
Molecular Dynamics with C++ report

Eslam Salah Zaki Elshiekh

5051183

University of Freiburg

Faculty of Engineering

Department of Microsystems Engineering

September 30th, 2022

Contents

1	Introduction	1
2	Methods	3
2.1	Velocity-Verlet integrator	3
2.2	Lennard-Jones potential force	4
2.3	Thermostat	4
2.4	Neighbor List	5
2.5	Embedded-atom method potential force	5
2.6	Parallelization	6
3	Implementation	7
3.1	Velocity-Verlet integrator	7
3.1.1	Test strategy for Verlet integrator	7
3.2	Lennard-Jones potential force	8
3.2.1	Derivation of the analytical expression for the forces of the Lennard-Jones potential	8
3.3	Berendsen thermostat	8
3.3.1	Test strategy for Berendsen thermostat	8
3.4	Embedded atom method	10
3.5	units and specification of the time unit	10
3.5.1	time step for the gold potential	10
3.6	Neighbor List	10

3.7	Parallelization using MPI	10
4	Results	11
4.1	Total energy as a function of time for different time steps	11
4.2	Snapshots sequence of LJ simulation	11
4.3	simulation time as a function of the size (number of atoms) without neighbor list	15
4.4	simulation time as a function of the size (number of atoms) with neighbor list	16
4.5	total energy vs temperature	16
4.6	melting point versus cluster size	17
4.7	heat capacity versus cluster size	18
4.8	latent heat versus cluster size	19
4.9	Energy conservation with MPI parallelization	20
4.10	Nanowire defects	20
5	Conclusion	21
	Bibliography	23

List of Figures

1	Total energy vs Time steps using parameters	11
2	LJ simulation snapshot1	12
3	LJ simulation snapshot2	12
4	LJ simulation snapshot3	13
5	LJ simulation snapshot4	13
6	LJ simulation snapshot5	14
7	Simulation time vs Atoms number	15
8	Simulation time vs Atoms number	16
9	Total energy vs Temp	17
10	Melting point vs cluster size	18
11	Heat capacity vs cluster size	19
12	Latent heat vs cluster size	20

List of Tables

List of Algorithms

1 Introduction

Molecular dynamics simulations are becoming an essential part of modern technology, because they allow to study the behavior of complex systems in a controlled way. The simulation of molecular systems is a very complex task, because the number of atoms in a system can be very large. Therefore, the simulation of a system requires a lot of computational power. In the last years, the computational power of computers has increased dramatically. This development has led to the fact that molecular dynamics simulations are now possible for systems with millions of atoms. However, the simulation of such systems is still a very demanding task. Therefore, the simulation of such systems is usually done on supercomputers or clusters of computers.

In this report, we will discuss the design and Implementation of a molecular dynamics simulations for atoms and molecules. We will also discuss the different potential forces that are used in the simulations, and how they are implemented, initializing atomic systems in random positions and preserve the atoms from evaporating using Thermostates, how to make the simulation run faster by using only neighbor list search, and how to parallelize the simulation using MPI.

2 Methods

Here we will discuss the different methods that we will use in our simulation.

2.1 Velocity-Verlet integrator

The velocity-Verlet algorithm is a numerical method for solving the equations of motion of a system of particles. It is a symplectic integrator, which means that it conserves the total energy of the system. The algorithm is based on the Taylor expansion of the position and velocity of the particles. The algorithm is as follows:

1. Calculate the acceleration of the particles at time t .
2. Calculate the velocity of the particles at time $t + \frac{1}{2}\Delta t$.
3. Calculate the position of the particles at time $t + \Delta t$.
4. Calculate the acceleration of the particles at time $t + \Delta t$.
5. Calculate the velocity of the particles at time $t + \Delta t$.

2.2 Lennard-Jones potential force

The Lennard-Jones potential is a potential that is used to model the interaction between atoms and molecules. It is a potential that is used in molecular dynamics simulations. LJ potential is a function of the distance between two particles. The LJ potential is given by the following equation:

$$V(r) = 4\epsilon \left[\left(\frac{\sigma}{r} \right)^{12} - \left(\frac{\sigma}{r} \right)^6 \right] \quad (1)$$

where ϵ is the depth of the potential well, σ is the distance at which the potential is zero, and r is the distance between the two atoms. The force is given by the following equation:

$$F(r) = -\frac{dV(r)}{dr} = 24\epsilon \left[\frac{2}{r} \left(\frac{\sigma}{r} \right)^{12} - \left(\frac{\sigma}{r} \right)^6 \right] \quad (2)$$

The force is calculated by using the following steps:

1. Calculate the distance between all pairs of atoms.
2. Calculate the potential between all pairs of atoms.
3. Calculate the force between all pairs of atoms.

