# Assignment 4: Fourier Approximations

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February 26, 2022

#### 1 Aim

The goal of this assignment is to fit the functions, exp(x) and cos(cos(x)) over the interval  $[0, 2\pi)$  using the trigonometric fourier series

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

where the coefficients are given by

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

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

## 2 The Assignment

## 2.1 Question 1: Getting the functions

I start by defining Python functions for each of the two functions. These functions return numpy arrays. Note that numpy has been imported under the alias np.

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

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

Now, I plot each of them over the range  $[-2\pi, 4\pi)$  in two ways - the actual function and the periodic extension with a  $2\pi$  period. Note that  $e^x$  has been plotted on a semilog scale. The following Python code does this (and similarly for  $\cos(\cos(x))$ ):

```
x = np.linspace(-2 * np.pi, 4 * np.pi, 300)
y1 = exponential(x)
y1_periodic = y1[100: 200]
y1_periodic = np.append(y1_periodic, y1_periodic)
y1_periodic = np.append(y1_periodic, y1[100: 200])
# because I am plotting over three periods

semilogy(x, y1, 'red', label='Main function')
semilogy(x, y1_periodic, label='Periodic over 0 to 2\u03C0')
grid()
xlabel("x")
ylabel("e^x")
legend()
title("Q1: Exponential(x)")
show()
```

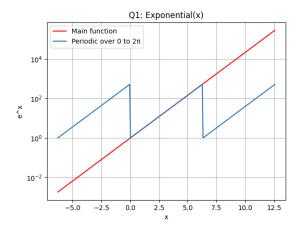


Figure 1: Q1: The Exponential Function

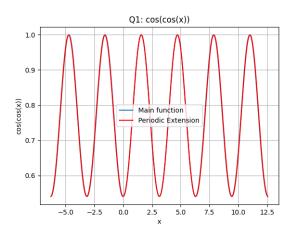


Figure 2: Q1: The cos(cos(x)) Function

As one may observe, the periodic extension of cos(cos(x)) matches the actual function because it is inherently periodic.

## 2.2 Question 2: Building the Fourier Series

The goal is to obtain 51 coefficients for each of the two functions. The Fourier coefficients, in order,  $a_0, a_1, b_1, \ldots, a_{25}, b_{25}$ . I have created two new functions to be integrated, namely f(x, k) and g(x, k). This is implemented for the exponential function in the below code block. Note that k is the second argument of the function. The integrator function integrate quad from the scipy module integrates its first argument (integrand) from limits 0 to  $2\pi$ ; the fourth argument passes k to the function f(x, k)

```
f0 = lambda a: np.exp(a)
a0_exp = (scipy.integrate.quad(f0, 0, 2 * np.pi)[0])/(2 * np.pi)

# 'a' coefficients for this series
a_n_exp = []
for i in range(1, 26):
    f = lambda x, k: np.exp(x) * np.cos(x * k)
    a_n_exp.append((scipy.integrate.quad(f, 0, 2 * np.pi,
        args=(i))[0]) / np.pi)

b_n_exp = []
for i in range(1, 26):
    g = lambda x, k: np.exp(x) * np.sin(x * k)
    b_n_exp.append((scipy.integrate.quad(g, 0, 2 * np.pi,
        args=(i))[0]) / np.pi)
```

And for the cos(cos(x)) function,

```
p0 = lambda x: np.cos(np.cos(x))
a0_cos = (scipy.integrate.quad(p0, 0, 2 * np.pi)[0]) / (2 * np.pi)

a_n_cos = []
for i in range(1, 26):
    p = lambda x, k: np.cos(np.cos(x)) * np.cos(x * k)
    a_n_cos.append((scipy.integrate.quad(p, 0, 2 * np.pi,
        args=(i))[0]) / np.pi)

b_n_cos = []
for i in range(1, 26):
    q = lambda x, k: np.cos(np.cos(x)) * np.sin(x * k)
    b_n_cos.append((scipy.integrate.quad(q, 0, 2 * np.pi,
        args=(i))[0]) / np.pi)
```

The coefficients  $a_n$  and  $b_n$  are plotted.

