

# The Basic Problem

Given a set of tabular values ( available from an experiment) of a function  $f$ .

$X$	$x_0$	$x_1$	$x_2$	...	$x_n$
$f(X)$	$f(x_0)$	$f(x_1)$	$f(x_3)$	...	$f(x_n)$

Where the **explicit nature of the function  $f$  is not known.**

**To find a function  $\phi$  such that  $\phi(x_i) = f(x_i)$  for all  $0 \leq i \leq n$ .**

Such a function  $\phi$  is called an ***interpolating function.***

# Geometrically

**The problem is :**

To find a function / polynomial whose graph passes through the given set of  $(n + 1)$ – points,

$$(x_0, f(x_0)), (x_1, f(x_1)), \dots, (x_n, f(x_n)).$$

# Polynomial interpolation

If  $\phi$  is a polynomial then the process is called **polynomial interpolation** and  $\phi$  is called an **interpolating polynomial**.

# Some types of interpolation

- Polynomial interpolation
- Piecewise Polynomial( Spline ) interpolation
- Rational interpolation
- Trigonometric interpolation
- Exponential interpolation

**We study :** Polynomial and Piecewise polynomial interpolations.

## Example

Find a polynomial which fits the following data.

$X$	$x_0$	$x_1$	$x_2$
$f(X)$	$f(x_0)$	$f(x_1)$	$f(x_2)$

# Example

**Solution:**

$$P_3(x) = \frac{(x - x_1)(x - x_2)}{(x_0 - x_1)(x_0 - x_2)} f(x_0) + \frac{(x - x_0)(x - x_2)}{(x_1 - x_0)(x_1 - x_2)} f(x_1) + \frac{(x - x_0)(x - x_1)}{(x_2 - x_0)(x_2 - x_1)} f(x_2).$$

This is a polynomial of **degree 2**.

- Called **Lagrange's** interpolation formula.

# Lagrange's Interpolation formula

Similarly,

$$\begin{aligned} p_n(x) = & \frac{(x - x_1)(x - x_2) \cdots (x - x_n)}{(x_0 - x_1)(x_0 - x_2) \cdots (x_0 - x_n)} f(x_0) \\ & + \frac{(x - x_2)(x - x_3) \cdots (x - x_n)}{(x_1 - x_0)(x_1 - x_2) \cdots (x_1 - x_n)} f(x_1) \\ & + \cdots + \frac{(x - x_0)(x - x_1) \cdots (x - x_{n-1})}{(x_n - x_0)(x_n - x_1) \cdots (x_n - x_{n-1})} f(x_n) \end{aligned}$$

is the “ **unique polynomial of degree  $\leq n$** ” which interpolates  $f$  at the  $n + 1$  points  $x_0, x_1, \cdots, x_n$ .

# Lagrange's Interpolation formula

If we write

$$l_i(x) = \frac{(x - x_0)(x - x_1) \cdots (x - x_{i-1})(x - x_{i+1}) \cdots (x - x_n)}{(x_i - x_0)(x_i - x_1) \cdots (x_i - x_{i-1})(x_i - x_{i+1}) \cdots (x_i - x_n)}$$

then

$$p_n(x) = \sum_{i=0}^n l_i(x) f(x_i)$$

is the “ **unique(how?) polynomial of degree  $\leq n$** ” which interpolates  $f$  at the  $n + 1$  points  $x_0, x_1, \cdots, x_n$ .



## We proved:

**Theorem 1. (Existence and Uniqueness theorem for interpolating polynomials.)** *If  $x_0, x_1, x_2, \dots, x_n$  are  $n + 1$  distinct numbers and  $f$  is a function whose values are given at these numbers, then a unique polynomial  $p(x)$  of **degree**  $\leq n$  with  $p(x_i) = f(x_i)$  for all  $i = 0, 1, 2, \dots, n$ .*

*This polynomial is given by*

$$p(x) = \sum_{i=0}^n l_i(x) f(x_i)$$

## Thus we have:

Given  $x_0, x_1, x_2, x_3, \dots, x_n$  and  $f(x_0), f(x_1), \dots, f(x_n)$ , we have:

Lagrange's polynomial	Degree at most	interpolates $f$ at
$p_1(x)$	1	$x_0, x_1$
$p_2(x)$	2	$x_0, x_1, x_2$
$p_3(x)$	3	$x_0, x_1, x_2, x_3$
$\vdots$	$\vdots$	$\vdots$
$p_n(x)$	$n$	$x_0, x_1, x_2, x_3, \dots, x_n$

## Example

Find the interpolating polynomial in Lagrangian form for the data.

