Python

# Python Concurrency

Part 3: Advanced Patterns and Hybrid Approaches

April 5, 2025

Source Code

## 1. Combining Approaches for Complex Systems

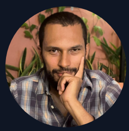
For real-world applications, you can combine different concurrency models to leverage their respective strengths:

```python
import asyncio
import concurrent.futures
import time

def cpu_bound(number):
    """CPU-bound task (runs in a process)"""
    total = sum(i * i for i in range(number))
    return total

def io_bound(number):
    """I/O-bound task (runs in a thread)"""
    time.sleep(1)  # Simulate I/O
    return number * 2

```



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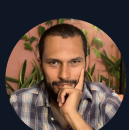
**PYTHON** | **CONCURRENCY (PART 3)**

```python
async def main():
    # Create executor pools
    process_pool = concurrent.futures.ProcessPoolExecutor(max_workers=4)
    thread_pool = concurrent.futures.ThreadPoolExecutor(max_workers=10)

    loop = asyncio.get_running_loop()

    # CPU-bound tasks (run in process pool)
    cpu_numbers = [5_000_000, 10_000_000, 15_000_000, 20_000_000]
    cpu_tasks = [
        loop.run_in_executor(process_pool, cpu_bound, number)
        for number in cpu_numbers
    ]

    # I/O-bound tasks (run in thread pool)
    io_numbers = list(range(1, 11))
    io_tasks = [
        loop.run_in_executor(thread_pool, io_bound, number)
```



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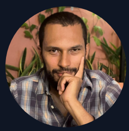
```python
        for number in io_numbers
    ]

    # Async I/O-bound tasks (native asyncio)
    async_tasks = [asyncio.sleep(1, result=f"async_{i}") for i in range(5)]

    # Gather all results
    print("Running all tasks concurrently...")
    start = time.time()

    cpu_results = await asyncio.gather(*cpu_tasks)
    io_results = await asyncio.gather(*io_tasks)
    async_results = await asyncio.gather(*async_tasks)

    end = time.time()

    # Show results
    print(f"\nTotal time: {end - start:.2f} seconds")
    print(f"CPU results: {len(cpu_results)} tasks completed")
```



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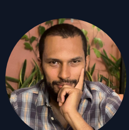
```python
    print(f"I/O results: {len(io_results)} tasks completed")
    print(f"Async results: {len(async_results)} tasks completed")

    # Clean up
    process_pool.shutdown()
    thread_pool.shutdown()

if __name__ == "__main__":
    asyncio.run(main())

# Running all tasks concurrently...

# Output:
# Total time: 2.02 seconds
# CPU results: 4 tasks completed
# I/O results: 10 tasks completed
# Async results: 5 tasks completed
```

This hybrid approach demonstrates how to:

- **Use ProcessPoolExecutor** for CPU-bound tasks to bypass the GIL



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- **Use ThreadPoolExecutor** for blocking I/O operations
- **Use native asyncio** for non-blocking I/O operations
- **Coordinate all approaches** through asyncio's event loop

The example shows how `run_in_executor` allows integration of traditional concurrency approaches with asyncio, creating a unified system that handles different types of workloads optimally.

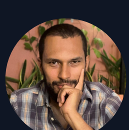
## 2. Advanced Concurrency Patterns

### 2.1. Fan-Out/Fan-In Pattern

The fan-out/fan-in pattern is ideal for data parallelism, where you split a large task into smaller subtasks, process them concurrently, and then combine the results:

```python
import asyncio
import time
import random

```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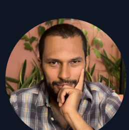
**PYTHON** | **CONCURRENCY (PART 3)**

```python
async def fan_out_fan_in_example():
    """
    Demonstrates the fan-out/fan-in pattern:
    1. Fan-out: Split a task into multiple subtasks
    2. Process each subtask concurrently
    3. Fan-in: Collect and combine results
    """
    async def process_chunk(chunk_id, data):
        """Process a chunk of data"""
        print(f"Processing chunk {chunk_id} with {len(data)} items")
        await asyncio.sleep(random.uniform(0.5, 1.5))  # Simulate processing

        # Simulate results (sum of items with processing artifact)
        result = sum(data) * random.uniform(0.9, 1.1)
        print(f"Chunk {chunk_id} processed, result: {result:.2f}")
        return result

    # Create a large dataset
    dataset = [i * 2 for i in range(1000)]
```

