# Advanced Programming

Process-based Parallelism

### IMPROVE APPLICATION PERFORMANCE

- ◆ Start reviewing all the key algorithms you have learned and their find their worst -case performance
- ◆ Look for different ways to improve the performance of an algorithm For example Bubble sort–>Merge Sort –> Quick Sort
- lacktriangle Choose the data structure wisely cost of accessing an element close to O(1)

### CACHING

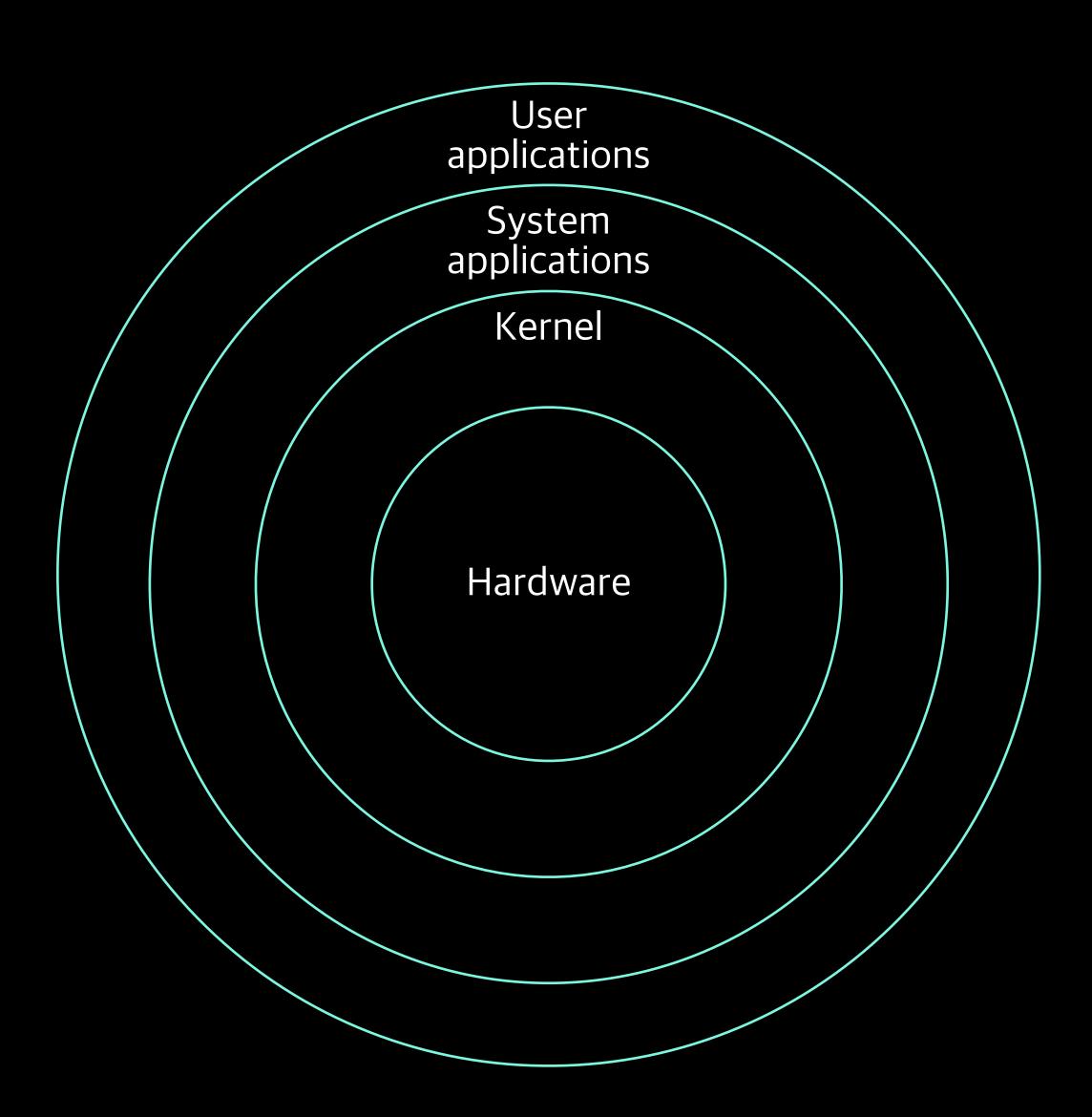
```
import time
# Function to calculate numbers recursively
def fibonacci_recursive(n):
    if n <= 1:
        return n
    else:
        return fibonacci_recursive(n-1) + \
               fibonacci_recursive(n-2)
# Function to calculate with caching
fib_cache = {}
def fibonacci_with_cache(n):
    if n in fib_cache:
        return fib_cache[n]
    if n <= 1:
        return n
    else:
        fib_cache[n] = fibonacci_with_cache(n-1) + \
                       fibonacci_with_cache(n-2)
        return fib_cache[n]
```

```
if __name__ == '__main__':
   # Calculate without caching
   start_time = time.time()
   fibonacci_recursive(35)
   end_time = time.time()
   print(f"Without caching = \
          {end time - start time: 4e} s")
   # Calculate with caching
   start time = time.time()
   fibonacci_with_cache(35)
   end_time = time.time()
   print(f"With caching =
         {end_time - start_time:.4e}")
 Without caching = 9.2852e-01 s
     With caching = 2.0981e-05 s
```

