

Molecular Dynamics Simulation of Argon

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Abstract Write the abstract here - what the project is and what we are going to show.

1 Introduction

There is an increasing demand for knowledge on the dynamic properties of physical systems. Where other methods such as MonteCarlo methods are useful to find static properties of such systems, Molecular Dynamics (MD) can be used to research the dynamics of such systems. MD is based on the equations of motion. For this reason MD allows us to track down the position and momentum of each particle at each moment in time. Hence besides static properties, MD can be used to research how particles effect eachother in multiple particle systems, which gives insight in the dynamics of such systems [2]. MD is applied in many fields, for example in biological physics in the study of macromolecules [1]

This report is broken into the following sections. In Section 2, we will briefly discuss the theory behind the calculations in this report. Section 3 presents the results of our calculations, including a thermostat, correlation functions for the three states of matter, and pressure calculations. We will then give a brief conclusion in section 4. An appendix is found at the end of this report to discuss the method used for error analysis.

2 Theory

2.1 Crystalline Array

The lattice structure of argon is face-centered cubic (fcc). The unit cell in an fcc crystal contains four atoms in the shape of a triangular pyramid, Figure ???. In the fcc lattice, the sheets of atoms stack on top of each other in such a way that every other layer is off-set by half of an atomic radius, Figure ??, left. (This is in contrast to a hpc lattice where every other sheet of atoms lines up, Figure ??, right.) We use the fcc configuration for our simulation because the equilibrium state of argon is an fcc lattice.

2.2 Interaction

To describe the interaction between each of the argon atoms, we use the Lennard-Jones potential, which accounts for the short-range repulsive force between the two atoms due to Coulomb repulsion and the longer-range attractive forces from dipole-dipole and dipole-induced dipole forces. The form of the potential is as follows:

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

where r is the distance between the two atoms, ϵ gives the depth of the potential well, and $\sigma = 2R$ where R is the internuclear distance between the two atoms - σ measures the distance at which the potential between the two particles is zero. An example of the Lennard-Jones potential can be found in Figure 1.

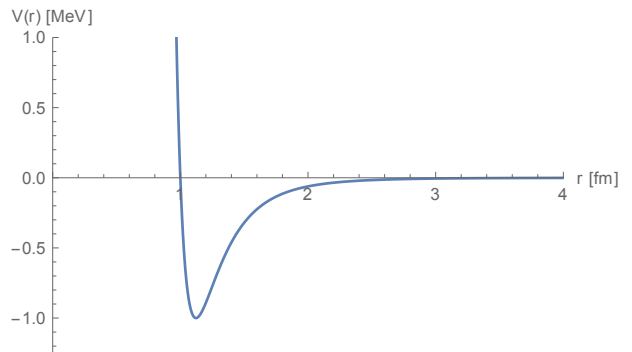


Figure 1: The shape of the Lennard-Jones potential. Here, $\sigma = 1$ (where the potential crosses the r-axis) and $\epsilon = 1$ (the depth of the interaction).

When the two atoms are far enough apart, they can be viewed as non-interacting. For this reason, we can introduced a cut-off radius, r_c , beyond which the potential is taken to be zero. Thus, we use the interaction

$$V_{LJ}(r) = \begin{cases} 4\epsilon \left[\left(\frac{\sigma}{r}\right)^{12} - \left(\frac{\sigma}{r}\right)^6 \right] & r \leq r_c \\ 0 & r > r_c \end{cases} \quad (2)$$

From the expression for the Lennard-Jones potential, (1), we immediately know the form of the force between the two atoms as well. The force is radial, acting to bring the atoms closer together if $r > r_{\min}$ and repelling them if $r < r_{\min}$. For our purposes, it is convenient to write the forces in terms of their Cartesian coordinates.

$$\begin{aligned} F_x &= 24\epsilon x \\ F_y &= 24\epsilon y \\ F_z &= 24\epsilon z \end{aligned} \quad (3)$$

From the force between each pair of particles, the momenta and positions can be calculated, leading to the dynamics of the system. This also allows calculation of thermodynamical quantities, such as pressure and correlation functions.

