## HIGH PERFORMANCE PROGRAMMING ASSIGNMENT 3

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# 1 Introduction of Galaxy Simulation

### 1.1 The N-Body Problem

The N-Body Problem is a problem in physics that involves considering N point masses moving in a Newtonian reference system, with all the points influenced only by their mutual gravitational attraction [1]. According to Newton's law of gravitation in two dimension states, the force acting on particle i due to particle j is expressed by:

$$\mathbf{f}_{ij} = -\frac{Gm_im_j}{r_{ij}^3}\mathbf{r}_{ij} = -\frac{Gm_im_j}{r_{ij}^2}\hat{\mathbf{r}}_{ij}$$

where:

G is the gravitational constant;

 $m_i$  and  $m_j$  are the masses of the particles;

 $r_{ij}$  is the distance between the particles;

 $\mathbf{r}_{ij}$  is the vector that represents the position of particle *i* relative to particle *j*;  $\hat{\mathbf{r}}_{ij}$  is the normalized distance vector.

If  $\mathbf{e}_x$  and  $\mathbf{e}_y$  are unit vectors in the x and y directions respectively, then we have:

$$\mathbf{r}_{ij} = (x_i - x_j)\mathbf{e}_x + (y_i - y_j)\mathbf{e}_y$$
$$r_{ij}^2 = (x_i - x_j)^2 + (y_i - y_j)^2$$
$$\mathbf{r}_{ij} = r_{ij}\hat{\mathbf{r}}_{ij}$$

Considering the distribution of N particles, We can straight-forwardly calculate the force exerted on particle i by the other N-1 particles (using C-style indexing), which is also the only force acting on particle i:

$$\mathbf{F}_i = -Gm_i \sum_{j=0, j \neq i}^{1} \frac{m_j}{r_{ij}^3} \mathbf{r}_{ij}$$

According to this formula, we can see that there is an inherent instability when  $r_{ij} \ll 1$ . Therefore, within the Plummer sphere model, we define a slightly modified force:

$$\mathbf{F}_{i} = -Gm_{i} \sum_{j=0, j \neq i}^{1} \frac{m_{j}}{(r_{ij} + \epsilon_{0})^{3}} \mathbf{r}_{ij}$$

where  $\epsilon_0$  is a small number and we will use  $10^{-3}$  in this case.

To preserve global properties of the gravitational system (e.g. total energy), we should use the symplectic Euler time integration method. The velocity  $u_i$  and position  $x_i$  of particle i can be updated with:

$$\mathbf{a}_i = \frac{\mathbf{F}_i^n}{m_i}$$
$$\mathbf{u}_i^{n+1} = \mathbf{u}_i^n + \Delta t \mathbf{a}_i$$
$$x_i^{n+1} == x_i^n + \Delta t \mathbf{u}_i^{n+1}$$

where:

 $\Delta t$  is the time step size;

 $\mathbf{a}_i$  is the acceleration of particle *i*.

The time complexity is determined by the number of operations required for each time step. If we have N particles, we need to calculate the gravitational force exerted on each particle by the other N-1 particles. According to the formula, calculating the force on particle i takes O(N-1) time. Then, we need to repeat this calculation for all N particles. Therefore, the total computation time is O(N\*(N-1)), which simplifies to  $O(N^2)$ . For large values of N, the computational cost becomes significantly higher.

#### 1.2 Problem Setting

In this simulation, we will implement a code that calculates the evolution of N particles in a gravitational simulation in two spatial dimensions, which approximates the evolution of a galaxy.

We have a file including an initial set of particles, with the particle masses and initial positions and velocities. And the positions of particles are in a  $L \times W$  dimensionless domain (Use L = W = 1),which means that the x and y values will be between 0 and 1.

After simulating for a specified number of time steps, we will obtain a results file containing the final masses, positions, and velocities. These results can be visualized through graphics to observe the evolution.

Considering that our simulation sample is relatively small, and the total particles mass is also small, to make the gravitational effect more noticeable, we need gravity to scale inversely with the number of bodies. Thus, we set:

$$G = 100/N$$

$$\epsilon_0 = 10^{-3}$$

$$\Delta t = 10^{-5}$$

#### 2 The Solution

#### 2.1 Data structures

The binary input file format we used consist of a sequence of double numbers giving the mass, position x and y, velocity in the x and y direction, and "brightness" of each particle. The result file will have the same format. So the total file size will be N \* 6 \* sizeof(double).

Among these properties, mass and "brightness" remain unchanged before and after execution.

