

GPUMD:
Graphics Processing Units
Molecular Dynamics

Reference Manual

Version 2.1

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Chapter 1

Introduction

1.1 How to read this manual?

- First read through chapters 1 and 2.
- Then glance over chapter 3. You can choose to either read through it or skip it.
- Then read the first section of chapter 4 and the sections you are interested.
- Then carefully read through chapter 5.
- Then study one or more examples in chapter 6.
- Go back to chapter 3 when necessary.

1.2 What is GPUMD?

GPUMD stands for **G**raphics **P**rocessing **U**nits **M**olecular **D**ynamics. It is a new molecular dynamics (MD) code fully implemented on graphics processing units (GPUs). It was firstly used for heat transport simulations only but we are now making it more and more general.

1.3 Citations

If you use GPUMD in your published work, we kindly ask you to cite the following paper which describes the central algorithms used in GPUMD:

- Zheyong Fan, Wei Chen, Ville Vierimaa, and Ari Harju,
Efficient molecular dynamics simulations with many-body potentials on graphics processing units,
Computer Physics Communications **218**, 10 (2017).
<https://doi.org/10.1016/j.cpc.2017.05.003>

If your work involves using heat current and virial stress formulas as implemented in GPUMD, the following paper can be cited:

- Zheyong Fan, Luiz Felipe C. Pereira, Hui-Qiong Wang, Jin-Cheng Zheng, Davide Donadio, and Ari Harju,
Force and heat current formulas for many-body potentials in molecular dynamics simulations with applications to thermal conductivity calculations,
Phys. Rev. B **92**, 094301 (2015).
<https://doi.org/10.1103/PhysRevB.92.094301>

You can cite the following paper if you use GPUMD to study heat transport using the in-out decomposition for 2D materials and/or the spectral decomposition method as described in it:

- Zheyong Fan, Luiz Felipe C. Pereira, Petri Hirvonen, Mikko M. Ervasti, Ken R. Elder, Davide Donadio, Tapio Ala-Nissila, and Ari Harju,
Thermal conductivity decomposition in two-dimensional materials: Application to graphene,
Phys. Rev. B **95**, 144309 (2017).
<https://doi.org/10.1103/PhysRevB.95.144309>

You can cite the following paper if you use GPUMD to study heat transport using the homogeneous nonequilibrium molecular dynamics (HNEMD) method:

- Zheyong Fan, Haikuan Dong, Ari Harju, and Tapio Ala-Nissila,
Homogeneous nonequilibrium molecular dynamics method for heat transport with many-body potentials,
<https://arxiv.org/abs/1805.00277>

1.4 Feedbacks

You can e-mail the first author if you find errors in the manual or bugs in the source code, or have any suggestions/questions about the manual and code. The following email addresses can be used:

- zheyong.fan(at)aalto.fi (valid at least up to the end of 2019)
- brucenju(at)gmail.com
- zheyongfan(at)163.com

1.5 Acknowledgments

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Chapter 2

Features of GPUMD

I suggest you read through this chapter on first reading.

2.1 GPU-accelerated force evaluation for many-body potentials

One of the major features of GPUMD is that force evaluation for many-body potentials has been significantly accelerated by using GPUs. Our efficient and flexible GPU-implementation of the force evaluation for many-body potentials relies on a set of simple expressions for force, virial stress, and heat current derived in Ref. [14]. Detailed algorithms for the efficient CUDA-implementation have been presented in Ref. [11].

Using the methods as described in Refs. [14, 11], we have implemented various many-body potentials in GPUMD, including:

- The EAM-type potential with some analytical forms [47, 6] (only for single-atom-type systems now).
- The Tersoff (1989) potential with single or double atom types [39].
- The Stillinger-Weber (1985) potential [37] up to three atom types.
- The Vashishta potential [42] for two-atom-type systems.
- The REBO potential for Mo-S systems [25, 26].

Apart from many-body potentials, there are also a few two-body potentials:

- The Lennard-Jones potential up to five atom types.
- The rigid-ion potential for two-atom-type systems, which consists of a Buckingham potential and a Coulomb potential. The Coulomb potential is evaluated using the method in Ref. [16].

One can also define multiple potentials for a complicated system.

More two-body and many-body potentials will be implemented in GPUMD in future versions.

2.2 Utilities for heat transport simulations

Apart from being highly efficient, another unique feature of GPUMD is that it has useful utilities to study heat transport. The current version of GPUMD can calculate the following quantities related to heat transport:

- It can calculate the phonon density of states (DOS) from the velocity autocorrelation function (VAC), using the method of Dickey and Paskin [8].
- It can calculate the equilibrium heat current autocorrelation (HAC), whose time integral gives the running thermal conductivity according to the Green-Kubo relation [17, 23]. As stressed in Ref. [14], the heat current as implemented in LAMMPS [32] does not apply to many-body potentials and significantly underestimates the thermal conductivity in 2D materials described by many-body potentials. GPUMD also contains the thermal conductivity decomposition method as introduced in Ref. [13], which is useful for 2D materials.
- It can calculate the thermal conductivity (or more exactly, the thermal conductance) of a system of finite length or the thermal boundary resistance (Kapitza resistance) of an interface or similar structures using non-equilibrium MD (NEMD) methods. The spectral decomposition method as described in Ref. [13] has also been implemented.
- It can calculate the thermal conductivity using the homogeneous nonequilibrium MD (HNEMD) method [10] for general many-body potentials, and the related spectral decomposition method [12].

2.3 Other features

2.3.1 Boundary conditions

GPUMD supports the following boundary conditions in each direction:

- free boundary conditions
- periodic boundary conditions (using the minimum image convention)
- fixed boundary conditions (by fixing some atoms)

The current version of GPUMD only supports orthogonal simulation boxes. Triclinic simulation boxes *might* be implemented in a future version.

2.3.2 Neighbor list construction

GPUMD has the following two versions for neighbor list construction and automatically chooses an appropriate one according to the inputs:

- an $O(N^2)$ method which builds the Verlet neighbor list by directly checking the distance between one particle and all the other particles in the simulation box
- an $O(N)$ method which first builds a cell list and then converts the cell list to the Verlet neighbor list

When the neighbor list is required to be updated during a run, one only has to specify a skin distance and the code will automatically determine when the neighbor list needs to be updated.

Warning: Neighbor list will not be updated for a run if the keyword “neighbor” is not used for that run. So use this keyword for each run unless you know for sure that there is no diffusion during the simulation. For more about keywords, see chapter 5.

2.3.3 Integration methods

The velocity-Verlet [38] integration scheme is used for all the ensembles. The supported ensembles and the adopted methods are

- the *NVE* ensemble
- the *NVT* ensemble
 - the Berendsen method [2]
 - the Nosé-Hoover chain method [31, 19, 29, 28, 41] with a fixed chain length of 4.
 - the Langevin thermostat using the algorithm by Bussi and Parrinello [4]
 - the Bussi-Donadio-Parrinello thermostat [3]
- the *NPT* ensemble
 - the Berendsen method [2] with the pressure in each direction controlled independently

We are working on implementing more integration methods.

2.4 Major changes compared to the previous versions

2.4.1 Major changes introduced in version 1.1

1. Added a potential model for single-layer black phosphorene as introduced by Xu *et al.* [46].

2.4.2 Major changes introduced in version 1.2

1. In the previous versions, the initial velocities have zero linear momentum but generally nonzero angular momentum. This will not result in rotation of the system if periodic boundary conditions are applied in two or three directions, but will result in rotation in systems with only one or no periodic direction. In version 1.2, both linear and angular momenta of the initial velocities are zeroed.

2.4.3 Major changes introduced in version 1.3

1. Added the REBO (reactive empirical bond-order) potential for Mo-S systems developed by Liang *et al.* [25, 26]. Note that the Lennard-Jones part was not included. We plan to include the Lennard-Jones part in a future version of GPUMD.

2.4.4 Major changes introduced in version 1.4

1. Added the Vashishta potential [42].

2.4.5 Major changes introduced in version 1.5

1. Added the general two-element Stillinger-Weber potential. The potential for single-layer black phosphorene introduced in version 1.1 is of this type and is thus removed.

2.4.6 Major changes introduced in version 1.6

1. Added the tabulated Vashishta potential, which is about two times as fast as the analytical version when the relative force error is about 10^{-6} .
2. The performance of the many-body potentials has been enhanced by about 50%.

2.4.7 Major changes introduced in version 1.7

1. Changed about half of the code from C style to C++ style.
2. Added the homogeneous nonequilibrium molecular dynamics (HNEMD) method for heat transport with many-body potentials [12, 44, 9]. Currently, we have only implemented this method for the Tersoff and the SW potentials. We will add this method to all the other potentials in the future.

2.4.8 Major changes introduced in version 1.8

1. Changed most of the code from C style to C++ style.
2. Added mixed potentials. That is, one can, if needed, use more than one potential for a system.
3. Implemented the HNEMD method [12, 44, 45, 9] for all the potentials.

2.4.9 Major changes introduced in version 1.9

1. Fixed some memory bugs.

2.4.10 Major changes introduced in version 2.0

1. Added the Langevin thermostat using the algorithm by Bussi and Parrinello [4].
2. Added the Bussi-Donadio-Parrinello thermostat [3].

2.4.11 Major changes introduced in version 2.1

1. Implemented (by the new contributor Alexander J. Gabourie) a general version of the spectral heat current (SHC) method. The previous version assumed that one atom from one group of the interface has at most one neighbor from the other group of the (imaginary) interface.

2.5 TODO List

Here I list a few features which I will hopefully implement the next couple of years:

- More Tersoff-type potentials
- Spline-based EAM potential
- The ADP potential
- Machine-learning potentials
- The NPT integrator based on the MTK equation
- Multi-GPU version of GPUMD
- Potential fitting
- Lattice dynamics (harmonic and anharmonic)

Chapter 3

Theoretical formalisms and numerical algorithms

This chapter could be skipped on first reading.

3.1 Physical units used in the program

The basic units in the numerical calculations are chosen to be

1. Energy: eV (electron volt)
2. Length: Å (angstrom)
3. Mass: amu (atomic mass unit)
4. Temperature: K (kelvin)
5. Charge: e (elementary charge)

The purpose of using these units is to make the values of most quantities in the code close to unity. The units for all the other quantities are thus fixed. Here are some examples:

1. Time: Å amu^{1/2} eV^{-1/2}, which is about 1.018051×10^1 fs
2. Velocity: eV^{1/2} amu^{-1/2}
3. Force: eV Å⁻¹
4. Pressure (stress): eV Å⁻³, which is about 1.602177×10^2 GPa
5. Thermal conductivity: eV^{3/2} amu^{-1/2} Å⁻² K⁻¹ which is about 1.573769×10^5 W m⁻¹ K⁻¹
6. Boltzmann's constant: $k_B \approx 8.617343 \times 10^{-5}$ eV K⁻¹
7. Electrostatic constant: $k_C = \frac{1}{4\pi\epsilon_0} \approx 1.441959 \times 10^1$ eV Å e⁻²

Important note: The input and output files do not necessarily adopt these units. For example, time step in the input file is in units of fs, rather than Å amu^{1/2} eV^{-1/2}. Details on the units adopted by the input and output files are presented in Chapter 5.

3.2 Overall structure of GPUMD

GPUMD is written using CUDA C/C++. Except for data initialization and some calculations that are very cheap or inherently serial, all the other calculations are done on the GPU.

The current version of GPUMD has about 15 000 lines of code. The structure of the code is as follows:

- The file `main.cu` contains the `main` function, which is the entrance of the code. The `main` function uses the GPUMD class. **There is one file here.**
- The GPUMD class is implemented in these files: `gpumd.cuh`, `gpumd.cu`, `run.cu`, `check_run.cu`, and `parse.cu`. **There are 5 files in total here.** This class controls the overall flow of the code. The GPUMD class uses all the other major classes: `Atom`, `Force`, `Integrate`, and `Measure`.
- The `Atom` class contains the major data related to the atoms (such as the various per-atom physical quantities, the simulation box, the neighbor list, etc) and functions used to initializing and finalizing these data. This class involves the files `atom.cuh`, `atom.cu`, `velocity.cu`, `neighbor.cu`, `neighbor_ON1.cu`, and `neighbor_ON2.cu`. **There are 6 files in total here.**
- The `Force` class (in files `force.cuh` and `force.cu`) deals with everything related to force evaluation. The `Force` class contains pointers of the abstract base class (ABC) `Potential` class (in files `potential.cuh` and `potential.cu`). This ABC is inherited by some classes, each corresponding to a specific empirical potential. The potential classes are `Pair` (in files `pair.cuh` and `pair.cu`), `EAM` (in files `eam.cuh` and `eam.cu`), `SW` (in files `sw.cuh` and `sw.cu`), `Tersoff` (in files `tersoff.cuh` and `tersoff.cu`), `Vashishta` (in files `vashishta.cuh` and `vashishta.cu`), and `REBO_MOS` (in files `rebo_mos2.cuh` and `rebo_mos2.cu`). **There are 16 files in total here.**
- The `Integrate` class (in files `integrate.cuh` and `integrate.cu`) deals with everything related to time integration. The `Integrate` class contains a pointer of the ABC `Ensemble` class (in files `ensemble.cuh` and `ensemble.cu`). This ABC is inherited by some classes, each corresponding to a specific integrator. The integrator classes are `Ensemble_NVE` (in files `ensemble_nve.cuh` and `ensemble_nve.cu`), `Ensemble_BER` (in files `ensemble_ber.cuh` and `ensemble_ber.cu`), `Ensemble_NHC` (in files `ensemble_nhc.cuh` and `ensemble_nhc.cu`), `Ensemble_LAN` (in files `ensemble_lan.cuh` and `ensemble_lan.cu`), and `Ensemble_BDP` (in files `ensemble_bdp.cuh` and `ensemble_bdp.cu`). **There are 14 files in total here.**
- The `Measure` class (in files `measure.cuh` and `measure.cu`) deals with everything related to measurement. The `Measure` class contains a few other classes, including the `VAC` class (in files `vac.cuh` and `vac.cu`), the `HAC` class (in files `hac.cuh` and `hac.cu`), `SHC` class (in files `shc.cuh` and `shc.cu`), the `HNEMD` class (in files `hnemd_kappa.cuh` and `hnemd_kappa.cu`), and the `Compute` class (in files `compute.cuh` and `compute.cu`). **There are 12 files in total here.**
- There is a header called `common.cuh` which is included in most of the other headers. This is a light-weighted header, which only contains some type definitions and forward declarations. **One file here.**

- There is a header called `ldg.cuh` which is included in some potential class files. This is used to choose between using the `__ldg()` function and the normal array indexing for some global memory reading. [One file here.](#)
- Some macros and functions for error handling are contained in the files `error.cuh` and `error.cu`. [Two files here.](#)
- There are two files named `validate.cuh` and `validate.cu` which are only used by the developers. [Two files here.](#)

How many files in total? **60!**

3.3 Coding style of GPUMD

Every programmer has his own coding style. Here is an incomplete list of conventions we tried to follow:

1. Use `snake_case` instead of `CamelCase` for naming.
2. Define variables as late as possible.
3. Use four spaces to indent.
4. Keep every line no longer than 80 characters.
5. Use the Allman style for brace placement. An example is:

```
for (int n = 0; n < 10; ++n)
{
    // do something
}
```

6. Use a similar style (perhaps I am the only one who uses this style?) for functions with many arguments:

```
my_function
(
    argument_1, argument_2, argument_3, argument_4,
    argument_5, argument_6, argument_7, argument_8
)
```

7. Remove dead code.
8. Keep GPUMD a standalone code. Only use standard C, C++ and CUDA libraries.

3.4 Neighbor list construction

We use the Verlet neighbor list when evaluating the forces between particles. Two methods for constructing the Verlet neighbor list are implemented, one is an $O(N^2)$ method and the other is an $O(N)$ method.

3.4.1 A simple quadratic-scaling method

In the $O(N^2)$ method, the Verlet neighbor list is constructed by directly checking the distance between every pair of particles. Therefore, the computational effort scales as N^2 , where N is the number of particles. This method only requires a single CUDA kernel. Algorithm 1 presents a pseudo code for the CUDA kernel.

Algorithm 1 The $O(N^2)$ method of neighbor list construction

Require: b is the block index
Require: t is the thread index
Require: S_b is the block size
Require: $i = S_b \times b + t$ is the particle index
Require: N is the number of particles
Require: r_c^2 is the square of the cutoff distance for building the neighbor list
Require: NN_i is the number neighbors for particle i
Require: NL_{ik} is the index of the k th neighbor of particle i
Require: \mathbf{r}_i is the position vector of particle i

```

1:  $k \leftarrow 0$ 
2: if  $i < N$  then
3:   load  $\mathbf{r}_i$  from the global memory
4:   for  $j = 0$  to  $N - 1$  do
5:     if  $j = i$  then
6:       continue
7:     end if
8:     load  $\mathbf{r}_j$  from the global memory and calculate  $\mathbf{r}_{ij} = \mathbf{r}_j - \mathbf{r}_i$ 
9:     apply the minimum image convention to  $\mathbf{r}_{ij}$ 
10:    if  $|\mathbf{r}_{ij}|^2 < r_c^2$  then
11:       $NL_{ik} \leftarrow j$ 
12:       $k \leftarrow k + 1$ 
13:    end if
14:  end for
15:   $NN_i \leftarrow k$ 
16: end if
```

In this kernel, the block size is S_b and the grid size is $\lceil N/S_b \rceil$. The **if** statement is used to avoid manipulating invalid memory.

3.4.2 A linear-scaling method

In this method, one partitions the system into cells and only searches for neighbors of a given particle in a small number of cells. In 3D, there are $3^3 = 27$ cells to be searched, which does not scale with N . The overall computational effort of this method scales as $27N_0N \sim N$, where N_0 is the average number of particles in one cell. Therefore, this is a linear-scaling, or $O(N)$ method.

When periodic boundary conditions are applied, the number of cells in each direction is determined as

$$N_x = \lfloor L_x/r_c \rfloor; \quad (3.1)$$

$$N_y = \lfloor L_y/r_c \rfloor; \quad (3.2)$$

$$N_z = \lfloor L_z/r_c \rfloor. \quad (3.3)$$

Here, L_x , L_y , and L_z are the box lengths. If a direction has free boundary conditions, we set the number of cells in that direction to 1. The total number of cells is thus $N_c = N_x N_y N_z$.

With the number of cells determined, we next determine the number of particles C_n ($n = 0, 1, \dots, N_c - 1$) in each cell. We need to use a CUDA kernel to do this. A pseudo code for the kernel is presented in Algorithm 2.

Algorithm 2 Determine the number of particles in each cell

Require: b is the block index

Require: t is the thread index

Require: S_b is the block size

Require: $i = S_b \times b + t$ is the particle index

Require: N is the number of particles

Require: C_n is the number of particles in cell n and has been initialized to 0

```

1: if  $i < N$  then
2:   calculate cell index  $n$  of particle  $i$ 
3:   atomic operation:  $C_n \leftarrow C_n + 1$ 
4: end if

```

Note that atomic operations (for integer data) are used to avoid write conflict. In the above kernel, we need to calculate the cell index of a given particle (with coordinate components x , y , and z). This is done by using a device function. The total cell index n is calculated from three indices:

$$n_x = \lfloor x/r_c \rfloor; \quad (3.4)$$

$$n_y = \lfloor y/r_c \rfloor; \quad (3.5)$$

$$n_z = \lfloor z/r_c \rfloor; \quad (3.6)$$

$$n = n_x + N_x n_y + N_x N_y n_z. \quad (3.7)$$

The cell index in the x -direction is required to be no less than 0 and no larger than $N_x - 1$. That is, when $n_x < 0$, we increase n_x by N_x ; when $n_x \geq N_x$, we decrease n_x by N_x . The other directions have similar requirements.

We then calculate the prefix sum (exclusive scan) S_n of C_n :

$$S_0 = 0; \quad (3.8)$$

$$S_n = \sum_{m=0}^{n-1} C_m \quad (1 \leq n \leq N_c - 1). \quad (3.9)$$

For this, we use the `thrust::exclusive_scan` function from the thrust library.

Now we can determine which particles are in which cells. We define a one-dimensional array I of length N , with I_{S_n} to $I_{S_n+C_n-1}$ being the indices of the particles in cell n . This array is constructed by using a CUDA kernel and a corresponding pseudo code is presented in Algorithm 3.

Up to now, the so-called cell list has been constructed. The remaining task is to convert the cell list to the Verlet neighbor list. This can be done by using a CUDA kernel similar to that for the $O(N^2)$ method. A pseudo code is presented in Algorithm 4.

Algorithm 3 Determine the array I containing the particle indices in the order of increasing cell index

Require: b is the block index

Require: t is the thread index

Require: S_b is the block size

Require: $i = S_b \times b + t$ is the particle index

Require: N is the number of particles

Require: C_n is the number of particles in cell n and has been initialized to 0

Require: S_n is the prefix sum of C_n as defined in the text

Require: I_{S_n} to $I_{S_n+C_n-1}$ are indices of the particles in cell n

```

1: if  $i < N$  then
2:   calculate cell index  $n$  of particle  $i$ 
3:    $I_{S_n+C_n} \leftarrow i$ 
4:   atomic operation:  $C_n \leftarrow C_n + 1$ 
5: end if

```

We note that the computation time used for the construction of the cell list is negligible compared to that used for the construction of the Verlet neighbor list from the cell list. However, we have to do this conversion because our efficient force evaluation algorithm [11] requires using the Verlet neighbor list rather than the cell list. Fortunately, the neighbor list usually only needs to be updated every tens of time steps with a typical skin distance (defined as the difference between the cutoff distance used for building the neighbor list and the cutoff distance used for force evaluation). In some simulations such as calculating the thermal conductivity of stable solids, the neighbor list even does not need to be updated during the simulation.

