



# Math 4543: Numerical Methods

## Lecture 7 — Newton's Divided Difference Method

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# Lecture Plan

## The agenda for today

- Represent interpolant polynomials using NDD
- Understand the advantage of NDD over the Direct Method
- Generalize the formula for finding the coefficients for  $n^{th}$  order interpolant
- Derive the formula for the quadratic NDD interpolant

# Newton's Divided Difference Interpolation

## What is it?

Newton represented the interpolant polynomial in such a manner so that the *coefficients* of the polynomial can be computed using the *division* of some *difference* values.

As given in Figure 1, data is given at discrete points such as  $(x_0, y_0), (x_1, y_1), \dots, (x_{n-1}, y_{n-1}), (x_n, y_n)$ .

A continuous function  $f(x)$  may be used to represent the  $n + 1$  data values with  $f(x)$  passing through the  $n + 1$  points.

Then one can find the value of  $y$  at any other value  $x$ .

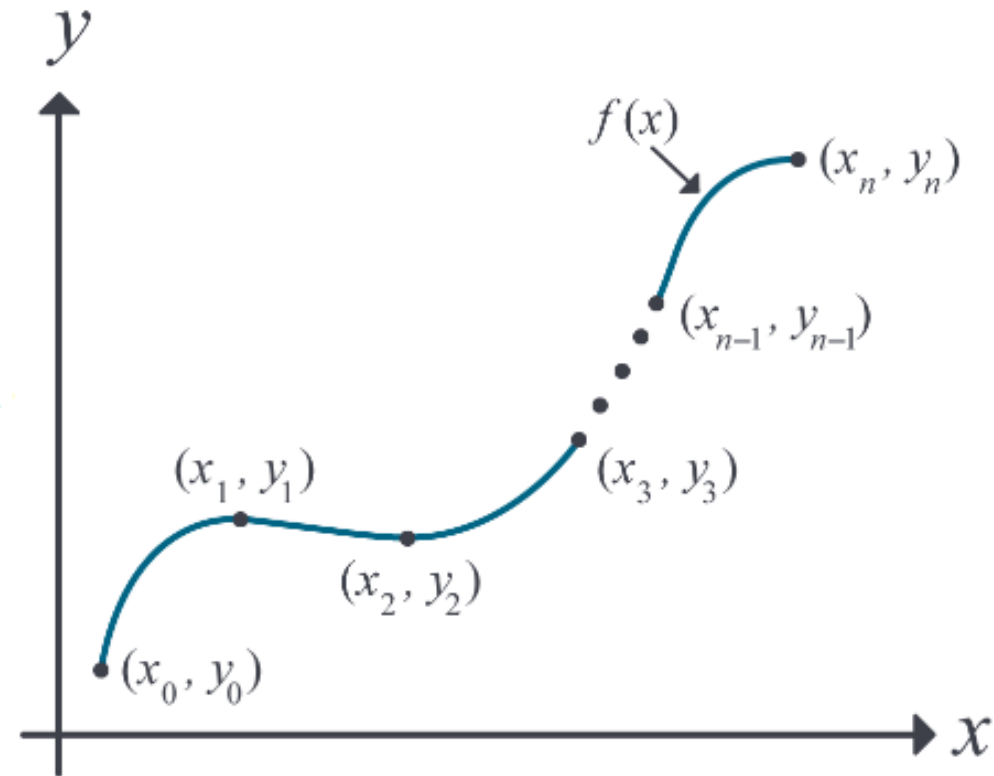


Figure 1. Interpolation of a function given at discrete points

# Newton's Divided Difference Interpolation

## Linear Interpolation

Given  $(x_0, y_0)$  and  $(x_1, y_1)$ , fit a linear interpolant through the data. Noting  $y = f(x)$  and  $y_1 = f(x_1)$ , assume the linear interpolant  $f_1(x)$  is given by (Figure 2)