2.3 Thermostat

The thermostat is an algorithm that is used to preserve the atoms from evaporating. The idea is when we initialize the atoms in random positions, the atoms will try to get to the equilibrium position. This means that the atoms will move very fast and will collide with each other and the temperature of the system will increase very fast. This will lead to the atoms evaporation. The thermostat is used to prevent this from happening. The thermostat is used to slow down the atoms and make them move

slower until they reach the equilibrium position and the temperature of the system is stable. The thermostat is based on the idea that the kinetic energy of the system is conserved. This means that if we change the velocity of the particles, then we must change the velocity of the other particles in the system in such a way that the total kinetic energy of the system is conserved. The algorithm is as follows:

2.4 Neighbor List

The neighbor list is an algorithm that is used to speed up the calculation of the forces that act on the particles. It is based on the idea that the forces between particles are only dependent on the distance between them. This means that if the distance between two particles is larger than a certain threshold value, then the force between them is zero. Therefore, we can calculate the forces between particles only if the distance between them is smaller than the threshold value. The algorithm is as follows:

1. Calculate the distance between all particles.
2. If the distance between two particles is smaller than the threshold value, then calculate the force between them.
3. If the distance between two particles is larger than the threshold value, then do not calculate the force between them.

2.5 Embedded-atom method potential force

The embedded-atom method (EAM) is a method for calculating the forces between atoms. It is based on the idea that the forces between atoms are dependent on the

density of the atoms. EAM is a potential force that is used in the simulation of metals, alloys, and intermetallic compounds. The EAM potential force is as follows:

2.6 Parallelization

The parallelization of the simulation is done using MPI. The algorithm is as follows:

1. Initialize the MPI environment.
2. Get the number of processes.
3. Get the rank of the process.
4. Initialize the atoms.
5. Calculate the forces.
6. Calculate the total energy.
7. Calculate the total momentum.
8. Calculate the temperature.
9. Calculate the pressure.
10. Print the results.
11. Finalize the MPI environment.

3 Implementation

3.1 Velocity-Verlet integrator

3.1.1 Test strategy for Verlet integrator

We test the Verlet integrator by comparing the results of the Verlet integrator with the results of the analytical solution of the equations of motion of a particle. the analytical solution of the equations of motion of a particle is given by the following equations:

$$\begin{aligned}x_i(t + dt) &= x_i(t) + v_i(t)dt + \frac{1}{2m}f_i(t)dt^2 \\v_i(t + dt) &= v_i(t) + \frac{1}{2m_i}(f_i(t) + f_i(t + dt))dt\end{aligned}\tag{3}$$

If we assume that there is no acting forces (constant and equal to zero) on the particles, then the analytical solution of the equations of motion of a particle after N time steps is given by the following equations:

$$\begin{aligned}x_i(t + N * dt) &= x_i(t) + \sum_{i=0}^N v_i(t)dt \\v_i(t + N * dt) &= v_i(t)\end{aligned}\tag{4}$$

where dt is the time step, N is the number of time steps, $x_i(t)$ is the position of the particle i at time t , $v_i(t)$ is the velocity of the particle i at time t , $f_i(t)$ is the force acting on the particle i at time t , and m_i is the mass of the particle i .

That means if we compare the result after all the integration steps of the two Verlet steps with the expected output of the analytical solution, then we can be sure that the Verlet integrator is working correctly.

3.2 Lennard-Jones potential force

3.2.1 Derivation of the analytical expression for the forces of the Lennard-Jones potential

3.3 Berendsen thermostat

talk a little bit about the Berendsen thermostat implementation

3.3.1 Test strategy for Berendsen thermostat

We have implemented two test cases for the Berendsen thermostat. The first test case is a test case where we test the Berendsen thermostat on a system with a single particle. The second test case is a test case where we test the Berendsen thermostat on a cubic lattice of size 5x5x5 with lattice constant=1.12. The test cases are implemented in the file.

Test case 1: Berendsen thermostat on a system with a single particle

Here, we make use of the derived equation in the lecture notes:

$$T(t) = T_0 + (T_1 - T_0) * \exp(-\frac{t}{\tau}) \quad (5)$$

where $T(t)$ is the temperature of the system at time t , T_1 is the initial temperature of the system, T_0 is the target temperature of the system, and τ is relaxation time constant. and using this equation we can calculate the temperature of the system at every iteration t before and after the thermostat, and then make sure that the temperature relaxes exponentially to the target temperature.

Test case 2: Berendsen thermostat on a cubic lattice of size 5x5x5 with lattice constant=1.12

This test is confirming if the temperature really converges to the target temperature after running the simulation for some time. So, the test is as follows:

1. Create a cubic lattice of size 5x5x5 with lattice constant=1.12.
2. using time_step=0.01 , target temperature=1.0, and relaxation time constant=10* time_step.
3. Run the simulation for 10000 time steps with the Berendsen thermostat.
4. In the last half of the simulation (5000 time steps), calculate the system temperature in every time step and make sure that the temperature converges to the target temperature.

