

# Lab-7

## Lagrange's and Newton Interpolation

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### Polynomial Interpolation Problem

#### Problem Statement:

Given the following table of data:

x	3.35	3.40	3.50	3.60
f(x)	0.298507	0.294118	0.285714	0.277778

#### Tasks:

- (a) Produce Lagrange polynomials of the linear, quadratic, and cubic orders with increasing values of x.
- (b) Produce Newton's divided-difference polynomial for all the three foregoing orders.
- (c) Plot the results of both methods on the same graph and compare them with the function

### Part - (a) -> Lagrange's Method

```
import matplotlib.pyplot as plt
import numpy as np

def mypolyint(data):
    n = len(data)
    x_values = [point[0] for point in data]
    y_values = [point[1] for point in data]
    coeffs = [0] * n
    for i in range(n):
        p = [1]
        for j in range(n):
            if i != j:
                p = poly_mul(p, [-x_values[j] / (x_values[i] - x_values[j]), 1
/ (x_values[i] - x_values[j])])
```

```

        coeffs = poly_add(coeffs, poly_mul_scalar(p, y_values[i]))

equation = "P(x) = "
for i, c in enumerate(coeffs):
    if i > 0:
        equation += " + "
        equation += f"{c:.5f}x^{i}"
print(equation)
print("Coefficients:", coeffs)

x_range=np.linspace(3, 5.5, 100)
y_interp = [lagrange_interpolation(x_val, x_values, y_values) for
x_val in x_range]
plt.plot(x_range, y_interp,color='green')
plt.scatter(x_values, y_values, color='red')
plt.title('Lagrange Interpolation')
plt.xlabel('x')
plt.ylabel('y')
plt.show()

return coeffs

def lagrange_interpolation(x, x_values, y_values):
    n = len(x_values)
    interpolated_y = 0
    for i in range(n):
        L_i = 1
        for j in range(n):
            if i != j:
                L_i *= (x - x_values[j]) / (x_values[i] - x_values[j])
        interpolated_y += y_values[i] * L_i
    return interpolated_y

def poly_mul(p1, p2):
    n1 = len(p1)
    n2 = len(p2)
    result = [0] * (n1 + n2 - 1)
    for i in range(n1):
        for j in range(n2):
            result[i + j] += p1[i] * p2[j]
    return result

def poly_add(p1, p2):
    n1 = len(p1)
    n2 = len(p2)
    result = [0] * max(n1, n2)
    for i in range(n1):
        result[i] += p1[i]
    for i in range(n2):

```

```

    result[i] += p2[i]
    return result

def poly_mul_scalar(p, s):
    return [c * s for c in p]

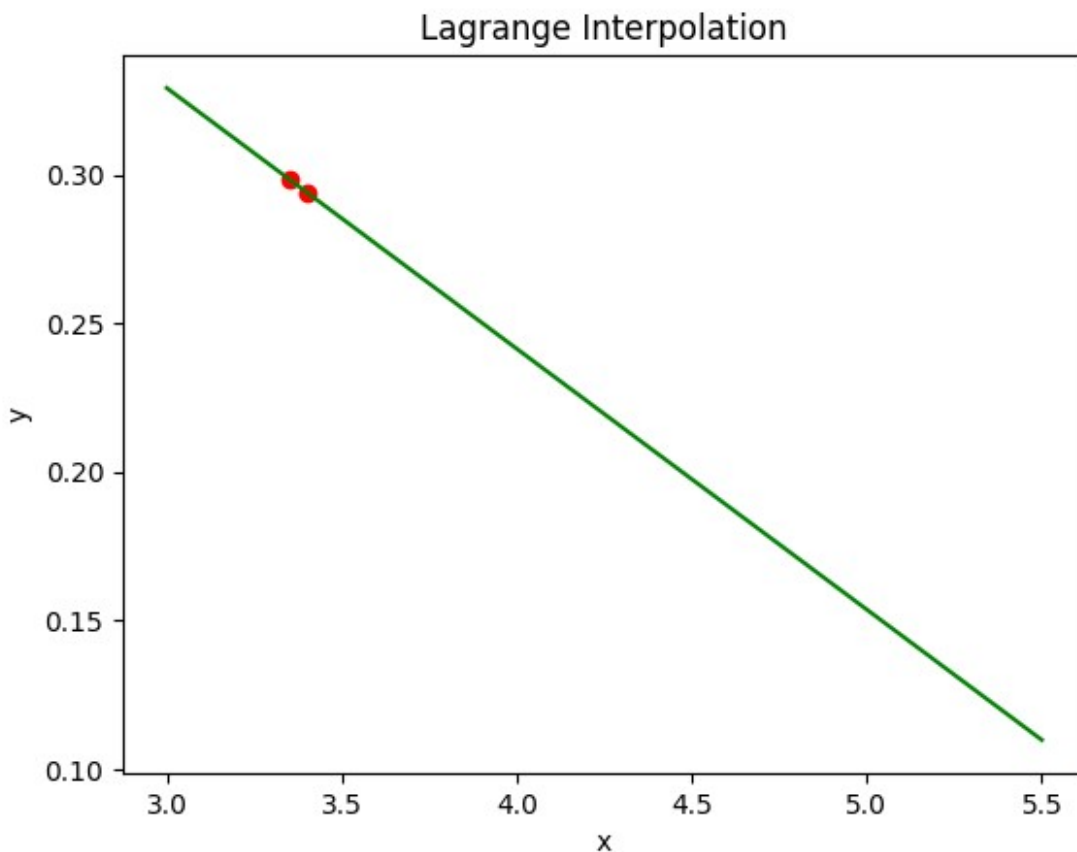
data1 = [[3.35,0.298507], [3.40,0.294118]]
coeffs1 = mypolyint(data1)

data2 = [[3.35,0.298507], [3.40,0.294118],[3.50,0.285714]]
coeffs2 = mypolyint(data2)

data3 = [[3.35,0.298507], [3.40,0.294118],[3.50,0.285714],
[3.60,0.277778]]
coeffs3 = mypolyint(data3)

P(x) = 0.59257x^0 + -0.08778x^1
Coefficients: [0.5925700000000002, -0.08778000000000013]

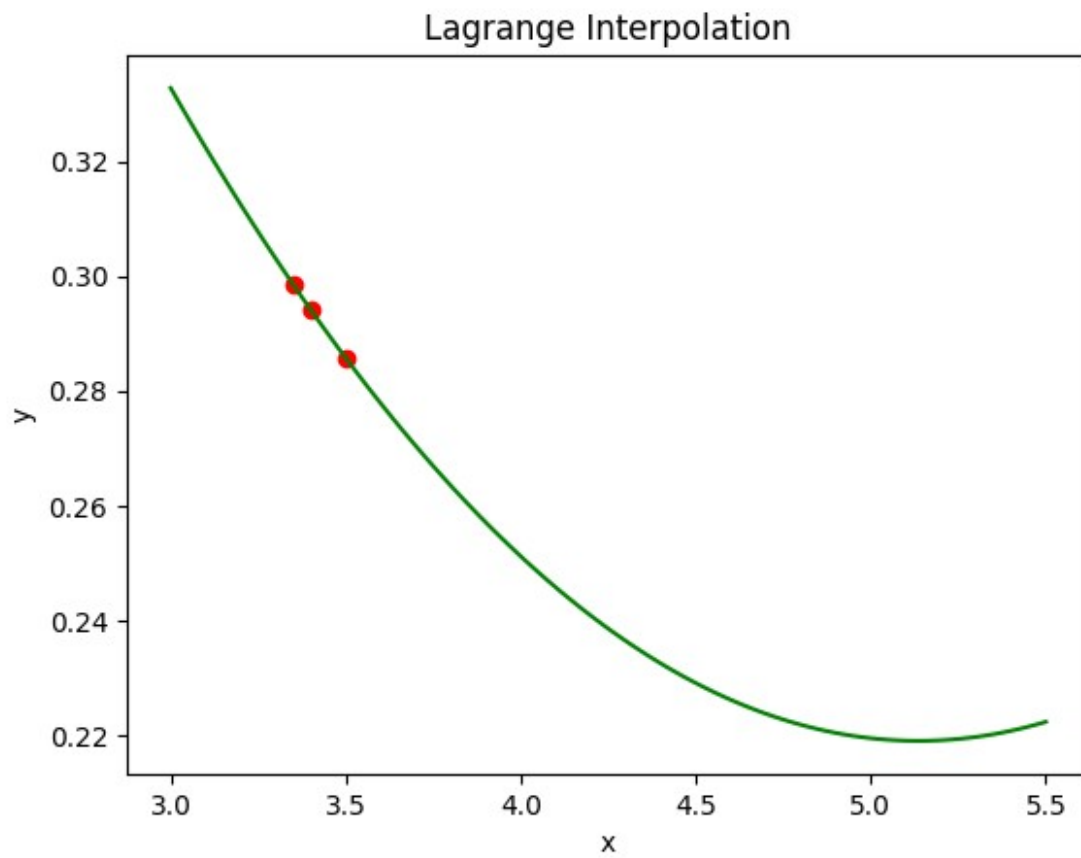
```



