

EE - 2703 Applied Programming Lab Assignment -3

Fourier Approximations

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1 Introduction

In this assignment is about approximating a function using Fourier series. Fourier Series is decomposition of a periodic function into sinusoids , cos and sin multiplied by fourier coefficients. These coefficients determine the component of that particular frequency in the function.

$$f(x) = \frac{a_o}{2} + \sum_{n=1}^{\infty} a_n \cos(nx) + b_n \sin(nx)$$

where a_n and b_n ,

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

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

for obtaining a_o , substitute $n = 0$ for a .

2 Assignment Problems

In this assignment e^x and $\cos(\cos(x))$ are the two functions that are taken and analyzed. First part is to evaluate first 50 coefficient using integral from definition. Second part of the problem is to do it again using least square method to solve matrix equation. Third part is the error comparison between both the methods. Finally to approximate both the functions from their Fourier series using coefficients derived from integral.

2.1 Function Definition

First part consist of defining the python functions to return e^x and $\cos(\cos(x))$. And then creating a vector form $[-2/\pi, 4/\pi]$ with 600 points. And the obtaining functional values at these points for both of these functions.

Listing 1: Defining Functions

```
def exp(x):
    return np.e**x

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

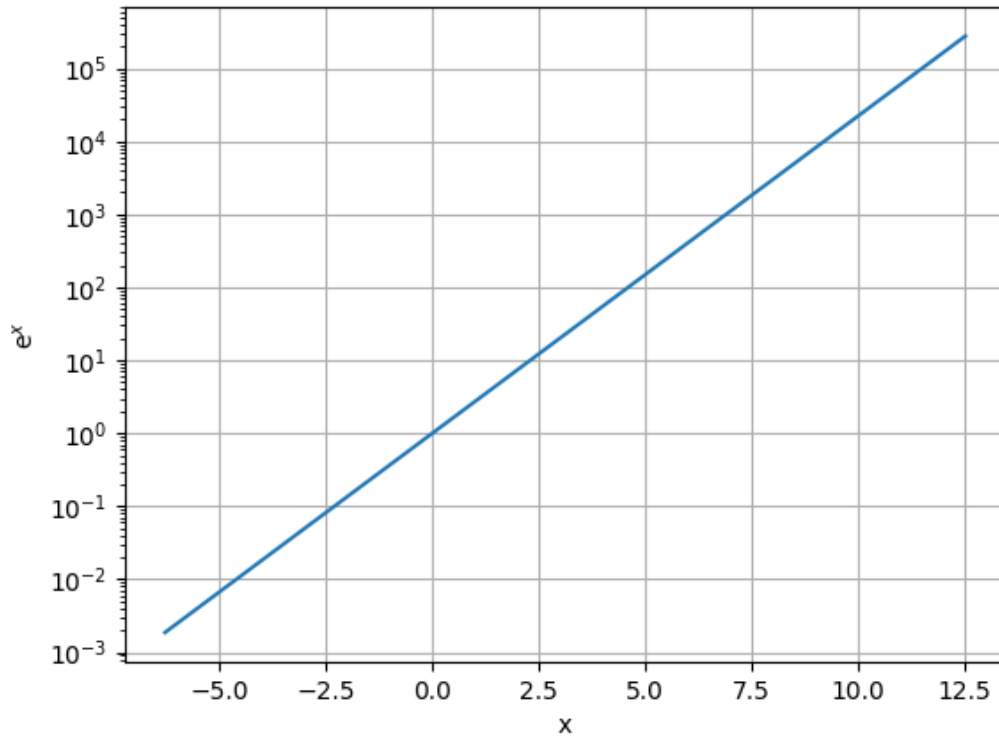


Figure 1: Semi-log scale plot of e^x

```
x = np.arange(-2*np.pi, 4*np.pi, np.pi/100)

exp_arr = exp(x)
cos_arr = coscos(x)
```

2.2 Plotting $\exp(x)$ and $\cos(\cos(x))$

In the above fragment `exp_arr` and `cos_arr` have the functional values and `x` has the points. Plotting both the values of e^x and $\cos(\cos(x))$ are plotted in semi-log and linear scale respectively. Look at Fig 1 and Fig 2.

Listing 2: Plotting the functions

```
#plotting exp(x) between 2pi to 4pi
plt.semilogy(x, exp_arr)
plt.xlabel("x")
plt.ylabel("e$^{x}$")
plt.grid(True)
plt.show()

