

# Polynomial interpolation

---

Nicholas Hu · Last updated on 2026-01-15

## Lagrange interpolation

---

Suppose that  $f$  is a real-valued function defined on a set of *distinct nodes*  $\{x_j\}_{j=0}^n \subseteq \mathbb{R}$  and let  $P_n$  denote the vector space of real polynomial functions of degree *at most*  $n$ . Then there exists a unique  $p_n \in P_n$  that *interpolates*  $f$  at the nodes in the sense that  $p_n(x_j) = f(x_j)$  for each  $j$ :

$$p_n = \sum_j f(x_j) \ell_j,$$

where  $\ell_j := \prod_{i \neq j} \frac{\bullet - x_i}{x_j - x_i} \in P_n$ . Clearly,  $p_n$  interpolates  $f$  since  $\ell_j(x_i) = \delta_{ij}$ , and it is unique because if  $q_n \in P_n$  also interpolates  $f$  (at the same nodes), then  $p_n - q_n \in P_n$  has  $n + 1$  distinct zeroes and must therefore be the zero polynomial. The polynomials  $\{\ell_j\}_{j=0}^n$  constitute a basis of  $P_n$  called the **Lagrange basis**.

Defining the **barycentric weights**  $w_j := \frac{1}{\prod_{i \neq j} (x_j - x_i)}$ , which notably depend only on the nodes, we can express  $p_n$  as  $\omega \sum_j \frac{w_j f(x_j)}{\bullet - x_j}$  (except at the nodes – where the values of  $p_n$  are given), where  $\omega := \prod_i (\bullet - x_i)$  is the **nodal polynomial**. In particular, taking  $f = 1$  implies that  $\omega \sum_j \frac{w_j}{\bullet - x_j} = 1$ , whence we obtain the **barycentric formula**

$$p_n = \frac{\sum_j \frac{w_j f(x_j)}{\bullet - x_j}}{\sum_j \frac{w_j}{\bullet - x_j}}.$$

## Hermite interpolation

---

More generally, suppose that  $f$  is a real-valued function defined on a multiset of nodes  $\{x_j^{1+\mu_j}\}_{j=0}^m \subseteq \mathbb{R}$ , where  $x_j$  has multiplicity  $1 + \mu_j$  and  $f$  is  $\mu_j$  times differentiable at  $x_j$ . Then there exists a unique polynomial  $p_n \in P_n$ , where  $n = m + \sum_j \mu_j$ , that interpolates  $f$  at the nodes in the sense that  $p_n^{(k)}(x_j) = f^{(k)}(x_j)$  for each  $j$  and each  $0 \leq k \leq \mu_j$ .

Indeed, such a polynomial can be constructed recursively as follows: if  $m > 0$ , let  $p_- \in P_{n-1}$  interpolate  $f$  at  $\{x_j^{1+\mu_j}\}_{j=0}^m \setminus \{x_m^1\}$  and  $p_+ \in P_{n-1}$  interpolate  $f$  at  $\{x_j^{1+\mu_j}\}_{j=0}^m \setminus \{x_0^1\}$ , and define  $p_n \in P_n$  as

$$\begin{aligned} p_n &:= \frac{x_m - \bullet}{x_m - x_0} p_- + \frac{\bullet - x_0}{x_m - x_0} p_+ \\ &= p_- + \frac{\bullet - x_0}{x_m - x_0} (p_+ - p_-). \end{aligned}$$

Clearly,  $p_n(x_j) = f(x_j)$  for each  $j$  since  $p_-(x_j) = f(x_j) = p_+(x_j)$  for  $0 < j < m$ ,  $p_-(x_0) = f(x_0)$ , and  $p_+(x_m) = f(x_m)$ . Moreover, for  $k > 0$ , we have

$$p_n^{(k)} = p_-^{(k)} + \frac{\bullet - x_0}{x_m - x_0} (p_+ - p_-)^{(k)} + k \cdot \frac{1}{x_m - x_0} (p_+ - p_-)^{(k-1)},$$

so, similarly,  $p_n^{(k)}(x_j) = f^{(k)}(x_j)$  for each  $j$  and each  $0 < k \leq \mu_j$ . In the base case  $m = 0$ , we take  $p_n$  to be the  $\mu_0^{\text{th}}$  degree Taylor polynomial of  $f$  centred at  $x_0$ . Uniqueness follows as above, counting zeroes with multiplicity.

