#### COURSE 3.

### Newton interpolation polynomial

Newton interpolation polynomial is given by

$$(L_m f)(x) := (N_m f)(x) = f(x_0) + \sum_{i=1}^m (x - x_0) \dots (x - x_{i-1})(D^i f)(x_0)$$
(1)

$$= f(x_0) + \sum_{i=1}^{m} (x - x_0)...(x - x_{i-1})[x_0, ..., x_i; f],$$

where  $(D^i f)(x_0)$  (or denoted  $[x_0, ..., x_i; f]$ ) is the *i*-th order divided difference of the function f at  $x_0$ .

### Newton interpolation formula is

$$f = N_m f + R_m f,$$

where the remainder (the error) is given by

$$(R_m f)(x) = (x - x_0)...(x - x_m)[x, x_0, ..., x_m; f].$$
 (2)

**Remark 1** The remainder for Lagrange interpolation formula is also given by

$$(R_m f)(x) = \frac{(x - x_0)...(x - x_m)}{(m+1)!} f^{(m+1)}(\xi),$$

with  $\xi$  between  $x, x_0, ..., x_m$ , so, by (2), it follows that the divided differences are approximations of the derivatives

$$[x, x_0, ..., x_m; f] = \frac{f^{(m+1)}(\xi)}{(m+1)!}.$$

Remark 2 We notice that

$$(N_i f)(x) = (N_{i-1} f)(x) + (x - x_0)...(x - x_{i-1})[x_0, ..., x_i; f]$$

so the Newton polynomials of degree 2,3,..., can be iteratively generated, similarly to Aitken's algorithm.

**Example 3** Find  $L_2 f$  for  $f(x) = \sin \pi x$ , and  $x_0 = 0, x_1 = \frac{1}{6}, x_2 = \frac{1}{2}$ , in both forms.

**Sol.** a) We have  $u(x) = x(x - \frac{1}{6})(x - \frac{1}{2})$ ;  $u_0(x) = (x - \frac{1}{6})(x - \frac{1}{2})$ ;  $u_1(x) = x(x - \frac{1}{2})$ ;  $u_2(x) = x(x - \frac{1}{6})$ 

$$(L_2 f)(x) = \sum_{i=0}^{2} l_i(x) f(x_i) = \sum_{i=0}^{2} \frac{u_i(x)}{u_i(x_i)} f(x_i)$$

$$= \frac{(x - \frac{1}{6})(x - \frac{1}{2})}{(-\frac{1}{6})(-\frac{1}{2})} 0 + \frac{x(x - \frac{1}{2})}{\frac{1}{6}(-\frac{1}{3})} \frac{1}{2} + \frac{x(x - \frac{1}{6})}{\frac{1}{2} \cdot \frac{1}{3}} 1$$

$$= -3x^2 + \frac{7}{2}x.$$

b)

$$(N_2 f)(x) = f(0) + \sum_{i=1}^{2} (x - x_0) \dots (x - x_{i-1}) (D^i f)(x_0)$$
  
=  $f(0) + (x - x_0) (Df)(x_0) + (x - x_0) (x - x_1) (D^2 f)(x_0)$   
=  $x(Df)(x_0) + x(x - \frac{1}{6}) (D^2 f)(x_0)$ 

The table of divided differences:

SO

$$(N_2 f)(x) = 3x - 3x(x - \frac{1}{6}) = -3x^2 + \frac{7}{2}x.$$

# 2.3. Hermite interpolation

**Example 4** In the following table there are some data regarding a moving car. We may estimate the position (and the speed) of the car when the time is t = 10 using Hermite interpolation.

Let  $x_k \in [a,b], \ k = 0,1,...,m$  be such that  $x_i \neq x_j$ , for  $i \neq j$  and let  $r_k \in \mathbb{N}, \ k = 0,1,...,m$ . Consider  $f:[a,b] \to \mathbb{R}$  such that there exist  $f^{(j)}(x_k), \ k = 0,1,...,m; \ j = 0,1,...,r_k$  and  $n = m + r_0 + ... + r_m$ .

