# Hw # 2 Solutions

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## Chapter 2

#### Section 2.2

1.

$$f(x) \approx \frac{1}{x_1 - x_0} \begin{vmatrix} f(x_0) & x_0 - x \\ f(x_1) & x_1 - x \end{vmatrix}$$
 (1)

Use (1) to calculate approximate values of f(x) when x = 1.1416, 1.1600, and 1.2000 from the following rounded data:

For  $\hat{x} = 1.1416$ , choose  $x_0 = 1.1275$  and  $x_1 = 1.1503$  so  $\hat{x} \in (x_0, x_1)$ . Thus

$$f(\hat{x}) \approx \frac{1}{x_1 - x_0} \begin{vmatrix} f(x_0) & x_0 - \hat{x} \\ f(x_1) & x_1 - \hat{x} \end{vmatrix}$$

$$= \frac{1}{1.1503 - 1.1275} \begin{vmatrix} 0.11971 & 1.1275 - 1.1416 \\ 0.13957 & 1.1503 - 1.1416 \end{vmatrix}$$

$$= \frac{1}{0.0228} [(0.11971)(0.0087) - (-0.0141)(0.13957)]$$

$$\approx \boxed{0.1320}$$

For  $\hat{x} = 1.1600$ , choose  $x_0 = 1.1503$  and  $x_1 = 1.1735$  so  $\hat{x} \in (x_0, x_1)$ . Thus

$$f(\hat{x}) \approx \frac{1}{x_1 - x_0} \begin{vmatrix} f(x_0) & x_0 - \hat{x} \\ f(x_1) & x_1 - \hat{x} \end{vmatrix}$$

$$= \frac{1}{1.1735 - 1.1503} \begin{vmatrix} 0.13957 & 1.1503 - 1.1600 \\ 0.15931 & 1.1735 - 1.1600 \end{vmatrix}$$

$$= \frac{1}{0.0232} [(0.13957)(0.0135) - (-0.0097)(0.15931)]$$

$$\approx \boxed{0.1478}$$

For  $\hat{x} = 1.2000$ , choose  $x_0 = 1.1735$  and  $x_1 = 1.1972$  because those are the two closes data points to  $\hat{x}$ . However, since  $\hat{x} \notin (x_0, x_1)$ , we cannot guarantee an upper bound on the error.

$$f(\hat{x}) \approx \frac{1}{x_1 - x_0} \begin{vmatrix} f(x_0) & x_0 - \hat{x} \\ f(x_1) & x_1 - \hat{x} \end{vmatrix}$$

$$= \frac{1}{1.1972 - 1.1735} \begin{vmatrix} 0.15931 & 1.1735 - 1.2000 \\ 0.17902 & 1.1972 - 1.2000 \end{vmatrix}$$
$$= \frac{1}{0.0237} [(0.15931)(-0.0028) - (-0.0265)(0.17902)]$$
$$\approx \boxed{0.1813}$$

**2**.

$$f(x) \approx f(x_0) + (x - x_0)f[x_0, x_1] \tag{2}$$

Calculate the three first divided differences relevant to successive pairs of data in Problem 1, and use (2) to determine approximate values of f(x) for

$$x \in \mathcal{X} = \{x \in [1.1600, 1.1700] \mid x = 1.1600 + 0.0020k, \ k \in \mathbb{Z}\}\$$
  
=  $\{1.1600, 1.1620, 1.1640, 1.1660, 1.1680, 1.1700\}$ 

The definition of first divided difference is

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

The data points in Problem 1 are  $\{(1.1275, 0.11971), (1.1503, 0.13957), (1.1735, 0.15931), (1.1972, 0.17902)\}$ . So,

$$f_1 = f[1.1275, 1.1503] = \frac{0.13957 - 0.11971}{1.1503 - 1.1275} \approx 0.8711$$

$$f_2 = f[1.1503, 1.1735] = \frac{0.15931 - 0.13957}{1.1735 - 1.1503} \approx 0.8509$$

$$f_3 = f[1.1735, 1.1972] = \frac{0.17902 - 0.15931}{1.1972 - 1.1735} \approx 0.8316$$

Since  $\mathcal{X} \subset (1.1503, 1.1735)$ , we use  $f_2$  to linearly interpolate the values of  $x \in \mathcal{X}$ . Per equation (2),

$$f(1.1600) \approx 0.13957 + (1.1600 - 1.1503)(0.8509) \approx 0.1478$$
  
 $f(1.1620) \approx 0.13957 + (1.1620 - 1.1503)(0.8509) \approx 0.1495$   
 $f(1.1640) \approx 0.13957 + (1.1640 - 1.1503)(0.8509) \approx 0.1512$   
 $f(1.1660) \approx 0.13957 + (1.1660 - 1.1503)(0.8509) \approx 0.1529$   
 $f(1.1680) \approx 0.13957 + (1.1680 - 1.1503)(0.8509) \approx 0.1546$   
 $f(1.1700) \approx 0.13957 + (1.1700 - 1.1503)(0.8509) \approx 0.1563$ 

Prove that  $f[x_0, x_1]$  is independent of  $x_0$  and  $x_1$  if and only if f(x) is a linear function of x.

