# Homework 5

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Github URL for homework 5: https://github.com/pach2648/APPM4600/tree/main/Homework/HW5

# Problem 1

#### Solution

## Github Link for Q1

### a) Github Link for pseudocode

Barycentric Lagrange interpolation is numerically stable and computationally efficient compared to other interpolation methods. The key idea behind barycentric Lagrange interpolation is to compute weighted sums of function values at interpolation points, where the weights depend only on the target points and the interpolation points.

#### b) Barycentric Lagrange Interpolation

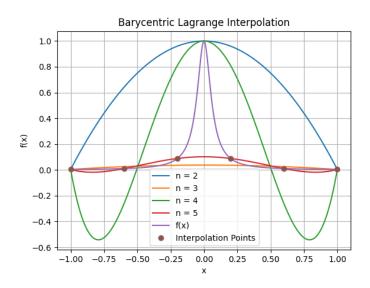


Figure 1: Barycentric Lagrange Interpolation (n = 2 - 5)

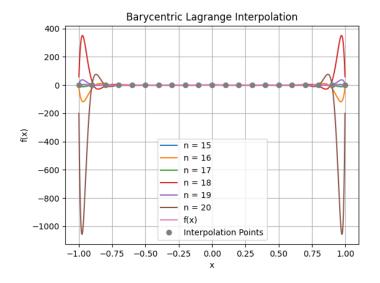


Figure 2: Barycentric Lagrange Interpolation (n = 15 - 20)

# c) Barycentric Lagrange Interpolation with Chebyshev

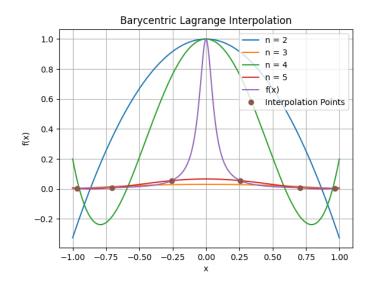


Figure 3: Barycentric Lagrange Interpolation with Chebyshev (n = 2 - 5)

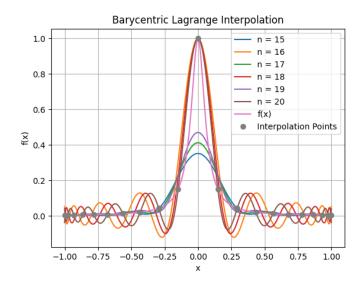


Figure 4: Barycentric Lagrange Interpolation with Chebyshev (n = 15 - 20)

d) Chebyshev nodes lead to smaller errors compared to equispaced nodes, especially near the endpoints of the interval. The equispaced nodes result in larger values of  $\psi(x)$  near the endpoints, contributing to higher interpolation errors in those regions while Chebyshev fixes this limitation.

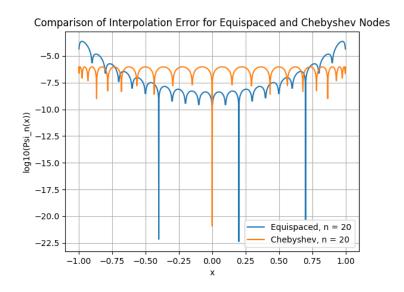
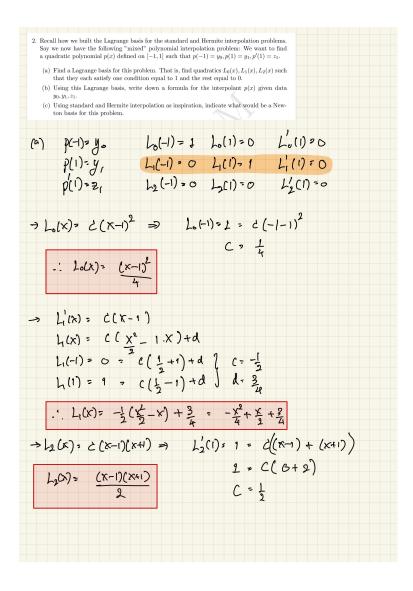