The Hermite interpolation problem (HIP) consists in determining the polynomial P of the smallest degree for which

$$P^{(j)}(x_k) = f^{(j)}(x_k), \quad k = 0, ..., m; \ j = 0, ..., r_k.$$

**Definition 5** A solution of (HIP) is called **Hermite interpolation polynomial**, denoted by  $H_nf$ .

**Hermite interpolation polynomial**,  $H_nf$ , satisfies the interpolation conditions:

$$(H_n f)^{(j)}(x_k) = f^{(j)}(x_k), \quad k = 0, ..., m; \ j = 0, ..., r_k.$$

Hermite interpolation polynomial is given by

$$(H_n f)(x) = \sum_{k=0}^{m} \sum_{j=0}^{r_k} h_{kj}(x) f^{(j)}(x_k) \in \mathbb{P}_n,$$
 (3)

where  $h_{kj}(x)$  denote the Hermite fundamental interpolation polynomials. They fulfill the relations:

$$h_{kj}^{(p)}(x_{\nu}) = 0, \quad \nu \neq k, \quad p = 0, 1, ..., r_{\nu}$$

$$h_{kj}^{(p)}(x_k) = \delta_{jp}, \quad p = 0, 1, ..., r_k, \quad \text{for } j = 0, 1, ..., r_k \text{ and } \nu, k = 0, 1, ..., m,$$
with  $\delta_{jp} = \begin{cases} 1, & j = p \\ 0, & j \neq p. \end{cases}$ 

We denote by

$$u(x) = \prod_{k=0}^{m} (x - x_k)^{r_k + 1}$$
 and  $u_k(x) = \frac{u(x)}{(x - x_k)^{r_k + 1}}$ .

We have

$$h_{kj}(x) = \frac{(x - x_k)^j}{j!} u_k(x) \sum_{\nu=0}^{r_k - j} \frac{(x - x_k)^{\nu}}{\nu!} \left[ \frac{1}{u_k(x)} \right]_{x = x_k}^{(\nu)}.$$
 (4)

**Example 6** Find the Hermite interpolation polynomial for a function f for which we know f(0) = 1, f'(0) = 2 and f(1) = -3 (equivalent with  $x_0 = 0$  multiple node of order 2 or double node,  $x_1 = 1$  simple node).

**Sol.** We have  $x_0 = 0, x_1 = 1, m = 1, r_0 = 1, r_1 = 0, n = m + r_0 + r_1 = 2$ 

$$(H_2f)(x) = \sum_{k=0}^{1} \sum_{j=0}^{r_k} h_{kj}(x) f^{(j)}(x_k)$$
  
=  $h_{00}(x) f(0) + h_{01}(x) f'(0) + h_{10}(x) f(1)$ .

We have  $h_{00}, h_{01}, h_{10}$ . These fulfills relations:

$$h_{kj}^{(p)}(x_{\nu}) = 0, \ \nu \neq k, \ p = 0, 1, ..., r_{\nu}$$
  
 $h_{kj}^{(p)}(x_k) = \delta_{jp}, \ p = 0, 1, ..., r_k, \quad \text{for } j = 0, 1, ..., r_k \text{ and } \nu, k = 0, 1, ..., m.$ 

We have  $h_{00}(x)=a_1x^2+b_1x+c_1\in\mathbb{P}_2$ , with  $a_1,b_1,c_1\in\mathbb{R}$ , and the system

$$\begin{cases} h_{00}(x_0) = 1 \\ h'_{00}(x_0) = 0 \\ h_{00}(x_1) = 0 \end{cases} \Leftrightarrow \begin{cases} h_{00}(0) = 1 \\ h'_{00}(0) = 0 \\ h_{00}(1) = 0 \end{cases}$$

that becomes

$$\begin{cases} c_1 = 1 \\ b_1 = 0 \\ a_1 + b_1 + c_1 = 0. \end{cases}$$

Solution is:  $a_1 = -1, b_1 = 0, c_1 = 1$  so  $h_{00}(x) = -x^2 + 1$ .

We have  $h_{01}(x) = a_2x^2 + b_2x + c_2 \in \mathbb{P}_2$ , with  $a_2, b_2, c_2 \in \mathbb{R}$ . The system is

$$\begin{cases} h_{01}(x_0) = 0 \\ h'_{01}(x_0) = 1 \\ h_{01}(x_1) = 0 \end{cases} \Leftrightarrow \begin{cases} h_{01}(0) = 0 \\ h'_{01}(0) = 1 \\ h_{01}(1) = 0 \end{cases}$$

and we get  $h_{01}(x) = -x^2 + x$ .

