

# Numerical Analysis

## Semester 4

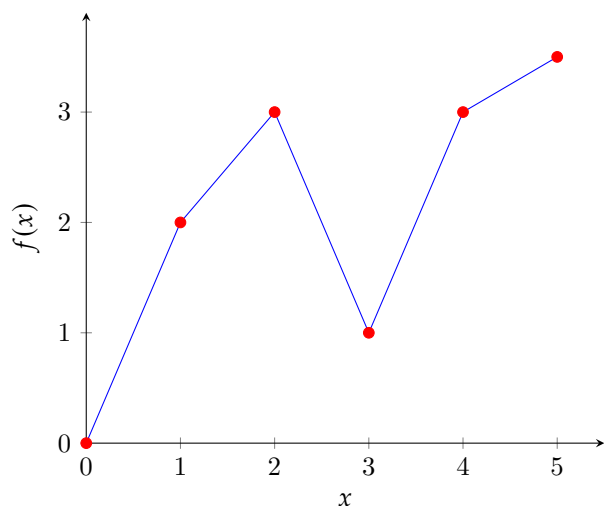
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# Chapter 1

## Interpolation

### 1.1 Linear Interpolation



$x$	$f(x)$
0	0
1	2
2	3
3	1
4	3
5	3.5

Linear interpolation is just drawing lines between the data points.

#### Definition 1.1.1: Linear Interpolation(lerp) equation

The equation of the lines between data points is

$$y = \frac{y_{i+1} - y_i}{x_{i+1} - x_i}(x - x_i) + y_i.$$

#### Theorem 1.1.1 Error due to linear interpolation

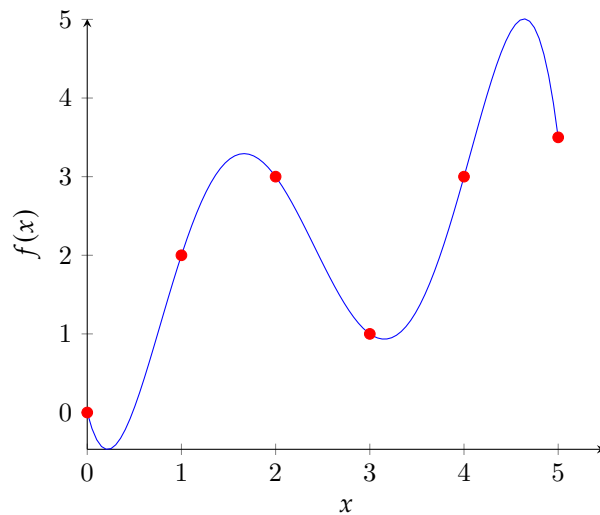
Let  $f$  be a continuous and differentiable on  $[a, b]$ . We define the error  $z(x)$  to be

$$|z(x)| \leq \frac{(b-a)^2}{8} \sup_{a \leq x \leq b} |f''(x)|.$$

### 1.2 Polynomial Interpolation

#### 1.2.1 Lagrange Polynomials

Really nice video [here](#) explaining Lagrange polynomials.



### Theorem 1.2.1 Lagrange polynomial equation

Consider a set of  $n$  points  $(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$ . The Lagrange polynomial for this set of data is

$$L(x) = \sum_{k=0}^n y_k \ell_k(x).$$

where

$$\ell_k(x) = \prod_{\substack{i=1 \\ i \neq k}}^n \frac{x - x_i}{x_k - x_i}.$$

### Case of equidistant points

If the set of  $x_i$  are equidistant from each other with a distance of  $h = x_{i+1} - x_i$ , then we can represent any point as  $x_k = x_0 + kh$  where  $k \in \mathbb{N}$  and any number  $x = x_0 + sh$  where  $s \in \mathbb{R}$ . We can rewrite the formula as

$$Q(s) = \sum_{k=0}^n \ell_k(s) f(x_k).$$

where

$$\ell_k(s) = \prod_{\substack{j=0 \\ j \neq k}}^n \frac{s - j}{k - j}.$$

by substitution

$$s = \frac{x - x_0}{h}.$$

### Existence

**Proof:**  $P(x)$  belongs to the vectorial space of polynomial of degree of, at most,  $n$ . Now, we must find a basis for this vectorial space. Find the polynomial  $\ell_k$  of degree  $\leq n$  such that

$$\ell_k(x_i) = \delta_{ki} = \begin{cases} 1 & \text{if } i = k \\ 0 & \text{if } i \neq k \end{cases}.$$