#### CACHING USING A DECORATOR

```
from functools import lru_cache
import time
# Function to calculate Fibonacci numbers
@lru_cache(maxsize=None) # No maximum cache size
def fibonacci(n):
  if n <= 1:
    return n
                                                                   Without caching: 7.8678131104e-06
  else:
    return fibonacci(n-1) + fibonacci(n-2)
                                                                   With caching : 0.0000000000e+00
if ___name__ == '___main___':
    # Without caching
    start_time = time.time()
    fibonacci(35)
    end_time = time.time()
    print(f"{'Without caching':<16}: {end_time - start_time:0.10e}")</pre>
    # With caching
    start_time = time.time()
    fibonacci(35)
    end_time = time.time()
    print(f"{'With caching':<16}: {end_time - start_time:0.10e}")</pre>
```

# ARCHITECTURE OF A LINUX SYSTEM



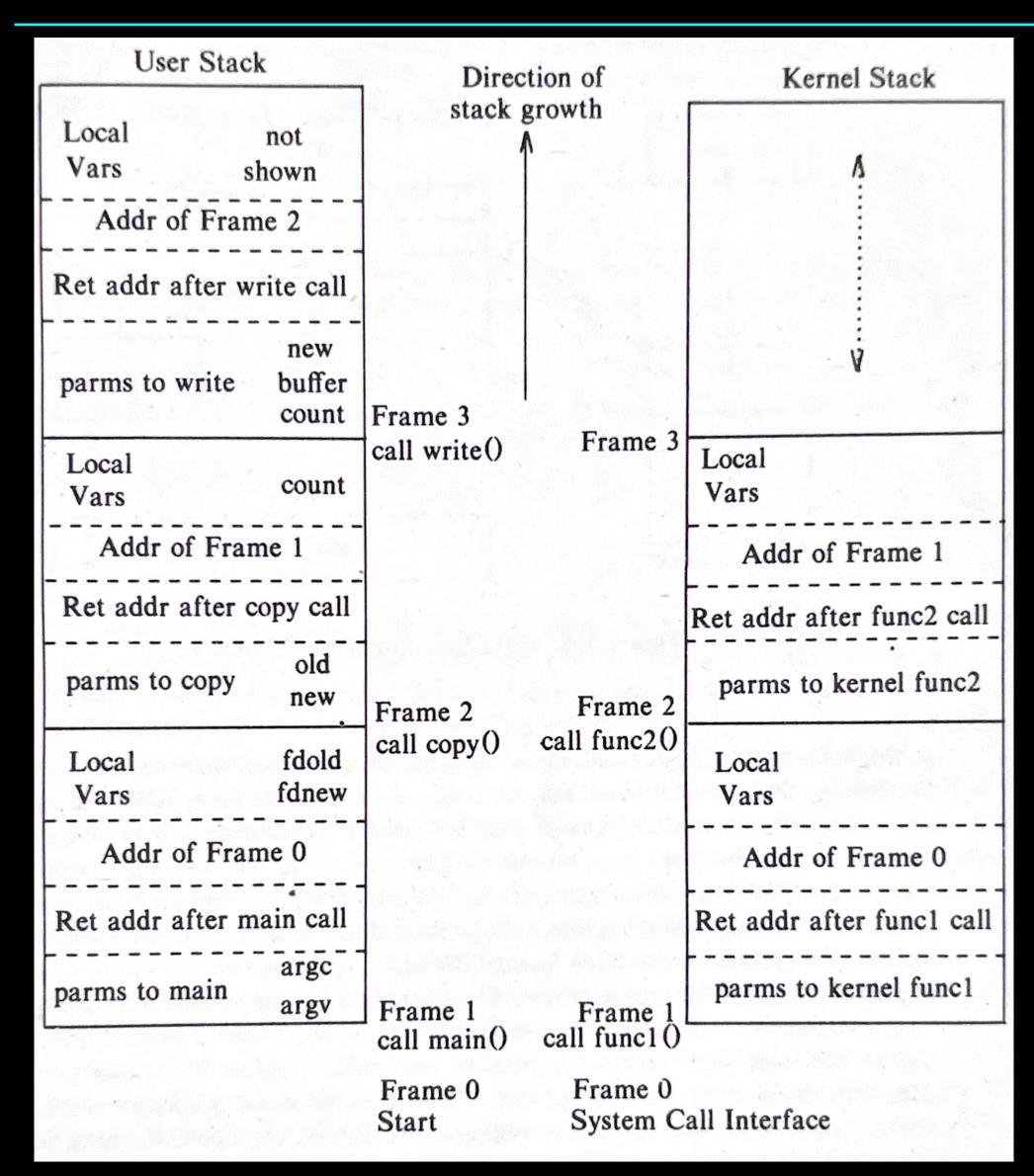
### **PROCESS**

- A process is an active entity in an OS
- Fundamental unit of a program in an execution state
  - Consists of regions
    - patterns of bytes interpreted as instructions by CPU, called as text
    - Data
    - Stack
- Self-contained
- Reads and writes its data and stack
- Cannot read and write another process directly
- Communicates with other processes through messages using system calls

### **PROCESS**

```
import sys
def copy(source_file, destination_file):
