Basics of Statistical Mechanics

• Review of ensembles
  – Microcanonical, canonical, Maxwell-Boltzmann
  – Constant pressure, temperature, volume,…

• Thermodynamic limit

• Ergodicity (see online notes also)
Newtonian molecular dynamics: fundamentals

- Pick particles, masses and potential (i.e. forces)
- Initialize positions and momentum (i.e., boundary conditions in time)
- Solve $F = ma$ to determine $r(t), v(t)$.
- Compute properties along the trajectory
- Estimate errors.
- Try to use the simulation to answer physical questions.

Also we need **boundary conditions** in space and time. Real systems are **not isolated**!

What about interactions with walls, stray particles?

How can we treat $10^{23}$ atoms at long times?
Statistical Ensembles

- Classical phase space is $6N$ variables $(q_i, p_i)$ with Hamiltonian $H(q, p, t)$.
- We may know a few constants of motion such as energy, momentum, number of particles, volume, ...
- The most fundamental way to understand the foundation of statistical mechanics is by using quantum mechanics:
  - In a finite system, there are a countable number of states with various properties, e.g. energy $E_i$.
  - For each energy interval we can define the density of states,
    \[ g(E) \, dE = \exp\left(\frac{S(E)}{k_B}\right) \, dE \]
    where $S(E)$ is the entropy.
  - If all we know is the energy, we have to assume that each state in the interval is equally likely. (Maybe we know the momentum or another property)
Environment

- Perhaps the system is isolated. No contact with outside world. This is appropriate to describe a cluster in vacuum.
- Or we have a heat bath: replace surrounding system with heat bath. All the heat bath does is occasionally shuffle the system by exchanging energy, particles, momentum,.....

The only distribution consistent with a heat bath is a canonical distribution:

$$\text{Prob}(q, p) \ dqdp = e^{-\beta H(q,p)}/Z$$
Interaction with environment: $E = E_1 + E_2$

- The number of energy states in thermodynamic system ($N \sim 10^{23}$) is very large! $g(E) =$ density of states. Combined density of states:

$$g(E; N_s + N_e, V_s + V_e) = g_s(E_1; N_s, V_s)g_e(E - E_1; N_e, V_e)$$

- Easier to use: \( \ln g(E) = \ln g_s(E_1) + \ln g_e(E - E_1) \)

- This is the entropy \( S(E) \): $g(E) \, dE = \exp(S(E)/k_B) \, dE$

- The most likely value of \( E_1 \) maximizes \( \ln g(E) \). This gives 2\textsuperscript{nd} law.

  - Temperatures of 1 and 2 the same:

\[
\beta = \frac{1}{k_B T} = \frac{d \ln g}{dE} = \frac{1}{k_B} \frac{dS}{dE}
\]

- Assuming that the environment has many degrees of freedom:

\[
P_s(E) = \exp(-\beta E_s)/Z \quad \langle A \rangle = \text{Tr}\{P_s(E)A\} \]
Statistical ensembles

• \((E, V, N)\) microcanonical, constant volume
• \((T, V, N)\) canonical, constant volume
• \((T, P, N)\) canonical, constant pressure
• \((T, V, \mu)\) grand canonical (variable particle number)

• Which is best? It depends on:
  – the question you are asking
  – the simulation method: MC or MD (MC better for phase transitions)
  – your code.
• Lots of work in recent years on various ensembles (later).
Maxwell-Boltzmann Distribution

\[ \text{Prob}(q,p) \, dqdp = e^{-\beta H(q,p)}/Z \]

- **Z**: partition function. Defined so that probability is normalized.
  \[ Z = \frac{1}{N!\hbar^{3N}} \int d^{3N}q \, d^{3N}p \, \exp(-\beta H(q,p)) \]
- Quantum expression:
  \[ Z = \sum_i \exp(-\beta E_i) \]
- Also \( Z = \exp(-\beta F) \), \( F \): free energy (more convenient since \( F \) is extensive)
- Classically:
  \[ H(q,p) = \sum_i \frac{1}{2m_i} p_i^2 + V(q) \]
- Then the momentum integrals can be performed. One has simply an uncorrelated Gaussian (Maxwell) distribution of momentum.
- On the average, there is no relation between position and velocity!
- Microcanonical is different: think about harmonic oscillator.
- **Equipartition Theorem**: Each quadratic variable carries \( \frac{1}{2}k_B T \) of energy

\[ \left\langle \frac{1}{2m_i} p_i^2 \right\rangle = \frac{3}{2} k_B T \]
Thermodynamic limit

• To describe a macroscopic limit we need to study how systems converge as \( N \to \infty \) and as \( t \to \infty \).

• Sharp, mathematically well-defined phase transitions only occur in this limit. Otherwise they are not perfectly sharp.

• It has been found that systems of \textit{as few as 20 particles with only thousand of steps} can be close to the limit \textit{if you are very careful with boundary conditions} (spatial BC).

