

## Problems

1.

- (i) Given  $x_0 = -0.2$ ,  $x_1 = 0$ , and  $x_2 = 0.2$  construct a second degree polynomial to approximate  $f(x) = e^x$  via Newton's divided differences.

We want to derive a polynomial of the form

$$p(x) = a_0 + a_1(x - x_0) + a_2(x - x_0)(x - x_1)$$

where  $a_i = [x_0, \dots, x_{i-1}]$  are the Newton Divided differences. For this problem, we have

$$\begin{aligned} a_0 &= f[x_0] = e^{x_0} = e^{-0.2}, \\ a_1 &= f[x_0, x_1] = \frac{e^{x_1} - e^{x_0}}{x_1 - x_0} = \frac{1 - e^{-0.2}}{0.2}, \end{aligned}$$

and

$$a_2 = f[x_0, x_1, x_2] = \frac{f[x_1, x_2] - f[x_0, x_1]}{x_2 - x_0} = \frac{\frac{e^{0.2} - 1}{0.2} - \frac{1 - e^{-0.2}}{0.2}}{0.4}$$

which makes our polynomial

$$\begin{aligned} p(x) &= e^{-0.2} + \frac{1 - e^{-0.2}}{0.2}(x + 0.2) + \frac{\frac{e^{0.2} - 1}{0.2} - \frac{1 - e^{-0.2}}{0.2}}{0.4}(x + 0.2)(x) \\ &= 1 + 1.00668x + 0.501669x^2. \end{aligned}$$

- (ii) Derive an error bound for  $p_2(x)$  when  $x \in [-0.2, 0.2]$ .

First, note that the third derivative of  $f$  is maximized over  $[-0.2, 0.2]$  when  $x = 0.2$ . Then, we can obtain a bound on our error as

$$\begin{aligned} E(t) &\leq \max_{t \in [-0.2, 0.2]} E(t) \\ &= \max_{t \in [-0.2, 0.2]} \frac{(t + 0.2)(t)(t - 0.2)}{6} e^{0.2} \\ &= \frac{(-\frac{\sqrt{3}}{15} + 0.2)(-\frac{\sqrt{3}}{15})(-\frac{\sqrt{3}}{15} - 0.2)}{6} e^{0.2} \\ &= 6.26824 \cdot 10^{-4}. \end{aligned}$$

- (iii) Compute the error  $E(0.1) = f(0.1) - p_2(0.1)$ . How does this compare with the error bound? Our error is

$$E(0.1) = |1.10517 - 1.10568| = 5.136621 \cdot 10^{-4}$$

which is within our error bound! So our error bound holds  $x = 0.1$ .

2.

(i) Show there is a unique cubic polynomial  $p(x)$  for which

$$\begin{aligned} p(x_0) &= f(x_0) & p(x_2) &= f(x_2) \\ p'(x_1) &= f'(x_1) & p''(x_1) &= f''(x_1) \end{aligned}$$

where  $f(x)$  is a given function and  $x_0 \neq x_2$ .Suppose  $p(x)$  and  $q(x)$  are two cubic polynomials satisfying

$$\begin{aligned} p(x_0) &= q(x_0) = f(x_0) & p(x_2) &= q(x_2) = f(x_2) \\ p'(x_1) &= q'(x_1) = f'(x_1) & p''(x_1) &= q''(x_1) = f''(x_1). \end{aligned}$$

Now let  $v(x) = p(x) - q(x)$ . Then, by linearity,  $v(x)$  is a cubic polynomial that satisfies

$$\begin{aligned} v(x_0) &= 0 & v(x_2) &= 0 \\ v'(x_1) &= 0 & v''(x_1) &= 0. \end{aligned}$$

Furthermore, because  $v(x)$  is a cubic polynomial, there exists constants  $a_0, a_1, a_2$ , and  $a_3$  such that

$$\begin{aligned} v(x) &= a_0 + a_1x + a_2x^2 + a_3x^3 \\ v'(x) &= a_1 + 2a_2x + 3a_3x^2 \\ v''(x) &= 2a_2 + 6a_3x. \end{aligned}$$

