#### 11/07/2022

- Today's lecture is a key step towards the following learning objective: Explain key concepts related to binding free energy calculations. Compare and contrast molecular docking and binding free energy calculations.
- Simulating and analyzing thermodynamic processes

## Simulating and analyzing thermodynamic processes

- The lecture will cover
  - importance sampling
  - umbrella sampling
  - thermodynamic processes
  - replica exchange
  - statistical estimators for thermodynamic properties
- At the end of this module, you should be able to answer the following questions:
  - What is a thermodynamic process?
  - What are the benefits of umbrella sampling?
  - What are the benefits of performing replica exchange?
  - How do you know if there are enough states along a thermodynamic process?
- You should be able to calculate
  - expectation values
  - free energy differences between thermodynamic states

## Importance Sampling

## Importance sampling

Sampling from one distribution and estimating quantities in another

$$\langle A \rangle_T = \int A(x) p_T(x) dx = \int A(x) \left( \frac{p_T(x)}{p_S(x)} \right) p_S(x) dx = \langle wA \rangle_S$$

- $p_T(x)$  is the probability density in the target distribution
- $p_S(x)$  is the probability density in the sampled distribution

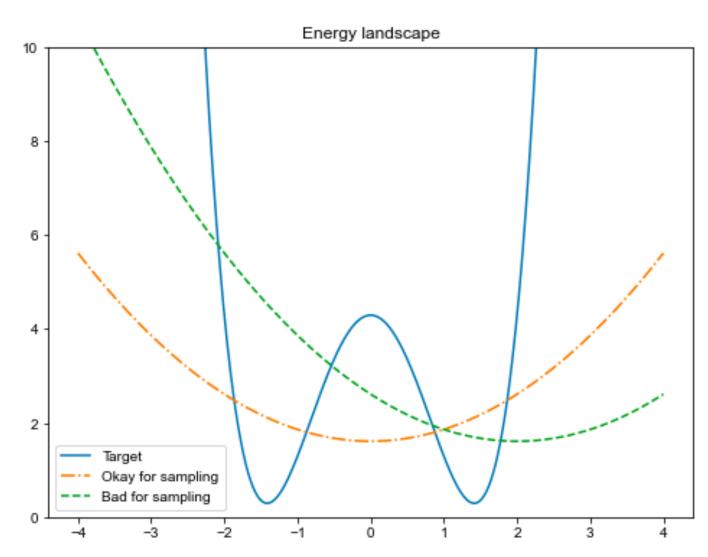
• 
$$w = \left(\frac{p_T}{p_S}\right)$$
 is the ratio of weights in the two distributions

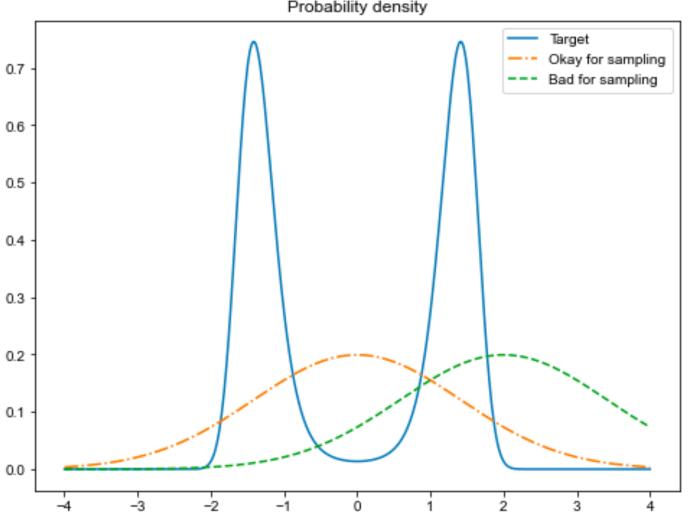
## Why do importance sampling?