Based on the file format, we construct the following structure to store the properties of the particles, and use an array to store all the particles:

```
typedef struct Particle {
                      // position x (assumed to be in [0,1])
2
      double x;
      double y;
                      // position y (assumed to be in [0,1])
3
                      // mass
      double mass;
      double v_x;
                      // velocity x
                      // velocity y
      double v_y;
      double bright; // brightness (can be used as a color
          value)
  } Particle;
```

Listing 1: Particle structure

#### 2.2 Code structure

```
// Measuring time, learn from Lab5_Task4
static double get_wall_seconds() {
   struct timeval tv;
   gettimeofday(&tv, NULL);
   return tv.tv_sec + (double)tv.tv_usec / 1000000;
}
```

Listing 2: Measuring execution time using gettimeofday

The get\_wall\_seconds() function is used to measure the program's wall time, which refers to the cost time from system startup to the current moment, which is used for performance testing.

```
exit(1);
6
            }
            size_t n = fread(particle, sizeof(Particle), N, file
            if(n != (size_t)N) {
                fprintf(stderr, "Error reading file. Expected %d
10
                    □particles, □got □ %zu. \n", N, n);
                exit(1);
11
            }
12
13
            fclose(file);
       }
15
```

The function readfile is used to read particle data from a binary file. It opens the specified file and reads the corresponding number N of particles. It also performs error checking in case the file cannot be opened or the expected data is not read properly.

```
// Perform one simulation step (update accelerations,
      velocities, positions)
   // Accept a_x and a_y as arguments, which avoids repeated
      memory allocation, improving performance. It comes from
      Deepseek.
   // Here we use inline to reduce function calls, learned from
       Lab5_Task5
   static inline void simulationStep(Particle *restrict
      particle, const int N, const double dt, double *restrict
      a_x, double *restrict a_y) {
       // F_i = -G * m_i * (m_j * r) / (r + eps)^3, precompute
          G to avoid redundant computations
       const double G = 100.0 / N;
6
       const double eps = 1e-3;
       // Reset accelerations to zero, learned from Lab5_Task4
       memset(a_x, 0, N * sizeof(double));
10
       memset(a_y, 0, N * sizeof(double));
12
       for (int i = 0; i < N; i++) {</pre>
13
           for (int j = 0; j < N; j++) {
14
               if (i == j)
15
                   continue; // Skip self-interaction
               // r_{ij} = (x_i - x_j) + (y_i - y_j)
               double r_x = particle[i].x - particle[j].x;
18
               double r_y = particle[i].y - particle[j].y;
19
               double r
                         = sqrt(r_x * r_x + r_y * r_y);
20
               double r_e = (r + eps) * (r + eps) * (r + eps);
21
               // Acceleration, a = -G * m_j * r / (r + eps)^3
22
               double temp = -G * particle[j].mass / r_e;
               a_x[i] += temp * r_x;
               a_y[i] += temp * r_y;
25
```

```
}
26
27
28
       for (int i = 0; i < N; i++) {</pre>
29
            // Velocity, v = dt * a
            particle[i].v_x += dt * a_x[i];
31
            particle[i].v_y += dt * a_y[i];
32
            // Position, x = dt * V
33
            particle[i].x
                            += dt * particle[i].v_x;
34
            particle[i].y
                            += dt * particle[i].v_y;
35
       }
   }
37
```

The simulationStep function is designed to update the state of each particle in the galaxy simulation for one time step.It use the modified force equation that corresponds to Plummer spheres and the symplectic Euler time integration to calculate and update the accelerations, velocities, and positions.

```
Write particles to a binary file, which we use chatgpt to
        modify our code
   void binary_particle(const char *filename, const Particle *
2
      particle, const int N) {
       FILE *file = fopen(filename, "wb");
3
       if (file == NULL) {
4
           fprintf(stderr, "Cannot open output file: "%s\n",
5
               filename);
           exit(1);
6
       }
       fwrite(particle, sizeof(Particle), N, file);
9
       fclose(file);
10
```

The binary\_particle function is used to save the current state of the particles (e.g. position, velocity, mass, etc.) into a binary file, making it easier for computation or further analysis.

```
// Main simulation loop
   while(step < nsteps) {</pre>
2
       simulationStep(particles, N, dt, a_x, a_y);
3
4
       step++;
5
       if (graphicsEnabled == 1) {
8
           ClearScreen();
9
            // when I use the input data's brightness, which is
10
               out of range, so we
            // need to normalize brightness to [0, 1]
11
            float min_bright = 0.0f;
12
            float max_bright = 1.0f;
13
```

```
14
           for (int i = 0; i < N; i++) {</pre>
15
                if (particles[i].bright < min_bright) min_bright</pre>
16
                     = particles[i].bright;
                if (particles[i].bright > max_bright) max_bright
                     = particles[i].bright;
            }
18
19
            for (int i = 0; i < N; i++) {</pre>
20
                particles[i].bright = (particles[i].bright -
21
                    min_bright) / (max_bright - min_bright);
22
            // Draw each particle as a circle.
23
            // Map particle coordinates (assumed in [0,1]) to
24
                screen coordinates, we learn it from chatgpt
            for (int i = 0; i < N; i++) {</pre>
25
                float screenX = (float)(particles[i].x *
                    windowWidth);
                float screenY = (float)(particles[i].y *
27
                    windowHeight);
                // making the radius proportional to the
28
                    particle's mass, which comes from deepseek
                float radius = 5.0f * (float)particles[i].mass ;
29
                // use particle's brightness field for rendering
                    , which comes from deepseek
                float color = (float)particles[i].bright;
31
32
                DrawCircle(screenX, screenY, (float)windowWidth,
33
                     (float)windowHeight, radius, color);
            }
34
            Refresh();
35
            // control the simulation speed, add a small delay,
36
               which comes from chatgpt
           usleep(20000);
37
38
            if (CheckForQuit()) break; // Exit loop if quit
39
                signal received
       }
40
41
```

This code snippet is the main simulation loop in the main function. It continues looping while step is less than the given N, simulating the particles' state over multiple steps, and displaying the results in a graphical window if needed.

