HPC OPENMP TUTORIAL 4 REPORT CS22B2015 – HARSHITH B

The parallel code for reduction and critical sections is provided in the main.c file. The key section of the code performs the following steps:

- 1. Threads are allocated in powers of 2 (1, 2, 4, 8, 16, 32, 64, 128) within a loop, and the sum is calculated using the OpenMP pragma reduction.
- 2. Two datasets consisting of approximately 15 million double-precision floating-point numbers is generated using Python and stored in the output1.txt and output2.txt files.
- 3. The threads versus time data is recorded into a text file. This data is then processed, and the respective results are plotted and analyzed using results.py.

1. Parallel Code Segment for Multiplication and Addition

```
printf("Vector Dot Product\n");
for(int t = 0; t < num options; t++){</pre>
   int num_threads = thread counts[t];
   omp set num threads(num threads);
   double dot product = 0.0;
   start = omp get wtime();
   #pragma omp parallel
       double local sum = 0.0;
       #pragma omp for
        for(int i = 0; i < count; i++){
           local sum += arr1[i] * arr2[i];
       #pragma omp critical
       dot product += local sum;
   end = omp_get_wtime();
   printf("%d\t%.6f\t%.12f\n", num_threads, end - start, dot product);
    fprintf(f dot, "%d %lf\n", num threads, end - start);
```

2. Terminal Output

```
(venv) harshith@harshithb:~/Projects /SEM 6/HPC/tutorial-4$ ./main
Vector Dot Product
1
        0.070798
                        3758873747356410368.0000000000000
2
        0.040675
                        3758873747355359232.0000000000000
        0.025236
                        3758873747355091968.0000000000000
        0.021679
                        3758873747355244544.0000000000000
                        3758873747355185664.0000000000000
8
        0.021532
10
        0.038120
                        3758873747355194368.0000000000000
12
                        3758873747355226624.0000000000000
        0.030346
16
                        3758873747355228672.0000000000000
        0.023518
20
        0.023104
                        3758873747355247104.0000000000000
        0.016972
                        3758873747355226624.0000000000000
32
64
        0.023649
                        3758873747355222016.0000000000000
```

3. Thread vs Time Plots and Observations

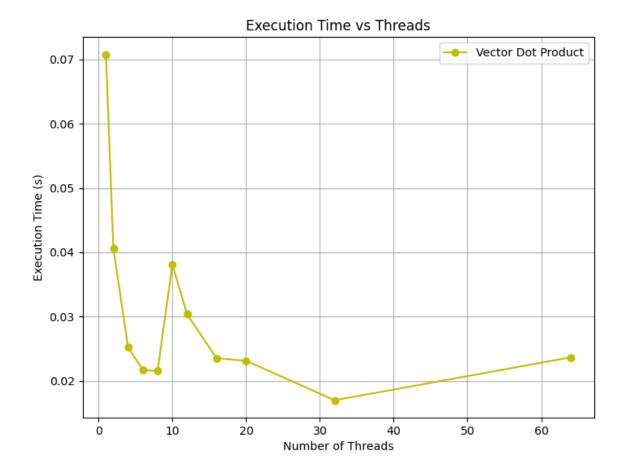
Vector Dot Product:

The execution time for the vector dot product decreases sharply as the number of threads increases from 1 to around 8 threads, indicating effective parallelization in this range.

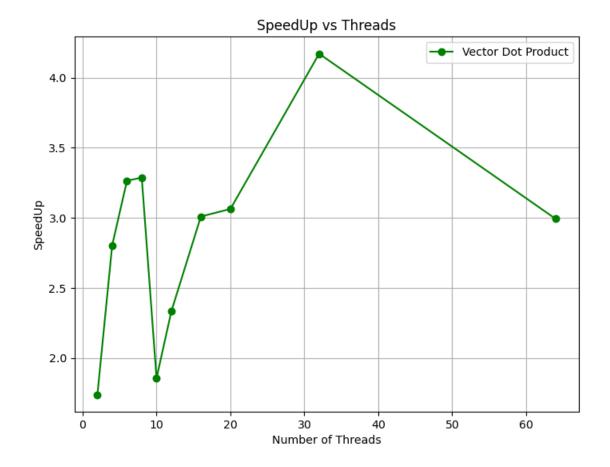
The best performance appears to occur around 32 threads, where the execution time is at its lowest.

Beyond 32 threads, the execution time increases slightly, likely due to overheads such as thread contention, synchronization, and communication costs among threads.

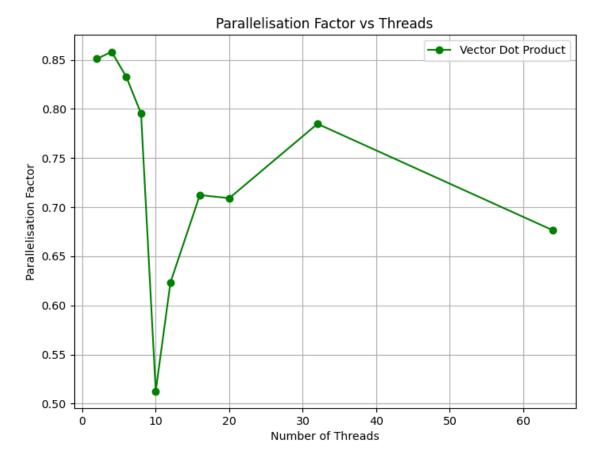
This suggests that the vector dot product scales well with increasing threads up to a point, after which diminishing returns and overhead costs start to dominate the performance.



4. SpeedUp vs Processors (SpeedUp == T(1) / T(n))



5. Parallelization Fraction and Inference



Parallel Vector Dot Product:

- The parallelization factor peaks initially but declines as the number of threads increases, stabilizing around **0.855**.
- This indicates diminishing returns for parallelization beyond a certain threshold (approximately 4 **threads**).
- Overhead due to thread management likely impacts performance at higher thread counts.