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Subject: Space Simulation Technical Interview Project

The example code provided is a particle system simulation where each particle affects every other particle. This is an N-body problem, which the implemented algorithm solves with a run-time complexity of O(n^2) as it uses a nested loop to compare each particle with every other. The solution is best optimized using the <u>Barnes-Hut</u> algorithm with a run-time complexity of O(n log n).

Note that the original code and iterative optimization approaches are archived via the following github repository.

UsefulEngines/NBodyDemos: Experimenting with C++ 20 standard library parallel constructs. (https://github.com/UsefulEngines/NBodyDemos)

An initial review of the solution indicates the following *semantic* observations:

Realism: A more realistic gravitational N-Body solution would account for additional variables including Particle Acceleration and perhaps Kinetic or Potential Energy. Here is example Particle struct with these member variables.

```
struct Particle {
  double x, y, z;  // position
  double mass;
  double vx, vy, vz;
  double ax, ay, az;  // adding Acceleration
  double kinetic_energy;  // adding Kinetic Energy
};
```

The classic N-Body solution demo leverages the acceleration inherent in Gravitational attraction (G) between large particles with Force calculations as follows.

With Newtonian mechanics, for every action, there is an equal and opposite reaction. This enables us to half the total number of *Force* calculations. If the force on particle q is due to particle k, Fqk, then the force on k due to q is -Fqk.

Initial Optimization Ideas

Use more efficient data structures: The Particles are stored in a std::vector. The vector provides contiguous memory, which is beneficial for cache locality and SIMD operations. It is usually the default choice for a collection of elements unless there is a specific reason to use another container type. Memory layout of the Particles data-structure requires consideration of CPU versus GPU architectures and potential struct packing and alignment. CPU cache lines are sensitive to data alignment with consideration of an "Array of Structures" versus a "Structure of Arrays" memory layout.

Different Algorithms: Consider using spatial partitioning data structures like <u>Octrees</u> for efficient force calculation. This can significantly reduce the number of particle interactions you need to compute. The Barnes-Hut algorithm leverages this approach.

Multi-threading: The current *update* function could easily be multi-threaded because each particle operation is independent. There should be one thread updating a particle's velocity and another updating its position. This way, while one thread is calculating the next position of a particle, the other is already updating the velocity of the next one. Force calculations between particles are independent and can be parallelized. Additionally, modern CPUs support SIMD (Single Instruction, Multiple Data) vectorization operations. Using libraries like Intel's MKL or OpenMP to distribute the computation across multiple cores. In the Update method, each particle's position is updated individually. This can be optimized using vectorized operations.

Optimize force calculation: The calculation of the force between particles is a bottleneck as it includes a square root operation in the distance calculation and division to determine the force. For the distance, you could calculate the squared distance and use it in the calculations. You could change the force formula to work with the squared distance, removing the cube root.

Use a simplified simulation (approximation): Accurately simulating the interaction of every particle with every other particle is computationally expensive. If 100% accuracy is not required for the use case, consider using approximation algorithms like the Barnes-Hut algorithm.

Reserve vector capacity: When you know how many particles you are going to simulate, you can use the reserve method to pre-allocate memory for the std::vector. This will avoid unnecessary reallocations as the vector grows.

Reducing function Calls: Within the *Update* function, avoid redundant calls to std::sqrt in the inner loop. Since the *force* calculation only requires the inverse square of the distance, you can eliminate the square root operation.

Improved Memory Access Patterns: Data structures may be reorganized to ensure better CPU cache utilization. Considerations include trade-offs between Structure of Arrays (SoA) versus Array of Structures (AoS) to improve cache locality. Additional considerations include accounting for CPU-group distribution across Non-Uniform Memory Access (NUMA) CPU sockets.

Avoid division by Zero: Within the original *Force* calculation, there's a potential for division by zero. This can be mitigated by adding a small epsilon value to the denominator.

Use a better random number generator: Use of rand() for particle initialization is not ideal. More recent pseudo-random number generators are available via the C++ <random> library.

Compile time optimizations: Utilize compiler optimizations when compiling code (e.g. *Whole Program Optimization, Favor Speed over Size*, etc.).

Profiling: Utilize the C++ Profiler to identify code segments that consume the most time and research optimizations accordingly.

Solution Scenarios Explored

The following solution scenarios have been explored corresponding to Part 1 of the exercise; that is, optimize the original serial solution using parallel algorithms. Note that the full source code with Visual Studio C++ 2022 project files is hosted via the github repository mentioned previously.

- 1. The original serial solution. See Appendix A.
- 2. Using the C++20 Standard Library. See Appendix B.
- 3. Using the OpenMP Library. See Appendix C.
- 4. Using C++20 and Structure of Arrays memory model. See Appendix D.

