## Parallel Computing

- Parallel computing
- Embarrassingly parallel
- Hadoop computing framework
- Python multiprocessing



# Python Multiprocessing





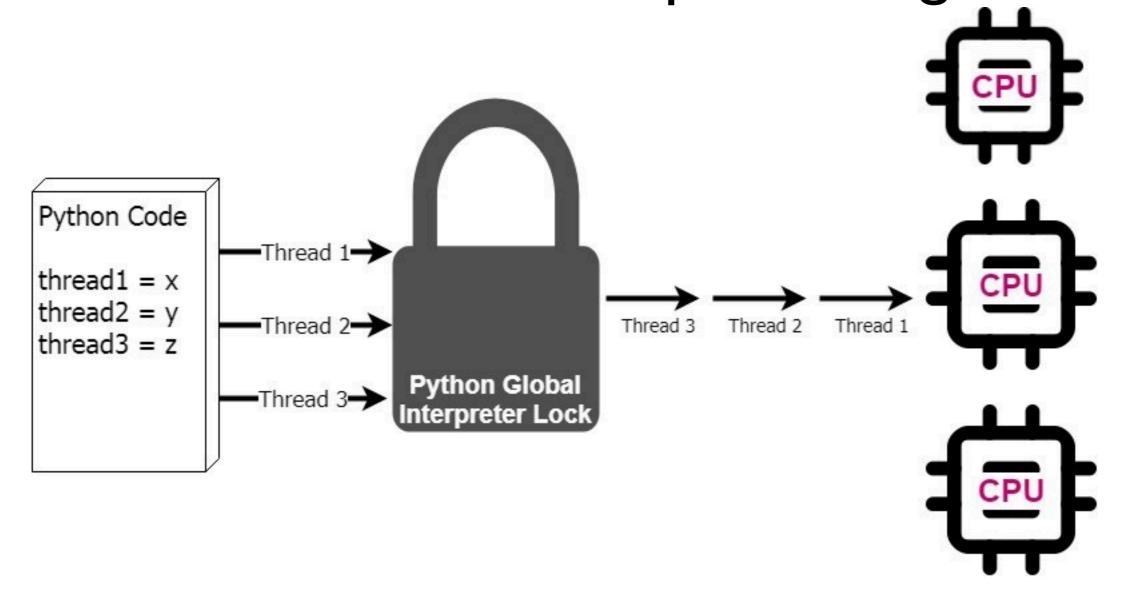
## Checking how many CPUs

```
MacBook Air
▼ Hardware
                             Hardware Overview:
    ATA
    Apple Pay
                              Model Name:
                                                     MacBook Air
    Audio
                              Model Identifier:
                                                     MacBookAir8,1
                                                     Intel Core i5
                              Processor Name:
    Bluetooth
                                                     1.6 GHz
                              Processor Speed:
    Camera
                              Number of Processors: 1
    Card Reader
                              Total Number of Cores: 2
    Controller
                              L2 Cache (per Core):
                                                     256 KB
                              L3 Cache:
                                                     4 MB
    Diagnostics
                                                     8 GB
                              Memory:
    Disc Burning
                              Boot ROM Version:
                                                     220.240.2.0.0 (iBridge: 16.16.3133.0.0,0)
    Ethernet Cards
                              Serial Number (system): C02XL7GUJK7C
    Fibre Channel
                              Hardware UUID:
                                                     E12FF4D3-EAF2-5A92-A6A0-1409E5E95ED4
    FireWire
    Graphics/Displays
    Hardware RAID
    Memory
    NVMExpress
    PCI
    Parallel SCSI
    Power
    Printers
    SAS
    SATA/SATA Express
    SPI
    Storage
    Thunderbolt
    USB
▼ Network
    Firewall
    Locations
Hardware
```

```
import multiprocessing as mp
nprocs = mp.cpu_count()
print(f"Number of CPU cores: {nprocs}")
```



