# EE2703 Applied Programming Lab - Assignment No 4

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## 1 Abstract

The goal of this assignment is the following.

- To fit two functions  $e^x$  and cos(cos(x)) using the Fourier series.
- To use least squares fitting to simplify the process of calculating Fourier series.
- Studying the effect of discontinuities on the Fourier Series, and the Gibbs Phenomenon.
- To plot graphs to understand the above

# 2 Assignment

### 2.1 Part 1

Importing the standard libraries

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

Defining the functions e<sup>x</sup> and cos(cos(x))

def exp(x):
    return np.exp(x)

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

We define a plotting function to help simplify the code.

```
def plotter(fig_no,plot_x,plot_y,label_x,label_y,type=None,kind='b-',title=""):
    plt.figure(fig_no)
    plt.grid(True)
    if type =="semilogy":
        plt.semilogy(plot_x,plot_y,kind)
    elif type =='ll':
        plt.loglog(plot_x,plot_y,kind)
    elif type ==None:
        plt.plot(plot_x,plot_y,kind)
    plt.xlabel(label_x,size =19)
    plt.ylabel(label_y,size =19)
    plt.title(title)
We evaluate the functions from -2\pi to 4\pi to plot, and we also plot a periodic
extension of the functions.
t = np.linspace(-2*np.pi, 4*np.pi, 1200)
fr_length = np.arange(1,52)#Fourier Coefficient Number
plotter(1,t,exp(t),r"$t$",r"exp(t)",
"semilogy", title = "Exponential Function on a semilog plot")
plotter(2, t, coscos(t)), r"$t$", r"cos(cos(t))",
title="Cos(Cos(t)) Function on a linear plot")
plotter(1, t,
np.concatenate((exp(t) [400:800], exp(t) [400:800], exp(t) [400:800])),
r"$t$", r"exp(t)", "semilogy", 'r-')
plt.legend(("true", "periodic extension"))
plotter(2, t,
np.concatenate((coscos(t)[400:800], coscos(t)[400:800], coscos(t)[400:800])),
r"$t$", r"exp(t)", "semilogy", 'r-')
plt.legend(("true", "periodic extension"))
2.2
     Part 2
We use the integrator in scipy to calculate the fourier integrals in a for loop.
And then we plot them.
def u1(x,k):
    return(coscos(x)*np.cos(k*x))
def v1(x,k):
    return(coscos(x)*np.sin(k*x))
def u2(x,k):
    return(exp(x)*np.cos(k*x))
```

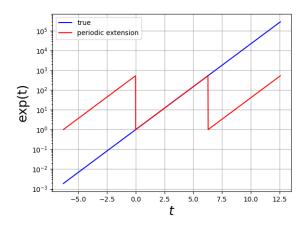


Figure 1:  $e^t$  vs t on a linear plot

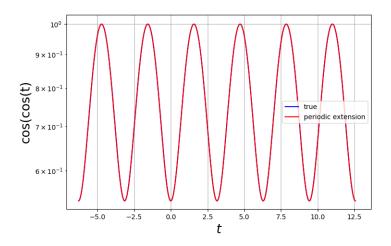


Figure 2: cos(cos(t)) vs t on a linear plot

```
def v2(x,k):
    return(exp(x)*np.sin(k*x))

def integrate():
    a = np.zeros(51)
    b = np.zeros(51)
    a[0] = scipy.integrate.quad(exp,0,2*np.pi)[0]/(2*np.pi)
    b[0] = scipy.integrate.quad(coscos,0,2*np.pi)[0]/(2*np.pi)
    for i in range(1,51,2):
```

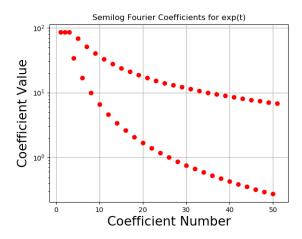
```
 a[i] = scipy.integrate.quad(u2,0,2*np.pi,args=(i//2+1))[0]/(np.pi) \\ b[i] = scipy.integrate.quad(u1,0,2*np.pi,args=(i//2+1))[0]/(np.pi) \\ a[i+1] = scipy.integrate.quad(v2,0,2*np.pi,args=(i//2+1))[0]/(np.pi) \\ b[i+1] = scipy.integrate.quad(v1,0,2*np.pi,args=(i//2+1))[0]/(np.pi) \\ return a,b
```

