EE2703 : Applied Programming Lab Assignment 4

Potta Muni Asheesh EE19B048

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1 Aim

We will be fitting two functions, namely, exp(x) and cos(cos(x)) over an interval of $[0, 2\pi)$ using Fourier series.

$$a_0 + \sum_{n=1}^{\infty} (a_n cos(nx) + b_n sin(nx))$$

The coefficients a_0, a_n, b_n are given by

$$a_0 = \frac{1}{2\pi} \int_0^{2\pi} f(x)dx$$

$$a_n = \frac{1}{\pi} \int_0^{2\pi} f(x)cos(nx)dx$$

$$b_n = \frac{1}{\pi} \int_0^{2\pi} f(x)sin(nx)dx$$

2 Assignment questions

2.1 Plotting the two functions

We can define functions which use functions from numpy to compute the required function for a given vector(or scalar) and returns a vector(or scalar). The functions are named as e and coscos in the code.

```
def coscos(x):
    return np.cos(np.cos(x))
```

The plot of these two functions over the interval $[-2\pi, 4\pi)$ is given below. It is seen that cos(cos(x)) is periodic and exp(x) is not periodic. When we compute the Fourier series of these two functions, the generated function is expected to be over the interval $[0, 2\pi)$, as shown below

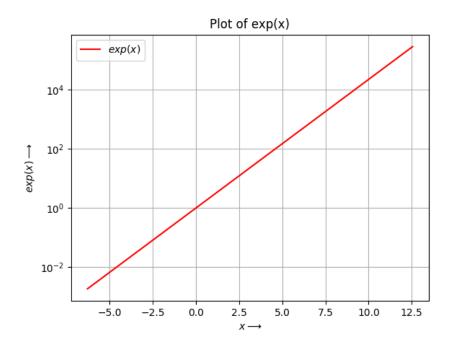


Figure 1: Plot of $\exp(x)$ (semi-log scale)

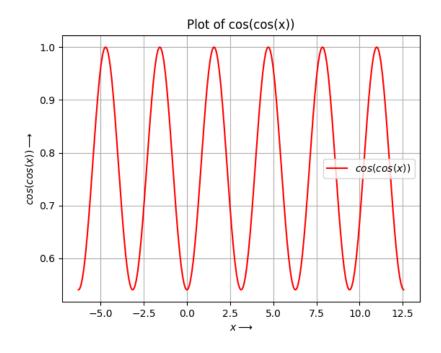


Figure 2: Plot of cos(cos(x))

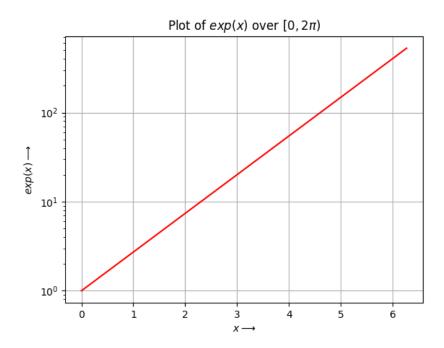


Figure 3: Plot of exp(x)

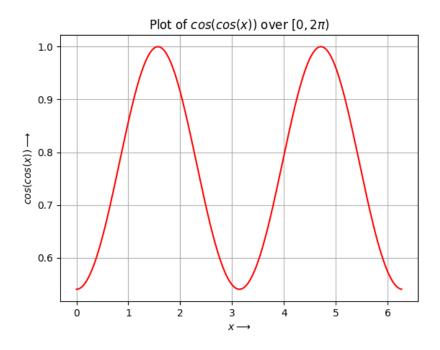


Figure 4: Plot of cos(cos(x))

2.2 Obtaining Fourier coefficients using direct integration

As discussed above, we can use integration to find the coefficients of the Fourier series. In Python, we can do definite integration using quad function from scipy.integrate. Two Python functions $u(x, k, f) = f(x)\cos(kx)$ and $v(x, k, f) = f(x)\sin(kx)$ are defined, to pass to quad function, as shown below

```
def u(x, k, f):
    return f(x)*np.cos(k*x)

def v(x, k, f):
    return f(x)*np.sin(k*x)
```