3.4 Embedded atom method

3.5 units and specification of the time unit

3.5.1 time step for the gold potential

3.6 Neighbor List

3.7 Parallelization using MPI

4 Results

4.1 Total energy as a function of time for different time steps

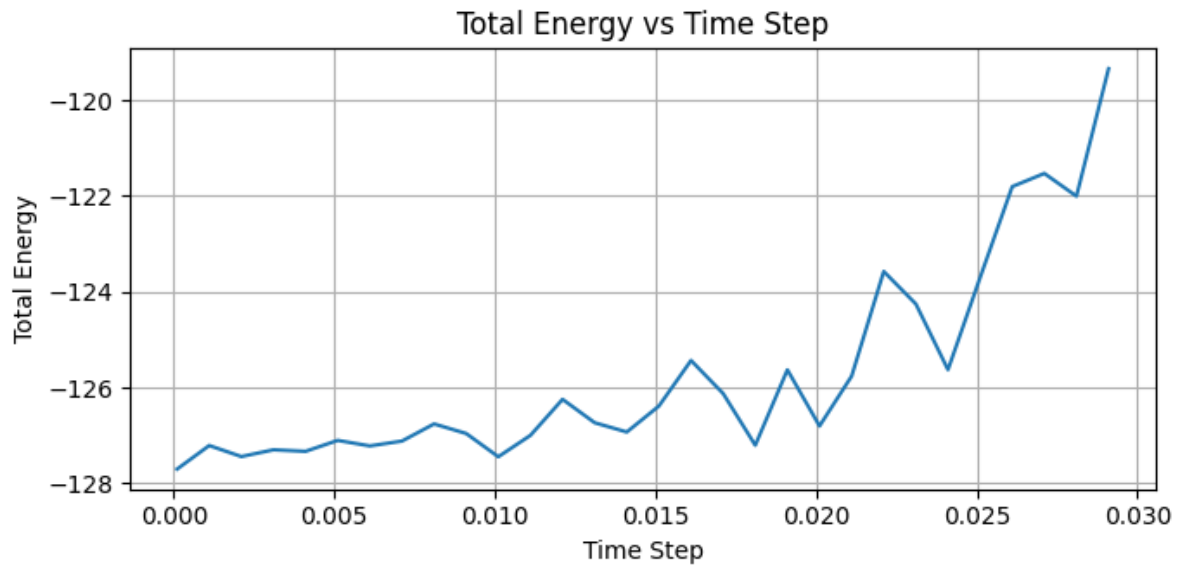


Figure 1: Total energy vs Time steps using parameters:
start_time_step=1e-4, end_time_step=30e-3, step=1e-3,
sigma=1, mass=1, epsilon=1, total_time=5000

