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

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

Lagrange interpolation polynomial

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

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

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

$$l_{j}(x) = \prod_{i=0, i\neq j}^{n} \frac{x - x_{i}}{x_{j} - x_{i}} = \frac{x - x_{0}}{x_{j} - x_{0}} \cdots \frac{x - x_{j-1}}{x_{j} - x_{j-1}} \frac{x - x_{j+1}}{x_{j} - x_{j+1}} \cdots \frac{x - x_{n}}{x_{j} - x_{n}}$$

Properties of Lagrange interpolation polynomial and Lagrange basis

They are the l_i 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:

$$\sum_{j=\cdot}^{n} \alpha_{j} l_{j}(x_{i}) = \sum_{j=\cdot}^{n} \alpha_{j} \delta_{ji} = \cdot \Rightarrow \alpha_{i} = \cdot$$

The set $(l_j)_{0 \le j \le n}$ is linearly independent and consists of n + 1 vectors. It is thus a basis of P_n . Finally, we can easily see that:

$$P_{n}(x_{i}) = \sum_{j=.}^{n} l_{j}(x_{i}) f(x_{i}) = \sum_{j=.}^{n} \delta_{ji} f(x_{i}) = f(x_{i})$$

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 :

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

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_i , 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

$$e_{1} = (x - x_{0})$$

$$e_{2} = (x - x_{0})(x - x_{1})$$

$$e_{3} = (x - x_{0})(x - x_{1})(x - x_{2})$$

$$\vdots$$

$$e_{n} = (x - x_{0})(x - x_{1})\cdots(x - x_{n-1})$$

The set of polynomials $(e_j)_{0 \le j \le n}$ (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 $\{(x_0, y_0), (x_1, y_1), ..., (x_n, y_n)\}$ 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

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

We shall see how to determine the coefficients $(\alpha_j)_{0 \le j \le n}$ 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:

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

 $f[x_0]$ 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 1st -order divided difference.

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

$$P_{n}(x_{2}) = \sum_{j=0}^{n} \alpha_{j} e_{j}(x_{2})$$

= $\alpha_{0} + \alpha_{1}(x_{2} - x_{0}) + \alpha_{2}(x_{2} - x_{0})(x_{2} - x_{1})$
= $f[x_{0}] + f[x_{0}, x_{1}](x_{2} - x_{0}) + \alpha_{2}(x_{2} - x_{0})(x_{2} - x_{1})$
= $f[x_{2}]$

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:

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 2nd-order divided difference. By recurrence, we obtain:

$$\alpha_{k} = \frac{f[x_{1}, \dots, x_{k}] - f[x_{0}, \dots, x_{k-1}]}{x_{k} - x_{0}} = f[x_{0}, \dots, x_{k}]$$

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

$$\begin{array}{l} x_0 \quad f[x_0] \\ x_1 \quad f[x_1] \quad f[x_0, x_1] \\ x_2 \quad f[x_2] \quad f[x_1, x_2] \quad f[x_0, x_1, x_2] \\ x_3 \quad f[x_3] \quad f[x_2, x_3] \quad f[x_1, x_2, x_3] \quad f[x_0, x_1, x_2, x_3] \end{array}$$

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 $E = \{0, 1, ..., n\}$ and σ be a permutation of $\mathcal{G}(E)$. Then

.

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

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.

$$x_{0}=0 \quad f[x_{0}]=1$$

$$x_{1}=2 \quad f[x_{1}]=5 \quad f[x_{0},x_{1}]=\frac{5-1}{2-0}=2$$

$$x_{2}=4 \quad f[x_{2}]=17 \quad f[x_{1},x_{2}]=\frac{17-5}{4-2}=6 \quad f[x_{0},x_{1},x_{2}]=\frac{6-2}{4-0}=1$$

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
```