# CS319 Week 06: Experimental Algorithm Analysis (in Python)

In this notebook we will

- revise using 'numpy' nad numpy arrays;
- do a little plotting with matplotlib.
- I'll assume you know about f-strings

#### numpy

numpy (pronoucned "/numb-pee/" or "/numb-pie/") is the primary module for working with arrays of data in Python.

We use the word array in the same sense it is used in C++: it is a collection of items all of the same data type. Working with numpy arrays is MUCH faster than with lists.

There are two reasons why

- 1. It is easy to locate any element in memory.
- 2. Most of the data types and functions are actually implemented in C++, rather than Python.

#### Loading the module

To use numpy you must import the module. By convention, this is done as:

```
In [1]: import numpy as np
```

There are lots of ways of creating an numpy array. For example, to make an array from a list, use the np.array() function:

```
In [2]: N=np.array([4, 8, 16, 32, 64, 128, 256, 512])
type(N)
```

Out[2]: numpy.ndarray

```
In [3]: T=np.array([8.9400e-03, 2.2367e-03, 5.5930e-04, 1.3983e-04, 3.4958e-05, 8.7396e-06, 2
    print(f"Contents of T: {T}")
    print(f"T is of type{type(T)}")
```

```
Contents of T: [8.9400e-03 2.2367e-03 5.5930e-04 1.3983e-04 3.4958e-05 8.7396e-06 2.1849e-06 5.4622e-07]
T is of type<class 'numpy.ndarray'>
```

13 of type-ctass flumpy.fluarray

Use the dtype method to check what the underlying type of the data is:

```
In [4]: print(f"Data in T is of type {T.dtype}")
print(f"Data in N is of type {N.dtype}")
```

```
Data in T is of type float64
Data in N is of type int64
```

## Functions of numpy arrays

One of the many benefits of using numpy is that we can avoid writing explicit loops, since numpy comes with functions that map arrays to other arrays.

Example:

```
In [5]: print(N)
print(np.log2(N))

[ 4 8 16 32 64 128 256 512]
[2. 3. 4. 5. 6. 7. 8. 9.]
```

Also, standard operators such as + and \* can be used with these arrays:

### Analysing the Quadrature Data

We'll need the matplotlib library, as well as numpy

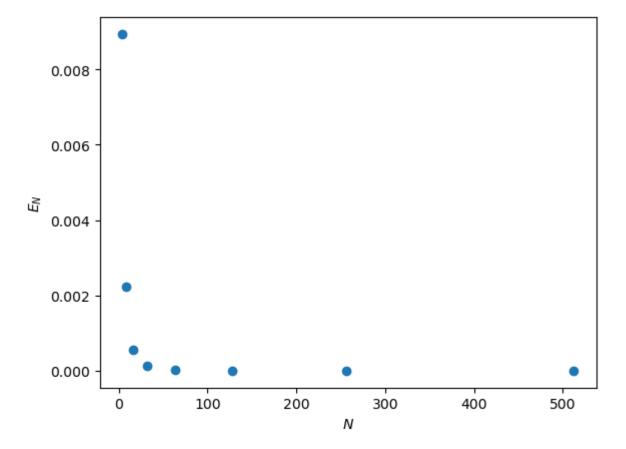
```
In [8]: import numpy as np
import matplotlib.pyplot as plt
```

We copy some data computed by 00CheckConvergence.cpp

- N is the set of numbers of intervals used in the calculations
- T is the set of values of E\_N for the Trapezium Rule.

```
In [9]: N=np.array([4, 8, 16, 32, 64, 128, 256, 512])
T=np.array([8.940076e-03, 2.236764e-03, 5.593001e-04, 1.398319e-04, 3.495839e-05, 8.7]
Plot the data, using matplotlib
In [10]: nlt.nlot(N T. 'o')
```

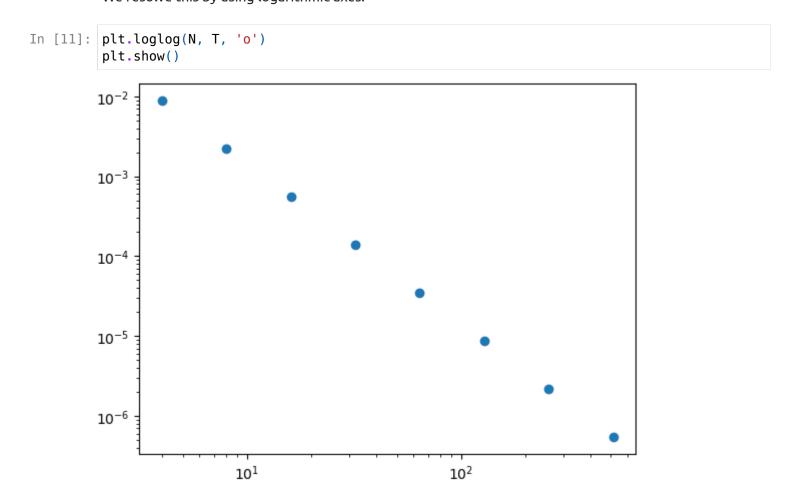
```
In [10]: plt.plot(N, T, 'o')
  plt.xlabel('$N$')
  plt.ylabel('$E_N$')
  plt.show()
```



That last figure is not very informative...

- on the horizontal axes, the data are too spread out
- ullet on the vertical axis the data is too compressed near T=0.

We resolve this by using logarithmic axes:



Looks like a straight line!

As discussed in class,  $E_N pprox CN^{-q}.$  Then, if we set

```
• Y = log(T),
```

- X = log(N), and
- K = log(C),

we get Y pprox K - qT. We have Y and X, so we want to estimate K and q, which are the slope and Y-intercept of the line.

We'll use the (depreciated) np.polyfit() function to compute the coefficients of the line that best fits (in a least squares sense) the points (X,Y).

If we set A=polyfit(x,y,n) then A is a np.array with the coefficients of the polynomial of degree n that best approximates the points (x,y). That is  $y \approx A[0]x^n + A[1]x^{n-1} + \cdots + A[-1]$ .

```
In [12]: X = np.log(N); Y = np.log(T)
Fit = np.polyfit( X, Y,1)
q = -Fit[0];
K = Fit[1];
print(f'We get K={K} and q={q}');
```

We get K=-1.9443306386200654 and q=1.9998490653205538

We can now recover the value of  ${\cal C}$ 

```
In [13]: C = np.exp(K)
print(f'C={C : .3f} and q={q : .3f}');
C= 0.143 and q= 2.000
```

Let's plot to check:

```
In [14]: plt.loglog(N, T, 'o', N, C*N**(-q), '--')
plt.show()
```