Suppose f(x) is a linear function. That is, f(x) = mx + b for some constants m and b. Then choose two arbitrary points,  $x_0$  and  $x_1$ .

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

$$= \frac{mx_1 + b - mx_0 - b}{x_1 - x_0}$$

$$= \frac{m(x_1 - x_0)}{x_1 - x_0}$$

$$= m$$

Since m is a given constant,  $f[x_0, x_1]$  is independent of  $x_0$  and  $x_1$ .

Now suppose  $f[x_0, x_1]$  is independent of  $x_0$  and  $x_1$ . That is,  $f[x_0, x_1]$  is a constant  $\forall x_0, x_1 \in \mathbb{R}$ . Let  $f[x_0, x_1] = m$  be that constant. Then

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

$$\implies f(x_1) - f(x_0) = m(x_1 - x_0)$$

This is the *point-slope form* of a line, and thus f(x) is a linear function of x.

4.

If f(x) = u(x)v(x), show that

$$f[x_0, x_1] = u[x_0]v[x_0, x_1] + u[x_0, x_1]v[x_1]$$

Using the definition of the first divided difference,

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

$$= \frac{u(x_1)v(x_1) - u(x_0)v(x_0)}{x_1 - x_0}$$

$$= \frac{u(x_1)v(x_1) - v(x_1)u(x_0) + v(x_1)u(x_0) - u(x_0)v(x_0)}{x_1 - x_0}$$

$$= \frac{u(x_0)(v(x_1) - v(x_0)) + v(x_1)(u(x_1) - u(x_0))}{x_1 - x_0}$$

$$= u(x_0)\frac{v(x_1) - v(x_0)}{x_1 - x_0} + v(x_1)\frac{u(x_1) - u(x_0)}{x_1 - x_0}$$

$$= u[x_0]v[x_0, x_1] + u[x_0, x_1]v[x_1]$$

**5**.

If f'(x) is continuous for  $x_0 \le x \le x_1$ , show that

$$f[x_0, x_1] = f'(\xi)$$

for some  $\xi$  between  $x_0$  and  $x_1$ , and hence also that

$$f[x_0, x_0] = \lim_{x_1 \to x_0} f[x_0, x_1] = f'(x_0)$$

Let f'(x) be continuous for  $x \in [x_0, x_1]$ . Then f(x) exists and is continuous on that interval. Then by the *Mean Value Theorem for Derivatives*,  $\exists \xi \in (x_0, x_1)$  such that  $f(x_1) - f(x_0) = (x_1 - x_0)f'(\xi)$ . In other words,

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

Now consider the constant sequence  $A = (a_i) = (x_0, x_0, ...)$  and the sequence  $B = (b_i) = (x_{11}, x_{12}, x_{13}, ...)$  where  $x_1 = x_{11}$  and  $a_i < b_i \ \forall i \in \mathbb{N}$ . Let B converge to  $x_0$ , and note that A also converges to  $x_0$ . Using the Mean Value Theorem for Derivatives for corresponding intervals  $(\min\{a_i, b_i\}, \max\{a_i, b_i\})$  define  $C = (\xi_i)$  where  $\xi_i$  is the value for which  $f(b_i) - f(a_i) = (b_i - a_i)f'(\xi_i)$ . By the Squeeze Theorem, C converges to  $x_0$ . Thus,

$$f[x_0, x_0] = \lim_{x_1 \to x_0} f[x_0, x_1] = \lim_{i \to \infty} f[a_i, b_i] = \lim_{i \to \infty} f'(\xi_i) = f'(x_0)$$

#### Section 2.3

7.

$$\alpha_i^{(k)} = \frac{1}{(x_i - x_0) \dots (x_i - x_{i-1})(x_i - x_{i+1}) \dots (x_i - x_k)}$$
(3)

$$f[x_0, \dots, x_k] = \sum_{i=0}^k \alpha_i^{(k)} f(x_i)$$
 (4)

Suppose that  $x_r = x_0 + rh$ , (r = 1, 2, ...), so that the abcissas are at a uniform spacing h. Show that (3) then becomes

$$\alpha_i^{(k)} = \frac{(-1)^{k-i}}{i!(k-i)!} \frac{1}{h^k} = \frac{(-1)^{(k-i)}}{h^k k!} \binom{k}{i}$$

where  $\binom{k}{i}$  is the binomial coefficient. Thus deduce that

$$f[x_0, \dots, x_k] = \frac{1}{h^k k!} \sum_{i=0}^k (-1)^{k-i} \binom{k}{i} f(x_i)$$
 (5)

in this case.

