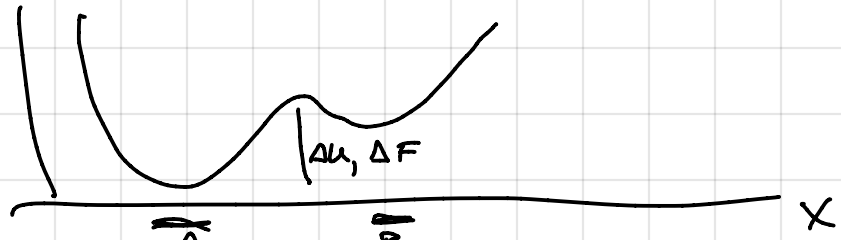


# Enhanced Sampling

Suppose  $U(x)$   
or  $F(x)$



$$F(x) = -kT \log \int \delta(M(\vec{q}) - \vec{x}) e^{-\beta U(\vec{q})} d\vec{q} - F_0$$

(Potential of mean force)

Rate  $A \rightarrow B \propto e^{-\beta \Delta U^\ddagger}$  or  $e^{-\beta \Delta F^\ddagger}$

So if  $1/\text{rate} \gg N\Delta t$ , then you will

be trapped in A (or B)

(rare event problem)

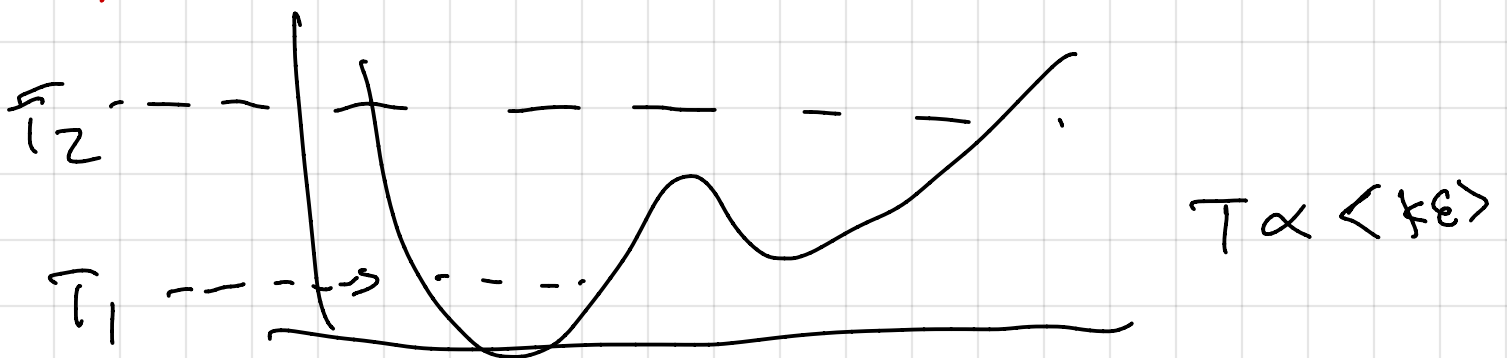


We need tricks to overcome this problem!

These are called enhanced or accelerated sampling methods, and generally estimate the (free) - energy diff between A & B too

( $\Delta A = -kT \log \cdot P_B/P_A$  & like equilibrium const)

**Idea!** increase temp  $\rightarrow$  rate faster



$$\langle A \rangle_{T_1} = \int dx P_1(x) A(x)$$

$$P_1(x) = \omega_1(x) / Z_1$$

eg  $\frac{e^{-u(x)/k_B T_1}}{\int dx e^{-u(x)/k_B T_1}}$

$$\langle A \rangle_{T_1} = \int dx A(x) \frac{\omega_1(x)}{Z_1} \cdot \frac{\omega_2(x)}{Z_2} \cdot \frac{\omega_2(x)}{Z_2}$$

$$= \int dx A(x) \frac{\omega_2(x)}{Z_2} \cdot \left( \frac{\omega_1(x)}{\omega_2(x)} \right) \cdot \frac{Z_2}{Z_1}$$

