# [544] Processes and Threads

Meenakshi Syamkumar

# Learning Objectives

- describe the interactions between schedulers, CPUs, threads, and address spaces
- decide for a given scenario whether to organize code as single-threaded, multi-threaded, or multi-process
- trace through different interleavings to identify race conditions

### **Motivation**

Modern CPUs have many cores (maybe dozens)

Trend: more cores rather than faster cores

Problem: a simple Python program can use at most ONE core (less if it accesses files or the Internet)

Understanding threads and processes will:

- let us write programs that fully utilize CPU resources
- decide the structure of our concurrent program (threads or processes) depending on the situation

# Outline

Virtual Address Spaces

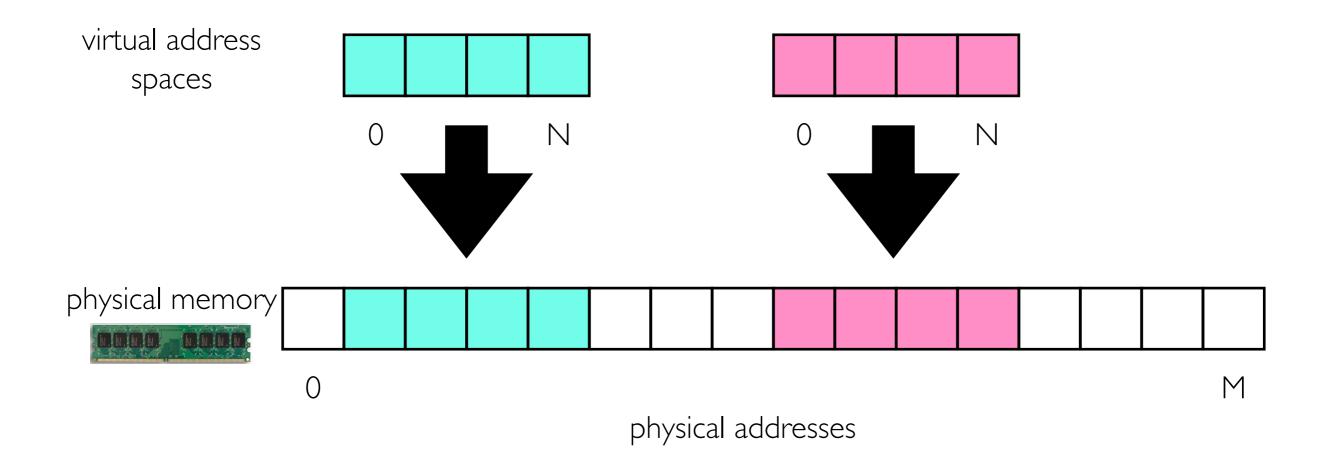
Threads

Demos and Worksheet

### Processes and Address Spaces

#### Address spaces

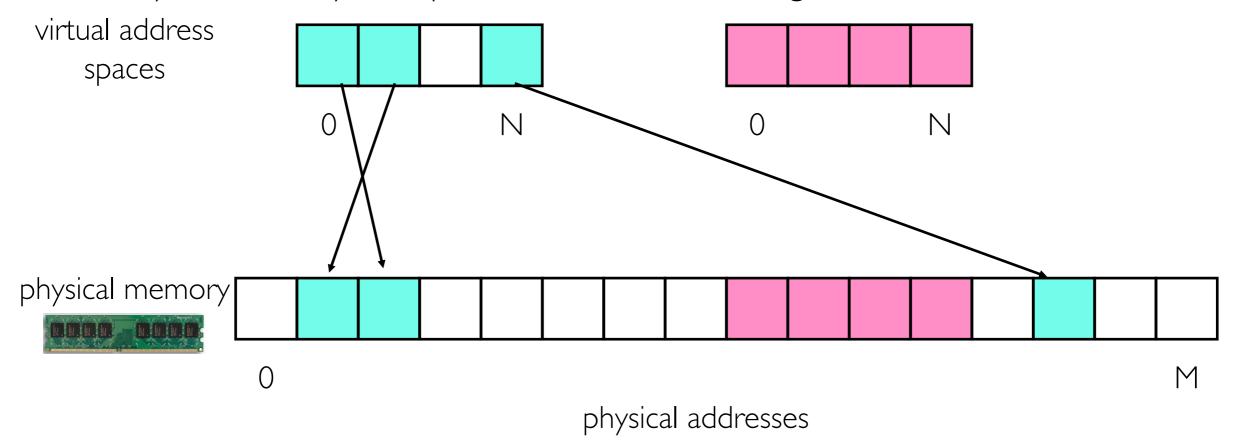
- A process is a running program
- Each process has it's own virtual address space
- The same virtual address generally refers to different memory in different processes
- Regular processes cannot directly access physical memory or other addr spaces