Figure 5: Comparison of Error for Equispaced and Chebyshev Nodes when n=20

# Problem 2

#### Solution



(c) 
$$P(x) = [y_0] | l_0(x) + (y_1) | l_1(x) + (y_1) | l_2(x)$$

P(x) =  $y_0(\frac{(x-1)^2}{4}) + y_1(\frac{-x^2}{x^2} + \frac{x}{2} + \frac{3}{4}) + \frac{2}{x}(\frac{(x-1)(x+1)}{2})$ 

(c) Hermite  $x = | 1 | 1$ 
 $y_0 = y_1$ 
 $y_1 = y_1$ 
 $y_1 = y_1$ 

1  $y_1 = y_1$ 

2  $y_1 = y_1$ 

1  $y_1 = y_1$ 

2  $y_1 = y_1$ 

2  $y_1 = y_1$ 

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4  $y_1 = y_1$ 

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6  $y_1 = y_1$ 

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7  $y_1 = y_1$ 

8  $y_1 = y_1$ 

9  $y_1 = y_1$ 

1  $y_1 = y_1$ 

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3  $y_1 = y_1$ 

4  $y_1 = y_1$ 

5  $y_1 = y_1$ 

6  $y_1 = y_1$ 

7  $y_1 = y_1$ 

9  $y_1 = y_1$ 

1  $y$ 

# Problem 3

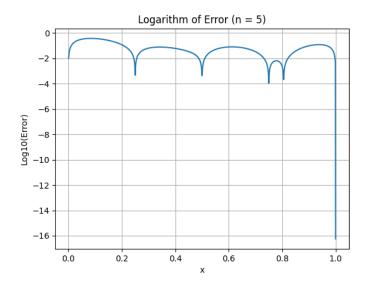
#### Solution

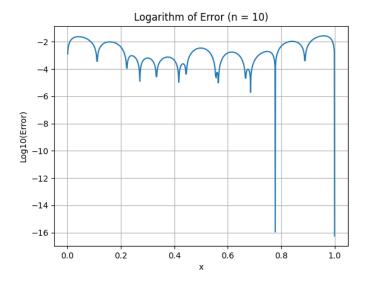
#### Github Link for Q3

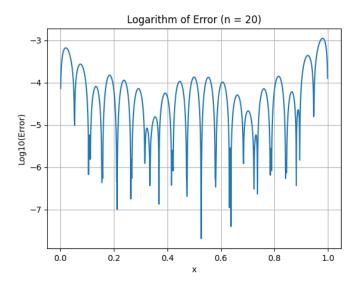
- a) A natural spline is one such that the boundary conditions of the second derivatives of the spline function at the endpoints of the interpolation interval are set to zero. s''(a) = s''(b) = 0.
  - A periodic spline is one such that the function values and their first and second derivatives match at both endpoints. s'(a) = s'(b), s''(a) = s''(b).
- b) For the periodic case, we need to adjust the boundary conditions. Specifically, we would need to ensure that the spline function "wraps around" at the endpoints. Then, we will get an  $n \times n$  linear system of the form:

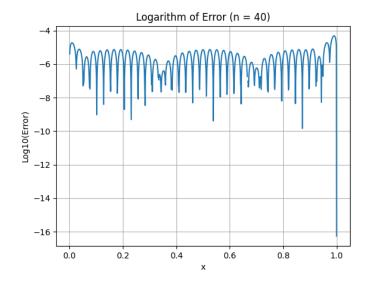
$$\frac{1}{12} \begin{bmatrix} 4 & 1 & 0 & \dots & 0 & 1 \\ 1 & 4 & 1 & \dots & 0 & 0 \\ 0 & 1 & 4 & \dots & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & \dots & 4 & 1 \\ 1 & 0 & 0 & \dots & 1 & 4 \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ a_2 \\ \vdots \\ a_{n-2} \\ a_{n-1} \end{bmatrix} = \begin{bmatrix} f[x_{-1}, x_0, x_1] \\ f[x_0, x_1, x_2] \\ f[x_1, x_2, x_3] \\ \vdots \\ f[x_{n-2}, x_{n-1}, x_0] \\ f[x_{n-2}, x_{n-1}, x_n] \end{bmatrix}$$

c) I have found the error while running the periodic type, so I do not think the figures below are the correct error graphs.