$$= \frac{Z_2}{Z_1} \cdot \int dx A(x) \frac{\omega_1(x)}{\omega_2(x)} \cdot \left[ \frac{\omega_2(x)}{Z_2} \right] \approx \frac{Z_2}{Z_1}$$

$$= \frac{Z_2}{Z_1} \langle A \omega_1 / \omega_2 \rangle_{T_2}$$

$$\text{for } N, U, T = \frac{Z_2}{Z_1} \langle A e^{-\frac{1}{k_B T_1} u(x)} e^{\frac{1}{k_B T_2} u(x)} \rangle_{T_2}$$

$$= \frac{Z_2}{Z_1} \langle A e^{-\frac{u(x)}{k_B} \left( \frac{1}{T_1} - \frac{1}{T_2} \right)} \rangle_{T_2}$$

$$\frac{Z_1}{Z_2} = \frac{\int dx \omega_1(x)}{Z_2} = \frac{\int dx \omega_1(x) \cdot \frac{\omega_2(x)}{\omega_2(x)}}{Z_2} = \langle \frac{\omega_1(x)}{\omega_2(x)} \rangle_{T_2}$$

Concisely:

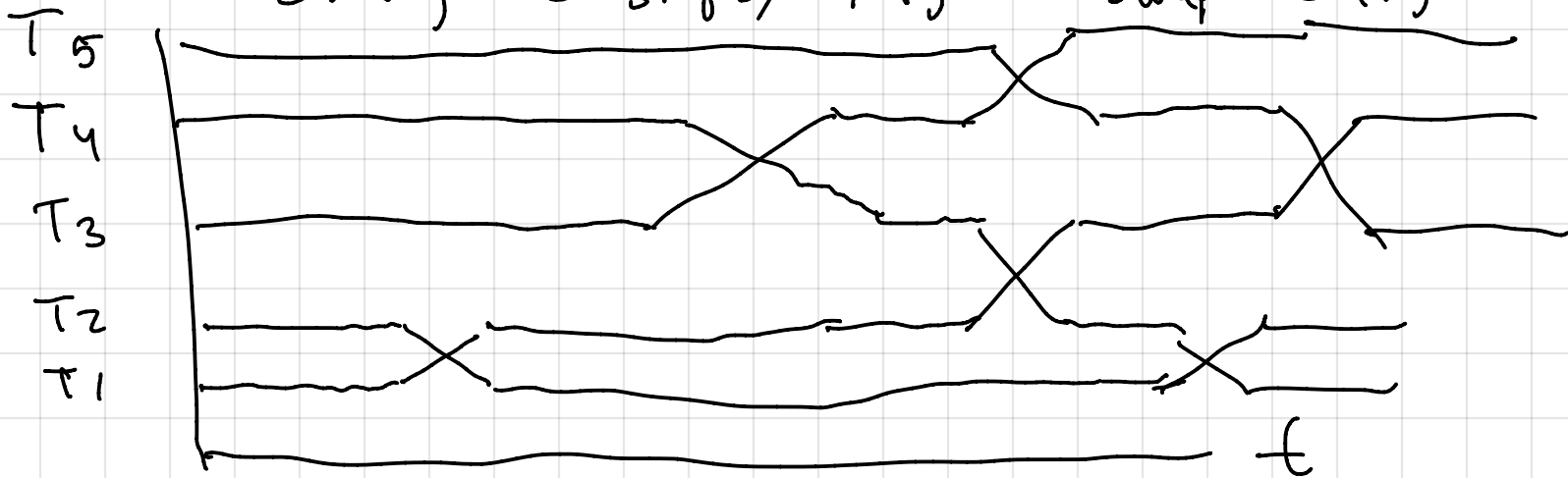
$$\langle A \rangle_{T_1} = \langle A \omega_1 / \omega_2 \rangle_{T_2} / \langle \omega_1 / \omega_2 \rangle_{T_2}$$

if  $T_2 \gg T_1$ , weights very small, numerical problem

Solution, run many sims @ diff T so  
 $\frac{1}{T_i} - \frac{1}{T_{i+1}}$  not too big

Replica exchange MD / Parallel tempering [didn't talk about const T yet?]