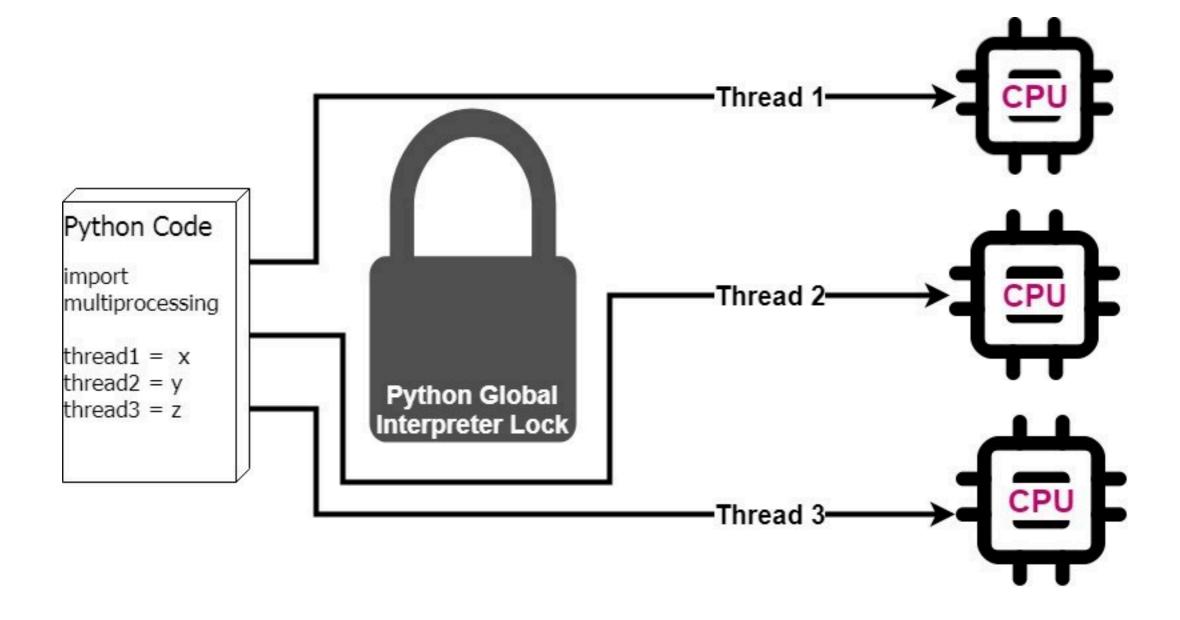
## Threads vs. Multiprocessing



https://medium.com/@urban\_institute/using-multiprocessing-to-make-python-code-faster-23ea5ef996bahttps://www.linuxjournal.com/content/multiprocessing-python



# Multiprocessing in Python





## Multiprocessing in Python

threading → multiprocessing

thread → Process

threads → processes

thread → process

### **Pool**

A class representing a pool of worker processes

- **e.g.** pool = Pool(processes=4)
  - pool has 4 worker processes

#### Tasks can be offloaded to these workers

- e.g. result = pool.map(f, data)
  - data is an array or list
  - function f will be applied to each member of data in turn



## Simple Pool Example

```
from multiprocessing import Pool
from math import sqrt

newPool = Pool(processes=2) # initialize the Pool.
res = newPool.map(sqrt,[4,5,6,7])
res
Out[22]:
[2.0, 2.23606797749979, 2.449489742783178, 2.6457513110645907]
```



### pool\_process

```
import time
import math
from multiprocessing import Pool
# A function for timing a job that uses a pool of processes.
# f is a function that takes a single argument
# data is an array of arguments on which f will be mapped
# pool size is the number of processes in the pool.
def pool_process(f, data, pool_size):
   tpl = time.time()
    pool = Pool(processes=pool_size) # initialize the Pool.
    result = pool.map(f, data) # map f to the data using the Pool
    pool.close() # No more processes
    pool.join() # Wait for the pool processing to complete.
    print("Results", result)
    print("Overall Time:", int(time.time()-tpl))
```

```
def my_func(x):
    return math.sqrt(x)

dataRange = range(20)
```

pool\_process(my\_func, dataRange, 2)



```
# This verbose version shows which process in the pool is running each task.
def my_func_verbose(x):
    s = math.sqrt(x)
    print("Task", multiprocessing.current_process(), x, s)
    return s
```

```
dataRange = range(10)
pool_process(my_func_verbose, dataRange)
Task <ForkProcess(ForkPoolWorker-7, started daemon)> 0 0.0
Task <ForkProcess(ForkPoolWorker-7, started daemon)> 1 1.0
Task <ForkProcess(ForkPoolWorker-7, started daemon)> 2 1.4142135623730951
Task <ForkProcess(ForkPoolWorker-7, started daemon) > 3 1.7320508075688772
Task <ForkProcess(ForkPoolWorker-7, started daemon)> 4 2.0
Task <ForkProcess(ForkPoolWorker-7, started daemon) > 5 2.23606797749979
Task <ForkProcess(ForkPoolWorker-7, started daemon) > 6 2.449489742783178
Task <ForkProcess(ForkPoolWorker-7, started daemon) > 7 2.6457513110645907
Task <ForkProcess(ForkPoolWorker-7, started daemon) > 8 2.8284271247461903
Task <ForkProcess(ForkPoolWorker-7/ started daemon)> 9 3.0
Results [0.0, 1.0, 1.4142135623730951, 1.7320508075688772, 2.0, 2.2360679
7749979, 2.449489742783178, 2.6457513110645907, 2.8284271247461903, 3.0]
Overall Time: 0
```