Here, the parameter f should be a function and it should return a scalar or vector. The functions defined previously are used here. for loop is used to find the coefficients iteratively as shown below. The coefficients for exp(x) and cos(cos(x)) are stored in arrays named coeffs_f1 and coeffs_f2 respectively.

```
coeffs_f1 = np.zeros(51)
coeffs_f2 = np.zeros(51)
coeffs_f1[0] = quad(u, 0, 2*pi, args=(0,e))[0]/(2*pi)
coeffs_f2[0] = quad(u, 0, 2*pi, args=(0,coscos))[0]/(2*pi)

for i in range(1, 26):
    coeffs_f1[2*i-1] = quad(u, 0, 2*pi, args=(i,e))[0]/pi
    coeffs_f1[2*i] = quad(v, 0, 2*pi, args=(i,e))[0]/pi
    coeffs_f2[2*i-1] = quad(u, 0, 2*pi, args=(i,coscos))[0]/pi
    coeffs_f2[2*i] = quad(v, 0, 2*pi, args=(i,coscos))[0]/pi
```

2.3 Plotting the coefficients

To plot the 51 coefficients, they are first arranged in a vector as shown below.

$$\begin{pmatrix} a_0 \\ a_1 \\ b_1 \\ \dots \\ a_{25} \\ b_{25} \end{pmatrix}$$

Now, this vector is plotted against n as n varies from 1 to 51. For each set of the coefficients, two plots are made, one with log scale on y-axis only and the other with log scale on both x and y axes. The plots of absolute values of the coefficients are shown below.

It is found that b_n for the function cos(cos(x)) are of the order $10^{-15} - 10^{-17}$ which for practical purposes can be approximated to 0. This happens because cos(cos(x)) is an even function and coefficients of sin terms in Fourier series are 0 for even functions.

The coefficients for exp(x) are given as

$$a_n = \frac{e^{2\pi} - 1}{\pi(1 + n^2)}$$
$$b_n = \frac{-n(e^{2\pi} - 1)}{\pi(1 + n^2)}$$

So, they decay as $\sim \frac{1}{n^2}$ and $\sim \frac{1}{n}$ which is not as quick as the coefficients of $\cos(\cos(x))$. This is because, as $\exp(x)$ is not periodic, higher frequency sinusoids are also required for good enough convergence. They look linear in the $\log\log$ plot because they decay as shown above. The coefficients of $\cos(\cos(x))$ are related to Bessel functions as they can be evaluated from the integral form of Bessels functions, so they look linear in semi-log-y plot.

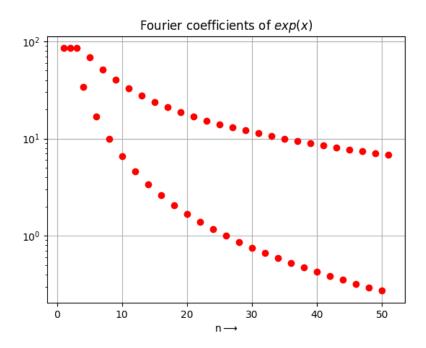


Figure 5: Coefficients for exp(x) (semilogy)

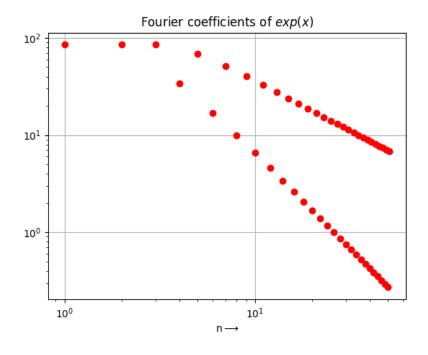


Figure 6: Coefficients for $\exp(x)$ (loglog)

