

Homework 3 Solution

4.1.12 (a) $L_0(x) + L_1(x) + L_2(x) + L_3(x)$ is the unique polynomial of degree ≤ 3 interpolating the data $(x_0, 1), (x_1, 1), (x_2, 1), (x_3, 1)$. Since the polynomial 1 also has degree ≤ 3 and interpolates the data, we conclude

$$L_0(x) + L_1(x) + L_2(x) + L_3(x) = 1.$$

(b) In general, $L_0(x) + \dots + L_n(x)$ and 1 are both degree $\leq n$ interpolants of $(x_0, 1), \dots, (x_n, 1)$, and therefore must be equal.

4.1.13 For $j \leq 3$, note that $x_0^j L_0(x) + x_1^j L_1(x) + x_2^j L_2(x) + x_3^j L_3(x)$ and x^j are both polynomials of degree ≤ 3 interpolating $(x_0, x_0^j), (x_1, x_1^j), (x_2, x_2^j), (x_3, x_3^j)$. Since such interpolants are unique, the two polynomials must be equal.

For $j > 3$, note that each $L_i(x)$ has degree ≤ 3 , so a linear combination of them can't have the same degree as x^j .

4.1.15 Note: this is the complicated solution.

In analogy to the Lagrange basis $\{L_i(x)\}$, we want to find a basis $\{H_i(x)\}$ for the (vector space of) polynomials of degree ≤ 3 so that the coordinates of $P(x)$ with respect to $\{H_i(x)\}$ are (y_0, y'_0, y_1, y'_1) . The analogous versions of the requirements

$$L_i(x_j) = \delta_{ij}$$

are

$$\begin{aligned} H_{2i}(x_j) &= \delta_{ij} & H'_{2i}(x_j) &= 0 \\ H_{2i+1}(x_j) &= 0 & H'_{2i+1}(x_j) &= \delta_{ij} \end{aligned}$$

These insure, for example, that $H_0(x)$ controls the value of $P(x)$ at 0, but doesn't affect the value of $P(x)$ at 1 or the derivative of $P(x)$ at 0 or 1.

The requirements are met by the polynomials

$$H_{2i}(x) = (1 - 2L'_i(x_i)(x - x_i))L_i^2(x) \text{ and } H_{2i+1}(x) = (x - x_i)L_i^2(x)$$

where $L_0(x)$ and $L_1(x)$ are the usual Lagrange basis elements. I.e.

$$\begin{aligned} H_0(x) &= \left(1 - \frac{2}{x_0 - x_1}(x - x_0)\right) \left(\frac{x - x_1}{x_0 - x_1}\right)^2 \\ H_1(x) &= (x - x_0) \left(\frac{x - x_1}{x_0 - x_1}\right)^2 \\ H_2(x) &= \left(1 - \frac{2}{x_1 - x_0}(x - x_1)\right) \left(\frac{x - x_0}{x_1 - x_0}\right)^2 \\ H_3(x) &= (x - x_1) \left(\frac{x - x_0}{x_1 - x_0}\right)^2 \end{aligned}$$

and

$$P(x) = y_0 H_0(x) + y'_0 H_1(x) + y_1 H_2(x) + y'_1 H_3(x)$$

4.1.16 The divided differences are easy to compute

i	x_i	$f[x_i]$	$f[x_i, x_{i+1}]$	$f[x_i, x_{i+1}, x_{i+2}]$
0	0	-1	0	2
1	1	-1	4	
2	1	-1		

(remember that $f[x, x] = f'(x)$!) and the polynomial just as easy to read off from the top row:

$$P(x) = -1 + 0 \cdot x + 2 \cdot x(x - 1)$$

The method of the previous problem is more complicated. We want polynomials $\{Q_i(x)\}$ for $i = 0, 1, 2$ which form a basis for the polynomials of degree ≤ 2 so that the coordinates of the polynomial we're looking for are $(-1, -1, 4)$. In particular, we require

$$\begin{aligned} Q_0(0) = 1 & \quad Q_0(1) = 0 & \quad Q'_0(1) = 0 \\ Q_1(0) = 0 & \quad Q_1(1) = 1 & \quad Q'_1(1) = 0 \\ Q_2(0) = 0 & \quad Q_2(1) = 0 & \quad Q'_2(1) = 1 \end{aligned}$$