Supposing  $x_r = x_0 + rh$  for r = 1, 2, ..., then specifically  $x_i = x_0 + ih$ , and by (3),

$$\alpha_{i}^{(k)} = \frac{1}{\left([ih][(i-1)h][(i-2)h]\dots[2h][h]\right)\left([-h][-2h]\dots[-(k-i-1)h][-(k-i)h]\right)}$$

$$= \frac{1}{\left(i!h^{i}\right)\cdot\left((-1)^{k-i}h^{k-i}(k-i)!\right)}$$

$$= \frac{(-1)^{k-i}}{i!(k-i)!h^{k}}$$

$$= \frac{(-1)^{k-i}}{k!h^{k}}\cdot\frac{k!}{i!(k-i)!}$$

$$= \frac{(-1)^{k-i}}{k!h^{k}}\binom{k}{i}$$

where  $\binom{k}{i}$  is the binomial coefficient given by  $\frac{k!}{i!(k-i)!}$ . Then by (4),

$$f[x_0, \dots, x_k] = \sum_{i=0}^k \frac{(-1)^{k-i}}{k!h^k} \binom{k}{i} f(x_i)$$
$$= \frac{1}{h^k k!} \sum_{i=0}^k (-1)^{k-i} \binom{k}{i} f(x_i)$$

9.

If 
$$f'(x) = \frac{df(x)}{dx}$$
, show that

$$\frac{d}{dx}f[x_0, x] \not\equiv f'[x_0, x]$$

unless f(x) is linear.

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

$$\implies \frac{d}{dx} f[x_0, x] = \frac{(x - x_0)(f'(x)) - (f(x) - f(x_0))}{(x - x_0)^2}$$

$$= \frac{1}{x - x_0} \left( f'(x) - \frac{f(x) - f(x_0)}{x - x_0} \right)$$

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

Setting these two equal yields

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

$$\implies f[x_0, x] = f'(x_0)$$

But  $f'(x_0)$  is a constant, and thus  $f[x_0, x]$  must be constant in order for the two quantities to be equal. However,  $f[x_0, x]$  is only constant if f(x) is a linear function of x (i.e. f(x) = mx + b for some  $m, b \in \mathbb{R}$ ). Thus  $\frac{d}{dx}f[x_0, x] \not\equiv f'[x_0, x]$  unless f(x) is linear.

#### 11.

If  $f(x) = \frac{ax + b}{cx + d}$ , obtain expressions for f[x, y], f[x, x, y], and f[x, x, y, y] in compact forms when  $x \neq y$ .

First note that  $f[x,x] \equiv f'(x) = \frac{ad - bc}{(cx+d)^2}$ . Then,

$$f[x,y] = \frac{\frac{ax+b}{cx+d} - \frac{ay+b}{cy+d}}{x-y} \cdot \frac{(cx+d)(cy+d)}{(cx+d)(cy+d)}$$

$$= \frac{(ax+b)(cy+d) - (ay+b)(cx+d)}{(x-y)(cx+d)(cy+d)}$$

$$= \frac{adx+bcy-ady-bcx}{(x-y)(cx+d)(cy+d)}$$

$$= \frac{ad-bc}{(cx+d)(cy+d)}$$

$$\Rightarrow f[x,x,y] = \frac{f[x,x] - f[x,y]}{x-y}$$

$$= \frac{\frac{ad-bc}{(cx+d)^2} - \frac{ad-bc}{(cx+d)(cy+d)}}{x-y}$$

$$= \frac{(ad-bc)(cy+d) - (ad-bc)(cx+d)}{(x-y)(cx+d)^2(cy+d)}$$

$$= \frac{c(bc-ad)}{(cx+d)^2(cy+d)}$$

Similarly, 
$$f[x, y, y] = \frac{c(bc - ad)}{(cx + d)(cy + d)^2}$$
. Thus,  

$$f[x, x, y, y] = \frac{f[x, x, y] - f[x, y, y]}{x - y}$$

$$= \frac{c(bc - ad)}{\frac{(cx + d)^2(cy + d)}{(cx + d)(cy + d)^2}} - \frac{c(bc - ad)}{\frac{(cx + d)(cy + d)^2}{(cx + d)(cy + d)^2}}$$

$$= \frac{c(bc - ad)(cy + d) - c(bc - ad)(cx + d)}{(x - y)(cx + d)^2(cy + d)^2}$$

$$= \frac{c^2(ad - bc)}{(cx + d)^2(cy + d)^2}$$

#### Section 2.5

23.

If  $f(x_1)$ ,  $f(x_2)$ , and  $f(x_3)$  are values of f(x) near a maximum or minimum point at  $x = \overline{x}$ , obtain the approximation

$$\overline{x} \approx \frac{x_1 + x_2}{2} - \frac{f[x_1, x_2]}{2f[x_1, x_2, x_3]}$$

and show that it can also be written in the more symmetrical form

$$\overline{x} \approx \frac{x_1 + 2x_2 + x_3}{4} - \frac{f[x_1, x_2] + f[x_2, x_3]}{4f[x_1, x_2, x_3]}$$

Show also that, when the abscissas are equally spaced, it becomes

$$\overline{x} \approx x_2 - \frac{h}{2} \left( \frac{f_3 - f_1}{f_1 - 2f_2 + f_3} \right)$$

where h is the common interval. Note  $f_i := f(x_i)$ .