Then,

$$v''(x_1) = 2a_2 + 6a_3x_1 = 0$$

which implies

$$a_2 = -3a_3.$$

Using this expression in our first derivative yields

$$v'(x_1) = a_1 - 6a_3x_1^2 + 3a_3x_1^2 = 0$$

which implies

$$a_1 = 3a_3x_1^2.$$

Finally,

$$\begin{aligned} v(x) &= a_0 + 3a_3x_1^2x - 3a_3x_1x^2 + a_3x^3 \\ &= a_0 + 3a_3x_1^2x - 3a_3x_1x^2 + a_3x^3 - a_3x_1^3 + a_3x_1^3 \\ &= (a_0 + a_3x_1^3) - a_3(x - x_1)^3. \end{aligned}$$

Then, using our last constraints, we have the system

$$\begin{aligned} v(x_0) &= (a_0 + a_3x_1^3) - a_3(x_0 - x_1)^3 = 0 \\ v(x_2) &= (a_0 + a_3x_1^3) - a_3(x_2 - x_1)^3 = 0 \end{aligned}$$

which yields

$$a_3(x_0 - x_1)^3 = a_3(x_2 - x_1)^3.$$

Then, because  $x_0 \neq x_2$ , we must have  $a_3 = 3$  which implies  $a_0 = 0$ . Therefore,

$$v(x) = 0$$

and so we must have

$$p(x) = q(x)$$

showing the uniqueness of our polynomial.

(ii) Derive a formula for  $p(x)$ .

We know  $p(x)$  has the form

$$p(x) = a_0 + a_1x + a_2x^2 + a_3x^3$$

for some constants  $a_0, a_1, a_2$ , and  $a_3$ . Then, using our second derivative information, we have

$$p''(x_1) = 2a_2 + 6a_3x_1 = f''(x_1)$$

which implies

$$a_2 = \frac{1}{2}f''(x_1) - 3a_3x_1.$$

Then, using our first derivative information, we have

$$p'(x_1) = a_1 + 2\left(\frac{1}{2}f''(x_1) - 3a_3x_1\right)x_1 + 3a_3x_1^2 = f'(x_1)$$

which implies

$$a_1 = f'(x_1) - f''(x_1)x_1 + 3a_3x_1^2.$$

Next, we can use our function information to get the system

$$\begin{aligned} p(x_0) &= a_0 + f'(x_1)x_0 - f''(x_1)x_0x_1 + 3a_3x_0x_1^2 + \frac{1}{2}f''(x_1)x_0^2 - 3a_3x_0^2x_1 + a_3x_0^3 = f(x_0) \\ p(x_2) &= a_0 + f'(x_1)x_2 - f''(x_1)x_1x_2 + 3a_3x_1^2x_2 + \frac{1}{2}f''(x_1)x_2^2 - 3a_3x_1x_2^2 + a_3x_2^3 = f(x_2) \end{aligned}$$

which implies

$$\begin{aligned} a_3 &= \frac{f(x_2) - f(x_0) + f'(x_1)(x_0 - x_2) - f''(x_1)x_1(x_0 - x_2) + \frac{1}{2}f''(x_1)(x_0^2 - x_2^2)}{3x_1^2(x_2 - x_0) + 3x_1(x_0^2 - x_2^2) + x_2^3 - x_0^3} \\ a_0 &= f(x_0) - f'(x_1)x_0 + f''(x_1)x_0x_1 + 3a_3x_0x_1^2 - \frac{1}{2}f''(x_1)x_0^2 + 3a_3x_0^2x_1 - a_3x_0^3. \end{aligned}$$

Now, we construct our polynomial by first computing  $a_3, a_0, a_1$ , and  $a_2$  in that order and then plugging them into our polynomial

$$p(x) = a_0 + a_1x + a_2x^2 + a_3x^3.$$

(iii) Let  $x_0 = -1, x_1 = 0$ , and  $x_2 = 1$ . Assuming  $f(x) \in C^4[-1, 1]$ , show that for  $x \in [-1, 1]$ ,

$$f(x) - p(x) = \frac{x^4 - 1}{4!}f^{(4)}(\eta_x)$$

for some  $\eta_x \in [-1, 1]$ .