By specifying appropriate values for  $f$  and its derivatives at the nodes, we can construct  $h_{jk} \in P_n$  such that  $p_n = \sum_j \sum_k f^{(k)}(x_j)h_{jk}$  for all  $f$ . These polynomials  $\{h_{jk}\}$  constitute a basis of  $P_n$  called the **Hermite basis**. For instance, if  $\mu_j = 0$  for each  $j$ , then  $h_{j,0} = \ell_j$  and we recover the Lagrange basis functions; if instead  $m = 0$ , then  $h_{0,k} = \frac{1}{k!}(\bullet - x_0)^k$ , which is sometimes called a **Taylor basis** function.

## Newton interpolation

Newton interpolation recasts Lagrange/Hermite interpolation in a more explicit basis in which the coefficients of the interpolating polynomial can still be efficiently computed. Let  $t_0, \dots, t_n$  be an enumeration of the nodes  $\{x_j^{1+\mu_j}\}_{j=0}^m$  (with multiplicity, in any order). The **Newton basis**  $\{\omega_j\}_{j=0}^n$  of  $P_n$  is then defined as  $\omega_j := \prod_{i < j} (\bullet - t_i)$ .

To compute the coefficients of the interpolating polynomial  $p_n$  in this basis, we define the **divided difference**  $f[t_0, \dots, t_n]$  as the  $(\bullet)^n$  coefficient of  $p_n$  in the **monomial basis**  $\{(\bullet)^j\}_{j=0}^n$  (which for brevity we will refer to as the "leading coefficient" despite the fact that it may be zero). The coefficients of  $p_n$  in the Newton basis are then successive divided differences of  $f$ .

### Coefficients of interpolating polynomial in Newton basis

$$p_n = \sum_j f[t_0, \dots, t_j] \omega_j$$

*Proof.* Write  $p_n = \sum_j c_j \omega_j$ . For each  $j$ , the polynomial  $\sum_{k \leq j} c_k \omega_k \in P_j$  interpolates  $f$  at  $\{t_0, \dots, t_j\}$  since  $\omega_k(t_i) = 0$  (with multiplicity) for all  $i \leq j < k$ . Clearly, its leading coefficient is  $c_j$ , so by definition,  $c_j = f[t_0, \dots, t_j]$ . ■

From Lagrange interpolation, we obtain an explicit formula for divided differences when the  $t_j$  are *distinct* in terms of the barycentric weights:

$$f[t_0, \dots, t_n] = \sum_j w_j f(t_j).$$

More generally, the recursive construction in Hermite interpolation shows that divided differences obey a recurrence relation.

### Recurrence relation for divided differences

Suppose that the nodes are ordered such that  $t_0 = t_n$  implies that  $t_0 = \dots = t_n$ . Then

$$f[t_0, \dots, t_n] = \begin{cases} \frac{f[t_1, \dots, t_n] - f[t_0, \dots, t_{n-1}]}{t_n - t_0} & \text{if } t_0 \neq t_n, \\ \frac{f^{(n)}(t_0)}{n!} & \text{if } t_0 = t_n. \end{cases}$$

*Proof.* This follows immediately from the construction above: namely, if  $t_0 \neq t_n$ , then  $p_n = p_- + \frac{\bullet - t_0}{t_n - t_0} (p_+ - p_-)$ , where  $p_- \in P_{n-1}$  interpolates  $f$  at  $\{t_0, \dots, t_{n-1}\}$  and  $p_+ \in P_{n-1}$  interpolates  $f$  at  $\{t_1, \dots, t_n\}$ ; otherwise, if  $t_0 = t_n$ , then  $p_n$  is the  $n^{\text{th}}$  degree Taylor polynomial of  $f$  centred at  $t_0$ . ■