#plotting cos(cos(x)) between 2pi tp 4pi
plt.scatter(x, cos_arr)
plt.xlabel("x")
plt.ylabel("cos(cos(x))")
plt.grid(True)
plt.show()
```

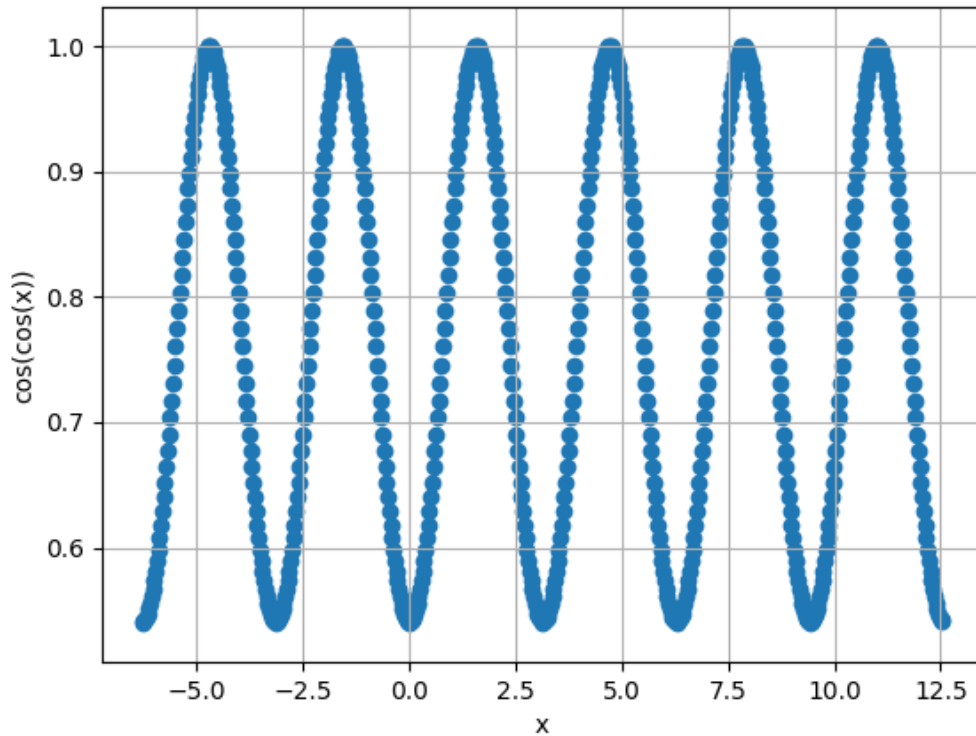


Figure 2: Plot of $\cos(\cos(x))$ in linear scale

2.3 Expected Fourier Approximation

Expected output for $\cos(\cos(x))$ would be the function itself as it is periodic and has no discontinuity. Whereas in case of e^x which is non-periodic the output expected would be a periodic part of function from 0 to 2π repeating in the domain here in this case 2π to 4π it will repeat thrice. See Figure 3 and Figure 4.

```
#expected output by fourier series for coscos(x)
plt.plot(x,cos_arr)
plt.title("Expected fourier approximation for cos(cos(x)) ")
plt.show()
#expected output by fourier series for exp(x)
x3 = np.arange(0,2*np.pi,np.pi/100)
expec_ex = np.array(list(exp(x3))+list(exp(x3))+list(exp(x3)))
plt.plot(x,expec_ex)
plt.title("Expected fourier approximation for e^{x}")
plt.show()
```

2.4 Obtaining Fourier Coefficients

First part of this subproblem involves defining the function inside the integral to calculate a_s and b_s . And then defining zero vector of size 26 each for a_n and b_n respectively. First element of b will be zero i.e. their index in the vector is the subscript value. Giving in total of fiftyone Fourier Coefficients including a_0 .

```
#defining a function to multiply cos and sine to parameter function. To be used to calculate
def four_a(x,f,n):
    return f(x)*np.cos(n*x)
def four_b(x,f,n):
```

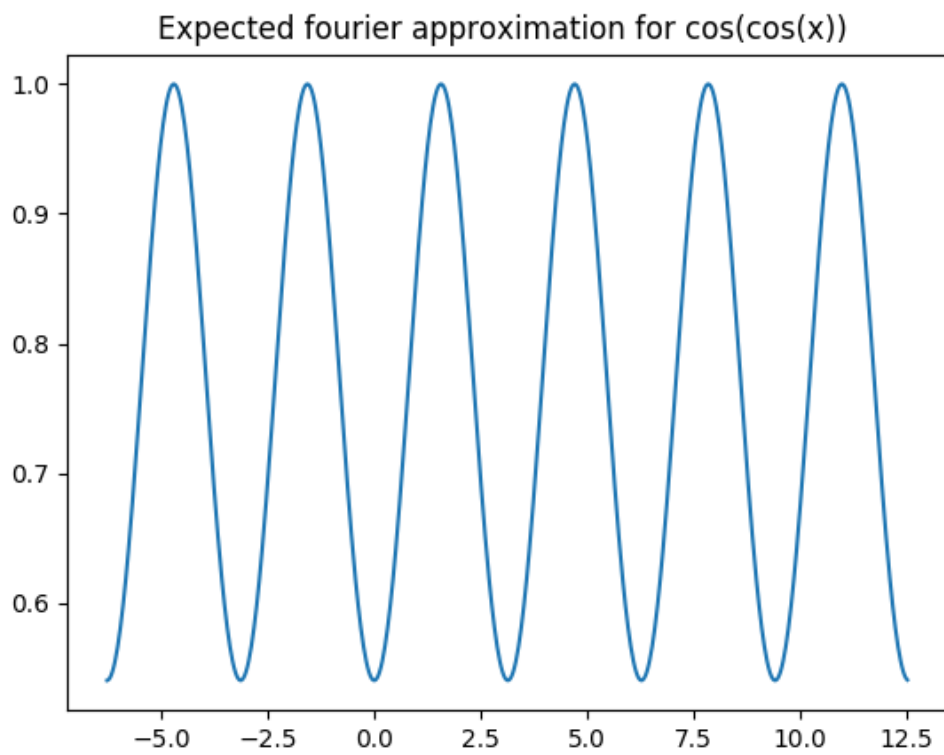


Figure 3: Expected Fourier Approximation of $\cos(\cos(x))$

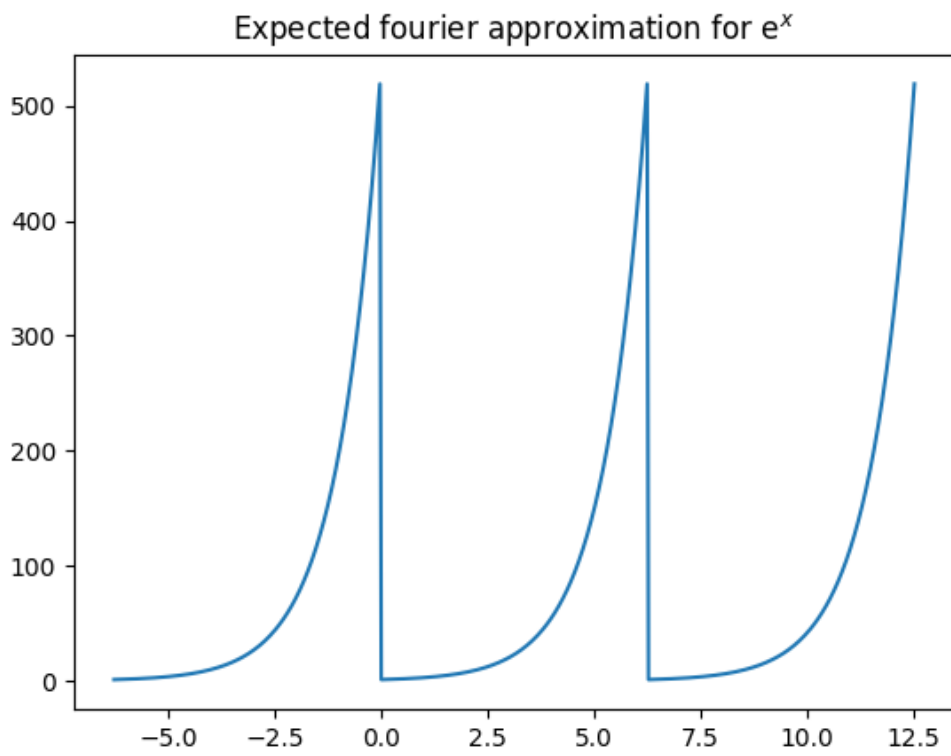


Figure 4: Expected Fourier approximation $\exp(x)$