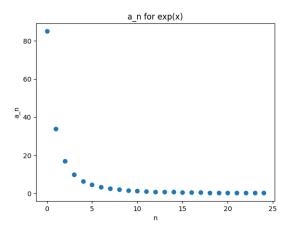


Figure 3: Q3:  $a_n$  for exp(x)

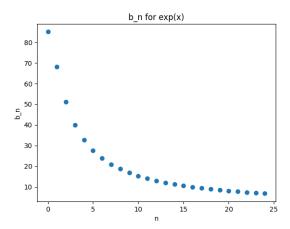


Figure 4: Q3:  $b_n$  for exp(x)

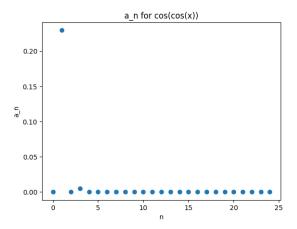


Figure 5: Q3:  $a_n$  for cos(cos(x))

## 2.3 Question 3: Plotting the coefficients

For each of the two functions, I have plotted the Fourier coefficients' magnitude - once on a semilog scale and once on a loglog scale. Note that the order of coefficients on the x-axis (For the figures with 'Q3' in the caption) is  $a_0, a_1, b_1, \ldots, a_{25}, b_{25}$ .

First, the Fourier coefficients of exp(x).

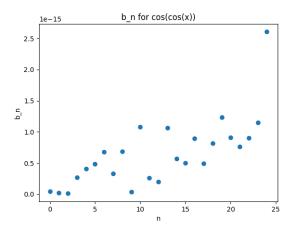


Figure 6: Q3:  $b_n$  for cos(cos(x))

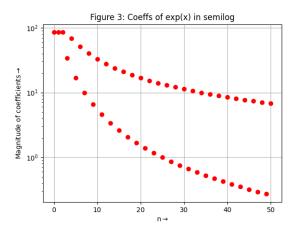


Figure 7: Q3: Coefficients of exp(x)

Next, the Fourier coefficients of cos(cos(x)).

The order of magnitude tells a lot about the nature of the series, where the harmonic contribution is significant and where it is not.

1.  $b_n$  coefficients are observed in Figure 6. The y-axis numbers are on a  $10^{-15}$  scale - nearly zero. The reason is, the function period is  $2\pi$  and I am integrating it over 0 to  $2\pi$ . If I change the integration variable to  $t+\pi$ , limits will change from  $-\pi$  to  $\pi$ , the integrand will remain the same (may get a +/- sign). The function  $\cos(\cos(x))\sin(nx)$  is now an odd function. Hence, the integral should vanish. I obtain very tiny values because I used a finite number of x for the integration function quad.

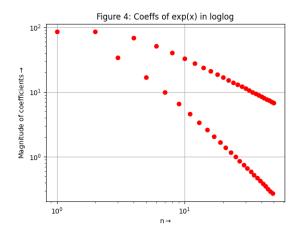


Figure 8: Q3: Coefficients of exp(x)

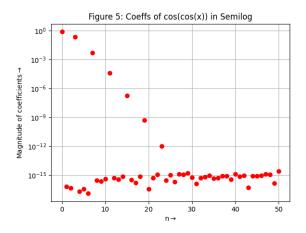


Figure 9: Q3: Coefficients of cos(cos(x))

- 2. The reason why the coefficients of  $e^x$  do not decay as quickly as that of  $\cos(\cos(x))$  is the latter is inherently periodic. In the Fourier domain, it will have a finite number of frequencies making up the signal. On the contrary,  $e^x$  is not, hence, it is discontinuous at  $2m\pi$ .
- 3. Fourier coefficients for  $e^x$  are

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

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

At large values of n,  $log(a_n)$  and  $log(b_n)$  can be approximated as -2 log(n)

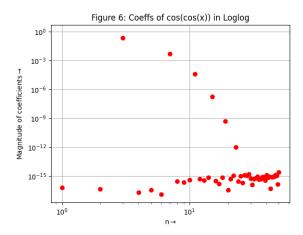


Figure 10: Q3: Coefficients of cos(cos(x))

and  $-\log(n)$  respectively. Hence the loglog plot is linear. Whereas the semilog plot in (caption) Figure 9 looks linear because the coefficients vary approximately as exponential n.