Then,  $\ell_k(x) = \lambda(x - x_0) \dots (x - x_{k-1})(x - x_{k+1}) \dots (x - x_n)$  where

$$\lambda = \frac{1}{(x_k - x_0) \dots (x_k - x_{k-1})(x_k - x_{k+1}) \dots (x_k - x_n)}.$$

The  $(n + 1)$  polynomials  $\ell_k(x)$  for a system of generators in the vectorial space of polynomials of degree at most  $n$ .

$$\lambda_0 \ell_0(x) + \lambda_1 \ell_1(x) + \cdots + \lambda_k \ell_k(x) + \cdots + \lambda_n \ell_n(x) = 0.$$

for  $x = x_k$

$$\lambda_0 \ell_0(x_k) + \lambda_1 \ell_1(x_k) + \cdots + \lambda_k \ell_k(x_k) + \cdots + \lambda_n \ell_n(x_k) = 0$$

$$0 + 0 + \cdots + \lambda_k 1 + \cdots + 0 = 0$$

$$\lambda_k = 0.$$

$\therefore$  the set of  $\ell_k$  for a basis in the vector space  $\Rightarrow$  there has to exist a polynomial passing through the given set of points. □

### Uniqueness

**Proof:** Let  $P$  and  $Q$  be 2 Lagrange polynomials of degrees  $\leq n$   $\left| P(x_i) = Q(x_i) = f(x_i) \quad \forall i = 0, 1, \dots, n. \right.$

Let

$$\left. \begin{array}{l} R = P - Q \text{ of degree } \leq n \\ R = 0 \text{ (n + 1) times} \end{array} \right\} R \equiv 0 \Rightarrow P = Q \quad \forall x.$$

□

## 1.2.2 Newton Polynomial

### Definition 1.2.1: Newton Polynomial equation

Consider a set of  $n$  points  $(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$ . The Newton polynomial for this set of data is

$$p_n(x) = \underbrace{a_0}_{A_0} + \underbrace{a_1(x - x_0)}_{A_1} + \underbrace{a_2(x - x_0)(x - x_1)}_{A_2} + \cdots + \underbrace{a_n \prod_{i=0}^{n-1} (x - x_i)}_{A_n}.$$

where

$$a_i = f[x_0, x_1, \dots, x_i].$$

Here  $f[\dots]$  is the divided difference of the inputted data.

### Definition 1.2.2: Backwards formula

$$P_n(x) = f_n + A_1 + A_2 + \cdots + A_n.$$

where

$$A_i = f[x_n, x_{n-1}, \dots, x_{n-i}] \prod_{j=n-i+1}^n (x - x_j).$$

The divided difference has 2 formulas, the recurrence formula

$$f[x_0, x_1, \dots, x_{n+1}] = \frac{f[x_1, x_2, \dots, x_{n+1}] - f[x_0, x_1, \dots, x_n]}{x_{n+1} - x_0}.$$

and a general formula

$$f[x_0, x_1, \dots, x_n] = \sum_{i=1}^n \frac{y_i}{\prod_{\substack{k=0 \\ k \neq i}}^n (x_i - x_k)}.$$

Now forget you ever saw those cause there is an easier method to finding the divided difference.

## Divided Difference Table

$x_0$	$y_0$	$\frac{y_1-y_0}{x_1-x_0} = f[x_0, x_1]$		
$x_1$	$y_1$	$\frac{y_2-y_1}{x_2-x_1} = f[x_1, x_2]$	$\frac{f[x_1, x_2] - f[x_0, x_1]}{x_2 - x_0}$	$\dots$
$x_2$	$y_2$	$\frac{y_3-y_2}{x_3-x_2} = f[x_2, x_3]$	$\frac{f[x_2, x_3] - f[x_1, x_2]}{x_3 - x_1}$	$\dots$
$x_3$	$y_3$	$\frac{y_4-y_3}{x_4-x_3} = f[x_3, x_4]$	$\frac{f[x_3, x_4] - f[x_2, x_3]}{x_4 - x_2}$	$\dots$
$x_4$	$y_4$			

After we have constructed the table we can find the divided difference we want by looking at the top diagonal

$x_0$	$y_0$	$f[x_0, x_1]$			
$x_1$	$y_1$		$f[x_0, x_1, x_2]$		
		$\dots$		$f[x_0, x_1, x_2, x_3]$	
$x_2$	$y_2$		$\dots$		$f[x_0, x_1, x_2, x_3, x_4]$
		$\dots$		$\dots$	
$x_3$	$y_3$		$\dots$		
		$\dots$			
$x_4$	$y_4$				