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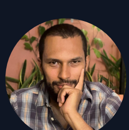
```python

    # 1. Fan-out: Split data into chunks
    chunk_size = 100
    chunks = [dataset[i:i+chunk_size] for i in range(0, len(dataset), chunk_size)]
    print(f"Split dataset into {len(chunks)} chunks of {chunk_size} items each")

    # 2. Process chunks concurrently
    print("\nFanning out processing to multiple tasks...")
    tasks = [
        process_chunk(i, chunk)
        for i, chunk in enumerate(chunks)
    ]

    # 3. Fan-in: Gather all results
    print("\nFanning in results...")
    start_time = time.time()
    results = await asyncio.gather(*tasks)
    end_time = time.time()
```

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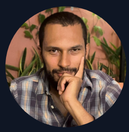
**PYTHON** | **CONCURRENCY (PART 3)**

```python

    # Combine results (in this case, take the average)
    final_result = sum(results) / len(results)

    print(f"\nAll chunks processed in {end_time - start_time:.2f} seconds")
    print(f"Final result (average): {final_result:.2f}")

# Run the example
if __name__ == "__main__":
    asyncio.run(fan_out_fan_in_example())

# Output:
# Split dataset into 10 chunks of 100 items each

# Fanning out processing to multiple tasks...

# Fanning in results...
# Processing chunk 0 with 100 items
# Processing chunk 1 with 100 items
```



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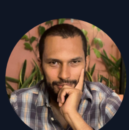
```
# Processing chunk 2 with 100 items
# Processing chunk 3 with 100 items
# Processing chunk 4 with 100 items
# Processing chunk 5 with 100 items
# Processing chunk 6 with 100 items
# Processing chunk 7 with 100 items
# Processing chunk 8 with 100 items
# Processing chunk 9 with 100 items
# Chunk 0 processed, result: 10331.82
# Chunk 5 processed, result: 106947.79
# Chunk 7 processed, result: 162176.67
# Chunk 3 processed, result: 71535.70
# Chunk 9 processed, result: 187058.54
# Chunk 6 processed, result: 117476.36
# Chunk 2 processed, result: 53288.83
# Chunk 1 processed, result: 31235.51
# Chunk 8 processed, result: 160851.37
# Chunk 4 processed, result: 88868.60

# All chunks processed in 1.38 seconds
```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```python
# Final result (average): 98977.12
```

This pattern applies to many real-world scenarios like map-reduce operations, batch processing, and parallel data analysis.

## 2.2. Task Queue with Priority

A priority-based task queue allows processing important tasks first:

```python
import asyncio
import random


class AsyncTaskQueue:
    """A task queue with priority and worker pool for asyncio"""

    def __init__(self, num_workers=3):
        self.queue = asyncio.PriorityQueue()
        self.num_workers = num_workers
```

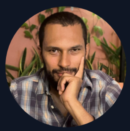
**PYTHON** | **CONCURRENCY (PART 3)**

```python
        self.workers = []
        self.running = False
        self._task_counter = 0  # Counter to ensure unique comparison

    async def add_task(self, task_func, priority=0):
        """Add a task to the queue with priority (lower is higher)

        Args:
            task_func: A callable that returns a coroutine (not a coroutine object itself)
            priority: Priority value (lower is higher priority)
        """
        # Increment counter to ensure FIFO behavior for same priority tasks
        self._task_counter += 1
        # Store priority, counter, and task function (not coroutine)
        await self.queue.put((priority, self._task_counter, task_func))

    async def worker(self, worker_id):
```

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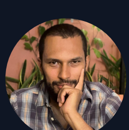
**PYTHON** | **CONCURRENCY (PART 3)**