$$f_1(x) = b_0 + b_1(x - x_0)$$

Since at  $x = x_0$ ,

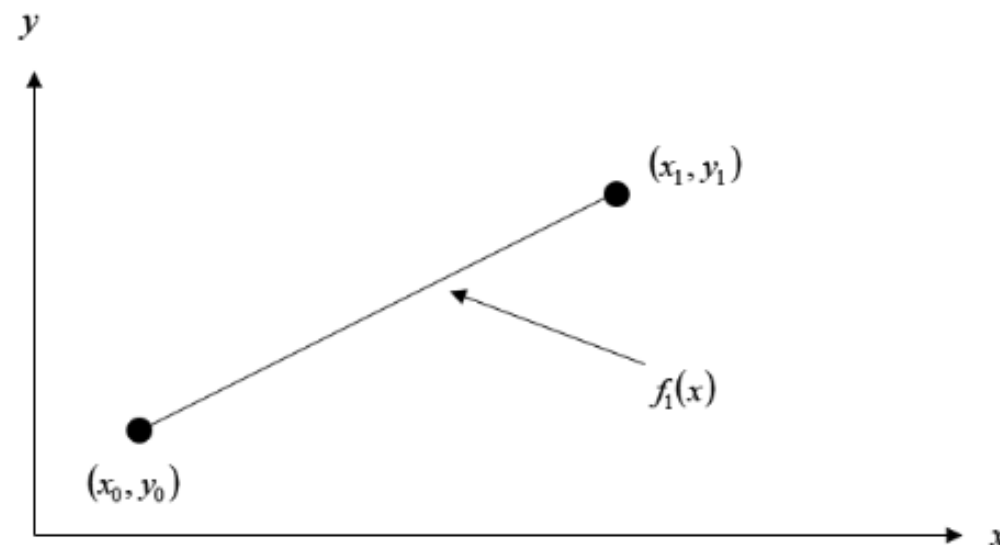
$$f_1(x_0) = f(x_0) = b_0 + b_1(x_0 - x_0) = b_0$$

and at  $x = x_1$ ,

$$\begin{aligned} f_1(x_1) &= f(x_1) = b_0 + b_1(x_1 - x_0) \\ &= f(x_0) + b_1(x_1 - x_0) \end{aligned}$$

giving

$$b_1 = \frac{f(x_1) - f(x_0)}{x_1 - x_0}$$



**Figure 2** Linear interpolation.

So

$$b_0 = f(x_0)$$

$$b_1 = \frac{f(x_1) - f(x_0)}{x_1 - x_0}$$

giving the linear interpolant as

$$f_1(x) = b_0 + b_1(x - x_0)$$

$$f_1(x) = f(x_0) + \frac{f(x_1) - f(x_0)}{x_1 - x_0}(x - x_0)$$

# Newton's Divided Difference Interpolation

## Quadratic Interpolation

Given  $(x_0, y_0)$ ,  $(x_1, y_1)$ , and  $(x_2, y_2)$ , fit a quadratic interpolant through the data. Noting  $y = f(x)$ ,  $y_0 = f(x_0)$ ,  $y_1 = f(x_1)$ , and  $y_2 = f(x_2)$ , assume the quadratic interpolant  $f_2(x)$  is given by

$$f_2(x) = b_0 + b_1(x - x_0) + b_2(x - x_0)(x - x_1)$$

At  $x = x_0$ ,

$$\begin{aligned} f_2(x_0) &= f(x_0) = b_0 + b_1(x_0 - x_0) + b_2(x_0 - x_0)(x_0 - x_1) \\ &= b_0 \\ b_0 &= f(x_0) \end{aligned}$$

At  $x = x_1$

$$f_2(x_1) = f(x_1) = b_0 + b_1(x_1 - x_0) + b_2(x_1 - x_0)(x_1 - x_1)$$

$$f(x_1) = f(x_0) + b_1(x_1 - x_0) \quad \text{Hence the quadratic interpolant is given by}$$