```

        return f(x)*np.sin(n*x)

#creating zero vectors to store fourier coefficients
#An : for a's
#Bn : for b's
An=np.zeros(26)
Bn=np.zeros(26)

```

2.5 Function to compute Fourier Coefficients

The defined function computes the integral of the defined functions in last code fragment from 0 to π to compute the a^n 's and b^n 's.

#Defied function calculates and returns fourier coefficients (a's and b's) of the para

```

def coeff(f):
    for i in range(26):
        if (i==0):
            An[i] = (quad(four_a,0,2*np.pi, args=(f,i,))[0]/(2*np.pi))#com
        else:
            An[i] = (quad(four_a,0,2*np.pi, args=(f,i,))[0]/(np.pi))#for tl
            Bn[i] = (quad(four_b,0,2*np.pi, args=(f,i,))[0]/(np.pi))
    return An,Bn

```

2.6 Plotting Fourier coefficients of cos(cos(x))

The function defined in last code fragment accepts function and returns a^n and b^n in separate arrays. Passing $\cos(\cos(x))$ to the function plotting the obtained coefficients against n in log-log and semi-log scale. See Figure 5 and Figure 6

```

a_cosc, b_cosc = coeff(coscos)

#plotting cos(cos(x)) coefficients on log-log scale
n = np.arange(0,26)
plt.loglog(n,abs(a_cosc),"r+")
plt.loglog(n[1:],abs(b_cosc[1:]),"r+")
plt.legend("a$_{n}$","b$_{n}$")
plt.grid(True)
plt.title("Log-Log plot of cos(cos(x)) coefficients")
plt.show()

#plotting cos(cos(x)) coefficients on semi-log scale
plt.semilogy(n,abs(a_cosc),"r+")
plt.semilogy(n[1:],abs(b_cosc[1:]),"r+")
plt.legend("a$_{n}$","b$_{n}$")
plt.grid(True)
plt.title("Semi-Log plot of cos(cos(x)) coefficients")
plt.show()

```

2.7 Plotting Fourier coefficients of exp(x)

.Similar to $\cos(\cos(x))$ passing e^x to the function which returns two arrays which have the values of a_0 to a_{25} and b_1 to b_{25} respectively. Plotting the obtained coefficients against n in log-log and semi-log scale. See Figure 7 and Figure 8

```

a_exp, b_exp = coeff(exp)

