

*“When you come to a fork in the road, take it”*

**Yogi Bera, 1925**

# Hands-on H.1: Python Multiprocessing

**CS205: Computing Foundations for Computational Science**  
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**Spring Term 2020**



**INSTITUTE FOR APPLIED  
COMPUTATIONAL SCIENCE**  
AT HARVARD UNIVERSITY



**HARVARD**  
School of Engineering  
and Applied Sciences

**Lectures developed by Dr. Ignacio Illorente**



# Before We Start

## Where We Are

Computing Foundations for Computational and Data Science

How to use modern computing platforms in solving scientific problems

Intro: Large-Scale Computational and Data Science

A. Parallel Processing Fundamentals

B. Parallel Computing

B.1. Foundations of Parallel Computing

B.2. Performance Optimization

B.3. Accelerated Computing

B.4. Shared-memory Parallel Processing

B.5. Distributed-memory Parallel Processing

C. Parallel Data Processing

Wrap-Up: Advanced Topics

# CS205: Contents

## APPLICATION SOFTWARE

A.3 APPLICATION PARALLELISM

A.4. PARALLEL PROGRAM DESIGN

## PROGRAMMING MODEL

Optimization

OpenACC

Python MP

OpenMP

MPI

Spark

Map-Reduce

B. BIG COMPUTE

PLATFORM

C. BIG DATA



A.2. LARGE-SCALE PROCESSING ON CLOUD



Open Nebula



FASRC



ODYSSEY  
HARVARD FAS  
RESEARCH COMPUTING



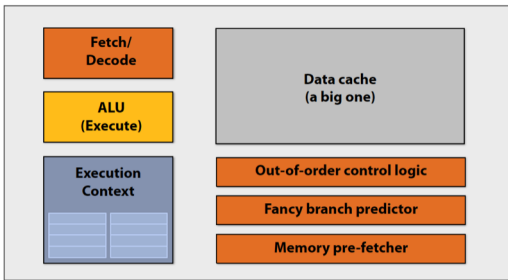
FASRC

A.1 PARALLEL ARCHITECTURES

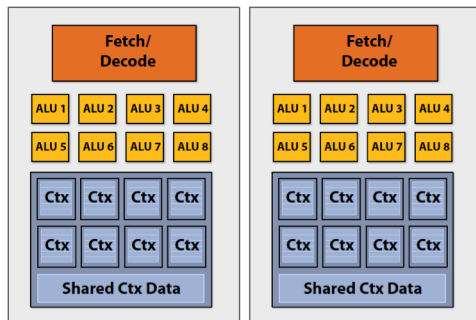
# Context

## Python Multiprocessing

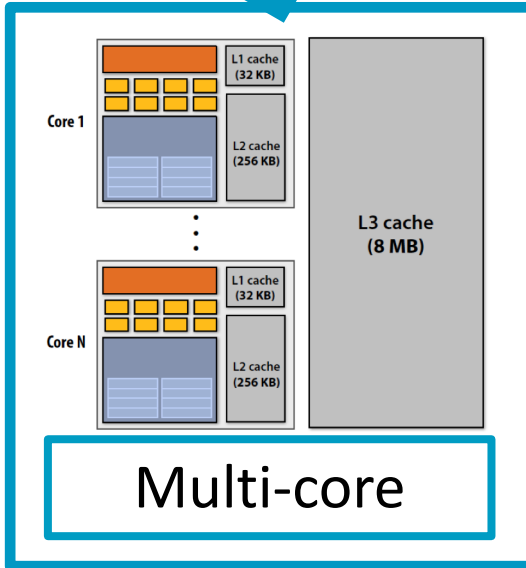
How to develop code that can make effective use of existing parallelism at different levels?



