Assignment 11 Multiprocessing

September 16, 2023

Q1. What is multiprocessing in python? Why is it useful?

ANS:

Multiprocessing in Python is a technique that involves the use of multiple processes, each with its own interpreter and memory space, to perform tasks concurrently. It is a form of parallelism that takes advantage of multiple CPU cores or processors in a system to execute tasks in parallel. Python's multiprocessing module provides a high-level API for creating and managing multiple processes.

Multiprocessing is useful for several reasons:

- 1. Parallelism: Multiprocessing allows you to perform multiple tasks concurrently, which can significantly improve the performance of CPU-bound tasks, such as data processing, simulations, and mathematical computations. Each process runs independently, utilizing its own CPU core.
- 2. Utilizing Multiple Cores: In modern computers, most processors have multiple cores. Multiprocessing helps fully utilize all available CPU cores, making your programs more efficient and reducing execution time.
- 3. Isolation: Each process in multiprocessing has its own memory space, which means that data and variables are isolated between processes. This isolation reduces the risk of data corruption and makes it easier to reason about concurrent code.
- 4. Fault Tolerance: If one process encounters an error or crashes, it typically does not affect other processes. This fault tolerance improves the overall stability and reliability of your applications.
- 5. Improved Responsiveness: In applications with user interfaces, multiprocessing can help maintain responsiveness by offloading time-consuming tasks to separate processes, ensuring that the main user interface remains responsive to user input.
- 6. Scalability: Multiprocessing is a scalable approach to parallelism. You can create as many processes as needed to match the computational demands of your application.

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Q2. What are the differences between multiprocessing and multithreading?

ANS:

Multiprocessing and multithreading are both techniques used for achieving concurrency in programs, but they have distinct differences in terms of their implementation and use. Here are the

key differences between multiprocessing and multithreading:

1. Processes vs. Threads:

Multiprocessing: In multiprocessing, the program runs as multiple independent processes. Each process has its own memory space, system resources, and Python interpreter. Processes do not share memory by default.

Multithreading: In multithreading, the program runs as multiple threads within a single process. Threads share the same memory space and resources, including variables and data structures.

2. Isolation:

Multiprocessing: Processes are isolated from each other. Any data sharing between processes requires explicit inter-process communication (IPC) mechanisms, such as pipes or queues.

Multithreading: Threads share the same memory space, which means they can easily access and modify shared data. This shared access can lead to synchronization issues like race conditions.

3. GIL (Global Interpreter Lock):

Multiprocessing: Each process has its own Python interpreter and GIL. This means that multiple processes can execute Python code in parallel, making multiprocessing well-suited for CPU-bound tasks.

Multithreading: All threads within a single Python process share the same GIL. This means that only one thread can execute Python code at a time, making multithreading less suitable for CPU-bound tasks. However, it can still be useful for I/O-bound tasks.

4. Complexity:

Multiprocessing: Implementing multiprocessing is often more complex than multithreading due to the need for inter-process communication and coordination between processes.

Multithreading: While multithreading can have its own complexities, it is generally considered easier to work with than multiprocessing, especially for tasks that benefit from shared memory.

5. Resource Overhead:

Multiprocessing: Creating and managing processes typically incurs higher resource overhead compared to threads. Each process has its own memory space and resources.

Multithreading: Threads have lower resource overhead since they share memory and resources within the same process.

6. Use Cases:

Multiprocessing: Multiprocessing is well-suited for CPU-bound tasks, parallelizing computations, and taking full advantage of multi-core processors.

Multithreading: Multithreading is often used for I/O-bound tasks, such as network communication or file I/O, where threads can perform tasks concurrently while waiting for I/O operations to complete.

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Q3. Write a python code to create a process using the multiprocessing module.

ANS:

```
[2]: import multiprocessing

# Function to be run in the process
def worker_function():
    print("Worker process is running.")

if __name__ == "__main__":
    # Create a new process
    process = multiprocessing.Process(target=worker_function)

# Start the process
process.start()

# Wait for the process to finish
process.join()

print("Main process finishes.")
```

Worker process is running. Main process finishes.

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Q4. What is a multiprocessing pool in python? Why is it used?

ANS:

A multiprocessing pool in Python refers to a group of worker processes that are created and managed for the purpose of distributing and parallelizing tasks or function calls across multiple processes. The multiprocessing module provides a Pool class that makes it easier to work with a pool of worker processes.

Here's why multiprocessing pools are used and their advantages:

- 1. Parallel Execution: Multiprocessing pools allow you to execute multiple instances of a function or task concurrently. Each worker process in the pool can execute a task independently, which can significantly improve the performance of CPU-bound or computationally intensive tasks.
- 2. Efficient Resource Utilization: Pools manage the creation and recycling of worker processes, which helps efficiently utilize system resources. You can specify the number of worker processes to match the available CPU cores or a desired level of parallelism.
- 3. Simplified API: The multiprocessing.Pool class provides a high-level and easy-to-use API for parallelism. You can use methods like map, map_async, apply, and apply_async to distribute tasks to the pool and retrieve results.
- 4. Load Balancing: Pools automatically distribute tasks among worker processes, ensuring that each worker receives roughly an equal share of the work. This load balancing helps maintain

good overall performance.

- 5. Fault Tolerance: If one worker process encounters an error or crashes, the pool can continue to execute other tasks with the remaining worker processes, improving the reliability of your application.
- 6. Reusability: You can reuse a pool to execute multiple tasks or function calls, which reduces the overhead of creating and destroying processes for each task.

Results: [1, 4, 9, 16, 25]

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Q5. How can we create a pool of worker processes in python using the multiprocessing module? ANS:

```
[4]: import multiprocessing

# Define the function to be executed in parallel

def my_function(x):
    # Example computation
    result = x * 2
    return result

if __name__ == "__main__":
    # Create a pool with 4 worker processes
    with multiprocessing.Pool(processes=4) as pool:
        # Define a list of data to be processed in parallel
        data = [1, 2, 3, 4, 5]

# Use the pool to apply the function to each element in the data
        results = pool.map(my_function, data)

print("Results:", results)
```

```
Results: [2, 4, 6, 8, 10]
```

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Q6. Write a python program to create 4 processes, each process should print a different number using the multiprocessing module in python.

ANS:

```
[5]: import multiprocessing
     # Function to print a number
     def print_number(number):
         print(f"Process {number}: {number}")
     if __name__ == "__main__":
         # Create a list of numbers to be printed
         numbers = [1, 2, 3, 4]
         # Create a list to hold the process objects
         processes = []
         # Create and start 4 processes, each printing a number
         for number in numbers:
             process = multiprocessing.Process(target=print_number, args=(number,))
             processes.append(process)
             process.start()
         # Wait for all processes to finish
         for process in processes:
             process.join()
         print("Main process finishes.")
```

```
Process 1: 1
Process 2: 2
Process 3: 3
Process 4: 4
Main process finishes.
```

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