

# CECC421: Numerical Analysis

## Lecture 6: Function Interpolation

**Eng. Aya Kherbek**  
**Eng. Baher Kherbek**  
**Faculty of Engineering**  
**Department of Informatics**  
**Manara University**

## Purpose of the Session:

**Study of Interpolation Algorithms Using Numerical Methods  
and Converting Them into Python Code**

## NEWTON'S DIVIDED-DIFFERENCE INTERPOLATING

This method requires extracting the divided differences table and then using it to compute Newton's polynomial.

## In Python:

1. Use a **function** to find the forward differences table, returning the **coefficients of Newton's interpolation polynomial**.
2. Use a **function** to find Newton's interpolation for the value 1.5.
3. Apply the above to find Newton's interpolation for the following table.

x	0	1	2	3
y	0	1	8	27

4. **Plot** the resulting polynomial curve using `matplotlib.pyplot`, showing how it passes through the points in the table.

## To find the coefficients:

`a_list = [a0, a1, a2, a3]`

The values of `a[i]` must be computed in the columns shown in the table.

Each element of each column is calculated using the formula:

$$\frac{y[i+1]-y[i]}{x[i+1+k]-x[i]}$$

- We observe that the number of elements decreases from one column to the next.
- The calculations stop when the number of elements in a rank reaches one. That means we need an **outer loop** to represent the ranks and an **inner loop** for the values of `x` and `y`, which are found at the core of the function.

X	y	الفروق الأولى	الفروق الثانية	الفروق الثالثة
1.0	2.0			
1.3	3.1	$f[x_1, x_2] = \frac{3.1 - 2.0}{1.3 - 1.0}$		
1.6	3.9	$f[x_2, x_3] = \frac{3.9 - 3.1}{1.6 - 1.3}$	$f[x_1, x_2, x_3] = \frac{f[x_2, x_3] - f[x_1, x_2]}{1.6 - 1.0}$	
1.9	4.5	$f[x_3, x_4] = \frac{4.5 - 3.9}{1.9 - 1.6}$	$f[x_2, x_3, x_4] = \frac{f[x_3, x_4] - f[x_2, x_3]}{1.9 - 1.3}$	$f[x_1, x_2, x_3, x_4] = \frac{f[x_2, x_3, x_4] - f[x_1, x_2, x_3]}{1.9 - 1.0}$

```
x=[0,1,2,3]
y=[0,1,8,27]
def diff(x,y):
    a_list=[y[0]]
    for j in range(len(x)):
        Temp=[]
        for i in range(len(y)-1):
            temp=(y[i+1]-y[i])/(x[i+1+j]-x[i])
            Temp.append(temp)
        y=Temp
        a_list.append(y[0])
    if (len(y)==1):
        break
    return a_list
print(diff(x,y))
```

The solution code using the **diff** function to compute the differences table.

OUTPUT

[0, 1.0, 3.0, 1.0]

$$\begin{aligned} P_n(\text{value}) = & \\ & a[0] \\ & + \\ & a[1] * (\text{value} - x[0]) \\ & + \\ & a[2] * (\text{value} - x[0]) * (\text{value} - x[1]) \\ & + \\ & a[3] * (\text{value} - x[0]) * (\text{value} - x[1]) * (\text{value} - x[2]) \\ & + \\ & \cdot \\ & \cdot \\ & + \\ & a[n] * (\text{value} - x[0]) * (\text{value} - x[1]) * (\text{value} - \\ & x[2]) * \dots * (\text{value} - x[n-1]) \end{aligned}$$

**Note:** The new value is the result of multiplying the old value by the next value.

```
a_value=diff(x,y)
def newton(a_value,value):
    sum=a_value[0]
    temp=1
    for i in range(len(a_value)-1):
        temp=temp*(value-x[i])
        sum=sum+a_value[i+1]*temp
    return sum
print(newton(a_value,1.5))
```