giving

$$b_1 = \frac{f(x_1) - f(x_0)}{x_1 - x_0}$$

$$f_2(x) = b_0 + b_1(x - x_0) + b_2(x - x_0)(x - x_1)$$

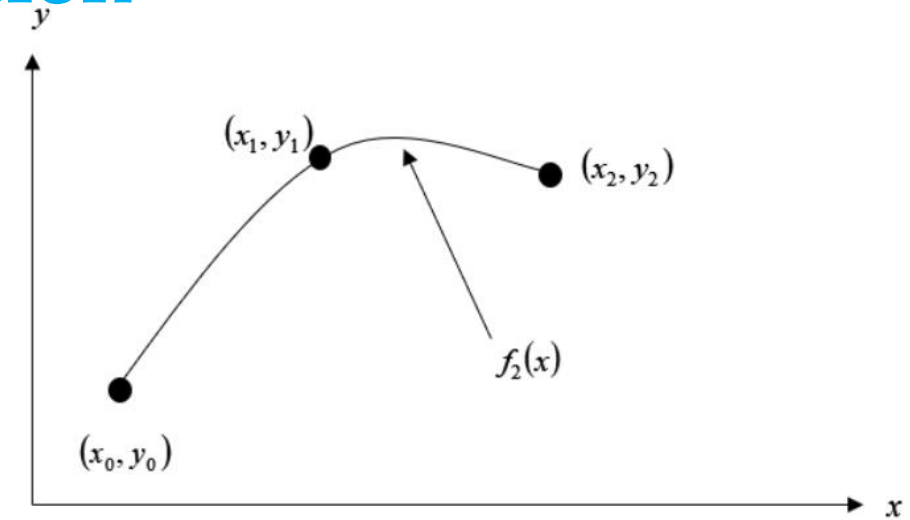
$$= f(x_0) + \frac{f(x_1) - f(x_0)}{x_1 - x_0}(x - x_0) + \frac{\frac{f(x_2) - f(x_1)}{x_2 - x_1} - \frac{f(x_1) - f(x_0)}{x_1 - x_0}}{x_2 - x_0}(x - x_0)(x - x_1)$$

At  $x = x_2$

$$f_2(x_2) = f(x_2) = b_0 + b_1(x_2 - x_0) + b_2(x_2 - x_0)(x_2 - x_1)$$

$$f(x_2) = f(x_0) + \frac{f(x_1) - f(x_0)}{x_1 - x_0}(x_2 - x_0) + b_2(x_2 - x_0)(x_2 - x_1)$$

$$b_2 = \frac{\frac{f(x_2) - f(x_1)}{x_2 - x_1} - \frac{f(x_1) - f(x_0)}{x_1 - x_0}}{x_2 - x_0}$$



# Mini Quiz

## NDD Interpolant vs Direct Method Interpolant

Do the  $n^{th}$  order interpolants obtained using **NDD** and **Direct method** differ?

If so, how do they differ? If not, why not?

$$y = a_0 + a_1x + \dots + a_nx^n$$

*Vs.*

$$f_n(x) = b_0 + b_1(x - x_0) + \dots + b_n(x - x_0)(x - x_1)\dots(x - x_{n-1})$$

# Newton's Divided Difference Interpolation

Why do we need it?

The advantages are —

- ✓ Need to solve *one equation* to get *one coefficient*.
- ✓ Overall time complexity to obtain the interpolant is  $O(n^2)$ .

For the Direct method, we needed to calculate the *inverse of a matrix* and simultaneously solve *all the equations* to obtain all the coefficients. The time complexities of the algorithms that are used to do this are,

- Naïve Gaussian Elimination —  $O(n^3 \log(||A|| + ||b||))$ , for  $Ax = b$  ([Wayne Eberly et al.](#))
- LU Decomposition —  $O(n^3)$
- Cramer's Rule —  $O((n + 1)!)$

