LAB-6

Newton and Lagrange's Polynomial Interpolation

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Lagrange Interpolation
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):
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result[i] += p1[i]
for i in range(n2):
    result[i] += p2[i]
    return result

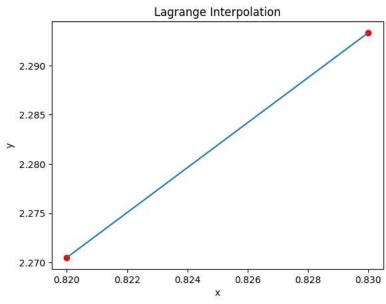
def poly_mul_scalar(p, s):
    return [c * s for c in p]

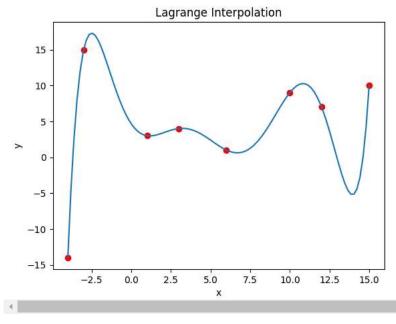
data1 = [[0.82, 2.2705], [0.83,2.293319]]
    coeffs1 = mypolyint(data1)

data2 = [[1,3], [3,4], [6,1], [-3, 15], [-4, -14], [15, 10], [10, 9], [12, 7]]
    coeffs2 = mypolyint(data2)

TYP P(x) = 0.39934x^0 + 2.28190x^1
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P(x) = 0.39934x^0 + 2.28190x^1 Coefficients: [0.3993420000004694, 2.28189999999999]





Newton Interpolation

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

def mynewtonint(data):
    n = len(data)
    x_values = [point[0] for point in data]
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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(min(x_values), max(x_values), 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)
 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
```

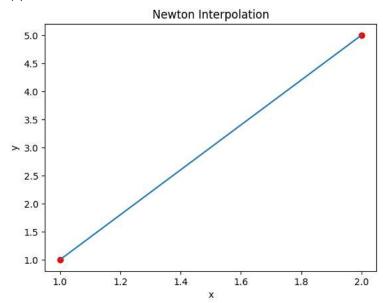
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def poly_add(p1, p2):
    n1 = len(p1)
    n2 = len(p2)
    result = [0] * max(n1, n2)
    for i in range(n1):
        result[i] += p1[i]
    for i in range(n2):
        result[i] += p2[i]
    return result

data1 = [[1, 1], [2,5]]
    coeffs1 = mynewtonint(data1)

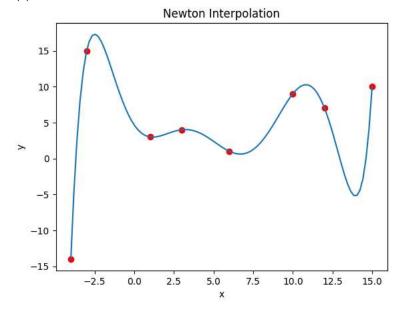
data2 = [[1,3], [3,4], [6,1], [-3, 15], [-4, -14], [15, 10], [10, 9], [12, 7]]
    coeffs2 = mynewtonint(data2)

    \( \rightarrow \text{P(x)} = 1.00 + 4.00(x - 1.00000) \)
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P(x) = 1.00 + 4.00(x - 1.00000) $P(x) = -3.00000x^0 + 4.00000x^1$



P(x) = 3.00 + 0.50(x - 1.00000) + -0.30(x - 1.00000)(x - 3.00000) + -0.10(x - 1.00000)(x - 3.00000)(x - 6.00000) + -0.11(x - 1.00000) $P(x) = 4.65535x^{0} - 3.23295x^{1} + 1.74638x^{2} - 0.07491x^{3} - 0.11718x^{4} + 0.02516x^{5} - 0.00189x^{6} + 0.00005x^{7}$



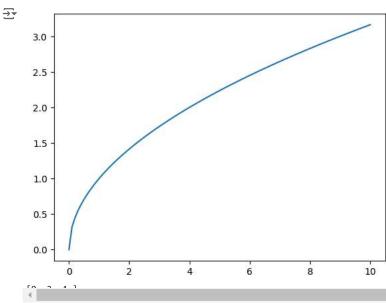
Newton Interpolation