So we find solutions:

$$Q_0(x) = (x - 1)^2, \quad Q_1(x) = 1 - (x - 1)^2, \quad \text{and} \quad Q_2(x) = x(x - 1)$$

and then write

$$P(x) = -1 \cdot (x - 1)^2 - 1 \cdot (1 - (x - 1)^2) + 4 \cdot x(x - 1)$$

4.1.28 Because we have 6 data points and the original polynomial has degree ≤ 5 , we know that the interpolating polynomial is the original polynomial.

The divided difference algorithm:

i	x_i	$f[x_i]$	\dots
0	-2	-5	6 -3 1 0 0
1	-1	1	0 1 0
2	0	1	0 3 1
3	1	1	6 6
4	2	7	18
5	3	25	

yields

$$p(x) = -5 + 6(x + 2) - 3(x + 2)(x + 1) + 1(x + 2)(x + 1)x$$

so $p(x)$ has degree 3.

4.1.31 The key observation is: for interpolation on $m + 1$ nodes, $f[x_0, \dots, x_m]$ is the coefficient of x^m in the interpolating polynomial.

For $m \geq n$, we know the interpolating polynomial $p(x)$ is just the original polynomial x^n , i.e. $p(x) = x^n$. If $m = n$, then the coefficient of x^m in x^n is 1 and thus $f[x_0, \dots, x_m] = 1$. If $m > n$, then the coefficient of x^m in x^n is 0 so $f[x_0, \dots, x_m] = 0$.

4.2.3 We just need to apply the error formula for Lagrange interpolation and then simplify.

(a) For $x \in [x_i, x_{i+1}]$ we know there is some $c \in [x_i, x_{i+1}]$ so that

$$f(x) - P_1(x) = \frac{(x - x_i)(x - x_{i+1})}{2} f''(x)$$

In our particular case,

$$|f''(c)| = |2ce^{-c^2}| \leq 2$$

and

$$\left| \frac{(x - x_i)(x - x_{i+1})}{2} \right| \leq \max_{x \in [0, 0.01]} \frac{x(0.01 - x)}{2} = \frac{0.05^2}{2}$$

so

$$|f(x) - P_1(x)| \leq 0.05^2 \approx 1.25 \times 10^{-3}$$

(b) As in (a), we know

$$f(x) - P_2(x) = \frac{(x - x_i)(x - x_{i+1})(x - x_{i+2})}{6} f'''(c)$$

and then for our specific case

$$|f'''(c)| = |(2 - 4c)e^{-c^2}| \leq 2$$

and

$$\left| \frac{(x - x_i)(x - x_{i+1})(x - x_{i+2})}{6} \right| \leq \max_{x \in [0, 0.02]} \frac{|x(x - 0.01)(x - 0.02)|}{6} = \frac{0.01^3}{9\sqrt{3}}$$

so

$$|f(x) - P_2(x)| \leq \frac{0.01^3}{9\sqrt{3}} 2 \approx 1.28 \times 10^{-7}$$

4.2.13 Induct on n . The base case already done, so consider

$$|(x - x_0) \cdots (x - x_{n+1})|$$

If $x \in [x_0, x_n]$, then by induction hypothesis

$$|(x - x_0) \cdots (x - x_n)| \leq n!h^{n+1}$$

and, since the x_i s have even spacing h ,

$$|x - x_{n+1}| \leq |x_0 - x_{n+1}| = (n+1)h$$

Putting these together yields

$$|(x - x_0) \cdots (x - x_n)| \leq (n+1)!h^{n+2}$$

If instead $x \in [x_1, x_{n+1}]$, a symmetric argument yields the same conclusion.

4.2.14 We compute

$$\begin{aligned} |e^x - P_n(x)| &\leq \left| \frac{(x - x_0) \cdots (x - x_n)}{n!} e^c \right| \\ &\leq \frac{n!(1/n)^{n+1}}{n!} e^c \leq (1/n)^{n+1} e^c \leq (1/n)^{n+1} e \rightarrow 0 \end{aligned}$$