In addition, there exist explicit formulas for the barycentric weights w_j when using equidistant or Chebyshev nodes, facilitating their computation:

1. Equidistant: $w_j = (-1)^j \binom{n}{j}$
2. Chebyshev: $w_j = (-1)^j \delta_j$, where $\delta_j = \frac{1}{2}$ when $j = 0$ or $j = n$ and $\delta_j = 1$ otherwise.

Runge Phenomenon

The use of equidistant nodes in any type of interpolation poses one big problem. For large N , the different weights w_j vary by exponentially large factors [1]. This makes polynomial interpolation with equidistant points *ill-conditioned*. Figures 1 and 2 demonstrate this problem. At the "edges" of the interpolation, the approximations do not converge with increasing N . As can be seen in the Figures, the maximum error eventually grows exponentially. It can be shown that the error can grow as fast as 2^N [4]. This phenomenon, which is more extreme than the Gibbs phenomenon, is called the *Runge phenomenon*.

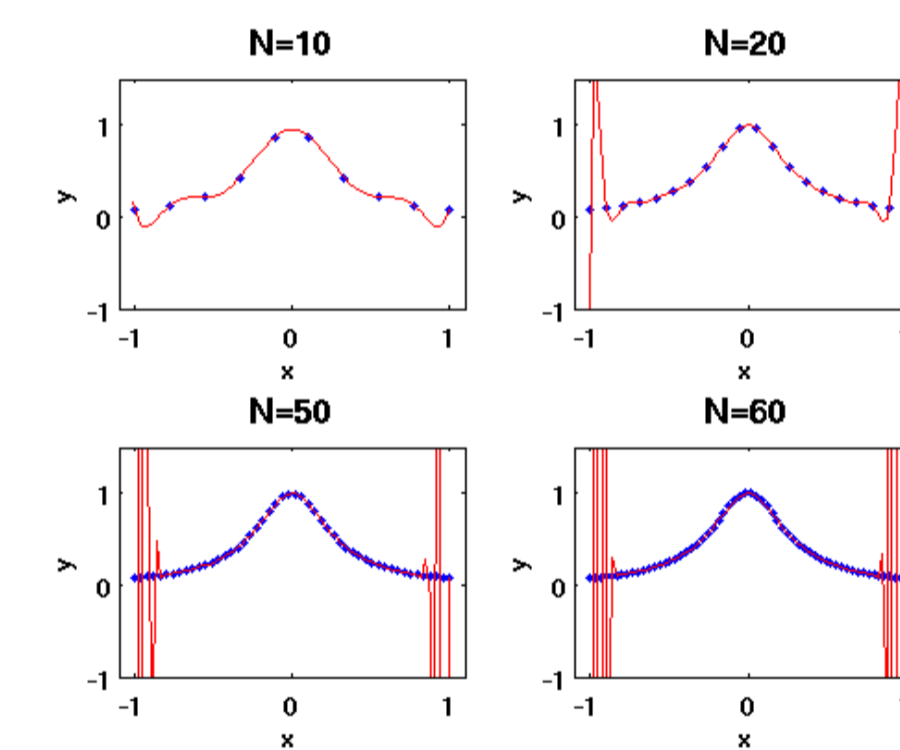


Figure 1: Barycentric Lagrange Interpolation with varying N

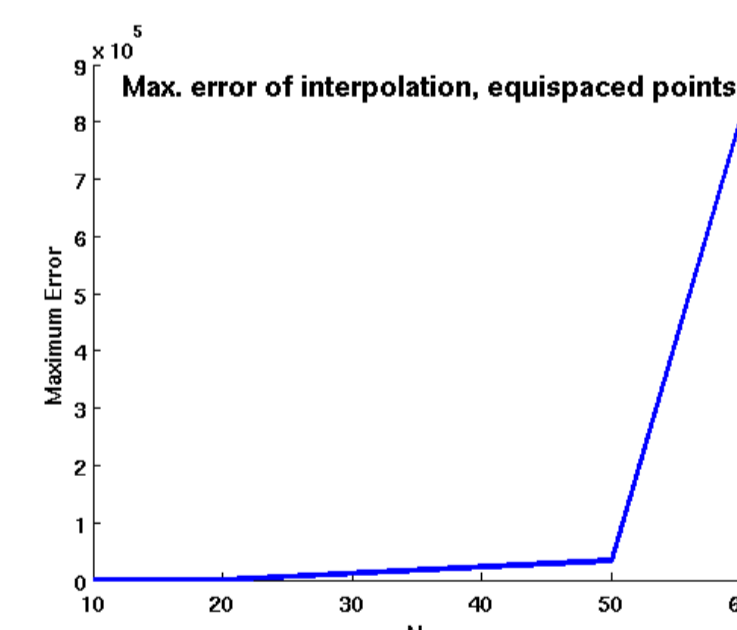


Figure 2: Maximum Error with increasing N

To overcome this problem, unevenly distributed interpolation points must be used. There are different types of such

points, but here we shall focus on the most common and simplest kind, the *Chebyshev points*. Figure 1 shows that the interpolation works well in the center, but fails at the edges. Chebyshev points, which have a density $\sim \frac{N}{\pi\sqrt{1-x^2}}$ can fix this issue (See [4] for more information).