- Use one simulation to estimate quantities in multiple thermodynamic states,
   e.g. different temperatures
- Less computational expense to sample from one distribution, e.g. sample with molecular mechanics and calculate quantities in QM/MM distribution
- Sample from distribution with smaller configuration space, e.g. harmonic restraint towards a crystal structure

## Caveat: Importance sampling

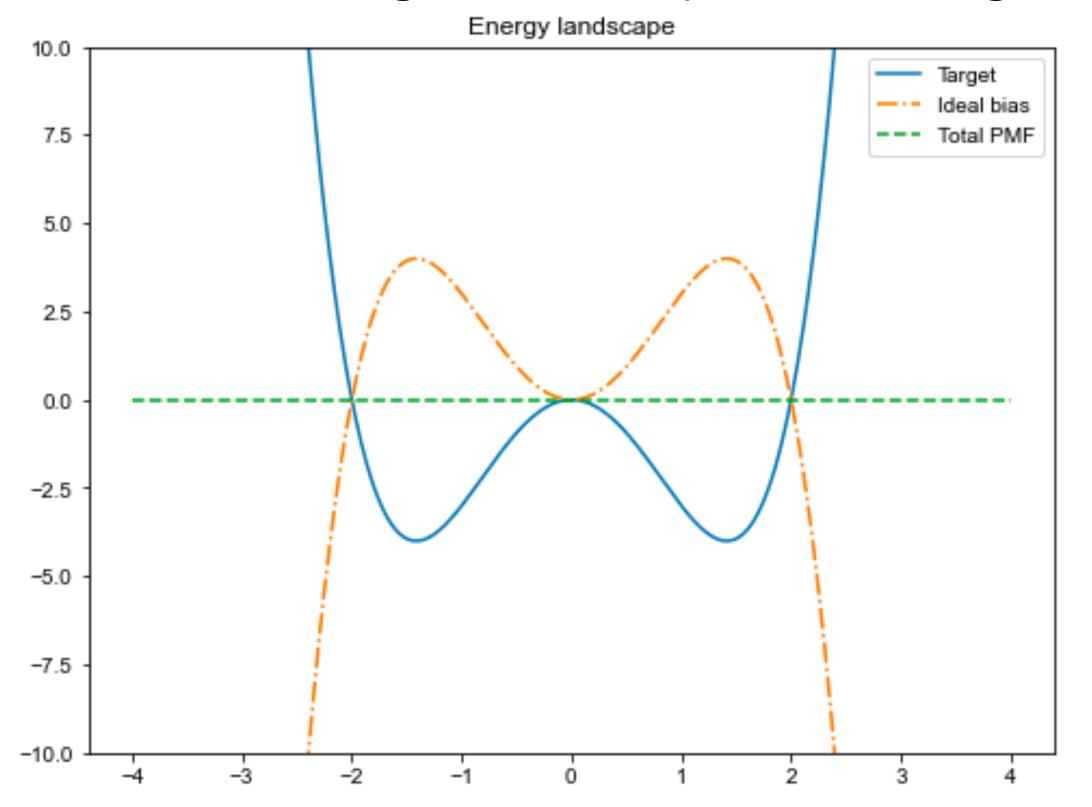
- The target and sampled distribution should be similar
  - If the *support* differs, important configurations may be missed
  - If the probability density significantly differs
    - the reweighing term can be noisy
    - thermodynamic expectations will require more samples to converge