#plotting Exp(x) coefficients on log-log scale
plt.loglog(n,abs(a_exp),"r+")

```

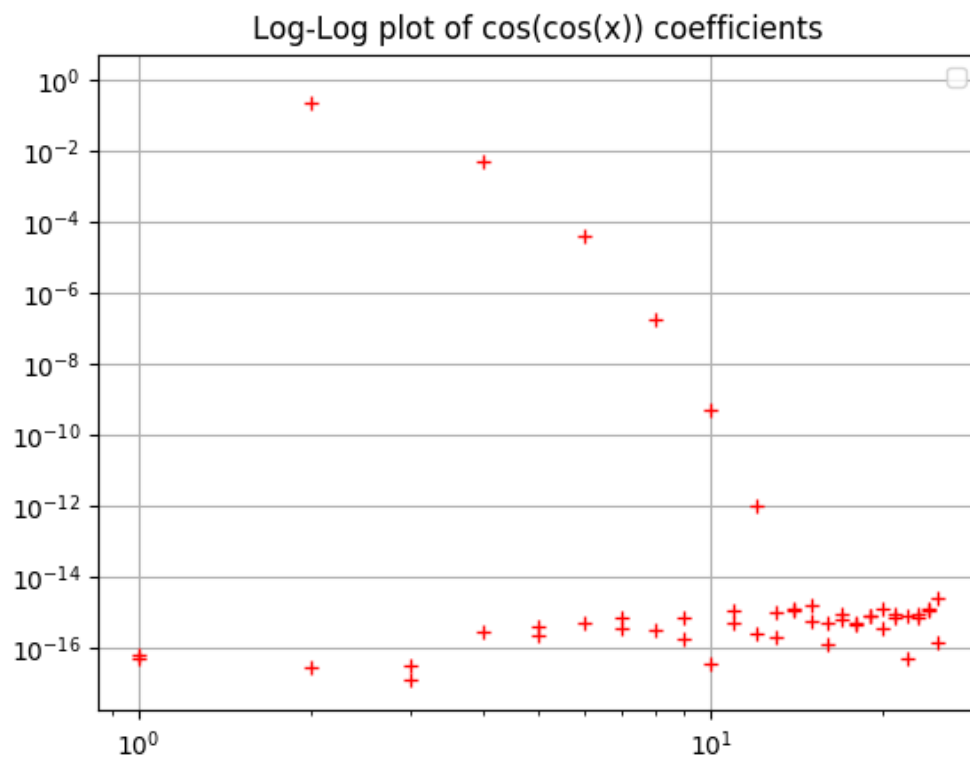


Figure 5: log-log plot of $\cos(\cos(x))$

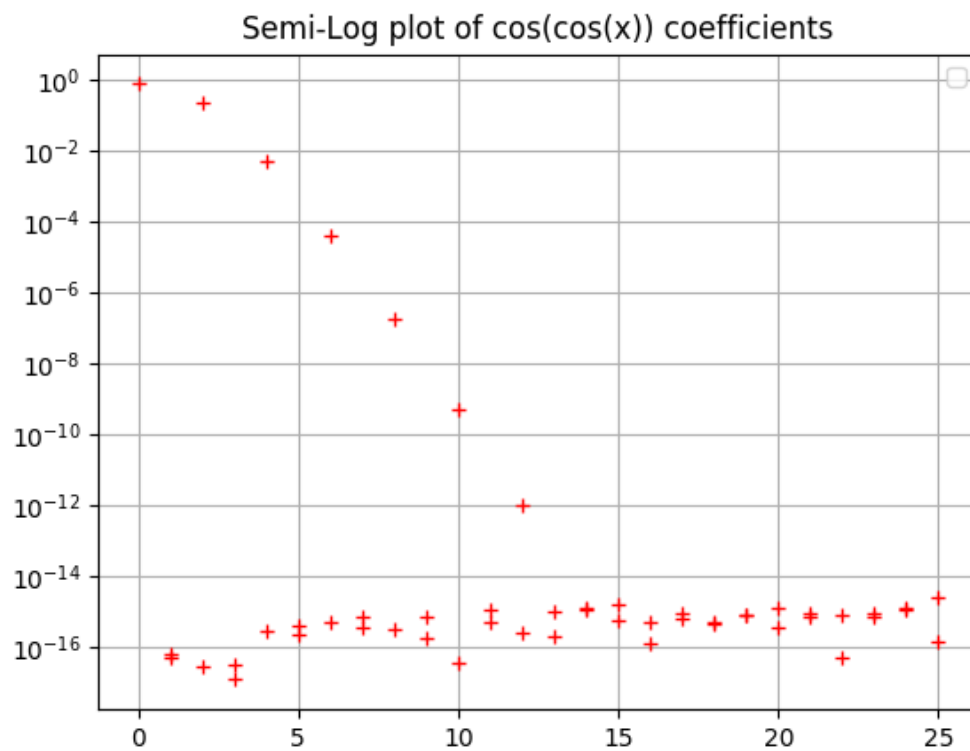


Figure 6: semi-log plot of $\cos(\cos(x))$

```
plt.loglog(n[1:],abs(b_exp[1:]),"r+")
plt.grid(True)
plt.title("Log-Log plot of ex's coefficients")
plt.show()

#plotting Exp(x) coefficients on semi-log scale
plt.semilogy(n,abs(a_exp),"r+")
plt.semilogy(n[1:],abs(b_exp[1:]),"r+")
plt.grid(True)
plt.title("Semi-Log plot of ex's coefficients")
plt.show()
```

2.8 Answer to problems

- 3) (a) For $\cos(\cos(x))$ b_n comes out close to zero because $\cos(\cos(x))$ is an even function and thus will have coefficient of even part i.e. $\sin(nx)$ close to zero.
- 3) (b) e^x is not a periodic signal and thus it made periodic by repeating and it is also not a sinusoidal function. Due to discontinuities (emerged due to its repetition) its Fourier series will have high frequency components and we are calculating only 25 terms. Where as $\cos(\cos(x))$ is a periodic sinusoidal function and thus its series converges much faster.
- 3) (c) Coefficients of e^x i.e.

$$a_n, b_n \propto \frac{1}{n^2 + 1}$$

and that is why slope = $-\log(n^2+1) \cdot \text{constant} \approx -2 \cdot \log(n) \cdot \text{constant}$. This shows its linear in log-log scale. Where as coefficients $\cos(\cos(x))$ are linearly in semi-log because it is infinitely differentiable and thus its

$$a^n, b^n \propto \frac{1}{c^n}$$

where c is less than 1. So log of the comes out to be n times constant which is negative since n is negative.

2.9 Alternative Method : Least Square Approach

$$f(x) = \frac{a_0}{2} + \sum_{n=1}^{\infty} a_n \cos(nx) + b_n \sin(nx)$$

The above equation in matrix form can be represented as

$$\begin{bmatrix} 1 & \cos(x_1) & \sin(x_1) & \dots & \cos(25x_1) & \sin(25x_1) \\ 1 & \cos(x_2) & \sin(x_2) & \dots & \cos(25x_2) & \sin(25x_2) \\ \dots & \dots & \dots & \dots & \dots & \dots \\ 1 & \cos(x_{400}) & \sin(x_{400}) & \dots & \cos(25x_{400}) & \sin(25x_{400}) \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ b_1 \\ \dots \\ a_{25} \\ b_{25} \end{bmatrix} = \begin{bmatrix} f(x_1) \\ f(x_2) \\ \dots \\ f(x_{400}) \end{bmatrix} \quad (1)$$

We can obtain the R.H.S. of the equation by evaluating functional values at 400 points. Similarly we can obtain \cos and \sin values at those points to obtain the first matrix. By using Least Squares we can obtain the coefficient matrix. In this assignment we are using in-built least square function of python.

