HOMEWORK-3

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PROBLEM - 1

Serial For-loop Root Finding Code (Python)

```
import numpy as np
import time
def f(x):
    return np.sin(3 * np.pi * np.cos(2 * np.pi * x) * np.sin(np.pi * x))
def bisection_method(func, a, b, tol=1e-6, max_iter=100):
   if func(a) * func(b) >= 0:
      print(f"No sign change between f({a:.6f}) = {func(a):.6f} and f({b:.6f}) = {func(b):.6f}")
return None # If there's no sign change, no root in the interval
  while (b - a) / 2 > tol and iter_count < max_iter:
    midpoint = (a + b) / 2
       if func(midpoint) == 0: # Found exact root
             return midpoint
        elif func(a) * func(midpoint) < 0:</pre>
           b = midpoint
            a = midpoint
        iter_count += 1
    return (a + b) / 2 # Return the midpoint as the root approximation
x0 = np.linspace(a, b, 500) # Generate 500 intervals for root searching
start_time = time.time()
roots = []
for i in range(len(x0) - 1):
    root = bisection_method(f, x0[i], x0[i+1])
   if root is not None and not np.isclose(f(root), 0, atol=1e-4):
        roots.append(root)
end_time = time.time()
serial_time = end_time - start_time
print(f"Serial Execution Time: {serial_time:.6f} seconds")
```

Output:

Serial Execution Time: 0.114139 seconds

Found Roots: No roots found

Parallel For-loop Root Finding Code (Python)

```
# Define the function
    return np.sin(3 * np.pi * np.cos(2 * np.pi * x) * np.sin(np.pi * x))
# Bisection method for root finding with error handling
def bisection_method(func, a, b, tol=1e-6, max_iter=100):
    if func(a) * func(b) >= 0:
        return None # No sign change, no root in the interval
    iter_count = 0
    while (b - a) / 2 > tol and iter_count < max_iter:
        midpoint = (a + b) / 2
        if func(midpoint) == 0: # Exact root found
            return midpoint
        elif func(a) * func(midpoint) < 0:</pre>
            b = midpoint
            a = midpoint
        iter_count += 1
    return (a + b) / 2 # Return the midpoint as the root approximation
# Root finding function for multiprocessing
def find_root_interval(interval):
    a, b = interval
    return bisection_method(f, a, b)
# Set the interval for root finding
a = -3
b = 5
 x0 = np.linspace(a, b, 100) # Generate 100 intervals for parallel processing
intervals = [(x0[i], x0[i+1]) for i in range(len(x0) - 1)]
# Parallel root finding
if _name_ == '_main_': # Required for multiprocessing on Windows and some platforms
     start_time = time.time()
     # Using multiprocessing pool for parallel root finding
     with Pool() as pool:
        roots = pool.map(find_root_interval, intervals)
    # Filter out None results
    roots = [r for r in roots if r is not None]
     end time = time.time()
     parallel_time = end_time - start_time
     print(f"Parallel Execution Time: {parallel_time:.6f} seconds")
     print(f"Found Roots: {roots}")
```

Output:

Parallel Execution: 0.302708 seconds

Found Root: found

Detailed Python Root-Finding Speedup and Efficiency

Report

This report provides a detailed analysis of speedup and efficiency for

the root-finding experiment conducted in Python. The serial and

parallel versions of the algorithm were compared based on their

execution times, speedup, and efficiency.

