## High-Performance Programming

Miles Erickson

February 27, 2017

### 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)
                                        <ロト (個) (重) (重) (重) のQで
```

#### **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 use 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 implemetation

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
def target function(url, results):
   results.append(scrape text(url)) # Thread-safe
threads = [] # List to keep track of threads
results = [] # Shared list to collect results
for i in range(num_threads):
   t = threading.Thread(target=target_function,
                         args=(url, results))
   threads.append(t)
   t.start()
for t in threads:
   t.join() # Wait until each thread finishes
```

# Summary

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