## Case of equidistant points

Bla bla bla the formula becomes

$$P(t) = a_0 + a_1(t-0) + a_2(t-0)(t-1) + \dots + a_n \prod_{i=0}^{n-1} (t-i).$$

where in this case

$$a_k = \frac{\nabla^k[y](x_k)}{k!}.$$

and

$$x = x_0 + th.$$

where  $\nabla^k[y]$  is the discrete difference.

$$\nabla[y](x_i) = y(x_i + h) - y(x_i).$$

and the backwards formula is

$$P(t) = f_n + A_1 + A_2 + \dots + A_n.$$

where

$$A_i = \frac{\bar{\nabla}^i f_n}{i!} \prod_{j=n-i+1}^n (t-j).$$

**Definition 1.2.3: Discrete Difference**

Forward discrete difference:

$$\begin{aligned}\nabla[y](x_i) &= y(x_i + h) - y(x_i) \\ \nabla^2[y](x_i) &= \nabla[y](x_i + h) - \nabla[y](x_i) \\ &= y(x_i + 2h) - 2y(x_i + h) + y(x_i) \\ \nabla^k[y](x_i) &= \nabla \left( \nabla^{k-1}[y](x_i) \right)\end{aligned}$$

Backwards discrete difference:

$$\begin{aligned}\bar{\nabla}[y](x_i) &= y(x_i) - y(x_i - h) \\ \bar{\nabla}^k[y](x_i) &= \bar{\nabla} \left( \bar{\nabla}^{k-1}[y](x_i) \right)\end{aligned}$$

$x_0$	$y_0$				
		$\nabla[y](x_i)$			
$x_1$	$y_1$		$\nabla^2[y](x_i)$		
		$\dots$		$\nabla^3[y](x_i)$	
$x_2$	$y_2$		$\dots$		$\nabla^4[y](x_i)$
		$\dots$		$\dots$	
$x_3$	$y_3$		$\dots$		
		$\dots$			
$x_4$	$y_4$				

**1.2.3 Error due to polynomial interpolation**

Let  $f(x)$  be of class  $C^{n+1}$   $\forall x \in [a, b]$  and let the polynomial  $P(x)$  interpolate it.

The error function is bounded by

$$|\text{Error}| = |f(x) - P(x)| \leq \frac{\left| \prod_{i=0}^n (x - x_i) \right|}{(n+1)!} \sup_{x \in [a, b]} \left| f^{(n+1)}(x) \right|.$$

**1.2.4 Hermite Interpolation****Definition 1.2.4: Hermite interpolation formula**

Consider  $(n+1)$  sets of point  $(x_i, y_i, y'_i)$  representing  $f(x)$  ( $y_i = f(x_i)$  and  $y'_i = f'(x_i)$ ), the hermite polynomial  $P(x)$  interpolates  $f(x)$  such that  $P'(x) = f'(x)$ .

$$P(x) = \sum_{i=0}^n h_i(x) y_i + \sum_{i=0}^n k_i(x) y'_i.$$

where

$$\begin{aligned}h_i(x) &= (1 - 2(x - x_i)\ell'_i(x_i)) \ell_i^2(x) \\ k_i(x) &= (x - x_i)\ell_i^2(x) \\ \ell_i(x) &= \prod_{\substack{j=0 \\ j \neq i}}^n \frac{x - x_j}{x_i - x_j}\end{aligned}$$

**Theorem 1.2.2** Error due to Hermite interpolation

$$|\text{Error}| = |f(x) - P(x)| \leq \frac{\left| \prod_{i=0}^n (x - x_i)^2 \right|}{(2n + 2)!} \sup_{x \in [a, b]} |f^{(2n+2)}(x)|.$$

**Existence***Proof:*

$$P(x) = \sum_{i=0}^n h_i(x) y_i + \sum_{i=0}^n k_i(x) y'_i.$$

where

$$\begin{aligned} h_i(x) &= (1 - 2(x - x_i)\ell'_i(x_i)) \ell_i^2(x) \\ k_i(x) &= (x - x_i)\ell_i^2(x) \\ \ell_i(x) &= \prod_{\substack{j=0 \\ j \neq i}}^n \frac{x - x_j}{x_i - x_j} \end{aligned}$$