Runtime Summary

With *no optimization profiling*, the release builds of each project indicate the following runtimes for particle systems of sizes 100, 1000, and 10000. The time series execution step was maintained at 100 iterations. Note that, as of this writing, the C++ Struct of Arrays example has not correctly compiled.

Immediate observations conclude that the C++ Standard Library example illustrates the most optimal runtime. Interestingly, the OpenMP example has a runtime that is even worse thant the Serial solution. And, as noted above, the example utilizing a Structure of Arrays instead of the Array of Structures memory model remains to be compiled.

See the following table that summarizes each runtime scenario.

Project	NumParticles	NumSteps	eps Runtime (microsecs)	
Serial	100	100	2081	
	1000	100	208210	
	10000	100	20823451	
C++20 Standard Lib	100	100	4540	
	1000	100	117024	
	10000	100	1301815	
OpenMP	100	100	2643	
	1000	100	267168	
	10000	100	27066009	
C++20 StructOfArrays	100	100		

1000	100	
10000	100	

Next Steps

- 1. Utilize the Visual Studio Profiler to assess thread distribution, time spent per code segment, and CPU cache miss issues.
- 2. Identify testing scenarios that validate the correctness of the solution or indicate implementation improvements.
- 3. Consider thread-safety considerations around serial loops that have not yet been parallelized. Investigate whether or not later versions of the standard libraries have introduced thread-safe implementations of data-structures and algorithms.
- 4. Redesign the solution to leverage the Barnes-Hut N-Body algorithm and validate the expected performance improvements.

Appendix A: Original Solution

```
0.0, 0.0, 0.0;
       particles.push_back(p);
     }
  }
  void update() {
    for (int i = 0; i < particles.size(); i++) {
       for (int j = i + 1; j < particles.size(); j++) {
         double dx = particles[i].x - particles[i].x;
         double dy = particles[i].y - particles[i].y;
         double dz = particles[j].z - particles[i].z;
         double dist = std::sqrt(dx * dx + dy * dy + dz * dz);
         double force = particles[i].mass * particles[j].mass / (dist * dist * dist);
         particles[i].vx += force * dx;
         particles[i].vy += force * dy;
         particles[i].vz += force * dz;
        particles[j].vx -= force * dx;
         particles[j].vy -= force * dy;
         particles[j].vz -= force * dz;
     }
    for (auto& p : particles) {
       p.x += p.vx;
       p.y += p.vy;
       p.z += p.vz;
  }
};
E:\GIT\UsefulEngines\NBodyDemos\x64\Release>demo00.exe 100
     Original nBody Demo without optimizations...
         NumParticles
                              NumberOfSteps
                                                     Runtime(microseconds)
         100
                          100
                                          2081
E:\GIT\UsefulEngines\NBodyDemos\x64\Release>demo00.exe 1000
     Original nBody Demo without optimizations...
         NumParticles
                              NumberOfSteps
                                                     Runtime(microseconds)
          1000
                                           208210
                           100
E:\GIT\UsefulEngines\NBodyDemos\x64\Release>demo00.exe 10000
     Original nBody Demo without optimizations...
         NumParticles
                              NumberOfSteps
                                                     Runtime(microseconds)
         10000
                           100
                                            20823451
```

Appendix B: Solution Using C++20 Standard Library Parallelization

```
class ParticleSystem
{
public:
   std::vector<Particle> particles;
```

```
ParticleSystem(int num particles)
     // Optimization: reserve space for all particles up front.
     particles.reserve(num particles);
  //
  // Thoughts on thread safety...
  //
  // The following code is not using a Parallel for loop because the std::vector::push_back() is not
thread-safe.
 // And, the std::random device is not thread-safe. We could rewrite this so that each thread has its own
 // std::random device, but that would be inefficient too. Research further...
  //
     // Use a better random number generation scheme than rand()
     std::random device rd;
     std::mt19937 gen(rd());
     std::uniform real distribution <> dis(0.0, 1000.0);
     // Initialize particles... Thread-safe, but not parallelized.
for (int i = 0; i < num particles; ++i) {
Particle p:
// random position
p.pos[0] = 1e18 * exp(-1.8) * (0.5 - dis(gen));
p.pos[1] = 1e18 * exp(-1.8) * (0.5 - dis(gen));
p.pos[2] = 1e18 * exp(-1.8) * (0.5 - dis(gen));
// random velocity
p.vel[0] = 1e18 * exp(-1.8) * (0.5 - dis(gen));
p.vel[1] = 1e18 * exp(-1.8) * (0.5 - dis(gen));
p.vel[2] = 1e18 * exp(-1.8) * (0.5 - dis(gen));
// random mass
p.mass = 1.98892e30 * dis(gen) * 10 + 1e20;
// initial force
p.acc[0] = 0.0;
p.acc[1] = 0.0;
p.acc[2] = 0.0;
particles.push_back(p); // push_back() thread safety issues?
  }
  void update()
     // Reset accelerations to zero
     for (auto& p : particles) {
       p.acc[0] = p.acc[1] = p.acc[2] = 0.0;
     }
     // Calculate forces in parallel using C++20 parallel algorithms
     std::for each(std::execution::par, particles.begin(), particles.end(),
       [&](Particle& pi) {
          for (auto& pj : particles) {
            if (&pi != &pj) {
               double dx = pi.pos[0] - pi.pos[0];
               double dy = pi.pos[1] - pi.pos[1];
               double dz = pj.pos[2] - pi.pos[2];
```