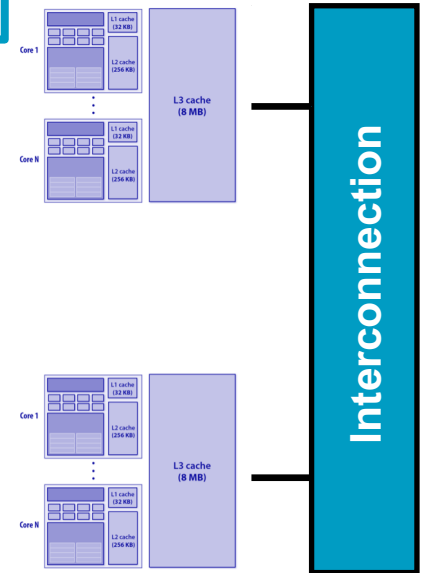
ILP/Data



Many-core



Multi-core



Multi-node

# Hands-on Examples

## Requirements

1. Unix-like shell (Linux, Mac OS or AWS VM)
2. Python installed

# Roadmap

## Python Multiprocessing

Multi-processing Basics

Process Creation and Synchronization

Process Communication

Process Synchronization

Work Distribution

Examples

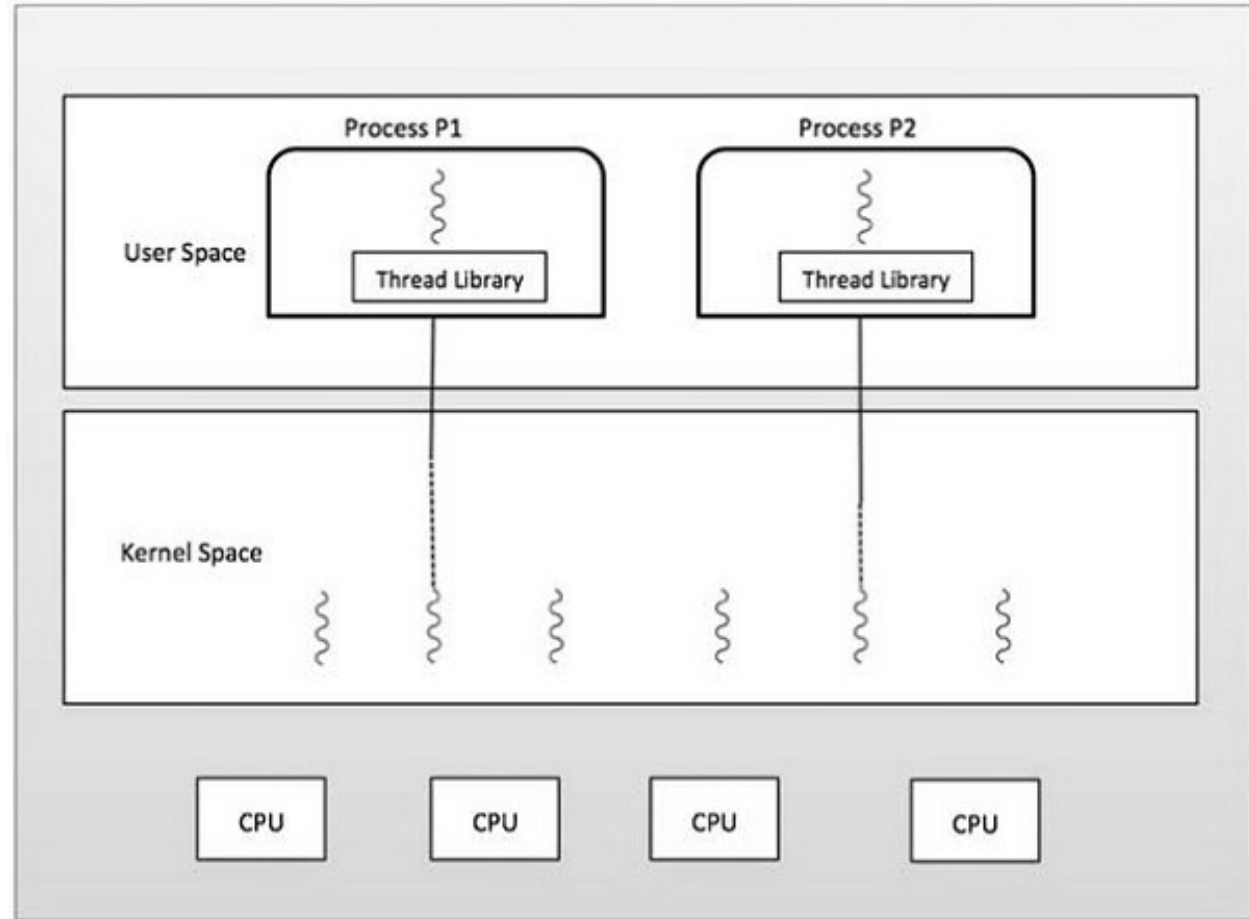
# Multi-processing Basics

## What is a Process?

- A process is an instance of a computer program that is being executed

- A process can have 1 or several threads (1 in this hands-on)