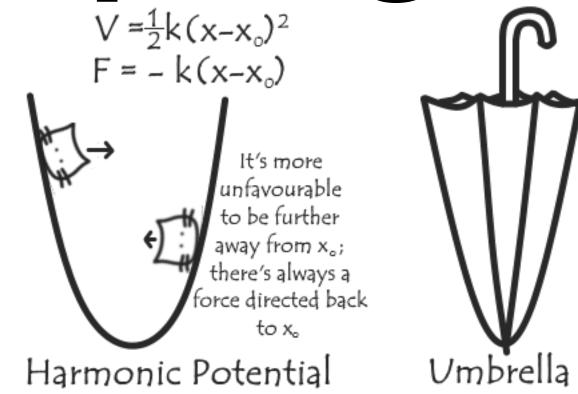
## Idealized importance sampling

Covers the entire relevant configuration space along an order parameter

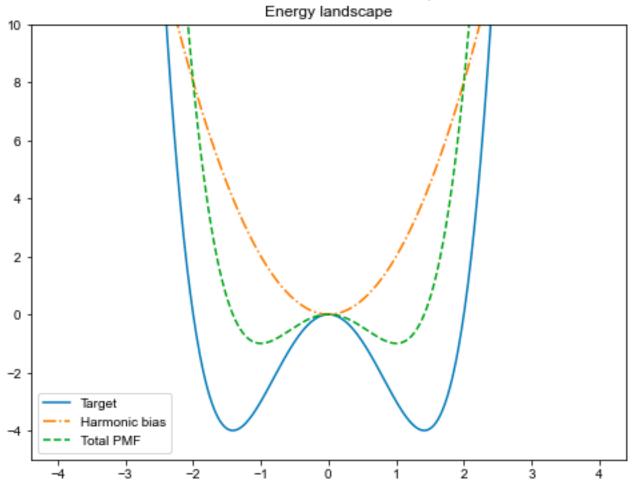


## Harmonic umbrella sampling

- Sampling that covers the entire relevant configuration space along an order parameter
- Typically involves
  - a harmonic restraint towards a specific value of the order parameter,  $U_b(x) = \frac{1}{2}k(z[x] z_o)^2$ 
    - k is the spring constant
    - z[x] is the value of the order parameter for configuration x
  - multiple simulations with different spring centers
    - the observed probability distributions should overlap



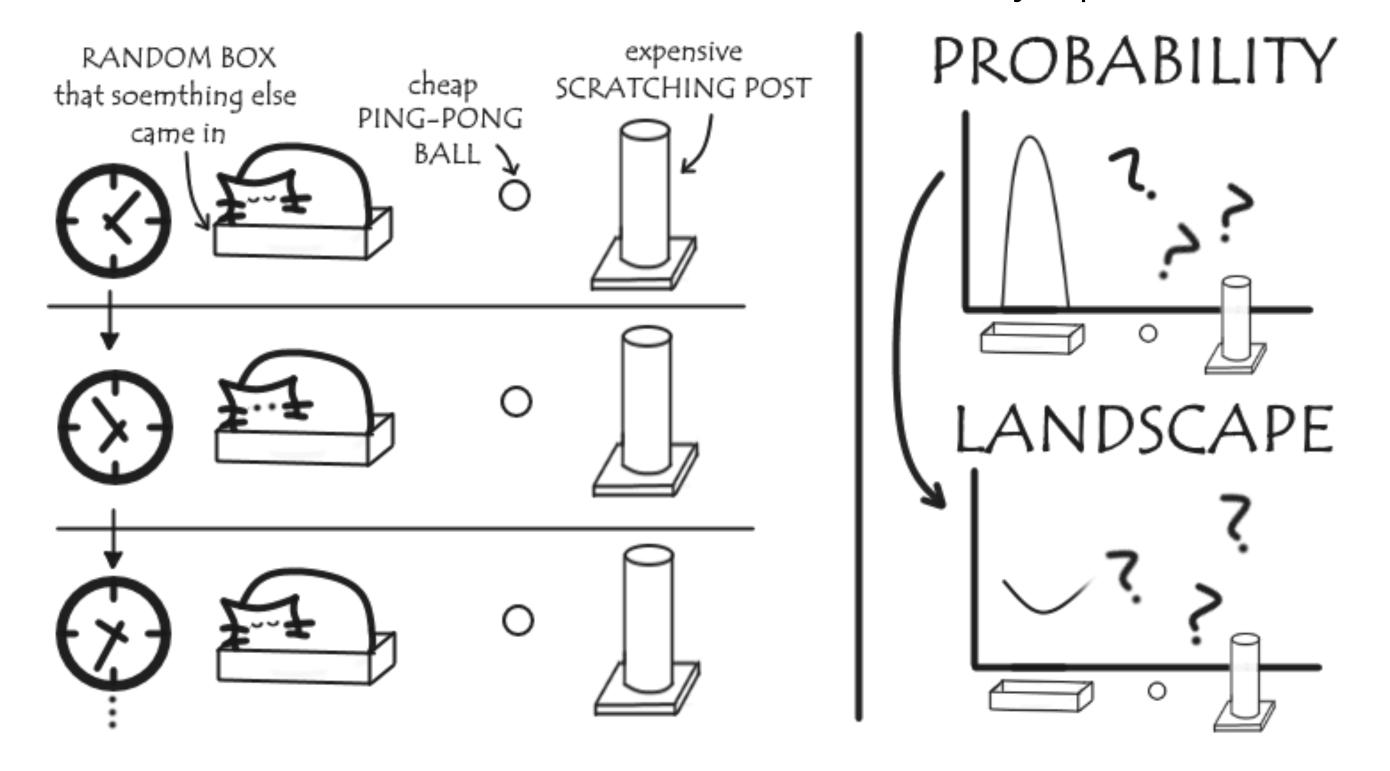
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<a href="http://fiona-naughton.github.io/blog/2016/05/25/">What-is-this-MD-thing-anyway</a>