This also yields a recursive algorithm for evaluating  $p_n$  known as **Neville's algorithm**. To wit, suppose that the nodes are ordered such that  $t_i = t_j$  for  $i < j$  implies that  $t_i = \dots = t_j$ , and let  $p_{i,j} \in P_{j-i}$  interpolate  $f$  at  $\{t_i, \dots, t_j\}$  so that  $p_n = p_{0,n}$ . Then

$$p_{i,j}(t) = \begin{cases} \frac{(t - t_i)p_{i+1,j}(t) - (t - t_j)p_{i,j-1}(t)}{t_j - t_i} & \text{if } t_i \neq t_j, \\ \sum_{k=0}^{j-i} \frac{f^{(k)}(t_i)}{k!} (t - t_i)^k & \text{if } t_i = t_j. \end{cases}$$

### Properties of divided differences

- **(Linearity)** If  $\alpha, \beta \in \mathbb{R}$ , then  $(\alpha f + \beta g)[t_0, \dots, t_n] = \alpha \cdot f[t_0, \dots, t_n] + \beta \cdot g[t_0, \dots, t_n]$ .
- **(Symmetry)** If  $\sigma$  is a permutation of  $\{0, \dots, n\}$ , then  $f[t_0, \dots, t_n] = f[t_{\sigma(0)}, \dots, t_{\sigma(n)}]$ .
- **(Factor property)** If  $n \geq 1$ , then  $((\bullet - t_0)f)[t_0, \dots, t_n] = f[t_1, \dots, t_n]$ .

*Proof.*

- (Linearity) If  $p_f \in P_n$  and  $p_g \in P_n$  interpolate  $f$  and  $g$ , respectively, at  $\{t_0, \dots, t_n\}$ , then  $\alpha p_f + \beta p_g \in P_n$  interpolates  $\alpha f + \beta g$  at  $\{t_0, \dots, t_n\}$ .
- (Symmetry) This is immediate since the definition of the divided difference is independent of the ordering of the nodes.
- (Factor property) If  $p_f \in P_{n-1}$  interpolates  $f$  at  $\{t_1, \dots, t_n\}$ , then  $(\bullet - t_0)p_f \in P_n$  interpolates  $(\bullet - t_0)f$  at  $\{t_0, \dots, t_n\}$ . ■

In fact, the recurrence relation, excluding the base case, can be derived solely from these three properties:

$$\begin{aligned} (t_n - t_0) \cdot f[t_0, \dots, t_n] &= (((\bullet - t_0) - (\bullet - t_n))f)[t_0, \dots, t_n] \\ &= ((\bullet - t_0)f)[t_0, \dots, t_n] - ((\bullet - t_n)f)[t_0, \dots, t_n] \\ &= f[t_1, \dots, t_n] - f[t_0, \dots, t_{n-1}] \end{aligned}$$

Thus, these properties along with the property  $f[t_0, \dots, t_n] = \frac{f^{(n)}(t_0)}{n!}$  when  $t_0 = \dots = t_n$  characterize divided differences.

The identity  $f[t_0, t_1] = \frac{f(t_1) - f(t_0)}{t_1 - t_0}$  for  $t_0 \neq t_1$  suggests another relationship between divided differences and derivatives: if, say,  $t_0 < t_1$ ,  $f$  is continuous on  $[t_0, t_1]$ , and  $f'$  exists on  $(t_0, t_1)$ , then the mean value theorem amounts to the assertion that there exists a  $\xi \in (t_0, t_1)$  such that  $f[t_0, t_1] = f'(\xi)$ . This generalizes readily to divided differences and derivatives of higher order.