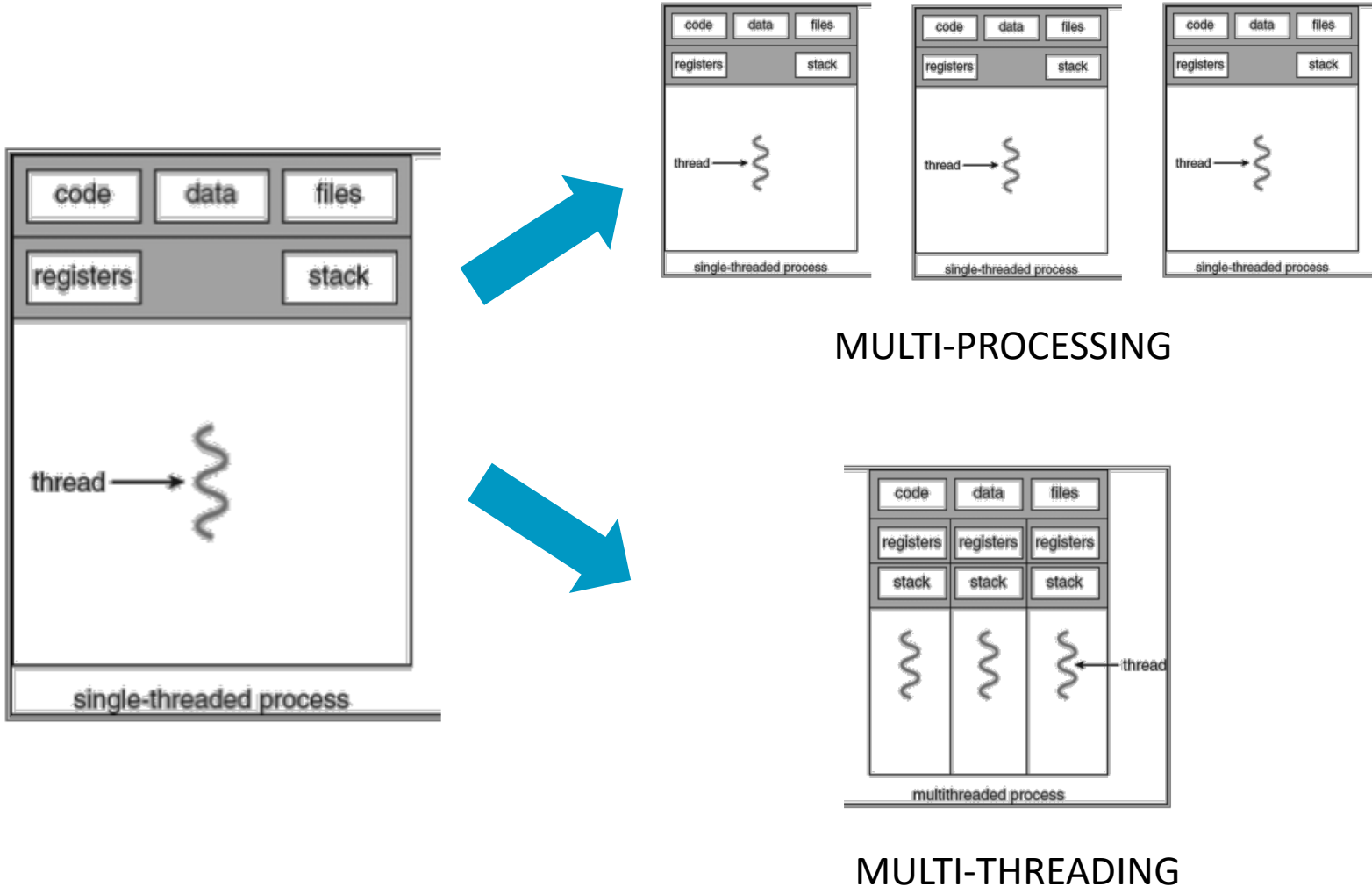
- The kernel of the OS schedule threads to multiple cores





# Multi-processing Basics

## Multi-Processing vs Multi-Threading



# Multi-processing Basics

## Multi-Processing vs Multi-Threading in Python

The CPython implementation (called CPI, C Python Interpreter) is not thread-safe and only permits a single thread to run at a time

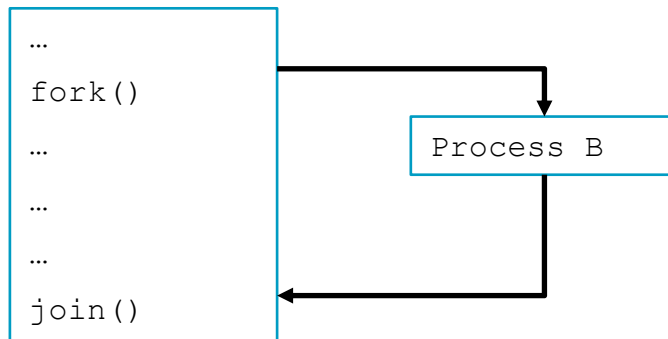
The standard Python library has two main modules for parallel computing:

- `threading`
  - Good for I/O bound tasks
  - Subject to the Global Interpreter Lock (GIL)
- `multiprocessing`
  - Circumvents the GIL
  - Useful for parallel computation
- We will focus on the multiprocessing module

# Multi-processing Basics

## Elements of Programming

- Memory Isolation
  - ✓ Processes do NOT share memory address space
- Fork/Join Execution Model
  - ✓ Fundamental way of expressing concurrency within a computation
  - ✓ Fork creates a new child process
  - ✓ Parent continues after the **Fork** operation
  - ✓ Child begins operation separate from the parent
  - ✓ Parent waits until child **joins** (continues afterwards)