We can use the following polynomial approximation for f(x):

$$f(x) \approx p_{1,2,3}(x) = f_1 + (x - x_1)f[x_1, x_2] + (x - x_1)(x - x_2)f[x_1, x_2, x_3]$$
 (6)

$$\implies \frac{d}{dx}p_{1,2,3}(x) = p'_{1,2,3}(x) = f[x_1, x_2] + [(x - x_1) + (x - x_2)]f[x_1, x_2, x_3]$$

Since  $\overline{x}$  is a minimum or maximum abcissa, then  $f'(\overline{x}) = 0$ , so

$$p'_{1,2,3}(\overline{x}) \approx 0$$

$$\implies f[x_1, x_2] + [(\overline{x} - x_1) + (\overline{x} - x_2)]f[x_1, x_2, x_3] \approx 0$$

Solving for  $\overline{x}$  yields

$$\overline{x} \approx \boxed{\frac{x_1 + x_2}{2} - \frac{f[x_1, x_2]}{2f[x_1, x_2, x_3]}}$$
 (7)

The definition of  $p_{1,2,3}$  in (6) is not unique, however, since the sum of n divided differences up to the n<sup>th</sup> divided difference can be summed in  $2^n$  ways. So define another polynomial approximation for f(x):

$$f(x) \approx \overline{p_{1,2,3}}(x) = f_3 + (x - x_3)f[x_2, x_3] + (x - x_2)(x - x_3)f[x_1, x_2, x_3]$$

$$\implies \frac{d}{dx}\overline{p_{1,2,3}}(x) = \overline{p'_{1,2,3}}(x) = f[x_2, x_3] + [(x - x_2) + (x - x_3)]f[x_1, x_2, x_3]$$

Again, since  $\overline{x}$  is a minimum or maximum abscissa, then  $f'(\overline{x}) = 0$ . Thus

$$\overline{p'_{1,2,3}}(\overline{x}) \approx 0$$

$$\implies f[x_2, x_3] + [(\overline{x} - x_2) + (\overline{x} - x_3)]f[x_1, x_2, x_3] \approx 0$$

Solving for  $\overline{x}$  yields

$$\overline{x} \approx \frac{x_2 + x_3}{2} - \frac{f[x_2, x_3]}{2f[x_1, x_2, x_3]}$$
 (8)

Since both (7) and (8) are valid approximations of  $\overline{x}$ , their arithmetic mean is also valid:

$$\overline{x} \approx \frac{\frac{x_1 + x_2}{2} - \frac{f[x_1, x_2]}{2f[x_1, x_2, x_3]} + \frac{x_2 + x_3}{2} - \frac{f[x_2, x_3]}{2f[x_1, x_2, x_3]}}{2}$$

$$\approx \left[ \frac{x_1 + 2x_2 + x_3}{4} - \frac{f[x_1, x_2] + f[x_2, x_3]}{4f[x_1, x_2, x_3]} \right]$$

Now suppose the abscissas are equally spaced, and h is the common difference  $x_{i+1} - x_i$ . Thus  $x_1 = x_2 - h$  and  $x_3 = x_2 + h$ . Then

$$\frac{x_1 + 2x_2 + x_3}{4} = \frac{x_2 - h + 2x_2 + x_2 + h}{4} = x_2$$

and

$$f[x_1, x_2] + f[x_2, x_3] = \frac{f_2 - f_1}{h} + \frac{f_3 - f_2}{h}$$
$$= \frac{f_3 - f_1}{h}$$

and

$$4f[x_1, x_2, x_3] = 4 \cdot \frac{f[x_2, x_3] - f[x_1, x_2]}{2h}$$
$$= \frac{2}{h} \left( \frac{f_3 - f_2}{h} - \frac{f_2 - f_1}{h} \right)$$
$$= \frac{2}{h^2} (f_1 - 2f_2 + f_3)$$

Thus,

$$\overline{x} \approx \frac{x_1 + 2x_2 + x_3}{4} - \frac{f[x_1, x_2] + f[x_2, x_3]}{4f[x_1, x_2, x_3]}$$
$$= \overline{\left[x_2 - \frac{2}{h} \left(\frac{f_3 - f_1}{f_1 - 2f_2 + f_3}\right)\right]}$$

### Section 2.6

#### 24.

Show that the truncation error associated with linear interpolation of f(x), using ordinates at  $x_0$  and  $x_1$  with  $x_0 \le x \le x_1$ , is not larger in magnitude than

$$\frac{1}{8}M_2(x_1 - x_0)^2$$

where  $M_2$  is the maximum value of |f''(x)| on the interval  $[x_0, x_1]$ . Does this result hold also for extrapolation?