$x$	-2	-1	1	3
$f(x)$	-15	-4	0	20

**Solution:**

We calculate  $p_3(x)$ .

$$p_3(x) = l_0(x).f(x_0) + l_1(x).f(x_1) + l_2(x).f(x_2) + l_3(x).f(x_3)$$

Where  $l_0(x) = \frac{(x-x_1)(x-x_2)(x-x_3)}{(x_0-x_1)(x_0-x_2)(x_0-x_3)} = \frac{(x+1)(x-1)(x-3)}{(-1)(-3)(-5)}$

$$= \frac{-1}{15}(x^3 - 3x^2 - x + 3)$$

Similarly,

$$l_1(x) = \frac{1}{8}(x^3 - 2x^2 - 5x + 6)$$

$$l_2(x) = \frac{-1}{12}(x^3 - 7x - 6)$$

$$l_3(x) = \frac{1}{40}(x^3 + 2x^2 - x - 2)$$

Therefore,

$$p_3(x) = -15\left\{\frac{-1}{15}(x^3 - 3x^2 - x + 3)\right\} - 4\left\{\frac{1}{8}(x^3 - 2x^2 - 5x + 6)\right\}$$

$$\begin{aligned} &+0 + 20\left\{\frac{1}{40}(x^3 + 2x^2 - x + -2)\right\} \\ &= x^3 - x^2 + x - 1. \end{aligned}$$

**Error:** Let  $f : [a, b] \longrightarrow \mathbb{R}$

$$E_n(x) = \frac{f^{n+1}}{n+1}(\xi)(x - x_0)(x - x_1) \cdots (x - x_n)$$

For some  $\xi \in (a, b)$ .

Problem: We do not know  $\xi$ , hence it would be difficult to compute error with this formula. However, this can be used in order to find the upper bound of the error, i.e., we have a number in hand such that error should not exceed by this number.

## A disadvantage

- If more data become available, the work performed to generate the original Lagrange form cannot be reused to compute a higher-degree polynomial.

This can be rectified by writing the Lagrange's interpolation polynomial in **Newton's form**.

For this we use the idea of **divided differences**.

# First order divided differences

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

$$f[x_1, x_2] = \frac{f(x_2) - f(x_1)}{x_2 - x_1}$$

⋮                    ⋮

In general,

$$f[x_i, x_{i+1}] = \frac{f(x_{i+1}) - f(x_i)}{x_{i+1} - x_i}$$

⋮                    ⋮

$$f[x_{n-1}, x_n] = \frac{f(x_n) - f(x_{n-1})}{x_n - x_{n-1}}$$

## Second order divided differences

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

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

⋮

⋮

$$f[x_i, x_{i+1}, x_{i+2}] = \frac{f[x_{i+1}, x_{i+2}] - f[x_i, x_{i+1}]}{x_{i+2} - x_i}$$

⋮

⋮

$$f[x_{n-2}, x_{n-1}, x_n] = \frac{f[x_{n-1}, x_n] - f[x_{n-2}, x_{n-1}]}{x_n - x_{n-2}}$$



## Divided differences of $n^{\text{th}} (\geq 3)$ order

- Can be defined recursively in a similar way.

$$f[x_i, x_{i+1}, \dots, x_{i+n}] = \frac{f[x_{i+1}, \dots, x_{i+n}] - f[x_i, x_{i+1}, \dots, x_{i+n-1}]}{x_{i+n} - x_i}$$

The numbers  $f(x_0), f(x_1), \dots, f(x_n)$  are called divided differences of **0-th order** and are denoted by  $f[x_0], f[x_1], \dots, f[x_n]$  respectively.

**We say:**  $f[x_0, x_1, \dots, x_n]$  as the divided difference of  $f$  at the points  $x_0, x_1, \dots, x_n$ .

## Divided differences in Tabular form (when $n=3$ .)

$x$	$f(x)$	First order	Second order	Third order
$x_0$	$f[x_0]$	$f[x_0, x_1]$		
$x_1$	$f[x_1]$	$f[x_1, x_2]$	$f[x_0, x_1, x_2]$	$f[x_0, x_1, x_2, x_3]$
$x_2$	$f[x_2]$	$f[x_2, x_3]$	$f[x_1, x_2, x_3]$	
$x_3$	$f[x_3]$			

# Example

Consider the data:

$x$	-2	-1	1	3
$f(x)$	-15	-4	0	20

The divided difference table:

$x$	$f(x)$	$f[ , ]$	$f[ , , ]$	$f[ , , , ]$
-2	-15			
		11		
-1	-4		-3	
		2		1
1	0		2	
		10		
3	20			

# The Newton's form of a polynomial

$$p_n(x) = A_0 + A_1(x-x_0) + A_2(x-x_0)(x-x_1) + \cdots + A_n(x-x_0)(x-x_1)\cdots(x-x_{n-1})$$

We need to compute  $A_i, i = 0, 1 \cdots n$ .