Let  $i \neq j$ .

$$\begin{aligned} k_i(x_j) &= (x_j - x_i)\ell_i^2(x_j) = 0 \\ k_i(x_i) &= (x_i - x_i)\ell_i^2(x_i) = 0 \end{aligned}$$

and

$$\begin{aligned} h_i(x_j) &= (1 - 2(x_j - x_i)\ell'_i(x_i))\ell_i^2(x_j) = 0 \\ h_i(x_i) &= (1 - 2(x_i - x_i)\ell'_i(x_i))\ell_i^2(x_i) = 1 \end{aligned}$$

We conclude that  $P(x_i) = f(x_i)$ Now we have to prove that  $P'(x_i) = f'(x_i)$ 

$$\begin{aligned} h'_i(x) &= -2\ell'_i(x_i)\ell_i^2(x) + 2(1 - 2(x - x_i)\ell'_i(x_i))\ell_i(x)\ell'_i(x) \\ k'_i(x) &= \ell_i^2(x) + 2(x - x_i)\ell_i(x)\ell'_i(x) \end{aligned}$$

$$\begin{aligned} h'_i(x_j) &= -2\ell'_i(x_i)\ell_i^2(x_j) + 2(1 - 2(x_j - x_i)\ell'_i(x_i))\ell_i(x_j)\ell'_i(x_j) = 0 \\ h'_i(x_i) &= -2\ell'_i(x_i)\ell_i^2(x_i) + 2(1 - 2(x_i - x_i)\ell'_i(x_i))\ell_i(x_i)\ell'_i(x_i) = 0 \end{aligned}$$

$$\begin{aligned} k'_i(x_j) &= \ell_i^2(x_j) + 2(x_j - x_i)\ell_i(x_j)\ell'_i(x_j) = 0 \\ k'_i(x_i) &= \ell_i^2(x_i) + 2(x_i - x_i)\ell_i(x_i)\ell'_i(x_i) = 1 \end{aligned}$$

$$\therefore P'(x_i) = f'(x_i)$$

□



### Uniqueness

**Proof:** Suppose that there exists 2 polynomials  $P$  and  $Q$  of degree  $n \leq 2n + 1$  such that  $P(x_i) = Q(x_i) = f(x_i)$  and  $P'(x_i) = Q'(x_i) = f'(x_i) \forall i = 0, 1, \dots, n$ .

Let  $R(x) = P(x) - Q(x)$ .

$R = 0$  ( $n + 1$ ) times  $\Rightarrow$  according to Rolle's theorem  $\exists n$  points  $\neq x_i \Big/ R' = 0$

$R' = 0$   $n$  times as a consequence of Rolle's theorem then

$$\left. \begin{array}{l} R'(x) = 0 \text{ (} 2n + 1 \text{) times} \\ R'(x) \text{ is of degree } 2n \end{array} \right\} R'(x) = 0 \forall x.$$

$$R'(x) = 0 \Rightarrow R(x) = \text{cnst} \quad \text{and} \quad R(x_i) = P(x_i) - Q(x_i) = 0 \Rightarrow \text{cnst} = 0.$$

$$R(x) = P(x) - Q(x) = 0 \forall x.$$

$$\therefore P(x) = Q(x)$$

□

# Chapter 2

## Finding $f(x) = 0$

We will assume that every function is defined in the interval  $I = [a, b]$  and that every  $x_0 \in I$

### 2.1 Bisection Method

Suppose that  $f$  is a continuous and monotone function on the domain  $I = [a, b]$  such that  $f(a)f(b) < 0 \Rightarrow \exists r \in ]a, b[ : f(r) = 0$ .

At each step in the algorithm, in an iteration we let  $c = (a + b)/2$ , then we check the value of  $f(c)$ , if it is 0 then we are done.

However when it is not, then we define a new interval such that

$$I = \begin{cases} [a, c] & \text{if } f(c)f(a) < 0 \\ [c, b] & \text{if } f(c)f(b) < 0 \end{cases}.$$

We repeat this step until the length of the interval reaches a certain number (for example  $|b - a| < 10^{-5}$ ), at this point we stop and the best guess for the root would be  $(a + b)/2$

#### Error of the Bisection Method

After  $n$  iterations, the error of the approximated root would be

$$\text{Error} \leq \frac{|b - a|}{2^{n+1}}.$$

### 2.2 Lagrange Method

Suppose that  $f$  is a continuous and monotone function on the domain  $I = [a, b]$  such that  $f(a)f(b) < 0 \Rightarrow \exists r \in ]a, b[ : f(r) = 0$ .