For n=1, the absolute value of the error is bounded by

$$|E(x)| \le \frac{M_2}{(1+1)!} |\pi(x)|$$

where  $M_2$  is the maximum value of |f''(x)| on  $[x_0, x_1]$ . But  $|\pi(x)| = |x - x_0| \cdot |x - x_1|$ , so

$$|E(x)| \le \frac{M_2}{2}|x - x_0| \cdot |x - x_1|$$

But since the sum of  $x_1 - x$  and  $x - x_0$  are constant, their product is maximized when they coincide, which is where x is at the midpoint of  $x_0$  and  $x_1$ , and thus the difference is half the distance from  $x_0$  to  $x_1$ , i.e.  $\frac{x_1 - x_0}{2}$ . Thus

$$|x - x_0| \cdot |x - x_1| \le \left(\frac{x_1 - x_0}{2}\right)^2$$

$$\implies |E(x)| \le \frac{M_2}{2} \left(\frac{x_1 - x_0}{2}\right)^2$$
$$= \frac{1}{8} M_2 (x_1 - x_0)^2$$

The argument presented above does *not* hold for extrapolation because if x is not between  $x_0$  and  $x_1$  then  $|x - x_0| \cdot |x - x_1|$  is not bounded, and is certainly not bounded by  $\left(\frac{x_1 - x_0}{2}\right)^2$ . As a counterexample, consider  $x_0 = 4$ ,  $x_1 = 10$ , and x = 12. Then  $\left(\frac{x_1 - x_0}{2}\right)^2 = \frac{36}{4} = 9$ , but  $|12 - 4| \cdot |12 - 10| = 8 \cdot 2 = 16 \nleq 9$ .

#### 29.

Obtain the formula

$$f(x) = f(x_0) + (x - x_0)f'(x_0) + (x - x_0)^2 f[x_0, x_0, x_1] + (x - x_0)^2 (x - x_1)f[x_0, x_0, x_1, x_1] + E(x)$$
where

$$E(x) = \frac{1}{24}(x - x_0)^2(x - x_1)^2 f^{iv}(\xi) \qquad (x_0 < x, \xi < x_1)$$

and show that

$$|E(x)| \le \frac{h^4}{384} \max_{x_0 \le x \le x_1} |f^{iv}(x)|$$

Newton's Fundamental Formula gives

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

where

$$E(x) = (x - x_0) \dots (x - x_n) f[x_0, \dots, x_n, x]$$

Suppose  $x_0$  and  $x_1$  coincide, and  $x_2$  and  $x_3$  coincide. For ease, let  $x_0$  and  $x_1$  be those values, respectively. Then

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

where

$$E(x) = (x - x_0)(x - x_0)(x - x_1)(x - x_1)f[x_0, x_0, x_1, x_1, x]$$

Thus,

$$f(x) = f(x_0) + (x - x_0)f'(x_0) + (x - x_0)^2 f[x_0, x_0, x_1] + (x - x_0)^2 (x - x_1)f[x_0, x_0, x_1, x_1] + E(x)$$
where

$$E(x) = (x - x_0)^2 (x - x_1)^2 f[x_0, x_0, x_1, x_1, x]$$

By Rolle's Theorem,  $\exists \xi \in [\min(x_0, x_1), \max(x_0, x_1)]$  such that

$$f[x_0, x_0, x_1, x_1, x] = \frac{1}{4!} f^{iv}(\xi) = \frac{1}{24} f^{iv}(\xi)$$

$$\implies E(x) = \frac{1}{24} (x - x_0)^2 (x - x_1)^2 f^{iv}(\xi)$$

By the same argument made in Problem 24,  $(x-x_0)^2(x-x_1)^2 \le \left(\frac{x_1-x_0}{2}\right)^4$ . Thus

$$E(x) \le \frac{1}{24} \cdot \frac{1}{16} (x_1 - x_0)^4 f^{iv}(\xi)$$

$$\le \frac{1}{384} (x_1 - x_0)^4 \max_{x_0 \le x \le x_1} |f^{iv}(x)|$$

$$= \frac{h^4}{384} \max_{x_0 \le x \le x_1} |f^{iv}(x)|$$

where h is the constant distance between  $x_0$  and  $x_1$ , namely  $|x_1 - x_0|$ .

30.

$$E(x) = \frac{1}{(n+1)!} f^{(n+1)}(\xi) \pi(x)$$
(9)

$$E(x) = \pi(x)f[x_0, \dots, x_n, x]$$

$$\tag{10}$$

If  $f(x) = \frac{1}{x+1}$  and y(x) is the polynomial approximation of degree n which agrees with f(x) when  $x = 0, 1, 2, \ldots, n$ , show that the use of (9) leads to the error bound

$$|E(x)| < |x(x-1)\dots(x-n)|$$

whereas (10) permits the less conservative bound

$$|E(x)| < \frac{1}{(n+1)!} |x(x-1)\dots(x-n)|$$

when  $x \geq 0$ 

First, note the following:

$$f^{(n+1)}(x) = \frac{(-1)^{n+1}(n+1)!}{(x+1)^{n+2}}$$

$$\implies f^{(n+1)}(\xi) = \frac{(-1)^{n+1}(n+1)!}{(\xi+1)^{n+2}}$$

Thus, by (9),

$$|E(x)| = \left| \frac{1}{(n+1)!} \cdot \frac{(-1)^{n+1}(n+1)!}{(\xi+1)^{n+2}} \cdot \pi(x) \right|$$
$$= \left| \frac{1}{(\xi+1)^{n+2}} \cdot (x-x_0)(x-x_1) \dots (x-x_n) \right|$$