//

```
// Optimization: Avoid redundant calls to std::sqrt in this inner loop
              double dist_squared = dx * dx + dy * dy + dz * dz;
             double dist sixth = dist squared * dist squared * dist squared;
              double force = pi.mass * pj.mass / (dist sixth + 1e-12); // Small epsilon to avoid division
by zero
              pi.acc[0] += force * dx;
              pi.acc[1] += force * dy;
              pi.acc[2] += force * dz;
         }
       }
    );
    // Update velocity and position. Parallelize this loop too?
    for (auto& p : particles) {
       for (int i = 0; i < 3; ++i) {
         p.vel[i] += p.acc[i];
         p.pos[i] += p.vel[i];
    }
};
E:\GIT\UsefulEngines\NBodyDemos\x64\Release>NBodyCpp20.exe 100
    Modified nBody Demo with C++20 standard library optimizations...
         NumParticles
                             NumberOfSteps
                                                  Runtime(microseconds)
         100
                         100
                                        4540
E:\GIT\UsefulEngines\NBodyDemos\x64\Release>NBodyCpp20.exe 1000
    Modified nBody Demo with C++20 standard library optimizations...
         NumParticles
                             NumberOfSteps
                                                  Runtime(microseconds)
         1000
                                         117024
E:\GIT\UsefulEngines\NBodyDemos\x64\Release>NBodyCpp20.exe 10000
    Modified nBody Demo with C++20 standard library optimizations...
         NumParticles
                             NumberOfSteps
                                                  Runtime(microseconds)
         10000
                          100
                                          1301815
Appendix C: Solution Using OpenMP Library Parallelization
```

```
class ParticleSystem
{
  public:
  std::vector<Particle> particles;

ParticleSystem(int num_particles)
{
  //
  // Optimization: reserve space for all particles up front.
  //
  particles.reserve(num_particles);
//
```

```
// Thoughts on thread safety...
//
// The following code is not using a Parallel for loop because the std::vector::push_back() is not thread-
safe.
// And, the std::random device is not thread-safe. We could rewrite this so that each thread has its own
// std::random device, but that would be inefficient too. Research further...
// Use a better random number generation scheme than rand()
std::random device rd;
std::mt19937 gen(rd());
std::uniform_real_distribution<> dis(0.0, 1000.0);
// Initialize particles...
for (int i = 0; i < num\_particles; ++i) {
Particle p;
// random position
p.pos[0] = 1e18 * exp(-1.8) * (0.5 - dis(gen));
p.pos[1] = 1e18 * exp(-1.8) * (0.5 - dis(gen));
p.pos[2] = 1e18 * exp(-1.8) * (0.5 - dis(gen));
// random velocity
p.vel[0] = 1e18 * exp(-1.8) * (0.5 - dis(gen));
p.vel[1] = 1e18 * exp(-1.8) * (0.5 - dis(gen));
p.vel[2] = 1e18 * exp(-1.8) * (0.5 - dis(gen));
// random mass
p.mass = 1.98892e30 * dis(gen) * 10 + 1e20;
// initial force
p.acc[0] = 0.0;
p.acc[1] = 0.0;
p.acc[2] = 0.0;
particles.push_back(p);
void update()
int n = particles.size();
// Reset accelerations to zero
#pragma omp parallel for
for (int i = 0; i < n; ++i) {
particles[i].acc[0] = particles[i].acc[1] = particles[i].acc[2] = 0.0;
// Calculate forces in parallel using OpenMP
#pragma omp parallel for
for (int i = 0; i < n; ++i) {
for (int j = 0; j < n; ++j) {
if (i != j) {
double dx = particles[i].pos[0] - particles[i].pos[0];
double dy = particles[i].pos[1] - particles[i].pos[1];
double dz = particles[j].pos[2] - particles[i].pos[2];
double dist_squared = dx * dx + dy * dy + dz * dz;
double dist sixth = dist squared * dist squared * dist squared;
double force = particles[i].mass * particles[j].mass / (dist_sixth + 1e-12);
particles[i].acc[0] += force * dx;
particles[i].acc[1] += force * dy;
```

```
particles[i].acc[2] += force * dz;
// Update velocity and position
#pragma omp parallel for
for (int i = 0; i < n; ++i) {
for (int k = 0; k < 3; ++k) {
particles[i].vel[k] += particles[i].acc[k];
particles[i].pos[k] += particles[i].vel[k];
};
E:\GIT\UsefulEngines\NBodyDemos\x64\Release>nbodyopenmp.exe 100
    Modified nBody Demo with OpenMP library optimizations...
         NumParticles
                            NumberOfSteps
                                                  Runtime(microseconds)
         100
                        100
                                        2643
E:\GIT\UsefulEngines\NBodyDemos\x64\Release>nbodyopenmp.exe 1000
    Modified nBody Demo with OpenMP library optimizations...
         NumParticles
                            NumberOfSteps
                                                  Runtime(microseconds)
         1000
                         100
                                         267168
E:\GIT\UsefulEngines\NBodyDemos\x64\Release>nbodyopenmp.exe 10000
    Modified nBody Demo with OpenMP library optimizations...
         NumParticles
                            NumberOfSteps
                                                  Runtime(microseconds)
         10000
                          100
                                         27066009
```

