Introduction

numpy is much faster than native Python code relying on loops. The sped up factor can be up to 10 or 100.

Example

Let's compare the running speed of native Python and numpy . The defined function perform the following operations:

- create a list (or array) fill with 1, of length n
- sum the values of this data container

Processing math: 100%

Let's compare running times:

```
In [2]: %timeit sum_python(10000)

299 \( \mu \times \time
```

Processing math: 100%

Caution

Every code comes with a constant overhead run time.

The use of numpy arrays comes with a pretty large overhead that is balanced only for large number of elements

The code below makes this overhead explicit.

Timing

```
In [8]: df = pd.DataFrame(results).set_index('N')
df
```

| Out[8]: | | Python | Numpy |
|---------|-----------|--------------|----------|
| | N | | |
| | 10 | 5.945477e-07 | 0.000004 |
| | 100 | 3.047114e-06 | 0.000005 |
| | 1000 | 3.083205e-05 | 0.000006 |
| | 10000 | 3.113286e-04 | 0.000010 |
| | 100000 | 3.114604e-03 | 0.000049 |
| | 1000000 | 3.154092e-02 | 0.001097 |
| | 10000000 | 3.319341e-01 | 0.015903 |
| | 100000000 | 3.358813e+00 | 0.140852 |

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Plotting

Analysis

Regarding the numpy version:

- it is 10 to 100 times faster than the Python version, **starting from n=300**
- for n < 300, the numpy running time is constant, which denotes a large overhead of numpy operations