

DS623 PE06 – Taylor Polynomial Visualization


Topic: Differentiating Univariate Functions and Taylor Approximations

This script visualizes $f(x) = \sin(x) + \cos(x)$ and its Taylor approximations T_0, T_1, T_3, T_5 on $[-4, 4]$ using step size 0.001.

It optionally finds the x values that yield min/max of $f(x)$.

School: City University of Seattle

Student: Verónica Elze

 Mission Briefing: We are Robotic Unit EL-ZE. Our directive is to mimic the curve $f(x) = \sin(x) + \cos(x)$ using polynomial approximations! Engage mathematical processors and deploy the Taylor series up to orders 0, 1, 3, and 5.

Imports

```
In [1]: import numpy as np
import matplotlib.pyplot as plt
from math import factorial, sin, cos
import pandas as pd
```

Define $f(x) = \sin(x) + \cos(x)$

 Directive 1: Define the original signal pattern (function)

```
In [2]: def f(x):
return np.sin(x) + np.cos(x)
```

Define Taylor series approximation centered at $x_0 = 0$


 Directive 2: Deploy Taylor Module Generator

This bot builds a Taylor copycat of $f(x)$, using up to n terms, all starting from $x = 0$.

```
In [3]: def taylor_series(f, x, n):
taylor_approx = np.zeros_like(x)
for k in range(n + 1):
    if k % 4 == 0:
        coeff = 1
    elif k % 4 == 1:
        coeff = -1 if k % 2 == 1 else 1
    elif k % 4 == 2:
        coeff = -1
    else:
        coeff = 1 if k % 2 == 1 else -1

    taylor_approx += coeff * x**k / factorial(k)
return taylor_approx
```

Define x values

 Directive 3: Set scanning range for x-values (our domain of operation)

```
In [4]: x_vals = np.arange(-4, 4.001, 0.001)
f_vals = f(x_vals)
```

Compute Taylor approximations

 Directive 4: Compute Taylor approximations with increasing order of intelligence

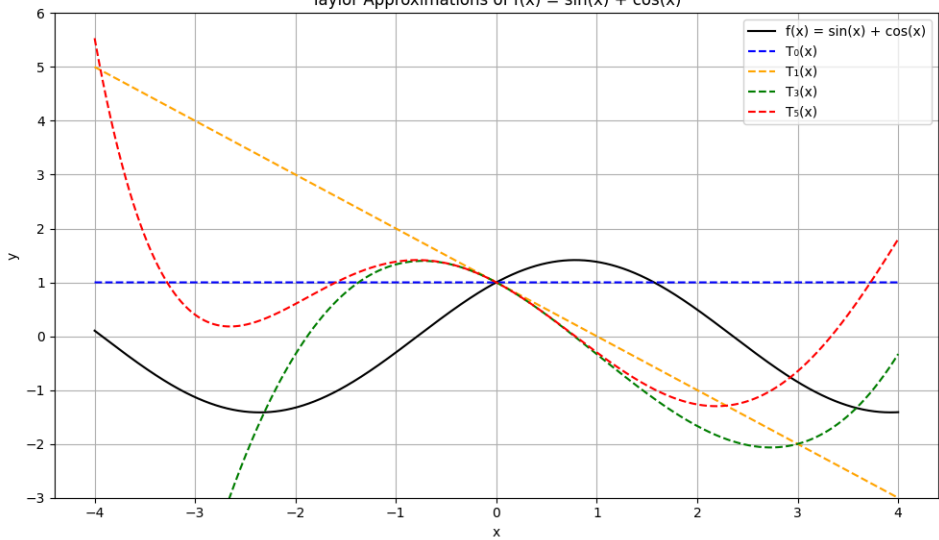
```
In [5]: t0 = taylor_series(f, x_vals, 0)
t1 = taylor_series(f, x_vals, 1)
t3 = taylor_series(f, x_vals, 3)
t5 = taylor_series(f, x_vals, 5)
```

Plotting

 Directive 5: Visualize approximations and compare to original pattern

```
In [6]: plt.figure(figsize=(10, 6))
plt.plot(x_vals, f_vals, label='f(x) = sin(x) + cos(x)', color='black')
plt.plot(x_vals, t0, '--', label='T0(x)', color='blue')
plt.plot(x_vals, t1, '--', label='T1(x)', color='orange')
plt.plot(x_vals, t3, '--', label='T3(x)', color='green')
plt.plot(x_vals, t5, '--', label='T5(x)', color='red')
plt.ylim([-3, 6])
plt.title("Taylor Approximations of f(x) = sin(x) + cos(x)")
plt.xlabel("x")
plt.ylabel("y")
plt.legend()
plt.grid(True)
plt.tight_layout()
plt.show()
```

Taylor Approximations of $f(x) = \sin(x) + \cos(x)$



Optional: Find max and min points

📡 Directive 6: Analyze terrain highs and lows in our operation range

```
In [7]: x_max = x_vals[np.argmax(f_vals)]
        x_min = x_vals[np.argmin(f_vals)]
```

Display extrema

📡 Report findings to mission control

```
In [8]: extrema_df = pd.DataFrame({
        "Label": ["x_max", "x_min"],
        "x_value": [x_max, x_min]
    })
```

```
In [9]: print("📡 Extrema Analysis Report:\n")
        print(extrema_df)
```

📡 Extrema Analysis Report:

```
Label  x_value
0  x_max    0.785
1  x_min    3.927
```

OpenAI. (2025). ChatGPT's assistance with Taylor series implementation [Large language model]. <https://openai.com/chatgpt>