    try:
        with open(source_file, 'r') as src, \
        open(destination_file, 'w') as dst: \
             dst.write(src.read())
        print(f"File copied successfully from
              {source_file} to {destination_file}")
    except FileNotFoundError:
        print(f"Error: Source file '{source_file}'
              not found.")
    except IOError as e:
        print(f"Error accessing files: {e}")
if __name__ == '__main__':
    if len(sys.argv) != 3:
        print(f"Usage: python {sys_argv[0]}
              <source_file> <destination_file>")
        sys.exit(1)
    else:
        copy(sys_argv[1], sys_argv[2])
```

### USER AND KERNEL STACK FOR COPY

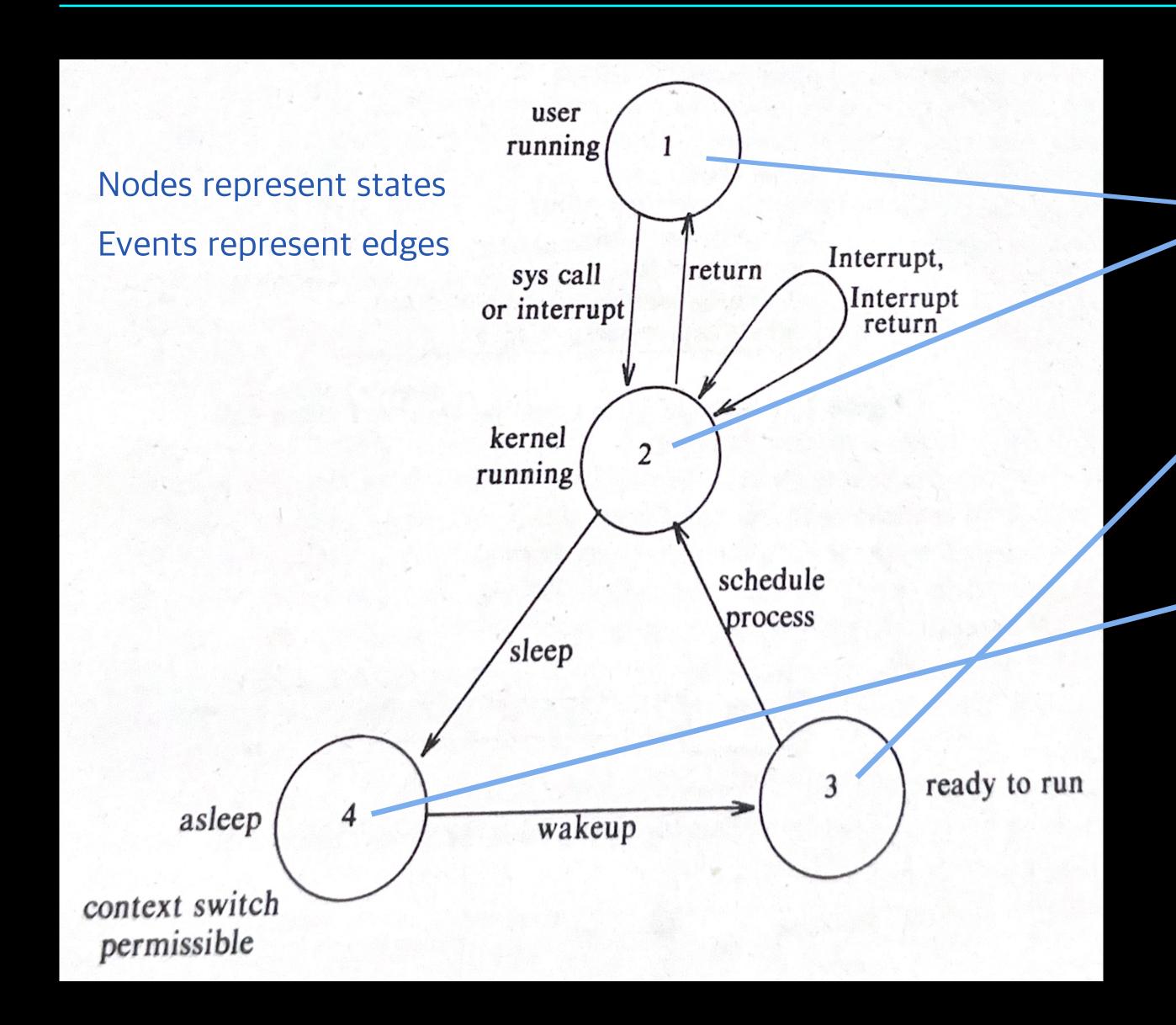


```
import sys
def copy(source file, destination file):
    try:
        with open(source_file, 'r') as src, \
        open(destination_file, 'w') as dst: \
             dst.write(src.read())
        print(f"File copied successfully from
              {source_file} to {destination_file}")
    except FileNotFoundError:
        print(f"Error: Source file '{source file}'
              not found.")
    except IOError as e:
        print(f"Error accessing files: {e}")
    name__ == '__main__':
   if len(sys.argv) != 3:
        print(f"Usage: python {sys.argv[0]}
              <source_file> <destination_file>")
        sys.exit(1)
    else:
        copy(sys.argv[1], sys.argv[2])
```