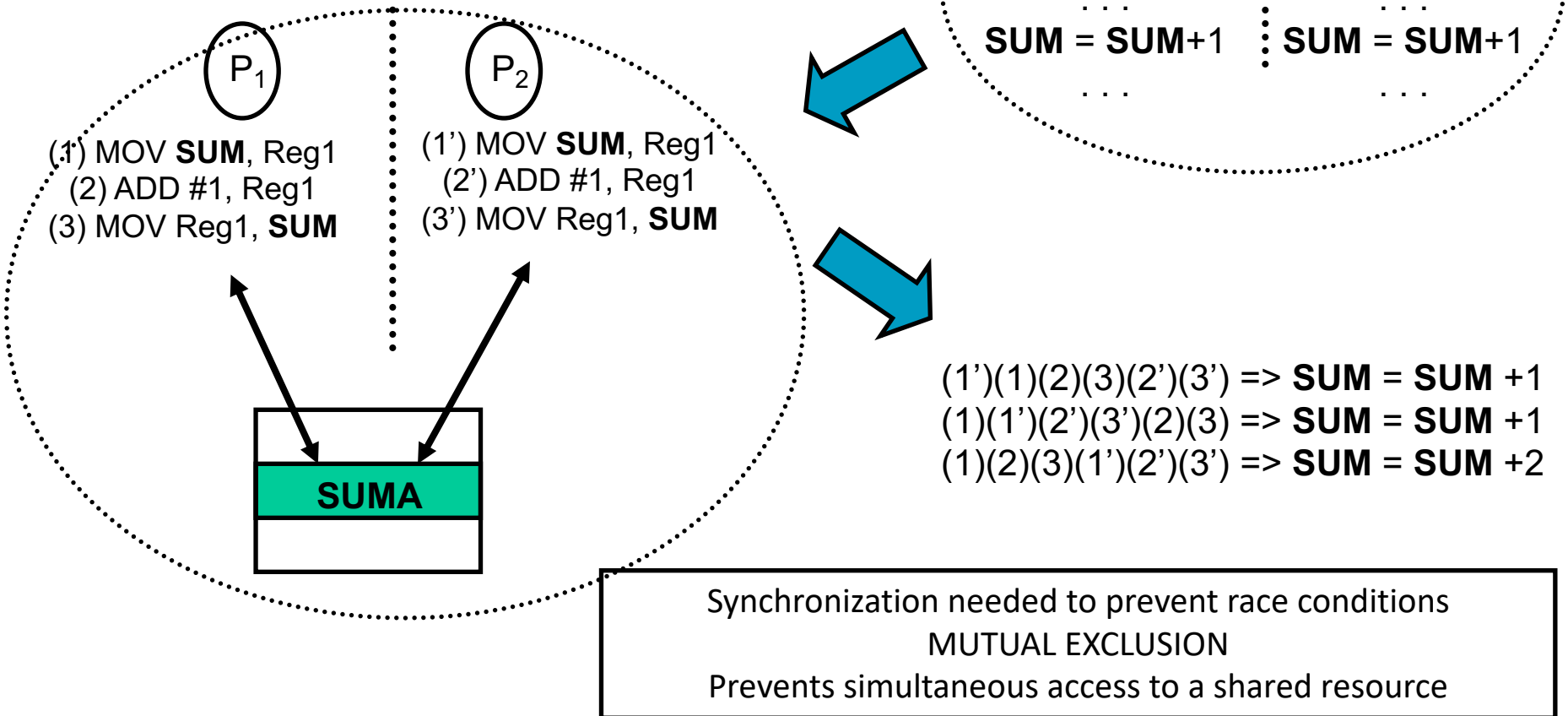


# Multi-processing Basics

## Race Conditions

A Race Condition occurs, if

- Two or more processes manipulate a shared resource concurrently, and
- The outcome of the execution depends on the particular order in which the access takes place



# Multi-processing Basics

## Synchronization

**Variable mutex: S**

*Boolean: 0 / 1*

*General: Integer  $\geq 0$*

**Functions:**

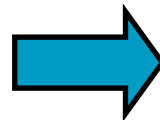
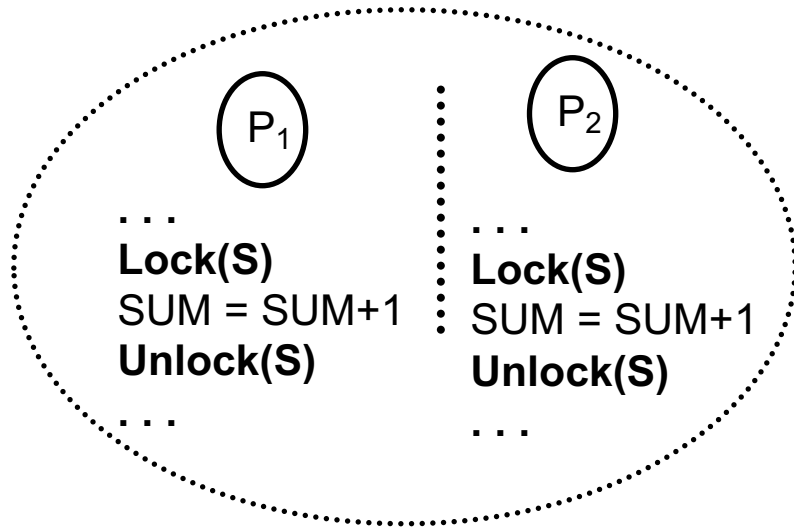
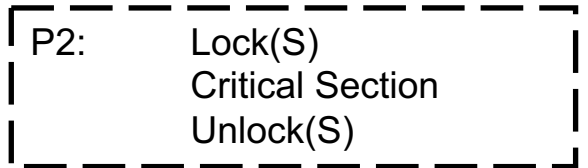
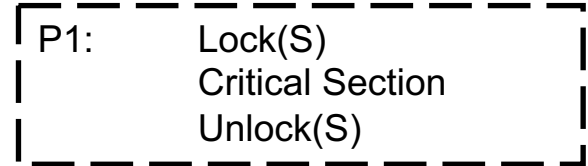
*Lock(S)*