```
x2 = np.linspace(0,2*np.pi,401)
x2=x2[:-1]
```

```
def coeff_2(f,x):
    b = f(x)
    A = np.zeros((400,51))
    A[:,0]=1
    for k in range(1,26):
```

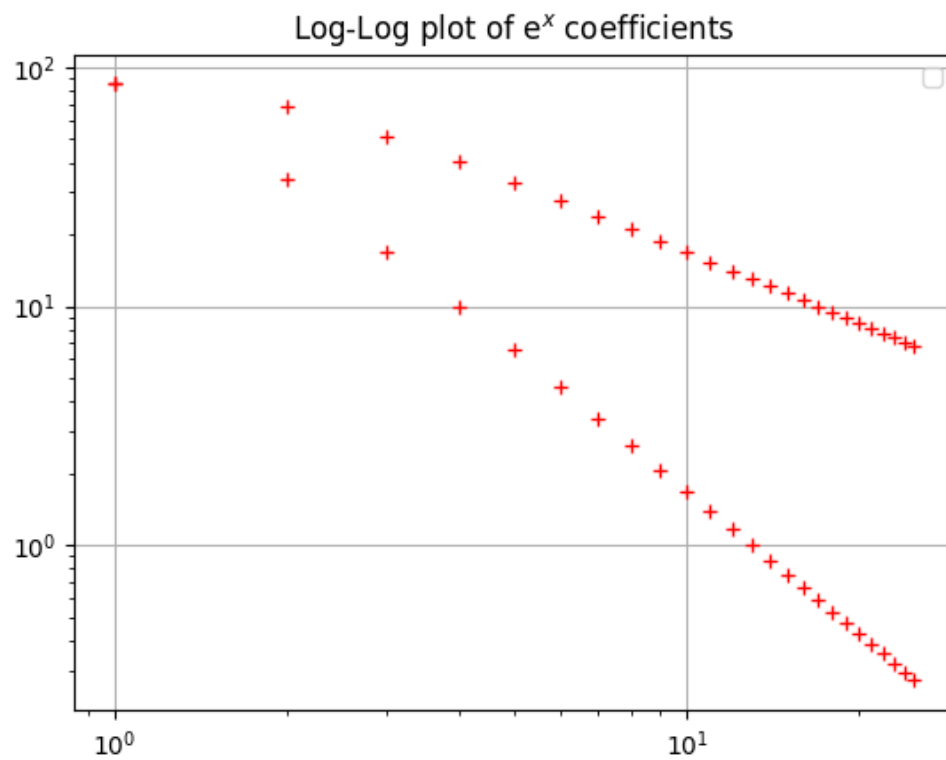



Figure 7: Log-Log plot of coefficient of $\exp(x)$

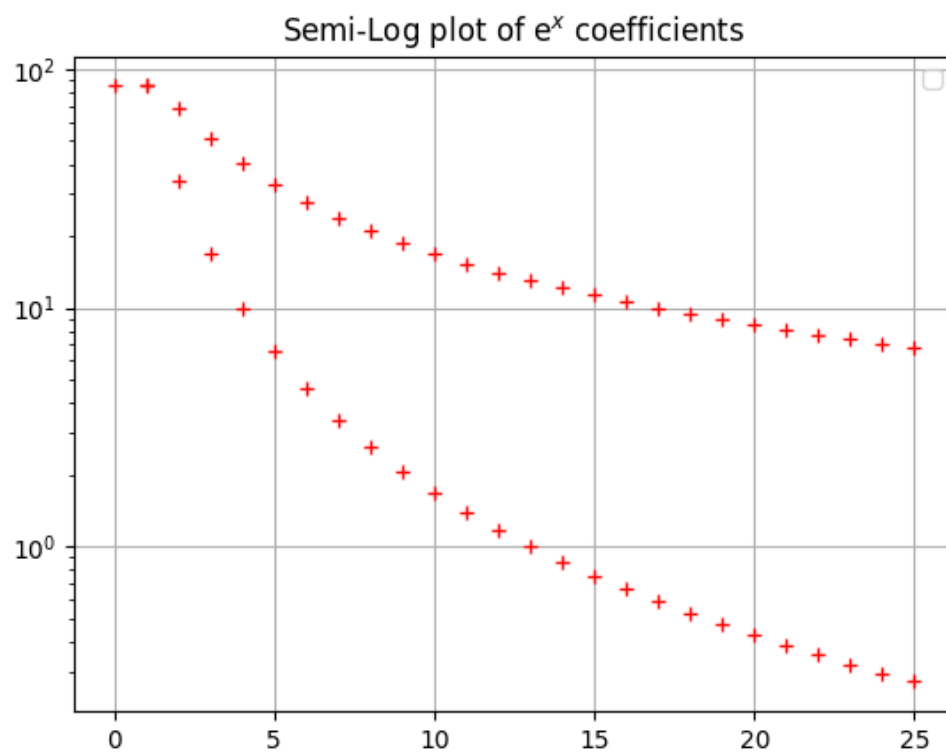


Figure 8: Semi-log plot of coefficient of $\exp(x)$

