

# Distributed and Parallel System Final Exam Report

# **Group Member:**

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## **Study Case**

Given a dataset containing NYC Taxi trip records. From the dataset, you are required to:

- Focus on the trip\_duration column.
- Perform sorting and filtering (e.g., filtering values greater than 1000).
- Analyze the processing time using three different approaches:
  - Sequential processing
  - Threading
  - Multiprocessing
- Apply the analysis on four dataset sizes: 25%, 50%, 75%, and 100%.
- Evaluate whether the performance is linear relative to data size.
- Include system specifications (Processor, RAM, Cores) as the basis for performance evaluation.

# **Dataset Description**

The dataset used: train.csv

Main column analyzed: trip\_duration

Filtering condition: Only trip durations greater than 1000 seconds are considered after sorting.

# **Program Workflow**

- a. Library Imports
  - pandas: For data manipulation.
  - time: For measuring execution time.
  - threading: For thread-based parallelism.
  - multiprocessing: For process-based parallelism.
  - platform, psutil: For fetching system specifications.

#### b. Data Loading

The dataset is read using pd.read\_csv().

Only the trip\_duration column is extracted for processing.

#### c. Data Splitting

The dataset is split into four parts to simulate increasing load:

- 25% of the data
- 50% of the data
- 75% of the data
- 100% (full) data

#### d. Data Processing Function

A function process\_data(data) performs two main operations:

- 1. Sorts the data
- 2. Filters values where trip duration > 1000

This function is the core workload that is executed in different processing models.

#### e. Processing Approaches

- 1. Sequential Processing
  - Directly calls process\_data(data) in a single thread and measures the time taken.
- 2. Threaded Processing
  - A thread is spawned to run process\_data(data) and joined to wait for completion.
  - Time is recorded before and after the thread runs.
- 3. Multiprocessing
  - The data is split into chunks based on the number of CPU cores.
  - Each chunk is processed in a separate process using multiprocessing. Process.
  - Results are stored in a shared dictionary.
  - All processes are joined before calculating the total time.

#### f. Execution and Timing

- Each processing model is executed for all data splits (25%–100%).
- Time taken is stored and presented in a comparison table.

#### g. System Information

The following system specs are collected and printed:

- CPU model
- Total RAM
- Number of cores

### Result

```
Processing split: 25%
Processing split: 50%
Processing split: 75%
Processing split: 100%
=== Performance Comparison ===
Split Sequential (s) Threading (s) Multiprocessing (s)
  25%
             0.0272
                      0.0344
                                              60.1730
  50%
             0.0594
                          0.0613
                                              63.9677
  75%
             0.0966
                          0.0902
                                              64.4062
 100%
             0.2085
                          0.1586
                                              66.8411
=== System Info ===
Processor: AMD64 Family 23 Model 96 Stepping 1, AuthenticAMD
RAM: 15.42 GB
CPU Cores: 16
```

# **Analysis & Arguments**

a. Is performance linear with data size?

Sequential and Threaded Processing:

- Performance is mostly linear. As the data size increases, processing time increases gradually.
- Threading does not show significant improvement due to Python's Global Interpreter Lock (GIL). Since the operation is CPU-bound (sorting and filtering), threading offers minimal benefit.

#### Multiprocessing:

- Has higher overhead due to process creation and inter-process communication.
- Only starts showing better performance for larger datasets.
- May not appear linear due to chunking and process overhead.

#### b. Conclusion

- Sequential is fastest for small datasets.
- Multiprocessing becomes useful only for very large data where parallelism offsets process creation overhead.
- Threading is more useful for I/O-bound tasks, not recommended for CPU-bound processing like sorting/filtering.
- Performance is not perfectly linear, especially for multiprocessing, because overhead increases with process count.

#### **Source Code**

https://github.com/Cadburyy/finalproj.git