## PROCESS STATES

- Currently executing in user mode
- Currently executing in kernel mode
- ♦ Waiting Not in execution mode waiting for the scheduler to allocate CPU
- Sleeping Waiting for an I/O to complete

# PROCESS STATES AS DIGRAPH



★A process may be in state 1 or 2

Many process may be in this state

➤ Waiting for I/O to complete

### CONTEXT SWITCH

- In any multitasking operating system, multiple processes can run simultaneously
- A single CPU can only execute one process at a time
- ◆ The context of a process its state its text, values, data structure and the machine register values, content of user and kernel stack frames
- Context switching saves the context of a currently running process at time  $t_1$  in order to continue at a future time  $t_n$
- Allows the OS to switch between processes efficiently

- Allows all processes to share a single CP
- Facilitates efficient CPU utilisation
- Increased CPU Usage
- frequency of context switching significantly affects performance

### CONCURRENCY AND PARALLELISM

- Creates an impression that multiple are running at the same time
  - Manages multiple tasks
  - Tasks seem to run at or
  - Achieved through contex
     switching on a single CPU
  - Concurrency exhibits nondeterministic control flow
- → Browser download and browsing To learn more about Concurrent programming, listen to this lecture

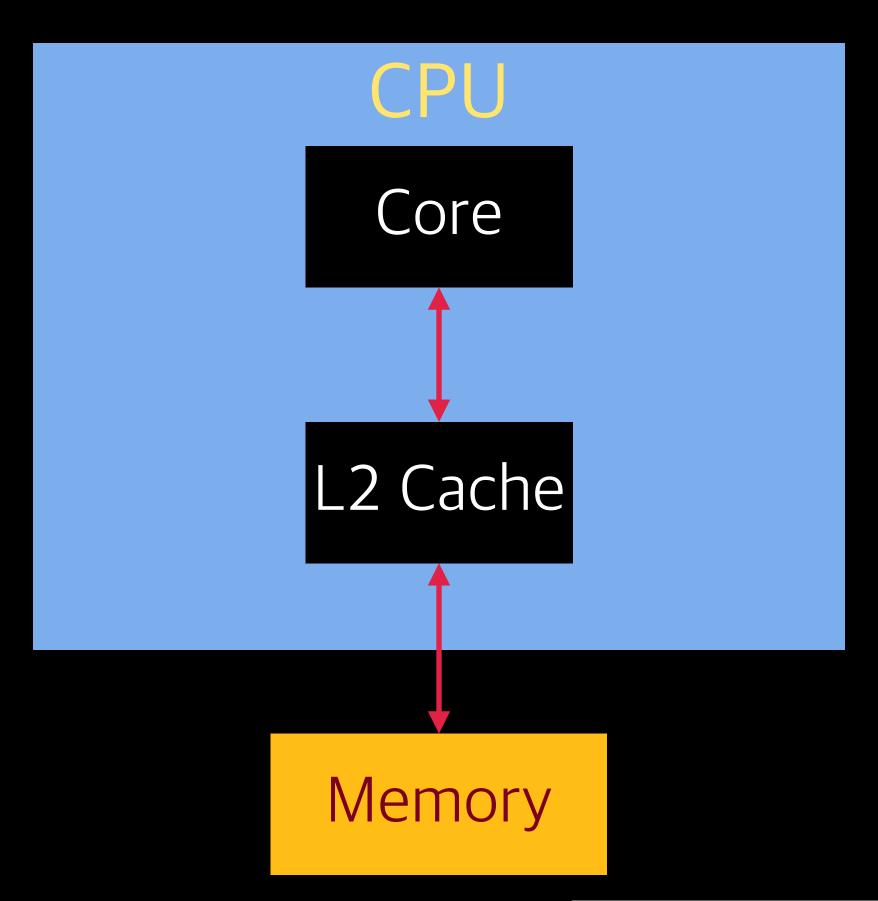
https://www.youtube.com/watch? v=XbdDSUI8NXE Due to