```

        A[:,2*k-1] = np.cos(k*x)
        A[:,2*k] = np.sin(k*x)
    return linalg.lstsq(A,b)[0]#using least square method built-in python function

```

The above defined function takes $f(x)$ and x vector i.e. the points at which we want to evaluate. For this we want points between 0 to 2π with 400 points.

2.10 Plotting $\cos(\cos(x))$ Fourier coefficients obtained by least squares

Calling the function defined in last code fragment and plotting the obtained Fourier coefficients against n in log-log scale. Obtained vector with coefficients is not indexed by n . To do so we have to divide a and b in two vectors. Adding a_0 to a vector named $a2_coscos$ and remaining even indexes where as odd indexes to b vector named $b2_coscos$. See Figure 9.

```

temp = coeff_2(coscos,x2)

#separating a and b coefficients
a2_coscoss = np.zeros(26)
a2_coscoss[0] = temp[0]
a2_coscoss[1:] = temp[1::2]
b2_coscoss = temp[2::2]

#plotting log-log of fourier coefficients of cos(cos(x)) by both the methods
plt.loglog(n,abs(a2_coscoss),"ro")
plt.loglog(n[1:],abs(b2_coscoss),"ro")
plt.loglog(n,abs(a_coscossnew),"gx")
plt.loglog(n[1:],abs(b_coscossnew[1:]),"gx")
plt.grid(True)
plt.legend("coefficients by least square","coefficients by integral")
plt.title("Log-Log plot of cos(cos(x)) coefficients")
plt.show()

```

2.11 Plotting $\exp(x)$ Fourier coefficients obtained by least squares

Calling the function defined in the code fragment one above and plotting the obtained Fourier coefficients against n in log-log scale. Obtained vector with coefficients is not indexed by n . To do so we have to divide a and b in two vectors. Adding a_0 to a vector named $a2_cosex$ and remaining even indexes where as odd indexes to b vector named $b2_cosex$. See Figure 10.

```

temp = coeff_2(exp,x2)
#separating a and b coefficients
a2_exp = np.zeros(26)
a2_exp[0] = temp[0]
a2_exp[1:] = temp[1::2]
b2_exp = temp[2::2]

#plotting log-log of fourier coefficients of exp(x) by both the methods
plt.loglog(n,abs(a2_exp),"ro")
plt.loglog(n,abs(a_exp),"gx")
plt.loglog(n[1:],abs(b2_exp),"ro")
plt.loglog(n[1:],abs(b_exp[1:]),"gx")
plt.grid(True)
plt.legend("coefficients by least square","coefficients by integral")
plt.title("Log-Log plot of exp(x) coefficients")
plt.show()

```

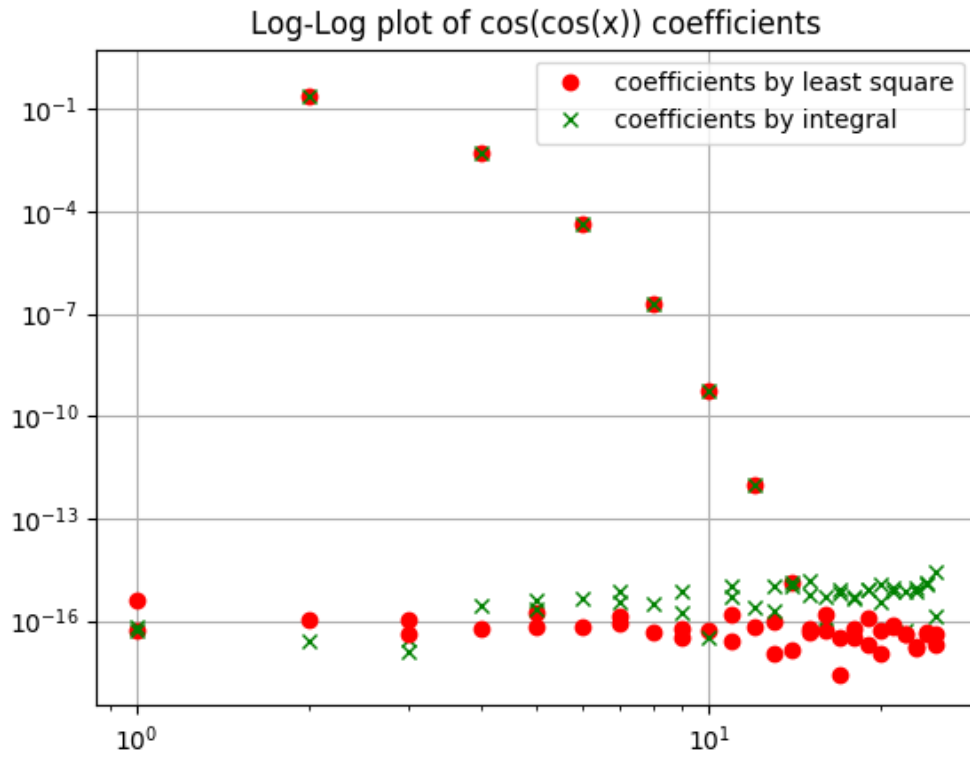


Figure 9: Green cross are obtained by integral and Red dots by least squares

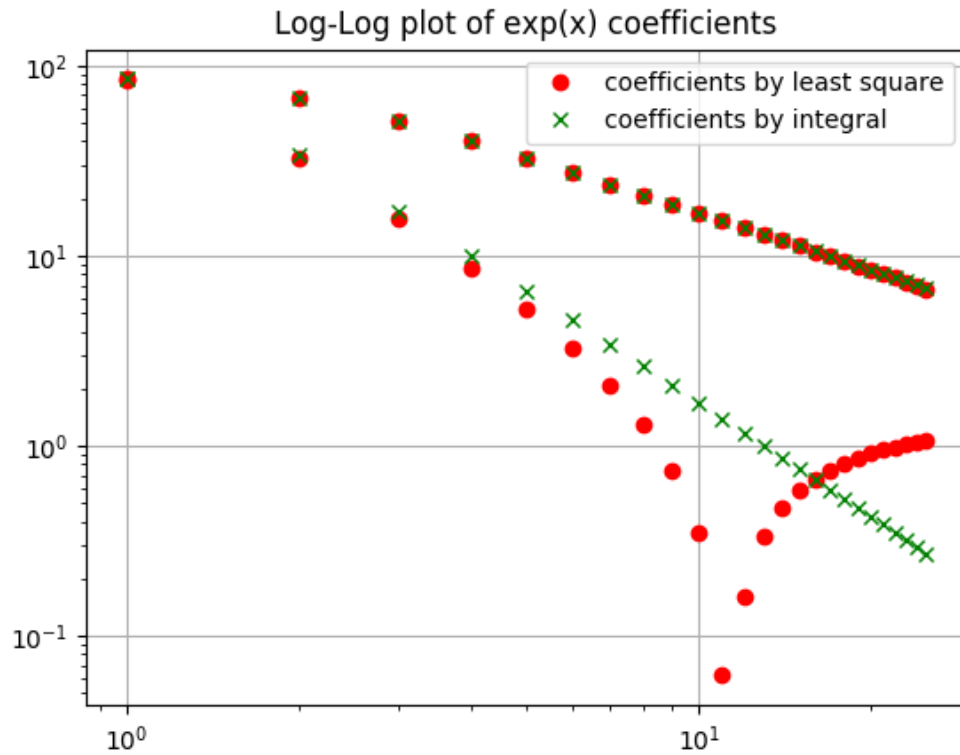


Figure 10: Green cross are obtained by integral and Red dots by least squares

2.12 Error Measurement Between The two Methods

For this case we can calculate estimated error i.e. the maximum value of the absolute difference between the coefficients obtained by the two methods.

2.12.1 Answer to the problem

• 6) No the expected output from both the method is not the same and should not be the same .First method each coefficient is calculated by solving integral which gives correct value.In the second case by least squares since we are taking only 25 coefficients approximation by this method just neglects high frequency components that the signal might have. Hence it gives just an approximate value of the coefficients.