4.2 Snapshots sequence of LJ simulation

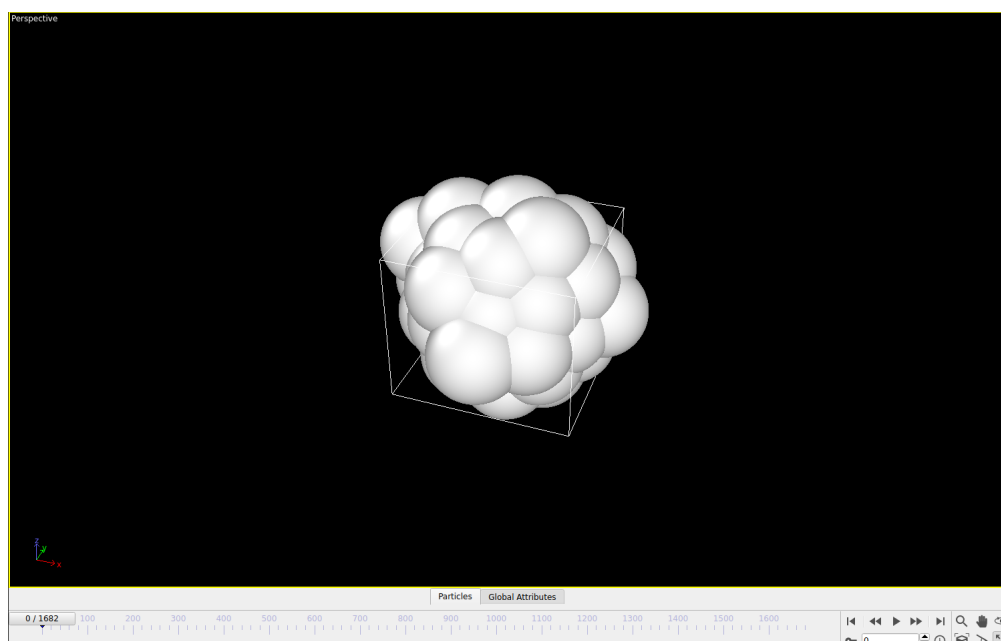


Figure 2: LJ simulation snapshot1

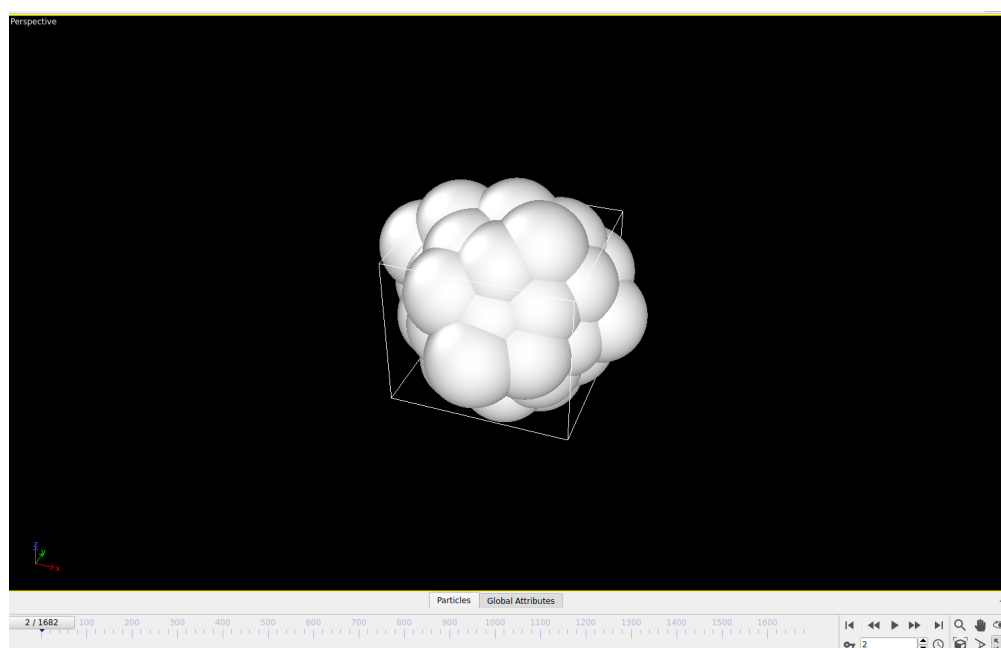


Figure 3: LJ simulation snapshot2

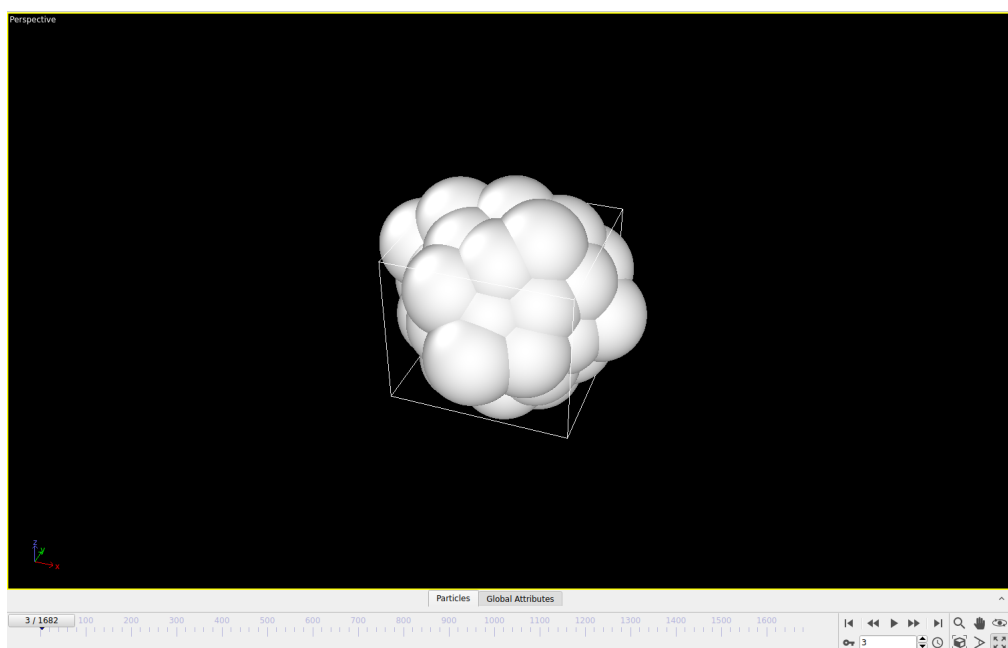


Figure 4: LJ simulation snapshot3

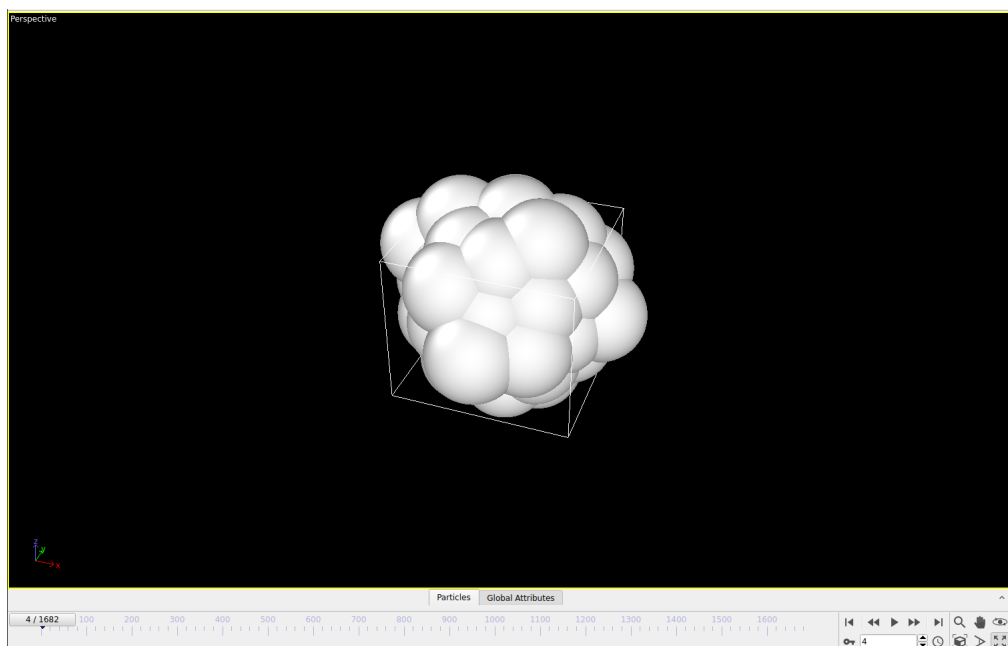


Figure 5: LJ simulation snapshot4

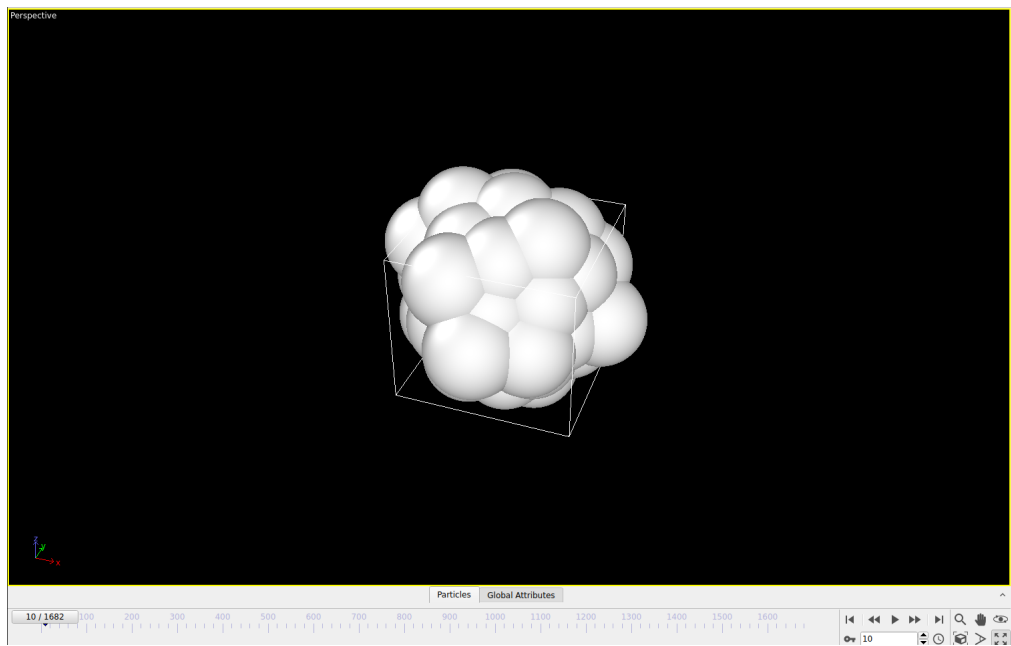


Figure 6: LJ simulation snapshot5

4.3 simulation time as a function of the size (number of atoms) without neighbor list

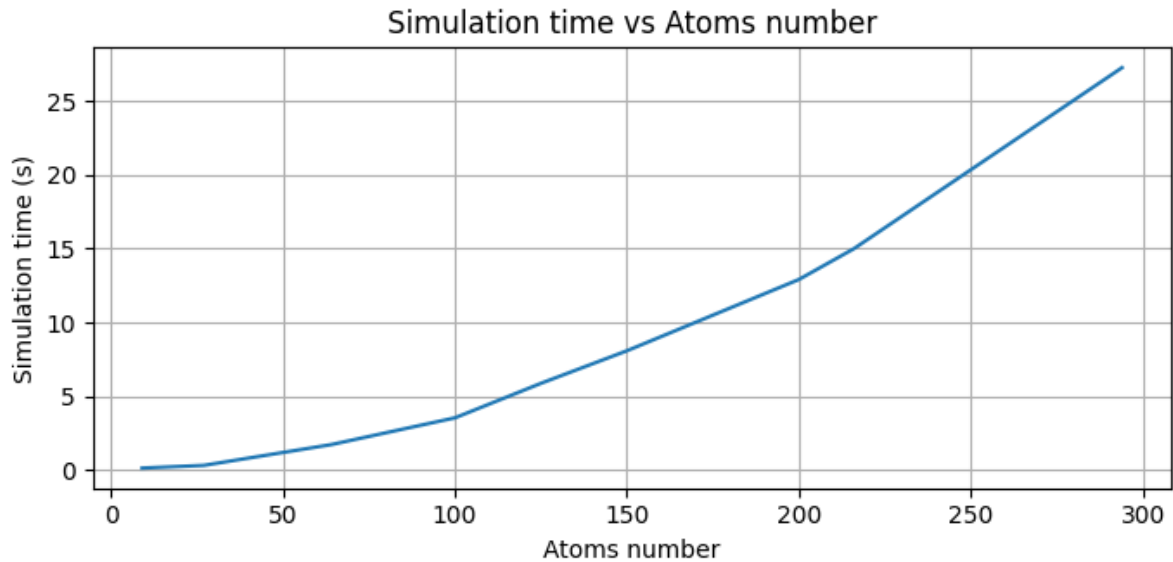


Figure 7: Simulation time vs Atoms number using parameters:
time_step=1e-3, sigma=1, mass=1, epsilon=1,
total_time=5000, relaxation_time_start = 10*
time_step, then after reaching equilibrium, it increases to
50*time_step.

Here we notice that the simulation time increases quadratically with the number of atoms in the system. This is because in the energy update step in the LJ simulation we have to calculate the energy between all the pairs of atoms in the system. So the time complexity of the energy update step is $O(N^2)$ where N is the number of atoms in the system. For example, In the figure above, The simulation time for 100 atoms is 3.5s and for 200 atoms is 12.9s which is almost 4 times more than the simulation time for 100 atoms.