OUTPUT

3.375

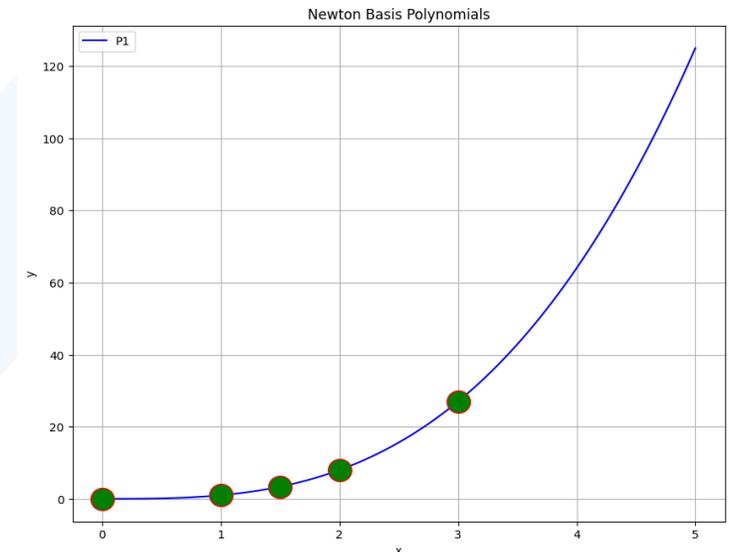
```
import numpy as np
import numpy.polynomial.polynomial as
poly
import matplotlib.pyplot as plt

x_new = np.linspace(0, 5, 100)

fig = plt.figure(figsize = (10,8))
plt.plot(x_new, newton(a_value,x_new),
'b', label = 'P1')

f=[0,1,1.5,2,3]
for i in range(len(f)):
    plt.plot(f[i], newton(a_value,f[i]),
marker="o", markersize=20,
markeredgecolor="red",
markerfacecolor="green")
```

```
plt.title('Newton Basis
Polynomials')
plt.xlabel('x')
plt.ylabel('y')
plt.grid()
plt.legend()
plt.show()
```



# LAGRANGE INTERPOLATING POLYNOMIAL

This method does not require extracting a table of divisors as in calculating Newton's polynomials.

### طريقة لاغرانج:

ليكن لدينا النقاط  $y_0 = f(x_0), y_1 = f(x_1), \dots, y_n = f(x_n)$   
فإن كثير حدود الاستيفاء لاغرانج يعطى بالعلاقة:

$$\begin{aligned} f_n(x) &= \sum_{i=0}^n L_i(x) f(x_i) = \\ &= L_0(x) f(x_0) + L_1(x) f(x_1) + \dots + L_n(x) f(x_n) \end{aligned}$$

$$L_i(x) = \prod_{\substack{j=0 \\ j \neq i}}^n \frac{x - x_j}{x_i - x_j}$$

حيث

1. Start
2. Read Data of x and y
3. Read value
4. Define function takes “value” as an argument
  - 4.1. Initialize list Y ,
  - 4.2. For i = 0 to n-1
    - 4.2.1. Initialize temp to 1
    - 4.2.2. For j = 0 to n-1
      - 4.2.2.1. If  $i \neq j$   
$$\text{temp} = \text{temp} * (x_p - X_j) / (X_i - X_j)$$
    - 4.2.3. Add temp to list Y
  - 4.3. For i = 0 to n-1  
$$Y = Y + \text{temp} * y_i$$
  - 4.4. Return Y
5. Print y
6. Stop