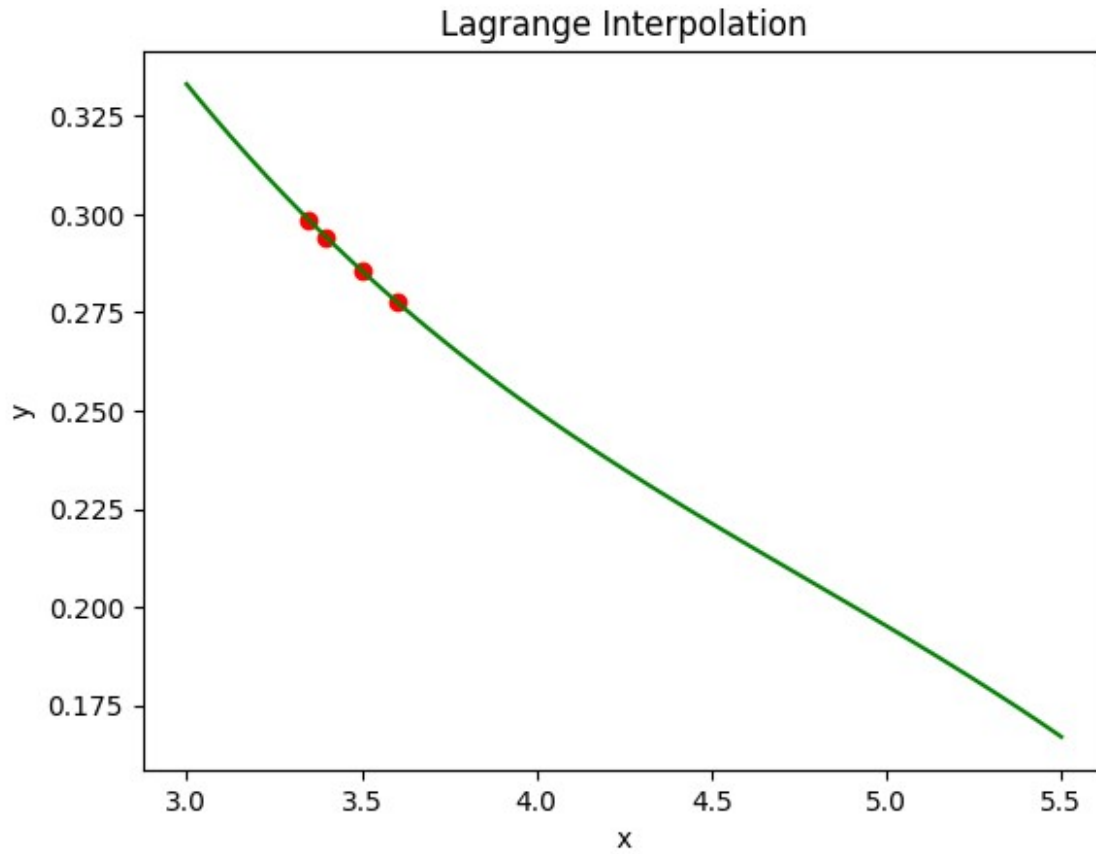
```

P(x) = 0.87656x^0 + -0.25608x^1 + 0.02493x^2
Coefficients: [0.87656066666667338, -0.2560800000000008246,
0.0249333333333334391]

```



$P(x) = 1.12107x^0 + -0.47084x^1 + 0.08780x^2 + -0.00613x^3$   
Coefficients: [1.1210660000010648, -0.4708386666675324,  
0.087800000000047039, -0.006133333333366409]



## Part - (b) -> Newton's Method

```
import numpy as np
import matplotlib.pyplot as plt

def mynewtonint(data):
    n = len(data)
    x_values = [point[0] for point in data]
    y_values = [point[1] for point in data]

    # Calculate divided differences
    f = np.zeros((n, n))
    f[:,0] = y_values
    for j in range(1,n):
        for i in range(n-j):
            f[i][j] = (f[i+1][j-1] - f[i][j-1]) / (x_values[i+j]-
x_values[i])

    # Extract coefficients
    coeffs = f[0,:]

    # Print the polynomial equation
```

```

equation = "P(x) = "
for i, c in enumerate(coeffs):
    if i > 0:
        equation += " + "
    equation += f"{c:.2f}"
    for j in range(i):
        equation += f"(x - {x_values[j]:.5f})"
print(equation)

# Expand the polynomial and collect coefficients
expanded_coeffs = expand_polynomial(coeffs, x_values)

# Print the expanded polynomial equation
expanded_equation = "P(x) = "
for i, c in enumerate(expanded_coeffs):
    if i > 0:
        if c >= 0:
            expanded_equation += " + "
        else:
            expanded_equation += " - "
            c = -c
        expanded_equation += f"{c:.5f}x^{i}"
print(expanded_equation)

# Generate points for plotting
x_range = np.linspace(3, 5.5, 100)
y_interp = [newton_interpolation(x_val, x_values, coeffs) for x_val
in x_range]

# Plot the polynomial and the data points
plt.plot(x_range, y_interp, color='pink')
plt.scatter(x_values, y_values, color='red')
plt.title('Newton Interpolation')
plt.xlabel('x')
plt.ylabel('y')
plt.show()

return coeffs

def newton_interpolation(x, x_values, coeffs):
    n = len(x_values)
    result = coeffs[n-1]
    for i in range(n-2, -1, -1):
        result = result * (x - x_values[i]) + coeffs[i]
    return result

def expand_polynomial(coeffs, x_values):
    n = len(coeffs)
    expanded_coeffs = [0] * n
    for i in range(n):

```

```

    term = [coeffs[i]]
    for j in range(i):
        term = poly_mul(term, [-x_values[j], 1])
    expanded_coeffs = poly_add(expanded_coeffs, term)
return expanded_coeffs

def poly_mul(p1, p2):
    n1 = len(p1)
    n2 = len(p2)
    result = [0] * (n1 + n2 - 1)
    for i in range(n1):
        for j in range(n2):
            result[i + j] += p1[i] * p2[j]
    return result

def poly_add(p1, p2):
    n1 = len(p1)
    n2 = len(p2)
    result = [0] * max(n1, n2)
    for i in range(n1):
        result[i] += p1[i]
    for i in range(n2):
        result[i] += p2[i]
    return result

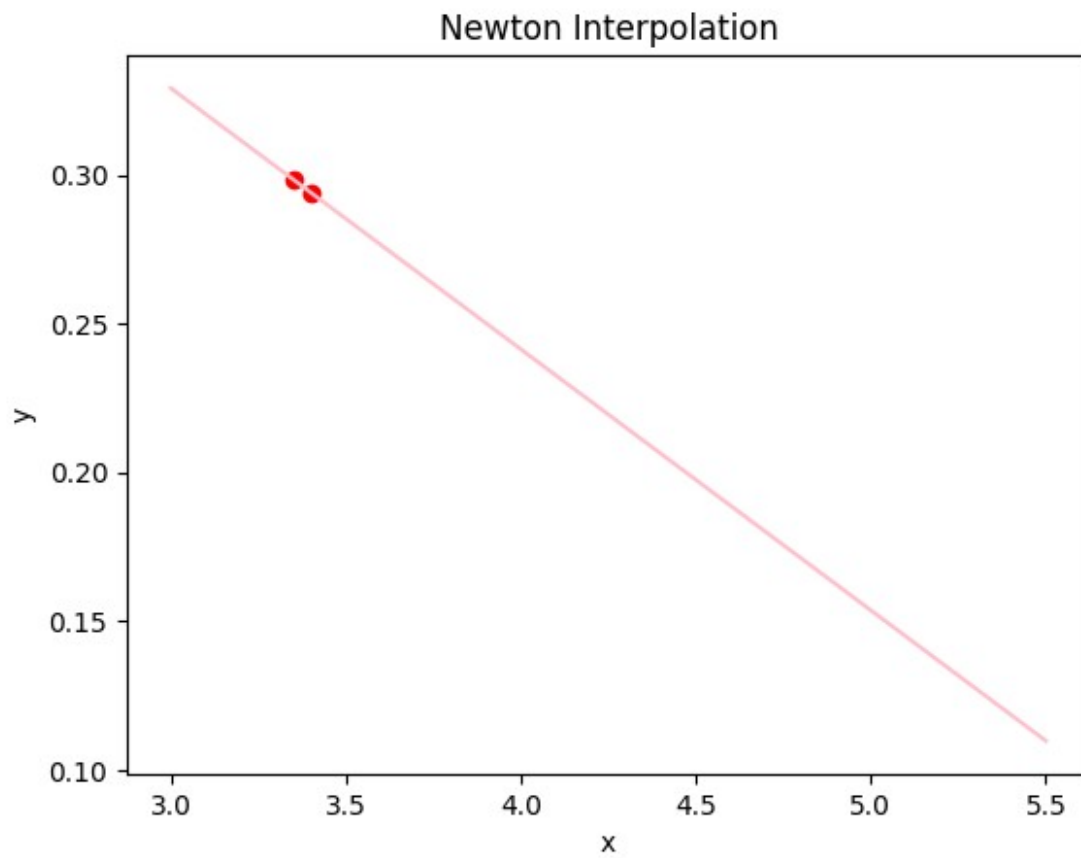
data1 = [[3.35,0.298507], [3.40,0.294118]]
coeffs1 = mynewtonint(data1)

data2 = [[3.35,0.298507], [3.40,0.294118],[3.50,0.285714]]
coeffs2 = mynewtonint(data2)

data3 = [[3.35,0.298507], [3.40,0.294118],[3.50,0.285714],
[3.60,0.277778]]
coeffs3 = mynewtonint(data3)

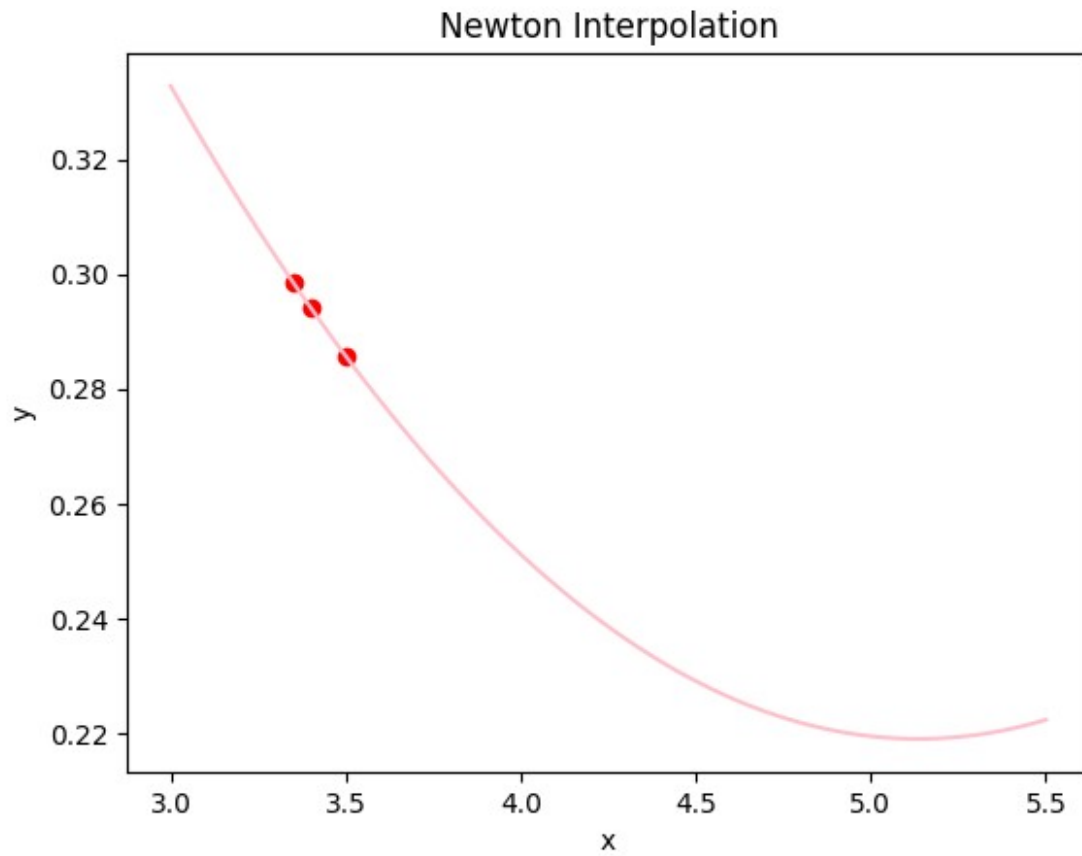
P(x) = 0.30 + -0.09(x - 3.35000)
P(x) = 0.59257x^0 - 0.08778x^1