2.3 Velocity Verlet Method

The method chosen for the position and momentum calculations is the Verlet method. While both can be calculated from introductory physics equations,

$$p_{i+1} = p_i + F\Delta t \quad (4)$$

and

$$x_{i+1} = x_i + \frac{p_i}{m} + \frac{1}{2} \frac{F(\Delta t)^2}{m} \quad (5)$$

this can lead to numerical inaccuracies in the position calculation because of the dependence on the square of the timestep and both the previous force and momentum. Therefore, in our algorithm we make use of the Verlet method due to its accuracy, stability and its tendency to traject the same path in phase space within numerical precision. In our simulation we will need the momenta and positions at every time step to calculate parameters such as energy. A variety of the Verlet method which accomplishes this is known as the Verlet velocity scheme ??.

$$\begin{aligned} p_{i+\frac{1}{2}} &= p_i + \frac{F_i}{2} \Delta t \\ r_{i+1} &= r_i + \frac{p_{i+\frac{1}{2}}}{m} \Delta t \\ F_{i+1} &= -\frac{1}{m} \frac{\partial V_{LJ}(r_i)}{\partial r} \\ p_{i+1} &= p_{i+\frac{1}{2}} + \frac{F_{i+1}}{2} \Delta t \end{aligned} \quad (6)$$

Along with taking $\epsilon = 1$ and $\sigma = 1$ in the Lennard-Jones potential, we will also take $m = 1$, with the motivation that these can easily be adjusted to comply with the users' wishes, but only scales the outcome.

2.4 Thermodynamics

Along with updating the position and momentum of each particle to see how the system equilibrates, there are also several thermodynamic quantities that we can calculate in order to learn more about the system we have created. Before calculating any quantities, we want to model our system as being in contact with a thermostat to keep it at a constant temperature. We do this by renormalizing the velocity.

Through the Equipartition Theorem, the average velocity and temperature of a system are related by

$$\frac{1}{2} m \langle v^2 \rangle = \frac{3}{2} k_B T \quad (7)$$

where m is the mass of the system, $\langle v^2 \rangle$ is the time average of the square of the velocity, k_B is Boltzmann's constant (equal to one in a system where we take temperature and energy to have the same unit), and T is the temperature of the system. Therefore, to enforce a target temperature, T_o in our system, the system must have a certain average velocity, $\langle v^2 \rangle_o$ given by

$$\frac{1}{2} m \langle v^2 \rangle_o = \frac{3}{2} k_B T_o \quad (8)$$

Dividing (7) by (8) and taking the square root of each side, we have

$$\sqrt{\frac{\langle v^2 \rangle}{\langle v^2 \rangle_o}} = \left(\frac{T}{T_o} \right)^{1/2} \quad (9)$$

Thus we can renormalize each velocity with

$$v = \lambda v_o \quad (10)$$

where $\lambda = (T/T_o)^{1/2}$ to ensure a constant temperature.

Once the system is at a constant temperature, we can calculate the correlation function. The correlation function measures the amount of disorder in a system, and therefore gives a distinct shape based on the state of matter - solid, liquid or gas. To calculate the correlation function, $g(r)$, we count the number of particles inside a spherical shell of $r + \delta r$ and normalize by the volume of that shell. Taking δr to be

much smaller than r , we can approximate the volume of each shell by

$$\begin{aligned}\Delta V &= \frac{4}{3}\pi(r + \delta r)^3 - \frac{4}{3}\pi r^3 \\ &= \frac{4}{3}\pi \left[r^3 \left(1 + \frac{\delta r}{r} \right)^3 - r^3 \right] \\ &= \frac{4}{3}\pi \left[r^3 \left(1 + 3\frac{\delta r}{r} + \mathcal{O}(\delta r^2) \right) - r^3 \right] \\ &= 4\pi r^2 \delta r\end{aligned}\tag{11}$$

where the third line uses the Taylor approximation of $(1 + z)^n = 1 + nz + \dots$. Thus the correlation function is

$$g(r) = \frac{N_r}{4\pi r^2 \delta r}\tag{12}$$

where N_r is the number of particles found in the shell $r + \delta r$.

The correlation function has distinct shapes for each state of matter. Solids are classified by period spikes which indicates a lattice of particles. Liquids look like a decaying periodic function. Lastly, gases show an exponential-like decay in their correlation function. These are the characteristics we will be looking to explore in our calculations.