Suppose we have a fixed  $t \in [-1, 1]$  such that  $t \neq x_0$  and  $t \neq x_2$ . Next, define

$$G(x) = E(x) - \frac{x^4 - 1}{t^4 - 1} E(t)$$

where  $E(x) = f(x) - p(x)$ . Then, by construction, we have  $G(x_0) = G(x_2) = G(t) = 0$ . Next, using Rolle's theorem, we must have at least two roots to  $G'(x)$ . But,  $G'(x_1) = 0$  by construction and so as long as the two roots from Rolle's theorem do not coincide with  $x_1$ ,  $G'(x)$  has three roots (Note it is highly unlikely that  $x_1$  will be the point given from Rolle's theorem). Then, using Rolle's theorem again implies that  $G''(x)$  also has two roots that are not equal to  $x_1$ . So, because  $G''(x_1) = 0$ , we know  $G''(x)$  also has at least three roots. Using Rolle's theorem twice more tells us that  $G^{(3)}(x)$  has two roots which implies  $G^{(4)}(x)$  has one root; call this root  $\eta_x \in [-1, 1]$ . Then, we have

$$\begin{aligned} G^{(4)} &= E^{(4)}(\eta_x) - \frac{4!}{t^4 - 1} E(t) \\ &= f^{(4)}(\eta_x) - \frac{4!}{t^4 - 1} E(t) = 0 \end{aligned}$$

which implies

$$E(t) = \frac{t^4 - 1}{4!} f^{(4)}(\eta_x).$$

Then, if we rename  $t$  to  $x$ , we have

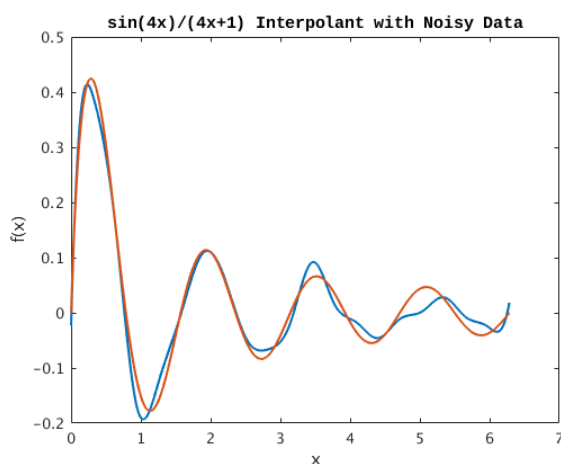
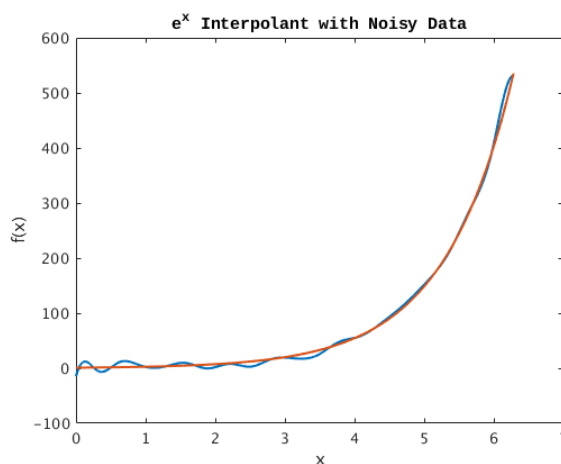
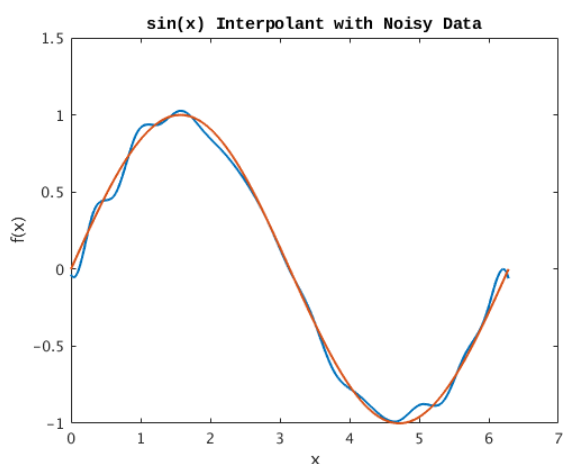
$$E(x) = f(x) - p(x) = \frac{x^4 - 1}{4!} f^{(4)}(\eta_x).$$