```

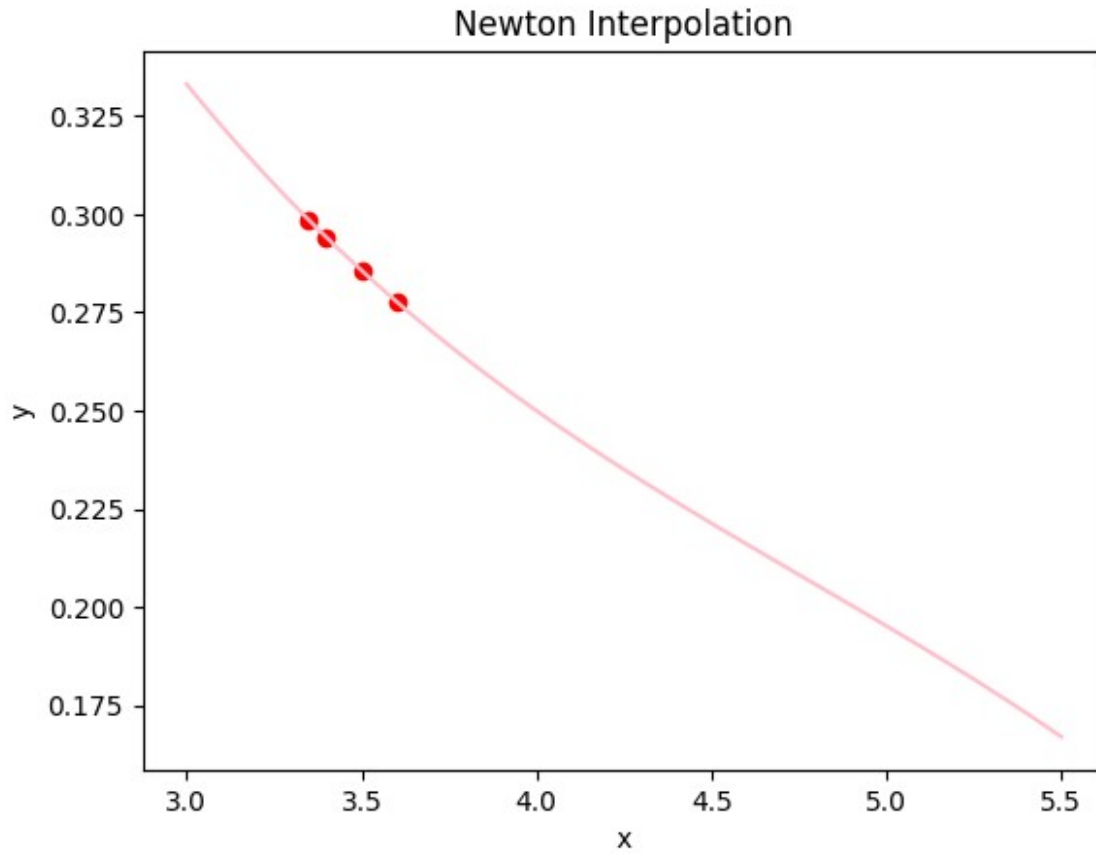


$$P(x) = 0.30 + -0.09(x - 3.35000) + 0.02(x - 3.35000)(x - 3.40000)$$
$$P(x) = 0.87656x^0 - 0.25608x^1 + 0.02493x^2$$





$$P(x) = 0.30 + -0.09(x - 3.35000) + 0.02(x - 3.35000)(x - 3.40000) + -0.01(x - 3.35000)(x - 3.40000)(x - 3.50000)$$
$$P(x) = 1.12107x^0 - 0.47084x^1 + 0.08780x^2 - 0.00613x^3$$



## Part - (c) Comparison with $y=1/x$

```
import numpy as np
import matplotlib.pyplot as plt

def mynewtonint(data, plot_title):
    n = len(data)
    x_values = [point[0] for point in data]
    y_values = [point[1] for point in data]

    # Calculate divided differences
    f = np.zeros((n, n))
    f[:, 0] = y_values
    for j in range(1, n):
        for i in range(n-j):
            f[i][j] = (f[i+1][j-1] - f[i][j-1]) / (x_values[i+j] -
x_values[i])

    # Extract coefficients
    coeffs = f[0, :]

    # Print the polynomial equation
```

```

equation = "P(x) = "
for i, c in enumerate(coeffs):
    if i > 0:
        equation += " + "
    equation += f"{c:.2f}"
    for j in range(i):
        equation += f"(x - {x_values[j]:.5f})"
print(equation)

# Expand the polynomial and collect coefficients
expanded_coefs = expand_polynomial(coefs, x_values)

# Print the expanded polynomial equation
expanded_equation = "P(x) = "
for i, c in enumerate(expanded_coefs):
    if i > 0:
        if c >= 0:
            expanded_equation += " + "
        else:
            expanded_equation += " - "
            c = -c
    expanded_equation += f"{c:.5f}x^{i}"
print(expanded_equation)

x_range = np.linspace(3, 5.52, 100)
y_interp = [newton_interpolation(x_val, x_values, coefs) for
x_val in x_range]

y_actual = 1 / x_range

plt.plot(x_range, y_interp, color='pink', label='Newton
Interpolation')
plt.plot(x_range, y_actual, color='blue', linestyle='--', label='y
= 1/x') # Add y=1/x
plt.scatter(x_values, y_values, color='red', zorder=5, label='Data
Points')

plt.title(plot_title)
plt.xlabel('x')
plt.ylabel('y')
plt.legend()
plt.show()

return coefs

def newton_interpolation(x, x_values, coefs):
    n = len(x_values)
    result = coefs[n-1]