```
import numpy as np
import matplotlib.pyplot as plt
def function():
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x=np.linspace(0,10,100)
y=(x)**(0.5)
plt.plot(x, y)
plt.show()

function()
start = 0
end = 4
num_points = 3

points = np.linspace(start, end, num_points)
print(points)
data1=[]
```



```
import numpy as np
import matplotlib.pyplot as plt
def f(x):
    return np.sqrt(x)
def newton_coefficients(x, y):
    n = len(x)
   coef = np.copy(y)
    for j in range(1, n):
        for i in range(n-1, j-1, -1):
            coef[i] = (coef[i] - coef[i-1]) / (x[i] - x[i-j])
    return coef
def newton_poly(x_vals, x, coef):
    n = len(coef)
    p = coef[-1]
    for i in range(n-2, -1, -1):
       p = p * (x_vals - x[i]) + coef[i]
    return p
x_interval = np.linspace(0, 4, 1000)
y_actual = f(x_interval)
n_{values} = [2, 4, 8, 16, 32]
max_errors = []
plt.figure(figsize=(10, 6))
plt.plot(x_interval, y_actual, label='f(x) = sqrt(x)', color='black')
for n in n_values:
    x_samples = np.linspace(0, 4, n+1)
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y_samples = f(x_samples)

coef = newton_coefficients(x_samples, y_samples)

y_interp = newton_poly(x_interval, x_samples, coef)

plt.plot(x_interval, y_interp, label=f'f_{n}(x)')

error = np.abs(y_actual - y_interp)
 max_errors.append(np.max(error))

plt.title("Function and Interpolations")
plt.xlabel("x")
plt.ylabel("f(x)")
plt.legend()
plt.grid(True)
plt.show()
```

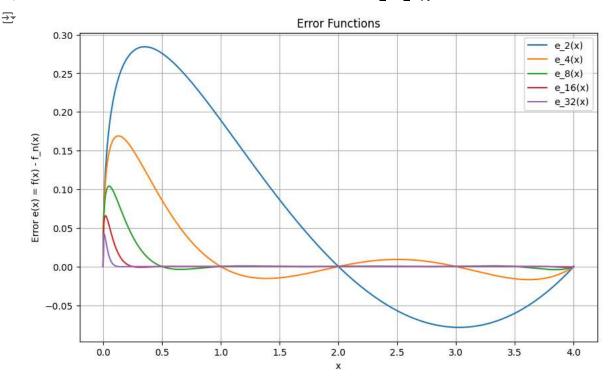


Function and Interpolations f(x) = sqrt(x)2.00 f 2(x) f 4(x) 1.75 f 8(x) f 16(x) f_32(x) 1.50 1.25 € 1.00 0.75 0.50 0.25 0.00 0.0 0.5 1.0 1.5 2.0 2.5 3.0 3.5 4.0

```
plt.figure(figsize=(10, 6))
for n, max_error in zip(n_values, max_errors):
    x_samples = np.linspace(0, 4, n+1)
    y_samples = f(x_samples)
    coef = newton_coefficients(x_samples, y_samples)
    y_interp = newton_poly(x_interval, x_samples, coef)

# Plot error
    error = y_actual - y_interp
    plt.plot(x_interval, error, label=f'e_{n}(x)')

plt.title("Error Functions")
plt.xlabel("x")
plt.ylabel("Error e(x) = f(x) - f_n(x)")
plt.legend()
plt.grid(True)
plt.show()
```



```
plt.figure(figsize=(10, 6))
plt.plot(n_values, max_errors, marker='o')
plt.title("Maximum Error vs n")
plt.xlabel("n")
plt.ylabel("Maximum Error E_n")
plt.grid(True)
plt.show()
```

