

# EE2703: Applied Programming Lab

## Assignment 4: Fourier Approximations

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### 1 Introduction

Two functions,  $\exp(x)$  and  $\cos(\cos(x))$  over the interval  $[0, 2\pi)$  will be modeled using the fourier series:

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

Where,

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

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

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

## 2 Subquestions

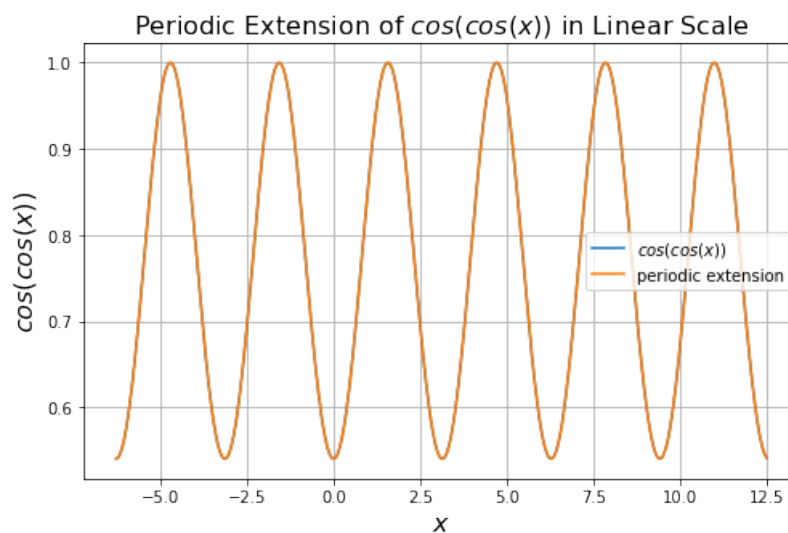
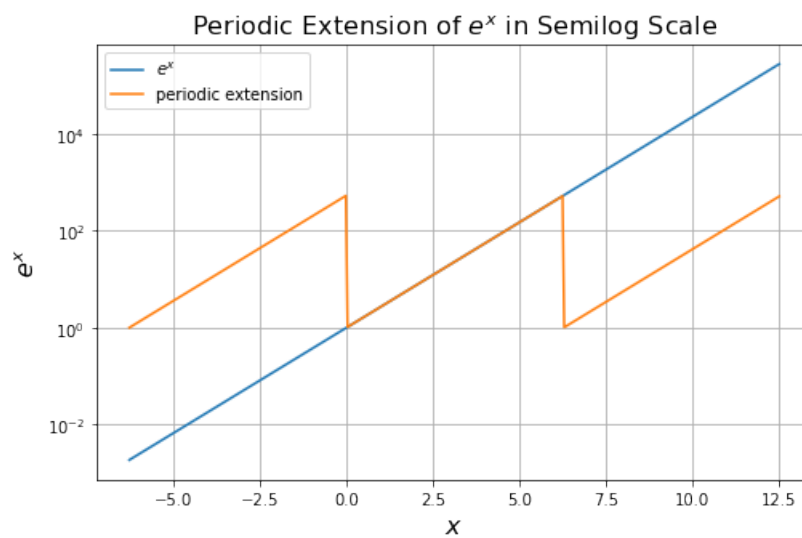
### 2.1 Define & Plot Python Functions

Two Python functions are defined:

```
def exp(x): # exponential function, supports value or vector
    return np.exp(x)

def coscos(x): # cos of cos function, supports value or vector
    return np.cos(np.cos(x))
```

The functions have been plotted with their periodic extensions ( $2\pi$  period):



## 2.2 Evaluate Integrals

The integrands have been defined as functions of  $x$ ,  $k$ , and  $f$ , where  $f$  is the Python function to be approximated, i.e. `exp` or `coscos`:

```
def u(x, k, f): # f is either exp or coscos
    return f(x) * np.cos(k * x)

def v(x, k, f): # f is either exp or coscos
    return f(x) * np.sin(k * x)
```