```

```

    for i in range(n-2, -1, -1):
        result = result * (x - x_values[i]) + coeffs[i]
    return result

def expand_polynomial(coeffs, x_values):
    n = len(coeffs)
    expanded_coeffs = [0] * n
    for i in range(n):
        term = [coeffs[i]]
        for j in range(i):
            term = poly_mul(term, [-x_values[j], 1])
        expanded_coeffs = poly_add(expanded_coeffs, term)
    return expanded_coeffs

def poly_mul(p1, p2):
    n1 = len(p1)
    n2 = len(p2)
    result = [0] * (n1 + n2 - 1)
    for i in range(n1):
        for j in range(n2):
            result[i + j] += p1[i] * p2[j]
    return result

def poly_add(p1, p2):
    n1 = len(p1)
    n2 = len(p2)
    result = [0] * max(n1, n2)
    for i in range(n1):
        result[i] += p1[i]
    for i in range(n2):
        result[i] += p2[i]
    return result

# Data sets for different interpolation orders
data1 = [[3.35, 0.298507], [3.40, 0.294118]]
coeffs1 = mynewtonint(data1, 'Linear Interpolation vs y = 1/x')

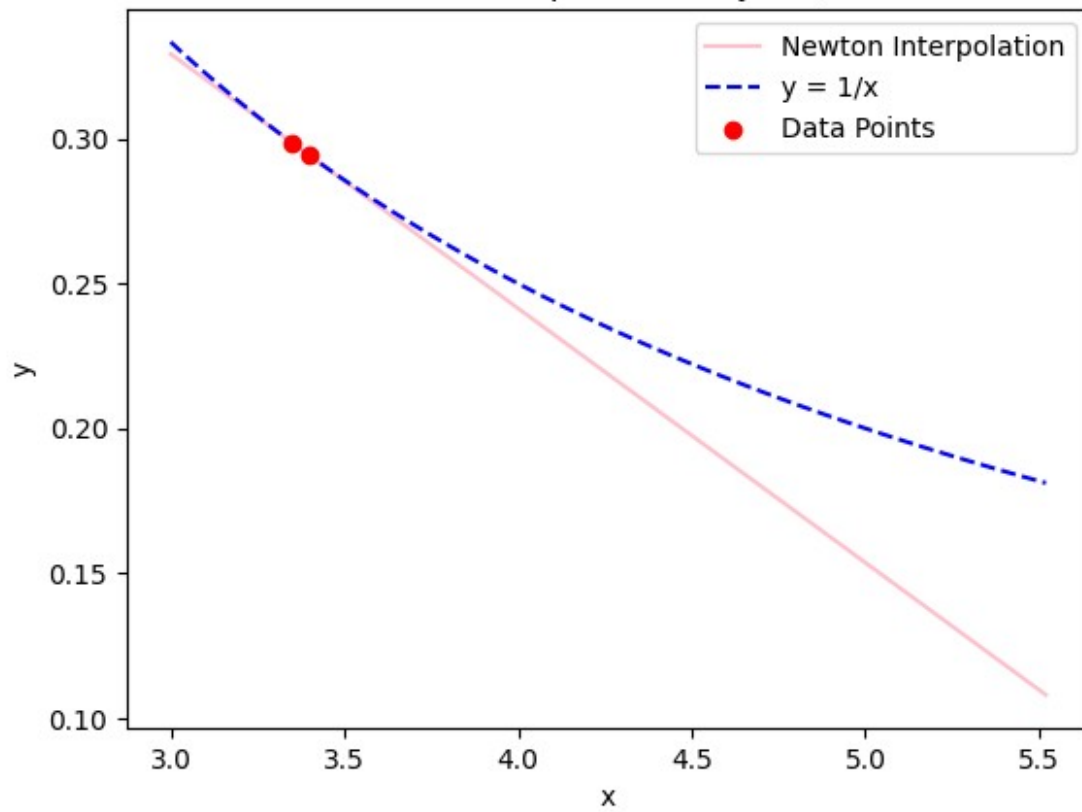
data2 = [[3.35, 0.298507], [3.40, 0.294118], [3.50, 0.285714]]
coeffs2 = mynewtonint(data2, 'Quadratic Interpolation vs y = 1/x')

data3 = [[3.35, 0.298507], [3.40, 0.294118], [3.50, 0.285714], [3.60,
0.277778]]
coeffs3 = mynewtonint(data3, 'Cubic Interpolation vs y = 1/x')

P(x) = 0.30 + -0.09(x - 3.35000)
P(x) = 0.59257x^0 - 0.08778x^1

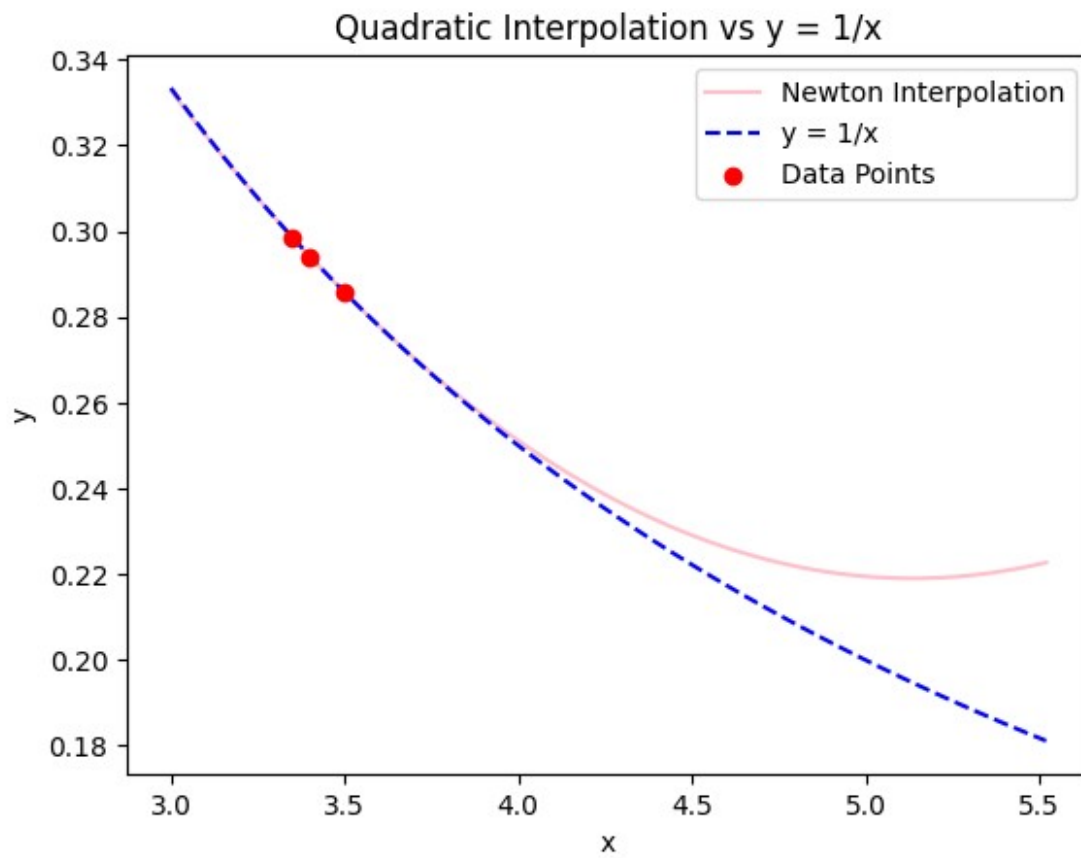
```

Linear Interpolation vs  $y = 1/x$



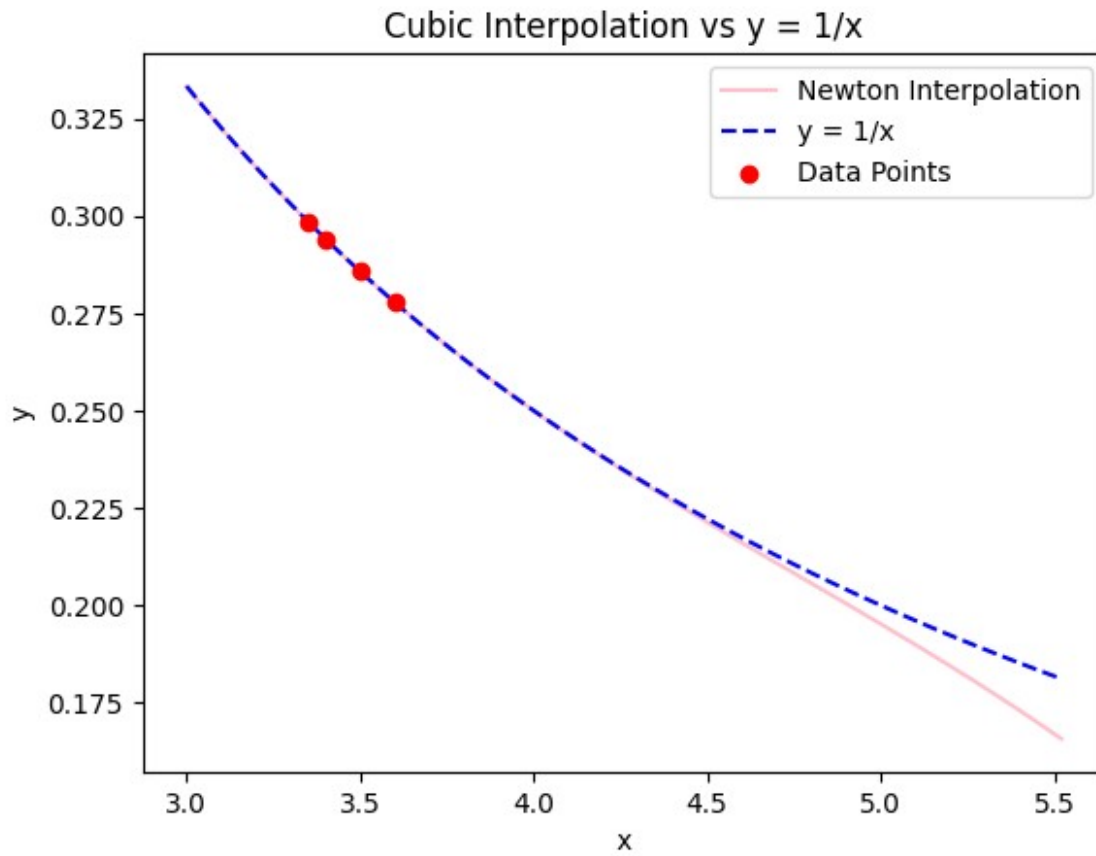
$$P(x) = 0.30 + -0.09(x - 3.35000) + 0.02(x - 3.35000)(x - 3.40000)$$

$$P(x) = 0.87656x^0 - 0.25608x^1 + 0.02493x^2$$



$$P(x) = 0.30 + -0.09(x - 3.35000) + 0.02(x - 3.35000)(x - 3.40000) + -0.01(x - 3.35000)(x - 3.40000)(x - 3.50000)$$

$$P(x) = 1.12107x^0 - 0.47084x^1 + 0.08780x^2 - 0.00613x^3$$



## Interpolation Problem

### Problem Statement:

Given the following table of data:

x	0	1	2	2.5	3	3.5	4
f(x)	2.5	0.5	0.5	1.5	1.5	1.125	0

### Tasks:

**(a)** Interpolate successive points by straight line segments. This method is known as **piecewise linear interpolation**.