3.4.3 How to choose between the two versions?

The $O(N)$ version is faster than the $O(N^2)$ version in most cases. But sometimes we still use the $O(N^2)$ version. Here are the choices made in GPUMD:

- If the number of cells in any direction with periodic boundary conditions is less than 3, the $O(N)$ version is not applicable (as some neighbors will be counted twice) and the $O(N^2)$ version will be used. Therefore, if you hope to use the $O(N)$ version, you should make sure that the number of cells in any direction with periodic boundary conditions is no less than 3.
- The $O(N^2)$ version is only faster when the number of cells is very small. Take a 3D system with periodic boundary conditions in each direction for example, when the number of cells in each direction is 3, the $O(N^2)$ version is definitely faster than the $O(N)$ version, but the $O(N)$ version is already faster when the number of cells in each direction is 4. After doing some tests, we have decided to use the $O(N)$ version whenever the total number of cells is larger than 50. This might not be always optimal but is not a bad choice.
- The $O(N^2)$ version is deterministic, but the $O(N)$ version contains randomness, due to the use of atomic operations in the CUDA kernels. Using atomic operations, the order of the neighbor particles for a given particle can be different from run to run, which is not desirable for the purpose of debugging. In view of this, we have

Algorithm 4 Construct the Verlet neighbor list from the cell list

Require: b is the block index

Require: t is the thread index

Require: S_b is the block size

Require: $i = S_b \times b + t$ is the particle index

Require: N is the number of particles

Require: r_c^2 is the square of the cutoff distance for building the neighbor list

Require: C_n is the number of particles in cell n

Require: S_n is the prefix sum of C_n as defined in the text

Require: I_{S_n} to $I_{S_n+C_n-1}$ are the indices of the particle in cell n

Require: NN_i is the number neighbors for particle i

Require: NL_{ik} is the index of the k th neighbor of particle i

Require: \mathbf{r}_i is the position vector of particle i

```
1:  $k \leftarrow 0$ 
2: if  $i < N$  then
3:   load  $\mathbf{r}_i$  from the global memory
4:   calculate cell index  $n$  of particle  $i$ 
5:   for  $m$  in all the neighbor cells of cell  $n$  (including cell  $n$ ) do
6:     for  $l = 0$  to  $C_m - 1$  do
7:        $j \leftarrow I_{S_m+l}$ 
8:       if  $j = i$  then
9:         continue
10:      end if
11:      load  $\mathbf{r}_j$  from the global memory and calculate  $\mathbf{r}_{ij} = \mathbf{r}_j - \mathbf{r}_i$ 
12:      apply the minimum image convention to  $\mathbf{r}_{ij}$ 
13:      if  $|\mathbf{r}_{ij}|^2 < r_c^2$  then
14:         $NL_{ik} \leftarrow j$ 
15:         $k \leftarrow k + 1$ 
16:      end if
17:    end for
18:     $NN_i \leftarrow k$ 
19:  end for
20: end if
```

provided a compiling option (see Chapter 5 for details) to switch on the debugging mode, where the $O(N^2)$ version is always used.

In summary, GPUMD chooses an appropriate method for neighbor list construction automatically and no input is expected from the users.

3.4.4 How often should the neighbor list be updated?

The frequency of updating the neighbor list depends on the applications. If one simulates a stable solid system with the initial neighbor list containing all the neighbors that has possible interactions with a given particle during the whole simulation, the neighbor list does not need to be updated at all. In other cases, the neighbor list needs to be updated during the simulation. Usually, one can set an updating frequency such as 10, which means that the neighbor list will be updated every 10 integration steps. However, this

can be either inefficient or unsafe. Another way is to determine at every integration step whether the neighbor list needs to be updated by checking how far each atom has moved since the last neighbor list updating. It can be argued that the neighbor list should be updated when the maximum traveling distance of the particles since the last updating exceeds half of the skin distance (set by the user). As this check takes negligible time, GPUMD uses this method to determine automatically when the neighbor list is to be updated and thus does not expect the users to specify an updating frequency.

3.5 General formalisms for force evaluation and related calculations

3.5.1 General form of empirical potential functions

In classical molecular dynamics, the total potential energy U of a system can be written as the sum of site potentials U_i :

$$U = \sum_{i=1}^N U_i. \quad (3.10)$$

The site potential can have different forms in different potential models. Although there are numerous potential models proposed to date, they can be largely classified into two groups: two-body potentials and many-body potentials.

3.5.2 Force

For two-body potentials, the site potential U_i can be expressed as

$$U_i = \frac{1}{2} \sum_{j \neq i} U_{ij}(r_{ij}). \quad (3.11)$$

Here, $r_{ij} = |\mathbf{r}_j - \mathbf{r}_i|$ is the distance between particles i and j and $U_{ij}(r_{ij})$ is the pair potential between them. The total force acting on particle i can be derived to be:

$$\mathbf{F}_i = -\nabla_i U = \sum_{j \neq i} \frac{\partial U_{ij}(r_{ij})}{\partial r_{ij}} \frac{\mathbf{r}_{ij}}{r_{ij}}. \quad (3.12)$$

In this manual, we use the symbol \mathbf{r}_{ij} to denote the position difference vector from particle i to particle j :

$$\boxed{\mathbf{r}_{ij} \equiv \mathbf{r}_j - \mathbf{r}_i}. \quad (3.13)$$

The reader should bear this in mind when comparing the formulas in this manual with those in the literature, because many authors have used the opposite sign convention. One can also write the total force on particle i in the following form:

$$\mathbf{F}_i = \sum_{j \neq i} \mathbf{F}_{ij}, \quad (3.14)$$

where

$$\mathbf{F}_{ij} = \frac{\partial U_{ij}(r_{ij})}{\partial r_{ij}} \frac{\mathbf{r}_{ij}}{r_{ij}}. \quad (3.15)$$

is the pairwise force acting on particle i by particle j . Newton's third law is apparently valid here, in the sense that

$$\mathbf{F}_{ij} = -\mathbf{F}_{ji}. \quad (3.16)$$

In some many-body potentials such as the embedded-atom method potential [7], the site potential can not be written in the form of Eq. (3.11). In some other many-body potentials such as the Tersoff potential, the site potential can be written in the form of Eq. (3.11), but the U_{ij} in this equation does not only depend on the distance between particles i and j . The force formulas for many-body potentials have confused the community a lot. Recently, a well-defined force expression for general many-body potentials that explicitly respects Newton's third law has been derived as [14]:

$$\mathbf{F}_i = \sum_{j \neq i} \mathbf{F}_{ij}, \quad (3.17)$$

where

$$\mathbf{F}_{ij} = -\mathbf{F}_{ji} = \frac{\partial U_i}{\partial \mathbf{r}_{ij}} - \frac{\partial U_j}{\partial \mathbf{r}_{ji}} = \frac{\partial (U_i + U_j)}{\partial \mathbf{r}_{ij}}. \quad (3.18)$$

Here, $\partial U_i / \partial \mathbf{r}_{ij}$ is a shorthand notation for a vector with cartesian components $\partial U_i / \partial x_{ij}$, $\partial U_i / \partial y_{ij}$, and $\partial U_i / \partial z_{ij}$. This pairwise force expression for many-body potentials has been confirmed by Hardy [18] as well as by Chen and Diaz [5]. We have also confirmed its correctness by comparing with finite-difference calculations. This simple pairwise force expression for many-body potentials is the key for deriving well-defined expressions for other useful quantities such as virial stress and heat current, as discussed below.

3.5.3 Stress

Stress (tensor) is an important quantity in MD simulations. It consists of two parts: a virial part which is related to the force and an ideal-gas part which is related to the temperature. The virial part must be calculated along with force evaluation.

The validity of Newton's third law is crucial in simplifying the calculation of the virial stress. We know that the virial stress tensor is defined as

$$\mathbf{W} = \sum_i \mathbf{W}_i, \quad (3.19)$$

$$\mathbf{W}_i = \mathbf{r}_i \otimes \mathbf{F}_i. \quad (3.20)$$

Here, \mathbf{W}_i can be regarded as the per-atom virial stress. For periodic systems, the presence of absolute positions \mathbf{r}_i would cause problems. However, when Newton's third law is valid, one can rewrite the per-atom virial stress as

$$\mathbf{W}_i = -\frac{1}{2} \sum_{j \neq i} \mathbf{r}_{ij} \otimes \mathbf{F}_{ij}, \quad (3.21)$$

where only relative positions \mathbf{r}_{ij} are involved. Because Newton's third law also applies to many-body potentials, the above expression of virial stress is valid for any classical potential.

The ideal-gas part of the stress is isotropic, which is given by the ideal-gas pressure:

$$p_{\text{ideal}} = \frac{Nk_B T}{V}, \quad (3.22)$$

where N is the number of particles, k_B is Boltzmann's constant, T is the absolute temperature, and V is the volume of the system.

Combining the ideal-gas part and the virial part, the total stress tensor $\sigma^{\alpha\beta}$ can be expressed as:

$$\sigma^{\alpha\beta} = -\frac{1}{2V} \sum_i \sum_{j \neq i} r_{ij}^\alpha F_{ij}^\beta + \frac{Nk_B T}{V} \delta^{\alpha\beta}. \quad (3.23)$$

Here, α and β can be x , y , and z and $\delta^{\alpha\beta}$ is the Kronecker symbol. We will denote the diagonal part of the total stress tensor as a “vector” \mathbf{p} with components $p_x = \sigma^{xx}$, $p_y = \sigma^{yy}$, and $p_z = \sigma^{zz}$. If the system is isotropic, we usually average the diagonal terms to get a scalar:

$$p = \frac{1}{3} (p_x + p_y + p_z) = -\frac{1}{6V} \sum_i \sum_{j \neq i} \mathbf{r}_{ij} \cdot \mathbf{F}_{ij} + \frac{Nk_B T}{V}. \quad (3.24)$$

3.5.4 Heat current

GPUMD can be used to compute the lattice thermal conductivity using the Green-Kubo [17, 23] formula, which requires calculating the heat current.

In classical physics, the total heat current vector \mathbf{J} of a system is defined to be¹ the time derivative of the sum of the energy moments:

$$\mathbf{J} = \frac{d}{dt} \sum_i \mathbf{r}_i E_i. \quad (3.25)$$

Here, E_i is the site energy of particle i , which is the sum of the kinetic and potential energies:

$$E_i = \frac{1}{2} m_i \mathbf{v}_i^2 + U_i. \quad (3.26)$$

Using Leibniz's rule, we have

$$\mathbf{J} = \sum_i \mathbf{v}_i E_i + \sum_i \mathbf{r}_i \frac{d}{dt} E_i. \quad (3.27)$$

The first term on the right hand side is usually called the convective term and we do not need to evaluate it in the force-evaluation kernel. The second term,

$$\mathbf{J}^{\text{pot}} = \sum_i \mathbf{r}_i \frac{dE_i}{dt}, \quad (3.28)$$

is usually called the potential term and needs to be evaluated in the force-evaluation kernel.

When using the Green-Kubo method, we need to use periodic boundary conditions (at least in the transport directions). For two-body potentials, we can arrive at the following expression which is suitable for implementation:

$$\mathbf{J}^{\text{pot}} = -\frac{1}{2} \sum_i \sum_{j \neq i} \mathbf{r}_{ij} (\mathbf{F}_{ij} \cdot \mathbf{v}_i). \quad (3.29)$$

¹Actually, it is \mathbf{J}/V that has the dimension of heat current density (also called heat flux), which has the units of W m^{-2} in the international unit system. However, it is tedious to add the factor of $1/V$ in many of the subsequent equations.

This equation can be expressed in an equivalent way:

$$\mathbf{J}^{\text{pot}} = -\frac{1}{2} \sum_i \sum_{j \neq i} (\mathbf{r}_{ij} \otimes \mathbf{F}_{ij}) \cdot \mathbf{v}_i. \quad (3.30)$$

Therefore, we can also write it in terms of the per-atom virial:

$$\mathbf{J}^{\text{pot}} = \sum_i \mathbf{W}_i \cdot \mathbf{v}_i. \quad (3.31)$$

We can also define the per-atom heat current $\mathbf{J}_i^{\text{pot}}$ for the potential part in the following way:

$$\mathbf{J}^{\text{pot}} = \sum_i \mathbf{J}_i^{\text{pot}}; \quad (3.32)$$

$$\mathbf{J}_i^{\text{pot}} = \mathbf{W}_i \cdot \mathbf{v}_i. \quad (3.33)$$

However, we note that the above formula only applies to two-body potentials. For many-body potentials, it has been demonstrated [14] that the above virial-based formula is wrong and the correct one is

$$\boxed{\mathbf{J}_i^{\text{pot}} = \sum_{j \neq i} \mathbf{r}_{ij} \left(\frac{\partial U_j}{\partial \mathbf{r}_{ji}} \cdot \mathbf{v}_i \right)}. \quad (3.34)$$

The above heat current formula is usually applied in equilibrium simulations. In nonequilibrium simulations, the following expression for the nonequilibrium heat current [13] from a subsystem A to a subsystem B is more useful:

$$\boxed{Q_{A \rightarrow B} = - \sum_{i \in A} \sum_{j \in B} \left\langle \left(\frac{\partial U_i}{\partial \mathbf{r}_{ij}} \cdot \mathbf{v}_j - \frac{\partial U_j}{\partial \mathbf{r}_{ji}} \cdot \mathbf{v}_i \right) \right\rangle}, \quad (3.35)$$

This formula applies to general many-body potentials. For two-body potentials, it reduces to the following one:

$$Q_{A \rightarrow B}^{\text{two-body}} = -\frac{1}{2} \sum_{i \in A} \sum_{j \in B} \langle \mathbf{F}_{ij} \cdot (\mathbf{v}_i + \mathbf{v}_j) \rangle. \quad (3.36)$$

3.6 Integration by one step

The aim of time evolution is to find the phase trajectory

$$\{\mathbf{r}_i(t_1), \mathbf{v}_i(t_1)\}_{i=1}^N, \{\mathbf{r}_i(t_2), \mathbf{v}_i(t_2)\}_{i=1}^N, \dots \quad (3.37)$$

starting from the initial phase point

$$\{\mathbf{r}_i(t_0), \mathbf{v}_i(t_0)\}_{i=1}^N. \quad (3.38)$$

The time interval between two time points $\Delta t = t_1 - t_0 = t_2 - t_1 = \dots$ is called the time step.

The algorithm for integrating by one step depends on the ensemble type and other external conditions. We discuss them in detail below. There are many ensembles used in MD simulations, but we only consider the following 3 in the current version:

- The *NVE* ensemble, where the particle number N , the system volume V , and the total energy E are kept constant. It is also called the micro-canonical ensemble.
- The *NVT* ensemble, where the particle number N , the system volume V , and the temperature T are kept constant. It is also called the canonical ensemble.
- The *NPT* ensemble, where the particle number N , the pressure p , and the temperature T are kept constant. There seems to be no simple name for this important ensemble, but it is usually called the isothermal-isobaric ensemble.

3.6.1 The NVE ensemble and the velocity-Verlet algorithm

In the *NVE* ensemble, the dynamics of the system is Hamiltonian and the equations of motion can be derived from Hamilton's equations. Because these equations of motion have the time-reversal symmetry, a good numerical integrating method (an integrator) should preserve this symmetry.

One of the most widely used integrators which has the property of time-reversibility is the so-called velocity-Verlet method [38]. This integrator is also symplectic. These two properties make the velocity-Verlet integrator very stable for long-time simulations. Here are the velocity and position updating equations in the velocity-Verlet method:

$$\mathbf{v}_i(t_{m+1}) \approx \mathbf{v}_i(t_m) + \frac{\mathbf{F}_i(t_m) + \mathbf{F}_i(t_{m+1})}{2m_i} \Delta t; \quad (3.39)$$

$$\mathbf{r}_i(t_{m+1}) \approx \mathbf{r}_i(t_m) + \mathbf{v}_i(t_m) \Delta t + \frac{1}{2} \frac{\mathbf{F}_i(t_m)}{m_i} (\Delta t)^2, \quad (3.40)$$

where m_i is the mass of particle i .

The above velocity-Verlet integrator can be derived by finite-difference method (Taylor series expansion), but a more general method, which can be generalized to more sophisticated situations, is the classical time-evolution operator approach, or the Liouville operator approach [41]. In this approach, the time-evolution of a classical system by one step can be formally expressed as

$$\begin{pmatrix} \mathbf{r}_i(t + \Delta t) \\ \mathbf{p}_i(t + \Delta t) \end{pmatrix} = e^{iL\Delta t} \begin{pmatrix} \mathbf{r}_i(t) \\ \mathbf{p}_i(t) \end{pmatrix}, \quad (3.41)$$

where \mathbf{p}_i is the momentum of particle i and $e^{iL\Delta t}$ is called the classical evolution operator, which is the classical counterpart of the quantum evolution operator. The operator iL in the exponent of the evolution operator is called the Liouville operator and is defined by

$$iL(\text{anything}) = \{\text{anything}, H\} \equiv \sum_{i=1}^N \left(\frac{\partial H}{\partial \mathbf{p}_i} \cdot \frac{\partial}{\partial \mathbf{r}_i} - \frac{\partial H}{\partial \mathbf{r}_i} \cdot \frac{\partial}{\partial \mathbf{p}_i} \right) (\text{anything}). \quad (3.42)$$

Here, H is the Hamiltonian of the system. Because

$$\frac{\partial H}{\partial \mathbf{p}_i} = \frac{\mathbf{p}_i}{m_i} \text{ and } -\frac{\partial H}{\partial \mathbf{r}_i} = \mathbf{F}_i, \quad (3.43)$$

we have

$$iL = iL_1 + iL_2, \quad (3.44)$$

$$iL_1 = \sum_{i=1}^N \frac{\mathbf{p}_i}{m_i} \cdot \frac{\partial}{\partial \mathbf{r}_i}, \quad (3.45)$$

$$iL_2 = \sum_{i=1}^N \mathbf{F}_i \cdot \frac{\partial}{\partial \mathbf{p}_i}. \quad (3.46)$$

Here, we have divided the Liouville operator into two parts. In general, iL_1 and iL_2 do not commute, and therefore $e^{iL\Delta t} \neq e^{iL_1\Delta t}e^{iL_2\Delta t}$. However, there is an important theorem called the Trotter theorem, which can be used to derive the following approximation:

$$e^{iL\Delta t} \approx e^{iL_2\Delta t/2}e^{iL_1\Delta t}e^{iL_2\Delta t/2}. \quad (3.47)$$

Now, we can express the one-step integration as

$$\begin{pmatrix} \mathbf{r}_i(t + \Delta t) \\ \mathbf{p}_i(t + \Delta t) \end{pmatrix} \approx e^{iL_2\Delta t/2}e^{iL_1\Delta t}e^{iL_2\Delta t/2} \begin{pmatrix} \mathbf{r}_i(t) \\ \mathbf{p}_i(t) \end{pmatrix}. \quad (3.48)$$

To make further derivations, we note that for an arbitrary constant c , we have

$$e^{c\frac{\partial}{\partial x}}x = x + c. \quad (3.49)$$

Applying this identity to the right most operator in the above equation, we have

$$\begin{pmatrix} \mathbf{r}_i(t + \Delta t) \\ \mathbf{p}_i(t + \Delta t) \end{pmatrix} \approx e^{iL_2\Delta t/2}e^{iL_1\Delta t} \begin{pmatrix} \mathbf{r}_i(t) \\ \mathbf{p}_i(t) + \frac{\Delta t}{2}\mathbf{F}_i(t) \end{pmatrix}. \quad (3.50)$$

Then, applying the operator $e^{iL_1\Delta t}$, we have

$$\begin{pmatrix} \mathbf{r}_i(t + \Delta t) \\ \mathbf{p}_i(t + \Delta t) \end{pmatrix} \approx e^{iL_2\Delta t/2} \begin{pmatrix} \mathbf{r}_i(t) + \Delta t \frac{\mathbf{p}_i(t) + \frac{\Delta t}{2}\mathbf{F}_i(t)}{m_i} \\ \mathbf{p}_i(t) + \frac{\Delta t}{2}\mathbf{F}_i(t) \end{pmatrix}. \quad (3.51)$$

Last, applying the remaining operator $e^{iL_2\Delta t/2}$, we have

$$\begin{pmatrix} \mathbf{r}_i(t + \Delta t) \\ \mathbf{p}_i(t + \Delta t) \end{pmatrix} \approx \begin{pmatrix} \mathbf{r}_i(t) + \Delta t \frac{\mathbf{p}_i(t) + \frac{\Delta t}{2}\mathbf{F}_i(t)}{m_i} \\ \mathbf{p}_i(t) + \frac{\Delta t}{2}\mathbf{F}_i(t) + \frac{\Delta t}{2}\mathbf{F}_i(t + \Delta t) \end{pmatrix}. \quad (3.52)$$

It is clear that this equation is equivalent to Eqs. (3.39) and (3.40).

We can see that in the velocity-Verlet integrator, the position updating can be done in one step, but the velocity updating can only be done by two steps, one before force updating and the other after it. Algorithm 5 gives the pseudo code for the complete time-stepping in the *NVE* ensemble, including force updating.

3.6.2 Berendsen thermostat and barostat

Using the Berendsen thermostat, the integration algorithm in the *NVT* ensemble only requires an extra scaling of all the velocity components, as shown in Algorithm 6. For the *NPT* ensemble, the Berendsen barostat requires an extra scaling of positions and box lengths, as shown in Algorithm 7. The Berendsen thermostat and barostat are very suitable for equilibrating the system to a target temperature and pressure.

Algorithm 5 The whole time-stepping in the *NVE* ensemble.

1: update the velocities partially

$$\mathbf{v}_i \leftarrow \mathbf{v}_i + \frac{1}{2} \frac{\mathbf{F}_i}{m_i} \Delta t \quad (3.53)$$

2: update the positions completely

$$\mathbf{r}_i \leftarrow \mathbf{r}_i + \mathbf{v}_i \Delta t \quad (3.54)$$

3: update the forces

$$\mathbf{F}_i \leftarrow \mathbf{F}_i(\{\mathbf{r}_i\}) \quad (3.55)$$

4: complete updating the velocities

$$\mathbf{v}_i \leftarrow \mathbf{v}_i + \frac{1}{2} \frac{\mathbf{F}_i}{m_i} \Delta t \quad (3.56)$$

Algorithm 6 The whole time-stepping in the *NVT* ensemble using the Berendsen method.