The integrals to calculate the values of  $a_n$  and  $b_n$  have been evaluated using the following loop:

```
F = [exp, coscos]
a_0 = np.zeros((2))
a_n = np.zeros((2, 25))
b_n = np.zeros((2, 25))

for i in range(2): # iterate over exp and coscos
    a_0[i] = quad(F[i], 0, 2 * np.pi)[0] / (2 * np.pi)

for j in range(25): # integration
    a_n[i, j] = quad(u, 0, 2 * np.pi, args=(j + 1, F[i]))[0] / np.pi
    b_n[i, j] = quad(v, 0, 2 * np.pi, args=(j + 1, F[i]))[0] / np.pi
```

## 2.3 Plot the Fourier Coefficients

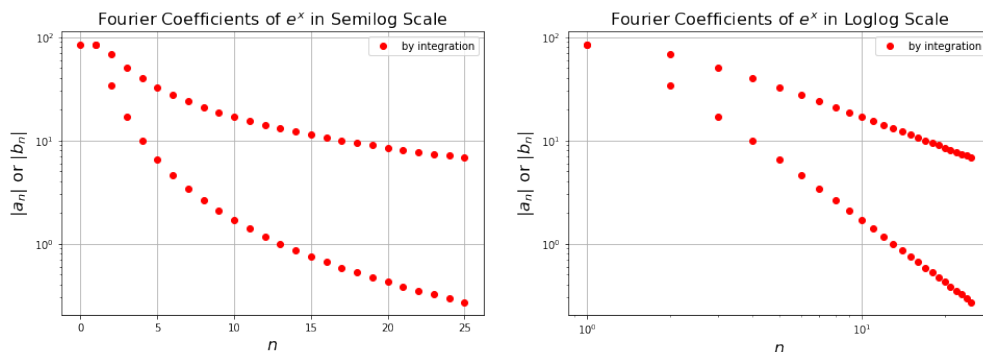
The answer vector  $c$  has been generated by:

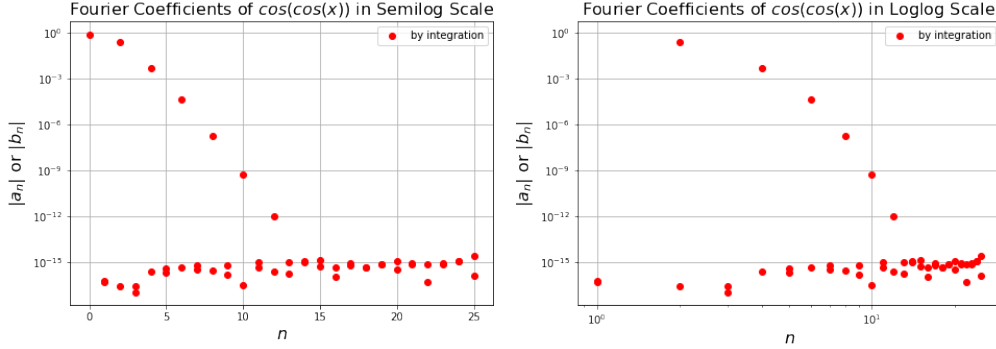
```
c_n = np.zeros((2, 51)) # c_n[0] and c_n[1] are the answer vectors
c_n[:, 0] = a_0
c_n[:, 1::2] = a_n # alternate elements starting at 1
c_n[:, 2::2] = b_n # alternate elements starting at 2
```

To plot in semilog scale, for example, we use:

```
plt.semilogy(0, np.abs(a_0[f_i]), 'ro', label="by integration")
plt.semilogy(range(1, 26), np.abs(a_n[f_i]), 'ro')
plt.semilogy(range(1, 26), np.abs(b_n[f_i]), 'ro')
```

The coefficients  $|a_n|$  and  $|b_n|$  have been plotted:





- (a) The  $b_n$  coefficients for  $\cos(\cos(x))$  are of the order of magnitude  $-16$ , i.e. nearly zero.  
This is because  $\cos(\cos(x))$  is an even function, and thus does not have any odd component.
- (b)  $\cos(\cos(x))$  is a periodic function, comprising not many frequencies.  $e^x$ , on the other hand, is a non-periodic function, and thus its periodic extension has multiple discontinuities. Hence, high frequency components are required to represent this function as a sum of trigonometric functions.
- (c) The *loglog* plot is linear for  $e^x$  because the *log* of its coefficients vary linearly with  $\log(n)$ , i.e. the coefficients depend on  $\frac{1}{n^a}$  for some  $a$ .  
On the other hand, the *semilog* plot is linear for  $\cos(\cos(x))$  because the *log* of its coefficients vary linearly with  $n$ , i.e. the coefficients decay exponentially.

