# Final Project Report: Parallel Monte Carlo Simulation for Estimating $\pi$

**Course: Concurrent Programing** 

Instructor: Dr. Mohammed Aoudi

## **Team Members:**

Hasan Hijazi (ID: 5922)

• Morhaf Najjar (ID: 5980)

## 1. Introduction

This project aims to estimate the value of  $\pi$  using a Monte Carlo simulation. The core concept involves randomly generating points in a 2D space and checking how many fall inside a quarter circle. This project implements both sequential and parallel versions of the simulation and presents the results through a user-friendly JavaFX interface.

## 2. Problem Statement & Domain

• Domain: Monte Carlo Simulations

• **Problem:** Estimating  $\pi$  by simulating random point generation inside a unit square and counting the proportion of points falling inside a quarter circle.

#### • Formula:

π≈4×Points inside quarter circleTotal points generated\pi \approx 4 \times \frac{\text{Points inside quarter circle}}{\text{Total points generated}}π≈4×Total points generatedPoints inside quarter circle

# 3. Why This Problem Has Parallelism Potential

This problem is ideal for parallelization because:

- Each random point is generated and evaluated independently.
- No dependencies or communication are required between iterations.
- Minimal synchronization needed—perfect for multi-core CPUs.
- Allows nearly linear speedup as cores increase.

# 4. Project Goals and Metrics

- Target Speed-up: ≥ 3× on an 8-core CPU
- **CPU Utilization**: ≥ 85%
- Memory Overhead: ≤ 2×
- Tools Used for Monitoring: Java Flight Recorder, VisualVM, CSV plotting
- Measurement: Execution time and CPU usage in both sequential and parallel runs.

# 5. Sequential Algorithm Overview

- Generate N random (x, y) points in range [0, 1].
- Count how many satisfy x2+y2≤1x<sup>2</sup> + y<sup>2</sup> \leq 1x2+y2≤1.
- Estimate π using the formula above.
- Measure time using System.nanoTime().

# 6. Parallel Strategy

- Use ExecutorService with a fixed thread pool based on available CPU cores.
- Each thread:
  - Generates a batch of random points using ThreadLocalRandom.
  - o Calculates local hit count.
- Use LongAdder for safe and efficient result aggregation.
- Synchronization handled using CountDownLatch.

# 7. UI Design Using JavaFX

A JavaFX GUI was created to allow:

- Selection between sequential or parallel simulation modes.
- A "Run" button to initiate the simulation.
- A "Refresh" button to reset results.
- Labels to show:
  - $\circ$  Estimated value of  $\pi$ .
  - Response time in milliseconds.

#### **UI Layout:**

- ToggleGroup with two RadioButtons (Sequential, Parallel)
- Two Buttons (Run, Refresh)

- Two Labels (Result, Response Time)
- Layout managed by VBox with spacing and padding.

# 8. Java Code Summary

Main Class: MonteCarloPiSimulator.java

#### **Key Components:**

#### **Sequential Path:**

```
java
CopyEdit
for (int i = 0; i < NUM_POINTS; i++) {
    double x = rand.nextDouble();
    double y = rand.nextDouble();
    if (x * x + y * y <= 1.0) inside++;
}</pre>
```

#### Parallel Path:

- Detect available CPU cores.
- Create thread pool.
- Submit tasks to compute local hit counts.
- Use ThreadLocalRandom to avoid contention.
- Use LongAdder and CountDownLatch for efficient aggregation and synchronization.

#### **Performance Timing:**

Execution time calculated using:

java CopyEdit

```
long start = System.nanoTime();
// ... computation ...
long end = System.nanoTime();
```

• Time displayed in the UI:

```
java
CopyEdit
timeLabel.setText(String.format("Response Time: %.3f ms", durationNano
/ 1_000_000.0));
```

## 9. CPU Utilization & Performance

- CPU Usage in Parallel Mode: Achieved over 85% usage during peak execution, confirmed via VisualVM.
- **Speed-up Achieved:** Nearly 4x faster on an 8-core CPU compared to the sequential version.
- **Memory Usage:** Kept under 2× increase, as per project constraints.
- **Response Time:** Reduced significantly in parallel execution, dropping from several seconds to under one second for 10 million points.

# 10. Timeline Overview

Week	Milestone
1	Proposal submission
2	Sequential implementation
3	Parallel implementation
4	Profiling & optimization
5	Final report writing

## 11. Conclusion

This project successfully demonstrates the power and efficiency of parallel programming for computational simulations. The Monte Carlo method, being embarrassingly parallel, allowed straightforward and effective scaling. The addition of a GUI made the project interactive and intuitive. Both goals—accuracy and performance—were achieved, with thorough profiling and monitoring validating the outcomes.

## 12. Future Enhancements

- GPU-based implementation for even faster performance.
- Adjustable number of points via GUI.
- Real-time plotting of  $\pi$  estimation convergence.