4.4 simulation time as a function of the size (number of atoms) with neighbor list

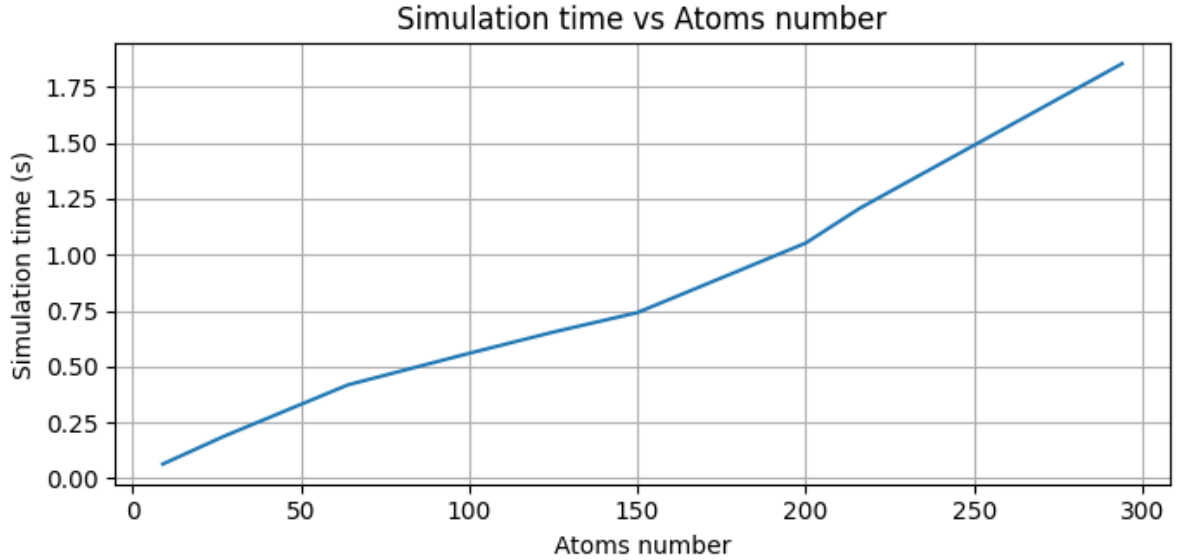


Figure 8: Simulation time vs Atoms number using parameters:
time_step=1e-3, sigma=1, mass=1, epsilon=1,
total_time=5000, relaxation_time_start = 10*
time_step, then after reaching equilibrium, it increases to
50*time_step, and cutoff_radius = 1.5.

Here, it's very clear the linearity of the simulation time with the number of atoms in the system. By introducing the concept of just calculating the energy between the atoms that are close to each other, we have reduced the time complexity of the energy update step from $O(N^2)$ to $O(N)$ where N is the number of atoms in the system. For example, In the figure above, The simulation time for 100 atoms is 0.56s and for 200 atoms is 1.05s which is almost 2 times more than the simulation time for 100 atoms.

4.5 total energy vs temperature

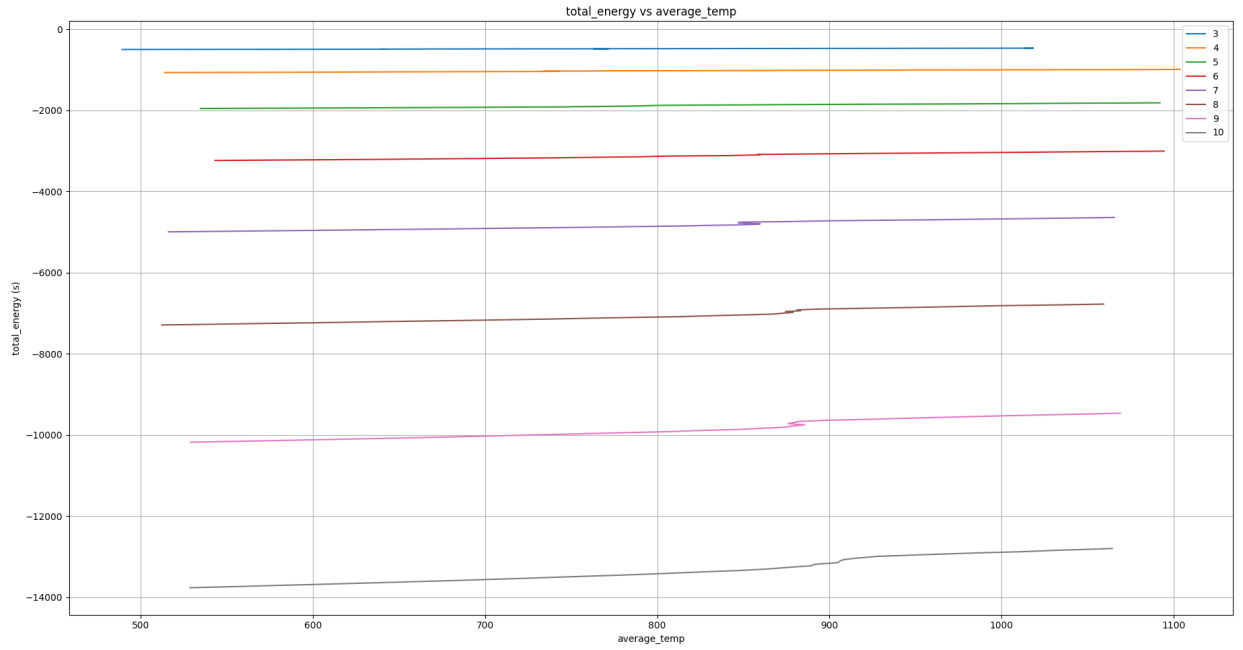


Figure 9: Total energy vs Temp using parameters: time_step=10fs, mass=20405.7294, initial_simulation_time=10000, relaxation_time_start = 10* time_step, then after 500 step, it increases to 1e10*time_step, and cutoff_radius = 10, target_temp = 500, number_experiments=26, experiment_simulation_time=500, added_energy_in_each_experiment=0.01*no_atoms, cluster_sizes=3:10, and atomic_distance = 2.885