## 2.4 Least Squares

The equation to be solved by `scipy.linalg.lstsq` is:

$$Ac = B$$

The matrix  $A$  has been generated by:

```
A = np.ones((400, 51)) # first column should be ones
for k in range(1, 26):
    A[:, 2 * k - 1] = np.cos(k * x)
    A[:, 2 * k] = np.sin(k * x)
```

The solution matrices are  $c[0]$  for  $e^x$  and  $c[1]$  for  $\cos(\cos(x))$ :

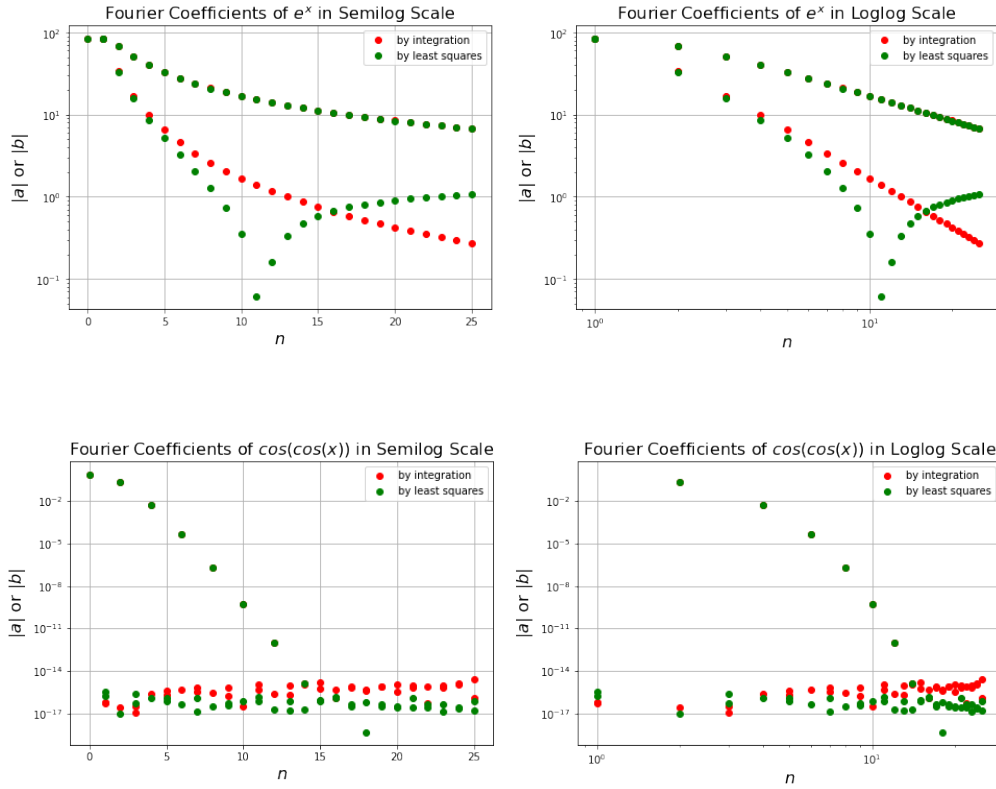
```
for i in range(2):
    b = F[i](x)
    c[i] = lstsq(A, b)[0]
```

## 2.5 Plot the Best Fit Coefficients

To plot in semilog scale, for example, we use:

```
plt.semilogy(0, np.abs(c[f_i][0]), 'go', label="by_least_squares")
plt.semilogy(range(1, 26), np.abs(c[f_i][1::2]), 'go')
plt.semilogy(range(1, 26), np.abs(c[f_i][2::2]), 'go')
```

The coefficients  $|a_n|$  and  $|b_n|$  by least squares method have been contrasted with those by integration:



## 2.6 Compare Least Squares and Direct Integration

The maximum errors between the least squares and direct integration methods has been calculated:

```
np.max(np.abs(c[0] - c_n[0])) # for  $e^x$ 
np.max(np.abs(c[1] - c_n[1])) # for  $\cos(\cos(x))$ 
```

The largest deviation of coefficients for  $e^x$  is : 1.3327

The largest deviation of coefficients for  $\cos(\cos(x))$  is : 2.6684e-15

The error is significant for  $e^x$  but negligible for  $\cos(\cos(x))$ .

