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# Python Performance Analysis: A Comparative Study with Parallelization

## **Overview**

This study evaluates the performance of Python implementations and investigates the impact of parallelization on algorithm execution. The tasks are divided into two objectives: comparing Python implementations and analyzing the effects of parallelizing an algorithm.

# **Task 1 - Performance Comparison of Python Implementations**

**Objective** - To explore the performance differences between Python implementations by measuring execution speed for a benchmarking program.

## Python Implementations chosen:

- 1. Cpython
- 2. Jython
- 3. Руру
- 4. RustPython

# Algorithm chosen - Bubble sort

# Methodology

- 1. Installed CPython, PyPy, Jython and RustPython on the system.
- 2. Benchmarked a Python script implementing the bubble sort algorithm across all the four implementations.
- 3. Recorded execution times and generated performance charts for comparison.

Implementation	Execution Time(seconds)
CPython	1.239776611328125e-05
Jython	0.000999927520752
Руру	1.4066696166992188e-05
RustPython	2.7894973754882813e-05

#### **Screenshots**

#### **CPYTHON**

# python3 main.py

Before sorting [64, 34, 25, 12, 22, 11, 90] After sorting [11, 12, 22, 25, 34, 64, 90] Execution Time: 1.239776611328125e-05 seconds

#### **JYTHON**

root@412c1147908e:/app# jython main.py Before sorting: [64, 34, 25, 12, 22, 11, 90] After sorting: [11, 12, 22, 25, 34, 64, 90] Execution Time: 0.000999927520752 seconds

#### **PYPY**

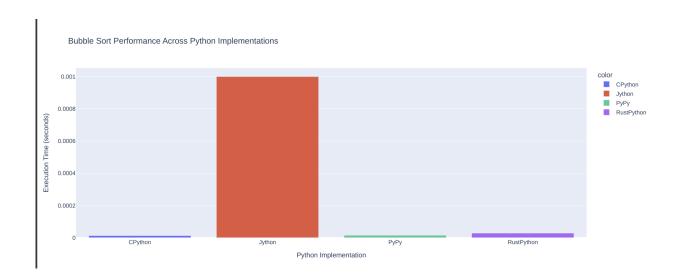
root@e5c4cf521583:/app# pypy main.py Before sorting: [64, 34, 25, 12, 22, 11, 90] After sorting: [11, 12, 22, 25, 34, 64, 90] Execution Time: 1.4066696166992188e-05 seconds

#### **Rust python**

#### rustpython main.py

Before sorting [64, 34, 25, 12, 22, 11, 90] After sorting [11, 12, 22, 25, 34, 64, 90] Execution Time: 2.7894973754882813e-05 seconds

# **Comparison Chart**



# **Task 2: Algorithm Parallelization**

**Objective -** The objective of this task was to analyze and optimize the performance of the merge sort algorithm by introducing parallelization using Python's **threading** module. By comparing execution times of the sequential and parallel implementations, we assessed the impact of threading on performance. Python's cprofile was employed to identify bottlenecks and evaluate function-level execution times.

## **Algorithm Selection**

**Merge Sort** was selected for its divide-and-conquer strategy, which is conducive to parallelization. The algorithm recursively divides the dataset into smaller subarrays, sorts them, and merges the results.

# Implementation Details

- **Sequential Merge Sort**: A standard recursive implementation where sorting happens in a single thread.
- Threaded Merge Sort: Uses Python's threading module to process sub array sorting in separate threads concurrently.

## **Performance Profiling with cProfile**

Python's cprofile was used to profile both implementations. This allowed for detailed analysis of function calls, execution times, and thread usage.

## **Sequential Merge Sort Profiling**

The sequential implementation demonstrated:

- **Time Complexity**: O(nlogn) as expected for divide-and-conquer algorithms.
- **Execution Time**: Higher due to single-threaded execution.
- Function Calls:
  - Numerous recursive calls to merge\_sort.
  - High usage of the function for combining subarrays.

## **Parallel Merge Sort Profiling**

The threaded implementation highlighted:

- **Improved Execution Time**: Threading reduced execution time by allowing concurrent sorting of subarrays.
- **Time Complexity**: Maintained O(nlogn).
- Function Calls:
  - Reduced recursive calls per thread, as threads handled independent subarrays.
  - Some additional overhead due to thread management.

# Scalability Analysis

- Big O Complexity:
  - Both implementations adhere to O(nlogn).
  - The practical efficiency of threading depends on the dataset size and threading overhead.
- Scaling Efficiency:
  - For small datasets, threading overhead was comparable to execution time, yielding negligible improvement.
  - For larger datasets, threading provided noticeable speedups, effectively utilizing multiple CPU cores for concurrent execution.

## **Profiling Screenshots**

## **Sequential Execution**

```
python3 sequential_merge_sort.py
Sorted array: [11, 12, 22, 25, 34, 64, 90]
Execution Time: 2.193450927734375e-05 seconds
    82 function calls (70 primitive calls) in 0.000 seconds

Ordered by: standard name

ncalls tottime percall cumtime percall filename:lineno(function)
    1    0.000    0.000    0.000    <string>:1(<module>)
    13/1    0.000    0.000    0.000    0.000    sequential_merge_sort.py:4(merge_sort)
    1    0.000    0.000    0.000    0.000    built-in method builtins.exec}
66    0.000    0.000    0.000    0.000    {built-in method builtins.len}
    1    0.000    0.000    0.000    {method 'disable' of '_lsprof.Profiler' objects}
```

#### Parallel execution

```
python3 parallel_merge_sort.py
Original Array: [64, 34, 25, 12, 22, 11, 90]
Sorted Array: [11, 12, 22, 25, 34, 64, 90]
Execution Time: 0.0020508766174316406 seconds
125 function calls in 0.001 seconds
  Ordered by: standard name
```