Asynchronous Execution
Context-switching
Resource sharing
Synchronisation

- Runs multiple computations simultaneously
- Tasks truly run at once
- Multiple CPU cores/distributed systems

# LECTURE 2 - MULTIPROCESSING

### SINGLE CORE CPU



L1 Cache – small memory – faster than main memory

L2 Cache > L1 Cache - faster than main memory

Store frequently accessed data and instructions to speed up the CPU's access to them

# BERNSTEIN'S CONDITIONS

- Conflict Serializability
  - One transaction reads and another writes the same data item
- Write-Write Conflict
  - ◆ Two different processes/transactions writing on the same critical section at the same time
- Write-Read Conflict

### THREADS

- A thread is a lightweight process with in a program
- ♦ Threads within a program share the same memory space. communication is efficient
- Threads can execute concurrently,
- Synchronization Due to sharing of memory, it is important to manage critical section
- OS manages the execution of threads using context switching
- ♦ Threads have a lifecycle consisting of creation, running, blocking, and termination phases

# PYTHON MODULES

- Threading
- Multiprocessing

### THREADS IN PYTHON

- Concurrency: Allows concurrent execution of tasks,
  - Improves (?) performance on multi-core systems
- ↑ Lightweight: efficient for smaller tasks suitable for tasks such as I/O operations
- Shared Memory: Threads within a program share the same memory space
- ◆ Synchronisation: Required to avoid race conditions
- Context Switching: OS manages thread execution using context switching
- Global Interpreter Lock (GIL) restricts true parallelism
  - Must acquire a lock to run a thread in the interpreter space
  - -Ensures only one thread at any point in time prhibits protrue parallelism

import threading

### THREAD EXAMPLE 1

#### THREAD EXAMPLE 2

```
import threading
def square(numbers):
    for num in numbers:
        print(f"Square of {num}: {num*num}")
if __name__ == '__main__':
    numbers = list(range(1, 11))
    half_len = len(numbers) // 2
    first_half = numbers[:half_len]
    second_half = numbers[half_len:]
    thread1 = threading.Thread(target=square, args=(first_half,))
    thread2 = threading.Thread(target=square, args=(second_half,))
    thread1.start();thread2.start()
    # Wait for both threads to finish
    thread1.join();thread2.join()
    print("Finished and exiting main thread")
```

### SERIAL COMPUTING

```
import numpy as np
import time
def square(seed):
    np random seed(seed)
    random_num = np.random.randint(0, 1000000)
    return random_num**2
def serial_main():
    start_time = time.time()
    for i in range(100000):
        square(i)
    end_time = time.time()
    print(f'Serial:Total time taken = {end_time - start_time:.2f} seconds')
if __name__ == '__main__':
   serial_main()
```

## CRITICAL SECTION

- ◆ A critical section is a specific part of a code that needs to be executed by multiple threads or processes atomically
  - To ensure data consistency and integrity
- Threats:
  - ♦ Race Conditions: If multiple threads access the same critical section concurrently

# MUTEX

Synchronisation primitives used for coordinating access to shared resources in multithreaded environments

### SEMAPHORE

- ◆ A semaphore is a variable that acts as a control mechanism for managing access to shared resources by multiple threads, preventing race conditions
- semaphore = 4 4 processes can run at a time concurrently
- If semaphore == 0, the process requesting will wait until semaphore > 0

# MUTEX VS SEMAPHORE

Feature	Mutex	Semaphore
FIINCTION	•	Signaling mechanism - Controls access to a shared resource by multiple threads
IJWNARCHIN	Yes - Thread that acquires the lock becomes the owner.	No - Any thread can wait on and signal the semaphore.
Signaling		Wait and signal operations - Thread waits until value > 0 (resource available) and increments when done.
Use Gases		Managing access to a limited pool of resources (e.g., database connections, network requests).