```
a_diff_coscoss = a_coscossnew - a2_coscoss
b_diff_coscoss = b_coscossnew[1:] - b2_coscoss

plt.plot(n,a_diff_coscoss,"ro")
plt.plot(n[1:],b_diff_coscoss,"g+")
plt.show()

a_diff_exp = a_exp - a2_exp
b_diff_exp = b_exp[1:] - b2_exp

plt.plot(n,a_diff_exp,"ro")
plt.plot(n[1:],b_diff_exp,"g+")
plt.show()

print "maximum a diff for e(x) ",max(a_diff_exp)
print "maximum b diff for e(x) ",max(b_diff_exp)
print "maximum a diff for coscos(x) ",max(a_diff_coscoss)
print "maximum b diff for coscos(x) ",max(b_diff_coscoss)
```

The output from the above code gives us,

```
>> maximum a diff for e(x)  1.33273087034
>> maximum b diff for e(x)  0.00349821093269
>> maximum a diff for coscos(x)  8.31811329195e-16
>> maximum b diff for coscos(x)  2.57586713575e-15
```

From this its clear that the error is extremely small for $\cos(\cos(x))$ coefficients . whereas the error is quite significant for e^x .

Listing 3: To generate and plot $1/(1+t^2)$

```
coeff_coscoss = []
coeff_coscoss.append(a_coscossnew[0])
for i in range(1,26):
    coeff_coscoss.append(a_coscossnew[i])
    coeff_coscoss.append(b_coscossnew[i])

coeff_exp = []
coeff_exp.append(a_exp[0])
for i in range(1,26):
    coeff_exp.append(a_exp[i])
    coeff_exp.append(b_exp[i])
```

2.13 Approximating the function back by Fourier series

From the evaluated coefficients we will regenerate the function and plot it. To obtain the Fourier series sum for convenience merging vector with $a^o, a^1, b^1, a^2...$ and

so on. After converting again substituting this vector in place of coefficient vector in the earlier mentioned matrix equation. After multiplying this to the matrix containing $\sin(nx)$ and $\cos(nx)$ evaluated and points whose values we desire to obtain. From this matrix multiplication we can obtain reconstructed functional values at 400 points in our case .

Listing 4: To generate and plot $1/(1+t^2)$

```
x3 = np.linspace(0,2*np.pi,401)
x3=x3[:-1]
B = np.zeros((400,51))
B[:,0]=1

#Constructing matrix with sinnx and cosnx
for k in range(1,26):
    B[:,2*k-1] = np.cos(k*x3)
    B[:,2*k] = np.sin(k*x3)

#multiplying fourier coefficients with sinosoids to approximate the function
re_coscoss = np.matmul(B,coeff_coscoss)
re_exp = np.matmul(B,coeff_exp)
```

The code below is to plot the values obtained above.For $\cos(\cos(x))$ and $\exp(x)$ see Figure 11 and Figure 12 respectively.

Listing 5: To generate and plot $1/(1+t^2)$