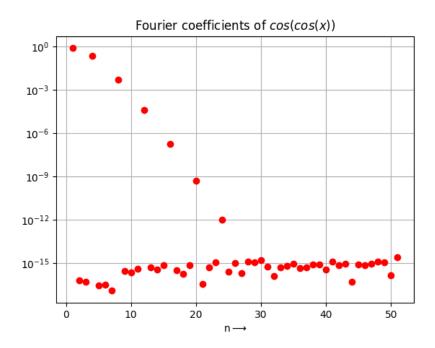


Figure 7: Coefficients for cos(cos(x)) (semilogy)

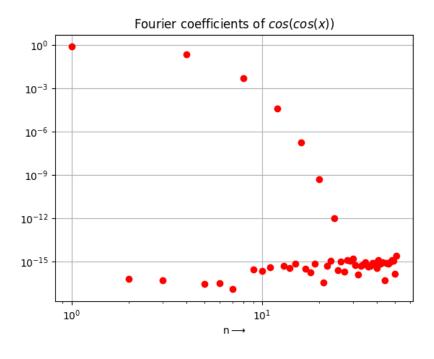


Figure 8: Coefficients for cos(cos(x)) (loglog)

2.4 Least squares approach

A finite set of Fourier coeffcients can be found using the **Least squares** approximation by looking at this as a matrix problem. 400 distinct x values over the interval $[0, 2\pi)$ are considered in this problem. The number of x values can be increased to get better approximations. The resultant matrix equation is

$$\begin{pmatrix} 1 & cos(x_1) & sin(x_1) & \dots & cos(25x_1) & sin(25x_1) \\ \dots & \dots & \dots & \dots & \dots \\ 1 & cos(x_{400}) & sin(x_{400}) & \dots & cos(25x_{400}) & sin(25x_{400}) \end{pmatrix} \begin{pmatrix} a_0 \\ a_1 \\ b_1 \\ \dots \\ a_{25} \\ b_{25} \end{pmatrix} = \begin{pmatrix} f(x_1) \\ \dots \\ f(x_{400}) \end{pmatrix}$$

Consider the matrix to be A, the coefficients vector be c and the function values vector be b. Then the matrix equation is

$$Ac = b$$

We can use lstsq function from scipy.linalg to approximate the vector c. The coefficients for exp(x) and cos(cos(x)) can be approximated as shown below

```
x = np.linspace(0, 2*pi, 401)
x = x[:-1]
b1 = e(x)
b2 = coscos(x)
A = np.zeros((400, 51))
A[:,0] = 1
for k in range(1, 26):
    A[:,2*k-1] = np.cos(k*x)
    A[:,2*k] = np.sin(k*x)
c1 = lstsq(A, b1)[0]
c2 = lstsq(A, b2)[0]
```

2.5 Plotting the approximated coefficients

The best fit coefficients for both the functions exp(x) and cos(cos(s)) are computed above. They can be plotted along the coefficients compute by direct integration. To differentiate them, green circles are used.

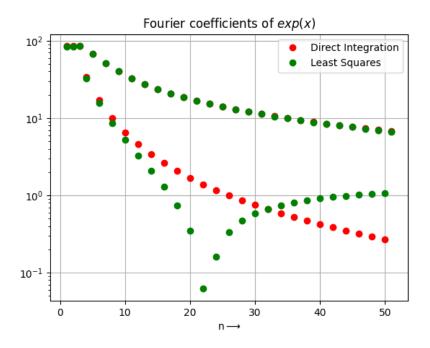


Figure 9: Coefficients for $\exp(x)$ (semilogy)

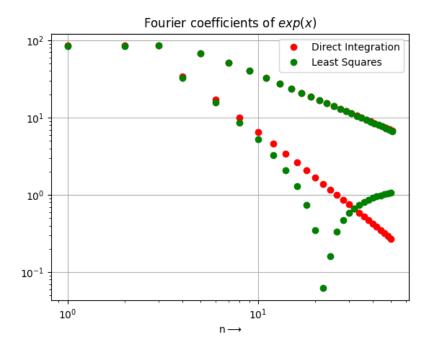


Figure 10: Coefficients for exp(x) (loglog)