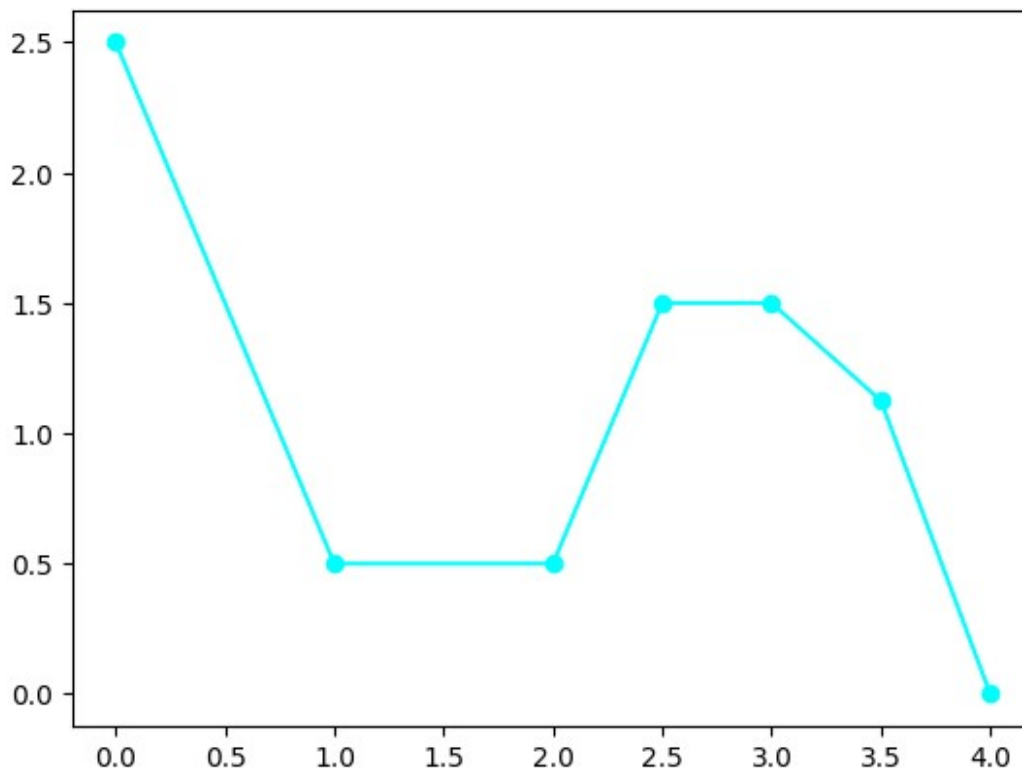
**(b)** On each of the following three subintervals of  $(x)$ :  $([0, 2])$ ,  $([2, 3])$ , and  $([3, 4])$ , interpolate using both **Lagrange's quadratic polynomial** and **Newton's divided-difference interpolation polynomial**.

**(c)** Plot the results of both methods covering all the three subintervals on the same graph, and compare them.

```
import numpy as np
import matplotlib.pyplot as plt

poly_fit=[2.5,0.5,0.5,1.5,1.5,1.125,0]
poly_fit_x=[0,1,2,2.5,3,3.5,4]
plt.plot(poly_fit_x, poly_fit, marker='o',color='cyan')

[<matplotlib.lines.Line2D at 0x7fed716bcc40>]
```



```
import matplotlib.pyplot as plt
import numpy as np

def mypolyint(data):
    n = len(data)
    x_values = [point[0] for point in data]
    y_values = [point[1] for point in data]
    coeffs = [0] * n
    for i in range(n):
        p = [1]
        for j in range(n):
            if i != j:
                p = poly_mul(p, [-x_values[j] / (x_values[i] - x_values[j]), 1
/ (x_values[i] - x_values[j])])
        coeffs = poly_add(coeffs, poly_mul_scalar(p, y_values[i]))
```



```

equation = "P(x) = "
for i, c in enumerate(coeffs):
    if i > 0:
        equation += " + "
        equation += f"{c:.5f}x^{i}"
print(equation)
print("Coefficients:", coeffs)

x_range=np.linspace(min(x_values), max(x_values), 100)
y_interp = [lagrange_interpolation(x_val, x_values, y_values) for
x_val in x_range]
plt.plot(x_range, y_interp)
plt.scatter(x_values, y_values, color='red')
plt.title('Lagrange Interpolation')
plt.xlabel('x')
plt.ylabel('y')
plt.show()

return coeffs

def lagrange_interpolation(x, x_values, y_values):
    n = len(x_values)
    interpolated_y = 0
    for i in range(n):
        L_i = 1
        for j in range(n):
            if i != j:
                L_i *= (x - x_values[j]) / (x_values[i] - x_values[j])
        interpolated_y += y_values[i] * L_i
    return interpolated_y

def poly_mul(p1, p2):
    n1 = len(p1)
    n2 = len(p2)
    result = [0] * (n1 + n2 - 1)
    for i in range(n1):
        for j in range(n2):
            result[i + j] += p1[i] * p2[j]
    return result

def poly_add(p1, p2):
    n1 = len(p1)
    n2 = len(p2)
    result = [0] * max(n1, n2)
    for i in range(n1):
        result[i] += p1[i]
    for i in range(n2):
        result[i] += p2[i]
    return result