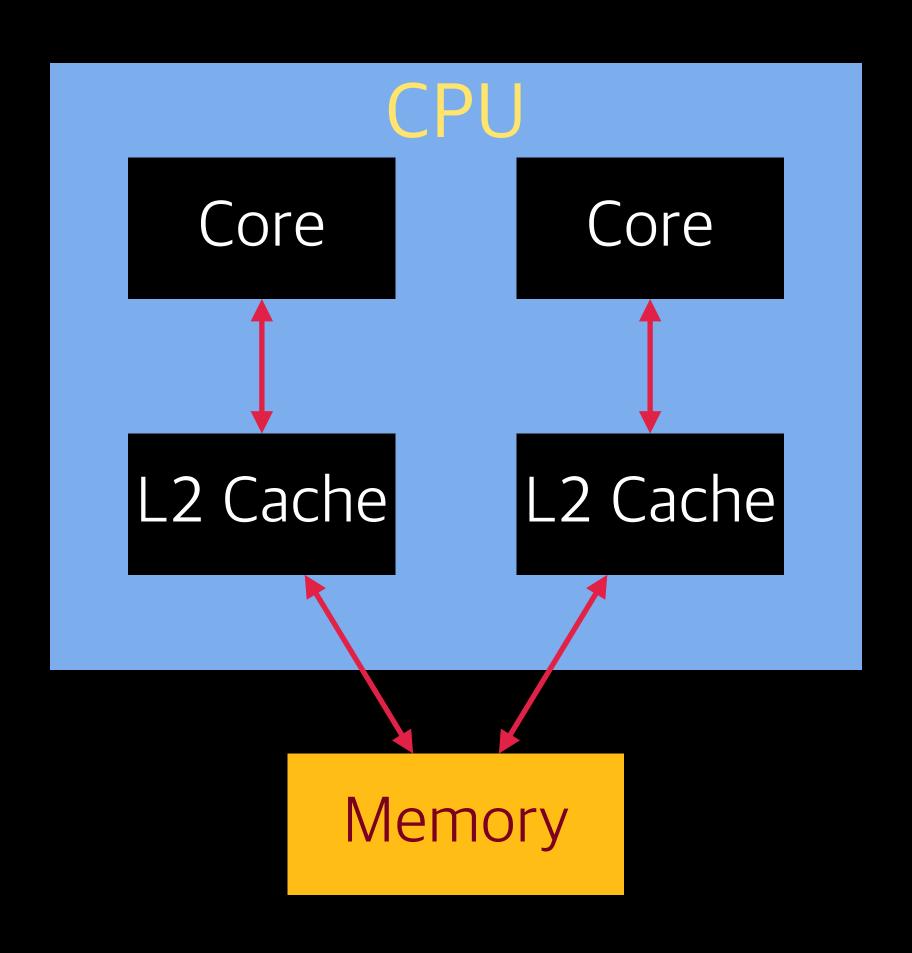
### THREAD SAFETY

- ↑ Thread safety refers to the ability of a program to function correctly when executed by multiple threads concurrently
- ◆ Unpredictable behavior happens when critical sections are accessed without proper synchronisation
- Multithreaded environment,
  - The program state must remain consistent
  - Data integrity is maintained

### THREADS COMPUTING:)

```
import numpy as np
import time
import threading
def threads_main():
    start_time = time.time()
    threads = []
    for i in range(100000):
        thread = threading.Thread(target=square,args=(i,))
        threads.append(thread)
        thread.start()
    for thread in threads:
        thread.join()
    end time = time.time()
    print(f"Total time taken: {end_time - start_time} seconds")
```

# DUAL CORE CPU



### **PROCESS**

- A process is an independent unit of execution in a computer program.
- Processes are independent have separate memory spaces.
- Creation Creation typically involves forking a new process from an existing one
- Communication Inter-Process Communication (IPC) using pipes, shared memory, message queues, and sockets
- ↑ Resource Management: Processes consume system resources such as memory, CPU time, and I/O resources.
- Process Scheduling The OS scheduler determines the order in which processes are executed on the CPU.
- Process Identification Each process is unique process

#### MULTIPROCESSING IN PYTHON

- Supports spawning process.
- Offers local and remote concurrency
- Multiprocessing utilises the physical cores the CPU
  - Creates subprocesses instead of threads
  - Leverages multiplecores

```
import multiprocessing as mp
def square(num, result):
  result.value = num * num
if ___name__ == "__main__":
    number = 5
   # Shared variable to store the result
    result = mp.Value('i', 0) # 'i' indicates integer type
    # Create a process to perform the calculation
    process = mp.Process(target=square, args=(number, result))
    process.start()
    process.join()
    print(f"The square of {number} is: {result.value}")
```

```
thread = threading.Thread(target=square,args=(i,))
```