The starting value of  $x_0$  depends on the value of  $f$

$$x_0 = \begin{cases} a & \text{if } f(a)f''(a) < 0 \\ b & \text{if } f(b)f''(b) < 0 \end{cases}.$$

then we can find a new guess  $x$  depending on the value of  $x_0$

- if  $x_0 = a$

$$x_1 = x_0 - \frac{b - x_0}{f(b) - f(x_0)} f(x_0).$$

- if  $x_0 = b$

$$x_1 = x_0 - \frac{a - x_0}{f(a) - f(x_0)} f(x_0).$$

## Error from Lagrange Method

For the first iteration

$$\text{Error} \leq \sup_{x \in [a, b]} |f''(x)| \frac{(b-a)^2}{8}.$$

For the second iteration

$$M_2 = \sup_{x \in [a, b]} |f''(x)|.$$

- if  $x_0 = a$

$$\text{Error} \leq \frac{M_2}{8} \left| \frac{(b-x_0)^3}{f(b)-f(x_0)} \right|.$$

- if  $x_0 = b$

$$\text{Error} \leq \frac{M_2}{8} \left| \frac{(a-x_0)^3}{f(a)-f(x_0)} \right|.$$

## 2.3 Newton Method

Suppose that  $f$  is a continuous and monotone function on the domain  $I = [a, b]$  such that  $f(a)f(b) < 0 \Rightarrow \exists r \in ]a, b[ : f(r) = 0$ .

The starting value of  $x_0$  depends on the value of  $f$

$$x_0 = \begin{cases} a & \text{if } f(a)f''(a) > 0 \\ b & \text{if } f(b)f''(b) > 0 \end{cases}.$$

Then the new guess for the root would be

$$x_1 = x_0 - \frac{f(x_0)}{f'(x_0)}.$$

### 2.3.1 Improved Newton Method

To improve the method we first let  $\eta = b - a$ , and we define condition

$$\frac{\eta M_2}{2|f'(x_0)|} < 1.$$

if the condition is not satisfied we need to choose another interval  $[a_1, b_1] \subset I$  where  $f(a_1)f(b_1) < 0$

## Error due to Newton Method

For one iteration

$$\text{Error} \leq \frac{\eta^2 M_2}{2|f'(x_0)|} \quad \text{where} \quad M_2 = \sup_{x \in [x_0 - \eta, x_0 + \eta]} |f''(x)|.$$

## 2.4 Fixed Point Iteration Method

If a function can be converted to the form  $x = g(x)$  along with the sequence  $x_{n+1} = g(x_n)$  with initial guess  $x_0$ , then it is called a fixed point scheme.

The scheme converges if

- $\forall x \in [a, b] : g(x) \in [a, b]$
- $g$  is strictly contracting meaning that  $\exists \varepsilon \in \mathbb{R} \ 0 < \varepsilon < 1$

$$\forall x, y \in [a, b], |g(x) - g(y)| \leq \varepsilon |x - y|.$$

then  $\forall x_0$  the sequence converges to  $l \in [a, b]$

**Note:-**

$$\sup_{x \in [a, b]} |g'(x)| = L < 1 \Rightarrow g(x) \text{ is strictly contracting.}$$

**Note:-**

Let  $l$  be the solution to  $g(l) = l$

- If  $|g'(l)| < 1$  then there exists an interval  $I$  containing  $l$  for which the sequence converges to  $l$
- If  $|g'(l)| > 1$  then the sequence diverges

## 2.5 Order of Convergence

Order of convergence (Rate of convergence) tells us how the error decreases between 2 iterations. The order of convergence  $p$  of a sequence is defined to be

$$\lim_{n \rightarrow +\infty} \left| \frac{x_{n+1} - l}{(x_n - l)} \right| \in \mathbb{R}_+^*.$$

**Note:-**

The order of convergence of

- Lagrange Method

$$g'(l) = \frac{(b-l)^2}{2f(b)} f''(c).$$

If  $f''(c) \neq 0$  then  $g'(l) \neq 0$  then the order is 1.

- Newton method, if  $g'(l) = 0$  then the order is at least 2.