- 1: perform the whole time-stepping for the *NVE* ensemble as shown in Algorithm 5
 - 2: scale the velocities
-

The velocities are scaled in the Berendsen thermostat in the following way:

$$\mathbf{v}_i^{\text{scaled}} = \mathbf{v}_i \sqrt{1 + \alpha_T \left(\frac{T_0}{T} - 1 \right)}. \quad (3.57)$$

Here, α_T is a dimensionless parameter, T_0 is the target temperature, and T is the instant temperature calculated from the current velocities $\{\mathbf{v}_i\}$. The parameter α_T should be positive and not larger than 1. When $\alpha_T = 1$, the above formula reduces to the simple velocity-scaling formula:

$$\mathbf{v}_i^{\text{scaled}} = \mathbf{v}_i \sqrt{\frac{T_0}{T}}. \quad (3.58)$$

A smaller α_T represents a weaker coupling between the system and the thermostat. Practically, any value of α_T in the range of $0.001 \sim 1$ can be used.

In the Berendsen barostat algorithm, the particle positions and box length in a given direction are scaled if periodic boundary conditions are applied to that direction. The scaling of the positions reads

$$\mathbf{r}_i^{\text{scaled}} = \mathbf{r}_i [1 - \alpha_p (\mathbf{p}_0 - \mathbf{p})]. \quad (3.59)$$

Here, α_p is a parameter and \mathbf{p}_0 (\mathbf{p}) is the target (instant) pressure in the three directions. The parameter α_p is not dimensionless, and it requires some try-and-error to find a good

Algorithm 7 The whole time-stepping in the *NPT* ensemble using the Berendsen method.

- 1: perform the whole time-stepping for the *NVE* ensemble as shown in Algorithm 5
 - 2: scale the velocities
 - 3: scale the positions and box lengths
-

value of it for a given system. A harder/softer system requires a smaller/larger value of α_p . In the unit system adopted by GPUMD, it is recommended that $\alpha_p = 10^{-4} \sim 10^{-2}$. Only directions with periodic boundary conditions will be affected by the barostat.

3.6.3 The Bussi-Donadio-Parrinello thermostat

The Berendsen thermostat does not generate a true NVT ensemble. As an extension of the Berendsen thermostat, the Bussi-Donadio-Parrinello (BDP) thermostat [3] incorporates a proper randomness into the velocity re-scaling factor and generates a true NVT ensemble. In the BDP thermostat, the velocities are scaled in the following way:

$$\mathbf{v}_i^{\text{scaled}} = \alpha \mathbf{v}_i; \quad (3.60)$$

$$\alpha^2 = e^{-\Delta t/\tau} + \frac{T_0}{TN_f} (1 - e^{-\Delta t/\tau}) \left(R_1^2 + \sum_{i=2}^{N_f} R_i^2 \right) + 2e^{-\Delta t/2\tau} R_1 \sqrt{\frac{T_0}{TN_f} (1 - e^{-\Delta t/\tau})}, \quad (3.61)$$

where τ is the relaxation time for this thermostat, N_f is the number of degrees of freedom in the thermostatted system, $\{R_i\}_{i=1}^{N_f}$ are N_f Gaussian distributed random numbers with zero mean and unit variance. We use the functions provided in the following site of Bussi to calculate the stochastic velocity re-scaling factor α : <https://sites.google.com/site/giovannibussi/Research/algorithms>.

3.6.4 The Langevin thermostat

We implement the Langevin thermostat using the integrator proposed by Bussi and Parrinello [4]. I will discuss the details of this integrator and its CUDA implementation (using the cuRAND library) when I have time. There is also a relaxation time τ which controls the strength of the coupling between the system and the thermostat.

3.6.5 Nosé-Hoover chain thermostat

The Nosé-Hoover chain method [31, 19, 29, 28, 41] is suitable for calculating equilibrium properties in a specific ensemble. In the current version of GPUMD, only the Nosé-Hoover chain thermostat is implemented. We hope to implement the Nosé-Hoover chain barostat in a future version.

The equations of motion in the Nosé-Hoover chain method with a chain length of M are

$$\frac{d}{dt} \mathbf{r}_i = \frac{\mathbf{p}_i}{m_i}, \quad (3.62)$$

$$\frac{d}{dt} \mathbf{p}_i = \mathbf{F}_i - \frac{\pi_0}{Q_0} \mathbf{p}_i, \quad (3.63)$$

$$\frac{d}{dt} \eta_k = \frac{\pi_k}{Q_k} \quad (k = 0, 1, \dots, M-1), \quad (3.64)$$

$$\frac{d}{dt} \pi_0 = 2 \left(\sum_i \frac{\mathbf{p}_i^2}{2m_i} - N_f \frac{k_B T_0}{2} \right) - \frac{\pi_1}{Q_1} \pi_0, \quad (3.65)$$

$$\frac{d}{dt} \pi_k = 2 \left(\frac{\pi_{k-1}^2}{2Q_{k-1}} - \frac{k_B T_0}{2} \right) - \frac{\pi_{k+1}}{Q_{k+1}} \pi_k \quad (k = 1, 2, \dots, M-2), \quad (3.66)$$

$$\frac{d}{dt}\pi_{M-1} = 2 \left(\frac{\pi_{M-2}^2}{2Q_{M-2}} - \frac{k_B T_0}{2} \right). \quad (3.67)$$

The optimal choice [29] for the thermostat masses is

$$Q_0 = N_f k_B T_0 \tau^2, \quad (3.68)$$

$$Q_k = k_B T_0 \tau^2 \quad (k = 1, 2, \dots, M-1), \quad (3.69)$$

where τ is a time parameter, whose value is usually chosen by try and error in practice. A good choice is $\tau = 100\Delta t$, where Δt is the time step for integration. Here η_k and π_k are the k th thermostat “position” and “momentum”, respectively.

An integration scheme for the NVT ensemble using the Nosé-Hoover chain can also be formulated using the approach of the time-evolution operator [28, 41]. The total Liouville operator for the equations of motion in the Nosé-Hoover chain method is [28, 41]

$$iL = iL_1 + iL_2 + iL_T; \quad (3.70)$$

$$iL_1 = \sum_{i=1}^N \frac{\mathbf{p}_i}{m_i} \cdot \frac{\partial}{\partial \mathbf{r}_i}; \quad (3.71)$$

$$iL_2 = \sum_{i=1}^N \mathbf{F}_i \cdot \frac{\partial}{\partial \mathbf{p}_i}; \quad (3.72)$$

$$iL_T = \sum_{k=0}^{M-1} \frac{\pi_k}{Q_k} \frac{\partial}{\partial \eta_k} + \sum_{k=0}^{M-2} \left(G_k - \frac{\pi_{k+1}}{Q_{k+1}} \pi_k \right) \frac{\partial}{\partial \pi_k} + G_{M-1} \frac{\partial}{\partial \pi_{M-1}} - \sum_{i=0}^{N-1} \frac{\pi_0}{Q_0} \mathbf{p}_i \cdot \frac{\partial}{\partial \mathbf{p}_i}; \quad (3.73)$$

$$G_0 = 2 \left(\sum_i \frac{\mathbf{p}_i^2}{2m_i} - N_f \frac{k_B T_0}{2} \right); \quad (3.74)$$

$$G_k = 2 \left(\frac{\pi_{k-1}^2}{2Q_{k-1}} - \frac{k_B T_0}{2} \right) \quad (k = 1, \dots, M-1). \quad (3.75)$$

That is, the Liouville operator for the NVT ensemble contains an extra term iL_T related to the thermostat variables, which is absent from that for the NVE ensemble.

The total time-evolution operator $e^{iL\Delta t}$ for one step can be factorized using the Trotter theorem as in the case of the NVE ensemble:

$$e^{iL\Delta t} \approx e^{iL_T\Delta t/2} e^{iL_2\Delta t/2} e^{iL_1\Delta t} e^{iL_2\Delta t/2} e^{iL_T\Delta t/2}. \quad (3.76)$$

Comparing this with the factorization in the NVE ensemble, we see that we only need to apply the operator $e^{iL_T\Delta t/2}$ before and after applying the usual velocity-Verlet integrator in the NVE ensemble.

The operator $e^{iL_T\Delta t/2}$ can be further factorized into some elementary factors using the Trotter theorem. First, we define the following decomposition of the operator iL_T :

$$iL_T = iL_{T1} + iL_{T2} + iL_{T3}, \quad (3.77)$$

$$iL_{T1} = \sum_{k=0}^{M-1} \frac{\pi_k}{Q_k} \frac{\partial}{\partial \eta_k}, \quad (3.78)$$

$$iL_{T2} = \sum_{k=0}^{M-2} \left(G_k - \frac{\pi_{k+1}}{Q_{k+1}} \pi_k \right) \frac{\partial}{\partial \pi_k} + G_{M-1} \frac{\partial}{\partial \pi_{M-1}}, \quad (3.79)$$

$$iL_{T3} = - \sum_{i=0}^{N-1} \frac{\pi_0}{Q_0} \mathbf{p}_i \cdot \frac{\partial}{\partial \mathbf{p}_i}. \quad (3.80)$$

We can then make the following factorization:

$$e^{iL_T \Delta t/2} \approx e^{iL_{T2} \Delta t/4} e^{iL_{T3} \Delta t/2} e^{iL_{T1} \Delta t/2} e^{iL_{T2} \Delta t/4}. \quad (3.81)$$

There are still a few terms in iL_{T2} and we need to factorize $e^{iL_{T2} \Delta t/4}$ further. We can factorize the $e^{iL_{T2} \Delta t/4}$ term on the right of the above equation as

$$e^{iL_{T2} \Delta t/4} \approx \prod_{k=0}^{M-2} \left(e^{-\frac{\Delta t}{8} \frac{\pi_{k+1}}{Q_{k+1}} \pi_k \frac{\partial}{\partial \pi_k}} e^{\frac{\Delta t}{4} G_k \frac{\partial}{\partial \pi_k}} e^{-\frac{\Delta t}{8} \frac{\pi_{k+1}}{Q_{k+1}} \pi_k \frac{\partial}{\partial \pi_k}} \right) e^{\frac{\Delta t}{4} G_{M-1} \frac{\partial}{\partial \pi_{M-1}}} \quad (3.82)$$

and correspondingly factorize that on the left as

$$e^{iL_{T2} \Delta t/4} \approx e^{\frac{\Delta t}{4} G_{M-1} \frac{\partial}{\partial \pi_{M-1}}} \prod_{k=M-2}^0 \left(e^{-\frac{\Delta t}{8} \frac{\pi_{k+1}}{Q_{k+1}} \pi_k \frac{\partial}{\partial \pi_k}} e^{\frac{\Delta t}{4} G_k \frac{\partial}{\partial \pi_k}} e^{-\frac{\Delta t}{8} \frac{\pi_{k+1}}{Q_{k+1}} \pi_k \frac{\partial}{\partial \pi_k}} \right). \quad (3.83)$$

Algorithm 8 The whole time-stepping in the NVT ensemble using the Nosé-Hoover chain method.

- 1: apply the operator $e^{iL_T \Delta t/2}$ except for $e^{iL_{T3} \Delta t/2}$ within it and save the value of $e^{-(\pi_0/Q_0) \Delta t/2}$
 - 2: scale the velocity components of all the particles by the factor $e^{-(\pi_0/Q_0) \Delta t/2}$
 - 3: perform the whole time-stepping for the NVE ensemble as shown in Algorithm 5
 - 4: apply the operator $e^{iL_T \Delta t/2}$ except for $e^{iL_{T3} \Delta t/2}$ within it and save the value of $e^{-(\pi_0/Q_0) \Delta t/2}$
 - 5: scale the velocity components of all the particles by the factor $e^{-(\pi_0/Q_0) \Delta t/2}$
-

It can be shown that the effect of the operator $e^{c x \frac{\partial}{\partial x}}$ on x is to scale it by a factor of e^c :

$$e^{c x \frac{\partial}{\partial x}} x = e^c x. \quad (3.84)$$

Therefore, the effect of the operator $e^{iL_{T3} \Delta t/2}$ is to scale the momenta of all the particles in the system by a uniform factor $e^{-(\pi_0/Q_0) \Delta t/2}$. Although this operator appears in the factorization of $e^{iL_T \Delta t/2}$, it does not affect the thermostat variables. Therefore, when applying the operator $e^{iL_T \Delta t/2}$, we only need to update the variables related to the thermostats and save this factor for later use when we update the variables for the particles. In this way, the update for the thermostat variables and that for the particle variables are separated. Algorithm 8 presents the pseudo code for the whole time-stepping in the NVT ensemble using the Nosé-Hoover chain method.

There are also some other tricks in the algorithm. For details, we refer to the excellent book by Tuckerman [41].

3.6.6 The isothermal-isobaric ensemble

We follow the excellent book by Tuckerman [41]. **This ensemble has not been implemented in GPUMD yet but I will do it in the near future.**

In this ensemble, we introduce a box matrix \mathbf{h} :

$$\mathbf{h} = \begin{pmatrix} a_x & b_x & c_x \\ a_y & b_y & c_y \\ a_z & b_z & c_z \end{pmatrix} \quad (3.85)$$

Generally, there are 6 nonzero matrix elements in the box matrix, but in the current version of GPUMD, we only consider the special case where the box matrix is diagonal. This corresponds to orthogonal box. We may consider the general triclinic box in the future. However, the formulations below are presented in a general form.

Important: we use bold face to represent (second rank) tensors and italic bold face to represent vectors. The dot product of a tensor and a vector is a vector and the dot product of two tensors is a tensor. The unit tensor is denoted as \mathbf{I} . The tensors are also regarded as matrices and various matrix operators (such as determinant, trace, transpose, and inverse) are used.

List of dynamic variables involved:

- particle positions: \mathbf{r}_i
- particle momenta: \mathbf{p}_i
- box matrix: \mathbf{h}
- box momenta: \mathbf{p}_g
- particle-thermostat positions: η_k
- particle-thermostat momenta: π_k
- box-thermostat positions: ξ_k
- box-thermostat momenta: β_k

The equations of motion for the particles are:

$$\frac{d}{dt}\mathbf{r}_i = \frac{\mathbf{p}_i}{m_i} + \frac{\mathbf{p}_g}{W_g} \cdot \mathbf{r}_i; \quad (3.86)$$

$$\frac{d}{dt}\mathbf{p}_i = \mathbf{F}_i - \frac{\pi_0}{Q_0}\mathbf{p}_i - \frac{\mathbf{p}_g}{W_g} \cdot \mathbf{p}_i - \frac{1}{N_f} \frac{\text{Tr}[\mathbf{p}_g]}{W_g} \mathbf{p}_i; \quad (3.87)$$

The equations of motion for the box are:

$$\frac{d\mathbf{h}}{dt} = \frac{\mathbf{p}_g}{W_g} \cdot \mathbf{h}; \quad (3.88)$$

$$\frac{d\mathbf{p}_g}{dt} = \det[\mathbf{h}](\boldsymbol{\sigma} - \boldsymbol{\sigma}^{\text{ext}}) + \frac{1}{N_f} \sum_i \frac{\mathbf{p}_i^2}{2m_i} \mathbf{I} - \frac{\pi'_0}{Q'_0} \mathbf{p}_g; \quad (3.89)$$

The equations of motion for the thermostat chain coupling to the particles are:

$$\frac{d}{dt}\eta_k = \frac{\pi_k}{Q_k} \quad (k = 0, 1, \dots, M-1), \quad (3.90)$$

$$\frac{d}{dt}\pi_0 = 2 \left(\sum_i \frac{\mathbf{p}_i^2}{2m_i} - dN \frac{k_B T}{2} \right) - \frac{\pi_1}{Q_1} \pi_0, \quad (3.91)$$

$$\frac{d}{dt}\pi_k = 2 \left(\frac{\pi_{k-1}^2}{2Q_{k-1}} - \frac{k_B T}{2} \right) - \frac{\pi_{k+1}}{Q_{k+1}} \pi_k \quad (k = 1, 2, \dots, M-2), \quad (3.92)$$

$$\frac{d}{dt}\pi_{M-1} = 2 \left(\frac{\pi_{M-2}^2}{2Q_{M-2}} - \frac{k_B T}{2} \right). \quad (3.93)$$

The equations of motion for the thermostat chain coupling to the box are:

$$\frac{d}{dt}\xi_k = \frac{\beta_k}{Q'_k} \quad (k = 0, 1, \dots, M-1), \quad (3.94)$$

$$\frac{d}{dt}\beta_0 = 2 \left(\frac{\text{Tr}[\mathbf{p}_g^T \mathbf{p}_g]}{2W_g} - (d^2 - d) \frac{k_B T}{2} \right) - \frac{\beta_1}{Q'_1} \beta_0, \quad (3.95)$$

$$\frac{d}{dt}\beta_k = 2 \left(\frac{\beta_{k-1}^2}{2Q'_{k-1}} - \frac{k_B T}{2} \right) - \frac{\beta_{k+1}}{Q'_{k+1}} \beta_k \quad (k = 1, 2, \dots, M-2), \quad (3.96)$$

$$\frac{d}{dt}\beta_{M-1} = 2 \left(\frac{\beta_{M-2}^2}{2Q'_{M-2}} - \frac{k_B T}{2} \right). \quad (3.97)$$

The optimal choice [29] for the box mass is

$$W_g = (dN + 1)k_B T \tau_b^2, \quad (3.98)$$

where τ_b is a time parameter, whose value is usually chosen by try and error in practice. A good choice is $\tau_b = 1000\Delta t$, where Δt is the time step for integration.

An integration scheme for the NPT ensemble using the Nosé-Hoover chain can also be formulated using the approach of the time-evolution operator [28, 41]. The total Liouville operator for the equations of motion in the Nosé-Hoover chain method is [28, 41]

$$iL = iL_1 + iL_2 + iL_{g1} + iL_{g2} + iL_{Tp} + iL_{Tb}, \quad (3.99)$$

where iL_{Tp} and iL_{Tb} are the Liouville operators for the Nosé-Hoover chain coupling to the particles and the box, respectively. They are similar to the operator iL_T presented in the last subsection. The other four operators are:

$$iL_1 = \sum_{i=1}^N \left[\frac{\mathbf{p}_i}{m_i} + \frac{\mathbf{p}_g}{W_g} \cdot \mathbf{r}_i \right] \cdot \frac{\partial}{\partial \mathbf{r}_i}; \quad (3.100)$$

$$iL_2 = \sum_{i=1}^N \left[\mathbf{F}_i - \left(\frac{\mathbf{p}_g}{W_g} + \frac{1}{N_f} \frac{\text{Tr}[\mathbf{p}_g]}{W_g} \mathbf{I} \right) \cdot \mathbf{p}_i \right] \cdot \frac{\partial}{\partial \mathbf{p}_i}; \quad (3.101)$$

$$iL_{g1} = \frac{\mathbf{p}_g \cdot \mathbf{h}}{W_g} \cdot \frac{\partial}{\partial \mathbf{h}}; \quad (3.102)$$

$$iL_{g2} = \left(\det[\mathbf{h}](\boldsymbol{\sigma} - \boldsymbol{\sigma}^{\text{ext}}) + \frac{1}{N_f} \sum_{i=1}^N \frac{\mathbf{p}_i^2}{m_i} \mathbf{I} \right) \cdot \frac{\partial}{\partial \mathbf{p}_g}. \quad (3.103)$$

The total time-evolution operator $e^{iL\Delta t}$ for one step can be factorized using the Trotter theorem (the factorization is not unique):

$$e^{iL} \approx e^{iL_{Tb}\Delta t/2} e^{iL_{Tp}\Delta t/2} e^{iL_{g2}\Delta t/2} e^{iL_2\Delta t/2} e^{iL_{g1}\Delta t} e^{iL_1\Delta t} e^{iL_2\Delta t/2} e^{iL_{g2}\Delta t/2} e^{iL_{Tp}\Delta t/2} e^{iL_{Tb}\Delta t/2}. \quad (3.104)$$

The two operators related to the box, $e^{iL_{g1}\Delta t}$ and $e^{iL_{g2}\Delta t/2}$, are simple translations. The operators related to the Nosé-Hoover chain have been discussed in detail in the previous subsection. The remaining operators related to the particles are more complicated. The operator $e^{iL_1\Delta t}$ gives

$$\mathbf{r}_i(\Delta t) = \mathbf{O}^T \cdot \mathbf{D} \cdot \mathbf{O} \cdot \mathbf{r}_i(0) + \Delta t \mathbf{O}^T \cdot \tilde{\mathbf{D}} \cdot \mathbf{O} \cdot \mathbf{v}_i; \quad (3.105)$$

The operator $e^{iL_2\Delta t/2}$ gives

$$\mathbf{v}_i(\Delta t/2) = \mathbf{O}^T \cdot \boldsymbol{\Delta} \cdot \mathbf{O} \cdot \mathbf{v}_i(0) + \frac{\Delta t}{2m_i} \mathbf{O}^T \cdot \tilde{\boldsymbol{\Delta}} \cdot \mathbf{O} \cdot \mathbf{F}_i. \quad (3.106)$$

Here, the operator \mathbf{O} diagonalizes the box velocity:

$$\begin{pmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{pmatrix} = \mathbf{O} \cdot \frac{\mathbf{p}_g}{W_g} \cdot \mathbf{O}^T, \quad (3.107)$$

where λ_1 , λ_2 , and λ_3 are the eigenvalues of the velocity matrix. The other matrices in the above equations are all diagonal with the elements:

$$D_{\alpha\alpha} = e^{\lambda_\alpha \Delta t}; \quad (3.108)$$

$$\tilde{D}_{\alpha\alpha} = e^{\lambda_\alpha \Delta t/2} \frac{\sinh[\lambda_\alpha \Delta t/2]}{\lambda_\alpha \Delta t/2}; \quad (3.109)$$

$$\Delta_{\alpha\alpha} = \exp \left[- \left(\lambda_\alpha + \frac{1}{N_f} \frac{\text{Tr}[\mathbf{p}_g]}{W_g} \right) \frac{\Delta t}{2} \right]; \quad (3.110)$$

$$\tilde{\Delta}_{\alpha\alpha} = \exp \left[- \left(\lambda_\alpha + \frac{1}{N_f} \frac{\text{Tr}[\mathbf{p}_g]}{W_g} \right) \frac{\Delta t}{4} \right] \frac{\sinh \left[\left(\lambda_\alpha + \frac{1}{N_f} \frac{\text{Tr}[\mathbf{p}_g]}{W_g} \right) \frac{\Delta t}{4} \right]}{\left(\lambda_\alpha + \frac{1}{N_f} \frac{\text{Tr}[\mathbf{p}_g]}{W_g} \right) \frac{\Delta t}{4}}. \quad (3.111)$$

As the argument in the sinh function here is small, it is better to evaluate the ratio $\sinh(x)/x$ using a power series expansion:

$$\frac{\sinh(x)}{x} \approx \sum_{n=0}^m \frac{1}{(2n+1)!} x^{2n}. \quad (3.112)$$

where $m = 5$ is accurate enough.

