

# Lagrange & Newton interpolation

In this section, we shall study the polynomial interpolation in the form of Lagrange and Newton. Given a sequence of  $(n + 1)$  data points and a function  $f$ , the aim is to determine an  $n$ -th degree polynomial which interpolates  $f$  at these points. We shall resort to the notion of divided differences.

## Interpolation

Given  $(n + 1)$  points  $\{(x_0, y_0), (x_1, y_1), \dots, (x_n, y_n)\}$ , the points defined by  $(x_i)_{0 \leq i \leq n}$  are called **points of interpolation**. The points defined by  $(y_i)_{0 \leq i \leq n}$  are the **values of interpolation**. To interpolate a function  $f$ , the values of interpolation are defined as follows:

$$y_i = f(x_i), \quad \forall i = 0, \dots, n.$$

## Lagrange interpolation polynomial

The purpose here is to determine the unique polynomial of degree  $n$ ,  $P_n$  which verifies

$$P_n(x_i) = f(x_i), \quad \forall i = 0, \dots, n.$$

The polynomial which meets this equality is Lagrange interpolation polynomial

$$P_n(x) = \sum_{j=0}^n l_j(x) f(x_j)$$

where the  $l_j$ 's are polynomials of degree  $n$  forming a basis of  $P_n$

$$l_j(x) = \prod_{i=0, i \neq j}^n \frac{x - x_i}{x_j - x_i} = \frac{x - x_0}{x_j - x_0} \dots \frac{x - x_{j-1}}{x_j - x_{j-1}} \frac{x - x_{j+1}}{x_j - x_{j+1}} \dots \frac{x - x_n}{x_j - x_n}$$

## Properties of Lagrange interpolation polynomial and Lagrange basis

They are the  $l_j$  polynomials which verify the following property:

$$l_j(x_i) = \delta_{ji} = \begin{cases} 1 & i = j \\ 0 & i \neq j \end{cases}, \quad \forall i = 0, \dots, n.$$

They form a basis of the vector space  $P_n$  of polynomials of degree at most equal to  $n$

$$\sum_{j=0}^n \alpha_j l_j(x) = 0$$

By setting:  $x = x_i$ , we obtain:

$$\sum_{j=0}^n \alpha_j l_j(x_i) = \sum_{j=0}^n \alpha_j \delta_{ji} = 0 \Rightarrow \alpha_i = 0$$

The set  $(l_j)_{0 \leq j \leq n}$  is linearly independent and consists of  $n + 1$  vectors. It is thus a basis of  $P_n$ .

Finally, we can easily see that:

$$P_n(x_i) = \sum_{j=0}^n l_j(x_i) f(x_j) = \sum_{j=0}^n \delta_{ji} f(x_j) = f(x_i)$$

## Example: computing Lagrange interpolation polynomials

Given a set of three data points  $\{(0, 1), (2, 5), (4, 17)\}$ , we shall determine the Lagrange interpolation polynomial of degree 2 which passes through these points.

First, we compute  $l_0, l_1$  and  $l_2$ :

$$l_0(x) = \frac{(x-2)(x-4)}{8}, \quad l_1(x) = -\frac{x(x-4)}{4}, \quad l_2(x) = \frac{x(x-2)}{8}$$

Lagrange interpolation polynomial is:

$$P_n = l_0(x) + 5l_1(x) + 17l_2(x) = 1 + x^2$$

## Scilab: computing Lagrange interpolation polynomial

The Scilab function `lagrange.sci` determines Lagrange interpolation polynomial.  $X$  encompasses the points of interpolation and  $Y$  the values of interpolation.  $P$  is the Lagrange interpolation polynomial.

### **lagrange.sci**

```
function[P]=lagrange(X,Y) //X nodes,Y values;P is the numerical Lagrange
polynomial interpolation
n=length(X); // n is the number of nodes. (n-1) is the degree
x=poly(0,"x");P=0;
for i=1:n, L=1;
    for j=[1:i-1,i+1:n] L=L*(x-X(j))/(X(i)-X(j));end
    P=P+L*Y(i);
end
endfunction
```

```
-->X=[0;2;4]; Y=[1;5;17]; P=lagrange(X,Y)
P = 1 + x^2
```

Such polynomials are not convenient, since numerically, it is difficult to deduce  $l_{j+1}$  from  $l_j$ . For this reason, we introduce Newton's interpolation polynomial.