This can be done easily by using

$$f(x_i) = p_n(x_i), i = 0, 1 \cdots n$$

Can you see a relation between  $A_i$  and divided difference?

# Idea

**We prove:**

$$A_k = f[x_0, x_1, \dots, x_k]$$

for all  $k = 0, 1, 2, \dots, n$ .

**Thus the Newton's form of  $p_n(x)$  is:**

$$p_n(x) = f[x_0]$$

$$+ f[x_0, x_1](x - x_0)$$

$$+ f[x_0, x_1, x_2](x - x_0)(x - x_1)$$

$$+ f[x_0, x_1, x_2, x_3](x - x_0)(x - x_1)(x - x_2)$$

$$\cdots + f[x_0, x_1, \cdots, x_n](x - x_0)(x - x_1)(x - x_2) \cdots (x - x_{n-1})$$

- *Look at the divided difference table and recognize the coefficients.*
- Thus the divided difference table helps us to write the interpolation

polynomial easily.

$x$	$f(x)$	First order	Second order	Third order
$x_0$	$f[x_0]$	$f[x_0, x_1]$		
$x_1$	$f[x_1]$	$f[x_1, x_2]$	$f[x_0, x_1, x_2]$	
$x_2$	$f[x_2]$	$f[x_2, x_3]$	$f[x_1, x_2, x_3]$	$f[x_0, x_1, x_2, x_3]$
$x_3$	$f[x_3]$			



# Example

Consider the data:

$x$	-2	-1	1	3
$f(x)$	-15	-4	0	20

The divided difference table:

$x$	$f(x)$	$f[ , ]$	$f[ , , ]$	$f[ , , , ]$
-2	-15			
		11		
-1	-4		-3	
		2		1
1	0		2	
		10		
3	20			

$$\begin{aligned} p_3(x) &= -15 + 11(x + 2) - 3(x + 2)(x + 1) + 1(x + 2)(x + 1)(x - 1) \\ &= x^3 - x^2 + x - 1. \end{aligned}$$

## When the points $x_0, x_1, \dots, x_n$ are equispaced.

- While forming the divided difference table, we look for the correct division, and there is chance of mistake. **Did you realize this?** Let  $x_{i+1} - x_i = h$  **for all**  $i = 0, 1, 2, \dots, n - 1$ .
- The Newton's divided difference formula can be further simplified to Newton's forward difference form.

We first define the forward differences.

# Forward differences

**First order forward differences:**

$$\Delta^1 f(x_0) = f(x_1) - f(x_0)$$

$$\Delta^1 f(x_1) = f(x_2) - f(x_1)$$

⋮            ⋮

$$\Delta^1 f(x_i) = f(x_{i+1}) - f(x_i)$$

⋮            ⋮

$$\Delta^1 f(x_{n-1}) = f(x_n) - f(x_{n-1})$$

We call  $\Delta^1 f(x_i)$  as: The first order forward difference of  $f$  at  $x_i$ .

Forward differences of higher order:

- can be computed using the recursive following formula.

If  $n > 1$  then we define,

$$\Delta^n f(x_i) = \Delta^{n-1} f(x_{i+1}) - \Delta^{n-1} f(x_i)$$

## Forward difference table

Consider the data:

$x$	-2	-1	0	1
$f(x)$	-15	-4	0	20

The forward difference table:

$x$	$f(x)$	$\Delta^1 f$	$\Delta^2 f$	$\Delta^3 f$
-2	-15			
		11		
-1	-4		-7	
		4		23
0	0		16	
		20		
1	20			

## Divided differences in terms of forward differences

Suppose  $f : [a, b] \rightarrow \mathbb{R}$  is given at the  $n + 1$  **equispaced** points  $x_0, x_1, \dots, x_n$  with spacing  $h = x_{i+1} - x_i$ .

$$f[x_0, x_1] = \frac{f(x_1) - f(x_0)}{x_1 - x_0} = \frac{1}{h} \Delta f(x_0)$$

$$\begin{aligned} f[x_0, x_1, x_2] &= \frac{f[x_1, x_2] - f[x_0, x_1]}{x_2 - x_0} \\ &= \frac{1}{2h} \left[ \frac{\Delta f(x_1) - \Delta f(x_0)}{h} \right] = \frac{1}{2h^2} \Delta^2 f(x_0) \end{aligned}$$

In general,

$$f[x_0, x_1, \dots, x_k] = \frac{1}{k! h^k} \Delta^k f(x_0).$$



## Newton's forward difference formula.

Suppose  $f : [a, b] \rightarrow \mathbb{R}$  is given at the  $n + 1$  **equispaced** points  $x_0, x_1, \dots, x_n$  with spacing  $h = x_{i+1} - x_i$ .