3.7 Heat transport

3.7.1 Heat current autocorrelation and lattice thermal conductivity

In MD simulations, a popular approach of computing the lattice thermal conductivity is to use the Green-Kubo formula [17, 23]. In this method, the running lattice thermal conductivity along the x -direction (similar expressions apply to other directions) can be expressed as an integral of the heat current autocorrelation (HAC):

$$\kappa_{xx}(t) = \frac{1}{k_B T^2 V} \int_0^t dt' \text{HAC}_{xx}(t'). \quad (3.113)$$

Here, k_B is Boltzmann's constant, V is the volume of the simulated system, T is the absolute temperature, and t is the correlation time. The HAC is

$$\text{HAC}_{xx}(t) = \langle J_x(0) J_x(t) \rangle, \quad (3.114)$$

where $J_x(0)$ and $J_x(t)$ are the total heat current of the system at two time points separated by an interval of t . The symbol $\langle \rangle$ means that the quantity inside will be averaged over different time origins.

The calculation of the heat current \mathbf{J} has been discussed earlier. Here, we assume that we have calculated the total heat current of the system at M number of time points and saved them into the global memory. The time interval $\Delta\tau$ between the time points here needs not to be the same as the time step Δt used in the time-stepping. Usually, $\Delta\tau = 10\Delta t$ is a good choice. From the N_d heat current data, we can calculate at most N_d HAC data $\text{HAC}_{xx}(t)$, with $t = 0, \Delta\tau, 2\Delta\tau, \dots, (N_d - 1)\Delta\tau$. However, a correlation function becomes more and more noisy as the correlation time increases and in practical applications, one has to make sure that the production time $N_d\Delta\tau$ is much larger than the maximum correlation time t_{\max} one needs. The number of HAC data N_c is related to the maximum correlation time by $t_{\max} = N_c\Delta\tau$. In most cases, $N_c = N_d/10$ is a good choice. It is also convenient to use the same number of time origins, $N_d - N_c$, to do the time-average for each correlation time. With these considerations, we arrive at the following explicit expression for the HAC:

$$\text{HAC}_{xx}(n_c\Delta\tau) = \frac{1}{N_d - N_c} \sum_{m=0}^{N_d - N_c - 1} J_x(m\Delta\tau) J_x((m + n_c)\Delta\tau), \quad (3.115)$$

where $n_c = 0, 1, 2, \dots, N_c - 1$.

Because the HAC at different correlation times can be calculated independently, we can simply use one CUDA-block for one point of the HAC data. Shared memory is used to do the summation in an binary-reduction way. An algorithm for the CUDA implementation can be found in Ref. [15].

Last, we note that for 2D materials, the heat current can be naturally decomposed into an in-plane component and an out-of-plane component. The lattice thermal conductivity can be decomposed accordingly. See Ref. [13] for more details.

3.7.2 NEMD method

The thermal conductivity of a finite-length system can also be computed by the NEMD method. In this method, a temperature gradient is established by externally generating

a nonequilibrium heat current. If the steady-state heat current is Q and the established temperature gradient is ∇T , the thermal conductivity is calculated according to Fourier's law as

$$\kappa = \frac{Q}{S|\nabla T|}, \quad (3.116)$$

where S is the cross-sectional area perpendicular to the transport direction.

The nonequilibrium heat current can be generated in various ways, such as the velocity re-scaling method [20, 21] and the momentum swapping method [30]. However, some guesswork is needed in choosing appropriate parameters in these methods. Another method is to couple the source/sink region to a thermostat with a higher/lower temperature. By setting the temperature difference, one has better control to the temperature gradient. When steady state is achieved, a temperature gradient will be established and the heat current Q can be calculated as the energy transfer rate dE/dt between the source/sink and the thermostat:

$$Q = \frac{dE}{dt}. \quad (3.117)$$

It has been shown [13] that Eq. (3.117) gives consistent heat current as calculated by Eq. (3.35). As in the case of the Green-Kubo method, the nonequilibrium heat current defined in Eq. (3.35) can also be decomposed into in-plane and out-of-plane components for 2D materials. See Ref. [13] for details.

We have tested that all these NEMD methods can give consistent results. To make the code simpler, we have only kept the version using thermostats in GPUMD.

In the framework of the NEMD method, one can also calculate spectrally decomposed thermal conductivity (or conductance) using the method as described in Ref. [13]. This method is based on the works by Sääskilahti *et al.* [35, 36] and Zhou *et al.* [48] but the formulation in Ref. [13] is simpler and does not use approximations for the heat current. In this method, one first calculates the following correlation function:

$$K_{A \rightarrow B}(t) = - \sum_{i \in A} \sum_{j \in B} \left\langle \left(\frac{\partial U_i}{\partial \mathbf{r}_{ij}}(0) \cdot \mathbf{v}_j(t) - \frac{\partial U_j}{\partial \mathbf{r}_{ji}}(0) \cdot \mathbf{v}_i(t) \right) \right\rangle, \quad (3.118)$$

which reduces to the nonequilibrium heat current as defined in Eq. (3.35) when $t = 0$. Then one can define the following Fourier transform pairs:

$$\tilde{K}_{A \rightarrow B}(\omega) = \int_{-\infty}^{\infty} dt e^{i\omega t} K_{A \rightarrow B}(t); \quad (3.119)$$

$$K_{A \rightarrow B}(t) = \int_{-\infty}^{\infty} \frac{d\omega}{2\pi} e^{-i\omega t} \tilde{K}_{A \rightarrow B}(\omega). \quad (3.120)$$

By setting $t = 0$ in the equation above, we can get the following spectral decomposition of the nonequilibrium heat current:

$$Q_{A \rightarrow B} = \int_0^{\infty} \frac{d\omega}{2\pi} \left[2\tilde{K}_{A \rightarrow B}(\omega) \right]. \quad (3.121)$$

From the spectral decomposition of the nonequilibrium heat current, one can deduce the spectrally decomposed thermal conductivity corresponding to a given finite system with

a temperature gradient ∇T ,

$$\kappa(\omega) = \frac{2\tilde{K}_{A \rightarrow B}(\omega)}{S|\nabla T|} \quad \text{with} \quad \kappa = \int_0^\infty \frac{d\omega}{2\pi} \kappa(\omega), \quad (3.122)$$

or a spectrally decomposed thermal conductance corresponding to a given segment or junction (interface) with a temperature difference ΔT ,

$$G(\omega) = \frac{2\tilde{K}_{A \rightarrow B}(\omega)}{S|\Delta T|} \quad \text{with} \quad G = \int_0^\infty \frac{d\omega}{2\pi} G(\omega). \quad (3.123)$$

For 2D materials, one can further consider the in-out decomposition. See Ref. [13] for details.

3.7.3 HNEMD method

The homogeneous nonequilibrium molecular dynamics (HNEMD) method for heat transport by Evans [10] has been recently generalized to general many-body potentials [12]. We have applied [45, 9] this method to study heat transport in various materials described by the Tersoff or the SW potential. All these manuscripts are still under review and we plan to discuss this method some time later.

3.8 Velocity autocorrelation and related quantities

Velocity autocorrelation (VAC) is an important quantity in MD simulations. On the one hand, its integral with respect to the correlation time gives the running diffusion constant, which is equivalent to that obtained by a time derivative of the mean square displacement (MSD). On the other hand, its Fourier transform is the phonon density of states (PDOS) [8].

3.8.1 Running diffusion coefficient

The VAC is a single-particle correlation function. This means that we can define the VAC for individual particles. For particle i , the VAC along the x direction is defined as

$$\langle v_{xi}(0)v_{xi}(t) \rangle. \quad (3.124)$$

Then, one can define the mean VAC for any number of particles. In the current version of GPUMD, it is assumed that one wants to calculate the mean VAC in the whole simulated system:

$$\text{VAC}_{xx}(t) = \frac{1}{N} \sum_{i=1}^N \langle v_{xi}(0)v_{xi}(t) \rangle. \quad (3.125)$$

The order between the time-average (denoted by $\langle \rangle$) and the space-average (the average over the particles) can be changed:

$$\text{VAC}_{xx}(t) = \left\langle \frac{1}{N} \sum_{i=1}^N v_{xi}(0)v_{xi}(t) \right\rangle. \quad (3.126)$$

Using the same conventions as in the case of HAC calculations, we have the following explicit expression for the VAC:

$$\text{VAC}_{xx}(n_c \Delta\tau) = \frac{1}{(N_d - N_c)N} \sum_{m=0}^{N_d - N_c - 1} \sum_{i=1}^N v_{xi}(m \Delta\tau) v_{xi}((m + n_c) \Delta\tau), \quad (3.127)$$

where $n_c = 0, 1, 2, \dots, N_c - 1$. The algorithm for calculating the VAC is quite similar to that for calculating the HAC and it thus omitted.

After obtaining the VAC, we can calculate the running diffusion constant $D_{xx}(t)$ as

$$D_{xx}(t) = \int_0^t dt' \text{VAC}_{xx}(t'). \quad (3.128)$$

One can prove that this is equivalent to the time-derivative of the MSD, i.e., the Einstein formula:

$$D_{xx}(t) = \frac{1}{2} \frac{d}{dt} \Delta x^2(t), \quad (3.129)$$

where the MSD $\Delta x^2(t)$ is defined as

$$\Delta x^2(t) = \left\langle \frac{1}{N} \sum_{i=1}^N [x_i(t) - x_i(0)]^2 \right\rangle = \frac{1}{N} \sum_{i=1}^N \langle [x_i(t) - x_i(0)]^2 \rangle. \quad (3.130)$$

Here is the proof. Starting from the relation between position and velocity,

$$x_i(t) - x_i(0) = \int_0^t dt' v_{xi}(t'), \quad (3.131)$$

we have

$$[x_i(t) - x_i(0)]^2 = \int_0^t dt' v_{xi}(t') \int_0^t dt'' v_{xi}(t'') = \int_0^t dt' \int_0^t dt'' v_{xi}(t') v_{xi}(t''). \quad (3.132)$$

Then, the MSD can be expressed as

$$\Delta x^2(t) = \frac{1}{N} \sum_{i=1}^N \int_0^t dt' \int_0^t dt'' \langle v_{xi}(t') v_{xi}(t'') \rangle. \quad (3.133)$$

Using Leibniz's rule, we have

$$D_{xx}(t) = \frac{1}{2} \frac{d}{dt} \Delta x^2(t) = \frac{1}{N} \sum_{i=1}^N \int_0^t dt' \langle v_{xi}(t) v_{xi}(t') \rangle, \quad (3.134)$$

which can be rewritten as

$$D_{xx}(t) = \frac{1}{N} \sum_{i=1}^N \int_0^t dt' \langle v_{xi}(0) v_{xi}(t' - t) \rangle. \quad (3.135)$$

Letting $\tau = t' - t$, we get (note that here t is considered as a constant)

$$D_{xx}(t) = \frac{1}{N} \sum_{i=1}^N \int_{-t}^0 d\tau \langle v_{xi}(0) v_{xi}(\tau) \rangle, \quad (3.136)$$

which can be rewritten as

$$D_{xx}(t) = \frac{1}{N} \sum_{i=1}^N \int_{-t}^0 d\tau \langle v_{xi}(-\tau) v_{xi}(0) \rangle. \quad (3.137)$$

Letting $t' = -\tau$, we finally get

$$D_{xx}(t) = \frac{1}{N} \sum_{i=1}^N \int_0^t dt' \langle v_{xi}(t') v_{xi}(0) \rangle = \int_0^t dt' \text{VAC}_{xx}(t'). \quad (3.138)$$

We thus have derived the Green-Kubo formula from the Einstein formula.

In summary,

- The derivative of half of the MSD gives the running diffusion coefficient.
- The integral of the VAC gives the running diffusion coefficient.
- One can obtain the MSD by integrating the VAC twice (numerically).

3.8.2 Phonon density of states

It is interesting that the same VAC can be used to compute the PDOS, as first demonstrated by Dickey and Paskin [8]. The PDOS is simply the Fourier transform of the normalized VAC:

$$\rho_x(\omega) = \int_{-\infty}^{\infty} dt e^{i\omega t} \text{VAC}_{xx}(t). \quad (3.139)$$

Here, $\text{VAC}_{xx}(t)$ should be understood as the normalized function $\text{VAC}_{xx}(t)/\text{VAC}_{xx}(0)$. Although it looks simple, it does not mean that one can get the correct PDOS by a naive fast Fourier transform (FFT) routine. Actually, this computation is very cheap and we do not need FFT at all. What we need is a discrete cosine transform. To see this, we first note that, by definition, $\text{VAC}_{xx}(-t) = \text{VAC}_{xx}(t)$. Using this, we have

$$\rho_x(\omega) = \int_{-\infty}^{\infty} dt \cos(\omega t) \text{VAC}_{xx}(t). \quad (3.140)$$

Because we only have the VAC data at the N_c discrete time points, the above integral is approximated by the following discrete cosine transform:

$$\rho_x(\omega) \approx \sum_{n_c=0}^{N_c-1} (2 - \delta_{n_c 0}) \Delta\tau \cos(\omega n_c \Delta\tau) \text{VAC}_{xx}(n_c \Delta\tau). \quad (3.141)$$

Here, $\delta_{n_c 0}$ is the Kronecker δ function and the factor $(2 - \delta_{n_c 0})$ accounts for the fact that there is only one point for $t = 0$ and there are two equivalent points for $t \neq 0$. Last, we note that a window function is needed to suppress the unwanted Gibbs oscillation in the calculated PDOS. In GPUMD, the Hann window $H(n_c)$ is applied:

$$\rho_x(\omega) \approx \sum_{n_c=0}^{N_c-1} (2 - \delta_{n_c 0}) \Delta\tau \cos(\omega n_c \Delta\tau) \text{VAC}_{xx}(n_c \Delta\tau) H(n_c); \quad (3.142)$$

$$H(n_c) = \frac{1}{2} \left[\cos\left(\frac{\pi n_c}{N_c}\right) + 1 \right]. \quad (3.143)$$

Here are some comments on the normalization of the PDOS. In the literature, one usually uses an arbitrary unit for the PDOS, but it actually has a dimension of [time], and an appropriate unit for it can be 1/THz or ps. The normalization of $\rho_x(\omega)$ can be determined by the inverse Fourier transform:

$$\text{VAC}_{xx}(t) = \int_{-\infty}^{\infty} \frac{d\omega}{2\pi} e^{-i\omega t} \rho_x(\omega). \quad (3.144)$$

As we have normalized the VAC, we have

$$1 = \text{VAC}_{xx}(0) = \int_{-\infty}^{\infty} \frac{d\omega}{2\pi} \rho_x(\omega). \quad (3.145)$$

Because $\rho_x(-\omega) = \rho_x(\omega)$, we have

$$\int_0^{\infty} \frac{d\omega}{2\pi} \rho_x(\omega) = \frac{1}{2}. \quad (3.146)$$

The calculated PDOS should meet this normalization condition (approximately).

Chapter 4

Potential models implemented in GPUMD

In this chapter, we discuss in detail the potential models that have been implemented in GPUMD. The formats of the potential files for the potential models are also presented. In the potential files, the unit of energy is eV and the unit of length is Å. One should be able to deduce the dimension of all parameters from the relevant equations defining them.

4.1 Conventions

The following conventions are used in this chapter:

- The position difference vector from particle i to particle j is denoted

$$\boxed{\mathbf{r}_{ij} \equiv \mathbf{r}_j - \mathbf{r}_i}. \quad (4.1)$$

The component form of this equation is

$$x_{ij} = x_j - x_i; \quad y_{ij} = y_j - y_i; \quad z_{ij} = z_j - z_i. \quad (4.2)$$

- The angle formed by \mathbf{r}_{ij} and \mathbf{r}_{ik} is denoted

$$\boxed{\cos \theta_{ijk} = \cos \theta_{ikj} = \frac{\mathbf{r}_{ij} \cdot \mathbf{r}_{ik}}{r_{ij} r_{ik}}}. \quad (4.3)$$

- The derivative $\partial/\partial\mathbf{r}_{ij}$ should be understood as a vector operator with components $\partial/\partial x_{ij}$, $\partial/\partial y_{ij}$, and $\partial/\partial z_{ij}$.
- It is easy to verify that

$$\boxed{\frac{\partial r_{ij}}{\partial \mathbf{r}_{ij}} = \frac{\mathbf{r}_{ij}}{r_{ij}}}. \quad (4.4)$$

- It is also easy to verify that

$$\boxed{\frac{\partial \cos \theta_{ijk}}{\partial \mathbf{r}_{ij}} = \frac{1}{r_{ij}} \left[\frac{\mathbf{r}_{ik}}{r_{ik}} - \frac{\mathbf{r}_{ij}}{r_{ij}} \cos \theta_{ijk} \right]}. \quad (4.5)$$

- If f is a function of x , then $f'(x)$ is understood as $\partial f / \partial x$.

In the force evaluation kernel for any potential model, the key quantities to be calculated are $\frac{\partial U_i}{\partial \mathbf{r}_{ij}}$ and $\frac{\partial U_j}{\partial \mathbf{r}_{ji}}$, the latter being related to the former by an exchange of the indices, $i \leftrightarrow j$. We thus need to derive an explicit expression of $\frac{\partial U_i}{\partial \mathbf{r}_{ij}}$ for each potential model. This is trivial for two-body potentials and we only present expressions for many-body potentials. We will call $\frac{\partial U_i}{\partial \mathbf{r}_{ij}}$ the **partial force**.

4.2 The Lennard-Jones potential

The Lennard-Jones potential is one the most simplest two-body potentials used in MD simulations. The pair potential between particle i and j is

$$U_{ij} = 4\epsilon_{ij} \left(\frac{\sigma_{ij}^{12}}{r_{ij}^{12}} - \frac{\sigma_{ij}^6}{r_{ij}^6} \right). \quad (4.6)$$

The parameters ϵ_{ij} are in units of eV and the parameters σ_{ij} are in units of Å. The subscripts in the parameters are used to denote atom types. For example, ϵ_{01} is a parameter applying to a pair of atoms which are of types 0 and 1. The current version of GPUMD supports up to five atom types.

If there is only one atom type, the potential file for this potential model reads

```
lj1
epsilon sigma cutoff
```

Here, `cutoff` is the cutoff distance.

If there are two atom types, the potential file reads

```
lj2
epsilon_00 sigma_00 cutoff_00
epsilon_01 sigma_01 cutoff_01
epsilon_10 sigma_10 cutoff_10
epsilon_11 sigma_11 cutoff_11
```

If there are three atom types, the potential file reads

```
lj3
epsilon_00 sigma_00 cutoff_00
epsilon_01 sigma_01 cutoff_01
epsilon_02 sigma_02 cutoff_02
epsilon_10 sigma_10 cutoff_10
epsilon_11 sigma_11 cutoff_11
epsilon_12 sigma_12 cutoff_12
epsilon_20 sigma_20 cutoff_20
epsilon_21 sigma_21 cutoff_21
epsilon_22 sigma_22 cutoff_22
```

I hope the reader can understand the pattern and figure out how to prepare potential files with four and five atom types.

4.3 The rigid-ion potential

By rigid-ion potential, we mean a potential model consisting of a short range part in the Buckingham form

$$U_{ij} = A_{ij} \exp(-b_{ij}r_{ij}) - \frac{C_{ij}}{r_{ij}^6} \quad (4.7)$$

and a Coulomb potential. The Coulomb potential is evaluated using the damped-shifted-force (DSF) method by Fennell and Gezelter [16], which is based on the Wolf summation method [43]. The DSF version of the pairwise Coulomb potential can be written as:

$$U_{ij} = \frac{q_i q_j}{4\pi\epsilon_0} \left[\frac{\text{erfc}(\alpha r_{ij})}{r_{ij}} - \frac{\text{erfc}(\alpha R_c)}{R_c} + \left(\frac{\text{erfc}(\alpha R_c)}{R_c^2} + \frac{2\alpha}{\sqrt{\pi}} \frac{\exp(-\alpha^2 R_c^2)}{R_c} \right) (r_{ij} - R_c) \right], \quad (4.8)$$

where α and R_c are the electrostatic damping factor and the cutoff radius, respectively. A good choice is $\alpha = 0.2$ and $R_c \geq 15$ Å. In GPUMD, we fix α to 0.2 and let the users to specify the cutoff distance.