## Newton's interpolation polynomial and Newton's basis properties

The polynomials of Newton's basis,  $e_j$ , are defined by:

$$e_j(x) = \prod_{i=0}^{j-1} (x - x_i) = (x - x_0)(x - x_1) \cdots (x - x_{j-1}), \quad j=1, \dots, n.$$

with the following convention:

$$e_0 = 1$$

Moreover

$$\begin{aligned} e_1 &= (x - x_0) \\ e_2 &= (x - x_0)(x - x_1) \\ e_3 &= (x - x_0)(x - x_1)(x - x_2) \\ &\vdots \\ e_n &= (x - x_0)(x - x_1) \cdots (x - x_{n-1}) \end{aligned}$$

The set of polynomials  $(e_j)_{0 \leq j \leq n}$  (Newton's basis) are a basis of  $P_n$ , the space of polynomials of degree at most equal to  $n$ . Indeed, they constitute an echelon-degree set of  $(n + 1)$  polynomials.

Newton's interpolation polynomial of degree  $n$  related to the subdivision  $\{(x_0, y_0), (x_1, y_1), \dots, (x_n, y_n)\}$  is:

$$P_n(x) = \sum_{j=0}^n \alpha_j e_j(x) = \alpha_0 + \alpha_1(x-x_0) + \alpha_2(x-x_0)(x-x_1) + \dots + \alpha_n(x-x_0)(x-x_1)\cdots(x-x_{n-1})$$

where

$$P_n(x_i) = f(x_i), \quad \forall i = 0, \dots, n.$$

We shall see how to determine the coefficients  $(\alpha_j)_{0 \leq j \leq n}$  in the following section entitled the **divided differences**.

## Divided differences

Newton's interpolation polynomial of degree  $n$ ,  $P_n(x)$ , evaluated at  $x_0$ , gives:

$$P_n(x_0) = \sum_{j=0}^n \alpha_j e_j(x_0) = \alpha_0 = f(x_0) = f[x_0]$$

Generally speaking, we write:

$$f[x_i] = f(x_i), \quad \forall i = 0, \dots, n$$

$f[x_0]$  is called a zero-order **divided difference**.

Newton's interpolation polynomial of degree  $n$ ,  $P_n(x)$ , evaluated at  $x_1$ , gives:

$$P_n(x_1) = \sum_{j=0}^n \alpha_j e_j(x_1) = \alpha_0 + \alpha_1(x_1 - x_0) = f[x_0] + \alpha_1(x_1 - x_0) = f[x_1]$$

Hence

$$\alpha_1 = \frac{f[x_1] - f[x_0]}{x_1 - x_0} = f[x_0, x_1]$$

$f[x_1, x_0]$  is called 1<sup>st</sup> -**order divided difference**.

Newton's interpolation polynomial of degree  $n$ ,  $P_n(x)$ , evaluated at  $x_2$ , gives:

$$\begin{aligned} P_n(x_2) &= \sum_{j=0}^n \alpha_j e_j(x_2) \\ &= \alpha_0 + \alpha_1(x_2 - x_0) + \alpha_2(x_2 - x_0)(x_2 - x_1) \\ &= f[x_0] + f[x_0, x_1](x_2 - x_0) + \alpha_2(x_2 - x_0)(x_2 - x_1) \\ &= f[x_2] \end{aligned}$$

Therefore:

$$\begin{aligned} \alpha_2(x_2 - x_0)(x_2 - x_1) &= f[x_2] - f[x_0] - f[x_0, x_1](x_2 - x_0) \\ \alpha_2 &= \frac{f[x_2] - f[x_0] - f[x_0, x_1](x_2 - x_0)}{(x_2 - x_0)(x_2 - x_1)} \\ \alpha_2 &= \frac{f[x_2] - f[x_0]}{(x_2 - x_0)(x_2 - x_1)} - \frac{f[x_0, x_1]}{x_2 - x_1} \\ \alpha_2 &= \frac{f[x_0, x_2] - f[x_0, x_1]}{x_2 - x_1} \end{aligned}$$

The following form is generally preferred:

$$\begin{aligned}
\alpha_2(x_2-x_0)(x_2-x_1) &= f[x_2]-f[x_0]-f[x_0,x_1](x_2-x_0) \\
\alpha_2(x_2-x_0)(x_2-x_1) &= f[x_2]-f[x_0]-f[x_0,x_1](x_2-x_0)-f[x_1]+f[x_1] \\
\alpha_2(x_2-x_0)(x_2-x_1) &= f[x_2]-f[x_1]+f[x_1]-f[x_0]-f[x_0,x_1](x_2-x_0) \\
\alpha_2(x_2-x_0)(x_2-x_1) &= f[x_2]-f[x_1]+(x_1-x_0)f[x_0,x_1]-f[x_0,x_1](x_2-x_0) \\
\alpha_2(x_2-x_0)(x_2-x_1) &= f[x_2]-f[x_1]+(x_1-x_2)f[x_0,x_1] \\
\alpha_2(x_2-x_0) &= \frac{f[x_2]-f[x_1]}{x_2-x_1}-f[x_0,x_1] \\
\alpha_2(x_2-x_0) &= f[x_1,x_2]-f[x_0,x_1]
\end{aligned}$$