3. Suppose we have  $m$  data points  $\{(t_i, y_i)\}_{i=1}^m$ , where the  $t$ -values all occur in some interval  $[x_0, x_n]$ . We subdivide the interval  $[x_0, x_n]$  into  $n$  subintervals  $\{[x_k, x_{k+1}]_{k=0}^{n-1}\}$  of equal length  $h$  and attempt to choose a spline function  $s(x)$  with nodes at  $\{x_k\}_{k=0}^n$  in such a way so that

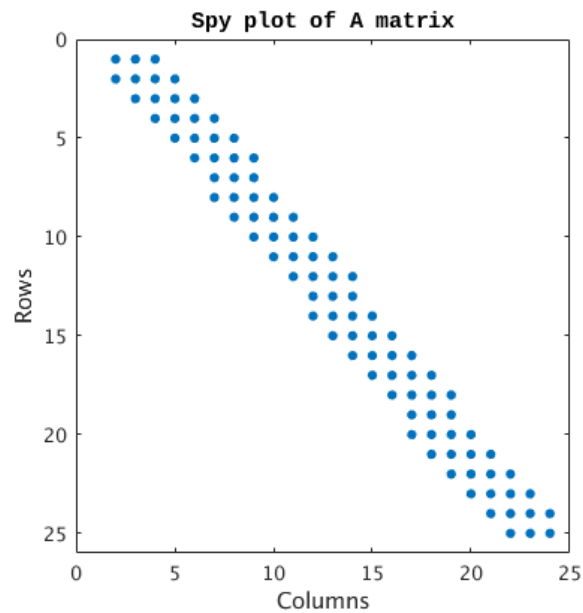
$$\sum_{i=1}^m |y_i - s(t_i)|^2$$

is *minimized*.

- (i) My B-Spline code is given at the end of the document  
(ii) Using my code, I was able to produce the following noisy plots:



- (iii) My spy plot is given below. We can see that we have nice and relatively uniform sparse structure throughout the matrix. Each row has at most 4 nonzero entries that are all consecutive. Furthermore, each new row either shifts the 4 consecutive non entries to the right by 1 or not at all.



## Code Used

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```
1 %%
2 % B-Spline interpolation routine for Problem 3 of HW08
3 % Author: Caleb Jacobs
4 % Date last modified: 20-10-2021
5
6 %% Clean workspace
7 clear
8
9 %% Settings
10 format longE
11 a = 0;           % Left side of interval
12 b = 2*pi;       % Right side of interval
13 n = 20;         % Number of intervals
14 m = 25;         % Number of data points
15
16 %% Initialize data
17 h = (b - a) / m; % Step size
18 x = linspace(a, b, n + 1); % Interval boundaries
19 t = linspace(a, b, m); % Nodes
20
21 A = constructA(t, x); % Find A matrix for computing weights
22
23 %% Interpololate sin(x)
24 f = @(x) sin(x); % Function definition
25 bb = f(t)' + 0.05*randn(m, 1); % Data function values
26 c = A \ bb; % Interpolation weights
27
28 figure(1)
29 T = linspace(a, b, 1000);
30 plot(T, S(T, x, c), T, f(T), 'LineWidth', 1.5)
31 title('sin(x) Interpolant with Noisy Data')
32 xlabel('x')
33 ylabel('f(x)')
34
35 %% Interpolate exp(x)
36 f = @(x) exp(x); % Function definition
37 bb = f(t)' + 6*randn(m, 1); % Data function values
38 c = A \ bb; % Interpolation weights
39
40 figure(2)
41 T = linspace(a, b, 1000);
42 plot(T, S(T, x, c), T, f(T), 'LineWidth', 1.5)
43 title('e^x Interpolant with Noisy Data')
44 xlabel('x')
45 ylabel('f(x)')
46
47 %% Interpolate sin(x)/x
```