$$x_i = i \text{ for } i = 0, 1, 2, \dots, n \implies \xi \in (0, n) \implies \xi + 1 \ge 1 \implies \left| \frac{1}{(\xi + 1)^{n+2}} \right| < 1. \text{ Thus,}$$

$$|E(x)| < |x(x - 1) \dots (x - n)|$$

An inductive attempt is made to prove the 'less conservative' bound

$$|E(x)| < \frac{1}{(n+1)!} |x(x-1)\dots(x-n)|$$

For n = 0,  $|f[x_0, x]| = \left| \frac{f(x) - f(x_0)}{x - x_0} \right| = \left| \frac{1 - (x+1)}{x(x+1)} \right| = \frac{1}{x+1}$ . If x = 0, the error is guaranteed to be 0, which is certainly less than  $\frac{1}{(0+1)!} = 1$ . If x > 0, then  $\frac{1}{x+1} < 1 = \frac{1}{(0+1)!}$ . Then by (10),

$$|E(x)| < \frac{1}{(0+1)!}|x| = |x|$$

Now assume  $\forall k \in \{0, \ldots, n-1\}$ ,  $|f[x_0, \ldots, x_k, x]| < \frac{1}{(k+1)!}$ , which would, by (10), show that  $|E(x)| < \frac{1}{(k+1)!}|x(x-1)\ldots(x-k)|$ . Also note that because our data points are equally spaced, (5) holds (with h=1), and thus  $f[x_0, \ldots, x_n] = \frac{1}{n!} \sum_{i=0}^n (-1)^{n-i} \binom{n}{i} f(x_i)$ . Thus,

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

$$< \frac{\frac{1}{n!} - \frac{1}{n!} \sum_{i=0}^{n} \left[ (-1)^{n-i} \binom{n}{i} f(i) \right]}{x - n}$$

$$= \frac{\frac{1}{n!} - \frac{1}{n!} \sum_{i=0}^{n} \left[ (-1)^{n-i} \frac{n!}{i!(n-i)!} \frac{1}{i+1} \right]}{x - n}$$

$$= \frac{\frac{1}{n!} - \sum_{i=0}^{n} \left[ (-1)^{n-i} \frac{1}{(i+1)!(n-i)!} \right]}{x - n}$$

$$= \frac{\frac{1}{n!} - \frac{1}{(n+1)!} \sum_{i=0}^{n} \left[ (-1)^{n-i} \binom{n+1}{i+1} \right]}{x - n}$$

At this time, I am unable to adequately bound this error function.

## Chapter 3

#### Section 3.2

1.

By noticing that the zeroth Lagrangian coefficient function of degree n takes on the value unity when  $x = x_0$  and the value zero when  $x = x_1, \ldots, x_n$ , and by considering the associated divided difference table (or otherwise), show that

$$l_0(x) = 1 + \frac{x - x_0}{x_0 - x_1} + \frac{(x - x_0)(x - x_1)}{(x_0 - x_1)(x_0 - x_2)} + \dots + \frac{(x - x_0)\dots(x - x_{n-1})}{(x_0 - x_1)\dots(x_0 - x_n)}$$
(11)

and that similar expressions can be written down by symmetry for the other coefficient functions.

I will show this by induction. First, note the base case. For n = 1,

$$1 + \frac{x - x_0}{x_0 - x_1} = \frac{x_0 - x_1 + x - x_0}{x_0 - x_1}$$
$$= \frac{x - x_1}{x_0 - x_1}$$

Now suppose the equation holds for and for n = k. For n = k + 1,

$$1 + \frac{x - x_0}{x_0 - x_1} + \dots + \frac{(x - x_0) \dots (x - x_{n-1})}{(x_0 - x_1) \dots (x_0 - x_n)} + \frac{(x - x_0) \dots (x - x_n)}{(x_0 - x_1) \dots (x_0 - x_{n+1})}$$

(by the induction assumption)

$$= \frac{(x-x_1)\dots(x-x_n)}{(x_0-x_1)\dots(x_0-x_n)} + \frac{(x-x_0)\dots(x-x_{n-1})(x-x_n)}{(x_0-x_1)\dots(x_0-x_n)(x_0-x_{n+1})}$$

$$= \frac{[(x-x_1)\dots(x-x_{n-1})]\cdot[x_0-x_{n+1}+x-x_0]}{(x_0-x_1)\dots(x_0-x_n)(x_0-x_{n+1})}$$

$$= \frac{(x-x_1)\dots(x-x_n)}{(x_0-x_1)\dots(x_0-x_{n+1})}$$

Thus (11) holds for all natural numbers. The corresponding formulas for i = 1, ..., n are:

$$l_1(x) = 1 + \frac{x - x_1}{x_1 - x_0} + \frac{(x - x_0)(x - x_1)}{(x_1 - x_0)(x_1 - x_2)} + \dots + \frac{(x - x_0)\dots(x - x_{n-1})}{(x_1 - x_0)(x_1 - x_2)\dots(x_1 - x_n)}$$

$$l_2(x) = 1 + \frac{x - x_2}{x_2 - x_0} + \frac{(x - x_0)(x - x_2)}{(x_2 - x_0)(x_2 - x_1)} + \frac{(x - x_0)(x - x_1)(x - x_2)}{(x_2 - x_0)(x_2 - x_1)(x_2 - x_3)}$$

$$+ \dots + \frac{(x - x_0)\dots(x - x_{n-1})}{(x_2 - x_0)(x_2 - x_1)(x_2 - x_3)\dots(x_2 - x_n)}$$