### Mean value theorem for divided differences

Suppose that  $a := \min \{t_j\}_{j=0}^n < \max \{t_j\}_{j=0}^n =: b$ . If  $f^{(n-1)}$  is continuous on  $[a, b]$  and  $f^{(n)}$  exists on  $(a, b)$ , then there exists a  $\xi \in (a, b)$  such that

$$f[t_0, \dots, t_n] = \frac{f^{(n)}(\xi)}{n!}.$$

*Proof.* Let  $p_n \in P_n$  interpolate  $f$  at  $\{t_0, \dots, t_n\}$ . Then  $f - p_n$  has  $n + 1$  zeroes in  $[a, b]$  (with multiplicity), so by repeated applications of Rolle's theorem,  $(f - p)^{(n)} = f^{(n)} - f[t_0, \dots, t_n]n!$  has a zero  $\xi \in (a, b)$ . ■

As a consequence, we can express the error in polynomial interpolation, which for  $t_0 = \dots = t_n$  reduces to the statement of Taylor's theorem.

### Polynomial interpolation error

Suppose that  $a := \min \{t_j\}_{j=0}^n \cup \{t\} < \max \{t_j\}_{j=0}^n \cup \{t\} =: b$ . If  $f^{(n)}$  is continuous on  $[a, b]$  and  $f^{(n+1)}$  exists on  $(a, b)$ , then there exists a  $\xi \in (a, b)$  such that

$$f(t) - p_n(t) = \frac{f^{(n+1)}(\xi)}{(n+1)!} \omega_{n+1}(t) = \frac{f^{(n+1)}(\xi)}{(n+1)!} \prod_j (t - t_j).$$

*Proof.* Let  $p_n \in P_n$  interpolate  $f$  at  $\{t_0, \dots, t_n\}$ . Then  $p_n + f[t_0, \dots, t_n, t]\omega_{n+1}$  interpolates  $f$  at  $\{t_0, \dots, t_n, t\}$ , so the result follows from the mean value theorem for divided differences. ■

We also have the identity  $(fg)[t_0, t_1] = f[t_0] \cdot g[t_0, t_1] + f[t_0, t_1] \cdot g[t_1]$ , where the case  $t_0 = t_1$  is the product rule for derivatives (which also follows from taking  $t_1 \rightarrow t_0$  in the case  $t_0 \neq t_1$ ). More generally, we have the following identity, which for  $t_0 = \dots = t_n$  reduces to the generalized product rule for derivatives  $(fg)^{(n)} = \sum_j \binom{n}{j} f^{(j)} g^{(n-j)}$ .

### Product rule for divided differences

$$(fg)[t_0, \dots, t_n] = \sum_j f[t_0, \dots, t_j] \cdot g[t_j, \dots, t_n]$$

*Proof.* Let  $p_n \in P_n$  interpolate  $f$  at  $\{t_0, \dots, t_n\}$ . Then  $(fg)[t_0, \dots, t_n] = (p_n g)[t_0, \dots, t_n]$  since  $fg$  agrees with  $p_n g$  on  $\{t_0, \dots, t_n\}$ . By linearity and the factor property, we have

$$\begin{aligned} (p_n g)[t_0, \dots, t_n] &= \left( \sum_j f[t_0, \dots, t_j] \omega_{jg} \right) [t_0, \dots, t_n] \\ &= \sum_j f[t_0, \dots, t_j] \cdot (\omega_{jg}) [t_0, \dots, t_n] \\ &= \sum_j f[t_0, \dots, t_j] \cdot g[t_j, \dots, t_n]. \quad \blacksquare \end{aligned}$$

Furthermore, from the recurrence relation for divided differences, we see that if  $f \in C^0$ , then  $f[t_0, \dots, t_n]$  is jointly continuous in  $t_0, \dots, t_n$  wherever they are distinct; if  $f \in C^n$ , the mean value theorem for divided differences implies that it is jointly continuous everywhere.