

### 3.2 Divided Differences

Previous section (Neville's / Lagrange) focused on interpolation at one point.

Divided difference methods generate the polynomials themselves.

Dof:  $n^{\text{th}}$  Lagrange Polynomial in divided diff form.

$$P_n(x) = a_0 + a_1(x - x_0) + a_2(x - x_0)(x - x_1) + \dots + a_n(x - x_0)(x - x_1) \dots (x - x_{n-1})$$

where the constants  $a_0$  are solved for.

Let's find  $a_0$ . Plug  $x_0$  into  $P_n(x)$  (Note  $P_n(x_0) = f(x_0)$  agree at this point)

$$P_n(x_0) = a_0 = f(x_0)$$

Let's find  $a_1$ . Plug  $x_1$  into  $P_n(x)$  (Remember,  $P_n(x_1) = f(x_1)$ )

$$P_n(x_1) = a_0 + a_1(x_1 - x_0) = f(x_1)$$

$$f(x_0) + a_1(x_1 - x_0) = f(x_1)$$

$$a_1(x_1 - x_0) = f(x_1) - f(x_0)$$

$$\boxed{a_1 = \frac{f(x_1) - f(x_0)}{x_1 - x_0}}$$

Let's find  $a_2$ . Plug  $x_2$  into  $P_n(x)$  [Remember  $P_n(x_2) = f(x_2)$ ]

$$P_n(x_2) = a_0 + a_1(x_2 - x_0) + a_2(x_2 - x_0)(x_2 - x_1) = f(x_2)$$

At this point, it becomes more difficult (but possible) to do the algebra involved. However, we introduce a new notation:

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

In particular, if we know the divided diff

$f[x_i, x_{i+1}, \dots, x_{i+k-1}]$  and  $f[x_{i+1}, x_{i+2}, \dots, x_{i+k}]$ , then

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

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With this, we can solve for each of the  $a_k$ 's easier.

So, for example, back to the example

$$P_2(x) = f(x_2) = a_0 + a_1(x_2 - x_0) + a_2(x_2 - x_0)(x_2 - x_1)$$

$\swarrow$  same  $\Rightarrow$  see def of  $a_1$  &  $f[x_0, x_1]$

$$f(x_2) = f[x_0] + f[x_0, x_1](x_2 - x_0) + a_2(x_2 - x_0)(x_2 - x_1)$$

$\downarrow$  notation

$$f[x_2] - f[x_0] - f[x_0, x_1](x_2 - x_0) = a_2(x_2 - x_0)(x_2 - x_1)$$

$$f[x_2] - f[x_0] + f[x_0] - f[x_0, x_1](x_2 - x_0) = a_2(x_2 - x_0)(x_2 - x_1)$$

trick

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

$$f[x_1, x_2](x_2 - x_1) + \underbrace{f[x_0, x_1](x_1 - x_0)}_{x_0\text{'s cancel.}} - f[x_0, x_1](x_2 - x_0) = a_2(x_2 - x_0)(x_2 - x_1)$$

$$f[x_1, x_2](x_2 - x_1) + f[x_0, x_1](x_1 - x_2) = a_2(x_2 - x_0)(x_2 - x_1)$$

So

$$a_2 = \frac{f[x_1, x_2](x_2 - x_1) + f[x_0, x_1](x_1 - x_2)}{(x_2 - x_0)(x_2 - x_1)} \stackrel{= (-x_2 - x_1)}{\leftarrow}$$

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

∴

In general,

$$P_n(x) = a_0 + a_1(x-x_0) + a_2(x-x_0)(x-x_1) + \cdots + a_n(x-x_0)(x-x_1)\cdots(x-x_{n-1})$$

$$= f[x_0] + \sum_{k=1}^n f[x_0, x_1, \dots, x_k](x-x_0)\cdots(x-x_{k-1})$$

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### Divided Difference Table

X	f(x)	First Divided Differences	Second DD
$x_0$	$f[x_0]$		
$x_1$	$f[x_1]$	$f[x_0, x_1] = \frac{f[x_1] - f[x_0]}{x_1 - x_0}$	$f[x_0, x_1, x_2] = \frac{f[x_1, x_2] - f[x_0, x_1]}{x_2 - x_0}$
		$f[x_1, x_2] = \frac{f[x_2] - f[x_1]}{x_2 - x_1}$	$\vdots$

$$\begin{array}{ll}
 x_2 & f[x_2] = \frac{x_2 - x_1}{x_2 - x_3} \\
 & f[x_2, x_3] = \frac{f[x_3] - f[x_2]}{x_3 - x_2} \\
 x_3 & f[x_3] = \frac{x_3 - x_2}{x_3 - x_4} \\
 & f[x_3, x_4] = \frac{f[x_4] - f[x_3]}{x_4 - x_3} \\
 x_4 & f[x_4] = \frac{x_4 - x_3}{x_4 - x_5} \\
 & f[x_4, x_5] = \frac{f[x_5] - f[x_4]}{x_5 - x_4} \\
 x_5 & f[x_5]
 \end{array}$$

$$\begin{aligned}
 f[x_1, x_2, x_3] &= \frac{f[x_2, x_3] - f[x_1, x_2]}{x_3 - x_1} \\
 f[x_2, x_3, x_4] &= \frac{f[x_3, x_4] - f[x_2, x_3]}{x_4 - x_2} \\
 f[x_3, x_4, x_5] &= \frac{f[x_4, x_5] - f[x_3, x_4]}{x_5 - x_3}
 \end{aligned}$$

you can also define

3 Third DD

2 Fourth DD

\* 1 Fifth DD

### Algorithm 32 Newton's Interpolatory Divided-Diff Formula

Note that  $P(x) = \sum_{i=0}^n F_{i,i} \prod_{j=0}^{i-1} (x - x_j)$   $\Rightarrow$  we're after the numbers on the diagonal. ( $F_{i,0} = f(x_i)$ )  
outputs the polynomial.