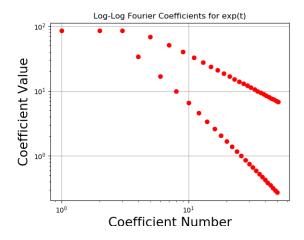
frexp,frcos = integrate()

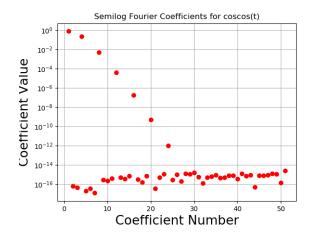
#### 2.3 Part 3

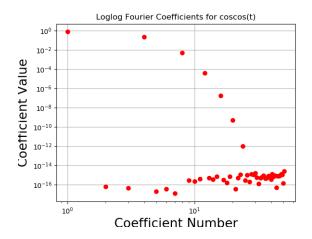
And now we plot the Fourier series terms.

```
plotter(3,fr_length,
np.absolute(frexp), "Coefficient Number",
"Coefficient Value", "semilogy", 'ro',
title="Semilog Fourier Coefficients for exp(t)")
plotter(4,fr_length,
np.absolute(frexp), "Coefficient Number",
"Coefficient Value", "ll", 'ro',
title="Log-Log Fourier Coefficients for exp(t)")
plotter(5,fr_length,
np.absolute(frcos), "Coefficient Number",
"Coefficient Value", "semilogy",
'ro',title="Semilog Fourier Coefficients for coscos(t)")
plotter(6,fr_length,
np.absolute(frcos), "Coefficient Number",
"Coefficient Value", "ll", 'ro',
title="Loglog Fourier Coefficients for coscos(t)")
plt.show()
```









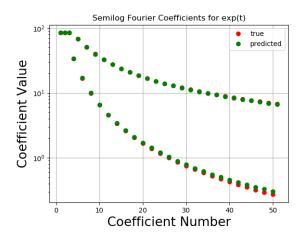
a. The  $b_n$  coefficients are nearly zero for cos(cos(t)) since it is an even function, and hence does not have an odd component.

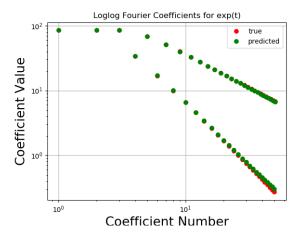
- b. The magnitude of the coefficients would represent how much of certain frequencies happen to be in the output. cos(cos(t)) does not have very many frequencies of harmonics, so it dies out quickly. However, since the periodic extension of  $e^t$  is discontinuous. To represent this discontinuity as a sum of continuous sinusoids, we would need high frequency components, hence coefficients do not decay as quickly.
- c. The loglog plot is linear for  $e^t$  since Fourier coefficients of  $e^t$  decay with 1/n or  $1/n^2$ . The semilog plot seems linear in the cos(cos(t)) case as its fourier coefficients decay exponentially with n.

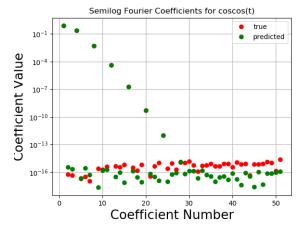
#### 2.4 Parts 4 and 5

We now use a least squares approach to calculate the Fourier series.

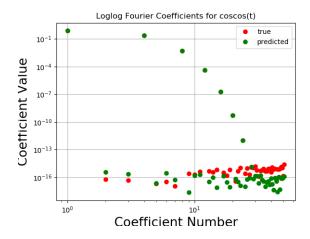
```
x =np.linspace(0,2*np.pi,400,endpoint =True)
bexp = exp(x)
bcoscos =coscos(x)
A = np.zeros((400,51))
A[:,0] = 1
for k in range (1,26):
    A[:,2*k-1] = np.cos(k*x)
    A[:,2*k] = np.sin(k*x)
cexp = np.linalg.lstsq(A,bexp)[0]
ccoscos = np.linalg.lstsq(A,bcoscos)[0]
cexp = np.linalg.lstsq(A,bexp)[0]
ccoscos = np.linalg.lstsq(A,bcoscos)[0]
plotter(3,fr_length,np.abs(cexp),"Coefficient Number",
"Coefficient Value", "semilogy", 'go',
title="Semilog Fourier Coefficients for exp(t)")
plt.legend(("true", "predicted"))
plotter(4,fr_length,np.abs(cexp),"Coefficient Number",
"Coefficient Value", "ll", 'go',
title="Loglog Fourier Coefficients for exp(t)")
plt.legend(("true", "predicted"))
plotter(5,fr_length,np.abs(ccoscos), "Coefficient Number",
"Coefficient Value", "semilogy", 'go',
title="Semilog Fourier Coefficients for coscos(t)")
plt.legend(("true", "predicted"))
plotter(6,fr_length,np.abs(ccoscos), "Coefficient Number",
"Coefficient Value", "ll", 'go',
title="Loglog Fourier Coefficients for coscos(t)")
plt.legend(("true", "predicted"))
```