Every 2 steps, try to swap configs



Then sample can be heated up & cooled down to  $T_1$ , overcoming barriers but still sampling @  $T_1$

Exchange prob?  $P(A \rightarrow B)P(A) = P(B \rightarrow A)P(B)$

$$A = \{ \vec{x} @ T_x, \vec{y} @ T_h \}$$

$$B = \{ \vec{x} @ T_h, \vec{y} @ T_x \}$$

$$P(A \rightarrow B) = \min(1, \frac{P(B)}{P(A)}) = \min(1, \frac{e^{-u(x)/kT_h} e^{-u(y)/kT_c}}{e^{-u(x)/kT_c} e^{-u(y)/kT_h}})$$

$$\Rightarrow P(A \rightarrow B) = \min(1, e^{-\frac{u(x)}{k}(\frac{1}{T_h} - \frac{1}{T_c}) - \frac{u(y)}{k}(\frac{1}{T_c} - \frac{1}{T_h})})$$

$$= \min(1, e^{-\left[\frac{u(x) - u(y)}{k} \cdot (\frac{1}{T_h} - \frac{1}{T_c})\right]})$$

if  $T_c < T_h$ ,  $\frac{1}{T_h} - \frac{1}{T_c} < 0$  and  $u(x) - u(y) \stackrel{\text{prob}}{\ll} 0$

so swaps usually have prob  $< 1$

Now since Swaps satisfy detailed balance and MD or MC @ each temp satisfies detailed balance, have a chain of

$$X_i @ T_i \quad \text{st} \quad P(X_i) \rightarrow e^{-u(X_i)/k_B T_i} \quad \&$$

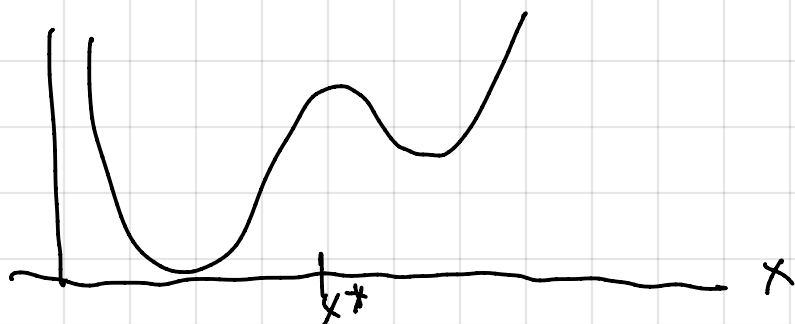
$$\langle A \rangle \sim \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{i=1}^N A(X_i^T)$$


---

Idea #2

Go back to this picture

$U_0(x)$



Idea: Torrie, Valleau 1977

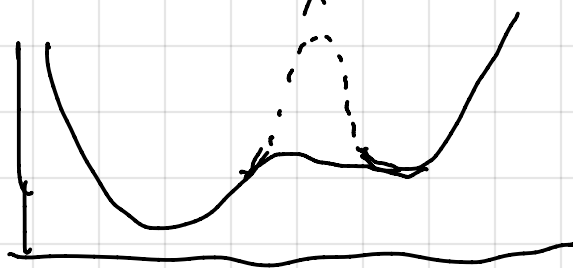
What if we add a potential to this to reduce the barrier and correct for the effect

