Thermostats in Molecular Dynamics Simulations

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Thermostat: A modification of the Newtonian MD scheme with the purpose of generating a *statistical ensemble* at a constant temperature.

- Match experimental conditions
- Manipulate temperatures in algorithms such as simulated annealing
- Avoid energy drifts caused by accumulation of numerical errors.

- Ensemble: a large collection of microscopically defined states of a system, with certain constant macroscopic properties
- Microcanonical (NVE)
 - Arises in Newtonian MD simulation
 - Conserves total energy
- Canonical (NVT)
 - Implement thermostats to sample from here
 - Relevant to real behavior in experiment

- Consider an isolated system
- Microstate: complete description of a state of the system, microscopically
- The probabilities of being in a certain microstate for the microcanonical ensemble are uniform over all possible states

 Follows a Gibbs distribution for probability p_j of being in a given microstate j with energy E_j

$$p_j = rac{e^{-eta E_j}}{Z_eta}, \qquad Z_eta = \sum_j e^{-eta E_j}$$
 $eta = rac{1}{k_B T}$

• Derivation follows from maximizing entropy

The ergodic hypothesis says the long time average of an observable *f* coincides with an ensemble average of the observable (*f*).

$$\overline{f} = \lim_{t o \infty} rac{1}{t} \int_0^t f(x(s)) \, \mathrm{d}s$$
 $\langle f
angle = \int_{\Gamma} f(x) \, \mathrm{d}\mu(x)$

where μ is the ensemble measure and Γ is the *phase space* of the observable.

- Phase space is collection of positions *q* and momenta *p* of particles in system
- The Hamiltonian Form

$$\begin{cases} dq_t = \nabla_p H(q_t, p_t) dt \\ dp_t = -\nabla_q H(q_t, p_t) dt \end{cases}$$
$$H(q, p) = E_{kin}(p) + V(q), \qquad E_{kin}(p) = \frac{1}{2} p^T M^{-1} p$$

- A canonical ensemble (constant average energy) is a distribution of microcanonical ensembles (constant energy)
- To sample from the canonical ensemble, the following thermostats modulate the energy entering and leaving the boundaries of the system

• Velocities are described by Maxwell-Boltzmann distribution

$$P(v_{i,\alpha}) = \left(\frac{m}{2\pi k_B T}\right)^{\frac{1}{2}} e^{-\frac{mv_{i,\alpha}^2}{2k_B T}}$$

- Adjust instantaneous temperature by scaling all velocities
- Average *E_{kin}* per degree of freedom related to T via the equipartition theorem

$$\left\langle \frac{m v_{i,\alpha}^2}{2} \right\rangle = \frac{1}{2} k_B T$$

 Ensemble average → average over velocities of all particles: define instantaneous temperature T_c for a finite system

$$k_B T_c = \frac{1}{N_f} \sum_{i,\alpha} m v_{i,\alpha}^2$$

• $T_c \neq T$ until rescaling

$$v_{i,\alpha}' = \sqrt{\frac{T}{T_c}} v_{i,\alpha}$$

Velocity Rescaling

Disadvantages

- Results do not correspond to any ensemble
 - ★ Does not allow the proper temperature fluctuations
- Localized correlation not removed
- Not time reversible
- Advantages
 - Straightforward to implement
 - Good for use in warmup / initialization phase

- Based on extended Lagrangian formalism
 - Deterministic trajectory
 - Simulated system contains virtual variables related to real variables
 - ***** Coordinates $\mathbf{q}'_i = \mathbf{q}_i$
 - ***** Momenta $\mathbf{p}'_i = \mathbf{p}_i / s$
 - ***** Time $t' = \int_0^t \frac{dt}{s}$
 - ★ s: additional degree of freedom, acts as external system

• Hamiltonian given by

$$H = \sum_{i=1}^{N} \frac{\mathbf{p}_{i}^{2}}{2m_{i}s^{2}} + V(\mathbf{q}) + \frac{Q\dot{s}^{2}}{2} + (3N+1)k_{B}T\ln s$$

- Logarithmic term required for proper time scaling: canonical ensemble
- Effective mass Q associated with s
 - Determines thermostat strength
 - Q too small: system not canonical
 - ► *Q* too large: temperature control inefficient
- Microcanonical dynamics on extended system give canonical properties