#### MULTIPROCESSING WITH MANAGER

```
if name == " main ":
 numbers = list(range(1, 11))
 num_processes = 4
 chunk_size = len(numbers) // num_processes
 sub_ranges = []
  for i in range(num_processes):
   start = i * chunk size
   end = min((i + 1) * chunk_size - 1, len(numbers) - 1)
   sub_ranges.append((start, end))
 manager = Manager()
 squared_list = manager.list()
 processes = []
  for start, end in sub_ranges:
    process = Process(target=square,
                      args=(numbers, start, end, squared_list))
    process.start()
    processes append(process)
  for process in processes:
    process.join()
 print(f"Square for {numbers}")
  for num in squared_list:
   print(num)
```

# MULTIPROCESSING USING POOL

- ♦ The Pool class of multiprocessing simplifies the coding using a map-reduce process
  - Distributes the work among the subprocess using map
  - Collect the return value as a list
  - Relieves the programer from the burner of
    - managing processes
    - Mainting a shared data/state

```
from multiprocessing import Pool
import multiprocessing as mp

pool = Pool(processes=4)
#OR
cpu_count = mp.cpu_count()
pool = Pool(processes=cpu_count)
```

### MULTIPROCESSING EXAMPLE

```
import numpy as np
import time
import multiprocessing as mp
def parallel_main():
   start_time = time.time()
    cpu_count = mp.cpu_count()
   pool = mp.Pool(processes=cpu_count)
   _ = [pool_map(square, range(100000))]
   end_time = time.time()
    print(f"Parallel:Total time taken: {end_time - start_time} seconds")
     name == ' main ':
   parallel_main()
```

# PRACTICAL EXAMPLES

- Reading files in parallel and combine the results
- lack Estimating  $\pi$  using Monte Carlo Simulation

### POOL

- One can create a pool of processes which will carry out tasks submitted to it with the Pool class
- If processes is None then the number returned by os.cpu\_count() is used

# ITERABLE - (LIST, DICT, TUPLE)

- An object capable of returning its members one at a time.
- List, str, tupe, dict, file objects, and objects of any classes you define with an \_\_iter\_\_() method or with a \_\_getitem\_\_() method that implements sequence semantics

```
text = "This is iterable"
for alpha in text:
    print(alpha)
tokens = ["This", "is", "iterable"]
for token in tokens:
    print(token)
dict_tokens = {"This":1, "is":2, "iterable":3}
for key in dict_tokens:
    print(f'key={key}, value ={dict_tokens[key]}')
trigram = ("This", "is", "iterable")
    for token in trigram:
        print(token)
```

### ITERABLE

- An object capable of returning its members one at a time.
- List, str, tupe, dict, file objects, and objects of any classes you define with an \_\_iter\_\_() method or with a \_\_getitem\_\_() method that implements sequence semantics

```
class FibSequence:
  def ___init___(self, max_count):
    self.max_count = max_count
    self.a, self.b = 0, 1
    self.count = 0
  def ___iter__(self):
    return self
  def next (self):
    if self.count > self.max count:
      raise StopIteration
    current_value = self.a
    self.a, self.b = self.b, self.a + self.b
    self.count += 1
    return (self.count,current_value)
if name == '__main__':
    fib_seq = FibSequence(20)
    for i,num in enumerate(fib_seq):
        print(f'{i}:{num}')
```

### ITERABLE

- An object capable of returning its members one at a time.
- List, str, tupe, dict, file objects, and objects of any classes you define with an \_\_iter\_\_() method or with a \_\_getitem\_\_() method that implements sequence semantics

```
class ColorList:
  def ___init___(self, color_names):
    self.color_names = color_names
  def ___getitem__(self, index):
    if 0 <= index < len(self.color names):</pre>
      return self.color_names[index]
    else:
      raise IndexError("Index out of range")
if __name__ == "__main__":
    color_names = ColorList(["Red", "Green", "Orange"])
    for color in color_names:
        print(color)
    # Access elements by index
    print(color_names[0])
    print(color_names[2])
```

### PYTHON OBJECT SERIALISATION - PICKLE

- Useful in transmitting Python objects
- Pickling The process of converting an object into a byte stream
- Unpickling: The process of reconstructing an object from a byte stream
- Picklable: An object that can be successfully pickled and unpickled

Importance of Piclability

- ◆ Data Persistence: Save objects to disk for later use
- ◆ Remote Procedure Calls (RPC): Python objects across processes or machines in distributed systems
- Machine Learning Pipelines: Saving trained models and deploying them in different environments.

What can be pickled and unpickled?