Hence

$$\alpha_2 = \frac{f[x_1,x_2]-f[x_0,x_1]}{x_2-x_0} = f[x_0,x_1,x_2]$$

$f[x_0, x_1, x_2]$  is called **2<sup>nd</sup>-order divided difference**. By recurrence, we obtain:

$$\alpha_k = \frac{f[x_1, \dots, x_k]-f[x_0, \dots, x_{k-1}]}{x_k-x_0} = f[x_0, \dots, x_k]$$

$f[x_0, \dots, x_k]$  is thus called a  **$k^{\text{th}}$ -order divided difference**. In practice, when we want to determine the **3<sup>rd</sup>-order divided difference**  $f[x_0, x_1, x_2, x_3]$  for instance, we need the following quantities

$$\begin{array}{ccccccc}
x_0 & f[x_0] & & & & & \\
x_1 & f[x_1] & f[x_0, x_1] & & & & \\
x_2 & f[x_2] & f[x_1, x_2] & f[x_0, x_1, x_2] & & & \\
x_3 & f[x_3] & f[x_2, x_3] & f[x_1, x_2, x_3] & f[x_0, x_1, x_2, x_3] & & 
\end{array}$$

Hence

$$f[x_0, x_1, x_2, x_3] = \frac{f[x_1, x_2, x_3]-f[x_0, x_1, x_2]}{x_3-x_0}$$

**Properties.** Let  $E = \{0, 1, \dots, n\}$  and  $\sigma$  be a permutation of  $G(E)$ . Then

$$f[x_{\sigma(0)}, \dots, x_{\sigma(n)}] = f[x_0, \dots, x_n]$$

## Newton's interpolation polynomial of degree $n$

Newton's interpolation polynomial of degree  $n$  is obtained via the successive divided differences:

$$P_n(x) = f[x_0] + \sum_{j=1}^n f[x_0, \dots, x_j] e_j(x)$$

## An example of computing Newton's interpolation polynomial

Given a set of 3 data points  $\{(0, 1), (2, 5), (4, 17)\}$ , we shall determine Newton's interpolation polynomial of degree 2 which passes through these points.

$$\begin{aligned}
x_0=0 & \quad f[x_0]=1 \\
x_1=2 & \quad f[x_1]=5 \quad f[x_0, x_1] = \frac{5-1}{2-0} = 2 \\
x_2=4 & \quad f[x_2]=17 \quad f[x_1, x_2] = \frac{17-5}{4-2} = 6 \quad f[x_0, x_1, x_2] = \frac{6-2}{4-0} = 1
\end{aligned}$$

Consequently:

$$P_2(x) = f[x_0] + f[x_0, x_1]x + f[x_0, x_1, x_2]x(x-2) = 1 + 2x + x(x-2) = 1 + x^2$$

### **Scilab: computing Newton's interpolation polynomial**

Scilab function `newton.sci` determines Newton's interpolation polynomial.  $X$  contains the points of interpolation and  $Y$  the values of interpolation.  $P$  is Newton's interpolation polynomial computed by means of divided differences.

#### **newton.sci**

```
function[P]=newton(X,Y) //X nodes,Y values;P is the numerical
Newton polynomial
n=length(X); // n is the number of nodes. (n-1) is the degree
for j=2:n,
    for i=1:n-j+1,Y(i,j)=(Y(i+1,j-1)-Y(i,j-1))/(X(i+j-1)-X(i));end,
end,
x=poly(0,"x");
P=Y(1,n);
for i=2:n, P=P*(x-X(i))+Y(i,n-i+1); end
endfunction;
```

Therefore, we obtain:

```
-->X=[0;2;4]; Y=[1;5;17]; P=newton(X,Y)
P = 1 + x^2
```

$$f(x) = a_0 \cdot 1 + a_1(x-x_0) + a_2(x-x_0)(x-x_1) + a_3(x-x_0)(x-x_1)(x-x_2) + \dots + a_k(x-x_0)(x-x_1)\dots(x-x_{k-1}) + \dots + a_n(x-x_0)(x-x_1)\dots(x-x_{n-2})(x-x_{n-1})$$