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

Thus  $x$  can be written in the form  $x = x_0 + sh$  for some  $s$

In particular,  $x_i = x_0 + ih$  for all  $i = 0, 1, 2, \dots, n$ .

Thus

$$x - x_i = x_0 + sh - (x_0 + ih) = (s - i)h$$

Now the interpolating polynomial  $p_n(x)$  can be written as:

$$\begin{aligned} p_n(x) &= p_n(x_0 + sh) = f[x_0] \\ &\quad + sh f[x_0, x_1] \\ &\quad + s(s-1)h^2 f[x_0, x_1, x_2] \\ &\quad + s(s-1)(s-2)h^3 f[x_0, x_1, x_2, x_3] \\ &\quad \cdots + s(s-1)(s-2)\cdots(s-n+1)f[x_0, x_1, \cdots, x_n] \\ &= \sum_{k=0}^n s(s-1)(s-2)\cdots(s-k+1)f[x_0, x_1, \cdots, x_k] \end{aligned}$$

$$\begin{aligned} &= \sum_{k=0}^n \binom{s}{k} k! h^k f[x_0, x_1, \dots, x_k] \\ &= \sum_{k=0}^n \binom{s}{k} \Delta^k f(x_0) \end{aligned}$$

Where

$$\binom{s}{k} = \frac{s(s-1)\cdots(s-k+1)}{k!}.$$

## Example

Prepare the forward difference table for the following data. Using Newton's forward interpolating polynomial, find approximate value of  $f(0.1)$ .

$x$	0	0.2	0.4	0.6	0.8
$f(x)$	0.12	0.46	0.74	0.90	1.2

$x$	$f(x)$	$\Delta^1 f$	$\Delta^2 f$	$\Delta^3 f$	$\Delta^4 f$
0	0.12				
		0.34			
0.2	0.46		-0.06		
		0.28		-0.06	
0.4	0.74		-0.12		0.32
		0.16		0.26	
0.6	0.90		0.14		
		0.30			
0.8	1.2				

To interpolate at  $x = 0.1$

$$s = \frac{x - x_0}{h} = \frac{0.1 - 0}{0.2} = \frac{1}{2}.$$

Therefore the binomial coefficients are:

$$\binom{s}{0} = 1, \quad \binom{s}{1} = \frac{1}{2}, \quad \binom{s}{2} = \frac{\frac{1}{2}(\frac{1}{2} - 1)}{1 \cdot 2} = -\frac{1}{8}.$$

$$\binom{s}{3} = \frac{s(s-1)(s-2)}{1 \cdot 2 \cdot 3} = \frac{\frac{1}{2}(-\frac{1}{2})(-\frac{3}{2})}{1 \cdot 2 \cdot 3} = \frac{1}{16}.$$

$$\binom{s}{4} = \frac{s(s-1)(s-2)(s-3)}{1.2.3.4} = \frac{\frac{1}{2}(-\frac{1}{2})(-\frac{3}{2})(-\frac{5}{2})}{1.2.3.4} = -\frac{5}{128}.$$

Therefore

$$\begin{aligned} p_4(x_0+sh) &= f(x_0) + s\Delta^1 f(x_0) + \binom{s}{2} \Delta^2 f(x_0) + \binom{s}{3} \Delta^3 f(x_0) + \binom{s}{4} \Delta^4 f(x_0). \\ &= 0.12 + \frac{1}{2}(0.34) + \frac{-1}{8}(-0.06) + \frac{1}{16}(-0.06) + \frac{-5}{128}(0.32) = 0.28125 \end{aligned}$$

Hence  $f(0.1) \approx 0.28125$

# Piece-wise polynomial interpolation

- Polynomials with higher degree grow / decay very fast.
- When the number of tabular points is more in number, the interpolating polynomial would be of higher degree and hence there is a chance of getting more error.

Therefore,

We divide the given interval into small subintervals and on each small subinterval we fit a polynomial of lower degree.