Currently, the potential only applies to systems with two atom types. The potential file for this potential model reads

```
ri
q_1    q_2    cutoff
A_11, b_11 C_11
A_22, b_22 C_22
A_12, b_12 C_12
```

4.4 The EAM potential

This is a simple many-body potential. The site potential energy is

$$U_i = \frac{1}{2} \sum_{j \neq i} \phi(r_{ij}) + F(\rho_i). \quad (4.9)$$

Here, the part with $\phi(r_{ij})$ is a pairwise potential and $F(\rho_i)$ is the embedding potential, which depends on the electron density ρ_i at site i . This density is contributed by the neighbors of i :

$$\rho_i = \sum_{j \neq i} f(r_{ij}). \quad (4.10)$$

The many-body part of the EAM potential comes from the embedding potential. The **partial force** in our formulation is

$$\frac{\partial U_i}{\partial \mathbf{r}_{ij}} = \frac{1}{2} \phi'(r_{ij}) \frac{\partial r_{ij}}{\partial \mathbf{r}_{ij}} + F'(\rho_i) f'(r_{ij}) \frac{\partial r_{ij}}{\partial \mathbf{r}_{ij}}. \quad (4.11)$$

All the three functions above, $\phi(r_{ij})$, $F(\rho_i)$, and $f(r_{ij})$ are usually given as splines, **but in the current version of GPUMD, we have only implemented two analytical forms.**

4.4.1 The analytical form by Zhou *et al.*

The pair potential between two atoms of the same type a is

$$\phi^{aa}(r) = \frac{A^a \exp[-\alpha(r/r_e^a - 1)]}{1 + (r/r_e^a - \kappa^a)^{20}} - \frac{B^a \exp[-\beta(r/r_e^a - 1)]}{1 + (r/r_e^a - \lambda^a)^{20}}. \quad (4.12)$$

The contribution of the electron density from an atom of type a is

$$f^a(r) = \frac{f_e^a \exp[-\beta(r/r_e^a - 1)]}{1 + (r/r_e^a - \lambda^a)^{20}}. \quad (4.13)$$

The pair potential between two atoms of different types a and b is then constructed as

$$\phi^{ab}(r) = \frac{1}{2} \left[\frac{f^b(r)}{f^a(r)} \phi^{aa}(r) + \frac{f^a(r)}{f^b(r)} \phi^{bb}(r) \right]. \quad (4.14)$$

The embedding energy function is piecewise:

$$F(\rho) = \sum_{i=0}^3 F_{ni} \left(\frac{\rho}{\rho_n} - 1 \right)^i, \quad (\rho < 0.85\rho_e) \quad (4.15)$$

$$F(\rho) = \sum_{i=0}^3 F_i \left(\frac{\rho}{\rho_e} - 1 \right)^i, \quad (0.85\rho_e \leq \rho < 1.15\rho_e) \quad (4.16)$$

$$F(\rho) = F_e \left[1 - \ln \left(\frac{\rho}{\rho_s} \right)^\eta \right] \left(\frac{\rho}{\rho_s} \right)^\eta, \quad (\rho \geq 1.15\rho_e) \quad (4.17)$$

For each element, there are 20 parameters, which are r_e , f_e , ρ_e , ρ_s , α , β , A , B , κ , λ , F_{n0} , F_{n1} , F_{n2} , F_{n3} , F_0 , F_1 , F_2 , F_3 , η , and F_e . Parameter values for 16 metals are tabulated in the paper by Zhou *et al.* [47]. **Although we have presented the potential in a form applicable to systems with more than one atom type, the current version of GPUMD only supports a single atom type for this potential model. Extensions will be made in a future version.**

The potential file for this potential model reads

```
eam_zhou_2004
r_e
f_e
rho_e
rho_s
alpha
beta
A
B
kappa
lambda
F_n0
F_n1
F_n2
F_n3
F_0
```

F_1
F_2
F_3
eta
F_e
cutoff

The last parameter `cutoff` is the cutoff distance which is not intrinsic to the model. The order of the parameters is the same as in Table III of Ref. [47].

4.4.2 The extended Finnis-Sinclair potential by Dai *et al.*

This is a very simple EAM-type potential which is an extension of the Finnis-Sinclair potential. The function for the pair potential is

$$\phi(r) = \begin{cases} (r - c)^2 \sum_{n=0}^4 c_n r^n & r \leq c \\ 0 & r > c \end{cases} \quad (4.18)$$

The function for the density is

$$\phi(r) = \begin{cases} (r - d)^2 + B^2(r - d)^4 & r \leq d \\ 0 & r > d \end{cases} \quad (4.19)$$

The function for the embedding energy is

$$F(\rho) = -A\rho^{1/2}. \quad (4.20)$$

For each element, there are 9 parameters, which are A , d , c , c_0 , c_1 , c_2 , c_3 , c_4 , and B . The original Finnis-Sinclair potential corresponds to the case where the parameters c_3 , c_4 , and B are all zero. Parameter values for 12 metals can be found from the paper by Dai *et al.* [6].

The potential file for this potential model reads

```

eam_dai_2006
A
d
c
c_0
c_1
c_2
c_3
c_4
B

```

4.4.3 Spline-based EAM potentials

To be written and implemented.

4.4.4 The ADP potential

To be written and implemented.

4.5 The Stillinger-Weber potential

Here we consider the original Stillinger-Weber potential as proposed by Stillinger and Weber in 1985 [37]. The total potential energy consists of a two-body part and a three-body part. The site potential is

$$U_i = \frac{1}{2}V_2(r_{ij}) + \frac{1}{2}\sum_{j \neq i} \sum_{k \neq i,j} h_{ijk}, \quad (4.21)$$

where the two-body part is

$$V_2(r_{ij}) = A_{ij}\epsilon_{ij} \left[B_{ij} \left(\frac{\sigma_{ij}}{r_{ij}} \right)^4 - 1 \right] \exp \left(\frac{1}{r_{ij}/\sigma_{ij} - a_{ij}} \right) \quad (4.22)$$

and the three-body part is

$$h_{ijk} = \epsilon_{ij}\lambda_{ijk} \exp \left[\frac{\gamma_{ij}}{r_{ij}/\sigma_{ij} - a_{ij}} + \frac{\gamma_{ik}}{r_{ik}/\sigma_{ik} - a_{ik}} \right] (\cos \theta_{ijk} - \cos \theta_0)^2. \quad (4.23)$$

Here, A_{ij} , B_{ij} , ϵ_{ij} , σ_{ij} , a_{ij} , λ_{ijk} , γ_{ij} , and $\cos \theta_0$ are parameters. Here is an explicit expression of the **partial force** for the three-body part:

$$\begin{aligned} \frac{\partial U_i}{\partial \mathbf{r}_{ij}} &= \epsilon_{ij}\lambda_{ijk} \exp \left[\frac{\gamma_{ij}}{r_{ij}/\sigma_{ij} - a_{ij}} + \frac{\gamma_{ik}}{r_{ik}/\sigma_{ik} - a_{ik}} \right] (\cos \theta_{ijk} - \cos \theta_0) \\ &\times \left[2 \frac{\partial \cos \theta_{ijk}}{\partial \mathbf{r}_{ij}} - \frac{\gamma_{ij}}{\sigma_{ij} \left(\frac{r_{ij}}{\sigma_{ij}} - a_{ij} \right)^2} (\cos \theta_{ijk} - \cos \theta_0) \frac{\partial r_{ij}}{\partial \mathbf{r}_{ij}} \right]. \end{aligned} \quad (4.24)$$

Parameters with two indices depend on a pair of atoms i and j ; parameters with three indices depend on a triplet of atoms i , j , and k , where i is the central atom of the triplet. The current version of GPUMD supports up to three atom types. For one-element systems, all the parameters have a single value and the potential file for this potential model reads

```
sw_1985
epsilon lambda A B a gamma sigma cos0
```

For two-element systems, the two-body parameters can generally have 3 different values depending on the types of the i and j atoms and the three-body parameters can generally have 8 different values depending on the types of the i , j , and k atoms. The potential file reads

```
sw_1985_2
A_00 B_00 a_00 sigma_00 gamma_00
A_01 B_01 a_01 sigma_01 gamma_01
A_11 B_11 a_11 sigma_11 gamma_11
lambda_000 cos0_000
lambda_001 cos0_001
lambda_010 cos0_010
lambda_011 cos0_011
lambda_100 cos0_100
lambda_101 cos0_101
lambda_110 cos0_110
lambda_111 cos0_111
```

Here, 0 and 1 refer to the atom types defined in the `xyz.in` file (see the next chapter). The numbers in the two-body parameters refer to the types of i and j atoms and those in the three-body parameters correspond to the types of the i , j , and k atoms. Note that we have removed the redundant parameter ϵ_{ij} in the potential file here and A_{ij} and λ_{ijk} actually stand for $\epsilon_{ij}A_{ij}$ and $\epsilon_{ij}\lambda_{ijk}$. The user should be aware of this when preparing the potential file.

If there are three atom types, the potential file reads:

```
sw_1985_3
A_00 B_00 a_00 sigma_00 gamma_00
A_01 B_01 a_01 sigma_01 gamma_01
A_02 B_02 a_02 sigma_02 gamma_02
A_10 B_10 a_10 sigma_10 gamma_10
A_11 B_11 a_11 sigma_11 gamma_11
A_12 B_12 a_12 sigma_12 gamma_12
A_20 B_20 a_20 sigma_20 gamma_20
A_21 B_21 a_21 sigma_21 gamma_21
A_22 B_22 a_22 sigma_22 gamma_22
lambda_000 cos0_000
lambda_001 cos0_001
lambda_002 cos0_002
lambda_010 cos0_010
lambda_011 cos0_011
lambda_012 cos0_012
lambda_020 cos0_020
lambda_021 cos0_021
lambda_022 cos0_022
lambda_100 cos0_100
lambda_101 cos0_101
lambda_102 cos0_102
lambda_110 cos0_110
lambda_111 cos0_111
lambda_112 cos0_112
lambda_120 cos0_120
lambda_121 cos0_121
lambda_122 cos0_122
lambda_200 cos0_200
lambda_201 cos0_201
lambda_202 cos0_202
lambda_210 cos0_210
lambda_211 cos0_211
lambda_212 cos0_212
lambda_220 cos0_220
lambda_221 cos0_221
lambda_222 cos0_222
```

Do we need to consider more than three atom types? I don't think so. So stop here.

4.6 The Vashishta potential

The Vashishta potential [42] is essentially a pairwise potential plus a modified form of the three-body part of the SW potential. Therefore, the site potential can be written in the same form as the SW potential:

$$U_i = \frac{1}{2}V_2(r_{ij}) + \frac{1}{2} \sum_{j \neq i} \sum_{k \neq i, j} h_{ijk}. \quad (4.25)$$

The two-body part reads

$$V_2(r_{ij}) = \frac{H_{ij}}{r_{ij}^{\eta_{ij}}} + \frac{1}{4\pi\epsilon_0} \frac{q_i q_j}{r_{ij}} e^{-r_{ij}/\lambda_{ij}} - \frac{1}{4\pi\epsilon_0} \frac{D_{ij}}{2r_{ij}^4} e^{-r_{ij}/\xi_{ij}} - \frac{W_{ij}}{r_{ij}^6}. \quad (4.26)$$

Here, H_{ij} , η_{ij} , q_i , q_j , λ_{ij} , D_{ij} , ξ_{ij} , and W_{ij} are material-specific parameters. The four terms on the right hand side of the above equation correspond to steric size effects, charge-charge interactions, charge-dipole interactions, and dipole-dipole interactions, respectively. The original paper [42] has used Gauss units for the middle two terms and we have used the SI units. The two-body part is shifted in terms of both potential and force:

$$V_2^{\text{shifted}}(r_{ij}) = V_2(r_{ij}) - V_2(r_c) - (r - r_c) \left. \frac{dV_2(r_{ij})}{dr_{ij}} \right|_{r=r_c}. \quad (4.27)$$

Therefore, both the potential and the force for the two-body part are continuous at the cutoff distance r_c .

The three-body part is

$$h_{ijk} = B_i \exp \left[\frac{\gamma}{r_{ij} - r_0} + \frac{\gamma}{r_{ik} - r_0} \right] \frac{(\cos \theta_{ijk} - \cos \theta_{0i})^2}{1 + C (\cos \theta_{ijk} - \cos \theta_{0i})^2}. \quad (4.28)$$

Here, B_i is a parameter of dimension [energy], γ is a redundant parameter which is chosen to be 1 Å, r_0 is the cutoff distance of the three-body part, C is a dimensionless parameter, and $\cos \theta_{0i}$ is the equilibrium value of $\cos \theta_{ijk}$. Only the triples with $j = k \neq i$ are considered. Here is an explicit expression of the **partial force** for the three-body part:

$$\begin{aligned} \frac{\partial U_i}{\partial \mathbf{r}_{ij}} = & B_i \exp \left[\frac{\gamma}{r_{ij} - r_0} + \frac{\gamma}{r_{ik} - r_0} \right] \frac{(\cos \theta_{ijk} - \cos \theta_{0i})}{1 + C (\cos \theta_{ijk} - \cos \theta_{0i})^2} \\ & \times \left[2 \frac{\partial \cos \theta_{ijk}}{\partial \mathbf{r}_{ij}} \frac{1}{1 + C (\cos \theta_{ijk} - \cos \theta_{0i})^2} - \frac{\gamma}{(r_{ij} - r_0)^2} (\cos \theta_{ijk} - \cos \theta_{0i}) \frac{\partial r_{ij}}{\partial \mathbf{r}_{ij}} \right]. \end{aligned} \quad (4.29)$$

The potential file for this potential model reads

```
vashishta
B_0  B_1  cos0_0  cos0_1  C  r0  rc
H_00 eta_00 q0*q0 lambda_00 D_00 xi_00 W_00
H_01 eta_01 q0*q1 lambda_01 D_01 xi_01 W_01
H_11 eta_11 q1*q1 lambda_11 D_11 xi_11 W_11
```

Note:

- The charge q_i was written as Z_i in the original reference [42];
- The parameter γ is always 1Å and is thus not treated as an input parameter;
- The parameters η_{ij} are always integers (as far as I know) and should be entered as integers;
- One has to define the atoms types (0 and 1) according to the conventions here when preparing the `xyz.in` file to be introduced in Chapter 5. The order is very important.

4.7 The Tersoff potential

There are many variants of the Tersoff potential. In the current version of GPUMD, we only consider the form as described in Ref. [39] published in 1989.

The site potential can be written as

$$U_i = \frac{1}{2} \sum_{j \neq i} f_C(r_{ij}) [f_R(r_{ij}) - b_{ij} f_A(r_{ij})]. \quad (4.30)$$

The function f_C is a cutoff function, which is 1 when $r_{ij} < R_{ij}$ and 0 when $r_{ij} > S_{ij}$ and takes the following form in the intermediate region:

$$f_C(r_{ij}) = \frac{1}{2} \left[1 + \cos \left(\pi \frac{r_{ij} - R_{ij}}{S_{ij} - R_{ij}} \right) \right]. \quad (4.31)$$

The repulsive function f_R and the attractive function f_A take the following forms:

$$f_R(r) = A_{ij} e^{-\lambda_{ij} r_{ij}}; \quad (4.32)$$

$$f_A(r) = B_{ij} e^{-\mu_{ij} r_{ij}}. \quad (4.33)$$

The bond-order is

$$b_{ij} = \chi_{ij} \left(1 + \beta_i^{n_i} \zeta_{ij}^{n_i} \right)^{-\frac{1}{2n_i}}, \quad (4.34)$$

where

$$\zeta_{ij} = \sum_{k \neq i, j} f_C(r_{ik}) g_{ijk}, \quad (4.35)$$

and

$$g_{ijk} = 1 + \frac{c_i^2}{d_i^2} - \frac{c_i^2}{d_i^2 + (h_i - \cos \theta_{ijk})^2}. \quad (4.36)$$

Here, A_{ij} , B_{ij} , λ_{ij} , μ_{ij} , β_i , n_i , c_i , d_i , h_i , R_{ij} , S_{ij} , and χ_{ij} are material-specific parameters. The parameter values can be found in Tersoff's original paper [39] and a later erratum [40]. Note that when $i \neq j$, the following mixing rules are used to determining some parameter values:

$$A_{ij} = \sqrt{A_{ii} A_{jj}}; \quad (4.37)$$

$$B_{ij} = \sqrt{B_{ii} B_{jj}}; \quad (4.38)$$

$$R_{ij} = \sqrt{R_{ii} R_{jj}}; \quad (4.39)$$

$$S_{ij} = \sqrt{S_{ii} S_{jj}}; \quad (4.40)$$

$$\lambda_{ij} = (\lambda_{ii} + \lambda_{jj})/2; \quad (4.41)$$

$$\mu_{ij} = (\mu_{ii} + \mu_{jj})/2. \quad (4.42)$$

The parameter χ_{ij} is 1 for $i = j$ and can deviate slightly from 1 for $i \neq j$.

In the current version of GPUMD, one can simulate systems with two atom types (such as SiC and SiGe) as well as those with a single atom type.

An expression of the **partial force** has been presented in Ref. [14] and we repeat it here:

$$\begin{aligned} \frac{\partial U_i}{\partial \mathbf{r}_{ij}} = & \frac{1}{2} f'_C(r_{ij}) [f_R(r_{ij}) - b_{ij} f_A(r_{ij})] \frac{\partial r_{ij}}{\partial \mathbf{r}_{ij}} \\ & + \frac{1}{2} f'_C(r_{ij}) [f'_R(r_{ij}) - b_{ij} f'_A(r_{ij})] \frac{\partial r_{ij}}{\partial \mathbf{r}_{ij}} \\ & - \frac{1}{2} f'_C(r_{ij}) \sum_{k \neq i, j} f_C(r_{ik}) f_A(r_{ik}) b'_{ik} g_{ijk} \frac{\partial r_{ij}}{\partial \mathbf{r}_{ij}} \\ & - \frac{1}{2} f'_C(r_{ij}) f_A(r_{ij}) b'_{ij} \sum_{k \neq i, j} f_C(r_{ik}) g'_{ijk} \frac{\partial \cos \theta_{ijk}}{\partial \mathbf{r}_{ij}} \\ & - \frac{1}{2} f'_C(r_{ij}) \sum_{k \neq i, j} f_C(r_{ik}) f_A(r_{ik}) b'_{ik} g'_{ijk} \frac{\partial \cos \theta_{ijk}}{\partial \mathbf{r}_{ij}}. \end{aligned} \quad (4.43)$$

The potential file for the single-element version reads

```
tersoff_1989_1
A B lambda mu beta n c d h R S
```

The potential file for the double-element version reads

```
tersoff_1989_2
A_1 B_1 lambda_1 mu_1 beta_1 n_1 c_1 d_1 h_1 R_1 S_1
A_2 B_2 lambda_2 mu_2 beta_2 n_2 c_2 d_2 h_2 R_2 S_2
chi
```

4.8 The modified Tersoff potential for silicon systems

Note: This potential has not been implemented in GPUMD yet, but is in my **TODO** list.

A modified Tersoff potential for silicon systems has been proposed by Kumagai *et al.* [24] and further extended by Pun *et al.* [33].

The site potential takes a similar form as in the case of the Tersoff potential:

$$U_i = \frac{1}{2} \sum_{j \neq i} f_C(r_{ij}) [f_R(r_{ij}) - b_{ij} f_A(r_{ij}) + c_0]. \quad (4.44)$$

Here, c_0 is a constant. The repulsive function $f_R(r)$ and the attractive function $f_A(r)$ are identical to those in the Tersoff potential:

$$f_R(r) = A e^{-\lambda r}; \quad (4.45)$$

$$f_A(r) = B e^{-\mu r}. \quad (4.46)$$

The bond-order reads

$$b_{ij} = (1 + \zeta_{ij}^\eta)^{-\delta}, \quad (4.47)$$

where (the function $f_C(r_{ik})$ in the following equation was mistakenly written as $f_C(r_{ij})$ in both references [24, 33])

$$\zeta_{ij} = \sum_{k \neq i,j} f_C(r_{ik}) g_{ijk} e^{\alpha(r_{ij}-r_{ik})^m}, \quad (4.48)$$

and

$$g_{ijk} = c_1 + \frac{c_2(h - \cos \theta)^2}{c_3 + (h - \cos \theta)^2} \left[1 + c_4 e^{-c_5(h - \cos \theta)^2} \right]. \quad (4.49)$$

They also adopted the modified cutoff function proposed by Murty and Atwater [34], which has a continuous second-order derivative:

$$f_C(r) = \frac{1}{2} + \frac{9}{16} \cos \left(\pi \frac{r_{ij} - R}{S - R} \right) - \frac{1}{16} \cos \left(3\pi \frac{r_{ij} - R}{S - R} \right). \quad (4.50)$$

There are 17 material-specific parameters: $A, B, \lambda, \mu, \eta, \delta, \alpha, m, c_0, c_1, c_2, c_3, c_4, c_5, h, R, S$.

The partial force can be derived to be

$$\begin{aligned} \frac{\partial U_i}{\partial \mathbf{r}_{ij}} = & \frac{1}{2} f'_C(r_{ij}) [f_R(r_{ij}) - b_{ij} f_A(r_{ij}) + c_0] \frac{\partial r_{ij}}{\partial \mathbf{r}_{ij}} \\ & + \frac{1}{2} f_C(r_{ij}) [f'_R(r_{ij}) - b_{ij} f'_A(r_{ij})] \frac{\partial r_{ij}}{\partial \mathbf{r}_{ij}} \\ & - \frac{1}{2} f'_C(r_{ij}) \sum_{k \neq i,j} f_C(r_{ik}) f_A(r_{ik}) b'_{ik} g_{ijk} e^{\alpha(r_{ik}-r_{ij})^m} \frac{\partial r_{ij}}{\partial \mathbf{r}_{ij}} \\ & - \frac{1}{2} f_C(r_{ij}) f_A(r_{ij}) b'_{ij} \sum_{k \neq i,j} f_C(r_{ik}) e^{\alpha(r_{ij}-r_{ik})^m} g'_{ijk} \frac{\partial \cos \theta_{ijk}}{\partial \mathbf{r}_{ij}} \\ & - \frac{1}{2} f_C(r_{ij}) f_A(r_{ij}) b'_{ij} \sum_{k \neq i,j} f_C(r_{ik}) g_{ijk} \alpha m (r_{ij} - r_{ik})^{m-1} e^{\alpha(r_{ij}-r_{ik})^m} \frac{\partial r_{ij}}{\partial \mathbf{r}_{ij}} \\ & - \frac{1}{2} f_C(r_{ij}) \sum_{k \neq i,j} f_C(r_{ik}) f_A(r_{ik}) b'_{ik} e^{\alpha(r_{ik}-r_{ij})^m} g'_{ijk} \frac{\partial \cos \theta_{ijk}}{\partial \mathbf{r}_{ij}} \\ & - \frac{1}{2} f_C(r_{ij}) \sum_{k \neq i,j} f_C(r_{ik}) f_A(r_{ik}) b'_{ik} (-\alpha) m (r_{ik} - r_{ij})^{m-1} e^{\alpha(r_{ik}-r_{ij})^m} g_{ijk} \frac{\partial r_{ij}}{\partial \mathbf{r}_{ij}}. \end{aligned} \quad (4.51)$$

4.9 The REBO potential for Mo-S systems

Here, we discuss the REBO (reactive empirical bond-order) potential for Mo-S systems developed by Liang *et al.* [25]. The erratum [26] should be consulted for improved and extra model parameters which are missing from the original article. This REBO potential is just slightly more complicated than the Tersoff potential. Note that there is also a long-range Lennard-Jones part.