```python
        """Worker that processes tasks from the queue"""
        while self.running:
            try:
                # Get a task from the queue
                priority, task_id, task_func = await self.queue.get()

                try:
                    print(
                        f"Worker {worker_id}: Processing task {task_id} with priority {priority}"
                    )
                    # Execute the task function to get the coroutine, then await it
                    result = await task_func()
                    print(
                        f"Worker {worker_id}: Task {task_id} completed with result: {result}"
                    )
                except asyncio.CancelledError:
                    # Handle cancellation properly
```



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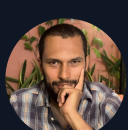
```python
                    self.queue.task_done()
                    raise  # Re-raise to propagate cancellation
                except Exception as e:
                    print(f"Worker {worker_id}: Task {task_id} failed with error: {e}")
                finally:
                    # Mark task as done
                    self.queue.task_done()
            except asyncio.CancelledError:
                # Don't call task_done() here as no task was retrieved
                break

    async def start(self):
        """Start the worker pool"""
        self.running = True
        self.workers = [
            asyncio.create_task(self.worker(i)) for i in range(self.num_workers)
        ]

```



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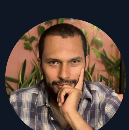
```python
    async def stop(self):
        """Stop the worker pool"""
        self.running = False

        # Give a small timeout for queue to process remaining tasks
        try:
            # Wait for all tasks to complete with a timeout
            await asyncio.wait_for(self.queue.join(), timeout=5.0)
            print("All tasks processed successfully")
        except asyncio.TimeoutError:
            print("Timed out waiting for tasks to complete")

        # Cancel all workers
        for worker in self.workers:
            worker.cancel()

        # Wait for all workers to complete cancellation
        await asyncio.gather(*self.workers, return_exceptions=True)
```



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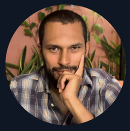
```python
# Example task implementations
def create_data_processing_task(task_id, duration):
    """Creates a data processing task (returns a task function)"""

    async def _task():
        await asyncio.sleep(duration)  # Simulate work
        return f"Data for task {task_id} processed"

    return _task


async def demo_task_queue():
    # Create a task queue
    task_queue = AsyncTaskQueue(num_workers=3)

    # Start the worker pool
    await task_queue.start()

    try:
        # Add tasks with different priorities
```



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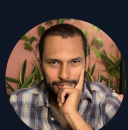
**PYTHON** | **CONCURRENCY (PART 3)**

```python
        for i in range(10):
            priority = random.randint(1, 3)  # 1=high, 3=low priority
            duration = random.uniform(0.5, 1.0)

            # Create a task function (not a coroutine) and add it to the queue
            task = create_data_processing_task(i, duration)
            await task_queue.add_task(task, priority)

            print(f"Added Task {i} with priority {priority}")

            # Small delay between adding tasks to better visualize execution
            await asyncio.sleep(0.1)
    finally:
        # Wait for all tasks to complete and stop the worker pool
        await task_queue.stop()
        print("Worker pool stopped")
```

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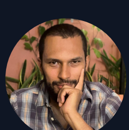
## PYTHON | CONCURRENCY (PART 3)

```python
# Run the demo
if __name__ == "__main__":
    asyncio.run(demo_task_queue())

# Output:
# Added Task 0 with priority 2
# Worker 0: Processing task 1 with priority 2
# Added Task 1 with priority 2
# Worker 1: Processing task 2 with priority 2
# Added Task 2 with priority 3
# Worker 2: Processing task 3 with priority 3
# Added Task 3 with priority 1
# Added Task 4 with priority 1
# Added Task 5 with priority 3
# Added Task 6 with priority 2
# Worker 1: Task 2 completed with result: Data for task 1 processed
# Worker 1: Processing task 4 with priority 1
# Added Task 7 with priority 2
# Added Task 8 with priority 2
# Worker 0: Task 1 completed with result: Data for task 0 processed
```



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```
# Worker 0: Processing task 5 with priority 1
# Added Task 9 with priority 2
# Worker 2: Task 3 completed with result: Data for task 2 processed
# Worker 0: Task 5 completed with result: Data for task 4 processed
# Worker 1: Task 4 completed with result: Data for task 3 processed
# Timed out waiting for tasks to complete
# Worker pool stopped
```

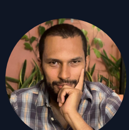
This pattern is useful for building job queues, task schedulers, and work distribution systems.