### 1 process, same as running in series in a loop



```
dataRange = range(10)
pool_process(my_func_verbose, dataRange
Task <ForkProcess(ForkPoolWorker 9) started daemon)> 2 1.4142135623730951
Task <ForkProcess(ForkPoolWorker-8, started daemon)> 0 0.0
Task <ForkProcess(ForkPoolWorker-8, started daemon)> 1 1.0
Task <ForkProcess(ForkPoolWorker-9, started daemon) > 3 1.7320508075688772
Task <ForkProcess(ForkPoolWorker-8, started daemon)> 4 2.0
Task <ForkProcess(ForkPoolWorker-9, started daemon)> 6 2.449489742783178
Task <ForkProcess(ForkPoolWorker-9, started daemon) > 7 2.6457513110645907
Task <ForkProcess(ForkPoolWorker-8, started daemon) > 5 2.23606797749979
Task <ForkProcess(ForkPoolWorker-8, started daemon) > 8 2.8284271247461903
Task <ForkProcess(ForkPoolWorker 8 started daemon) > 9 3.0
Results [0.0, 1.0, 1.4142135623739951, 1.7320508075688772, 2.0, 2.2360679
7749979, 2.449489742783178, 2.6457513110645907, 2.8284271247461903, 3.0]
Overall Time: 0
```

### 2 processes

Allows both cores to be used in parallel Asynchronous behaviour



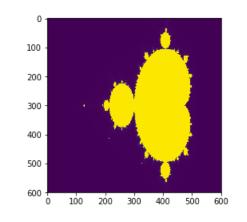
```
dataRange = range(10)
pool_process(my_func_verbose, datarange(10)

Task <ForkProcess(ForkPoolWorker-12, started daemon)> 2 1.4142135623730951
Task <ForkProcess(ForkPoolWorker-10, started daemon)> 0 0.0
Task <ForkProcess(ForkPoolWorker-15, started daemon)> 5 2.23606797749979
Task <ForkProcess(ForkPoolWorker-16, started daemon)> 6 2.449489742783178
Task <ForkProcess(ForkPoolWorker-14, started daemon)> 4 2.0
Task <ForkProcess(ForkPoolWorker-11, started daemon)> 1 1.0
Task <ForkProcess(ForkPoolWorker-18, started daemon)> 8 2.8284271247461903
Task <ForkProcess(ForkPoolWorker-13, started daemon)> 3 1.7320508075688772
Task <ForkProcess(ForkPoolWorker-17, started daemon)> 7 2.6457513110645907
Task <ForkProcess(ForkPoolWorker-19, started daemon)> 9 3.0
Results [0.0, 1.0, 1.41421356237.0911, 1.7320508075688772, 2.0, 2.23606797749979, 2.449489742
783178, 2.6457513110645907, 2.8284771247461903, 3.0]
Overall Time: 0
```