# Newton's Divided Difference Interpolation

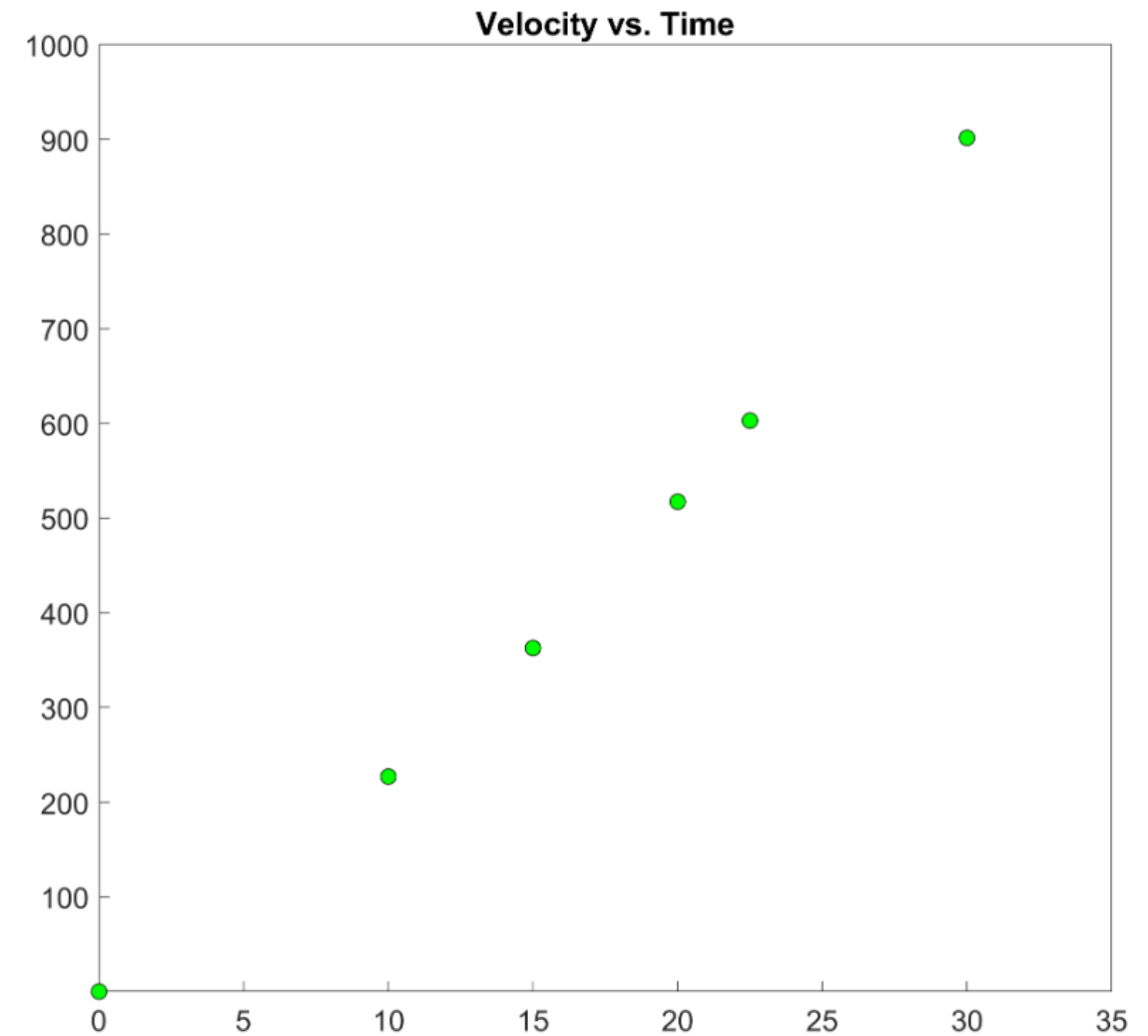
## A first-order polynomial example

The upward velocity of a rocket is given as a function of time in Table 1.

**Table 1.** Velocity as a function of time.

$t$ (s)	$v(t)$ (m/s)
0	0
10	227.04
15	362.78
20	517.35
22.5	602.97
30	901.67

Determine the value of the velocity at  $t = 16$  seconds using first order polynomial interpolation by Newton's divided difference polynomial method.