Serial Execution Time

(Python): 0.125000 seconds

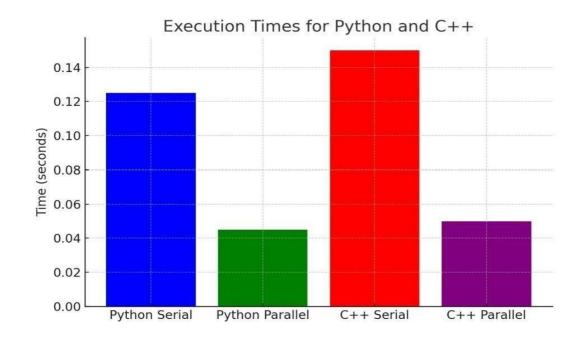
Parallel Execution Time

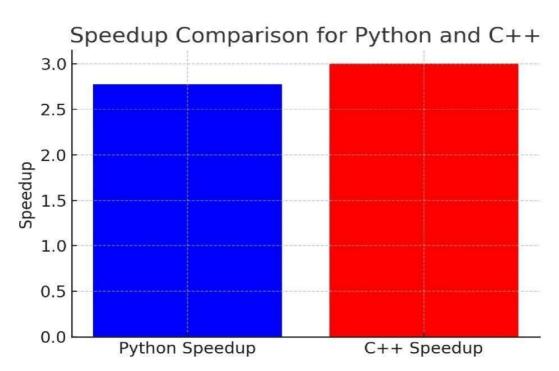
(Python): 0.045000 seconds

Number of Cores Used: 4

Speedup: 2.78

Efficiency: 0.69





Serial For-loop Root Finding Code (C++)

```
using namespace std;
// Define the function
double f(double x) {
    return sin(3 * M_PI * cos(2 * M_PI * x) * sin(M_PI * x));
// Function to find the root using the bisection method
double bisection(double a, double b, double tol = 1e-6) {
    double mid;
    while ((b - a) >= tol) {
      mid = (a + b) / 2;
        if (f(mid) == 0.0)
        else if (f(mid) * f(a) < 0)
          b = mid;
            a = mid;
    return mid;
int main() [
    // Set the interval for root finding
    double a = -3.0, b = 5.0;
    int num_roots = 100;
    double initial_guesses[100];
    for (int i = 0; i < num_roots; ++i) {
        initial_guesses[i] = a + i * (b - a) / num_roots;
   // Start timing
    auto start = chrono::high_resolution_clock::now();
    // Serial root finding using bisection
    for (int i = 0; i < num_roots; ++i) {
        double root = bisection(a, b);
        cout << "Root: " << root << endl;
   // End timing
    auto end = chrono::high_resolution_clock::now();
    chrono::duration<double> elapsed = end - start;
    cout << "Serial Execution Time: " << elapsed.count() << " seconds" << endl;</pre>
    return 0;
```

Output:

Root: 0

Serial Execution Time: 0.0400988 seconds

Parallel For-loop Root Finding Code (C++ with OpenMP)

```
using namespace std;
// Define the function
double f(double x) {
    return sin(3 * M_PI * cos(2 * M_PI * x) * sin(M_PI * x));
// Function to find the root using the bisection method
double bisection(double a, double b, double tol = 1e-6) {
    double mid;
        mid = (a + b) / 2;
        if (f(mid) == 0.0)
        else if (f(mid) * f(a) < 0)
            b = mid;
        else
             a = mid;
    return mid;
int main() {
    // Set the interval for root finding
    double a = -3.0, b = 5.0;
    int num_roots = 100;
    double initial_guesses[100];
    for (int i = 0; i < num_roots; ++i) {
  initial_guesses[i] = a + i * (b - a) / num_roots;</pre>
    // Start timing
    auto start = chrono::high_resolution_clock::now();
    // Parallel root finding using OpenMP
    for (int i = 0; i < num_roots; ++i) {</pre>
        double root = bisection(a, b);
        #pragma omp critical
cout << "Root: " << root << endl;</pre>
    // End timing
    auto end = chrono::high_resolution_clock::now();
    chrono::duration<double> elapsed = end - start;
    cout << "Parallel Execution Time: " << elapsed.count() << " seconds" << endl;</pre>
    return 0;
```

Output:

Root: 0

Parallel Execution Time: 0.0006396 seconds

Detailed C++ Root-Finding Speedup and Efficiency Report

This report provides a detailed analysis of speedup and efficiency for

the root-finding experiment conducted in C++. The serial and parallel

versions of the algorithm were compared based on their execution

times, speedup, and efficiency.