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# Problem 4

#### Solution

a) 
$$\boldsymbol{G} = \boldsymbol{M}^T \cdot \boldsymbol{M}$$

$$\boldsymbol{M} = \begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 2 \\ 1 & 3 \end{bmatrix}$$

$$\boldsymbol{\vec{y}} = \begin{bmatrix} 1 \\ 4 \\ 2 \\ 6 \end{bmatrix}$$

$$\boldsymbol{\vec{a}} = \begin{bmatrix} a_0 \\ a_1 \end{bmatrix} = \begin{bmatrix} 1.3 \\ 1.3 \end{bmatrix}$$

$$m{M}^T \cdot m{M} \cdot m{ec{a}} = m{M}^T m{ec{y}}$$
  $egin{bmatrix} 1 & 1 & 1 & 1 \ 0 & 1 & 2 & 3 \end{bmatrix} egin{bmatrix} 1 & 0 \ 1 & 1 \ 1 & 2 \ 1 & 3 \end{bmatrix} egin{bmatrix} a_0 \ a_1 \end{bmatrix} = egin{bmatrix} 1 & 1 & 1 & 1 \ 0 & 1 & 2 & 3 \end{bmatrix} egin{bmatrix} 1 \ 4 \ 2 \ 6 \end{bmatrix}$ 

b) 
$$\mathbf{w} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 \\ 0 & 0 & 3 & 0 \\ 0 & 0 & 0 & \sqrt{6} \end{bmatrix}$$

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 \\ 0 & 0 & 3 & 0 \\ 0 & 0 & 0 & \sqrt{6} \end{bmatrix} \begin{pmatrix} \begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 2 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \end{bmatrix} - \begin{bmatrix} 1 \\ 4 \\ 2 \\ 6 \end{bmatrix} \end{pmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$1(a_0 - 1) = 0 \rightarrow a_0 = 1$$

$$2(a_0 + a_1 - 4) = 0 \rightarrow 2a_0 + 2a_1 = 8$$

$$2(a_0 + a_1 - 4) = 0 \rightarrow 2a_0 + 2a_1 = 8$$
$$3(a_0 + 2a_1 - 2) = 0 \rightarrow 3a_0 + 6a_1 = 6$$
$$\sqrt{6}(a_0 + 3a_1 - 6) = 0 \rightarrow \sqrt{6}a_0 + 3\sqrt{6}a_1 = 6\sqrt{6}$$

Normal equation is:

$$\underbrace{\begin{bmatrix} 1 & 0 \\ 2 & 2 \\ 3 & 6 \\ \sqrt{6} & 3\sqrt{6} \end{bmatrix}}_{A} \underbrace{\begin{bmatrix} a_0 \\ a_1 \end{bmatrix}}_{a_w} = \underbrace{\begin{bmatrix} 1 \\ 8 \\ 6 \\ 6\sqrt{6} \end{bmatrix}}_{\vec{y_2}}$$

c) 
$$\boldsymbol{a_w} = (\boldsymbol{M}^T \cdot \boldsymbol{A})^{-1} (\boldsymbol{M}^T \cdot \boldsymbol{\vec{y_2}})$$
  
 $\boldsymbol{a_w} = \begin{bmatrix} 1.2573 \\ 1.2427 \end{bmatrix}$ 

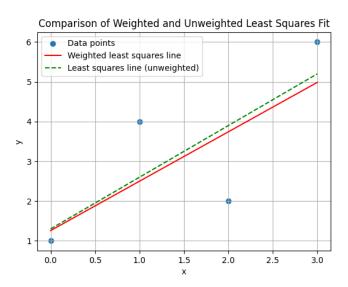


Figure 6: Comparison of Weighted and Unweighted Least Squares Fit

The effect of the weights on the linear fit is to adjust the influence of each data point on the fitting process. In the weighted least squares approach, data points with higher weights contribute more to the fitting process, while data points with lower weights have less impact.

For the code, click Github Link for 4