Thm 3.6

Suppose  $f \in C^n[a, b]$ ,  $x_i$ 's are distinct  $\in [a, b]$ .  $\exists \xi \in$

$$f[x_0, x_1, \dots, x_n] = \frac{f^{(n)}(\xi)}{n!}$$

Now, reformulate

$$(*) \quad P_n(x) = f[x_0] + \sum_{k=1}^n f[x_0, x_1, \dots, x_k] (x - x_0) \cdots (x - x_{k-1})$$

with equal spacing Let  $h = x_{i+1} - x_i$  (common for all)

Let  $x = x_0 + sh$  (point to interpolate)

$x_i = x_0 + ih$  and

$$x - x_i = (s - i)h$$

Then (\*) becomes

$$\begin{aligned} P_n(x) &= P_n(x_0 + sh) = f[x_0] + f[x_0, x_1](x - x_0) + f[x_0, x_1, x_2](x - x_0)(x - x_1) \\ &\quad + \cdots + f[x_0, x_1, \dots, x_k](x - x_0)(x - x_1) \cdots (x - x_{k-1}) + \cdots + \text{last one}. \end{aligned}$$

Note:  $(x - x_0)(x - x_1) \cdots (x - x_{k-1}) = (sh)(s-1)h(s-2)h \cdots (s-k+1)h$

$$= s(s-1) \dots (s-k+1) h^k$$

To simplify notation, define an extension of the binomial coefficient

as  $\binom{s}{k} = \frac{s(s-1)\dots(s-k+1)}{k!}$ , where  $s \in \mathbb{R}$  (any real #)

Thus,

$$(x-x_0)\dots(x-x_{k-1}) = \binom{s}{k} k! h^k \text{ and}$$

$$P_n(x) = \sum_{k=0}^n f[x_0, \dots, x_k] \binom{s}{k} k! h^k$$

Newton's Forward divided diff formula

If we use Aiken's  $\Delta^2$  operator, we can make a simplification to the notation for the formula. First, note

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

$$f[x_0, x_1, x_2] = \frac{f[x_1, x_2] - f[x_0, x_1]}{x_2 - x_0} = \frac{\frac{\Delta f(x_1)}{h} - \frac{\Delta f(x_0)}{h}}{2h} = \frac{\Delta^2 f(x_0)}{2h^2}$$

In general

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

becomes

$$P_n(x) = \sum_{k=0}^n f[x_0, x_1, \dots, x_k] \binom{s}{k} k! h^k$$

$$\boxed{P_n(x) = \sum_{k=0}^n \binom{s}{k} \Delta^k f(x_0)}$$

Newton  
Forward-Difference  
formula.

You can also reorder the indices from the back to the front.  $x_n, x_{n-1}, \dots, x_0$ . In this case, we get

Newton's backward divided difference formula

Newton backward divided difference formula

$$P_n(x) = f[x_n] + f[x_{n-1}, x_n](x-x_n) + f[x_{n-2}, x_{n-1}, x_n](x-x_n)(x-x_{n-1}) \\ + \dots + f[x_0, \dots, x_n](x-x_n)(x-x_{n-1}) \dots (x-x_1)$$

Using equal spacing (like before) yields

$$P_n(x) = f[x_n] + s_0 h f[x_{n-1}, x_n] + s_1(s+1) h^2 f[x_{n-2}, x_{n-1}, x_n] + \dots + s_{(s+1)} \dots (s+n-1) h^n f[x_0, \dots, x_n]$$

DEF:  $\nabla P_n \equiv P_n - P_{n-1}$  and

Backwards

DIFF operator  $\nabla^k P_n = \nabla(\nabla^{k-1} P_n)$

This makes  $f[x_{n-1}, x_n] = \frac{1}{n} \nabla f(x_n)$ ,  $f[x_{n-2}, x_{n-1}, x_n] = \frac{1}{2h^2} \nabla^2 f(x_n)$

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

which means

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

Note that the binomial coefficient idea doesn't seem to work because the multiply is going up (not down).  
But it can still be done using the same extension!

So

$$\binom{-s}{k} = \frac{-s(-s-1)(-s-2)\dots(-s-k+1)}{k!} = \frac{(-1)^k s(s+1)(s+2)\dots(s+k-1)}{k!}$$

Thus, Newton's Backward Difference formula is

$$P_n(x) = \sum_{k=0}^n (-1)^k \binom{-s}{k} \nabla^k f(x_n)$$

$$P_n(x) = \sum_{k=0}^{\infty} (-1)^k (F)^k T(x_k)$$

Examples

Choose Normal table from 1 to 2 separated by .2