Appendix D: Solution Using C++20 and a Structure of Arrays Memory Model

```
class ParticleSystem
public:
  std::vector<std::array<double, 3>> positions;
  std::vector<std::array<double, 3>> velocities;
  std::vector<std::array<double, 3>> accelerations;
  std::vector<double> masses;
  ParticleSystem(int num particles)
    // Reserve space for all particles up front
     positions.reserve(num_particles);
     velocities.reserve(num particles);
     accelerations.reserve(num_particles);
     masses.reserve(num_particles);
    // Use a better random number generation scheme
     std::random device rd;
     std::mt19937 gen(rd());
     std::uniform_real_distribution<> dis(0.0, 1000.0);
    // Initialize particles...
     for (int i = 0; i < num\_particles; ++i) {
```

```
std::array<double, 3 > pos = \{ \{1e18 * exp(-1.8) * (0.5 - dis(gen)), \} \}
                            1e18 * exp(-1.8) * (0.5 - dis(gen)),
                            1e18 * exp(-1.8) * (0.5 - dis(gen)) \} ;
       std::array < double, 3 > vel = \{ \{1e18 * exp(-1.8) * (0.5 - dis(gen)), \} \}
                            1e18 * \exp(-1.8) * (0.5 - \text{dis(gen)}),
                            1e18 * \exp(-1.8) * (0.5 - \operatorname{dis}(\operatorname{gen})) \} ;
       double mass = 1.98892e30 * dis(gen) * 10 + 1e20;
       positions.push_back(pos);
       velocities.push_back(vel);
       accelerations.push_back({ 0.0, 0.0, 0.0 });
       masses.push back(mass);
     }
  }
  void update()
     // Reset accelerations to zero
     for (auto& acc: accelerations) {
       acc.fill(0.0);
     }
    // NOTE: The std::execution::par algorithm below doesn't yet work... debugging...
    // Calculate forces in parallel using C++20 parallel algorithms
     std::for each(std::execution::par, positions.begin(), positions.end(),
       [&](std::array<double, 3>& pi_pos, size_t i) {
          for (size_t j = 0; j < positions.size(); ++j) {
             if (i != j) {
               std::array<double, 3>& pj_pos = positions[j];
               double dx = pj_pos[0] - pi_pos[0];
               double dy = pi_pos[1] - pi_pos[1];
               double dz = pj_pos[2] - pi_pos[2];
               double dist squared = dx * dx + dy * dy + dz * dz;
               double dist_sixth = dist_squared * dist_squared * dist_squared;
               double force = masses[i] * masses[j] / (dist sixth + 1e-12);
               accelerations[i][0] += force * dx;
               accelerations[i][1] += force * dy;
               accelerations[i][2] += force * dz;
          }
       }
     );
     // Update velocity and position
     for (size_t i = 0; i < positions.size(); ++i) {
       for (int i = 0; i < 3; ++i) {
          velocities[i][j] += accelerations[i][j];
          positions[i][j] += velocities[i][j];
       }
     }
  }
};
```