The site potential takes the same general form as in the case of the Tersoff potential:

$$U_i = \frac{1}{2} \sum_{j \neq i} f_C(r_{ij}) [f_R(r_{ij}) - b_{ij} f_A(r_{ij})]. \quad (4.52)$$

The cutoff function $f_C(r)$ and the attractive function $f_A(r)$ are also identical to those in the Tersoff potential. The repulsive function $f_R(r)$ takes a different form than that in the Tersoff potential:

$$f_R(r) = \left(1 + \frac{Q}{r}\right) A e^{-\alpha r}, \quad (4.53)$$

where Q , A , and α are material-specific parameters.

The bond order function in this REBO potential is

$$b_{ij} = (1 + \zeta_{ij})^{-1/2}, \quad (4.54)$$

where

$$\zeta_{ij} = \left[\sum_{k \neq i,j} f_C(r_{ik}) g_{ijk}(\cos \theta_{ijk}) \right] + P_{ij}(N_{ij}); \quad (4.55)$$

$$N_{ij} = \sum_{k \neq i,j} f_C(r_{ik}); \quad (4.56)$$

$$P(x) = -a_0(x-1) - a_1 e^{-a_2 x} + a_3. \quad (4.57)$$

The angular function is given by

$$g(x) = G(x) + \psi(x)[\gamma(x) - G(x)], \quad (4.58)$$

where both $G(x)$ and $\gamma(x)$ are 6th order polynomials and

$$\psi(x) = 0 \text{ when } x < 0.5; \quad (4.59)$$

$$\psi(x) = \frac{1}{2} [1 - \cos(2\pi(x - 0.5))] \text{ when } x \geq 0.5. \quad (4.60)$$

The partial force can be derived to be

$$\begin{aligned} \frac{\partial U_i}{\partial \mathbf{r}_{ij}} = & \frac{1}{2} f'_C(r_{ij}) [f_R(r_{ij}) - b_{ij} f_A(r_{ij})] \frac{\partial r_{ij}}{\partial \mathbf{r}_{ij}} \\ & + \frac{1}{2} f'_C(r_{ij}) [f'_R(r_{ij}) - b_{ij} f'_A(r_{ij})] \frac{\partial r_{ij}}{\partial \mathbf{r}_{ij}} \\ & - \frac{1}{2} f'_C(r_{ij}) \sum_{k \neq i,j} f_C(r_{ik}) f_A(r_{ik}) b'_{ik} g_{ijk} \frac{\partial r_{ij}}{\partial \mathbf{r}_{ij}} \\ & - \frac{1}{2} f'_C(r_{ij}) \sum_{k \neq i,j} f_C(r_{ik}) f_A(r_{ik}) b'_{ik} P'_{ik} \frac{\partial r_{ij}}{\partial \mathbf{r}_{ij}} \\ & - \frac{1}{2} f'_C(r_{ij}) f_A(r_{ij}) b'_{ij} \sum_{k \neq i,j} f_C(r_{ik}) g'_{ijk} \frac{\partial \cos \theta_{ijk}}{\partial \mathbf{r}_{ij}} \\ & - \frac{1}{2} f'_C(r_{ij}) \sum_{k \neq i,j} f_C(r_{ik}) f_A(r_{ik}) b'_{ik} g'_{ijk} \frac{\partial \cos \theta_{ijk}}{\partial \mathbf{r}_{ij}}. \end{aligned} \quad (4.61)$$

Here, b'_{ij} should be understood as $\partial b_{ij} / \partial \zeta_{ij}$.

This potential was added to GPUMD in version 1.3. When preparing the xyz.in file to be introduced in Chapter 5, one has to define Mo atoms as type 0 and S atoms as type 1. The potential file for this potential model reads

```
rebo_mos2
# Nothing here, because this is a special potential model and
# the parameters are thus hard coded
```

4.10 The EDIP potential for silicon

Note: This potential has not been implemented in GPUMD yet, but is in my **TODO** list.

We next consider the EDIP potential for silicon [1, 22]. The total potential energy consists of a two-body part and a three-body part. Site potential

$$U_i = U_i^{(2)} + U_i^{(3)}; \quad (4.62)$$

$$U_i^{(2)} = \frac{1}{2} \sum_{j \neq i} V_{ij}^{(2)}(r_{ij}, Z_i); \quad (4.63)$$

$$U_i^{(3)} = \frac{1}{2} \sum_{j \neq i} \sum_{k \neq i, j} V_{ijk}^{(3)}(\mathbf{r}_{ij}, \mathbf{r}_{ik}, Z_i). \quad (4.64)$$

The effective coordination number Z_i is defined as

$$Z_i = \sum_{k \neq i} f(r_{ik}). \quad (4.65)$$

where $f(r)$ is a cutoff function which is 1 when $r < c$ and 0 when $r > a$, and takes the value in between:

$$f(r) = \exp\left(\frac{\alpha}{1 - x^{-3}}\right); \quad x = \frac{r - c}{a - c}. \quad (4.66)$$

The two-body part is

$$V_{ij}^{(2)}(r_{ij}, Z_i) = A \left[\left(\frac{B}{r_{ij}} \right)^\rho - e^{-\beta Z_i^2} \right] \exp\left(\frac{\sigma}{r_{ij} - a}\right) \quad (4.67)$$

and the three-body part is

$$V_{ijk}^{(3)}(\mathbf{r}_{ij}, \mathbf{r}_{ik}, Z_i) = \exp\left(\frac{\gamma}{r_{ij} - a}\right) \exp\left(\frac{\gamma}{r_{ik} - a}\right) h(\cos \theta_{ijk}, Z_i). \quad (4.68)$$

The angular function is

$$h(x, Z) = \lambda \left[1 - e^{-Q_0 e^{-\mu Z} (x + \tau(Z))^2} + \eta Q_0 e^{-\mu Z} (x + \tau(Z))^2 \right], \quad (4.69)$$

where

$$\tau(Z) = u_1 + u_2(u_3 e^{-u_4 Z} - e^{-2u_4 Z}). \quad (4.70)$$

Both $U^{(2)}$ and $U^{(3)}$ contain three-body effects. The partial force for the $U^{(2)}$ part is

$$\begin{aligned} \frac{\partial U_i^{(2)}}{\partial \mathbf{r}_{ij}} &= \frac{1}{2} \frac{\partial V_{ij}^{(2)}}{\partial r_{ij}} \frac{\partial r_{ij}}{\partial \mathbf{r}_{ij}} \\ &+ \frac{1}{2} \frac{\partial V_{ij}^{(2)}}{\partial Z_i} f'(r_{ij}) \frac{\partial r_{ij}}{\partial \mathbf{r}_{ij}} \\ &+ \frac{1}{2} \sum_{k \neq i, j} \frac{\partial V_{ik}^{(2)}}{\partial Z_i} f'_{ij}(r_{ij}) \frac{\partial r_{ij}}{\partial \mathbf{r}_{ij}}. \end{aligned} \quad (4.71)$$

The partial force for the $U^{(3)}$ part is

$$\begin{aligned} \frac{\partial U_i^{(3)}}{\partial \mathbf{r}_{ij}} = & \sum_{k \neq i,j} \left[\left(\frac{\partial V_{ijk}^{(3)}}{\partial r_{ij}} + \frac{\partial V_{ijk}^{(3)}}{\partial Z_i} f'(r_{ij}) \right) \frac{\partial r_{ij}}{\partial \mathbf{r}_{ij}} + \frac{\partial V_{ijk}^{(3)}}{\partial \cos \theta_{ijk}} \frac{\partial \cos \theta_{ijk}}{\partial \mathbf{r}_{ij}} \right] \\ & + \frac{1}{2} \sum_{k \neq i,j} \sum_{m \neq i,j} \frac{\partial V_{ikm}^{(3)}}{\partial Z_i} f'(r_{ij}) \frac{\partial r_{ij}}{\partial \mathbf{r}_{ij}}. \end{aligned} \quad (4.72)$$

4.11 Mixed potentials

This is explained in the next chapter when we discuss the `potentials` keyword in the `run.in` file.

Chapter 5

Using GPUMD

After downloading and unpacking GPUMD, one can see the following folders:

- `src`, which contains all the source files of GPUMD and a `makefile`
- `potentials`, which contains all the potential files we have prepared
- `examples`, which contains all the examples (each in a sub-folder) we have prepared
- `doc`, which contains the source files for the manual

5.1 Compile and run GPUMD

5.1.1 Compile GPUMD

To compile GPUMD, one just needs to go to the `src` directory and type `make`. One may want to first do `make clean`. When the compilation finishes, an executable named `gpumd` will be generated in the `src` directory.

In the `makefile`, the default compiling flag is

```
CFLAGS = -std=c++11 -O3 -arch=sm_35
```

This means that

- The C++ part uses the C++11 standard.
- The C++ part has an optimization level of 3.
- The CUDA part is compiled with a target compute capability of 3.5. **Note that GPUs with compute capability less than 3.5 are not supported by GPUMD.**

One can specify some compiling flags in the `makefile` in this way:

```
CFLAGS = -std=c++11 -O3 -arch=sm_35 -DFLAG
```

Here are the possible `-DFLAGs`:

- `-DUSE_SP`
When this is specified, the code will be compiled as a single-precision version. We do not recommend to use this version, but the users can try.

- **-DDEBUG**

When this is specified, the code will use the same seed to initialize the random number generator for initializing the velocities. This will make sure to get exactly the same output for the same input. However, due to some limitations of the current implementation of GPUMD, the code will always use the $O(N^2)$ version to generate the neighbor list, which is not efficient for large systems.

5.1.2 Run simulations with GPUMD

To run a simulation or a set of simulations with GPUMD, one needs to first prepare some input files. For any individual simulation, one needs to prepare a file named `xyz.in` and a file named `run.in`, and put them into the same directory. Then, one just needs an extra input file, which we call a “driver input file”, to specify the path(s) of the folder(s) containing the input files. This “driver input file” should have the following format:

```
number_of_simulations
path_1
path_2
...
```

Let us consider a few explicit examples. Consider a “driver input file” which reads

```
1
examples/ex1
```

This means that there will be one simulation and the input files (`xyz.in` and `run.in`) are prepared in the folder `examples/ex1`. One can also run multiple simulations using a single “driver input file”. An example is:

```
4
examples/ex1
examples/ex2
examples/ex3
examples/ex4
```

In this case, it means that four sets of inputs will be processed consecutively. If you want to do 10 independent calculations for the `examples/ex3` input folder, the “driver input file” can be:

```
10
examples/ex3
examples/ex3
examples/ex3
examples/ex3
examples/ex3
examples/ex3
examples/ex3
examples/ex3
examples/ex3
examples/ex3
```

Now we are ready to run the code. Suppose that the “driver input file” is named as `input` and is in the folder where we can see the `src` folder, we can run the code using the following command:

```
src/gpumd < input
```

Output files will be created in the folders containing the corresponding input files.

The next two sections are devoted to describing the structures of the input and output files, respectively.

5.2 Input files of GPUMD

5.2.1 The xyz.in input file

A file named `xyz.in` should be prepared and should have the following format (empty lines and comments are not allowed):

```
N M cutoff has_velocity has_layer number_of_grouping_methods
pbc_x pbc_y pbc_z L_x L_y L_z
t1 x1 y1 z1 m1 [vx1] [vy1] [vz1] [layer1] [group1_1] [group2_1] ...
t2 x2 y2 z2 m2 [vx2] [vy2] [vz2] [layer1] [group1_2] [group2_2] ...
...
```

We explain line by line:

- In the first line,
 - `N` is the number of atoms.
 - `M` is the maximum possible number of neighbor atoms for one atom. **Warning!** It's the responsibility of the user to use a large enough value of `M`. If some atom has more than `M` neighbors during the simulation, unexpected error may occur and there may be no useful error message. If `M` is too large, it just wastes some memory.
 - `cutoff` is the **initial** cutoff distance used for building the neighbor list.
 - `has_velocity` can be 1 or 0, which means this file contains or does not contain the initial velocities, respectively. See the explanations for the third line below for further details.
 - `has_layer` can be 1 or 0, which means this file contains or does not contain the layer index used for some mixed potentials, respectively. See the **potentials** keyword in the `run.in` file. See the explanations for the third line below for further details.
 - `number_of_grouping_methods` can be integers from 0 to 10, which is the number of grouping methods that can be referred to in some keywords in the `run.in` file. See the explanations for the third line below for further details.
- In the second line, `pbc_x`, `pbc_y`, and `pbc_z` can only be 1 or 0. If `pbc_x` is 1, it means that periodic boundary conditions will be applied to the x direction; if `pbc_x` is 0, it means that free boundary conditions will be applied to the x direction. Similar descriptions apply to the other two directions. The next three items in the second line, `L_x`, `L_y`, and `L_z`, are the initial lengths of the (rectangular) simulation box along the x , y , and z directions, respectively.

- In the third line,
 - `t1` is the type of atom 1. The atom type will be used to determine which potential parameters to use. **We use integers to record the atom types and the indices start from 0.**
 - `x1`, `y1`, and `z1` are the coordinates of this atom.
 - `m1` is the mass of this atom.
 - if `has_velocity` is 1, the next 3 items are the initial velocities of this atom.
 - if `has_layer` is 1, the next item is the layer label of this atom. That is, the number specifies which “layer” of the material this atom belongs to. This is only used in some mixed potentials when you use the `potentials` keyword in the `run.in` file. **We use integers to record the layer labels and the numbers can be arbitrary because they are only used to distinguish the different layers.**
 - if `number_of_grouping_methods` is larger than zero, the next items are the group labels of this atom in the `number_of_grouping_methods`-th grouping methods. The group labels will be used for some other keywords in the `run.in` file. That is, `group1_1` specifies which group this atom belongs to using the first grouping method; `group2_1` specifies which group this atom belongs to using the second grouping method... **We use integers to record the group labels and the indices start from 0.**
- Similarly, the $(m+2)$ th line gives the information for the m th atom. This file should have $N+2$ lines.

The mass should be given in units of the unified atomic mass unit (amu). The cutoff distance, box lengths and atom coordinates should be given in units of angstrom (\AA). Velocities can be in arbitrary units because we will scale them according to the initial temperature specified in the `run.in` file.

5.2.2 The `run.in` input file

Then, a file named `run.in` should also be prepared. In this input file, blank lines and lines starting with `#` are ignored. All the other lines should be of the following form:

```
keyword parameter_1 parameter_2 ...
```

The overall structure of a `run.in` file is as follows:

```
#-----
# First, write these two keywords in any order (group-1)
potential (or potentials)
velocity

# Then, write (all or part of) these keywords in any order (group-2)
ensemble
time_step
neighbor
fix
deform
```

```

dump_thermo
dump_position
dump_velocity
dump_force
dump_potential
dump_virial
dump_heat
compute
compute_shc
compute_vac
compute_hac
compute_hnemd

# Then write the keyword run (group-3)
run

# Now one can repeat the last two groups as many times as one wants
#-----

```

5.3 Keywords in the run.in input file

We now describe the use of the keywords in detail.

5.3.1 The potential keyword

This keyword is used to specify a single potential for the system. It only has one parameter, which is the file name (including the absolute or relative path) containing the information of the potential that the user wants to use. An example is

```
potential potentials/tersoff/si_tersoff_1989_1.txt
```

By writing this, one has to make sure that the file `si_tersoff_1989_1.txt` has been prepared in the folder `potentials/tersoff/`. The potential files are discussed in the previous chapter. **It is very important to make sure that the atom types in the xyz.in file are consistent with the potential file.**

5.3.2 The potentials keyword

This keyword is used to specify multiple potentials for the system. The number of parameters can only be 5 or 8.

When it is 5, two potentials need to be specified and the command reads:

```
potentials file1 flag file2 n1 n2
```

Here, `file1` is the filename for a two-body potential acting on the whole system (for all the atom types), `file2` is the filename for a many-body potential acting on atoms from type `n1` to type `n2`. The parameter `flag` can only be set to 0 or 1, where 1 means that the two-body potential will not act on atoms from the same “layer”. In this case, one has to set the parameter `has_layer_in_xyz` to 1 in `xyz.in` and provide the required data there.

When the number of parameters is 8, three potentials need to be specified and the command reads:

```
potentials file1 flag file2 n1 n2 file3 m1 m2
```

The only difference compared to the case of 5 parameters is that here we have an extra many-body potential specified by the file `file3` which acts on atoms from type `m1` to type `m2`.

We give two explicit examples below.

In the first example, let us consider simulating bilayer graphene. There is only one atom type and we can use the Tersoff potential for all the carbon atoms. However, one usually assumes that there is an inter-layer Lennard-Jones potential (van der Waals interaction) which should not act between atoms from the same layer. In this case, one can use the following command to specify two potentials for the system:

```
potentials lj_file 1 tersoff_file 0 0
```

Here, the first parameter `lj_file` specifies the Lennard-Jones potential file which reads

```
lj1
epsilon sigma cutoff
```

The second parameter `1` means that we need to prepare “layer label” data in the `xyz.in` file. Then, atoms with the same “layer label” will not experience the Lennard-Jones potential. The third parameter specifies the Tersoff potential file and the last two parameters mean that the Tersoff potential will act on atoms from type `0` to type `0`.

In the second example, let us consider simulating a single-layer graphene on top of a SiO_2 substrate. There are three atom types and we define carbon atoms as type `0`, silicon atoms as type `1` and oxygen atoms as type `2`. We can then use the following command to specify three potentials for the system:

```
potentials lj_file 0 tersoff_file1 0 0 tersoff_file2 1 2
```

Here, the first parameter `lj_file` specifies the Lennard-Jones potential file which reads

```
lj3
epsilon_00 sigma_00 cutoff_00
epsilon_01 sigma_01 cutoff_01
epsilon_02 sigma_02 cutoff_02
epsilon_10 sigma_10 cutoff_10
epsilon_11 sigma_11 cutoff_11
epsilon_12 sigma_12 cutoff_12
epsilon_20 sigma_20 cutoff_20
epsilon_21 sigma_21 cutoff_21
epsilon_22 sigma_22 cutoff_22
```

If we want to exclude the Lennard-Jones interactions between C-C, Si-Si, Si-O, O-Si, and O-O pairs, we can simply set the corresponding cutoff distance to `0`. Therefore, we do not need to use an extra layer label to achieve this. This explains why we set the second parameter to `0`. The remaining parameters simply mean that we apply a single-element Tersoff potential to atoms from type `0` to type `0` (graphene) and a double-element Tersoff potential to atoms from type `1` to type `2` (SiO_2). **It is very important to make sure that, in the `xyz.in` file, atoms for the first many-body potential all come before (or after) atoms for the second many-body potential.**

5.3.3 The velocity keyword

This keyword only has one parameter, which is the initial temperature of the system. For example, the command

```
velocity 10
```

means that one wants to set the initial temperature to 10 K.

5.3.4 The ensemble keyword

This keyword specifies the ensemble type (including external conditions such as local heating and cooling) and the relevant parameters. The number of parameters depends on the first parameter, which can be:

- **nve**. This corresponds to the *NVE* ensemble and there is no need to further specify any other parameters. Therefore, the full command is

```
ensemble nve
```

- **nvt_ber**. This corresponds to the *NVT* ensemble using the Berendsen method. In this case, one needs to specify an initial target temperature **T_1**, a final target temperature **T_2**, and a parameter **T_coup** which reflects the strength of the coupling between the system and the thermostat. The full command is

```
ensemble nvt_ber T_1 T_2 T_coup
```

The target temperature (not the instant system temperature) will vary linearly from **T_1** to **T_2** during a run.

- **nvt_nhc**. This corresponds to the *NVT* ensemble using the Nosé-Hoover chain method. The full command is

```
ensemble nvt_nhc T_1 T_2 T_coup
```

- **nvt_bdp**. This corresponds to the *NVT* ensemble using the BDP method. The full command is

```
ensemble nvt_bdp T_1 T_2 T_coup
```

- **nvt_lan**. This corresponds to the *NVT* ensemble using the Langevin method. The full command is

```
ensemble nvt_lan T_1 T_2 T_coup
```

- **npt_ber**. This corresponds to the *NPT* ensemble using the Berendsen method. In this case, apart from the same parameters as in the case of **nvt_ber**, one needs to further specify 3 target pressures, **Px**, **Py**, and **Pz**, and a pressure coupling constant **P_coup**. The full command is

```
ensemble npt_ber T_1 T_2 T_coup Px Py Pz P_coup
```

- **heat_nhc.** This corresponds to heating a source region and simultaneously cooling a sink region using local Nosé-Hoover chain thermostats. The full command is

```
ensemble heat_nhc T T_coup delta_T label_source label_sink
```

The target temperatures in the source region (with label `label_source`) and sink region (with label `label_sink`) are $T + \text{delta_T}$ and $T - \text{delta_T}$, respectively. Therefore, the temperature difference between the two regions is twice of `delta_T`.