4.6 melting point versus cluster size

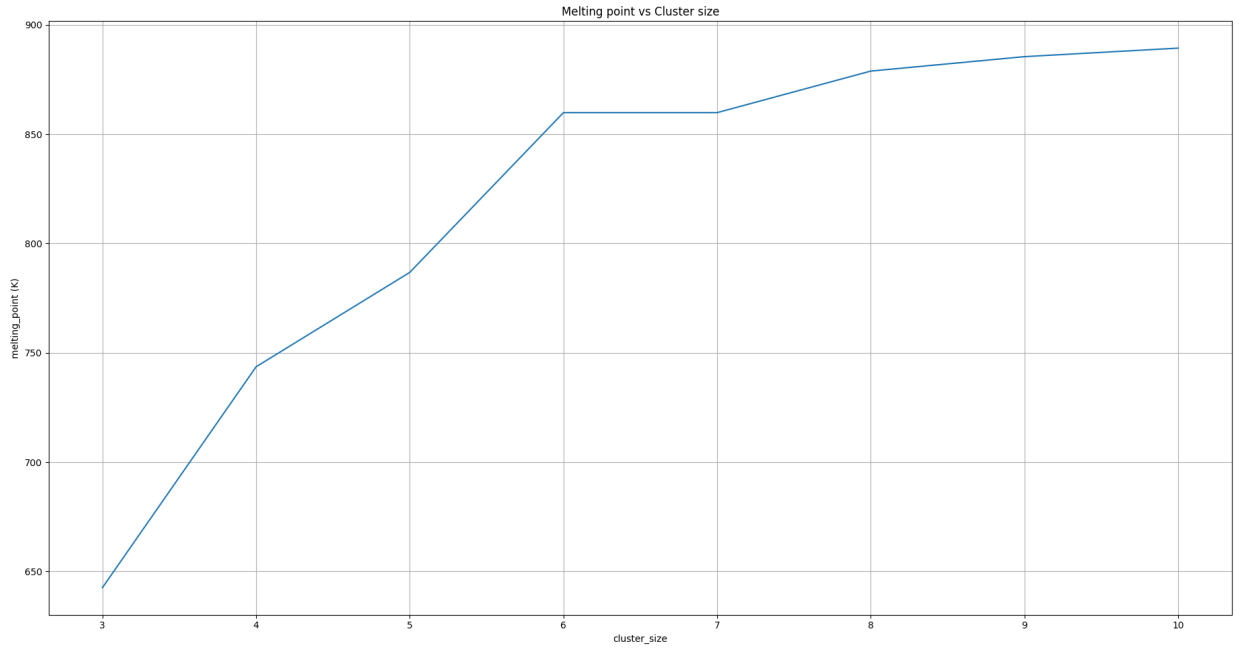


Figure 10: Melting point vs cluster size using parameters:
time_step=10fs, mass=20405.7294,
initial_simulation_time=10000, relaxation_time_start =
10* time_step, then after 500 step, it increases to
1e10*time_step, and cutoff_radius = 10, target_temp = 500,
number_experiments=26,
experiment_simulation_time=500,
added_energy_in_each_experiment=0.01*no_atoms,
cluster_sizes=3:10, and atomic_distance = 2.885