If  $S == 0$  then wait to  $S > 0$

If  $S > 1$  then  $S = S - 1$

*Unlock(S):*

$S = S + 1$



(1)(2)(3)(1')(2')(3') => **SUM = SUM + 2**

(1')(2')(3')(1)(2)(3) => **SUM = SUM + 2**

# Process Creation and Synchronization

## Join/Fork Model

- The `multiprocessing` module includes a very simple and intuitive API for dividing work between multiple processes

```
import multiprocessing
```

```
def print_cube(num):
```

```
    ...
```

```
def print_square(num):
```

```
    ...
```

```
if __name__ == "__main__":
```

```
    # creating processes
```

```
    p1 = multiprocessing.Process(target=print_square, args=(10, ))
```

```
    p2 = multiprocessing.Process(target=print_cube, args=(10, ))
```

```
    # starting process 1 and 2
```

```
    p1.start()
```

```
    p2.start()
```

```
    # wait until process 1 and 2 are finished
```

```
    p1.join()
```

```
    p2.join()
```

```
    # both processes finished
```

```
    print("Done!")
```

**example1.py**

Creates a process structure to execute a target function with args

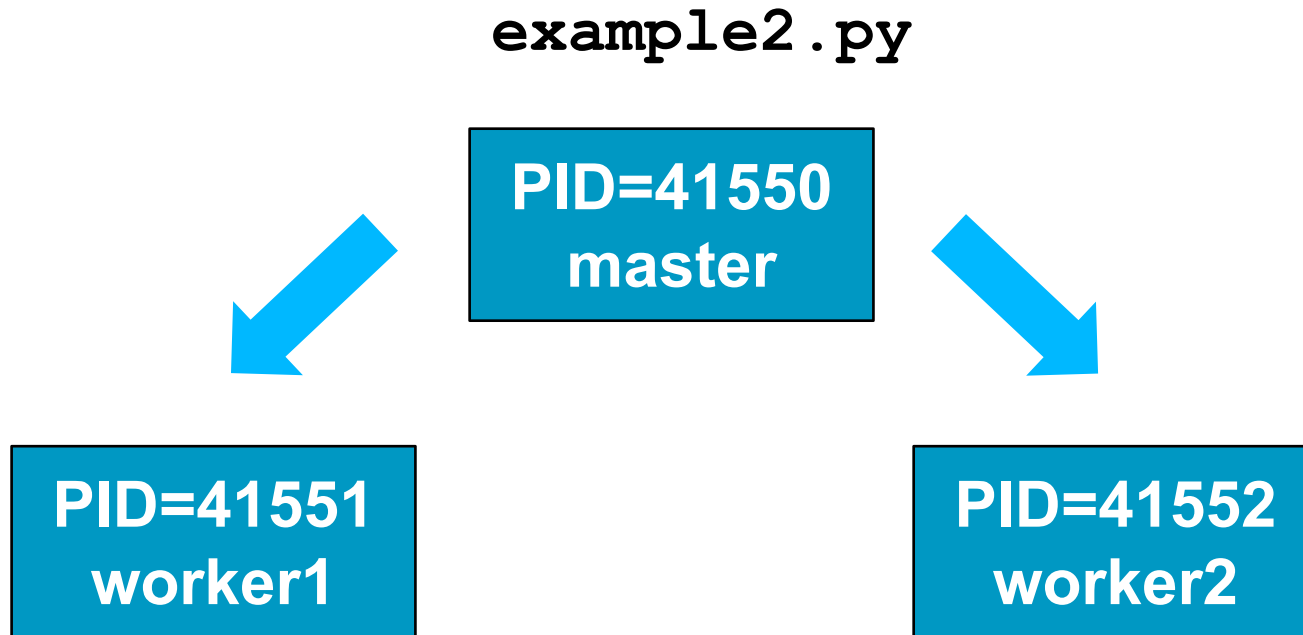
Start a process

Wait for process termination

# Process Creation and Synchronization

## Join/Fork Model

- Each process is completely independent



# Process Communication

## Shared Memory

- Each process runs independently and has its own memory space

```
import multiprocessing
# empty list with global scope
result = []
def square_list(mylist):
    global result
    # append squares of mylist to global list result
    for num in mylist:
        result.append(num * num)
    # print global list result
    print("Result(in process p1): {}".format(result))
if __name__ == "__main__":
    # input list
    mylist = [1,2,3,4]
    p1 = multiprocessing.Process(target=square_list, args=(mylist,))
    p1.start()
    p1.join()
    # print global result list
    print("Result(in main program): {}".format(result))
```

**example3.py**



# Process Communication

## Shared Memory

multiprocessing module provides **Array** and **Value** objects to share data between processes.