**Note:-**

We stop the iteration method when

- First case  $g'(x) < 0$ , then we stop iteration when

$$|x_{n+1} - r| < \varepsilon.$$

- Second case  $g'(x) > 0$ , then we stop iteration when

$$|f(x_n)| < \eta.$$

where

$$\eta = \varepsilon \inf |f'(x)|.$$

## 2.6 Polynomial Shenanigans

### 2.6.1 Roots of $x^3 + px + q = 0$

Let  $y_1(x) = x^3 + px$  and  $y_2(x) = -q$

- $p \geq 0 \Rightarrow \exists 1$  root
- $p < 0$  then we have 3 separate cases

$$27q^2 + 4p^3 \begin{cases} = 0 & \text{we have 2 separate real roots (one double and one single)} \\ > 0 & \text{we have one real root} \\ < 0 & \text{we have 3 separate real roots} \end{cases}.$$

### 2.6.2 Roots of $x^3 + ax^2 + bx + c = 0$

If we replace  $x$  with  $X + h$  where  $h = -\frac{a}{3}$ , we can get the cubic in the form

$$X^3 + PX + Q = 0.$$

where

$$P = -\frac{a^2}{3} + b$$
$$Q = \frac{2a^3}{27} - \frac{ab}{3} + c$$

### 2.6.3 Roots of $x^4 + ax^3 + bx^2 + cx + d = 0$

If we replace  $x$  with  $X + h$  where  $h = -\frac{a}{4}$ , we can get the quartic in the form

$$X^4 + PX^2 + QX + R = 0.$$

where

$$P = -\frac{3a^2}{8} + b$$
$$Q = \frac{a^3}{8} - \frac{ab}{2} + c$$
$$R = -\frac{3a^4}{256} - \frac{ac}{4} + d$$

Let the circle  $C$  be the circle of radius  $\left(-\frac{Q}{2}, \frac{1-P}{2}\right)$  and of radius  $\sqrt{\left(\frac{P-1}{2}\right)^2 + \frac{Q^2}{4}} - R$ .

The roots of the polynomial  $X^4 + PX^2 + QX + R = 0$  are the intersection of the circle  $C$  and the parabola  $Y = X^2$

## Chapter 3

# Numerical Intergration

Let  $f$  be a continuous function on  $[a, b]$  and  $I = \int_a^b f(x) dx$

### 3.1 Rectangle method

We sample the domain of  $f$  in to  $n$  equal subintervals ( $x_i - x_{i+1} = \frac{b-a}{n} = h$ ). The approximated value of  $I$  becomes

$$\int_a^b f(x) dx \approx \frac{b-a}{n} \left[ \sum_{i=0}^{n-1} f(x_i) \right].$$

such that  $x_0 = a$  and  $x_1 = b$

The error associated with this approximation is

$$|\varepsilon| \leq \frac{M_1}{2n} (b-a)^2.$$

where

$$M_1 = \sup_{[a,b]} |f'(x)|.$$

### 3.2 Trapezoid method

Same sampling as before.

$$\int_a^b f(x) dx \approx \frac{b-a}{2n} \left( f(a) + f(b) + 2 \sum_{i=1}^{n-1} f(x_i) \right).$$

The error associated with this approximation is

$$|\varepsilon| \leq \frac{M-2}{12n^2} |(b-a)^3|.$$

where

$$M_2 = \sup_{[a,b]} |f''(x)|.$$

### 3.3 Simpson's rule

You get the point by now

$$\int_a^b f(x) dx \approx \frac{b-a}{6n} \left( f(a) + f(b) + 2 \sum_{i=1}^{n-1} f(x_i) + 4 \left( f\left(\frac{a+x_1}{2}\right) + f\left(\frac{x_1+x_2}{2}\right) + \dots + f\left(\frac{x_{n-1}+b}{2}\right) \right) \right).$$

The error associated with this approximation is

$$|\varepsilon| \leq \frac{|(b-a)^5|}{n^4} \frac{M_4}{2880}.$$

where

$$M_4 = \sup_{[a,b]} |f^{(4)}(x)|.$$

### 3.4 Newton Cote's method

Let  $P$  be the Lagrange polynomial that interpolates the function  $f$  at  $(n+1)$  points  $(x_0, f_0), (x_1, f_1), \dots, (x_n, f_n)$ .

$$P(s) = \sum_{i=0}^n \ell_i(s) f(x_i).$$

The approximated value of  $I$  is

$$\int_a^b f(x) dx \approx (b-a) \sum_{i=0}^n H_i f_i.$$

where

$$H_i = \frac{1}{n} \int_0^n \ell_i(s) ds.$$