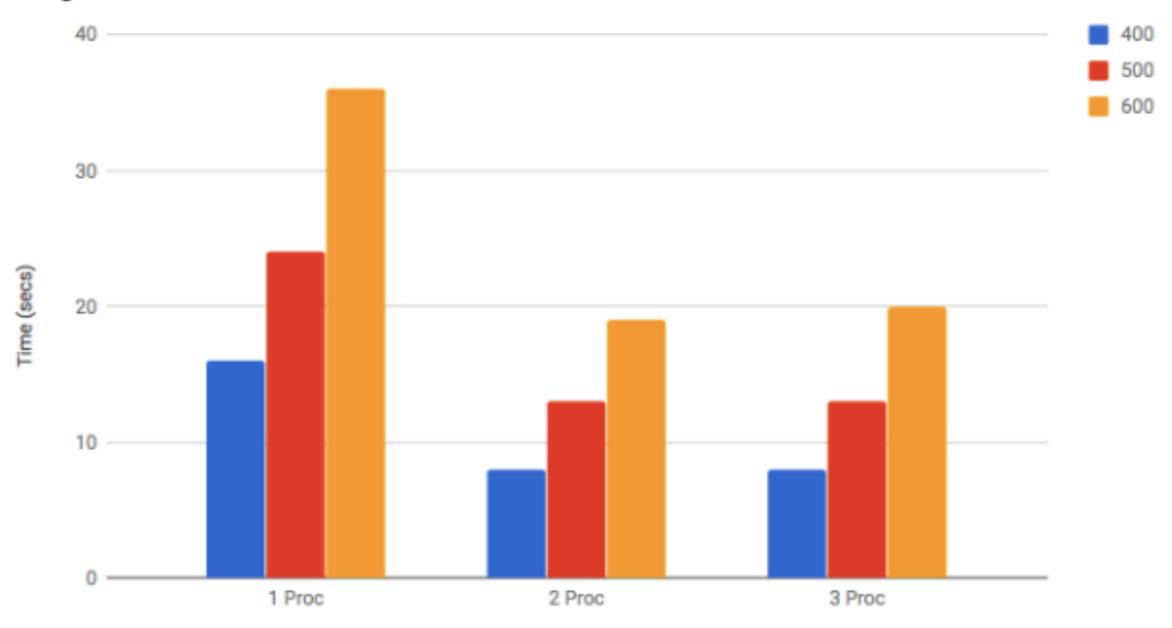
### 10 processes but 2 cores



## Testing with Mandelbrot Images



#### Image Size & No of Processes



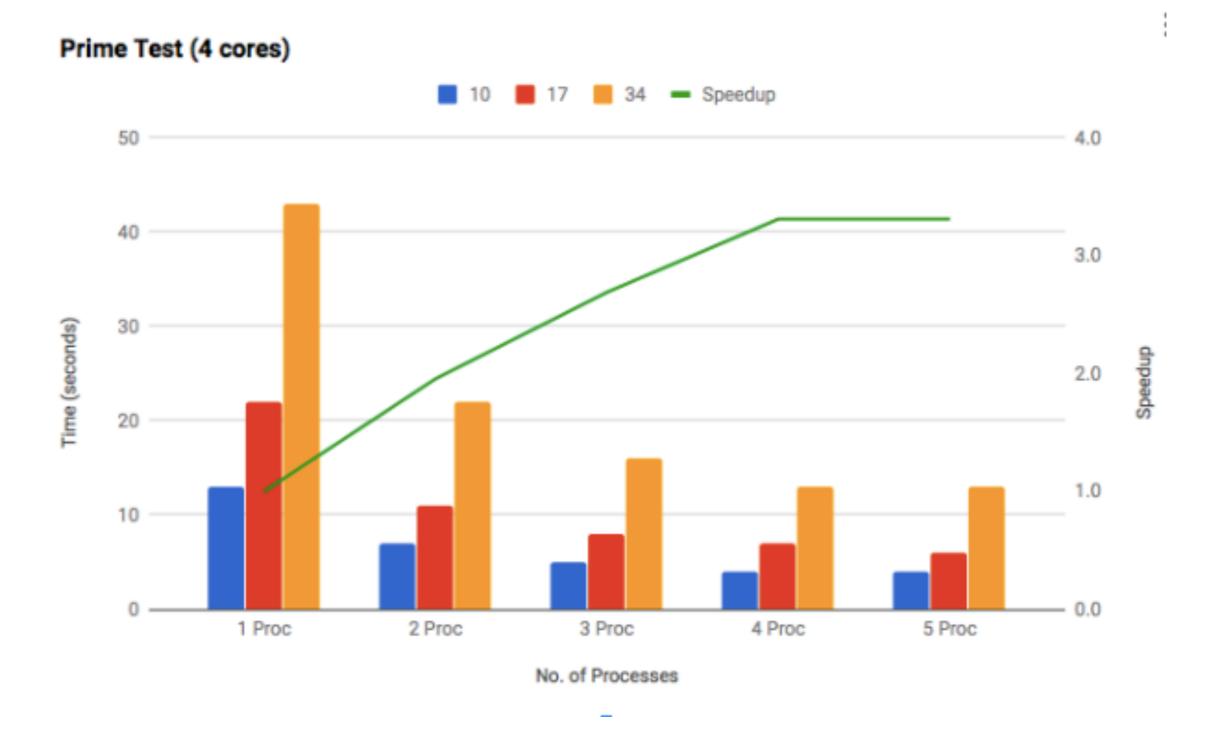
Number of Processes

 $\underline{https://timothyawiseman.wordpress.com/2012/12/21/a-really-simple-multiprocessing-python-example/}$ 





### 4 Core CPU





# Assignment I

#### Multiprocessing

Objective

The objective of this exercise is to evaluate speedup derived from using multiple CPU cores through the multiprocessing facility in Python.

#### What is provided?

A Python Notebook (MultiprocessingCore) is available on the Moodle page that provides a basic multiprocessing framework. The Notebook contains code to set up a multiprocessing Pool and use that Pool to process tasks. If multiple cores are available the Pool can use those cores to speed up processing. The function in the sample code is a simple square-root function so it is not possible to see the impact of using multiple cores.

A naive function for checking primes is also provided. If this is used to check large numbers (8 digit) it takes time and the speedup from using multiple cores will be evident.

There are plenty of primes to be found here: https://primes.utm.edu/lists/small/millions/



# Further reading

https://www.linuxjournal.com/content/multiprocessing-python

https://www.quantstart.com/articles/Parallelising-Python-with-Threading-and-Multiprocessing

https://www.youtube.com/watch? v=Lu5LrKh1Zno&list=PLeo1K3hjS3uub3PRhdoCTY8BxMKSW7Rj N&index=2

https://medium.com/@urban\_institute/using-multiprocessing-to-make-python-code-faster-23ea5ef996ba