- **Array**: a ctypes array allocated from shared memory.
- **Value**: a ctypes object allocated from shared memory.

```
import multiprocessing

def square_list(mylist, result, square_sum):
    ...

if __name__ == "__main__":
    ...

    # creating Array of int data type with space for 4 integers
    result = multiprocessing.Array('i', 4)
    # creating Value of int data type
    square_sum = multiprocessing.Value('i')
    # creating new process
    p1 = multiprocessing.Process(target=square_list, args=(mylist, result, square_sum))
```

**example4.py**

Shared variables passed as arguments

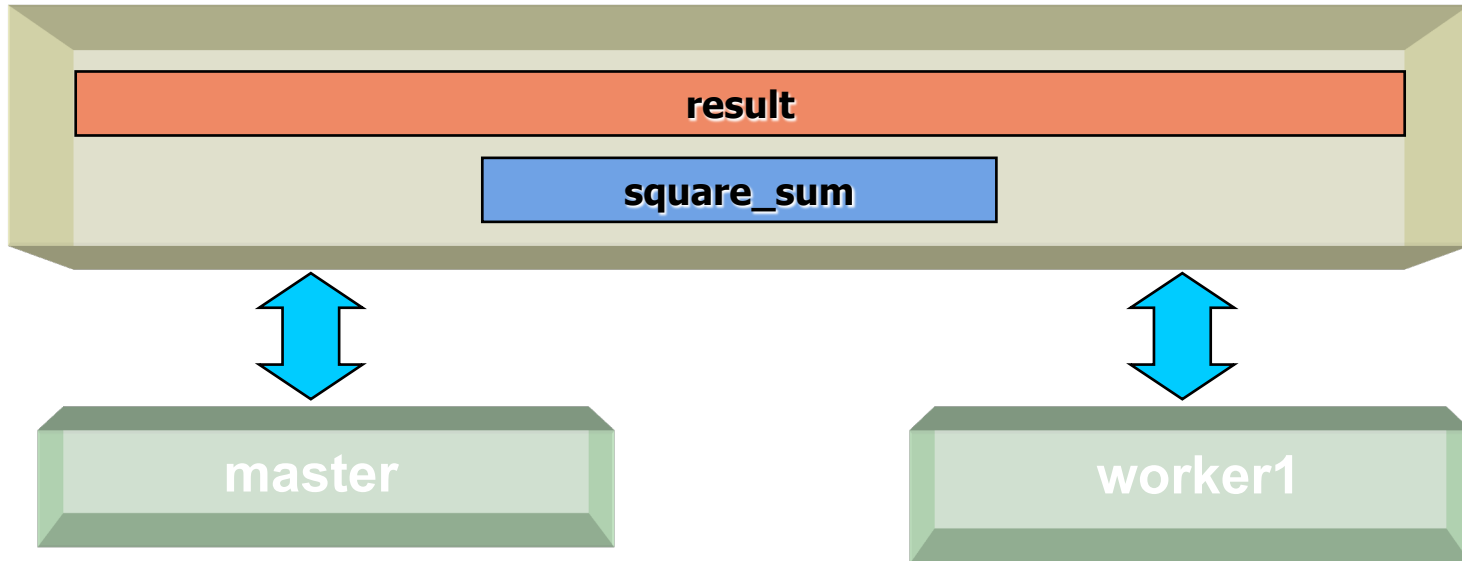
Array integers of size 4

Single variable type integers

# Process Communication

## Shared Memory

Most efficient way to share memory across processes



`multiprocessing` module provides `manager` class (**Advanced!**) that

- Share arbitrary object types like lists, dictionaries, Queue, Array, etc.
- A single manager can be shared by processes on different computers
- However, they are slower than using shared memory.

# Process Communication

## Queues

- A simple way to communicate between process with multiprocessing is to use a Queue to pass messages back and forth. Any Python object can pass through a Queue.

```
def square_list(mylist, q):
    # append squares of mylist to queue
    for num in mylist:
        q.put(num * num)
def print_queue(q):
    print("Queue elements:")
    while not q.empty():
        print(q.get())
    print("Queue is now empty!")
if __name__ == "__main__":
    ...
    # creating multiprocessing Queue
    q = multiprocessing.queue()
    # creating new processes
    p1 = multiprocessing.Process(target=square_list, args=(mylist, q))
    p2 = multiprocessing.Process(target=print_queue, args=(q,))
```