```
plt.plot(x3,re_coscoss)
plt.title("Approximation_of_cos(cos(x))_by_fourier_series")
plt.grid(True)
plt.show()

plt.plot(x3,re_exp)
plt.title("Approximation_of_e^{x}_by_fourier_series")
plt.grid(True)
plt.show()
```

2.14 Answer to the problem

As already mentioned in the answer to last problem. The approximation least squares takes by taking only first 25 terms is to neglect functions higher frequency components. e^x has more higher frequency components and can not have an accurate approximation with just 25 coefficients compared to $\cos(\cos(x))$ which converges much faster as it has only low frequency components and higher order coefficients tend to zero. The deviation that are at the edges in figure 12 are due to Gibbs phenomena . Since e^x is not periodic , for calculating Fourier series the function is repeated from $0 - 2\pi$ and has discontinuity at both the ends.

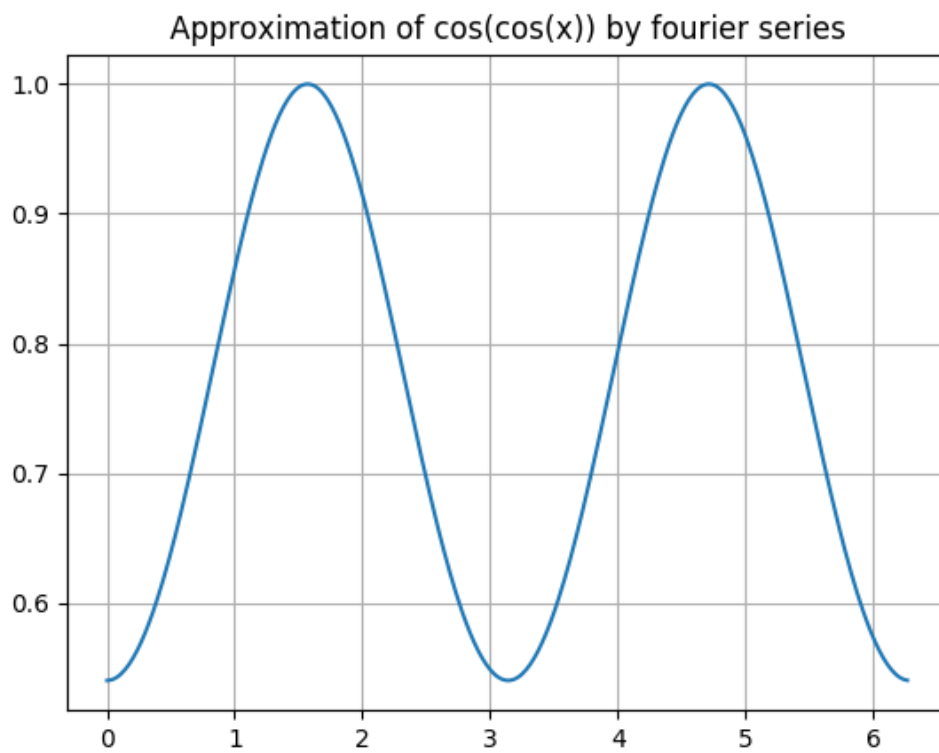


Figure 11: Approximation of $\cos(\cos(x))$

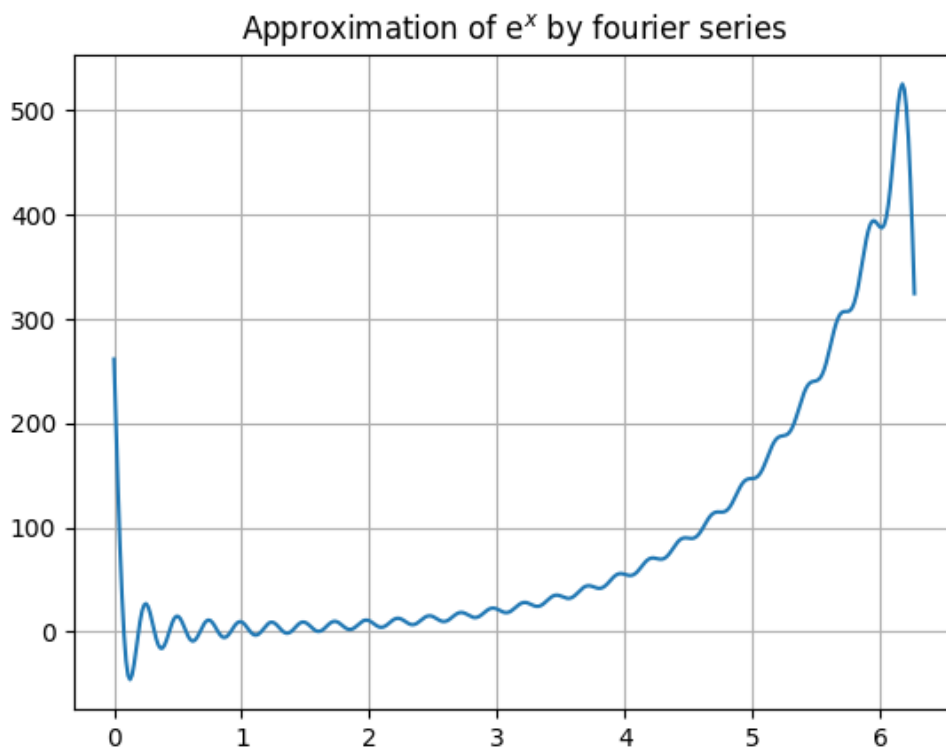


Figure 12: Approximation of $\exp(x)$