- **heat_bdp.** This corresponds to heating a source region and simultaneously cooling a sink region using the BDP thermostats. The full command is

```
ensemble heat_bdp T T_coup delta_T label_source label_sink
```

- **heat_lan.** This corresponds to heating a source region and simultaneously cooling a sink region using the Langevin thermostats. The full command is

```
ensemble heat_lan T T_coup delta_T label_source label_sink
```

The units of temperature and pressure for this keyword are K and GPa, respectively. The temperature coupling constant in the Berendsen method can be any positive number less than or equal to 1 and we recommend a value in the range of [0.01, 1]. A larger number results in a faster control of the temperature. The temperature coupling constant (the relaxation time) in the Nosé-Hoover chain, BDP, and Langevin methods is in units of the time step and is recommended to be in the range of [100, 1000]. Here, a larger number results in a slower control of the temperature. The pressure coupling constant in the Berendsen method should be a small positive number in the unit system adopted by GPUMD. We recommend a value in the range of [0.01, 0.0001]. For a stiffer material (such as diamond or graphene), one should use a smaller value. In practice, all these parameters should be determined by try and error.

5.3.5 The time step keyword

This keyword only requires a single parameter, which is the time step for integration in units of fs (10^{-15} s). For example, the command

```
time_step 1
```

means that the time step for the current run is 1 fs. Note that the value of time step does not need to be set for each run in a “run.in” file. **If you do not set a new value of time step in a run, the value in the previous run will be used.**

5.3.6 The neighbor keyword

This keyword only requires a single parameter, which is the skin distance (the difference between the cutoff distance used in neighbor list construction and that used in force evaluation) in units of Å. For example, the command

```
neighbor 1
```

means that the neighbor list will be updated when necessary (the code determines automatically when the neighbor list needs to be updated) and the skin distance is 1 Å. **If this keyword is absent in a run, the neighbor list will not be updated during the run.**

5.3.7 The dump thermo, dump position, dump velocity, dump force, dump potential, and dump virial keywords

These keywords only requires a single parameter, which is the output frequency for the relevant quantities: global thermodynamic quantities for `dump_thermo`, per-atom positions for `dump_position`, per-atom velocities for `dump_velocity`, per-atom forces for `dump_force`, per-atom potential energies for `dump_potential`, and per-atom virial for `dump_virial`. For example, the command

```
dump_thermo 1000
```

means that the thermodynamic quantities will be written into the file `thermo.out` (in the folder which contains the `run.in` file) every 1000 steps. **By default, GPUMD does not dump these quantities.** For example, if there is no `dump_position` command for one run, positions will not be output for that run.

5.3.8 The fix keyword

This keyword requires a single parameter which is the label of the group in which the atoms are to be fixed (velocities and forces are set to zero such that the atoms in the group do not move). For example, command

```
fix 0
```

means that atoms in group 0 will be fixed during the current run.

5.3.9 The deform keyword

This keyword is used to deform the simulation box, which can be used to do tensile tests. It requires 4 parameters:

```
deform strain_rate deform_x deform_y deform_z
```

Here, `strain_rate` specifies the speed of the increase of the box length, which is in units of Å /step. For example, suppose the box length (in a given direction) in the beginning of a run is 100 Å and this parameter is 10^{-5} Å /step, then a run with 10^6 steps will change the box length by 10%. This gives a strain rate of 10^8 s⁻¹ if the time step is 1 fs. The second parameter `deform_x` can be 0 or 1, where 0 means do not deform the *x* direction and 1 means deform the *x* direction. The last two parameters have similar meanings.

5.3.10 The compute keyword

This keyword is used to compute and output space and time averages of various per-atom quantities. It is used in the following way:

```
compute sample_interval output_interval quantity_1 quantity_2 ...
```

Here, the first parameter `sample_interval` means the sampling interval, i.e., sampling the quantities every so many time steps and the second parameter `output_interval` means averaging over so many sampled data before giving one output. For example,

```
compute 100 10 quantity_1
```

means sampling `quantity_1` every 100 time steps and averaging over 10 data points before making an output. That is, there is only one output every $100 \times 10 = 1000$ time steps.

The allowed names for the quantities are:

- `temperature`
- `potential`
- `force`
- `virial`
- `heat_current`

One can write one or more (distinct) names for this keyword.

5.3.11 The `compute shc` keyword

This keyword is used to compute the nonequilibrium heat flux correlation function $K_{A \rightarrow B}(t)$ define in Eq. (3.118) and requires 5 parameters. The first parameter is the sampling interval between two correlation steps. The second parameter is the total correlation steps. The third parameter is the number of steps used to calculate one correlation function. The last two parameters are the labels of the groups A and B as used in Eq. (3.118). For example, the command

```
compute_shc 1 500 5000 10 11
```

means that (1) you want to do this calculation; (2) the relevant data will be sampled every step; (3) the maximum number of correlation steps is 500; (4) the correlation function $K_{A \rightarrow B}(t)$ will be calculated every 5000 steps (If the number of steps in this run is 100000, there will be $100000/5000 = 20$ independent correlation functions calculated); (5) the group label of A is 10 and that of B is 11. The results will be written into a file named `shc.out` in the same folder where you put your `run.in` file in.

5.3.12 The `compute vac` keyword

This keyword is related to the calculations of VAC (velocity autocorrelation) and two other related quantities: RDC (running diffusion coefficient) and DOS (phonon density of states). If this keyword appears in a run, VAC and related quantities will be calculated in the run. This keyword requires 3 parameters. The first parameter for this keyword is the sampling interval of the velocity data. The second parameter is the maximum number of correlation steps. The third parameter is the maximum angular frequency $\omega_{\max} = 2\pi\nu_{\max}$ used in the DOS calculations. For example, the command

```
compute_vac 5 200 350
```

means that (1) you want to calculate the VAC and related quantities; (2) the velocity data will be recorded every 5 steps; (3) the maximum number of correlation steps is 200; (4) the maximum angular frequency you want to consider is $\omega_{\max} = 2\pi\nu_{\max} = 350$ THz. The results will be written into a file named `vac.out` in the same folder where you put your `run.in` file in.

5.3.13 The compute hac keyword

The `compute_hac` keyword is similar to the `compute_vac` keyword. It is used to calculate HAC (heat current autocorrelation) and RTC (running thermal conductivity). It has 3 parameters. The first parameter is the sampling interval for the heat current data. The second parameter is the maximum correlation steps. These two parameters are similar to those for the `compute_vac` keyword. The third parameter for `compute_hac` is the output interval of the HAC and RTC data. For example, the command

```
compute_hac 20 50000 10
```

means that (1) you want to calculate the thermal conductivity using the Green-Kubo method; (2) the heat current data will be recorded every 20 steps; (3) the maximum number of correlation steps is 50000; (4) the HAC/RTC data will be averaged for every 10 data and the number of HAC/RTC data output in a given direction is then $50000/10 = 5000$. The results will be written into a file named `hac.out` in the same folder where you put your `run.in` file in.

5.3.14 The compute hnmmd keyword

The `compute_hnmmd` keyword is used to calculate the RTC (running thermal conductivity) using the HNEMD method [12]. For the meanings of the RTC and the running average of this quantity, see Ref. [12]. This keyword has 4 parameters. The first parameter is the output interval. The next three parameters are the x , y , and z components of the external driving force \mathbf{F}_e in units of \AA^{-1} . For example, the command

```
compute_hnmmd 1000 0.00001 0 0
```

means that (1) you want to calculate the thermal conductivity using the HNEMD method; (2) the RTC will be averaged and output every 1000 steps (the heat current is sampled for every step); (3) the external driving force is along the x direction and has a magnitude of 10^{-5}\AA^{-1} . The results will be written into a file named `kappa.out` in the same folder where you put your `run.in` file in. Note that one should control the temperature when using this keyword. Otherwise, the system will be heated up by the external driving force.

5.3.15 The run keyword

This keyword only requires a single parameter, which is the number of steps for the current run. The time-evolution will only start when a `run` keyword has been reached. Before reaching this keyword, the code just collects the parameters for the current run. In the case where the VAC or the HAC is calculated, the number of steps should be larger than the product of the sampling interval and the number of correlation data. For example, the parameters in the commands

```
compute_hac 10 100000 10
run 10000000
```

are reasonably good because the number of steps (10^7) is 10 times as large as the product of the sampling interval and the number of correlation data ($10 \times 10^5 = 10^6$). In the case of calculating the VAC, it is important to first estimate the amount of memory to be used. Denote the number of steps as N_{run} and the sampling interval as N_{samp} , the

memory to be used for holding the velocity data is $(N_{\text{run}}/N_{\text{samp}}) \times N \times 3 \times 8$ bytes if using double-precision. If the number of atoms is $N = 10^4$, $N_{\text{samp}} = 5$, and $N_{\text{run}} = 10^5$, the memory to be used for holding the velocity data is about 4.8 GB. This is ok for Tesla K40 and K80, but may be too much for older GPUs.

5.4 Output files of GPUMD

In this section, we describe the formats of the output files. The output files and the corresponding keywords (used in the `run.in` file) generating them are:

- `thermo.out` is generated by `dump_thermo`
- `xyz.out` is generated by `dump_position`
- `velocity.out` is generated by `dump_velocity`
- `force.out` is generated by `dump_force`
- `potential.out` is generated by `dump_potential`
- `virial.out` is generated by `dump_virial`
- `heat.out` is generated by `dump_heat`
- `compute.out` is generated by `compute`
- `vac.out` is generated by `compute_vac`
- `hac.out` is generated by `compute_hac`
- `shc.out` is generated by `compute_shc`
- `kappa.out` is generated by `compute_hnemd`

5.4.1 An important note

For all the output files, data from a new simulation will be appended to existing data. Therefore, if you do not intend to append new data to existing ones, you'd better first remove the existing output file or rename it.

5.4.2 The thermo.out file

This file is generated by using the `dump_thermo` keyword in the `run.in` file. There are 9 columns in the `thermo.out` file, each containing the values of a quantity at increasing time points. The quantities are as follows:

- column 1: temperature (in units of K)
- column 2: kinetic energy (in unit of eV) of the system
- column 3: potential energy (in unit of eV) of the system
- column 4: pressure (in units of GPa) in the x direction

- column 5: pressure (in units of GPa) in the y direction
- column 6: pressure (in units of GPa) in the z direction
- column 7: box length (in units of Å) in the x direction
- column 8: box length (in units of Å) in the y direction
- column 9: box length (in units of Å) in the z direction

5.4.3 The xyz.out file

There are 3 columns in the xyz.out file, corresponding to the x , y , and z coordinates of the system at increasing time points. For example, if there are 4 atoms (labelled from 0 to 3) and you have saved 2 frames (corresponding to t_0 and t_1) of the configuration into the xyz.out file, the data will be arranged in the following way:

```
x_0(t_0) y_0(t_0) z_0(t_0)
x_1(t_0) y_1(t_0) z_1(t_0)
x_2(t_0) y_2(t_0) z_2(t_0)
x_3(t_0) y_3(t_0) z_3(t_0)
x_0(t_1) y_0(t_1) z_0(t_1)
x_1(t_1) y_1(t_1) z_1(t_1)
x_2(t_1) y_2(t_1) z_2(t_1)
x_3(t_1) y_3(t_1) z_3(t_1)
```

5.4.4 The velocity.out file

Similar to the xyz.out file, but for the velocities, in the natural unit used in GPUMD.

5.4.5 The force.out file

Similar to the xyz.out file, but for the forces, in unit of eV/Å.

5.4.6 The potential.out file

Similar to the xyz.out file, but for the potentials in unit of eV. Note that there is only one column, as potential is a scalar.

5.4.7 The virial.out file

Similar to the xyz.out file, but for the virial stresses, in unit of eV. Note that the three columns correspond to the three diagonal elements of the virial stress tensor. The off-diagonal elements are not dumped.

5.4.8 The vac.out file

This file contains the data of VAC (velocity autocorrelation) and related quantities, namely, the RDC (running diffusion coefficient) and the PDOS (phonon density of states). The data in this file are organized as follows:

- column 1: correlation time (in units of ps)
- column 2: VAC (in units of $\text{\AA}^2/\text{ps}^2$) in the x direction
- column 3: VAC (in units of $\text{\AA}^2/\text{ps}^2$) in the y direction
- column 4: VAC (in units of $\text{\AA}^2/\text{ps}^2$) in the z direction
- column 5: RDC (in units of $\text{\AA}^2/\text{ps}$) in the x direction
- column 6: RDC (in units of $\text{\AA}^2/\text{ps}$) in the y direction
- column 7: RDC (in units of $\text{\AA}^2/\text{ps}$) in the z direction
- column 8: angular frequency ω in units of THz
- column 9: DOS (in units of $1/\text{THz}$) in the x direction
- column 10: DOS (in units of $1/\text{THz}$) in the y direction
- column 11: DOS (in units of $1/\text{THz}$) in the z direction

5.4.9 The hac.out file

This file contains the data of HAC (heat current autocorrelation) and RTC (running thermal conductivity), organized in the following way:

- column 1: correlation time (in units of ps)
- column 2: $\langle J_x^{\text{in}}(0)J_x^{\text{tot}}(t) \rangle$ (in units of eV^3/amu)
- column 3: $\langle J_x^{\text{out}}(0)J_x^{\text{tot}}(t) \rangle$ (in units of eV^3/amu)
- column 4: $\langle J_y^{\text{in}}(0)J_y^{\text{tot}}(t) \rangle$ (in units of eV^3/amu)
- column 5: $\langle J_y^{\text{out}}(0)J_y^{\text{tot}}(t) \rangle$ (in units of eV^3/amu)
- column 6: $\langle J_z^{\text{tot}}(0)J_z^{\text{tot}}(t) \rangle$ (in units of eV^3/amu)
- column 7: $\kappa_x^{\text{in}}(t)$ (in units of $\text{Wm}^{-1}\text{K}^{-1}$)
- column 8: $\kappa_x^{\text{out}}(t)$ (in units of $\text{Wm}^{-1}\text{K}^{-1}$)
- column 9: $\kappa_y^{\text{in}}(t)$ (in units of $\text{Wm}^{-1}\text{K}^{-1}$)
- column 10: $\kappa_y^{\text{out}}(t)$ (in units of $\text{Wm}^{-1}\text{K}^{-1}$)
- column 11: $\kappa_z^{\text{tot}}(t)$ (in units of $\text{Wm}^{-1}\text{K}^{-1}$)

Note that the HAC and the RTC have been decomposed as described in Ref. [13]. This decomposition is useful for 2D materials but is not necessary for 3D materials. For 3D materials, one can sum up some columns to get the conventional data. For example:

$$\langle J_x^{\text{tot}}(0)J_x^{\text{tot}}(t) \rangle = \langle J_x^{\text{in}}(0)J_x^{\text{tot}}(t) \rangle + \langle J_x^{\text{out}}(0)J_x^{\text{tot}}(t) \rangle. \quad (5.1)$$

$$\kappa_x^{\text{tot}}(t) = \kappa_x^{\text{in}}(t) + \kappa_x^{\text{out}}(t). \quad (5.2)$$

Note that the “cross term” introduced in Ref. [13] has been evenly attributed to the in- and out-of-plane components. This has been justified in Ref. [12].

5.4.10 The compute.out file

This file contains data produced by using the `compute` command. Assuming that the system is divided into M groups, then

- if temperature is computed:
 - M columns of group temperatures (in units of K) from left to right
 - the next column is the total energy of the thermostat coupling to the heat source region (in units of eV)
 - the next column is the total energy of the thermostat coupling to the heat sink region (in units of eV)
- if potential is computed:
 - M columns of group potentials (in units of eV) from left to right
- if force is computed:
 - $3M$ columns of group forces (in units of eV/Å) from left to right
- if virial is computed:
 - $3M$ columns of group virials (in units of eV) from left to right
- if heat current is computed:
 - $3M$ columns of group heat currents (in units of eV^{3/2} amu^{-1/2}) from left to right

5.4.11 The shc.out file

This file contains data for the nonequilibrium heat current correlation function $K_{A \rightarrow B}(t)$ as defined in Eq. (3.118), in units of eV^{3/2} amu^{-1/2} Å⁻¹. There are two columns, which are $K_{A \rightarrow B}^{\text{in}}(t)$ and $K_{A \rightarrow B}^{\text{out}}(t)$ defined in Ref. [13]. If one does not need the decomposition, one can sum up the two columns to get the total correlation function. Note that an individual correlation function corresponds to N_c consecutive rows.

5.4.12 The kappa.out file

This file is generated by the `compute_hnemd` command and contains the running thermal conductivity calculated by using the HNEMD method. There are five columns, which are κ_x^{in} , κ_x^{out} , κ_y^{in} , κ_y^{out} , and κ_z from the first to the last column. For the meanings of these thermal conductivity components, see Ref. [12, 44, 9].

Chapter 6

Examples

In this chapter, we give some examples to illustrate the usage of GPUMD. All the results presented here are obtained by using the **double-precision** version of the code. The single-precision version is faster but we are not sure whether it is always as safe to use. Note that even the double-precision version of GPUMD is highly efficient. For details of the performance of GPUMD, see Ref. [11].

For each example, we have provided a `run.in` file and two or more MATLAB scripts:

- `create_xyz.m`, which can be used to create the `xyz.in` input file
- script(s) starting with `plot_`, which can be used to analyze the output data

After gaining some experiences in using GPUMD, one can use any other program to do these pre-processing and post-processing jobs.

6.1 Thermal expansion of silicon crystal

A given crystal should have a well defined average lattice constant at a given pressure and temperature. Here we use silicon as an example to show how to calculate lattice constants using GPUMD. We use a cubic system (of diamond structure) consisting of $10^3 \times 8 = 8000$ silicon atoms and use the Tersoff-1989 potential.

The first few lines of the `xyz.in` file created by the `create_xyz.m` MATLAB script are:

```
8000 4 3 0 0 0
1 1 1 54.3 54.3 54.3
0 0 0 0 28
0 0 2.715 2.715 28
0 2.715 0 2.715 28
0 2.715 2.715 0 28
```

The first line tells that the number of particles is 8000, the neighbor list size will be 8000×4 , and the initial cutoff distance for the neighbor list construction is 3 Å. The remaining three zeros in the first line means: 1) initial velocities are not contained in this file; 2) there is no need to use layer labels; 3) there is no need to use any grouping method. The second parameter 4 in this line is good for silicon crystal described by the Tersoff potential, because no atom can have more than 4 neighbor atoms in the temperature

range studied. One can make the second number larger, which only results in using more memory. If this number is not large enough, GPUMD will give an error message and exit.

In the second line, the first three ones means that all three directions are periodic. The remaining three numbers are the box lengths in the three directions.

Starting from the third line, the numbers in the first column are all 0 here, which means that all the atoms are of type 0 (single atom-type system). The next three columns are the initial coordinates of the atoms. The last column gives the masses of the atoms. Here, we consider isotopically pure Si-28 crystal. In some applications, one can consider mass disorder in a flexible way.

The “run.in” input file is given below. The first line of command tells that the potential to be used is specified in the file `potentials/si_tersoff_1989_1.txt`. The second line of the command tells that the velocities will be initialized with a temperature of 1 K. Then, the next 4 lines tell how to do the first run. This run will be in the *NPT* ensemble, using the Berendsen method. The temperature is 1 K and the pressures are zero in all the directions. The coupling constants are 0.01 (dimensionless) and 0.0005 (in the natural unit system adopted by GPUMD) for the thermostat and the barostat, respectively. The time step for integration is 1 fs. There are 10^5 steps for this run and the thermodynamic quantities will be output every 100 steps. After this run, there are 5 other runs with the same parameters but the target temperature. Note that the time step only needs to be set once if one is intended to use the same time step in the whole simulation. In contrast, one has to use the `dump_thermo` keyword for each run in order to get outputs for each run.

```
#-----
potential    potentials/si_tersoff_1989_1.txt
velocity     1

ensemble     npt_ber 1 1 0.01 0 0 0 0.0005
time_step    1
dump_thermo  100
run          100000

ensemble     npt_ber 200 200 0.01 0 0 0 0.0005
dump_thermo  100
run          100000

ensemble     npt_ber 400 400 0.01 0 0 0 0.0005
dump_thermo  100
run          100000

ensemble     npt_ber 600 600 0.01 0 0 0 0.0005
dump_thermo  100
run          100000

ensemble     npt_ber 800 800 0.01 0 0 0 0.0005
dump_thermo  100
run          100000

ensemble     npt_ber 1000 1000 0.01 0 0 0 0.0005
```

```

dump_thermo 100
run          100000
#-----

```

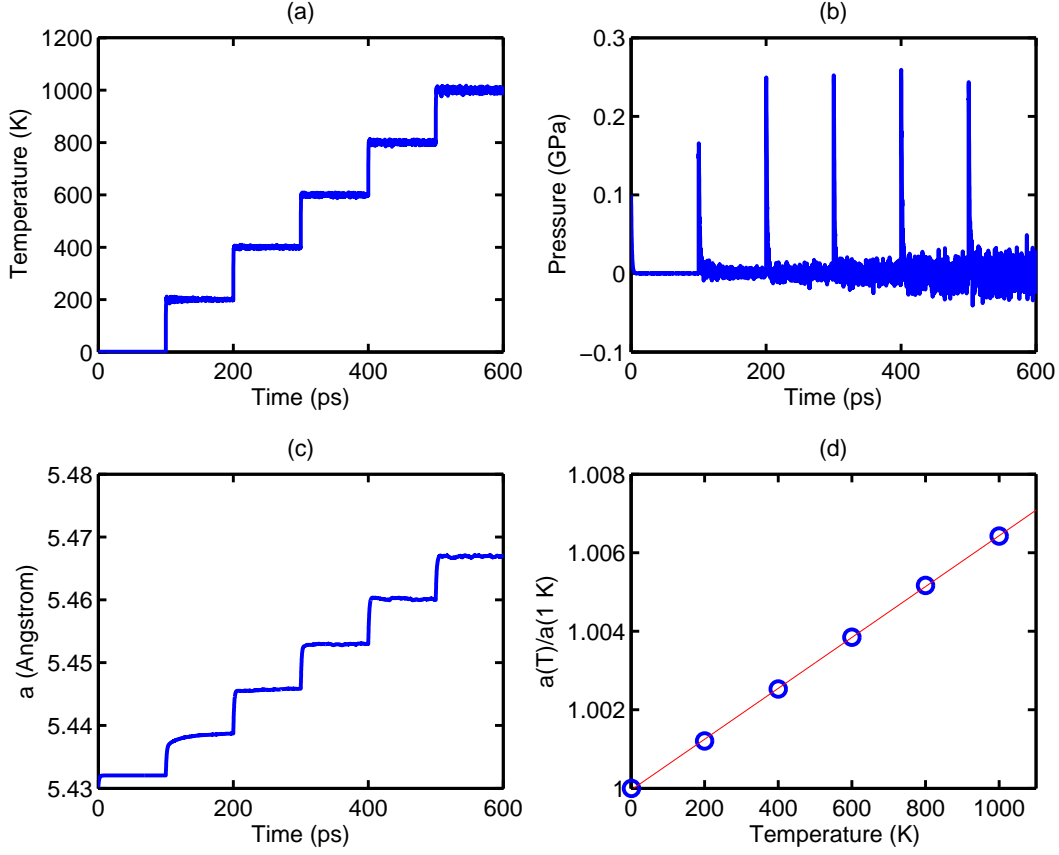


Figure 6.1: (a) Instant temperature as a function of simulations time. (b) Instant pressure as a function of simulation time. (c) Instant lattice constant as a function of simulations time. (d) Normalized average lattice constant (over the last 50 ps in each run for a given temperature) as a function of temperature.