```

48 f = @(x) sin(4*x) ./ (4*x + 1); % Function definition
49 bb = f(t)' + 0.02*randn(m, 1); % Data function values
50 c = A \ bb; % Interpolation weights
51
52 figure(3)
53 T = linspace(a, b, 1000);
54 plot(T, S(T, x, c), T, f(T), 'LineWidth', 1.5)
55 title('sin(4x)/(4x+1) Interpolant with Noisy Data')
56 xlabel('x')
57 ylabel('f(x)')
58
59 %% Plot matrix structure
60 figure(4)
61 spy(A)
62 title('Spy plot of A matrix')
63 xlabel('Columns')
64 ylabel('Rows')
65
66 %% Routine definitions
67 % Basis spline
68 function val = B(t, x, i)
69     h = x(2) - x(1);
70     n = length(x);
71
72     if i - 2 > 0 && t < x(i - 2)
73         val = 0;
74     elseif i - 2 > 0 && x(i - 2) <= t && t <= x(i - 1)
75         val = (t - x(i - 2)).^3 / h^3;
76     elseif i - 1 > 0 && x(i - 1) <= t && t <= x(i)
77         xi = x(i - 1);
78         val = 1 + 3*(t - xi) / h + 3*(t - xi).^2 / h^2 - 3*(t - xi)^3 / h^3;
79     elseif i + 1 <= n && x(i) <= t && t <= x(i + 1)
80         xi = x(i + 1);
81         val = 1 + 3*(xi - t) / h + 3*(xi - t).^2 / h^2 - 3*(xi - t)^3 / h^3;
82     elseif i + 2 <= n && x(i + 1) <= t && t <= x(i + 2)
83         val = (x(i + 2) - t).^3 / h^3;
84     else
85         val = 0;
86     end
87 end
88
89 % Get interval indices of data
90 function idx = getIdx(t, x)
91     m = length(t); % Number of nodes
92     n = length(x); % Number of intervals + 1
93
94     idx = ones(m, 1);
95     for i = 1 : m
96         for j = 2 : n
97             if t(i) < x(j)
98                 break

```



```
99         end
100     end
101     idx(i) = j - 1;
102 end
103 end
104
105 % Constuct A matrix
106 function A = constructA(t, x)
107     m = length(t);           % Number of nodes
108     n = length(x);           % Number of intervals + 1
109     h = x(2) - x(1);         % Data spacing
110
111     idx = getIdx(t, x) + 2;   % Interval indices of data
112
113     x = [(x(1) - h*[2 1]) x (x(n) + h*[1 2])]; % Append extra intervals
114
115     A = spalloc(m, n + 2, 4 * m); % Initialize sparse A
116     for i = 1 : m
117         for j = -1 : 2
118             A(i, idx(i) + j) = B(t(i), x, idx(i) + j);
119         end
120     end
121 end
122
123 % Evaluate Interpolant
124 function f = S(t, x, c)
125     m = length(t);
126     h = x(2) - x(1);
127     n = length(x);
128     x = [(x(1) - h*[2 1]) x (x(n) + h*[1 2])]; % Append extra intervals
129     n = length(x);
130
131     f = zeros(m, 1);
132     for j = 1 : m
133         for i = 1 : n - 1
134             f(j) = f(j) + c(i) * B(t(j), x, i);
135         end
136     end
137 end
```

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