Convergence Rates of Smooth Functions

Let f be analytic on an inside an ellipse in the complex plane with foci $[-1, 1]$ and axis lengths $2L$ and $2l$. When Chebyshev points are used, the interpolant converges exponentially as $N \rightarrow \infty$ [4]. In addition, the interpolants p_N satisfy the error estimate

$$\max_{x \in [-1, 1]} |f(x) - p_N(x)| \leq CK^{-N}$$

for some constants C and $K > 1$ [1, 2, p. 508, p. 173]. If the conditions above are satisfied, then $K = L + l$, and K denotes the convergence rate. Note that the convergence rate depends on the poles of f , and that a larger region of analyticity also results in a higher convergence rate [1]. Figure 3 shows the convergence of several functions with increasing N . We will also calculate their convergence rates:

1. $f(x) = \frac{\exp(x)}{\cos(x)}$: $K = \frac{\pi}{2} + \sqrt{\frac{\pi^2}{4} - 1} \approx 2.7822$.

Note that there is a mistake in [1], where the term under the square root is $\pi^2 - 1$.

2. $f(x) = \frac{1}{1+12x^2}$: $K = \frac{1}{\sqrt{12}} + \sqrt{\frac{13}{12}} \approx 1.3295$

3. $f(x) = \frac{1}{1+50x^2}$: $K = \frac{1}{\sqrt{50}} + \sqrt{\frac{51}{50}} \approx 1.1514$

4. $f(x) = \text{abs}(x)$ is not analytic on $[-1, 1]$ because it is not differentiable at $x = 0$. The interpolation does not seem to converge (at least exponentially).

5. $f(x) = \text{tan}(x)$: we would expect this function to have a very similar convergence rate as $\frac{\exp(x)}{\cos(x)}$, since it also has nodes at ± 1 . Indeed, Figure 3 shows this!

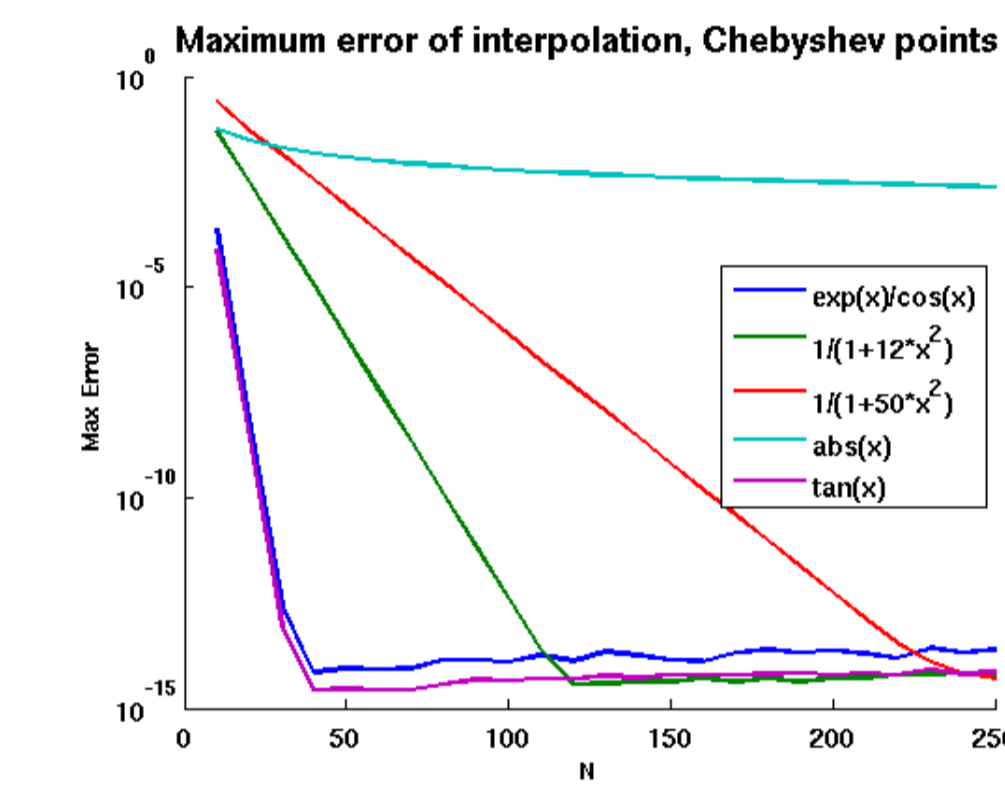


Figure 3: Convergence Rates of Different Functions

References

- [1] Jean-Paul Berrut and Lloyd N. Trefethen. Barycentric Lagrange Interpolation. *SIAM REVIEW*, 46(3):501–517, 2004.
- [2] Bengt Fornberg. *A Practical Guide to Pseudospectral Methods*. Cambridge University Press, 1996.
- [3] Nabil R. Nassif and Dolly K. Fayyad. *Numerical Analysis and Scientific Computing*. CRC Press, 2014.
- [4] Lloyd N. Trefethen. *Spectral Methods in Matlab*, chapter 5. Polynomial Interpolation and Clustered Grids, pages 41–50. SIAM, 2000.