

Lecture 7 - Intro to sampling

Reminder: goal of stat mech
is to compute average observables

$$\langle O \rangle_{\text{ensemble}} = \int dx O(x) P(x)$$

If const NVT, $P(x) = e^{-\beta H(x)} / Z$

Some problems can solve exactly,
but most not. Will talk about
how to solve numerically

First: finish canonical ensemble

$$A = -k_B T \ln Z$$

$$P = -\partial A / \partial \nu \text{ etc}$$

Also know

$$E = - \frac{\partial \ln Z}{\partial \beta}$$

first finish ideal gas:

$$Q = \frac{1}{N!} \left(\frac{V}{\lambda^3} \right)^N = \frac{1}{N!} V^N (2\pi m k_B T)^{3/2 N}$$

$$= \beta^{-3/2 N} \cdot \text{stuff}$$

$$E = -\frac{\partial \ln Q}{\partial \beta} = \frac{3}{2} N \frac{\partial \ln \beta}{\partial \beta} = \frac{3}{2} N k_B T \quad \checkmark$$

one more thermodynamic quantity of importance!

$$C = \frac{\partial E}{\partial T} \quad C_V = \left(\frac{\partial E}{\partial T} \right)_V \quad C_P = \left(\frac{\partial E}{\partial T} \right)_P$$

$$\text{Ideal gas: } C_V = \frac{3}{2} N k_B$$

$$\begin{aligned} \text{In general, } C &= \frac{\partial E}{\partial T} = \frac{\partial E}{\partial \beta} \frac{\partial \beta}{\partial T} = \\ &= -\frac{1}{k_B T^2} \frac{\partial E}{\partial \beta} = -k_B \beta^2 \frac{\partial E}{\partial \beta} \end{aligned}$$

Now, for NVT

$$C_V = -k_B \beta^2 \frac{\partial \mathcal{E}}{\partial \beta} \quad \text{but} \quad \mathcal{E} = - \frac{\partial \ln Z}{\partial \beta}$$

$$= k_B \beta^2 \frac{\partial}{\partial \beta} \left[\frac{1}{Z} \frac{\partial Z}{\partial \beta} \right]$$

$$= k_B \beta^2 \left[-\frac{1}{Z^2} \frac{\partial Z}{\partial \beta} \frac{\partial Z}{\partial \beta} + \frac{1}{Z} \frac{\partial^2 Z}{\partial \beta^2} \right]$$

$$= k_B \beta^2 \left[\langle \mathcal{E}^2 \rangle - \langle \mathcal{E} \rangle^2 \right] \quad \star$$

$$= k_B \beta^2 \text{Var}(\mathcal{E})$$

$$Z = \int dx e^{-\beta \mathcal{H}(x)}, \quad \frac{\partial Z}{\partial \beta^2} = \int dx \mathcal{H}(x)^2 e^{-\beta \mathcal{H}(x)}$$

$$\frac{\partial \mathcal{E}}{\partial T} = \frac{1}{k_B T^2} \text{Var}(\mathcal{E}) \quad \star$$

Example of fluctuation dissipation

theorem: fluctuations @ ϵ_f
prop to how quantity responds to ϵ_f
(Onsager regression)

What does $\text{var}(\epsilon)$ mean physically?



Energy goes in & out of bath

How big is fluct relative to ϵ ?

$$\sigma_\epsilon / \epsilon = \sqrt{\delta^2 \epsilon} / \epsilon \propto \sqrt{C_V} / \epsilon \propto \frac{1}{N}! \quad \text{b/c } C_V \propto N$$

These energy fluctuations are what allow chemical systems to overcome barriers and undergo reactions. But how do we calculate properties if we can't explicitly get Z ?

Consider the problem of computing

$$\langle O \rangle = \int d\vec{x} O(\vec{x}) P(\vec{x})$$

first, let's consider that most observables may only depend on position

$$\langle O \rangle = \int dx^3 \int dp^{3N} O(\vec{x}) e^{-\beta H(\vec{x}, \vec{p})} / Z$$

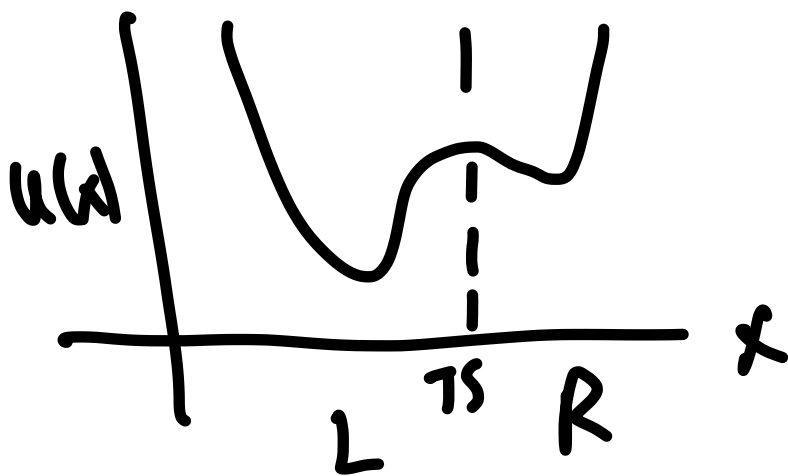
and $H = \sum p_i^2 / 2m + U(\vec{x})$

Then $\langle O \rangle = \frac{\text{const}}{\tilde{Z}} \int d\vec{x} e^{-\beta H(\vec{x})} O(\vec{x})$

Can define another pos. $\tilde{Z} = \int d\vec{x} e^{-\beta H(\vec{x})}$

$\langle O \rangle = \frac{\int d\vec{x} O(\vec{x}) e^{-\beta H(\vec{x})}}{\tilde{Z}}$ $\tilde{Z} = \text{const} \tilde{Z}$

Now consider 1d problem



Can compute $\langle O \rangle$ numerically
 "by quadrature" i.e.

$$\langle O \rangle \approx \sum_{j=1}^k O(x_j) P(x_j) \Delta x$$

Seems great, what's the problem

For each dimension:

$$\# \text{ points} = L/\Delta x$$

So for d dimensions

$$\# = (L/\Delta x)^d = e^{d \ln(L/\Delta x)}$$

exponentially large ($d=3N$ gets big fast)

Idea of sampling:

Generate representative cfs

s.t. $x_t \propto P(x_t)$ somehow

$$\text{Then } \langle O \rangle \approx \frac{1}{T} \sum_{t=1}^T O(x_t)$$

One way to do this is "Molecular

Dynamics", mimic Newton + temp

First discuss Monte Carlo

generate a "Markov Chain"

$x_t \rightarrow x_{t+1}$, depends only on
current state

If this satisfies "detailed balance"

$$P(x_t) P(x_t \rightarrow x_{t+1}) = P(x_{t+1}) P(x_{t+1} \rightarrow x_t)$$

then can be proven that

chain converges s.t. $x \sim P(x)$