4.7 heat capacity versus cluster size

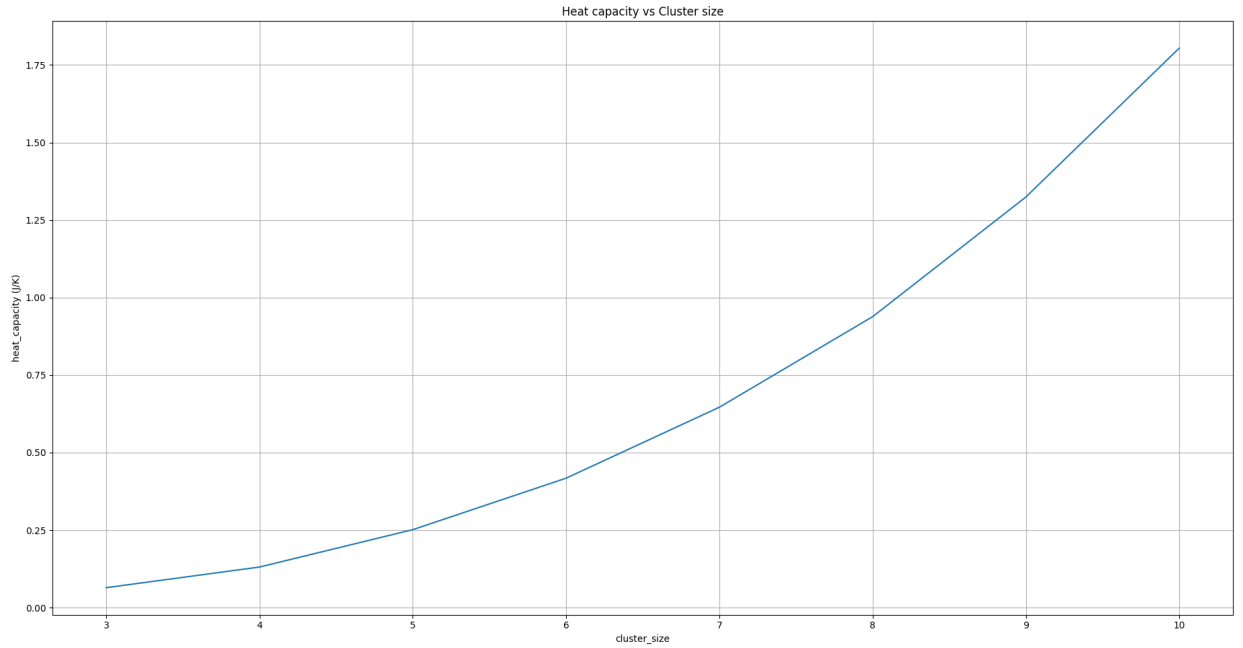


Figure 11: Heat capacity vs cluster size using parameters:
time_step=10fs, mass=20405.7294,
initial_simulation_time=10000, relaxation_time_start =
10* time_step, then after 500 step, it increases to
1e10*time_step, and cutoff_radius = 10, target_temp = 500,
number_experiments=26,
experiment_simulation_time=500,
added_energy_in_each_experiment=0.01*no_atoms,
cluster_sizes=3:10, and atomic_distance = 2.885

4.8 latent heat versus cluster size

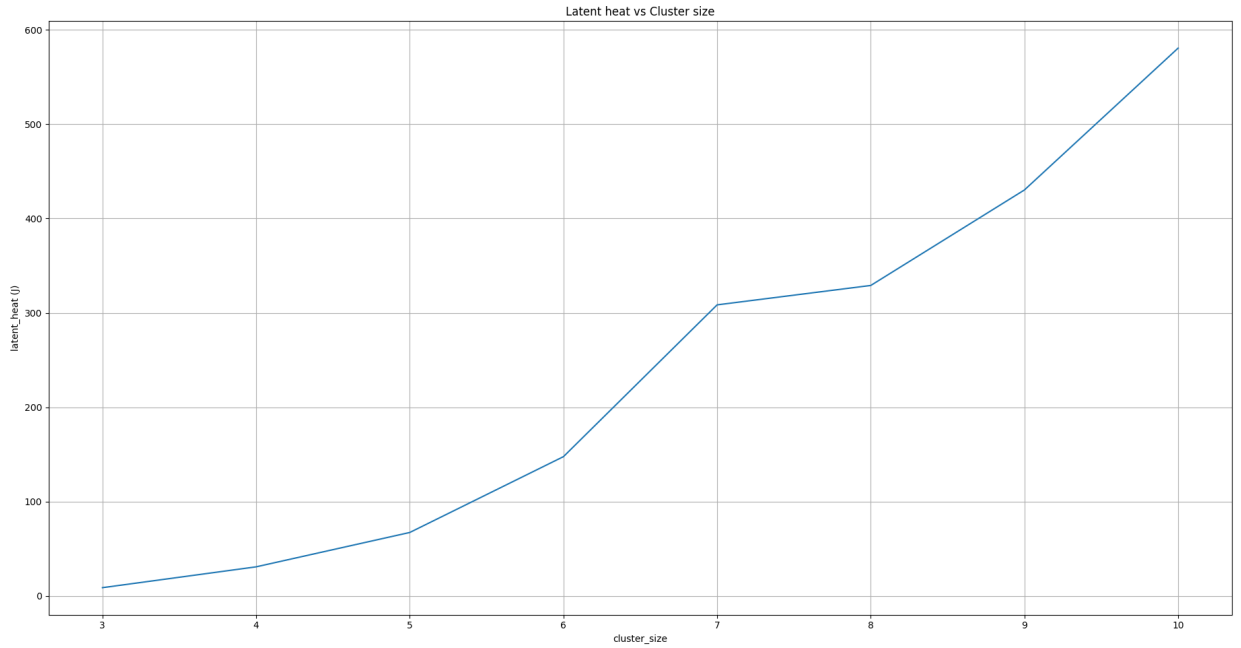


Figure 12: Latent heat vs cluster size using parameters: $\text{time_step}=10\text{fs}$, $\text{mass}=20405.7294$, $\text{initial_simulation_time}=10000$, $\text{relaxation_time_start} = 10 * \text{time_step}$, then after 500 step, it increases to $1e10 * \text{time_step}$, and $\text{cutoff_radius} = 10$, $\text{target_temp} = 500$, $\text{number_experiments}=26$, $\text{experiment_simulation_time}=500$, $\text{added_energy_in_each_experiment}=0.01 * \text{no_atoms}$, $\text{cluster_sizes}=3:10$, and $\text{atomic_distance} = 2.885$

4.9 Energy conservation with MPI parallelization

4.10 Nanowire defects

5 Conclusion

Bibliography