```

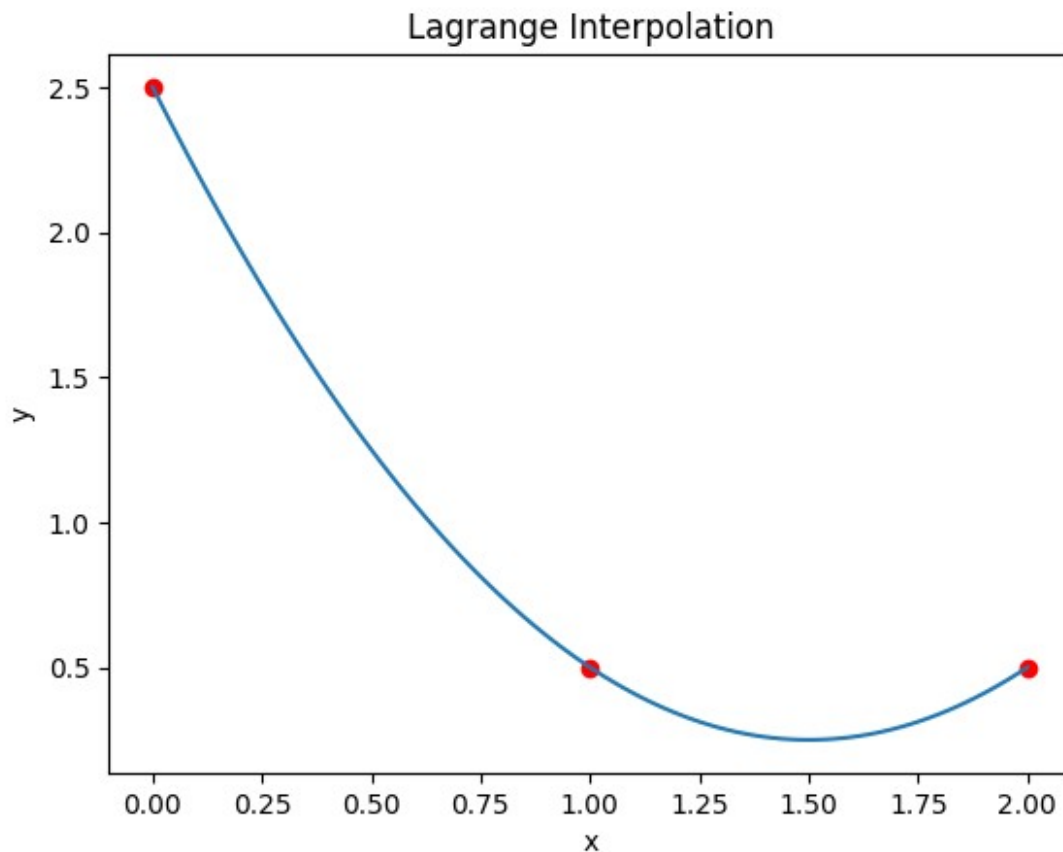
```
def poly_mul_scalar(p, s):
    return [c * s for c in p]

data1 = [[0,2.5],[1,0.5],[2,0.5]]
coeffs1 = mypolyint(data1)

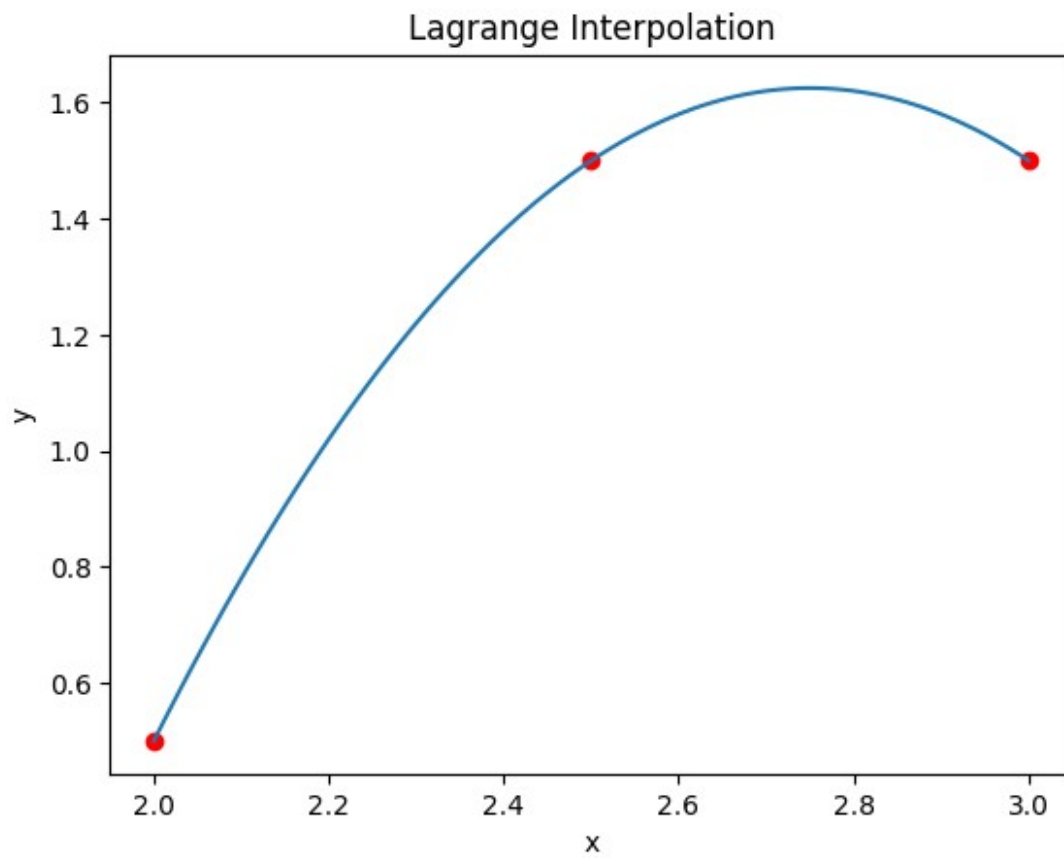
data2 = [[2,0.5],[2.5,1.5],[3,1.5]]
coeffs2 = mypolyint(data2)

data3 = [[3,1.5],[3.5,1.125],[4,0]]
coeffs3 = mypolyint(data3)
```

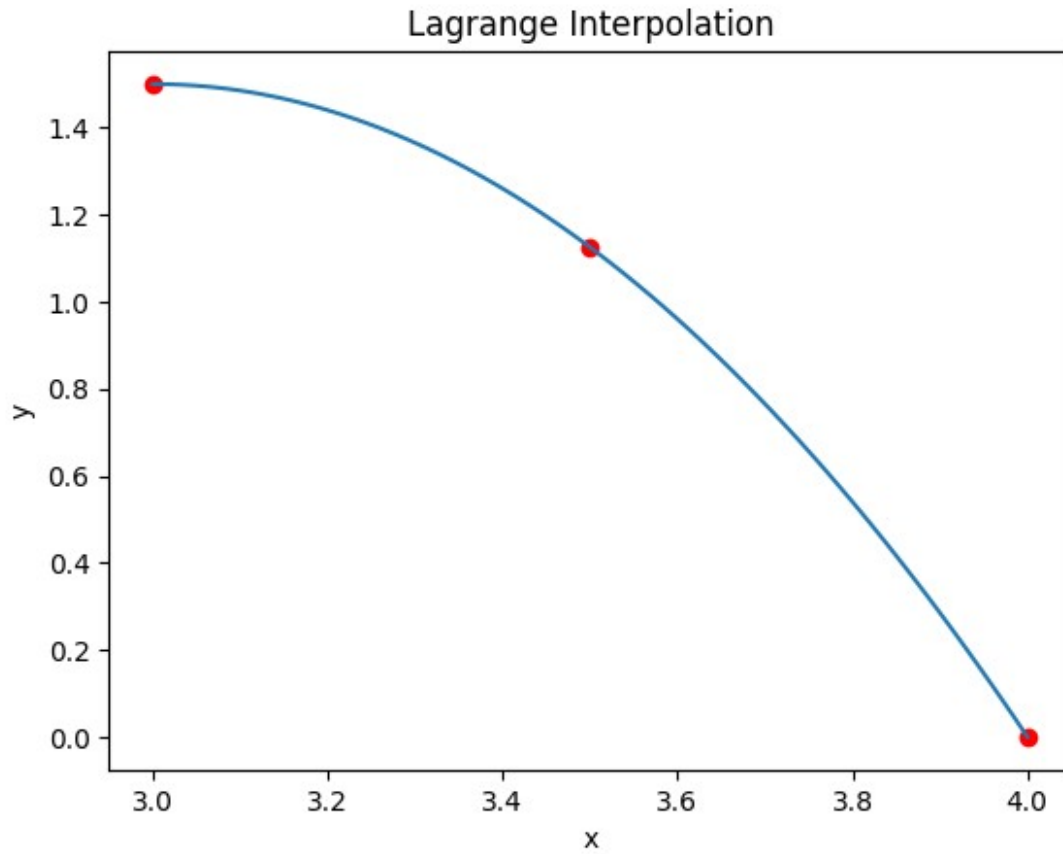
$P(x) = 2.50000x^0 + -3.00000x^1 + 1.00000x^2$   
Coefficients: [2.5, -3.0, 1.0]



$P(x) = -13.50000x^0 + 11.00000x^1 + -2.00000x^2$   
Coefficients: [-13.5, 11.0, -2.0]