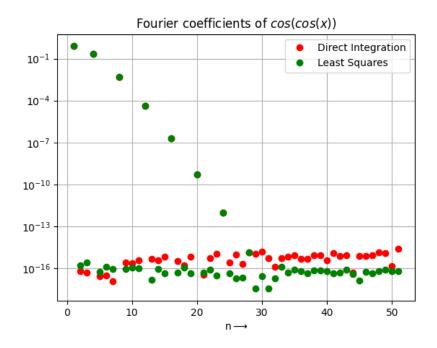


Figure 11: Coefficients for cos(cos(x)) (semilogy)

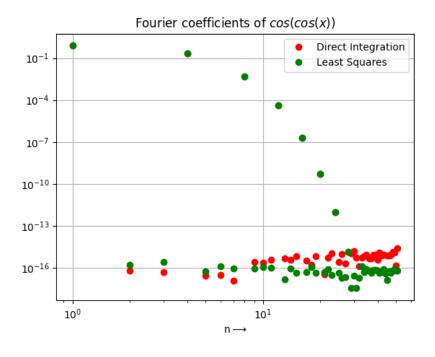


Figure 12: Coefficients for cos(cos(x)) (loglog)

2.6 Comparing the coefficients

The results from direct integration and least squares approximation agree for cos(cos(x)), but there is significant deviation in coefficients of cos terms for exp(x). To find the maximum deviation from actual values, np.max can be used.

```
dev_f1 = np.max(np.abs(c1 - coeffs_f1))
dev_f2 = np.max(np.abs(c2 - coeffs_f2))
```

Here, c1, c2 are the approximated coefficients and coeffs_f1, coeffs_f2 are the coefficients from direct integration.

- The maximum absolute deviation for exp(x) is 1.3327308703353964
- The maximum absolute deviation for cos(cos(x)) is 2.6566469738662125e-15

2.7 Plotting the approximated functions

The function values for the approxiamted coefficients can be found from the multiplication of matrix A and vector c. This will the function values for x over the interval $[0, 2\pi)$. For matrix multiplication, np.dot function can be used. The plots of the functions are given below.

High deviation is seen in case of exp(x), but almost no deviation is seen for cos(cos(x)). This happens because 2π -periodic extension of exp(x) has a finite discontinuity and the Fourier series deviates from the actual function at the point of discontinuity. This is called *Gibbs phenomenon*. This deviation seen in the plot has very little to do with error in least squares approximation as similar deviation can be seen for direct integration coefficients as well. This does not happen for cos(cos(s)) as there is no discontinuity.

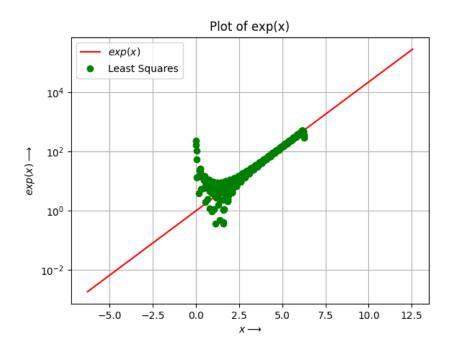


Figure 13: Approximated exp(x)

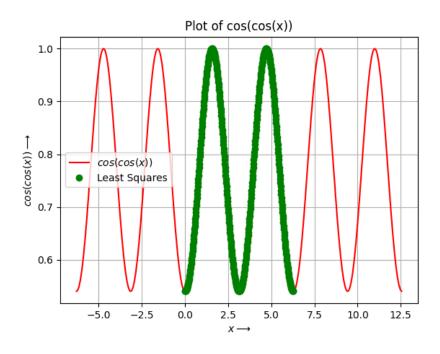


Figure 14: Approximated cos(cos(x))

3 Conclusion

51 Fourier coefficients of exp(x) and cos(cos(x)) are computed by integration and their variation with n is observed. It was seen that even functions have sine-coefficients as 0 in Fourier series. The coefficients are also approximated using least squares method and the deviation from integration coefficients was observed. The peculiar deviation of Fourier series from actual function after jump discontinuity, called Gibbs phenomenon, was observed.