• To get this behavior consider whether:
  – Have your BCs introduced anything that shouldn’t be there? (walls, defects, voids, etc.)
  – Is your box bigger than the natural length scale? (for a liquid/solid it is the interparticle spacing)
  – The system starts \((t=0)\) in a reasonable state (BC in time!).
Thermodynamic limit

Binder, Landau: spontaneous magnetization of Ising model
Ergodicity

- In MD we often use microcanonical ensemble: just $F=ma!$ $E$ is conserved.
- Replace ensemble or heat bath with a single very long trajectory.
- This is OK only if system is ergodic.
- Ergodic Hypothesis: a phase point for any isolated system passes in succession through every point compatible with the energy of the system before finally returning to its original position in phase space. (a Poincare cycle).
- The Ergodic hypothesis: each state consistent with our knowledge is equally “likely”.
  - Implies the average value does not depend on initial conditions.
  - Is $<A>_{\text{time}} = <A>_{\text{ensemble}}$ a good estimator? $<A> = \frac{1}{N_{\text{MD}}} \sum_{t=1,N} A_t$
  - True if: $<A> = <\langle A \rangle_{\text{ens}}>_\text{time} = <\langle A \rangle_{\text{time}}>_\text{ens} = <A>_{\text{time}}$.
    - First equality is true if the distribution is stationary.
    - Second equality is true if interchanging averages does not matter.
    - Third equality is only true if system is ERGODIC.
- Are systems in nature really ergodic? Not always!
  - Non-ergodic examples are glasses, folding proteins (in practice), harmonic crystals (in principle), the solar system.
Different aspects of Ergodicity

- The system relaxes on a *reasonable time scale* towards a unique equilibrium state.
- This state is the microcanonical state. It differs from the canonical distribution by corrections of order \(1/N\).
- There are no hidden variable (conserved quantities) other than the energy, linear and angular momentum, number of particles. (systems which do have conserved quantities might be integrable.)
- Trajectories wander irregularly through the energy surface, eventually sampling all of accessible phase space.
- Trajectories initially close together separate rapidly. They are extremely sensitive to initial conditions; the “butterfly effect.” The coefficient is the Lyapunov exponent.

Ergodic behavior makes possible the use of **statistical methods** on MD of **small systems**.
Small round-off errors and other mathematical approximations **may not matter**! They may even help.
Ergodicity: pictorially

- MD:
  - Trajectory in time

- MC:
  - Individual samples
Particle in a smooth/rough circle

From J.M. Haile: MD Simulations
Fermi-Pasta-Ulam “experiment” (1954)

- 1-D anharmonic chain:
  \[ V(q) = \sum_i \left[ (q_{i+1} - q_i)^2 + \alpha (q_{i+1} - q_i)^3 \right] \]

- The system was started out with energy with the lowest energy mode. Equipartition implies that energy would flow into the other modes.
- Systems at low temperatures never come into equilibrium. The energy sloshes back and forth between various modes forever.
- At higher temperature many-dimensional systems become ergodic.
- The field of non-linear dynamics is devoted to these questions.
Let us say here that the results of our computations were, from the beginning, surprising us. Instead of a continuous flow of energy from the first mode to the higher modes, all of the problems show an entirely different behavior. ... Instead of a gradual increase of all the higher modes, the energy is exchanged, essentially, among only a certain few. It is, therefore, very hard to observe the rate of “thermalization” or mixing in our problem, and this was the initial purpose of the calculation.

Fermi, Pasta, Ulam (1954)
Distribution of normal modes.

High energy \((E=1.2)\)

- Time averages \(E_k(T), k = 1, \ldots, 32\), for the 32-particle FPU with \(\alpha = 0.1\) and higher specific energy \(\varepsilon = 1.2\).

Low energy \((E \sim 0.07)\)

Fig. 1. - Time averages \(E_1(T), \ldots, E_4(T)\) (solid lines, top to bottom), and \(\sum_{k=5}^{32} E_k(T)\) (dashed line), for the 32-particle FPU model with \(r = 3, \alpha = 0.1\) and \(\varepsilon \sim 0.07\). First mode initially excited.
Distribution of normal modes vs time-steps

- 20K steps

- 400K steps
  - Energy SLOWLY oscillates from mode to mode--never coming to equilibrium

$E_i(t)$, $E_j(t)$ and $E_k(t)$ (solid, dashed and dotted lines, respectively), functions of fig. 1.
Distribution of normal modes vs time-steps

- 20K steps
- 400K steps
- Energy SLOWLY oscillates from mode to mode—never coming to equilibrium

https://en.wikipedia.org/wiki/Fermi%E2%80%93Pasta%E2%80%93Ulam%E2%80%93Tsingou_problem
Aside from these mathematical questions, there is always a practical question of convergence: \textit{How do you judge if your results are converged?}

There is no sure way. Why?

There are only empirical tests for convergence such as:

- Occasionally do very long runs.
- Use different starting conditions. For example “quench” from higher temperature/higher energy states.
- Shake up the system.
- Use different algorithms such as MC and MD
- Compare to experiment or to another well-studied system.

\textbf{Convergence: practical issues}
Continuum of dynamical methods with different dynamics and ensembles

Path Integral Monte Carlo (quantum nuclei)
Ab initio Molecular Dynamics (no randomness)
semi-empirical Molecular Dynamics
Langevin Equation (heat bath adds more forces)
Brownian Dynamics (heat bath sets velocities)
Kinetic Monte Carlo (random walk biased by rates)
Metropolis Monte Carlo (unbiased random walk)
Smart Monte Carlo (random walk biased by force)

The general procedure is to average out fast degrees of freedom. Which is correct?