### Processes and Address Spaces

#### Address spaces

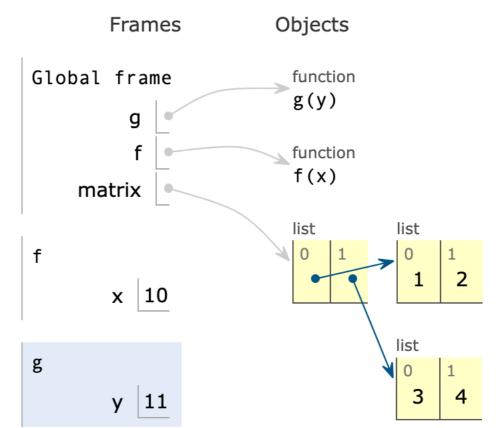
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- Each process has it's own virtual address space
- The same virtual address generally refers to different memory in different processes
- Regular processes cannot directly access physical memory or other addr spaces
- Address spaces can have holes (N is usually MUCH bigger than M)
- Physical memory for a process need not be contiguous



# What goes in an address space?

```
→ 1 def g(y):
    return y * 2

3
4 def f(x):
    return g(x+1)
6
7 matrix = [[1,2], [3,4]]
8 f(10)
```

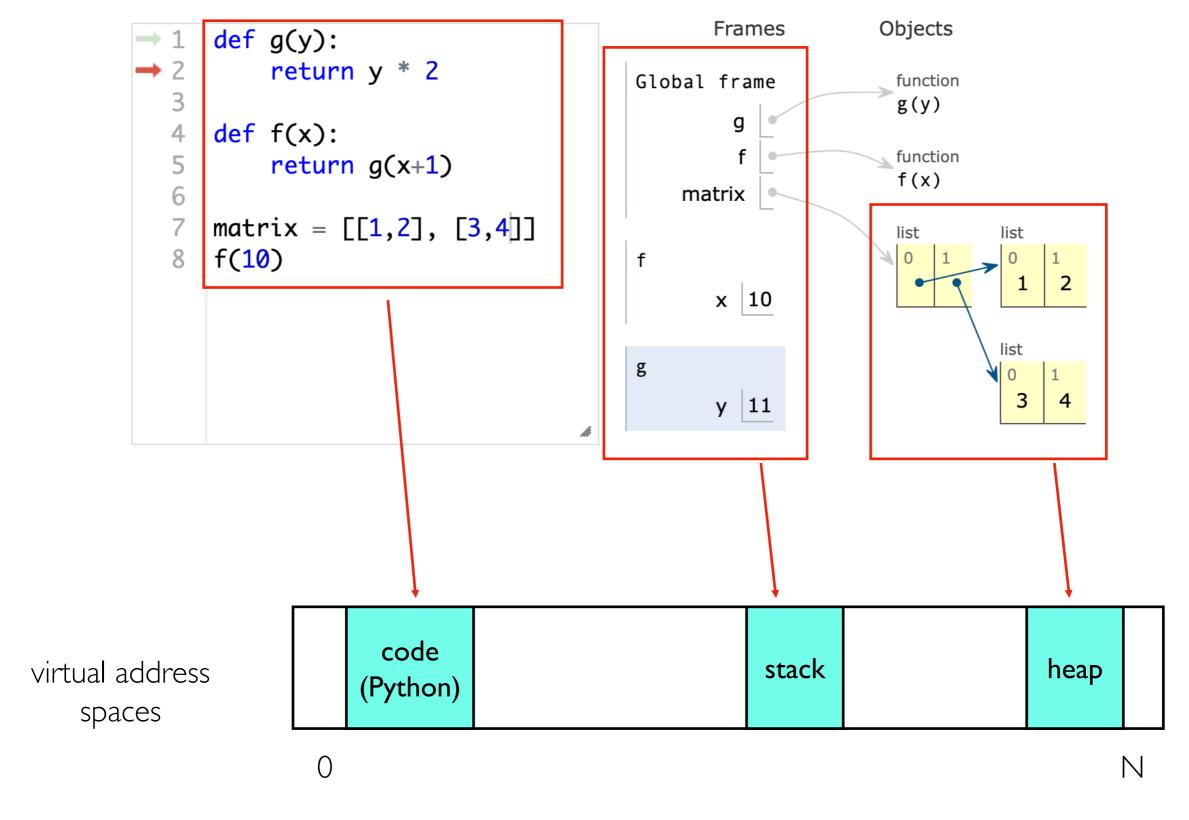


https://pythontutor.com/

virtual address spaces

what goes here?

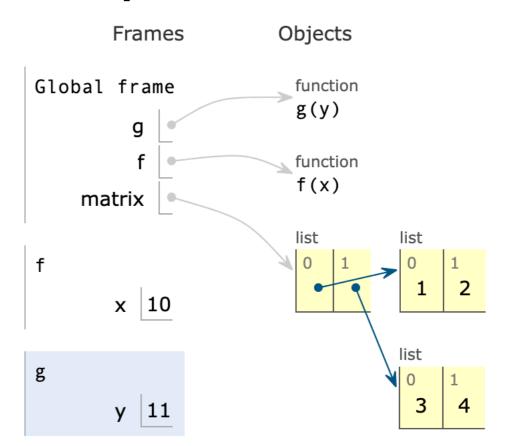
# What goes in an address space?



Note: code and heap generally not contiguous

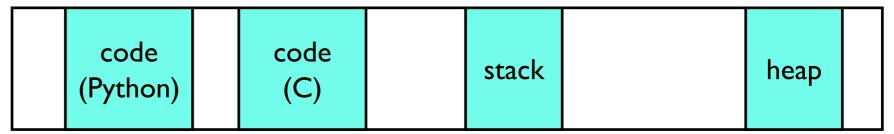
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some packages (like numpy)

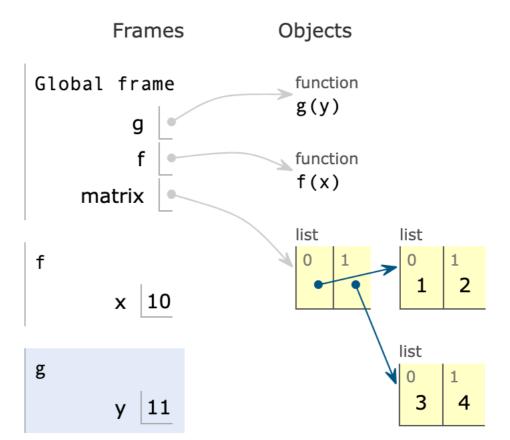
virtual address spaces



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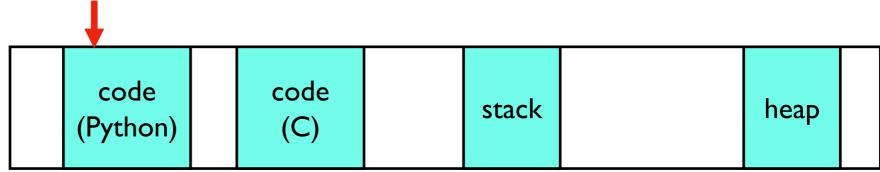
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instruction pointer