It takes about 4 min to run this example when a Tesla K40 card is used. The speed of the run is about 1.9×10^7 atom \times step/second.

The output file `thermo.out` contains many useful data, which can be analyzed by the MATALB script `plot_results.m`. The results are shown in Fig. 6.1:

- (a): The temperature for each run quickly reaches the target temperature (with fluctuations).
- (b): The pressure (averaged over the three directions) for each run quickly reaches the target pressure zero (with fluctuations).
- (c): The lattice constant (averaged over the three directions) for each run reaches a plateau (with fluctuations) after some steps.

- (d): We calculate the average lattice constant at each temperature by averaging the second half of the data for each run. The average lattice constants at different temperatures can be well fit by a linear function, with the thermal expansion coefficient being estimated to be $\alpha \approx 6.5 \times 10^{-6} \text{ K}^{-1}$.

6.2 Phonon density of states of graphene

In this example, we calculate the phonon density of states of graphene at 300 K and zero pressure. The simulated cell size is about $15 \text{ nm} \times 15 \text{ nm}$ (8 640 atoms). The first few lines of the `xyz.in` file are:

```
8640 3 2.1 0 0 0
1 1 0 149.649 155.52 3.35
0 1.24708 0 0 12
0 0 0.72 0 12
0 0 2.16 0 12
0 1.24708 2.88 0 12
```

This is a stable structure with 3 neighbors for each atom when periodic boundary conditions are applied in the planar directions (x and y). In the z direction, free boundary conditions are used. Every atom is of type 0.

The `run.in` file reads:

```
#-----
potential    potentials/c_tersoff_fan_2017.txt
velocity     300

ensemble     npt_ber 300 300 0.01 0 0 0 0.0005
time_step    1
dump_thermo  1000
run          1000000

ensemble     nve
compute_vac  5 200 400
run          100000

ensemble     nve
compute_vac  5 200 400
run          100000

ensemble     nve
compute_vac  5 200 400
run          100000

ensemble     nve
compute_vac  5 200 400
run          100000

ensemble     nve
```

```

compute_vac 5 200 400
run          100000
#-----

```

The potential model is Tersoff-1989, but some parameters are those reparameterized by Lindsay and Broido [27]. In the version by Lindsay and Broido, the carbon-carbon bond length at zero temperature is 1.44 Å, which is larger than the experimental value, 1.42 Å. Because the only relevant length parameters in the Tersoff-1989 potential are λ and μ in the repulsive and attractive functions, we can simply correct the bond length by a proper scaling of these two parameters. All the other parameters are not affected by this scaling.

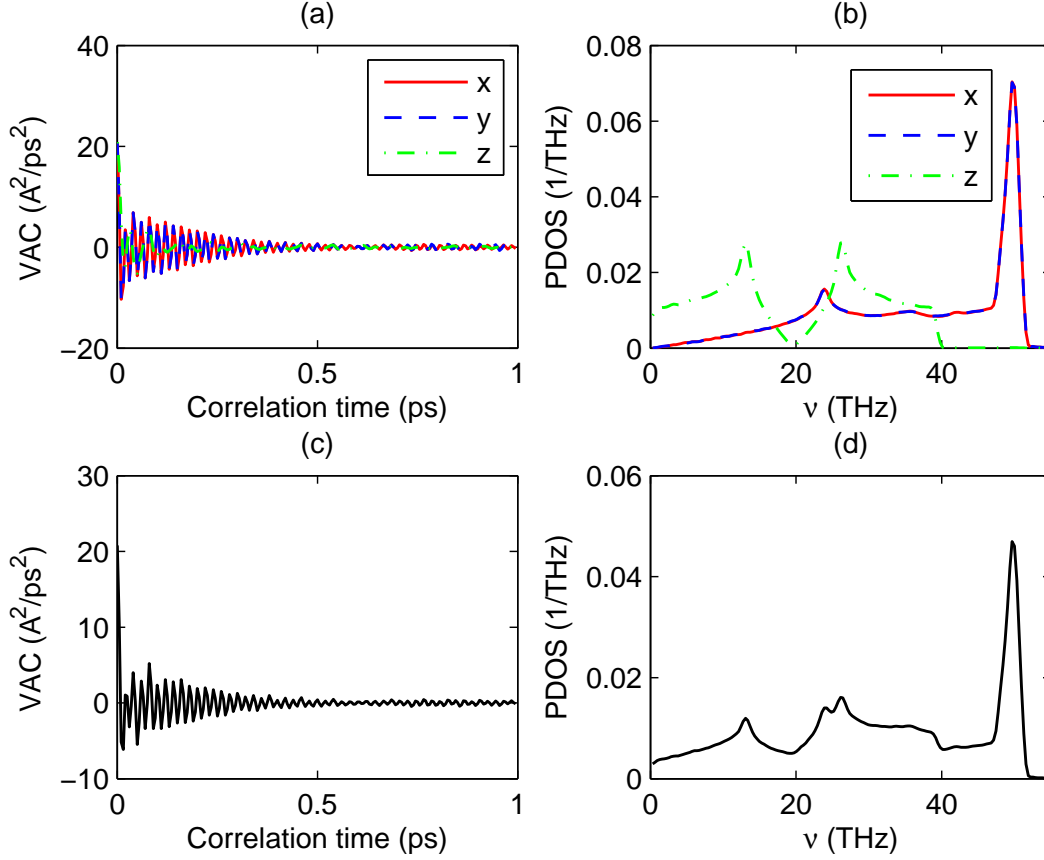


Figure 6.2: (a) VAC as a function of correlation time for the separate directions. (b) PDOS as a function of the phonon frequency for the separate directions. (c) VAC as a function of correlation time averaged over the separate directions. (d) PDOS as a function of the phonon frequency average over the separate directions.

There are 6 runs. The first run serves as the equilibration stage, where the *NPT* ensemble is used. This run lasts 1 ns. The other 5 runs are identical production runs. In each production run, the *NVE* ensemble is used. The line with `compute_vac` means that velocities will be recorded every 5 steps (5 fs) and 200 VAC data (the maximum correlation time is then about 1 ps) will be calculated. The last parameter in this line is the maximum angular frequency considered, $\omega_{\text{max}} = 2\pi\nu_{\text{max}} = 400$ THz, which is large enough for graphene. Each production run lasts 100 ps. The major reason for using

multiple production runs rather a single one is that computing the VAC requires a lot of memory, which prevents using very long runs.

This simulation takes about 7 min when a Tesla K40 is used. The speed of this simulation, being about 3.6×10^7 atom \times step/second, is higher than that of the previous example because the number of neighbors for each atom is smaller here (numbers of atoms are comparable and the potential models are the same).

Figure 6.2 shows the calculated VAC and PDOS. For 3D isotropic systems, the results along different directions are equivalent and can be averaged, but for 2D materials like graphene, it is natural to consider the in-plane part (the x and y directions in the simulation) and the out-of-plane part (the z direction) separately. It can be seen that the two components behave very differently. We can see that the cutoff frequency for the out-of-plane component (~ 40 THz) is smaller than that for the in-plane component (~ 52 THz), which means that the two components have different Debye temperatures.

6.3 Thermal conductivity of graphene

In this example, we use the Green-Kubo method to calculate the lattice thermal conductivity of graphene at 300 K and zero pressure. The `xyz.in` file and the potential parameters used are the same as in the last example. Note that the thickness of the graphene sheet is set to 3.35 Å according to the convention in the literature. This thickness is needed to calculate an effective 3D thermal conductivity for a 2D material.

The `run.in` file for this simulation reads:

```
#-----
potential          potentials/c_tersoff_fan_2017.txt
velocity           300

ensemble           npt_ber 300 300 0.01 0 0 0 0.0005
time_step          1
dump_thermo        1000
run                1000000

ensemble           nve
compute_hac         20 50000 10
run                10000000
#-----
```

The equilibrium stage is the same as in the last example. In the production stage, we use the NVE ensemble and calculate the HAC (heat current autocorrelation) and RTC (running thermal conductivity). The sampling interval is 20, the number of correlation steps is 50000 (such that the maximum correlation time is about 10^6 fs = 1 ns), and the HAC and RTC are averaged for every 10 data points before written out. The production time is 10 ns, which is 10 times as long as the maximum correlation time. This is a reasonable choice.

Figure 6.3 shows the results from three independent simulations, which took about two hours in total using a Tesla K40 card. Note that the output file `hac.out` contains decomposed HACs and RTCs as described in Ref. [13], but without the “cross term”. As the system is essentially isotropic in the planar directions, we can average over the two in-plane directions. From Fig. 6.3(a), we can see that the in-plane component and the

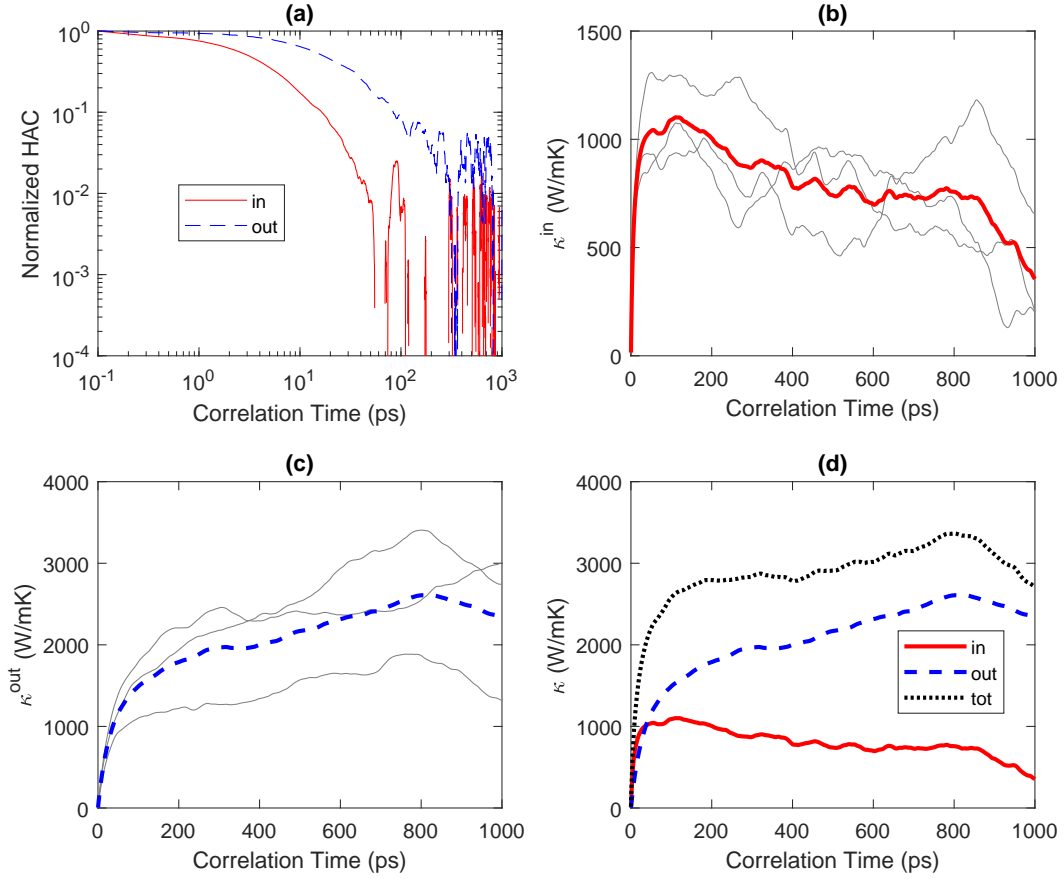


Figure 6.3: Thermal conductivity results for pristine graphene at 300 K. (a) Normalized HAC as a function of correlation time for the in-plane and out-of-plane components. (b) Individual (thin lines) and averaged (thick line) RTC as a function of correlation time for the in-plane component. (c) Individual (thin lines) and averaged (thick line) RTC as a function of correlation time for the out-of-plane component. (d) Averaged RTC as a function of correlation time for various components.

out-of-plane component of the HAC have different time scales. The latter decays much more slowly. Figure 6.3(b) shows the individual and averaged RTCs for the in-plane component $\kappa^{\text{in}}(t)$. The averaged RTC converges to about 1000 W/mK at around 200 ps. Figure 6.3(c) shows the individual and averaged RTCs for the out-of-plane component $\kappa^{\text{out}}(t)$, and the convergence property is not very clear here. This is because the out-of-plane component converges very slowly [13] and three independent simulations (each with 10 ns) are not enough to give accurate results. Summing up $\kappa^{\text{in}}(t)$ and $\kappa^{\text{out}}(t)$, we get $\kappa^{\text{tot}}(t)$, as shown in Fig. 6.3(d).

Accurately calculating thermal conductivity of graphene can be a very time consuming task. The results we presented are from three independent simulations with a total production time of 30 ns. It can be seen that the HAC data already become very noisy when the correlation time is 100 ps. To obtain accurate results, one needs to do many independent simulations. Much more accurate data were presented in Fig. 2 of Ref. [13]. Here are the simulation parameters used in Ref. [13] which differ from those used in this

example:

- The simulation cell size used in Ref. [13] is larger, which is about $25 \text{ nm} \times 25 \text{ nm}$ (24000 atoms).
- The maximum correlation time used in Ref. [13] is larger, which is 10 ns.
- The production time used in Ref. [13] for one independent simulation is larger, which is 50 ns.
- There are 100 independent simulations in Ref. [13], not only three.

Each independent simulation in Ref. [13] took about 10 GPU hours (using Tesla K40) and about 1000 GPU hours were used to obtain the results shown in Fig. 2 of Ref. [13].

6.4 Ballistic thermal conductance of graphene

In this example, we show how to study heat transport using the NEMD method combined with the spatial and spectral decompositions as described in Ref. [13]. We aim to obtain similar results for the case of unstrained graphene as presented in Fig. 4 of Ref. [13].

In the NEMD simulation, periodic boundary conditions are applied to the transverse direction (chosen as the zigzag direction) and fixed boundary conditions are applied to the transport direction (chosen as the armchair direction). The width of the simulated system is about 10 nm and the total length in the transport direction is about 70 nm. One major difference from the previous simulations is that here the group labels are not identically 0. We divide the system into 8 groups along the transport direction and label the groups from 1 to 8. Groups 1 and 8 are taken as the source and sink regions, respectively. The number of atoms in groups 1 to 8 are 8000, 1600, 1600, 1600, 1600, 1600, 1600, and 8000, respectively. Some extra fixed atoms are put into group 0.

The `run.in` file for this example reads:

```
#-----
potential      potentials/c_tersoff_fan_2017.txt
velocity       300

ensemble       nvt_ber 300 300 0.01
fix            0
time_step      1
dump_thermo    1000
run            1000000

ensemble       heat_nhc 300 100 10 1 8
fix            0
compute        100 10 temperature
compute_shc    2 250 100000 4 5
run            2000000
#-----
```

In this simulation, we fix the lattice constant and only control the temperature in the equilibration stage. The `fix 0` command is used to realize the fixed boundary conditions

by fixing the atoms in group 0. In the production stage, the `heat_nhc` “ensemble” type is used to generate the nonequilibrium heat current. The heat source (group 1) and the heat sink (group 8) will be maintained at 310 K and 290 K, respectively. The block temperatures will be output every 1000 integration steps. The command with the keyword `compute_shc` means that the nonequilibrium heat current correlation function defined in Eq. (3.118) will be computed: the sampling interval is 2, the number of correlation steps is 250, the number of steps for calculating one correlation function is 100000, and the heat current considered flows from group 4 to group 5 (the middle interface of the simulated system).

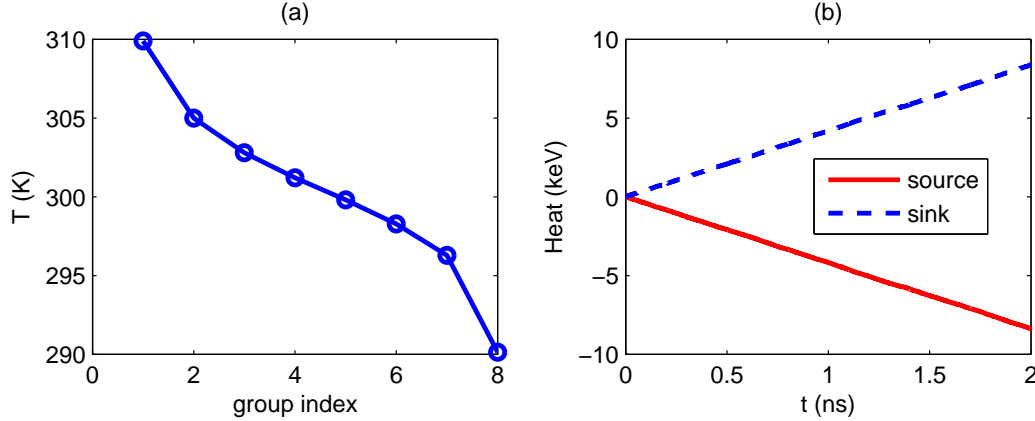


Figure 6.4: (a) Temperature profile in the NEMD simulation. (b) Total energy of the thermostats as a function of simulation time in the production stage.

This simulation takes about 40 min using a Tesla K40 card. Figure 6.4 shows the results obtained from the `temperature.out` file:

- (a): The block temperatures show a relatively smooth profile, but no clear linear region can be identified. Actually, heat transport here is ballistic and we do not expect to find a well defined temperature gradient (which is needed for computing the thermal conductivity). What is important is that a well defined temperature difference (20 K), which is needed for computing the ballistic thermal conductance, can be established.
- (b): A steady energy exchange between the system and the thermostats in the source and sink regions has been well established. The nonequilibrium heat current can be estimated to be about $Q = 4.2$ eV/ps. Then a ballistic conductance of about $G = 10.2$ GW m⁻² K⁻¹ can be obtained. This classical value overestimates the correct one and we will add discussion about quantum corrections after some of our submitted manuscripts get published.

The `shc.out` file contains data for the nonequilibrium heat current autocorrelation function $K(t)$ as defined in Eq. (3.118). Also, the in-out decompositions introduced in Ref. [13] is considered. The calculated $K(t)$ and the spectrally decomposed conductance $g(\omega)$ for the in-plane and the out-of-plane components are shown in Fig. 6.5. One can see that they are similar to the velocity autocorrelation and the phonon density of states discussed in a previous example. This is reasonable because the ballistic conductance is proportional to the product of the phonon density of states and the group velocity.

One can check the consistency of the results at least in the following two ways:

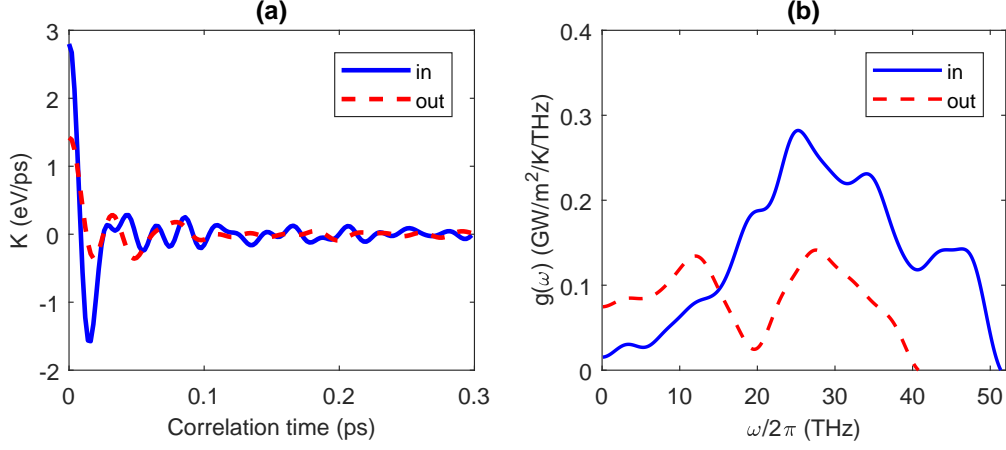


Figure 6.5: (a) The nonequilibrium heat current autocorrelation function as define in Eq. (3.118) as a function of correlation time. (b) The spectrally decomposed ballistic conductance as a function of phonon frequency.

- When steady state is achieved, the correlation function $K(t)$ evaluated at zero correlation time should be consistent with the heat current calculated from the energy exchange rate between the system and the thermostats. That is, $K(0) = Q$.
- The total thermal conductance G should equal the integration of the spectral conductance. That is, $G = \int_0^\infty \frac{d\omega}{2\pi} g(\omega)$.

One should always make sure that the obtained data pass these tests approximately.

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