We have  $h_{10}(x) = a_3x^2 + b_3x + c_3 \in \mathbb{P}_2$ , with  $a_3, b_3, c_3 \in \mathbb{R}$ . The system is

$$\begin{cases} h_{10}(x_0) = 0 \\ h'_{10}(x_0) = 0 \\ h_{10}(x_1) = 1 \end{cases} \Leftrightarrow \begin{cases} h_{10}(0) = 0 \\ h'_{10}(0) = 0 \\ h_{10}(1) = 1 \end{cases}$$

and we get  $h_{10}(x) = x^2$ .

The Hermite polynomial is

$$(H_2f)(x) = -x^2 + 1 - 2x^2 + 2x - 3x^2 = -6x^2 + 2x + 1.$$

## The Hermite interpolation formula is

$$f = H_n f + R_n f,$$

where  $R_n f$  denotes the remainder term (the error).

**Theorem 7** If  $f \in C^n[\alpha, \beta]$  and  $f^{(n)}$  is derivable on  $(\alpha, \beta)$ , with  $\alpha = \min\{x, x_0, ..., x_m\}$  and  $\beta = \max\{x, x_0, ..., x_m\}$ , then there exists  $\xi \in (\alpha, \beta)$  such that

$$(R_n f)(x) = \frac{u(x)}{(n+1)!} f^{(n+1)}(\xi).$$
 (5)

Proof. Consider

$$F(z) = \begin{vmatrix} u(z) & (R_n f)(z) \\ u(x) & (R_n f)(x) \end{vmatrix}.$$

 $F \in C^n[\alpha, \beta]$  and there exists  $F^{(n+1)}$  on  $(\alpha, \beta)$ .

We have

$$F(x) = 0$$
,  $F^{(j)}(x_k) = 0$ ,  $k = 0, ..., m$ ;  $j = 0, ..., r_k$ ;

because

$$u(x) = \prod_{k=0}^{m} (x - x_k)^{r_k + 1} \Rightarrow u^{(j)}(x_k) = 0, \ j = 0, ..., r_k$$

and

$$(R_m f)^{(j)}(x_k) = f^{(j)}(x_k) - (H_n f)^{(j)}(x_k) = f^{(j)}(x_k) - f^{(j)}(x_k) = 0.$$

So, F and its derivatives have n+2 distinct zeros in  $(\alpha,\beta)$ . Applying successively Rolle's theorem it follows that F' has at least n+1 zeros in  $(\alpha,\beta) \Rightarrow ... \Rightarrow F^{(n+1)}$  has at least one zero  $\xi \in (\alpha,\beta)$ ,  $F^{(n+1)}(\xi) = 0$ .

We have

$$F^{(n+1)}(z) = \begin{vmatrix} u^{(n+1)}(z) & (R_n f)^{(n+1)}(z) \\ u(x) & (R_n f)(x) \end{vmatrix},$$

with 
$$u(z) = \prod_{k=0}^{m} (z - z_k)^{r_k + 1} \in \mathbb{P}_{n+1} \Rightarrow u^{(n+1)}(z) = (n+1)!$$
, and  $(R_n f)^{(n+1)}(z) = f^{(n+1)}(z) - (H_n f)^{(n+1)}(z) = f^{(n+1)}(z)$  (as,  $H_n f \in \mathbb{P}_n$ )

 $\mathbb{P}_n$ ). We get

$$F^{(n+1)}(\xi) = \begin{vmatrix} (n+1)! & f^{(n+1)}(\xi) \\ u(x) & (R_n f)(x) \end{vmatrix} = 0,$$

whence it follows (5).

Corolar 8 If  $f \in C^{n+1}[a,b]$  then

$$|(R_n f)(x)| \le \frac{|u(x)|}{(n+1)!} ||f^{(n+1)}||_{\infty}, \quad x \in [a,b]$$

where  $\|\cdot\|_{\infty}$  denotes the uniform norm  $(\|f\|_{\infty} = \max_{x \in [a,b]} |f(x)|)$ .

Remark 9 In case of m=0, i.e.,  $n=r_0$ , (HIP) becomes Taylor interpolation problem. Taylor interpolation polynomial is

$$(T_n f)(x) = \sum_{j=0}^n \frac{(x-x_0)^j}{j!} f^{(j)}(x_0).$$