**Figure 1.** Graph of velocity vs. time data for the rocket example.



# Newton's Divided Difference Interpolation

## A first-order polynomial example

### Solution

For linear interpolation, the velocity is given by

$$v(t) = b_0 + b_1(t - t_0)$$

Since we want to find the velocity at  $t = 16$ , and we are using a first order polynomial, we need to choose the two data points that are closest to  $t = 16$  that also bracket  $t = 16$  to evaluate it. The two points are  $t = 15$  and  $t = 20$ .

$$\begin{aligned} b_0 &= v(t_0) \\ &= 362.78 \end{aligned}$$

$$t_0 = 15, v(t_0) = 362.78$$

$$t_1 = 20, v(t_1) = 517.35$$

$$\begin{aligned} b_1 &= \frac{v(t_1) - v(t_0)}{t_1 - t_0} \\ &= \frac{517.35 - 362.78}{20 - 15} \\ &= 30.914 \end{aligned}$$

Hence

$$\begin{aligned} v(t) &= b_0 + b_1(t - t_0) \\ &= 362.78 + 30.914(t - 15), \quad 15 \leq t \leq 20 \end{aligned}$$

At  $t = 16$ ,

$$\begin{aligned} v(16) &= 362.78 + 30.914(16 - 15) \\ &= 393.69 \text{ m/s} \end{aligned}$$

If we expand

$$v(t) = 362.78 + 30.914(t - 15), \quad 15 \leq t \leq 20$$

we get

$$v(t) = -100.93 + 30.914t, \quad 15 \leq t \leq 20$$

and this is the same expression as obtained in the direct method.