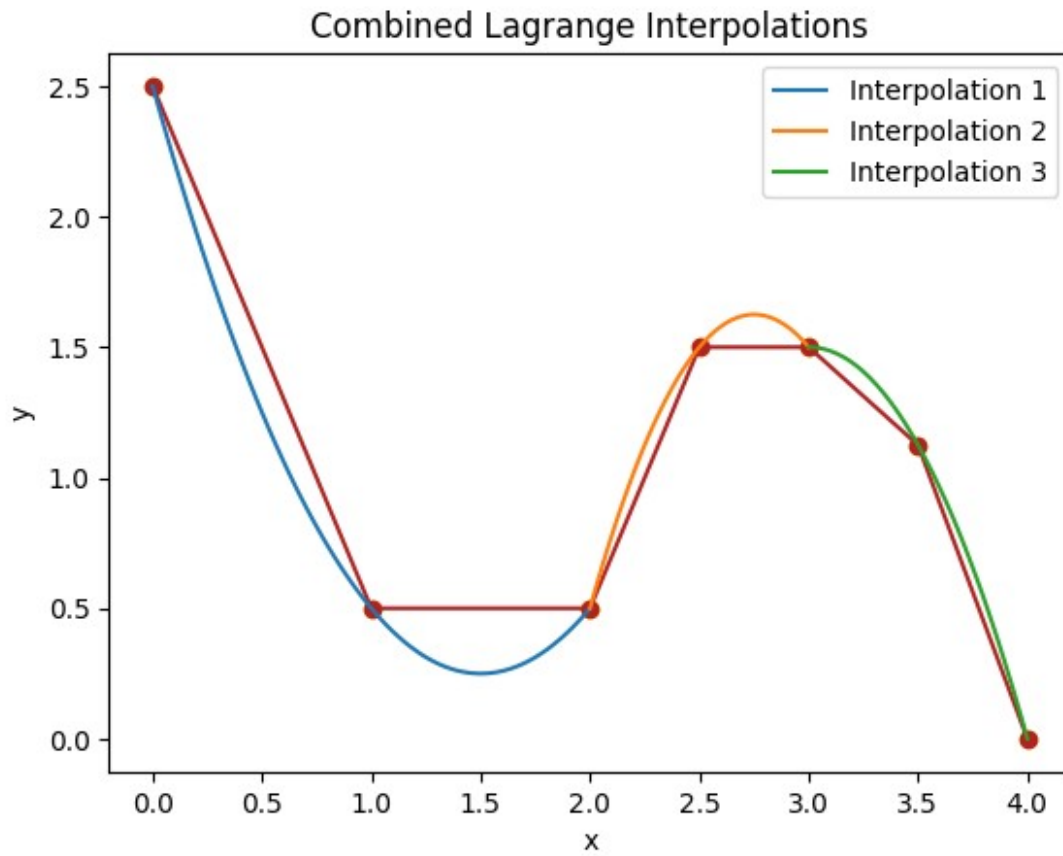


$P(x) = -12.00000x^0 + 9.00000x^1 + -1.50000x^2$   
Coefficients: [-12.0, 9.0, -1.5]



```
poly_fit=[2.5,0.5,0.5,1.5,1.5,1.125,0]
poly_fit_x=[0,1,2,2.5,3,3.5,4]
plt.plot(poly_fit_x, poly_fit, marker='o',color='firebrick')

plt.plot(x_range1, y_interp1, label='Interpolation 1')
plt.plot(x_range2, y_interp2, label='Interpolation 2')
plt.plot(x_range3, y_interp3, label='Interpolation 3')
plt.scatter([point[0] for point in data1] + [point[0] for point in
data2] + [point[0] for point in data3],
            [point[1] for point in data1] + [point[1] for point in
data2] + [point[1] for point in data3], color='yellow')
plt.title('Combined Lagrange Interpolations')
plt.xlabel('x')
plt.ylabel('y')
plt.legend()
plt.show()
```



```
import numpy as np
import matplotlib.pyplot as plt

def mynewtonint(data):
    n = len(data)
    x_values = [point[0] for point in data]
    y_values = [point[1] for point in data]

    # Calculate divided differences
    f = np.zeros((n, n))
    f[:,0] = y_values
    for j in range(1,n):
        for i in range(n-j):
            f[i][j] = (f[i+1][j-1] - f[i][j-1]) / (x_values[i+j]-
x_values[i])

    # Extract coefficients
    coeffs = f[0,:]

    # Print the polynomial equation
    equation = "P(x) = "
    for i, c in enumerate(coeffs):
        if i > 0:
            equation += " + " + str(c) + "x^" + str(i)
    equation += " = 0"
```

```

        equation += " + "
        equation += f"{c:.2f}"
        for j in range(i):
            equation += f"(x - {x_values[j]:.5f})"
    print(equation)

    # Expand the polynomial and collect coefficients
    expanded_coefs = expand_polynomial(coefs, x_values)

    # Print the expanded polynomial equation
    expanded_equation = "P(x) = "
    for i, c in enumerate(expanded_coefs):
        if i > 0:
            if c >= 0:
                expanded_equation += " + "
            else:
                expanded_equation += " - "
                c = -c
            expanded_equation += f"{c:.5f}x^{i}"
    print(expanded_equation)

    # Generate points for plotting
    x_range = np.linspace(min(x_values), max(x_values), 100)
    y_interp = [newton_interpolation(x_val, x_values, coefs) for x_val
in x_range]

    # Plot the polynomial and the data points
    plt.plot(x_range, y_interp, color='peachpuff')
    plt.scatter(x_values, y_values, color='red')
    plt.title('Newton Interpolation')
    plt.xlabel('x')
    plt.ylabel('y')
    plt.show()

    return coefs

def newton_interpolation(x, x_values, coefs):
    n = len(x_values)
    result = coefs[n-1]
    for i in range(n-2, -1, -1):
        result = result * (x - x_values[i]) + coefs[i]
    return result

def expand_polynomial(coefs, x_values):
    n = len(coefs)
    expanded_coefs = [0] * n
    for i in range(n):
        term = [coefs[i]]
        for j in range(i):
            term = poly_mul(term, [-x_values[j], 1])

```

```

    expanded_coeffs = poly_add(expanded_coeffs, term)
    return expanded_coeffs

def poly_mul(p1, p2):
    n1 = len(p1)
    n2 = len(p2)
    result = [0] * (n1 + n2 - 1)
    for i in range(n1):
        for j in range(n2):
            result[i + j] += p1[i] * p2[j]
    return result

def poly_add(p1, p2):
    n1 = len(p1)
    n2 = len(p2)
    result = [0] * max(n1, n2)
    for i in range(n1):
        result[i] += p1[i]
    for i in range(n2):
        result[i] += p2[i]
    return result

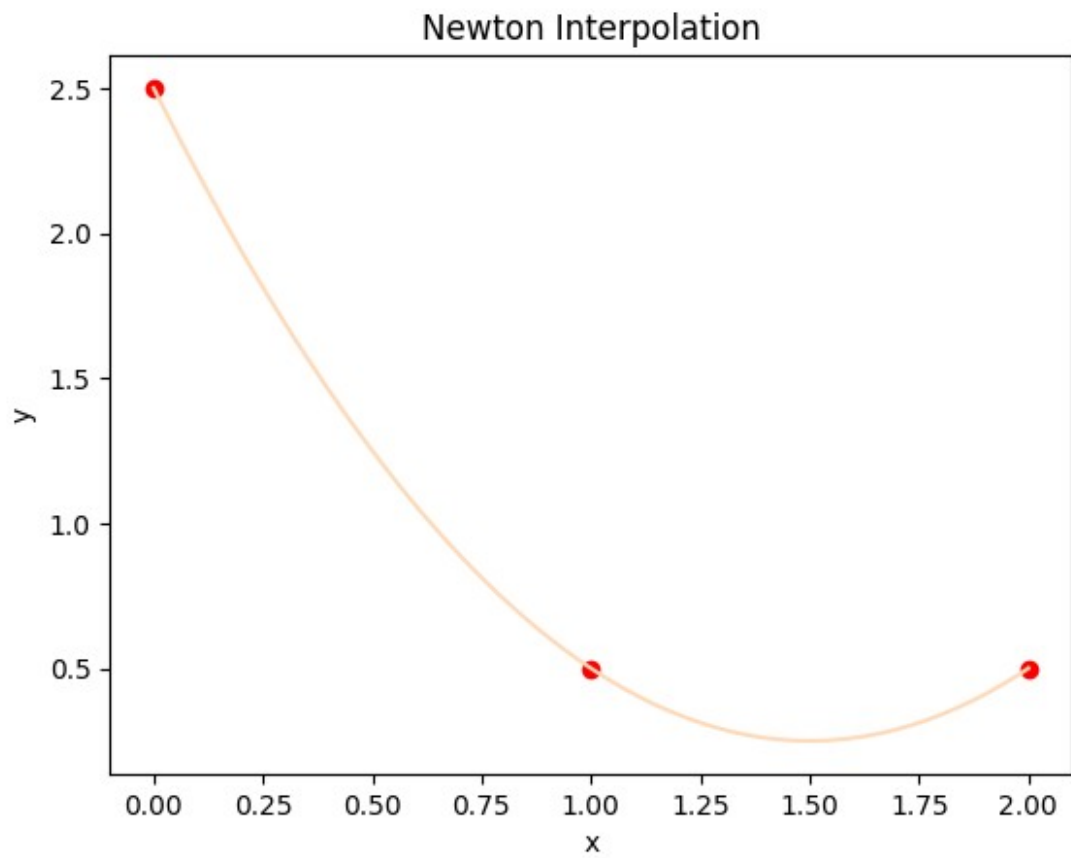
data1 = [[0,2.5],[1,0.5],[2,0.5]]
coeffs1 = mynewtonint(data1)

data2 = [[2,0.5],[2.5,1.5],[3,1.5]]
coeffs2 = mynewtonint(data2)

data3 = [[3,1.5],[3.5,1.125],[4,0]]
coeffs3 = mynewtonint(data3)

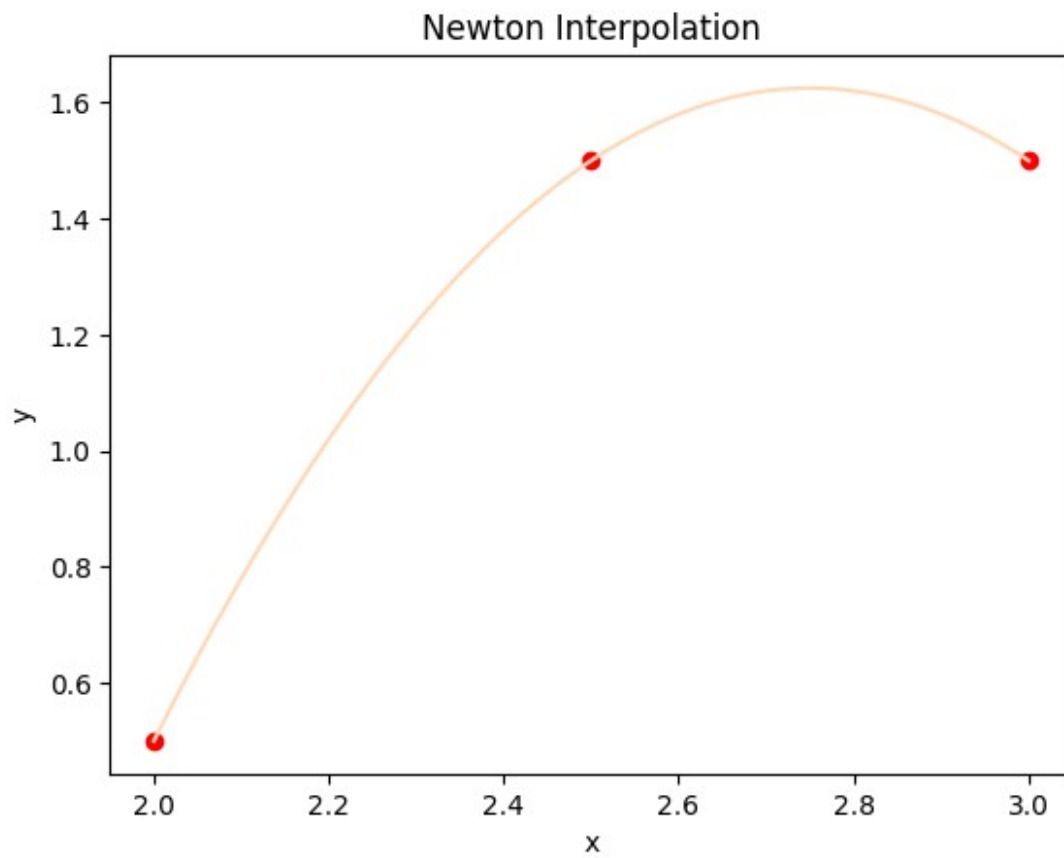
P(x) = 2.50 + -2.00(x - 0.00000) + 1.00(x - 0.00000)(x - 1.00000)
P(x) = 2.50000x^0 - 3.00000x^1 + 1.00000x^2