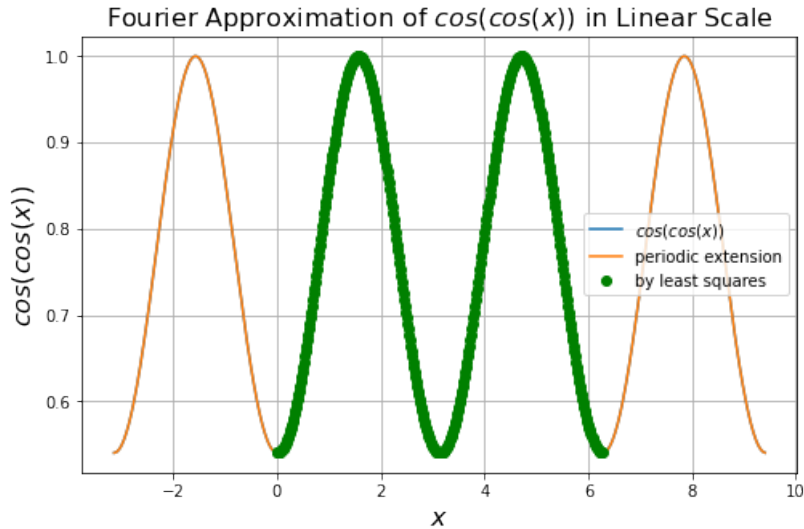
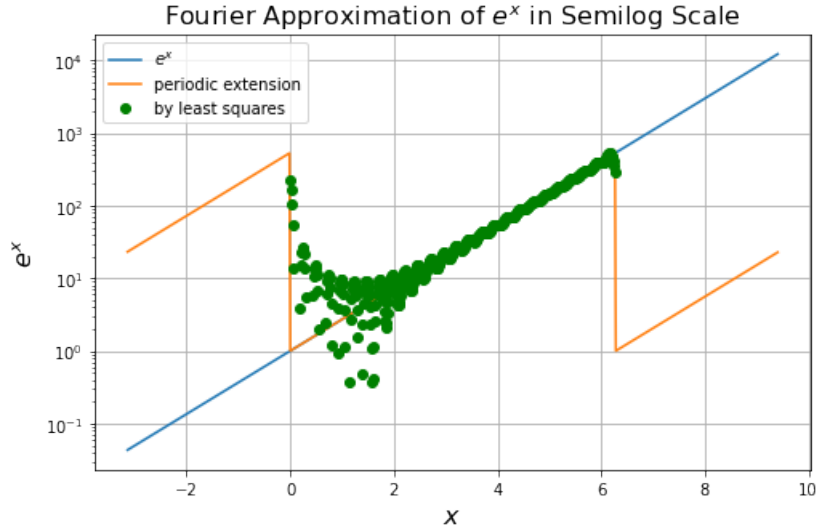
This is because there are multiple discontinuities in the periodic extension of  $e^x$ , and thus would require a much higher number of coefficients to be somewhat accurately represented.

This lack in accuracy is more apparent close to those discontinuities.

## 2.7 Plot the Fourier Approximations

The functions as represented by the Fourier coefficients calculated through the least squares method have been plotted:

```
plt.plot(x[200:600], np.dot(A, c[1]), 'go', label="by least squares")
```



The  $\cos(\cos(x))$  plot agrees nearly perfectly, but the  $e^x$  plot has a large deviation. This is because unlike  $\cos(\cos(x))$ , which has periodic components, the  $e^x$  function is not inherently comprised of periodic trigonometric functions. The periodic extension of  $e^x$  has multiple discontinuities, which causes the Fourier representation to overshoot around them due to the Gibbs phenomenon.