# Newton's Divided Difference Interpolation

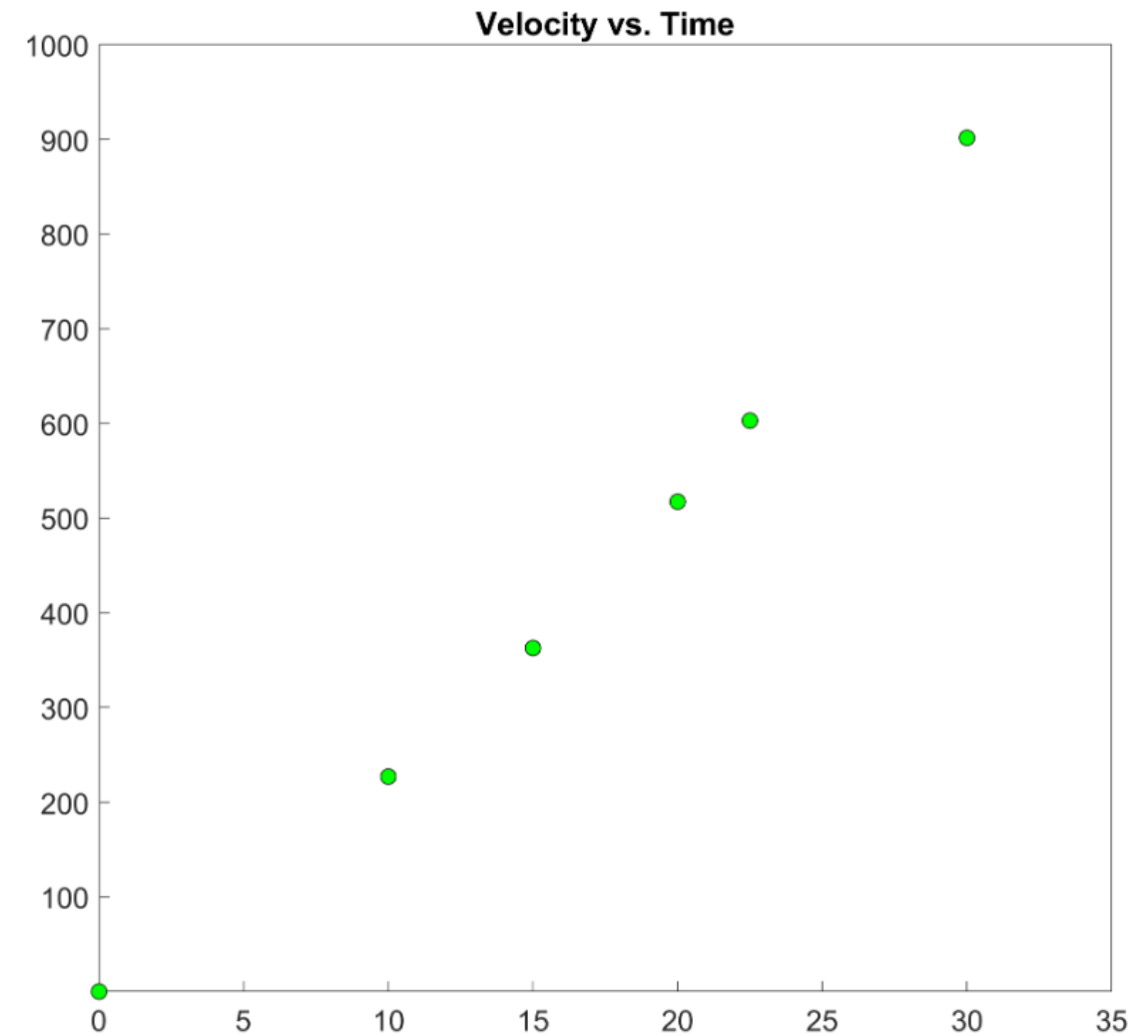
## A second-order polynomial example

The upward velocity of a rocket is given as a function of time in Table 1.

**Table 1.** Velocity as a function of time.

$t$ (s)	$v(t)$ (m/s)
0	0
10	227.04
15	362.78
20	517.35
22.5	602.97
30	901.67

Determine the value of the velocity at  $t = 16$  seconds using second order polynomial interpolation using Newton's divided difference polynomial method.



**Figure 1.** Graph of velocity vs. time data for the rocket example.

# Newton's Divided Difference Interpolation

## A second-order polynomial example

### Solution

For quadratic interpolation, the velocity is given by

$$v(t) = b_0 + b_1(t - t_0) + b_2(t - t_0)(t - t_1)$$

Since we want to find the velocity at  $t = 16$ , and we are using a second order polynomial, we need to choose the three data points that are closest to  $t = 16$  that also bracket  $t = 16$  to evaluate it. The three points are  $t_0 = 10$ ,  $t_1 = 15$ , and  $t_2 = 20$ .

$$t_0 = 10, v(t_0) = 227.04$$

$$t_1 = 15, v(t_1) = 362.78$$

$$t_2 = 20, v(t_2) = 517.35$$

$$\begin{aligned} b_0 &= v(t_0) \\ &= 227.04 \end{aligned}$$

$$\begin{aligned} b_1 &= \frac{v(t_1) - v(t_0)}{t_1 - t_0} \\ &= \frac{362.78 - 227.04}{15 - 10} \\ &= 27.148 \end{aligned}$$

$$\begin{aligned} b_2 &= \frac{\frac{v(t_2) - v(t_1)}{t_2 - t_1} - \frac{v(t_1) - v(t_0)}{t_1 - t_0}}{t_2 - t_0} \\ &= \frac{\frac{517.35 - 362.78}{20 - 15} - \frac{362.78 - 227.04}{15 - 10}}{20 - 10} \\ &= \frac{30.914 - 27.148}{10} \\ &= 0.37660 \end{aligned}$$

# Newton's Divided Difference Interpolation

## A second-order polynomial example

Hence

$$\begin{aligned}v(t) &= b_0 + b_1(t - t_0) + b_2(t - t_0)(t - t_1) \\&= 227.04 + 27.148(t - 10) + 0.37660(t - 10)(t - 15), \quad 10 \leq t \leq 20\end{aligned}$$

At  $t = 16$ ,

$$\begin{aligned}v(16) &= 227.04 + 27.148(16 - 10) + 0.37660(16 - 10)(16 - 15) \\&= 392.19 \text{ m/s}\end{aligned}$$

If we expand

$$v(t) = 227.04 + 27.148(t - 10) + 0.37660(t - 10)(t - 15), \quad 10 \leq t \leq 20$$

we get

$$v(t) = 12.05 + 17.733t + 0.37660t^2, \quad 10 \leq t \leq 20$$

This is the same expression obtained by the direct method.

# Newton's Divided Difference Interpolation

## The general form

Note that  $b_0$ ,  $b_1$ , and  $b_2$  are finite divided differences.  $b_0$ ,  $b_1$ , and  $b_2$  are the first, second, and third finite divided differences, respectively. We denote the first divided difference by

0<sup>th</sup> DD

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

the second divided difference by

1<sup>st</sup> DD

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

and the third divided difference by

2<sup>nd</sup> DD

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

where  $f[x_0]$ ,  $f[x_1, x_0]$ , and  $f[x_2, x_1, x_0]$  are called bracketed functions of their variables enclosed in square brackets.