Another physical quantity that can be calculated is the pressure. With a constant temperature, the pressure of a system can be found using the Virial Theorem. Using Hamilton's equations

$$\begin{aligned}\dot{\mathbf{r}}_a &= \mathbf{v}_a \\ \dot{\mathbf{p}}_a &= \mathbf{F}_a\end{aligned}\tag{13}$$

and taking the time average of $\sum_a (\mathbf{r}_a \cdot \mathbf{p}_a)$, we have

$$\begin{aligned}\overline{\frac{d}{dt} \sum_a \mathbf{r}_a \cdot \mathbf{p}_a} &= \overline{\sum_a \mathbf{v}_a \cdot \mathbf{p}_a + \mathbf{r}_a \cdot \mathbf{F}_a} \\ &= 0\end{aligned}\tag{14}$$

For non-relativistic particles

$$K = \sum_a \frac{\mathbf{p}_a^2}{2m_a}\tag{15}$$

and

$$\begin{aligned}\overline{\sum_a \mathbf{v}_a \cdot \mathbf{p}_a} &= 2\bar{K} \\ &= -\overline{\sum_a \mathbf{r}_a \cdot \mathbf{F}_a}\end{aligned}\tag{16}$$

This gives the Virial term between two particles. We can add this term to our expected energy-pressure relation for non-interacting particles, $\bar{K} = \frac{3}{2}PV$, to find the total average kinetic energy of a system of interacting particles:

$$\bar{K} = \frac{3}{2}PV - \frac{1}{2} \overline{\sum_{a \neq b} r_{ab} F(r_{ab})}\tag{17}$$

from here, straight-forward algebraic manipulations give the pressure.

$$P = \frac{2}{3V} \left[\bar{K} + \frac{1}{2} \overline{\sum_{a \neq b} r_{ab} F(r_{ab})} \right]\tag{18}$$

Here, $r_{ab} = |\mathbf{r}_a - \mathbf{r}_b|$, the distance between any pair of particles.

3 Discussion

In the following subsections, we present the results of our simulation.

3.1 Initial Conditions

We started with our system in a **crystal name** crystalline configuration (recall Figure **fig**). Our simulation modeled 108 (**probably more at the end of the project**) argon atoms with a unit cell length of **2 fm**. As mentioned above, we will take $\sigma = \epsilon = k_B = 1$, which also gives us a system where temperature and energy are measured in the same units.

3.2 Thermostat

Give a target temperature of each thermostat (whatever's used below) and an error on it. Also include a plot or two of the temperature over time to see how quickly the system reaches the target temperature - based on how often the velocity is renormalized.

3.3 Correlation Function

One for each of the three states of matter - can either keep the same density and run the calculation for different temperatures or take one temperature and see if we can find different densities at which each state of matter exists. Both are probably interesting.

3.4 Pressure

Find the pressure of different systems under different conditions - temperature and density. Make sure to include errors on each number.

System paramters	p	$\sigma_{pressure}$	$\sigma_{pressure_{block}}$
$T = 3, \rho = 0.5$	5.4862	0.0836	0.0089
$T = 1, \rho = 0.88$	2.6982	0.0171	0.0067
$T = 0.5, \rho = 1.2$	1.1105	0.0056	0.0021

4 Conclusion

A Error Calculations

Like experimental measurements, numerical calculations should also include error bars. There are many ways that these uncertainties can be calculated, but in this paper, we will be using **name of method - or some description**. Just as in an experiment where we would ideally take many measurements and calculate an error based on their standard deviations from the mean, we can use the same concept to put error bars on our calculated values. There are several ways that we can do this. Perhaps, the most obvious one is to run the calculation several times to generate a mean and a standard deviation. However, this process is lengthy and can be heavily dependent on the seed of the random number generator used in initializing positions and velocities the particles.

It is, therefore, much better to run one simulation and average over many blocks of time. In each simulation, after a long enough period of time, the

quantity of interests will settle on some mean value with every time step oscillating around this value, as show in Figure **plot of total energy or something like that which shows the oscillations from the mean and the cuts defining each time step block**. While each value depends on the previous value, averaging over a large enough number of time steps will produce a value independent of the previous block of time steps. For each of these blocks, say 1500 (**put in actual number that we used**) time steps, an average is calculated (this is equivalent to having many measurements from an experiment), and then from each of these "measurements", we can calculate an average value for our simulation and an error for that value. The average value is defined as

$$\bar{C} = \frac{1}{N} \sum_{i=1}^N C_i \quad (19)$$

and the error is calculated from

$$E = \frac{1}{\sqrt{N}} \sum_{i=1}^N (C_i - \bar{C})^2 \quad (20)$$

where E is the error on a given quantity, N is the number of time-step blocks, and C_i is the quantity value for each time step block. **I'm not sure if this equation is actually correct.**

This ensures that we include fluctuations in the numbers that we report, as well as account for a decreasing error with an increasing number of "measurements".

References

- [1] B. de Groot. Computational biomolecular dynamics group.
- [2] J.M. Thijssen. *Computational Physics*, volume 4 of 10. Cambridge University Press, Cambridge, 3 edition, 7 2007. An optional note.