Example

$U'(x)$



Then  $U(x) = U_0(x) + U'(x)$



This will give fast transitions from L to R

But how do we get  $\langle A \rangle = \int A(x) P_0(x)$  when

We are simulating with  $U_1$  and

hence  $P_1(x) = e^{-\beta U_1(x)} / Z_1 = w_1(x) / Z_1$

We actually did this before, perturbation theory

$$\langle A \rangle_0 = \frac{1}{Z_0} \int dx A(x) e^{-\beta u_0(x)}$$

like before, mult by  $e^{-\beta u_1(x)} / Z_1$  in top and

$$= \frac{1}{Z_0} \int dx A(x) e^{-\beta u_0(x)} \cdot \frac{e^{-\beta u_1(x)}}{e^{-\beta u_1(x)}} \cdot \frac{Z_1}{Z_1}$$

$$\begin{aligned} \exp(-\beta u_0(x)) / \exp(-\beta u_1(x)) &= \exp(-\beta u_0(x) + \beta u_1(x)) \\ &= \exp(-\beta u_0(x) + \beta(u_0(x) + u'(x))) \end{aligned}$$

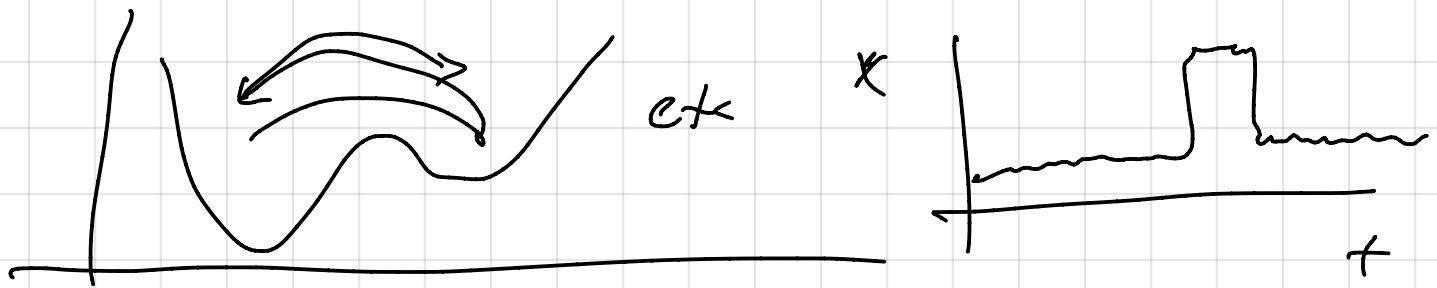
$$= Z_1 / Z_0 \int dx A(x) e^{+\beta u'(x)} \cdot \frac{e^{-\beta u_0(x)}}{Z_1}$$

$$= \langle A e^{+\beta u'(x)} \rangle_1 / \langle e^{+\beta u'(x)} \rangle_1 = \frac{\langle A / w \rangle_1}{\langle 1 / w \rangle_1}$$

What is this weight doing? correcting for energy  
time there is something near  $x^*$  it should  
have less weight

This works well for a simple 1d barrier  
crossing, but it does not necessarily  
make the simulation averages converge very  
fast.

I said for the average to be computed well, you should see every state multiple times



Now, diffusion

Diffusion  $\langle s_x^2 \rangle \propto Dt$

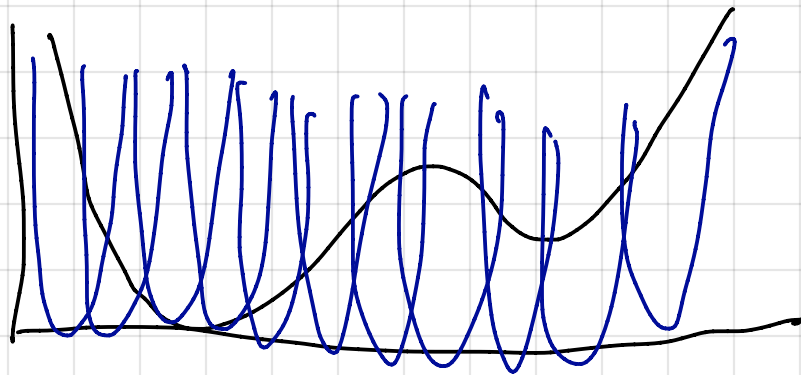
so to explore whole space,  $t \sim L^2/D$

if  $D$  is small or space is big will take a long time

Later idea: Umbrella Sampling (Umbrella, cover all the space)