# 3. Performance Considerations and Best Practices

## 3.1. Benchmarking Concurrent Code

When optimizing Python code with concurrency, it's essential to measure actual performance gains:

```python
import time
import concurrent.futures
```



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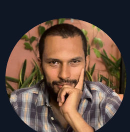
**PYTHON** | **CONCURRENCY (PART 3)**

```python
def benchmark(func, data, executor_class, max_workers=None):
    """Benchmark a function using different execution methods"""
    # Sequential execution (baseline)
    start = time.time()
    sequential_result = [func(item) for item in data]
    sequential_time = time.time() - start
    print(f"Sequential: {sequential_time:.4f}s")

    # Concurrent execution
    start = time.time()
    with executor_class(max_workers=max_workers) as executor:
        concurrent_result = list(executor.map(func, data))
    concurrent_time = time.time() - start
    print(f"Concurrent: {concurrent_time:.4f}s")
    print(f"Speedup: {sequential_time / concurrent_time:.2f}x")


# Example CPU-bound task
```

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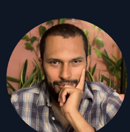
```python
def cpu_task(n):
    """CPU-intensive calculation"""
    return sum(i * i for i in range(n * 100000))


# Example I/O-bound task
def io_task(n):
    """I/O-bound operation (simulated)"""
    time.sleep(0.1)  # Simulate I/O delay
    return n * 2


if __name__ == "__main__":
    # This is critical for multiprocessing to work properly

    # Data for benchmarking
    data = list(range(1, 9))

    # Demo for CPU-bound tasks
    print("CPU-bound task with ProcessPoolExecutor:")
```



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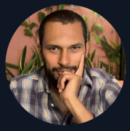
```python
    benchmark(cpu_task, data, concurrent.futures.ProcessPoolExecutor)

    # Demo for I/O-bound tasks
    print("\nI/O-bound task with ThreadPoolExecutor:")
    benchmark(io_task, data, concurrent.futures.ThreadPoolExecutor)


# Output:
# CPU-bound task with ProcessPoolExecutor:
# Sequential: 0.1448s
# Concurrent: 0.1072s
# Speedup: 1.35x

# I/O-bound task with ThreadPoolExecutor:
# Sequential: 0.8338s
# Concurrent: 0.1098s
# Speedup: 7.59x
```



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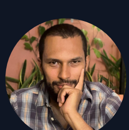
## 3.2. Best Practices for Production Code

For production-grade concurrent Python applications, follow these guidelines:

- **Tool Selection:**
    - I/O-bound → asyncio or threading (asyncio preferred for new code)
    - CPU-bound → multiprocessing
    - Mixed workloads → hybrid approach
- **Resource Management:**
    - Reuse thread/process pools rather than creating new ones
    - Use context managers or ensure proper cleanup in `finally` blocks
    - Monitor memory usage, especially with multiprocessing
- **Error Handling:**
    - Properly catch and handle exceptions in worker functions
    - Use timeouts to prevent hanging operations
    - Implement graceful shutdown mechanisms
- **Avoiding Common Pitfalls:**
    - Thread Safety: Always protect shared resources with locks
    - Deadlocks: Acquire locks in a consistent order
    - Oversubscription: Don't create too many threads or processes



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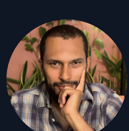
## 4. Conclusion (Part 3)

In this final part of our exploration of concurrency and parallelism in Python, we've covered advanced techniques that build on the fundamentals:

- We've learned how to combine different concurrency models (threading, multiprocessing, and asyncio) to create hybrid solutions that leverage the strengths of each approach.
- We've explored advanced concurrency patterns like the fan-out/fan-in pattern and priority-based task queues that solve real-world parallelization problems.
- We've examined practical benchmarking approaches to quantify performance improvements and make data-driven decisions.
- We've identified best practices and common pitfalls to create production-ready concurrent code.

By mastering these techniques, you can develop Python applications that efficiently utilize system resources, respond quickly to events, and process data in parallel. The key is selecting the right concurrency model for each specific task and combining them when needed for complex applications.
Remember that concurrency is not always the answer—sometimes a simpler sequential solution is more maintainable and even faster for small datasets. Al-



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ways benchmark your code to ensure that your concurrent solution actually improves performance in your specific use case.

## 5. References (Part 3)


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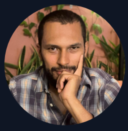
- Translated, Edited and written in collaboration with AI.

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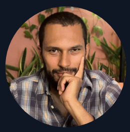
**Python Concurrency Part 2: Asyncio and HTTP Operations**

Learn how to use Python's asyncio for efficient I/O operations, working with HTTP requests, and implementing producer-consumer patterns. Master the modern approach to concurrency.



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# Ready to Level Up Your Python?

You've mastered powerful concurrency techniques.

Apply these patterns in your next project

and watch your code become faster and more efficient!