- Disadvantages
 - Extended system not guaranteed to be ergodic
- Advantages
 - Easy to implement and use
 - ★ Implement as a chain
 - $\star\,$ Each link: apply thermostatting to the previous thermostat variable
 - Increasing Q lengthens decay time of response to instantaneous temperature jump
 - Deterministic and time reversible

Langevin

 Consider the motion of large particles through a continuum of smaller particles

$$\frac{d\mathbf{q}_i}{dt} = \frac{\mathbf{p}_i}{m_i}$$
$$\frac{d\mathbf{p}_i}{dt} = -\frac{\delta V(\mathbf{q})}{\delta \mathbf{q}_i} - \gamma \mathbf{p}_i + \sigma G_i$$

- Viscous drag force proportional to velocity $-\gamma \mathbf{p}_i$
- Smaller particles give random pushes to large particle
- Fluctuation-dissipation relation

$$\sigma^2 = 2\gamma m_i k_B T$$

Langevin

Disadvantages

- \blacktriangleright Difficult to implement drag for non-spherical particles: γ related to particle radius
- Momentum transfer lost: cannot compute diffusion coefficients
- Advantages
 - Damping + random force \mapsto correct canonical ensemble
 - Ergodic
 - Can use larger time step

- Couple a system to a heat bath to impose desired temperature
- Equations of motion are Hamiltonian with stochastic collision term
- Strength of coupling specified by ν , the stochastic collision frequency
- When particle has collision, new velocity is sampled from $\mathcal{N}(0,\sqrt{\mathcal{T}})$

Disadvantages

- $\blacktriangleright \ \ \mathsf{Newtonian} \ \ \mathsf{dynamics} + \mathsf{stochastic} \ \mathsf{collisions} \to \mathsf{Markov} \ \mathsf{chain}$
- Algorithm randomly decorrelates velocity: dynamics are not physical
- Advantages
 - Allows sampling from canonical ensemble

Thermostat	Description	Canonical?	Stochastic?
Velocity Rescaling	KE fixed to match T	no	no
Nosé-Hoover	extra degree of freedom	yes	no
	acts as thermal reser-		
	voir		
Langevin	noise and drag balance	yes	yes
	to give correct T		
Andersen	momenta occasionally	yes	yes
	re-randomized		

Applied Math Masters Project

- Molecular Dynamics Simulation
 - Implemented in MATLAB and C++
 - Simulations in 2 and 3 dimensions
 - Periodic and walled boundary conditions
 - External fields such as gravity
 - Optimization
 - ★ OpenMP
 - ★ CUDA

Lennard-Jones Potential



Numeric Integration

• Verlet Algorithm

$$p^{n+1/2} = p^n - rac{\Delta t}{2}
abla V(q^n)$$

$$q^{n+1} = q^n + \Delta t M^{-1} p^{n+1/2}$$

$$p^{n+1} = p^{n+1/2} - \frac{\Delta t}{2} \nabla V(q^{n+1})$$

Preserves modified Hamiltonian

Thermostat Implementation

Velocity Rescaling

•
$$p_i \rightarrow \sqrt{\frac{T}{T_c}} p_i$$

- Anderson
 - ▶ $\nu = 1\%, 0.1\%, 0.05\%$
 - Velocities are sampled from $\mathcal{N}(0, \sqrt{T})$
- Langevin
 - BBK algorithm

$$p^{n+1/2} = p^n - \frac{\Delta t}{2} \nabla V(q^n) - \frac{\Delta t}{2} \gamma(q^n) M^{-1} p^n + \sqrt{\frac{\Delta t}{2}} \sigma(q^n) G^n$$
$$q^{n+1} = q^n + \Delta t M^{-1} p^{n+1/2}$$

$$p^{n+1} = p^{n+1/2} - \frac{\Delta t}{2} \nabla V(q^{n+1}) - \frac{\Delta t}{2} \gamma(q^{n+1}) M^{-1} p^{n+1} + \sqrt{\frac{\Delta t}{2}} \sigma(q^{n+1}) G^{n+1}$$

- \blacktriangleright γ chosen to be constant
- $\sigma = \sqrt{2\gamma M k_B T}$
- G sampled from a standard normal distribution

Velocity Rescaling - Instantaneous Temperature



Anderson - Instantaneous Temperature



Anderson - Histogram of Velocity Distribution



Langevin Instantaneous Temperature



Langevin - Histogram of Velocity Distribution



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Brief introduction to the thermostats

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