Many biased simulations with different  $w'$  to make sure all space is covered

So Eq:



Common to use  $U'_i(x) = \frac{1}{2}k(x-x_i)^2$

if instead  $U'_i(x) = \begin{cases} 0 & |x-x_i| < a \\ \infty & \text{boxes} \end{cases}$

Potential  $U_c(x) = U_0(x) + U'_i(x)$

$= U_0(x)$  inside box

Now if  $U_c(x)$  is actually PMF we don't know, to get that we need to compute

$$A_i = -k_B T \log \left[ \frac{\int dx \chi(x) e^{-\beta U_0(x)}}{\int dx e^{-\beta U_0(x)}} \right]$$

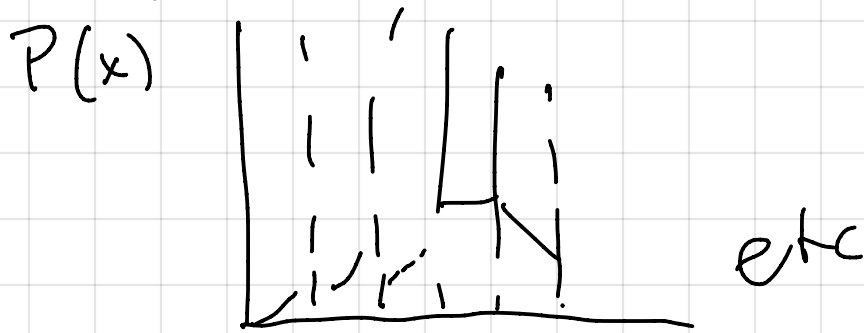
where  $\chi = \begin{cases} 1 & |x-x_i| < a \\ 0 & \text{otherwise} \end{cases}$

can do this with

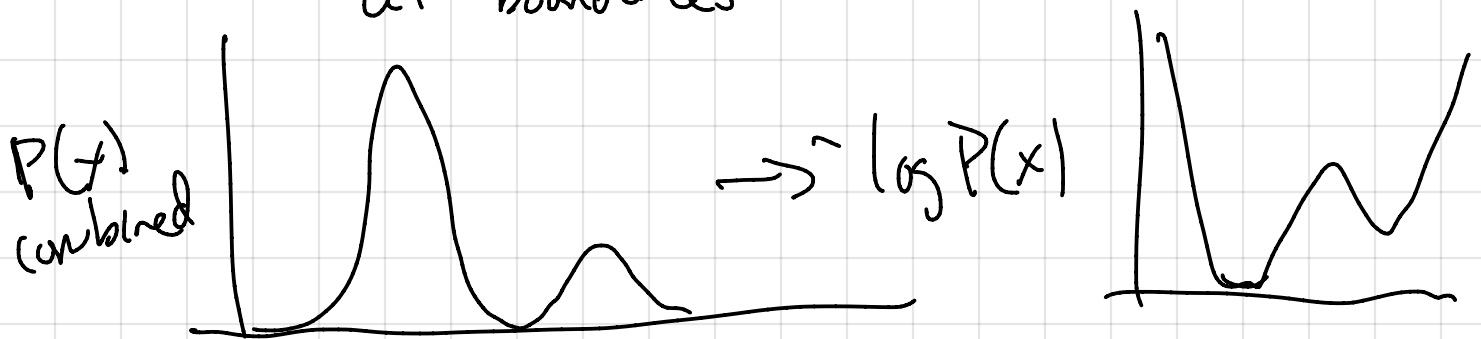
$$\langle \chi(x)/w_i(x) \rangle_i / \langle 1/w_i(x) \rangle_i$$



will get something like



don't know offsets but must be equal  
at boundaries



Methods exist to combine samples with  
other biasing functions  
Next: Use this for phase transitions