# High-Performance Programming

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### Afternoon Objectives

- Compare and contrast processes vs. threads
- Compare and contrast parallelism and concurrency
- Identify problems that require parallelism or concurrency
- Implement parallel and concurrent solutions
- Measure the run time of different approaches to see the benefit of threading/parallelism.

# Agenda

### Morning - AWS

#### Afternoon

- Discuss computer resources
- Talk about processes
- Talk about threads

### Motivation

- Process biggish data (≥ 5GB depending on task)
- More efficient use of CPU resources
- Saves time

### Computing Resources

- Central Processing Unit (CPU) Clock speed measured in GHz (cycles/second)
- Random Access Memory (RAM) Size measured in GB (gigabytes)
- Persistent Storage (disk) Size measured in GB (gigabytes)
- Graphics Processing Unit (GPU)

### **CPU Cores**

- A CPU can have multiple cores
- Each core is a self-contained processor that can execute programs
- GPUs have many cores

#### **Processes**

An instance of a computer program that is being executed.

Each process has its own memory, program text, filehandles, permissions, etc. and can run on any core.

A computer runs many, many processes, most just waiting.

```
$ ps aux | wc -1
233
```

### Multiprocessing in python

```
from multiprocessing import Pool
import os
# Count the number of words in a file
def word count(f):
    return len(open(f).read().split())
# Use a regular loop to count words in files
def sequential word count(lst of files):
    return sum([word count(f) for f in lst of files])
# Use a multiple cores to count words in files
def parallel word count(lst of files):
    pool = Pool(processes=4)
    results = pool.map(word count, lst of files)
    return sum(results)
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```

#### **Threads**

Each process contains one or more threads of execution

Threads are lighter-weight than processes

- faster to create
- less memory overhead
- inter-thread communication easier (shared memory)
- faster to context switch

Can we used these for parallel programming?

Python processes have a Global Interpreter Lock (GIL) that prevents multiple thread from running at once.\*

Python threads are concurrent but not parallel

Why use threads?

\*In the most common implementation

Useful when the program has to wait on resources outside of the python code

- I/O
- Database queries
- Certain libraries (e.g., image processing)

```
import threading
jobs = []
# Tnitial and start threads
for i in xrange(num threads):
    t = threading.Thread(target=target_function, args=(arg1, a
    jobs.append(t)
    t.start
# "join" will wait until the thread terminates
results = []
    t.join()
    # Access the result of the thread (if any) and append
    results.append(t.result)
```

# Summary

What?	Library	Cores	Why?
Parellelism	multiprocessing	multiple	CPU-bound problems I/O-bound problems
Concurrency	threading	single	