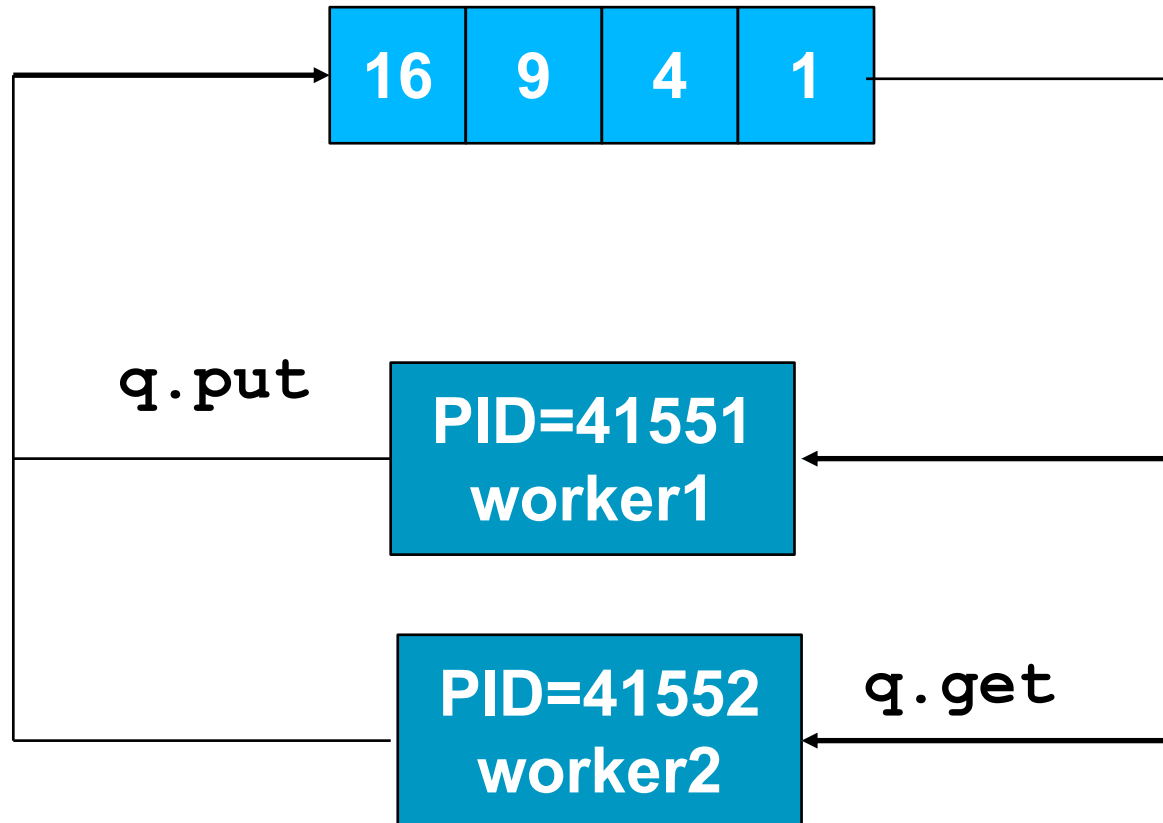
**example5.py**

Put value in queue

Read values from queue

# Process Communication

## Queues



A more efficient two-way communication can be performed with the `pipe` class  
**(Advanced!)**

# Process Synchronization

## Race Conditions

- Simultaneous access to shared variable: Why is not 100 the final value?

```
# function to withdraw from account
def withdraw(balance):
    for _ in range(10000):
        balance.value = balance.value - 1

# function to deposit to account
def deposit(balance):
    for _ in range(10000):
        balance.value = balance.value + 1

def perform_transactions():
    # initial balance (in shared memory)
    balance = multiprocessing.Value('i', 100)
    # creating new processes
    p1 = multiprocessing.Process(target=withdraw, args=(balance,))
    p2 = multiprocessing.Process(target=deposit, args=(balance,))
```

**example6.py**

# Process Synchronization

## Locks

- `multiprocessing` module provides a `Lock` class to deal with the race conditions. `Lock` is implemented using a Semaphore object provided by the Operating System

```
# function to withdraw from account
def withdraw(balance, lock):
    for _ in range(10000):
        lock.acquire()
        balance.value = balance.value - 1
        lock.release()
...
...

# initial balance (in shared memory)
balance = multiprocessing.Value('i', 100)
# creating a lock object
lock = multiprocessing.Lock()
# creating new processes
p1 = multiprocessing.Process(target=withdraw, args=(balance, lock))
p2 = multiprocessing.Process(target=deposit, args=(balance, lock))
```

**example7.py**

# Work Distribution

## Pools

- Simple serial program to calculate squares of elements of a given list

```
# Python program to find squares of numbers in a given list
```

```
def square(n):
```

```
    return (n*n)
```

```
if __name__ == "__main__":
```

```
    # input list
```

```
    mylist = [1,2,3,4,5]
```

```
    # empty list to store result
```

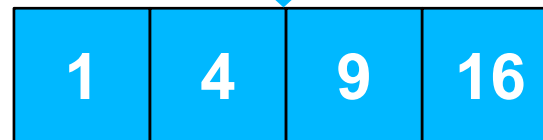
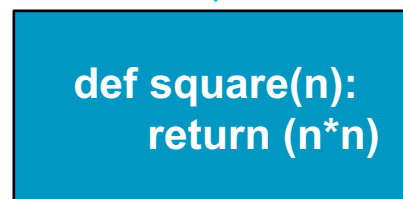
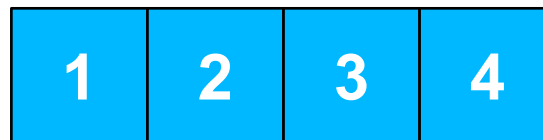
```
    result = []
```

```
    for num in mylist:
```

```
        result.append(square(num))
```

```
    print(result)
```

**example8.py**



# Work Distribution

## Pools

- multiprocessing module provides a `Pool` class that represents a pool of worker processes. It has methods which allows tasks to be offloaded to the worker processes in a few different ways

```
# Python program to find squares of numbers in a given list
```

```
import multiprocessing
```

```
import os
```

```
def square(n):
```

```
    print("Id for {0}: {1}".format(n, os.getpid()))
```

```
    return (n*n)
```

```
if __name__ == "__main__":
```

```
    # input list
```

```
    mylist = [1,2,3,4,5]
```