### 3 Performance and Discussion

#### 3.1 Implementation Time Measurement

```
int main(int argc, char *argv[]) {
   double startTime1 = get_wall_seconds();
   if (argc != 6) {
       printf("There_should_be_galsim_N_filename_nsteps_delta_t_graphics\n");
       return 1;
   }
}
```

The variable startTime1 is initialized at the beginning of the main function, where the get\_wall\_seconds function is called to obtain the current wall time. And the program then continues to execute.

```
double totalTimeTaken = get_wall_seconds() - startTime1
;
printf("totalTimeTaken_=_%f\n", totalTimeTaken);
return 0;
```

The variable totalTimeTaken calculates the difference between the current time at the end of the program execution and startTime1, which means the total running time of the program.

# 3.2 Experiments and Performance

Particles Number (N)	Runtime (s)		
10	0.0073		
20	0.0088		
30	0.0083		
40	0.0093		
50	0.0110		
60	0.0121		
70	0.0148		
80	0.0146		
90	0.0179		
100	0.0202		
150	0.0370		
200	0.0566		
300	0.1106		
400	0.2025		
500	0.3123		
600	0.4582		
700	0.6090		
800	0.7747		
900	0.9843		
1000	1.2390		
1500	2.6919		
2000	4.7969		
3000	10.7788		
4000	19.2148		
5000	29.2542		
6000	42.4580		
7000	42.4580		
8000	75.4925		
9000	95.0584		
10000	117.1811		

Table 1: Simulation Runtimes for Different Particle Numbers

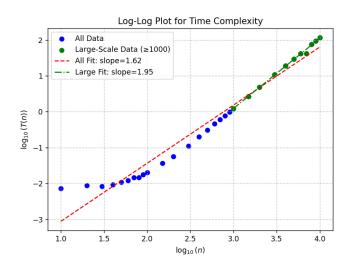


Figure 1: Log plot of Time complexity

From the log-log plot, we can see that the slope of all particle number is smaller than the expected slope, this is because At small particle counts, the entire dataset fits into the L1/L2 cache, resulting in extremely fast execution. As the particle count increases and the data moves into slower memory (L3 cache or RAM), the performance starts to reflect the expected  $O(n^2)$  (the explanation of cache effects we learned from Chatgpt).

We calculate the running time for N=500,1000,2000 and 5000. From the figure above, we can see the relationship between the number of particles N and the simulation time. The curve follows a quadratic trend, indicating that the algorithm has a time complexity of  $O(n^2)$ .

#### 3.3 Optimizations of Simulations

The optimization attempts as below:

**inline**: show a significant improvement, which reaches the improvement rate (74.59%), beacuse it can reduce functions calls.

**loop unrolling**: improve the performance by (25.07%), beacuse it can reduce loop overhead.

**buffer**: improve the performace slightly, it can store array contiguously in memory.

Techniques like **restrict**, **const**, **-ffast-math**, and **-march=native** have minimal impact.

We applied the optimization methods mentioned above. The following table shows a comparison of the running times for the optimized and original algorithms with different values of N.

The table above compares the performance improvement of different optimiza-

Table 2: Performance Improvement with Different Optimization Techniques (1500 Particles)

	Inline	Data Locality (Buffer)	Loop Unrolling	restrict	const	-ffast-math
Original	10.778086	2.750036	3.666598	2.782311	2.75322	2.751732
Optimization	2.738427	2.730311	2.747539	2.744423	2.74132	2.744206
Improvement Rate	74.59%	0.75%	25.07%	1.36%	0.77%	0.27%

tion methods applied to 1500 particles, all optimizations improve the performance.

## 3.4 Computer environment

CPU Model:Intel i7-14700HX

Compiler Version: gcc (Ubuntu 11.4.0-1ubuntu1 22.04) 11.4.0

Compiler Optimization option:

- -O3 (for high-level optimizations)
- -march = native
- -ffast-math
- -funroll-loops (loop unrolling)

# References

- [1] Meyer, K. R., Hall, G. R., & Offin, D. C. (2009). *Introduction to Hamiltonian dynamical systems and the N-body problem* (2nd ed.). Springer Science+Business Media. https://doi.org/10.1007/978-0-387-09724-4 p. 27
- [2] High Performance Programming. (2022). Assignment 3: The Gravitational N-Body Problem. Uppsala University, Spring 2022.