## A cat analogy of umbrella sampling

Which toy does the cat like best?

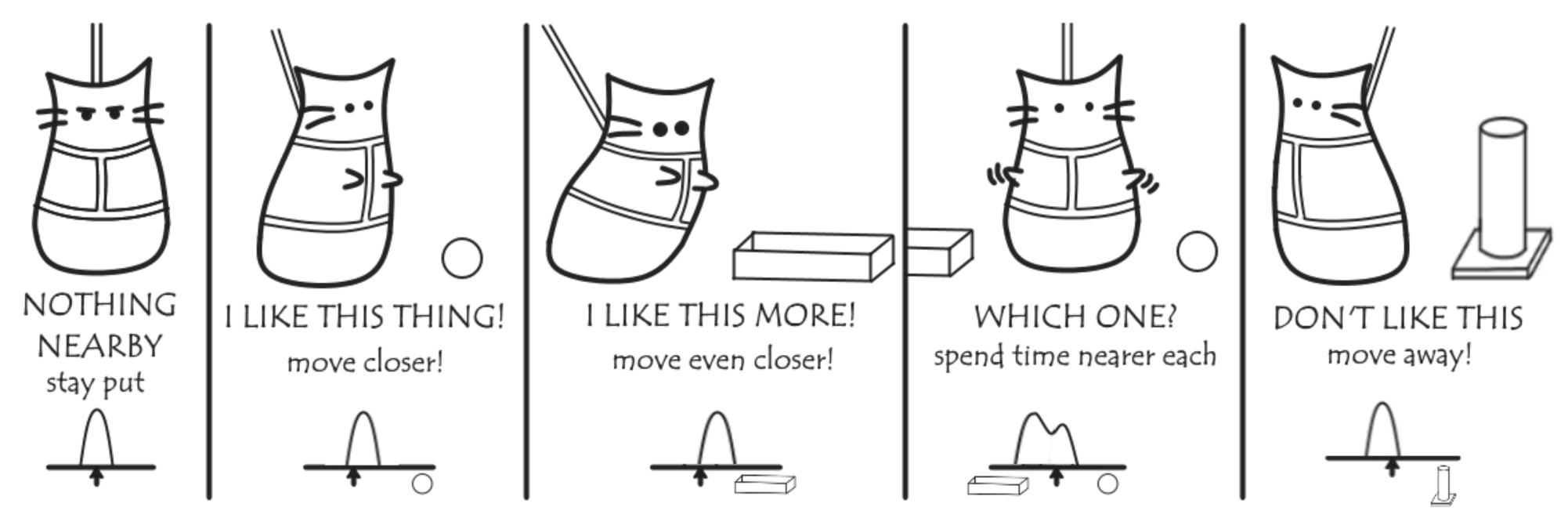
To quantify this, we can look at the fraction of time they spend near a specific toy.



But we can be watching the cat for a long time...

## Umbrella sampling with a cat

If we put a restraint on the cat, we can determine what they prefer in a smaller area. Overall, we don't need to watch as long.