## Newton's backward difference formula

$$\nabla f(x_1) = f(x_1) - f(x_0)$$

$$\nabla f(x_2) = f(x_2) - f(x_1)$$

$$\nabla f(x_i) = f(x_i) - f(x_{i-1})$$

$$\nabla f(x_n) = f(x_n) - f(x_{n-1})$$

meaning full for all  $i = 1, 2, \dots, n$

Note that

$$\Delta^1 f(x_2) = f(x_3) - f(x_2) = \nabla^1 f(x_3)$$

Thus in general,

$$\Delta^1 f(x_i) = \nabla^1 f(x_{i+1})$$

for all  $i = 1, 2, \dots, n - 1$

Thus we can use the forward difference table to determine the backward differences.

## Newton's backward difference formula

Given  $f$  at the equispaced points  $x_0, x_1, \dots, x_n$  with spacing  $h$ .

Let  $x$  be in between  $x_0$  and  $x_n$ .

In this case, we take  $s = \frac{x-x_n}{h}$

Now

$$\begin{aligned} p_n(x) &= f[x_n] \\ &+ f[x_n, x_{n-1}](x - x_n) \\ &+ f[x_n, x_{n-1}, x_{n-2}](x - x_n)(x - x_{n-1}) \\ &+ f[x_n, x_{n-1}, x_{n-2}, x_{n-3}](x - x_n)(x - x_{n-1})(x - x_{n-2}) \end{aligned}$$

$$\begin{aligned}
& \cdots + f[x_n, x_{n-1}, \cdots, x_0](x - x_n)(x - x_{n-1})(x - x_{n-2}) \cdots (x - x_1) \\
& = f[x_n] + sh f[x_n, x_{n-1}] + s.(s + 1).h^2 f[x_n, x_{n-1}, x_{n-2}] \\
& \quad + s.(s + 1).(s + 2).h^3 f[x_n, x_{n-1}, x_{n-2}, x_{n-3}] \\
& \quad \cdots + s.(s + 1) \cdots (s + n - 1).h^n .f[x_n, x_{n-1}, \cdots, x_0]
\end{aligned}$$

We can now prove that:

$$f[x_n, x_{n-1}, \cdots, x_{n-k}] = \frac{1}{k! h^k} \nabla^k f(x_n).$$

Thus

$$p_n(x) = f[x_n] + s \cdot \nabla^1 f(x_n) + \frac{s(s+1)}{2} \nabla^2 f(x_n) + \dots + \frac{s(s+1)(s+2) \cdots (s+n-1)}{n!} \nabla^n f(x_n)$$

If we write

$$\begin{aligned} \binom{-s}{k} &= \frac{-s(-s-1)(-s-2) \cdots (-s-k+1)}{k!} \\ &= (-1)^k \frac{s(s+1)(s+2) \cdots (s+k-1)}{k!} \end{aligned}$$

then

$$p_n(x) = f[x_n] + \sum_{k=1}^n (-1)^k \binom{-s}{k} \cdot \nabla^k f(x_n)$$

## Example

Interpolate at  $x = 0.65$  using Newton's backward interpolation formula.

$x$	0	0.2	0.4	0.6	0.8
$f(x)$	0.12	0.46	0.74	0.90	1.2

$x$	$f(x)$	$\Delta^1 f$	$\Delta^2 f$	$\Delta^3 f$	$\Delta^4 f$
0	0.12				
		0.34			
0.2	0.46		-0.06		
		0.28		-0.06	
0.4	0.74		-0.12		<b>0.32</b>
		0.16		<b>0.26</b>	
0.6	0.90		<b>0.14</b>		
		<b>0.30</b>			
0.8	<b>1.2</b>				



Give point  $x = 0.65$ .

In the case of Newton's backward form:

$$s = \frac{x - x_n}{h} = \frac{x - x_4}{h} = \frac{0.65 - 0.8}{0.2} = -\frac{3}{4}.$$

The binomial coefficients :

$$\binom{-s}{0} = 1, \quad \binom{-s}{1} = \frac{3}{4}, \quad \binom{-s}{2} = \frac{-3}{32}$$

$$\binom{-s}{3} = \frac{5}{128}, \quad \binom{-s}{4} = \frac{-45}{2048}$$

$$p_4(x) = f[x_4] + \sum_{k=1}^4 (-1)^k \binom{-s}{k} \cdot \nabla^k f(x_4)$$

$$\begin{aligned} f(0.65) &= p_4\left(0.8 - \frac{3}{4} \times 0.2\right) \\ &= f[x_0] - \frac{3}{4} \nabla^1(x_4) - \frac{3}{32} \nabla^2(x_4) - \frac{5}{128} \nabla^3(x_4) - \frac{45}{2048} \nabla^4(x_4) \\ &= 0.944687 \end{aligned}$$

# We Use the interpolating polynomial

- To estimate the missing data.

OR

- To approximate  $f$  at some intermediate points.