```

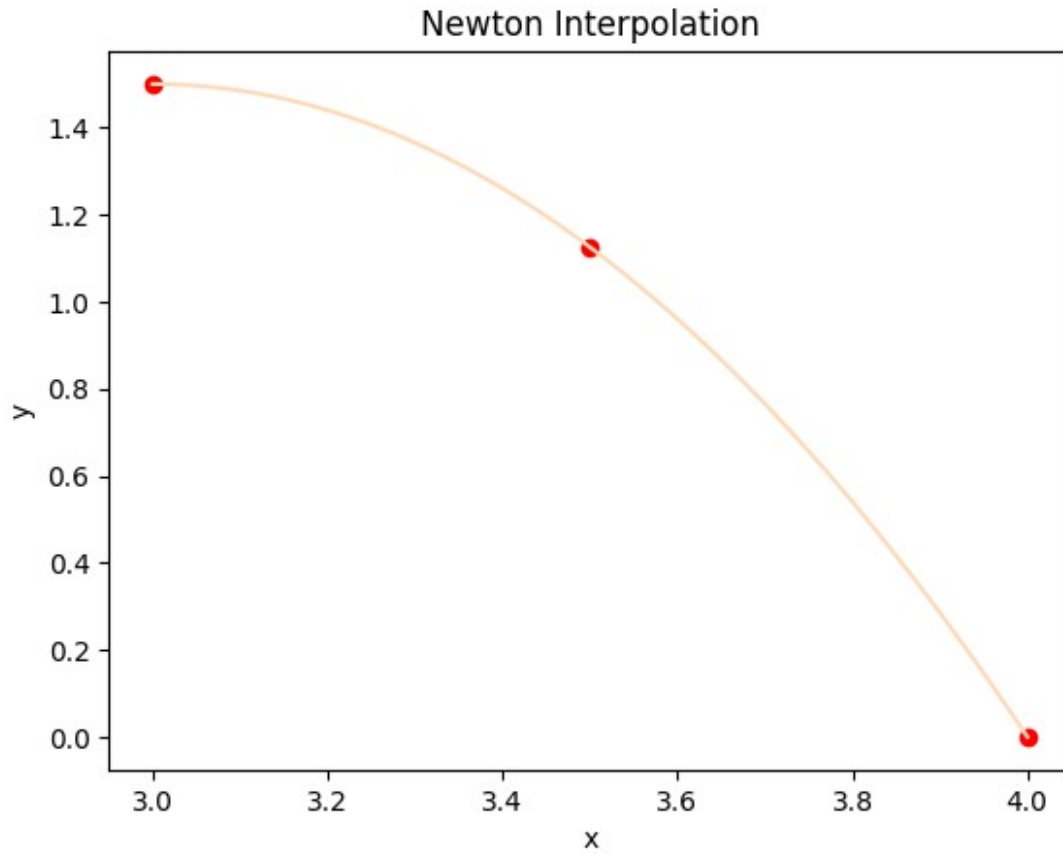


$$P(x) = 0.50 + 2.00(x - 2.00000) + -2.00(x - 2.00000)(x - 2.50000)$$
$$P(x) = -13.50000x^0 + 11.00000x^1 - 2.00000x^2$$



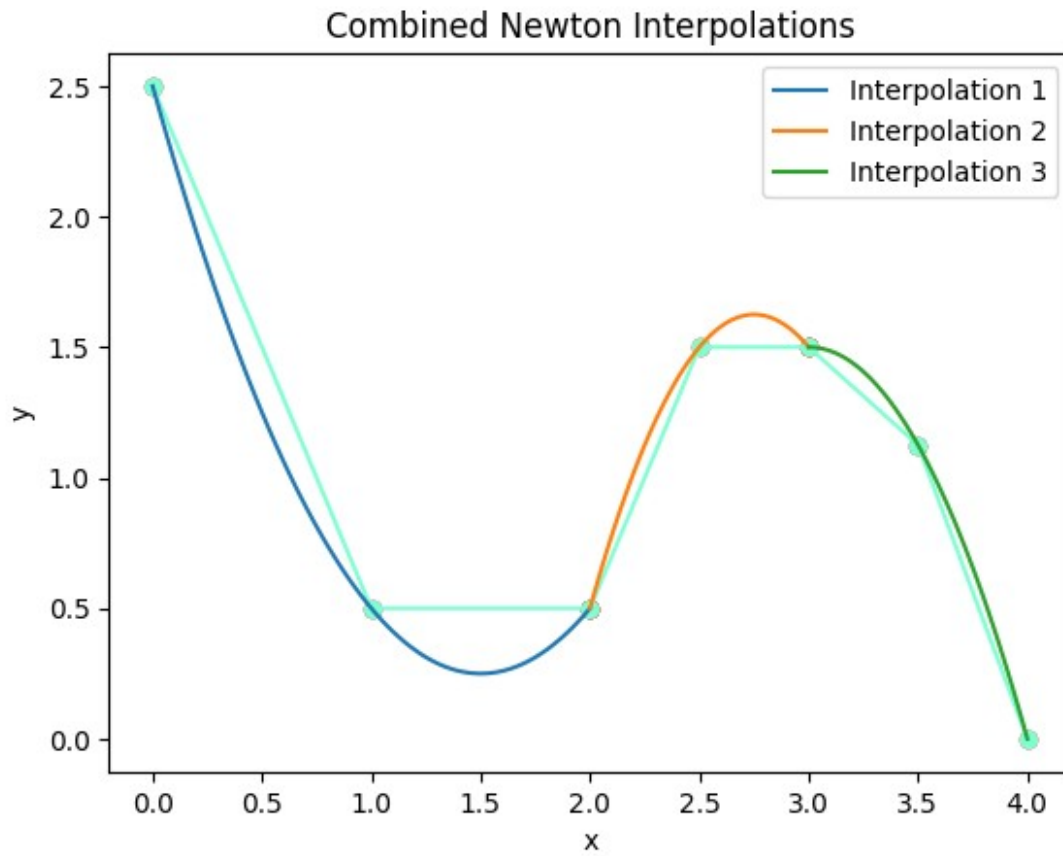


$$P(x) = 1.50 + -0.75(x - 3.00000) + -1.50(x - 3.00000)(x - 3.50000)$$
$$P(x) = -12.00000x^0 + 9.00000x^1 - 1.50000x^2$$



```
poly_fit=[2.5,0.5,0.5,1.5,1.5,1.125,0]
poly_fit_x=[0,1,2,2.5,3,3.5,4]
plt.plot(poly_fit_x, poly_fit, marker='o',color='aquamarine')

plt.plot(x_range1, y_interp1, label='Interpolation 1')
plt.plot(x_range2, y_interp2, label='Interpolation 2')
plt.plot(x_range3, y_interp3, label='Interpolation 3')
plt.scatter([point[0] for point in data1] + [point[0] for point in
data2] + [point[0] for point in data3],
            [point[1] for point in data1] + [point[1] for point in
data2] + [point[1] for point in data3], color='red')
plt.title('Combined Newton Interpolations')
plt.xlabel('x')
plt.ylabel('y')
plt.legend()
plt.show()
```



Part - (c)

Write a function that implements a least square polynomial fit. The function should take in two vectors  $x$  and  $y$  (each of size  $n \times 1$ ), and the degree of the polynomial ( $m < n$ ) and output the coefficient vector of the polynomial that minimizes the least square error.

```
import numpy as np
import matplotlib.pyplot as plt

def least_squares_polyfit_manual(x, y, degree):
    A = np.zeros((len(x), degree + 1))
    for i in range(degree + 1):
        A[:, i] = x**i

    ATA = np.dot(A.T, A)
    ATy = np.dot(A.T, y)

    coeffs = np.dot(np.linalg.inv(ATA), ATy)
    return coeffs
```

```
x = np.array([0, 1, 2, 3, 4])
y = np.array([2.5, 0.5, 0.5, 1.5, 0.0])

degree = 2
coeffs = least_squares_polyfit_manual(x, y, degree)

print("Polynomial coefficients:", coeffs)

x_range = np.linspace(min(x), max(x), 100)
y_fit = np.polyval(coeffs[::-1], x_range)

plt.plot(x_range, y_fit, label='Polynomial Fit')
plt.scatter(x, y, color='red', label='Data Points')
plt.xlabel('x')
plt.ylabel('y')
plt.legend()
plt.title('Least Squares Polynomial Fit (Degree=2)')
plt.show()

Polynomial coefficients: [ 2.08571429 -0.97142857  0.14285714]
```