#### 2.4 Question 4 and 5: The Least Squares Approach

I have defined a vector X going from 0 to  $2\pi$  in 400 steps. Lets deal with  $e^x$  first. The function  $e^x$  at these X values is given by another vector b\_matrix\_for\_exp. Now I build the matrix A\_mat\_for\_exp; the code is as follows:

```
X = np.linspace(0, 2*np.pi, 400)
b_matrix_for_exp = exponential(X)
X = np.linspace(0, 2*np.pi, 401)
X = X[:-1] \# drop last term to have a proper periodic
   integral
b_matrix_for_exp = exponential(X)
# f has been written to take a vector
A_mat_for_exp = np.zeros((400, 51)) # allocate space for A
A_mat_for_exp[:, 0] = 1
                        # col 1 is all ones
for k in range (1, 26):
   A_mat_for_exp[:,2*k-1] = np.cos(k*X) # cos(kx) column
   A_{mat_for_exp[:,2*k]} = np.sin(k*X)
                                        # sin(kx) column
coeff_for_exp_lstsq = np.linalg.lstsq(A_mat_for_exp,
b_matrix_for_exp)[0]
```

Similar to an application in the previous assignment, the function nplinalglstsq

returns (the first return, hence, [0]) the vector x that minimizes the Euclidean norm  $|Ax - b|^2$ , implying the closest solution x. This is what I am looking for. In this case the vector x would be the fourier coefficients in order  $a_0, a_1, b_1, \ldots, a_{25}, b_{25}$ .

```
(A_mat_for_exp).(coeff_for_exp_lstsq) = (b_matrix_for_exp)
```

Then I plot these coefficients. Note that a lighter shade has been used for the plots to show overlap.

```
semilogy(np.array(range(51)), (coeffs_of_coscos), 'ro',
   alpha=0.4, label='Original Value')
semilogy(np.array(range(51)), abs(coeff_for_coscos_lstsq),
   'go', alpha=0.4, label='Least Squares Approach')
xlabel(r'n$\rightarrow$')
ylabel(r'Magnitude of coefficients$\rightarrow$')
title('Figure 8: Coeffs of cos(cos(x)) lstsq approach')
legend()
grid()
show()
```

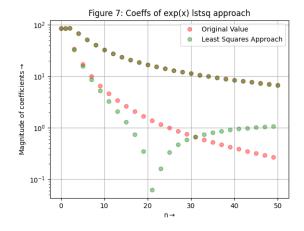


Figure 11: Q5: 1stsq coefficients for  $e^x$  (semilog)

Similar procedure for cos(cos(x)) gives the following plots.

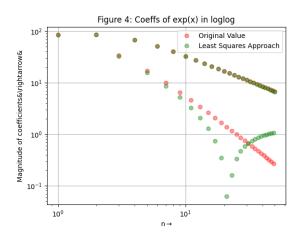


Figure 12: Q5: 1stsq coefficients for  $e^x$  (loglog)

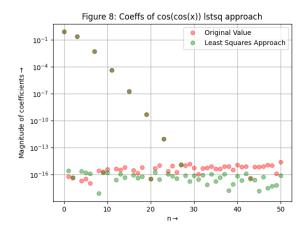


Figure 13: Q5: 1stsq coefficients for cos(cos(x)) (semilog)