Rewriting,

$$f_2(x) = f[x_0] + f[x_1, x_0](x - x_0) + f[x_2, x_1, x_0](x - x_0)(x - x_1)$$

This leads us to writing the general form of the Newton's divided difference polynomial for  $n + 1$  data points,

$(x_0, y_0), (x_1, y_1), \dots, (x_{n-1}, y_{n-1}), (x_n, y_n)$ , as

$$f_n(x) = b_0 + b_1(x - x_0) + \dots + b_n(x - x_0)(x - x_1)\dots(x - x_{n-1})$$

# Newton's Divided Difference Interpolation

The general form

where

$$b_0 = f[x_0]$$

$$b_1 = f[x_1, x_0]$$

$$b_2 = f[x_2, x_1, x_0]$$

$$\vdots$$

$$b_{n-1} = f[x_{n-1}, x_{n-2}, \dots, x_0]$$

$$b_n = f[x_n, x_{n-1}, \dots, x_0]$$

where the definition of the  $m^{\text{th}}$  divided difference is

$$\begin{aligned} b_m &= f[x_m, \dots, x_0] \\ &= \frac{f[x_m, \dots, x_1] - f[x_{m-1}, \dots, x_0]}{x_m - x_0} \end{aligned}$$

# Mini Quiz

## Strategy for the implementation of the NDD method

What can be an *optimal approach* to write a computer program that calculates all the divided differences?

Justify your choice.

# Newton's Divided Difference Interpolation

## The general form

$$b_0 = f[x_0]$$

$$b_1 = f[x_1, x_0]$$

$$b_2 = f[x_2, x_1, x_0]$$

$$\vdots$$

$$b_{n-1} = f[x_{n-1}, x_{n-2}, \dots, x_0]$$

$$b_n = f[x_n, x_{n-1}, \dots, x_0]$$

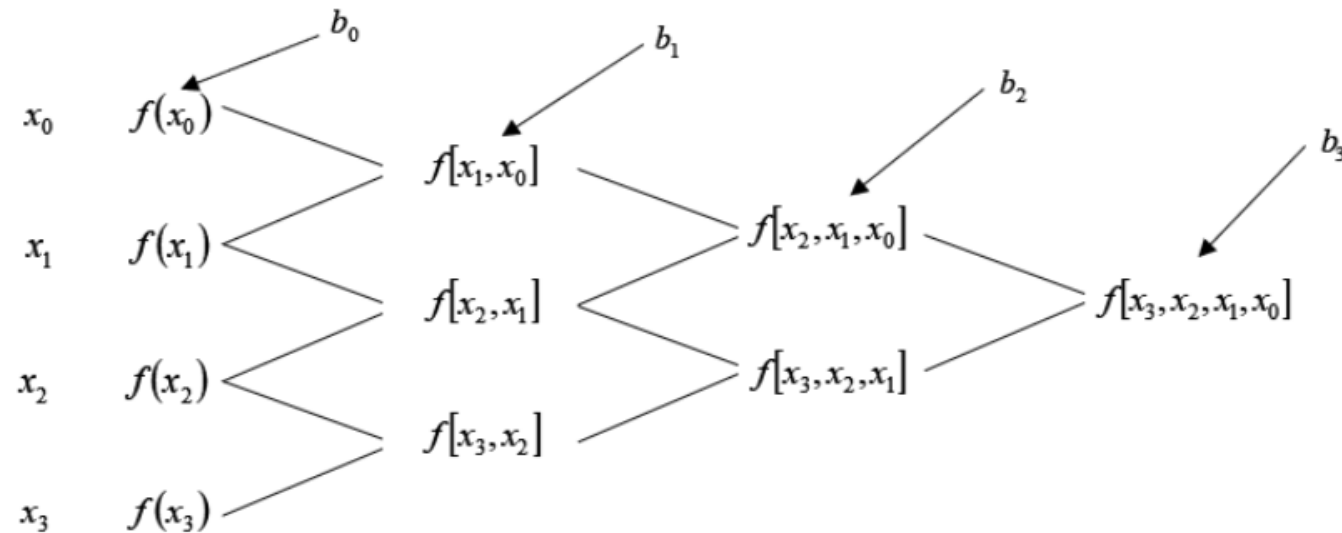
where the definition of the  $m^{\text{th}}$  divided difference is