### PICKLE - EXAMPLE

```
if __name__ == '__main__':
    fib_seq = FibSequence(21)
    with open('fib_seq.pickle', 'wb') as f:
        pickle.dump(fib_seq, f)

with open('fib_seq.pickle', 'rb') as f:
        load_fib_seq:FibSequence = pickle.load(f)

for fib in load_fib_seq:
    print(fib)
```

# PROCES TIME

Chunk size	Elapsed Time	# processes
None	11.7	8
2	12.08	8
4	11.01	8
8	11.62	8
16	11.87	8
32	11.64	8
64	10.75	8
128	10.47	8
256	10.76	8

Chunk size	Elapsed Time	# processes
None	10.66	16
2	11.46	16
4	11.97	16
8	9.21	16
16	9.03	16
32	8.83	16
64	8.54	16
128	9.30	16
256	9.97	16

Elapsed time for converting a json text to plain text using multiprocessing

Seq\_Process: elapsed time = 40.385626792907715

### JSON2TEXT

```
def par_write(files):
    cpu_count = os.cpu_count()
    file_count = len(files)
    p = Pool(processes=16)
    p.map(write_file, files, chunksize=16)
    p.close()
```

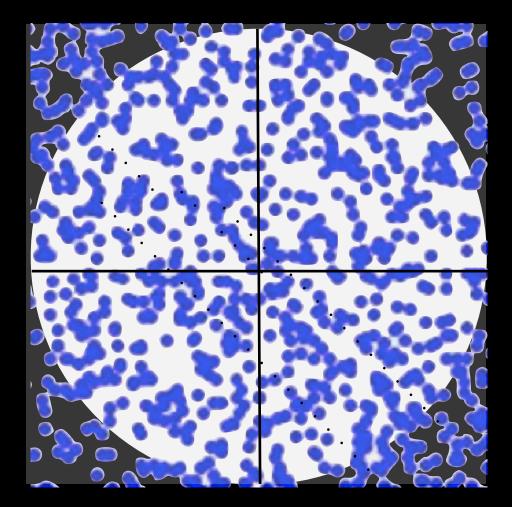
```
def par_main():
    json_path = '/Users/ram/Teaching/Demos/NLP/Python/COVID19/pdf_json'
    json_files = glob.glob(f'{json_path}/**/*.json',recursive=True)
    start_time = time.time()
    par_write(json_files)
    end_time = time.time()
    print(f'elapsed time = {end_time - start_time :0.2f}')
```

### MONTE CARLO SIMULATION

- ♦ Developed in the 1940s during the creation of the atomic bomb
- Monte Carlo methods are a class of computational algorithms
- They can be applied to a wide range of problems, not limited to statistics
- ♦ Instead of relying on pure statistical analysis, repeated random sampling to arrive at solutions is used
- ♦ Monte Carlo methods typically provide approximate solutions, not exact answers
- ♦ Valuable when analytical or numerical solutions are either unavailable or too complex to implement

# ESTIMATION OF II

- Define the Simulation Space:
  - Imagine a square with a side length 2 units, centred at the origin (0, 0)
  - This square will enclose a circle with radius 1 unit
- Set the total number of random points (num\_points)
  - The points ether lie in the square or in the circle
- ightharpoonup Set the *points\_in\_circle* = 0
- Iterate for all num\_points times
  - In each iteration a random point (x, y) is thrown
  - ◆ Generate a random x-coordinate (x) between 0 and 1
  - Generate a random y-coordinate (y) between 0 and 1
  - If the point (x, y) is with in the circle,
  - points\_in\_circle ← points\_in\_circle + 1
    - Calculate the Euclidean distance,  $E(x, y) = \sqrt{(x^2 + y^2)}$
    - If  $E(x, y) \le 1$ , then the point falls inside the circle.
- $\bullet \quad E_{\Pi} = \frac{points\_in\_circle}{num\_points} \times 4$
- The more points simulated (higher num\_points), the closer the pi\_estimate will be to the actual value of Pi (approximately 3.14159).



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

Area of the square =  $4r^2$ 

Area of the circle  $= \pi r^2 = \pi$ Area of the square  $= 4r^2 = 4$ 

### ESTIMATION OF PI

```
def calculate_pi(iterations):
    points_in_circle = 0
    for _ in range(iterations):
        x = random.random()
        y = random.random()
        if (x**2 + y**2) <= 1:
            points_in_circle += 1
        pi_estimate = (4 * points_in_circle) / iterations
        return pi_estimate</pre>
```

```
def main():
    num_processes = os.cpu_count()
    num_iterations = 100000

with Pool(processes=num_processes) as pool:
    pi_estimates = pool.map(calculate_pi, [num_iterations] * num_processes)

average_pi = sum(pi_estimates) / len(pi_estimates)
    print(f"Estimated pi using Monte Carlo simulation: {average_pi}")

if __name__ == "__main__":
    main()
```