:

$$l_{i}(x) = 1 + \frac{x - x_{i}}{x_{i} - x_{0}} + \frac{(x - x_{i})(x - x_{0})}{(x_{i} - x_{0})(x_{i} - x_{1})} + \frac{(x - x_{i})(x - x_{0})(x - x_{1})}{(x_{i} - x_{0})(x_{i} - x_{1})(x_{i} - x_{2})} + \dots + \frac{(x - x_{0}) \dots (x - x_{i})}{(x_{i} - x_{0}) \dots (x_{i} - x_{i-1})(x_{i} - x_{i+1})} + \dots + \frac{(x - x_{0}) \dots (x - x_{n-1})}{(x_{i} - x_{0}) \dots (x_{i} - x_{i-1})(x_{i} - x_{i+1}) \dots (x_{i} - x_{n})}$$

:

$$l_n(x) = 1 + \frac{x - x_n}{x_n - x_0} + \frac{(x - x_n)(x - x_0)}{(x_n - x_0)(x_n - x_1)} + \dots + \frac{(x - x_0)(x - x_1)\dots(x - x_{n-2})(x - x_n)}{(x_n - x_0)\dots(x_n - x_{n-1})}$$

4.

$$\begin{vmatrix} y & 1 & x & x^2 & \dots & x^n \\ f(x_0) & 1 & x_0 & x_0^2 & \dots & x_0^n \\ \vdots & \vdots & \vdots & \vdots & & \vdots \\ f(x_n) & 1 & x_n & x_n^2 & \dots & x_n^n \end{vmatrix} = 0$$
(12)

Show that

$$\begin{vmatrix} 1 & a_1 & a_1^2 \\ 1 & a_2 & a_2^2 \\ 1 & a_3 & a_3^2 \end{vmatrix} = (a_2 - a_1)(a_3 - a_1)(a_3 - a_2)$$

and use this fact to express the result of expanding the left-hand member of (12) with respect to the elements of the first column, and equating the result to zero, in Lagrangian form when n=2.

In the second line of the following calculation, the term  $a_1a_2a_3$  is added and subtracted so that the correct factoring can occur.

$$\begin{vmatrix} 1 & a_1 & a_1^2 \\ 1 & a_2 & a_2^2 \\ 1 & a_3 & a_3^2 \end{vmatrix} = a_2 a_3^2 - a_2^2 a_3 - a_1 a_3^2 + a_1^2 a_3 + a_1 a_2^2 - a_1^2 a_2$$

$$= a_2 a_3^2 - a_2^2 a_3 - a_1 a_2 a_3 + a_1 a_2^2 - a_1 a_3^2 + a_1 a_2 a_3 + a_1^2 a_3 - a_1^2 a_2$$

$$= (a_2 a_3 - a_1 a_2 - a_1 a_3 + a_1^2)(a_3 - a_2)$$

$$= (a_2 - a_1)(a_3 - a_1)(a_3 - a_2)$$

Consider (12) for n=2:

$$\begin{vmatrix} y & 1 & x & x^2 \\ f(x_0) & 1 & x_0 & x_0^2 \\ f(x_1) & 1 & x_1 & x_1^2 \\ f(x_2) & 1 & x_2 & x_2^2 \end{vmatrix} = 0$$

$$y \cdot \begin{vmatrix} 1 & x_0 & x_0^2 \\ 1 & x_1 & x_1^2 \\ 1 & x_2 & x_2^2 \end{vmatrix} - f(x_0) \cdot \begin{vmatrix} 1 & x & x^2 \\ 1 & x_1 & x_1^2 \\ 1 & x_2 & x_2^2 \end{vmatrix} + f(x_1) \cdot \begin{vmatrix} 1 & x & x^2 \\ 1 & x_0 & x_0^2 \\ 1 & x_2 & x_2^2 \end{vmatrix} - f(x_2) \cdot \begin{vmatrix} 1 & x & x^2 \\ 1 & x_0 & x_0^2 \\ 1 & x_1 & x_1^2 \end{vmatrix} = 0$$

Each of the four 3x3 matrices can be expanded in a similar fashion as above, so

$$y[(x_1 - x_0)(x_2 - x_0)(x_2 - x_1)] - f(x_0)[(x_1 - x)(x_2 - x)(x_2 - x_1)]$$
  
+  $f(x_1)[(x_0 - x)(x_2 - x)(x_2 - x_0)] - f(x_2)[(x_0 - x)(x_1 - x)(x_1 - x_0)] = 0$ 