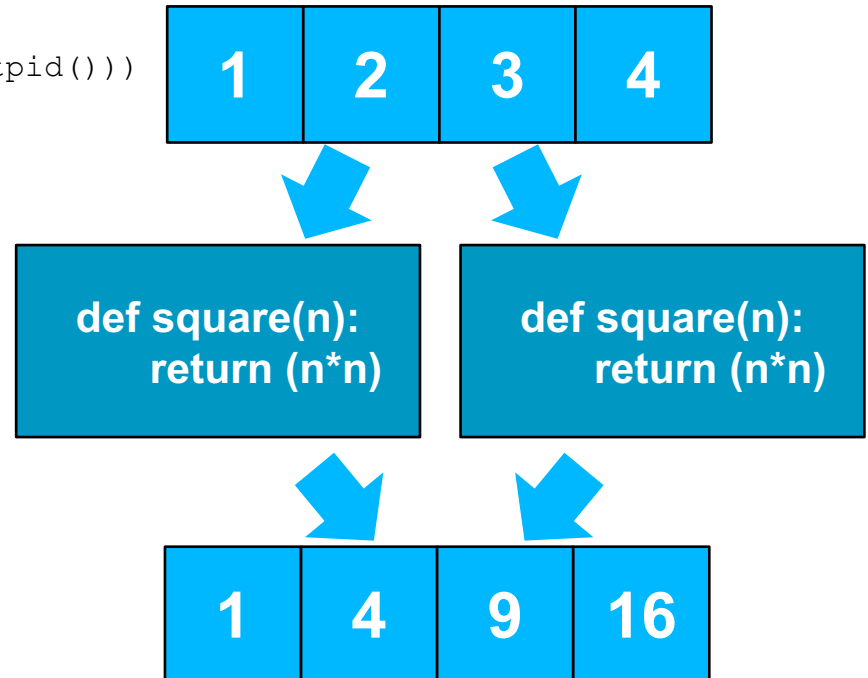
```
    # creating a pool object
```

```
    p = multiprocessing.Pool()
```

```
    # map list to target function
```

```
    result = p.map(square, mylist)
```

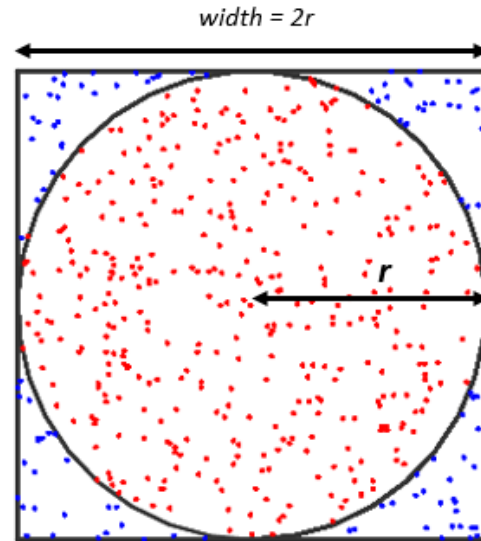
**example9.py**





# Examples

## Example: pi Using Montecarlo Method



`pi1.py`

Area of the circle is  $\pi r^2$

The area of the square is  $4r^2$

If we divide the area of circle by the area of square we get  $\pi/4$

Same ratio can be used between the number of points within the square and the number of points within the circle.

Hence we can use the following formula to estimate Pi:

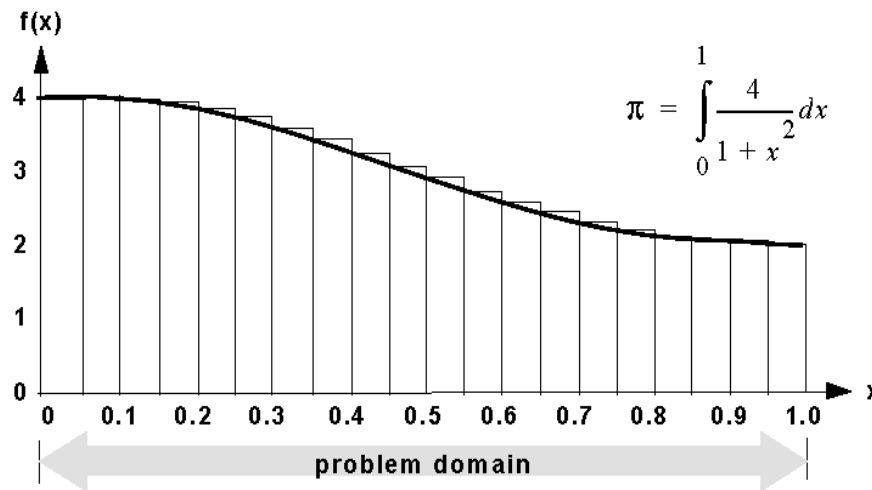
$$\pi \approx 4 \times (\text{number of points in the circle} / \text{total number of points})$$

**What is the speed-up in your system?**

# Example

## Exercise: pi Using Numerical Approximation to Integral

pi2.py



# Questions

## Python Multiprocessing



Some of the examples have been downloaded from <https://www.geeksforgeeks.org/> (Nikhil Kumar)