

# 1 Hermite Interpolation (Section 6.3)

In Hermite interpolation, the derivatives are included in this interpolation condition. This is different from Lagrange interpolation, where derivatives aren't used.

Now, let's suppose we have  $x_0, x_1$  with interpolation conditions such that

$$P(x_i) = f(x_i), \quad P'(x_i) = f'(x_i), \quad i = 0, 1.$$

There are four<sup>1</sup> conditions in total. Thus, we want to seek polynomial of degree at most 3; in other words, we want to find

$$P(x) = a + b(x - x_0) + c(x - x_0)^2 + d(x - x_0)^2(x - x_1).$$

Finding the derivative,

$$P'(x) = b + 2c(x - x_0) + 2d(x - x_0)(x - x_1) + d(x - x_0)^2.$$

Then,

$$P(x_0) = a = f(x_0).$$

$$P'(x_0) = b = f'(x_0).$$

$$P(x_1) = a + b(x_1 - x_0) + c(x_1 - x_0)^2 = f(x_1).$$

$$P'(x_1) = b + 2c(x_1 - x_0) + d(x_1 - x_0)^2 = f'(x_1).$$

To simplify things, let  $h = x_1 - x_0$ . Then, we can write out a matrix like so

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 1 & h & h^2 & 0 \\ 0 & 1 & 2h & h^2 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} = \begin{bmatrix} f(x_0) \\ f'(x_0) \\ f(x_1) \\ f'(x_1) \end{bmatrix}.$$

In general, the system for solving these coefficients with derivative data does not have to be nonsingular.

## 1.1 Hermite Interpolation Conditions

We need a condition to ensure that the polynomials can be found. In particular, this condition is related to the highest derivative that is included in the interpolation problem. So, for example, if at  $x_0$  we want to interpolate at the third derivative, then the problem must also include the 0th, 1st, and 2nd derivatives. This can happen for a set of different  $x$ -values. Thus, the condition given is

$$P^{(j)}(x_i) = \underbrace{c_{ij}}_{\text{Data}} \quad (0 \leq j \leq k_i - 1, 0 \leq i \leq m).$$

Here,  $k_i$  is the number of prescribed interpolatory conditions associated with the node  $x_i$ . So, the total condition is that

$$k_0 + k_1 + \dots + k_m \equiv m + 1.$$

So, the interpolating polynomial  $P$  is of degree *at most*  $m$ .

(Example.) Suppose  $m = 1$ ,  $k_0 = 2$ ,  $k_1 = 2$ . Then,  $m = 3$  since  $k_0 + k_1 = 4 = 3 + 1 = m + 1$ . With a polynomial of degree  $m = 3$ , for each  $i$ , the data that we need to have to get an interpolating polynomial is:

- $i = 0$ :  $c_{00}, c_{01}$
- $i = 1$ :  $c_{10}, c_{11}$

<sup>1</sup>Note that  $i = 0$ ,  $i = 1$  are two conditions.

(Example.) Suppose we want to use Hermite Interpolation with just one point,  $x_0$ . Then,  $i$  will be fixed at 0 but we can have  $k$  derivatives. Thus,  $p^{(j)}(x_0) = c_{0j}$  (where  $c_{0j}$  is data that's given) for  $0 \leq j \leq k$ . The degree  $k$  polynomial can fit this data:

$$P(x) = a_0 + a_1(x - x_0) + a_2(x - x_0)^2 + \dots + a_k(x - x_0)^k.$$

Using  $x = x_0$ , we have

$$\begin{aligned} P(x_0) &= a_0 = c_{00} \\ P'(x_0) &= a_1 = c_{01} \\ &\vdots \\ P^{(k)}(x_0) &= k!a_k = c_{0k}. \end{aligned}$$

So, in particular, the coefficients are  $a_0 = c_{00}$ ,  $a_1 = c_{01}/1$ , and so on until  $a_k = c_{0k}/k!$ . Now,

$$P(x) = c_{00} + c_{01}(x - x_0) + \frac{c_{02}}{2!}(x - x_0)^2 + \dots + \frac{c_{0k}}{k!}(x - x_0)^k.$$

## 1.2 Extended Divided Differences

We can account for repeated nodes,  $x_i$ , by creating the extended divided difference table. As is the case, the coefficients for each term of the interpolating polynomial will be in the first row of the table.

(Example.) Suppose we are *given* three data points,

$$P(x_0) = c_{00}, \quad P'(x_0) = c_{01}, \quad P(x_1) = c_{10}.$$

We can create a divided difference table, shown below:

x	P(x)	P'(x)
x <sub>0</sub>	c <sub>00</sub>	c <sub>01</sub>
x <sub>0</sub>	c <sub>00</sub>	
x <sub>1</sub>	c <sub>10</sub>	

Diagram illustrating the construction of the extended divided difference table. The table has columns for x, P(x), and P'(x). The first row contains x<sub>0</sub>, c<sub>00</sub>, and c<sub>01</sub>. The second row contains x<sub>0</sub>, c<sub>00</sub>, and is empty. The third row contains x<sub>1</sub>, c<sub>10</sub>, and is empty. Arrows indicate the calculation of divided differences: a diagonal arrow from the first c<sub>00</sub> to the second c<sub>00</sub> leads to a yellow box labeled P[x<sub>0</sub>, x<sub>0</sub>]. Another diagonal arrow from the second c<sub>00</sub> to c<sub>10</sub> leads to an orange box labeled P[x<sub>0</sub>, x<sub>0</sub>, x<sub>1</sub>].

Then, we find that

$$\begin{aligned} P[x_0, x_1] &= \frac{P[x_1] - P[x_0]}{x_1 - x_0} = \frac{c_{10} - c_{00}}{x_1 - x_0}. \\ P[x_0, x_0, x_1] &= \frac{P[x_0, x_1] - P[x_0, x_0]}{x_1 - x_0} = \frac{\frac{c_{10} - c_{00}}{x_1 - x_0} - c_{01}}{x_1 - x_0}. \end{aligned}$$

Note that, for the denominator, we want the highest index minus the lowest index of the nodes. Then, we find that

$$P(x) = c_{00} + c_{01}(x - x_0) + P[x_0, x_0, x_1](x - x_0)^2.$$