Solving for y yields

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

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

$$= l_0(x) f(x_0) + l_1(x) f(x_1) + l_2(x) f(x_2)$$

By considering the limit of the three-point Lagrangian interpolation formula relative to  $x_0$ ,  $x_0 + \epsilon$ , and  $x_1$ , as  $\epsilon \to 0$ , obtain the formula

$$f(x) = \frac{(x_1 - x)(x + x_1 - 2x_0)}{(x_1 - x_0)^2} f(x_0) + \frac{(x - x_0)(x_1 - x)}{x_1 - x_0} f'(x_0) + \frac{(x - x_0)^2}{(x_1 - x_0)^2} f(x_1) + E(x)$$

where

$$E(x) = \frac{1}{6}(x - x_0)^2(x - x_1)f'''(\xi)$$

The Lagrangian coefficient of  $f(x_k)$  in  $p_{0,1,\ldots,n}$  is

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

We will denote  $x_0$ ,  $x_1$ , and  $x_2$  as  $x_0$ ,  $x_0 + \epsilon$ , and  $x_1$ , respectively. Thus,

$$l_0(x) = \frac{(x - x_0 - \epsilon)(x - x_1)}{(x_0 - x_0 - \epsilon)(x_0 - x_1)} = \frac{-1}{\epsilon} \cdot \frac{(x - x_0 - \epsilon)(x - x_1)}{x_0 - x_1}$$

$$= \frac{-1}{\epsilon} \cdot \left[ \frac{(x - x_1)(x - x_0)}{x_0 - x_1} - \frac{\epsilon(x - x_1)}{x_0 - x_1} \right]$$

$$l_{0,\epsilon}(x) = \frac{(x - x_0)(x - x_1)}{(x_0 + \epsilon - x_0)(x_0 + \epsilon - x_1)} = \frac{1}{\epsilon} \cdot \frac{(x - x_0)(x - x_1)}{x_0 + \epsilon - x_1}$$

$$= \frac{1}{\epsilon} \cdot \left[ \frac{(x - x_0)(x - x_1)}{x_0 + \epsilon - x_1} + \frac{(x - x_1)(x - x_0)}{x_0 - x_1} - \frac{(x - x_1)(x - x_0)}{x_0 - x_1} \right]$$

$$l_1(x) = \frac{(x - x_0)(x - x_0 - \epsilon)}{(x_1 - x_0)(x_1 - x_0 - \epsilon)}$$

Using these coefficients, we can write

$$p_{0,1,2}(x) = l_0(x)f(x_0) + l_{0,\epsilon}(x)f(x_0 + \epsilon) + l_1(x)f(x_1)$$

$$= \frac{-f(x_0)}{\epsilon} \left[ \frac{(x - x_1)(x - x_0)}{x_0 - x_1} - \frac{\epsilon(x - x_1)}{x_0 - x_1} \right]$$

$$+ f(x_0 + \epsilon) \left[ \frac{(x - x_0)(x - x_1)}{\epsilon(x_0 + \epsilon - x_1)} + \frac{(x - x_1)(x - x_0)}{\epsilon(x_0 - x_1)} - \frac{(x - x_1)(x - x_0)}{\epsilon(x_0 - x_1)} \right]$$

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

$$= \frac{(x - x_1)(x - x_0)}{x_0 - x_1} \left[ \frac{f(x_0 + \epsilon) - f(x_0)}{\epsilon} \right] + f(x_0) \frac{x - x_1}{x_0 - x_1}$$

$$+ f(x_0 + \epsilon)(x - x_0)(x - x_1) \left[ \frac{g(x_0 + \epsilon) - g(x_0)}{\epsilon} \right] + f(x_1) \frac{(x - x_0)(x - x_0 - \epsilon)}{(x_1 - x_0)(x_1 - x_0 - \epsilon)}$$

where

$$g(x) = \frac{1}{x - x_1}$$
Since  $f'(x) := \lim_{\epsilon \to 0} \frac{f(x + \epsilon) - f(x)}{\epsilon}$ ,  $g'(x) := \lim_{\epsilon \to 0} \frac{g(x + \epsilon) - g(x)}{\epsilon}$ , and  $g'(x) = \frac{-1}{(x - x_1)^2}$ 

$$\lim_{\epsilon \to 0} p_{0,1,2} = f'(x_0) \frac{(x - x_1)(x - x_0)}{x_0 - x_1} + f(x_0) \left(\frac{x - x_1}{x_0 - x_1} - \frac{(x - x_0)(x - x_1)}{(x_0 - x_1)^2}\right) + f(x_1) \frac{(x - x_0)^2}{(x_1 - x_0)^2}$$

$$= \frac{(x_1 - x)(x + x_1 - 2x_0)}{(x_1 - x_0)^2} f(x_0) + \frac{(x - x_0)(x_1 - x)}{x_1 - x_0} f'(x_0) + \frac{(x - x_0)^2}{(x_1 - x_0)^2} f(x_1)$$

$$= \frac{(x_1 - x)(x + x_1 - 2x_0)}{(x_1 - x_0)^2} f(x_0) + \frac{(x - x_0)(x_1 - x)}{x_1 - x_0} f'(x_0) + \frac{(x - x_0)^2}{(x_1 - x_0)^2} f(x_1) + E(x)$$