$$\begin{aligned} b_m &= f[x_m, \dots, x_0] \\ &= \frac{f[x_m, \dots, x_1] - f[x_{m-1}, \dots, x_0]}{x_m - x_0} \end{aligned}$$

From the above definition, it can be seen that the divided differences are calculated recursively.

For an example of a third order polynomial, given  $(x_0, y_0)$ ,  $(x_1, y_1)$ ,  $(x_2, y_2)$ , and  $(x_3, y_3)$ ,

$$\begin{aligned} f_3(x) &= f[x_0] + f[x_1, x_0](x - x_0) + f[x_2, x_1, x_0](x - x_0)(x - x_1) \\ &\quad + f[x_3, x_2, x_1, x_0](x - x_0)(x - x_1)(x - x_2) \end{aligned}$$



**Figure 5** Table of divided differences for a cubic polynomial.



# Newton's Divided Difference Interpolation

## Deriving the 2<sup>nd</sup> order NDD polynomial

Given  $(x_0, y_0)$ ,  $(x_1, y_1)$ , and  $(x_2, y_2)$ , pass a quadratic interpolant through the data. Noting  $y = f(x)$ ,  $y_0 = f(x_0)$ ,  $y_1 = f(x_1)$ , and  $y_2 = f(x_2)$ , assume the quadratic interpolant  $f_2(x)$  is given by

$$f_2(x) = b_0 + b_1(x - x_0) + b_2(x - x_0)(x - x_1) \quad (1)$$

At  $x = x_2$

$$f_2(x_2) = f(x_2) = b_0 + b_1(x_2 - x_0) + b_2(x_2 - x_0)(x_2 - x_1)$$

$$f(x_2) = f(x_0) + \frac{f(x_1) - f(x_0)}{x_1 - x_0}(x_2 - x_0) + b_2(x_2 - x_0)(x_2 - x_1)$$

$$b_2 = \frac{f(x_2) - f(x_0) - \frac{f(x_1) - f(x_0)}{x_1 - x_0}(x_2 - x_0)}{(x_2 - x_0)(x_2 - x_1)}$$

$$= \frac{\frac{f(x_2) - f(x_0)}{x_2 - x_0} - \frac{f(x_1) - f(x_0)}{x_1 - x_0}}{x_2 - x_1}$$

Doesn't match the pattern!

# Newton's Divided Difference Interpolation

## Deriving the 2<sup>nd</sup> order NDD polynomial

But if we want to write this in the form where  $(x_2 - x_0)$  is in the denominator so as to express it in the divided difference form of  $f[x_2, x_1, x_0]$ , we need to do the following manipulations.

Add 0 in the form of  $\{-f(x_1) + f(x_1)\}$  to the numerator of equation (4)

$$b_2 = \frac{f(x_2) + \{-f(x_1) + f(x_1)\} - f(x_0) - \frac{f(x_1) - f(x_0)}{x_1 - x_0}(x_2 - x_0)}{(x_2 - x_0)(x_2 - x_1)}$$

Collecting  $\{f(x_1) - f(x_0)\}$  terms together

$$b_2 = \frac{f(x_2) - f(x_1) + \{f(x_1) - f(x_0)\}\left(1 - \frac{x_2 - x_0}{x_1 - x_0}\right)}{(x_2 - x_0)(x_2 - x_1)}$$

Dividing the numerator and denominator by  $(x_2 - x_1)$

$$b_2 = \frac{\frac{f(x_2) - f(x_1)}{x_2 - x_1} + \frac{\{f(x_1) - f(x_0)\}(x_1 - x_2)}{(x_1 - x_0)(x_2 - x_1)}}{x_2 - x_0}$$

Matches the pattern!

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