In the case of the  $e^t$ , if we did not include  $2\pi$  as one of the times to be included in the A matrix, there would be quite a big discrepancy between



the fourier transforms calculated in the two methods. The reason for this is that at a discontinuity, the value of the function should be approximated as the average of the two endpoints, but if we did not include the endpoint at  $2\pi$ , the least square fitting would not be aware of it, leading to the discrepancy. On the other hand, this doesn't matter at all for  $\cos(\cos(t))$  since it is continuous, and the final point contributes very little to the final fourier coefficients. Hence we include the value of  $t = 2\pi$  in the matrix.

# 2.5 Part 6

We can find the absolute difference between the two answers by subtracting the two vectors in vector form, take the absolute value and finding the maximum.

```
diffexp = np.absolute(cexp-frexp)
diffcos = np.absolute(ccoscos-frcos)
print(np.amax(diffexp),np.amax(diffcos))
```

 $0.0881216977876722\ 2.5466388061776397e-15$ 

There is very good agreement in values in the case of cos(cos(x)) but a significant amount of difference in the case of  $e^t$ . The reason for this is that the periodic extension of the exponential function is discontinuous, and hence would require a lot more samples to accurately determine its Fourier coefficients. If we increased the number of samples to  $10^6$ , the maximum deviation would reduce, but not vanish. The effect of this lack of samples is felt more near the discontinuity of the signal, which can be attributed to the Gibbs Phenomenon.

#### 2.6 Part 7

The function is evaluated using the fourier coefficients calculated via the least squares method.

```
Acexp = A@cexp
Accos = A@ccoscos
plotter(1,t,np.concatenate((np.zeros(400),Acexp,np.zeros(400))),
r"$t$",r"exp(t)","semilogy",'go')
plt.legend(("true","periodic extension","predicted"))
plotter(2,t,np.concatenate((np.zeros(400),Accos,np.zeros(400))),
r"$t$",r"coscos(t)",None,'go')
plt.legend(("true","periodic extension","predicted"))
```

The cos(cos(t)) vs t graph, agrees almost perfectly, beyond the scope of the precision of the least squares fitter. The Fourier approximation of  $e^t$  does not agree very well to the ideal case near the discontinuity. The cause for this is the Gibbs Phenomenon, which can be described as below. The partial sums of the Fourier series will have large oscillations near the discontinuity of the function. These oscillations do not die out as n increases, but approaches a finite limit. This is one of the causes of ringing artifacts in signal processing, and is very undesirable. Plotting the output on a linear graph would make this ringing much more apparent.

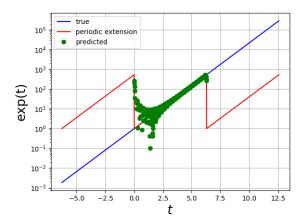


Figure 3:  $e^t$  vs t on a linear plot

# 3 Conclusions

• We saw two different ways to calculate the Fourier series of a periodic signal.

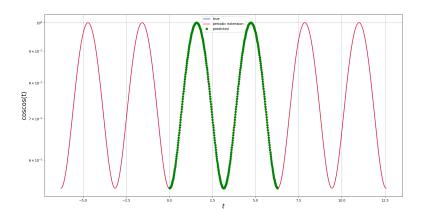


Figure 4: cos(cos(t)) vs t on a linear plot

- We saw how least squares fitting can be used to simplify the process of calculating the Fourier Series.
- ullet We observed Gibbs phenomenon at the discontinuity in the Fourier approximation of  $e^t$ .