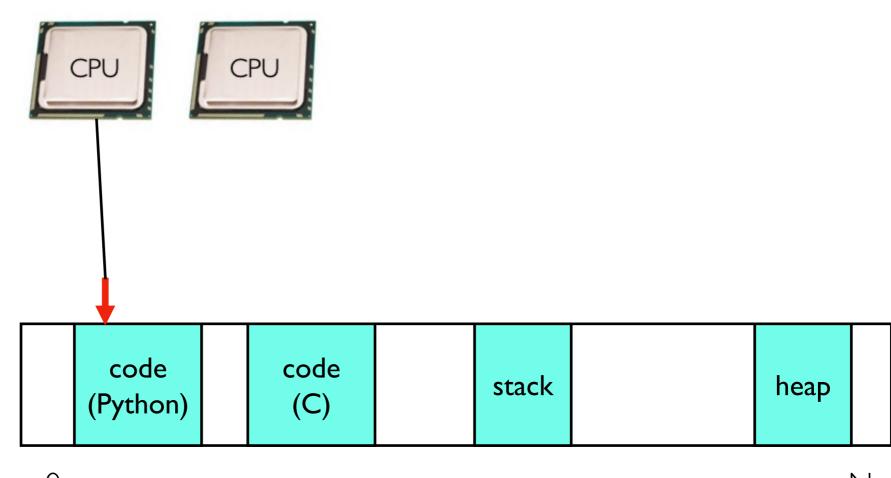
virtual address spaces



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#### **CPUs**

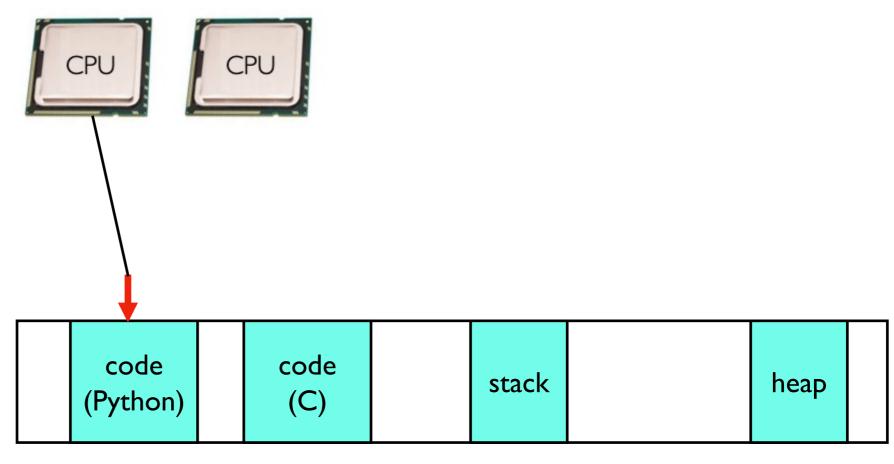
- CPUs are attached to at most one instruction pointer at any given time
- they run code by executing instructions and advancing the instruction pointer
- Note: interpreter left out for simplicity (CPU points to interpreter code, which points to Python bytecode)



virtual address spaces

#### **CPUs**

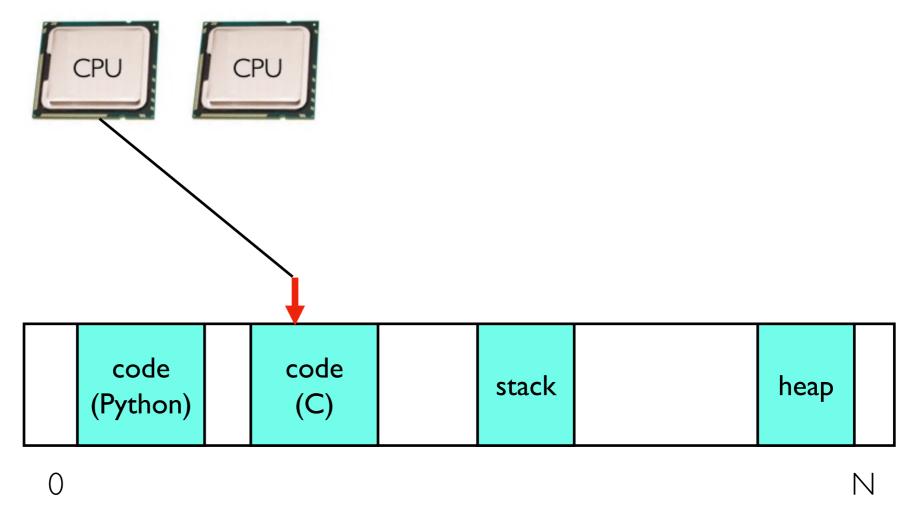
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virtual address spaces

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virtual address spaces

# Outline

Virtual Address Spaces

Threads

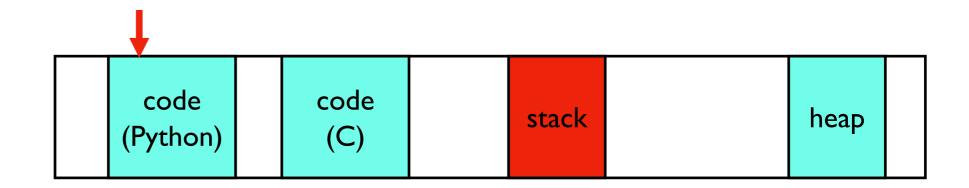
Demos and Worksheet

### **Threads**

Threads have their own instruction pointers and stacks, but share the heap.