#### 2.5 Question 6: Least Squares Vs Integration!

I have calculated the absolute difference between the coefficients obtained from Integration and the least squares approach. The plots and code for both the functions follow.

```
# PART 6: Deviation
# First the exponential
error_exp = coeffs_of_exp - abs(coeff_for_exp_lstsq)
plot(range(51), abs(error_exp), 'ro')
xlabel('n')
ylabel('real coeffs - lstsq coeffs')
title('Deviation in coeffs of exp(x)')
```

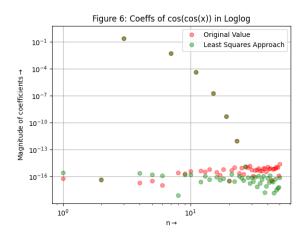


Figure 14: Q5: 1stsq coefficients for cos(cos(x)) (loglog)

```
grid()
show()
print(np.amax(abs(error_exp)))

# Second the coscos
error_coscos = coeffs_of_coscos - abs(coeff_for_coscos_lstsq)
plot(range(51), abs(error_coscos), 'ro')
xlabel('n')
ylabel('real coeffs - lstsq coeffs')
grid()
title('Deviation in coeffs of cos(cos(x))')
show()
print(np.amax(abs(error_coscos)))
```

Maximum deviations are printed by the python program.

```
Max deviation for e^x = 1.3327308703353395
Max deviation for cos(cos(x)) = 2.5195381376685242e-15
```

There is appreciable deviation for some of the coefficients of  $e^x$  (there is a row of dots above 1.2). The deviation observed for cos(cos(x)) coefficients is of the order  $10^{-15}$  implying that the least squares and integration coefficients are amazingly close. This difference is because the former is discontinuous at the boundaries but the latter is essentially periodic.

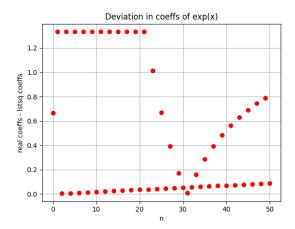


Figure 15: Q6:  $e^x$  deviation

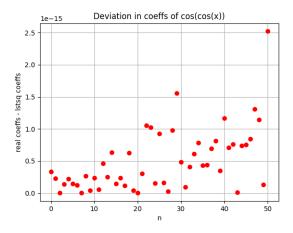


Figure 16: Q6: cos(cos(x)) deviation

## 2.6 Question 7: Function values' deviation

Computing Ac (The function values at  $x_i$ ) from the estimated values of c. The Python code:

```
legend()
show()

plot(X, Ac_coscos, 'go', label='Obtained from lstsq
    and Fourier', alpha=0.4)
plot(X, coscos(X), label='True Value')
xlabel('X')
ylabel('Function values at Xi')
legend()
show()
```

The function values hence obtained are plotted along with the 'true' values. Significant inaccuracy can be observed for the  $e^x$  plot because it would need a larger number of high frequency components to obtain the exact function. On the other hand, the estimated cos(cos(x)) function is extremely close to the real values because, in frequency domain, it was almost completely covered up by the first few fourier coefficients (harmonics).

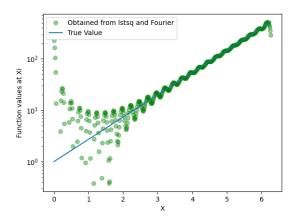


Figure 17: Q7:  $e^x$  (semilog)

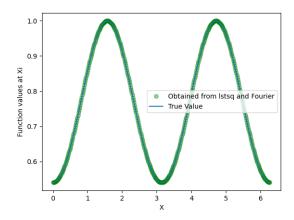


Figure 18: Q7: cos(cos(x))

## 3 Conclusion

The aims that I enlisted at the beginning of this report are now seen to have been accomplished. Through every subsection, I analysed how Fourier coefficients will be calculated using scientific Python and every plot gave a deeper insight into their nature. I infer that for a function inherently periodic, the fourier coefficients obtained by both integration and least squares will give accurate results for a finite number of coefficients. However, non-periodic functions when given a periodic extension will create a discontinuity at every integral multiple of Time period. Thus, it will take a greater amount of fourier coefficients to reverse-realise the original function.