## Use Python:

1. Write a function that returns the Lagrange polynomial corresponding to the function  $y = f(x)$ .

2. Apply the above to find Newton's interpolation for the given table and for the value **1.5**.

x	0	1	2	3
f(x)	1	2	9	28

3. Plot the resulting polynomial curve using `matplotlib.pyplot`, showing how it passes through the points in the table.

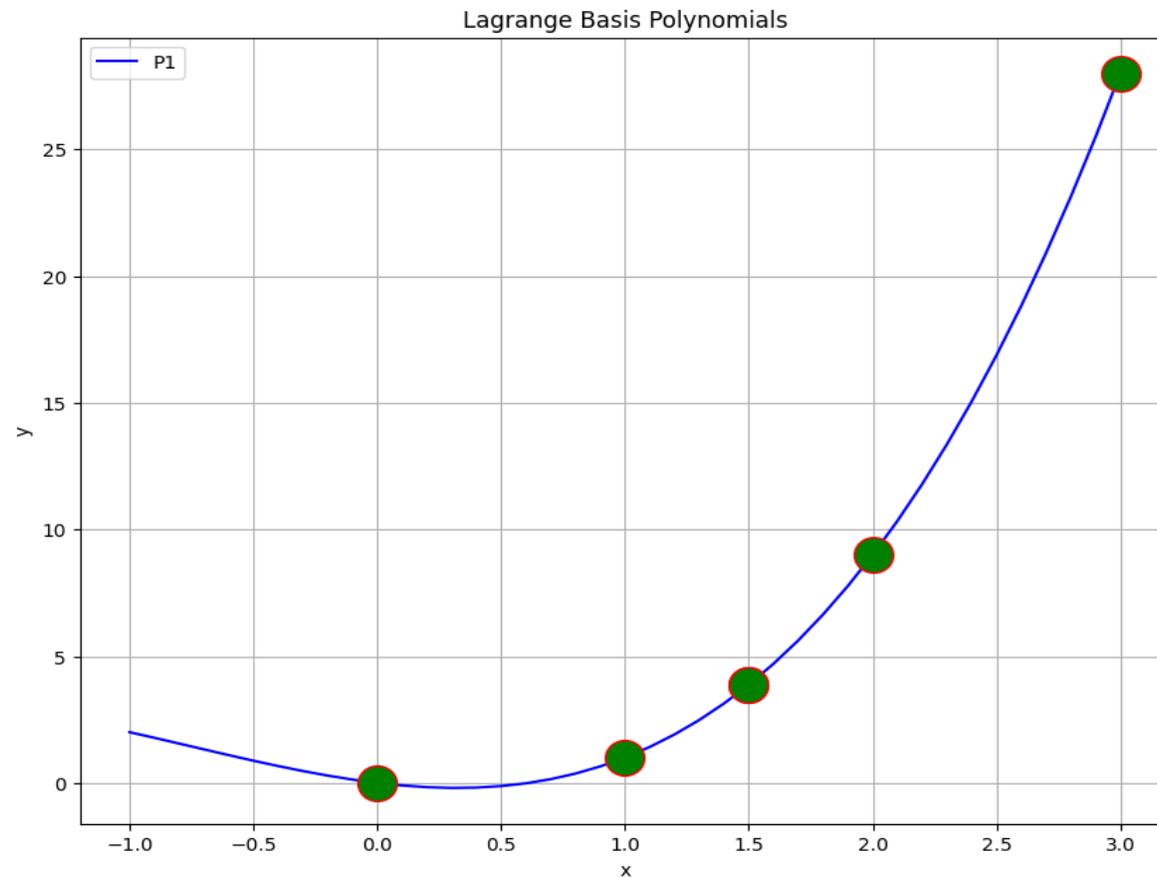
```
import numpy as np
import
numpy.polynomial.polynomi
al as poly
import matplotlib.pyplot
as plt

x=[0,1,2,3]
y=[0,1,9,28]
```

```
def Lagrange_Interpolating(value):
    L=[]
    for i in range(len(x)):
        temp=1
        for j in range(len(x)):
            if i!=j :
                temp=temp*(value-x[j])/(x[i]-x[j])
        L.append(temp)
    Y=0
    for i in range(len(y)):
        Y=Y+L[i]*y[i]
    return Y
print(Lagrange_Interpolating(1.5))
```

```
x_new = np.arange(-1.0, 3.1, 0.1)
fig = plt.figure(figsize = (10,8))
plt.plot(x_new, Lagrange_Interpolating(x_new), 'b', label = 'P1')
f=[0,1,1.5,2,3]
for i in range(len(f)):
    plt.plot(f[i], Lagrange_Interpolating(f[i]), marker="o", markersize=20,
markeredgecolor="red", markerfacecolor="green")
plt.plot(1.5, Lagrange_Interpolating(1.5), marker="o", markersize=20,
markeredgecolor="red", markerfacecolor="green")
plt.title('Lagrange Basis Polynomials')
plt.xlabel('x')
plt.ylabel('y')
plt.grid()
plt.legend()
plt.show()
```

output



**Thanks for Listening**