# EE2703: Report

# Assignment 6 - Speeding up with Cython EE23B092

# 1 Implementation

## 1.1 Python implementation

The basic idea is to sum up all the areas of the trapeziums. The function accepts 4 parameters - the function to integrate (f), lower bound (a), upper bound (b) and the number of trapeziums (n).

## 1.2 Cython Implementation

It is similar in structure to the python implementation. It uses cdef for function and variable definitions along with C data types. Some Cython decorators are called to speed up the performance. The function f is made inline (as we call it multiple times) which makes it run faster. We also use a wrapper function which calls the actual function cy\_trapz(f, a, b, n) to measure the time taken.

# 2 Performance and Accuracy on different functions (10 million trapeziums)

# 2.1 $f(x) = x^2$ from 0 to 1

### 2.1.1 Cython implementation

Time taken:- 14.9 ms  $\pm$  439 µs per loop — Result:- 0.3333333333333358258

#### 2.1.2 Numpy Implementation

Time taken:- 88.3 ms  $\pm$  1.61 ms per loop — Result:- 0.333333333333333504

#### 2.1.3 Python Implementation

Time taken:- 1.99 s  $\pm$  18.9 ms per loop — Result:- 0.3333333333333358258

2.2 f(x) = sin(x) from 0 to  $\pi$ 

#### 2.2.1 Cython implementation

Time taken:  $490 \text{ ms} \pm 15 \text{ ms} \text{ per loop} - \text{Result:- } 2.000000000413601$ 

#### 2.2.2 Numpy Implementation

Time taken:- 94.3 ms  $\pm$  775 µs per loop — Result:- 1.9999999999999

#### 2.2.3 Python Implementation

Time taken:- 1.94 s  $\pm$  73.3 ms per loop — Result:- 2.000000000413601

2.3  $f(x) = e^x$  from 0 to 1

#### 2.3.1 Cython implementation

Time taken:- 432 ms  $\pm$  9.02 ms per loop — Result:- 1.7182818283253474

#### 2.3.2 Numpy Implementation

Time taken:- 83.2 ms  $\pm$  514 µs per loop — Result:- 1.718281828459046

#### 2.3.3 Python Implementation

Time taken:- 1.79 s  $\pm$  50.1 ms per loop — Result:- 1.7182818283253474

### 2.4 f(x) = 1/x from 1 to 2

### 2.4.1 Cython Implementation

Time taken:- 14.9 ms  $\pm$  338  $\mu s$  per loop — Result:- 0.6931471804472792

#### 2.4.2 Numpy Implementation

Time taken:- 92.5 ms  $\pm$  2.94 ms per loop — Result:- 0.6931471805599465

#### 2.4.3 Python Implementation

Time taken:- 1.44 s  $\pm$  87.5 ms per loop — Result:- 0.6931471804472792

### 3 Conclusion

Numpy is much more accurate than the other approaches and is faster in most cases. Cython seems to be the fastest for polynomial functions. To measure the time, I had to use a wrapper function for the Cython implementation otherwise timeit was throwing an error.