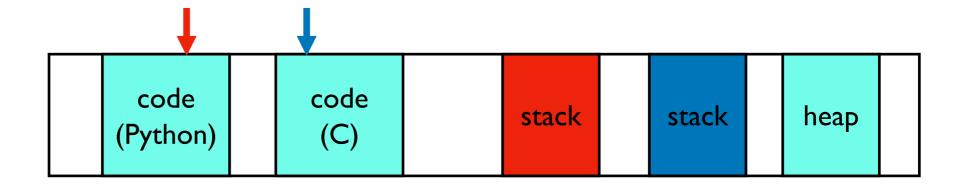
#### Single-threaded process:

virtual address spaces



#### Multi-threaded process:

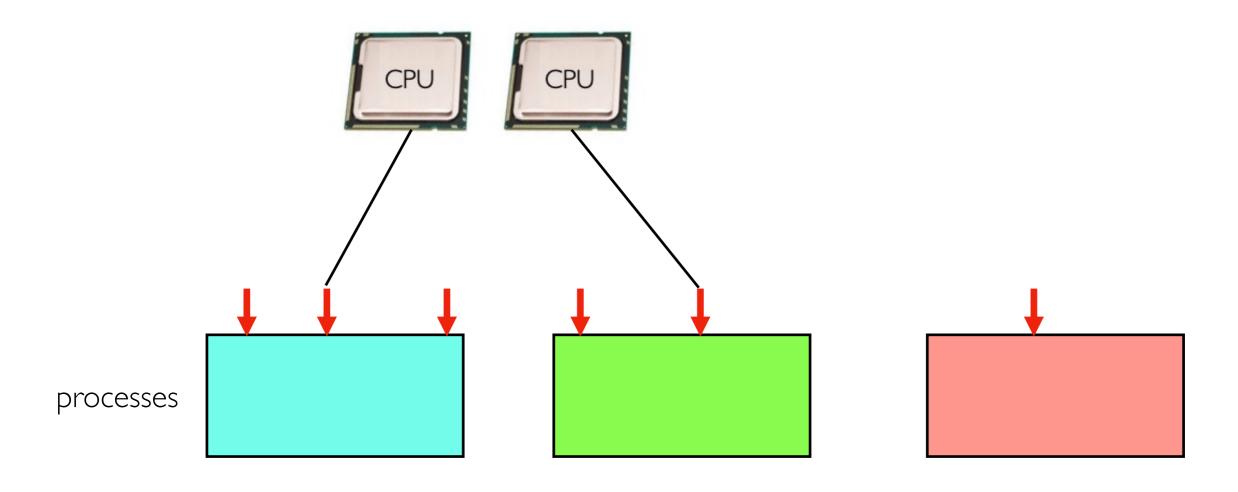
virtual address spaces



### Context Switch

#### Schedulers

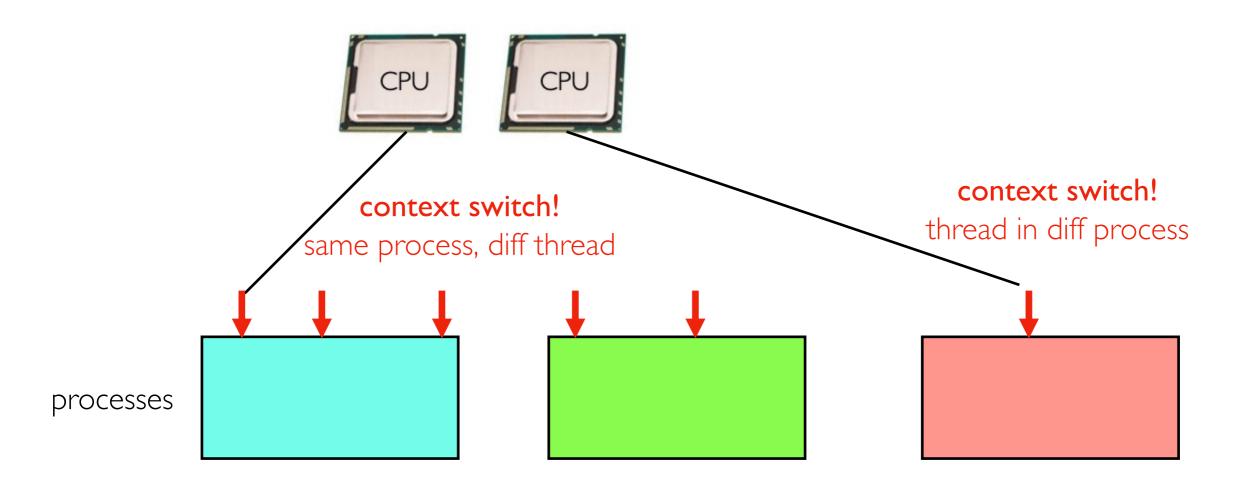
- CPU scheduler is an important sub system in an operating system
- schedulers decide when to context switch between threads
- context swich: change which thread a CPU is running