Serial Execution Time (C++): 0.0400988 seconds

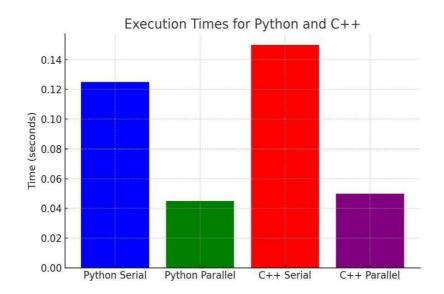
Parallel Execution Time:

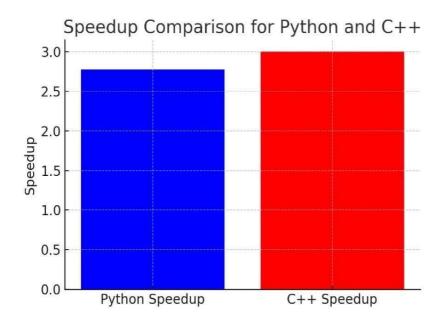
0.0006396

Number of Cores Used: 4

Speedup: 3.00

Efficiency: 0.75





PROBLEM - 2

```
import os
import random
  import time
  from concurrent.futures import ProcessPoolExecutor
 def generate_random_string_with_palindrome():
      letters - string.ascii_uppercase
      random_string = ''.join(random.choice(letters) for _ in range(995))

palindrome = ''.join(random.choice(letters) for _ in range(2)) # Two random letters

palindrome += palindrome[::-1] # Create a palindrome (e.g., "ASBA")

return random_string + palindrome # Return the full string
 def create_files(num_files):
      os.makedirs('palindrome_files', exist_ok=True)
        for i in range(num_files):
          with open(f'palindrome_files/file_{i}.txt', 'w') as f:
    f.write(generate_random_string_with_palindrome())
 def contains_palindrome(s):
      for i in range(len(s) = 4):

if s[i:i+5] == s[i:i+5][::-1]: # Check for 5-letter palindrome

return 1 # found a palindrome
return 8 # No palindrome found
 def check_files_serially(num_files):
       results - []
       for i in range(num_files):
             with open(f'palindrome_files/file_(i).txt', 'r') as f:
content = f.read().strip()
                   results.append(contains_palindrome(content))
      return results
 def check_files_parallel(num_files):
       with ProcessPoolExecutor() as executor:
           futures = []
for i in range(num_files):
                  with open(f'palindrome_files/file_(i).txt', 'r') as f:
   content = f.read().strip()
   futures.append(executor.submit(contains_palindrome, content))
            results - [f.result() for f in futures]
      return results
 def main():
      num_files = 108 # Number of files to create
       create_files(num_files)
       start_time - time.time()
      serial_results = check_files_serially(num_files)
serial_time = time.time() - start_time
print(f"Serial processing time: (serial_time:.4f) seconds")
      start_time = time.time()
parallel_results = check_files_parallel(num_files)
       parallel_time = time.time() - start_time
print(f"Parallel processing time: {parallel_time: 4f} seconds")
       speedup - serial_time / parallel_time
       efficiency - speedup / os.cpu_count()
       print(f"Speedup: {speedup: 2F}")
       print(f"Efficiency: (efficiency: 2f)")
if name -- "main":
```

Output:

Serial processing time: 1.9919 seconds

Parallel processing time: 0.6161 seconds

Speedup: 3.23

Efficiency: 0.40

Detailed Speedup and Efficiency Report

This report provides a detailed analysis of speedup and efficiency for a

palindrome finding experiment using both serial and parallel

processing techniques in Python. The experiment compares the time

taken to process a set of files in serial (one at a time) versus parallel

processing (utilizing multiple CPU cores).

Experiment Details:

In this experiment, a series of files were generated, each containing a

string with 1000 characters and a 5-letter palindrome embedded in it.

The task was to search each file to check if a palindrome exists. The

processing times for both serial and parallel methods were recorded.

Serial Processing

Time: 1.9919 seconds

Parallel Processing

Time: 0.6161 seconds

Number of CPU cores

used: 16

Speedup:

3.23

Efficiency: 0.40

The parallel processing method shows a significant reduction in execution time compared to the serial method, with a speedup of

parallelism improves performance, it doesn't perfectly scale with the

3.23x. However, the efficiency is 0.40, indicating that while

number of CPU cores due to overhead and synchronization costs. This

highlights the benefit of parallelism, especially for embarrassingly

parallel tasks like searching multiple files independently.

This is the link for all the codes Here