### Context Switch

#### Schedulers

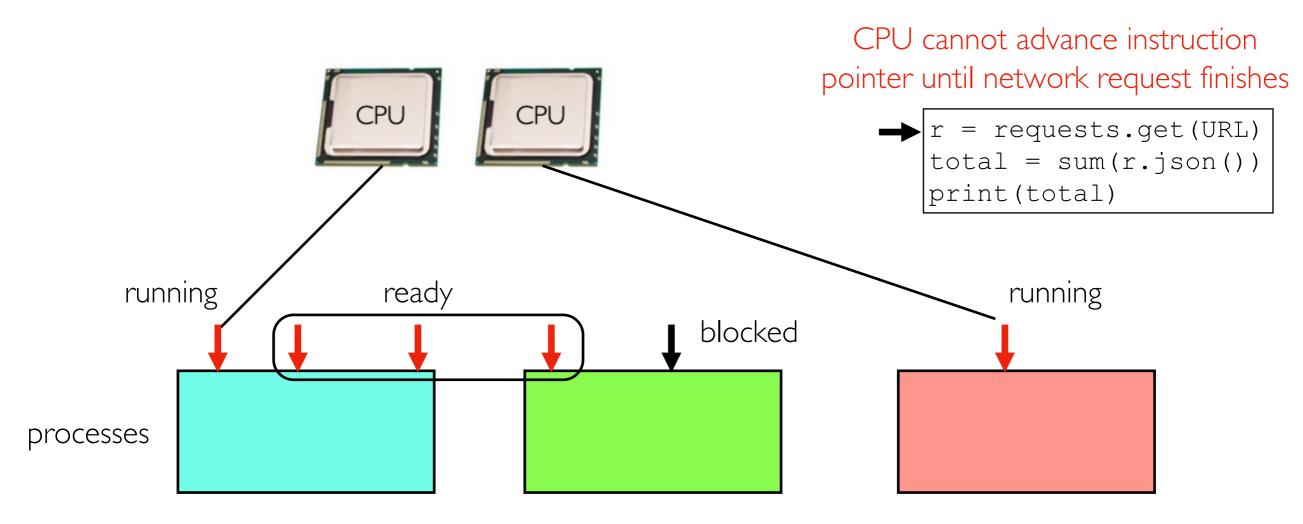
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# Scheduling Restrictions: Blocked Threads

Threads can be in one of three states

- running: CPU is executing it
- blocked: waiting on something other than CPU (network, input, disk, etc)
- ready: scheduler can choose to context switch to it



# Efficient Use of Compute Resources

Wasted cores: (1) not enough threads (2) blocked threads

For 100% CPU utilization (difficult goal)

- need at least one ready/running thread for each CPU core
- generally need more threads than cores (threads are often blocked)
- threads could be in one process (or many)

#### Multi-threaded applications

- good when multiple threads need to access frequently modified data structures
- new kinds of bugs possible (race conditions, deadlock)

Multi-process applications (<a href="https://docs.python.org/3/library/multiprocessing.html">https://docs.python.org/3/library/multiprocessing.html</a>)

- easier to program (or just manually launch several processes in background)
- better at keeping multiple cores busy simultaneously (Python specific)

Both approaches work well for dealing with blocked threads

# Coding Demos, Worksheet

#### Thread operations

- t = threading.Thread()
- t.start(target=????, args=[????])
- t.join()
- t.get\_native\_id()