Used with permission from Fiona Naughton: <a href="http://fiona-naughton.github.io/blog/2016/05/25/What-is-this-MD-thing-anyway">http://fiona-naughton.github.io/blog/2016/05/25/What-is-this-MD-thing-anyway</a>

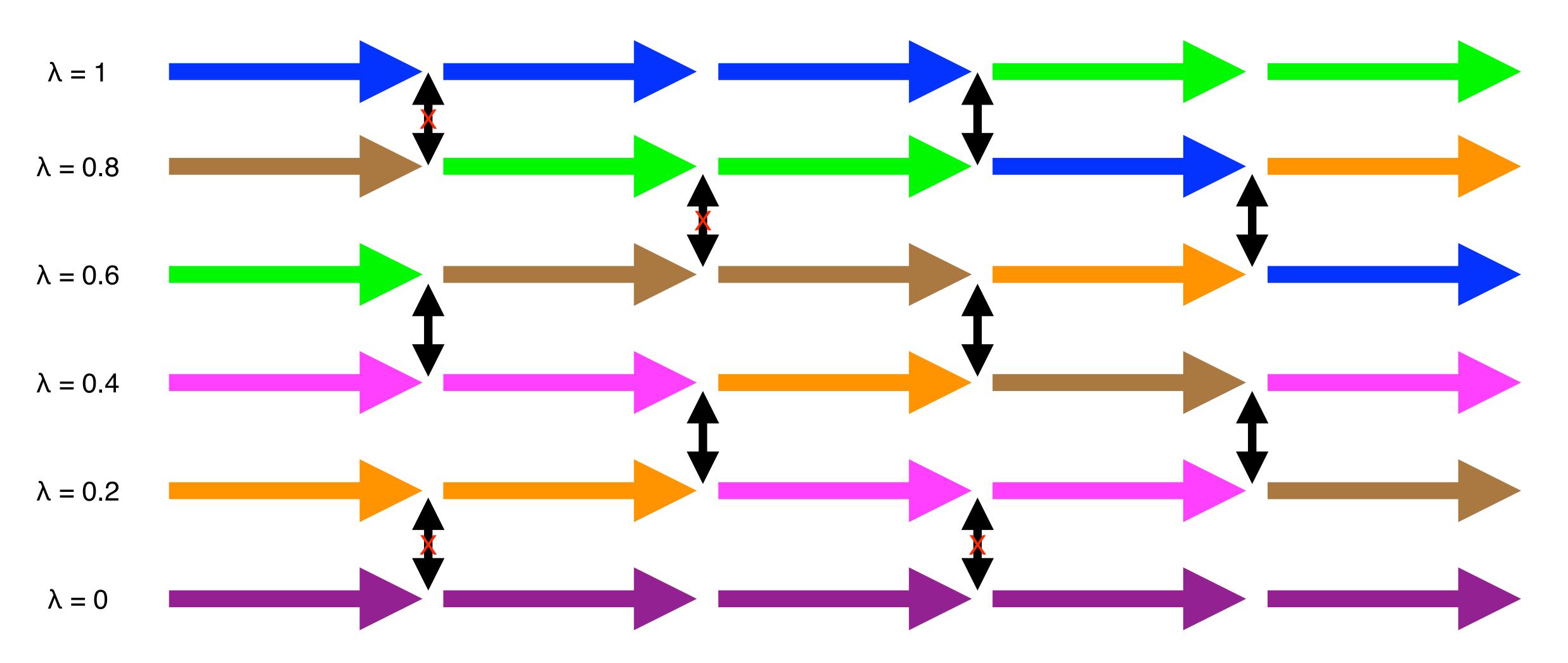
## Thermodynamic processes

- A thermodynamic process involves a change in one or more variables that specify a thermodynamic state
- In general and physical chemistry, you learn about specifying a state with
  - temperature
  - pressure/volume
  - number of particles/chemical potential
- Processes include
  - isothermal expansion a change in volume but not temperature
  - adiabatic expansion a change in volume without heat transfer in and out of a system
- In molecular simulations, additional variables can define the potential energy
  - harmonic spring constant and center (umbrella sampling)
  - alchemical parameter for
    - transforming one substituent into another
    - decoupling a molecule from solvent

## Replica exchange

- Simulations of multiple thermodynamic states with different parameters
  - originally, variation in temperature
  - often used for alchemical coupling
- Configurations from thermodynamic states are periodically swapped
  - Equivalently, thermodynamic parameters are swapped
  - Swapping satisfies detailed balance; both states sample from respective Boltzmann distributions

## Swapping

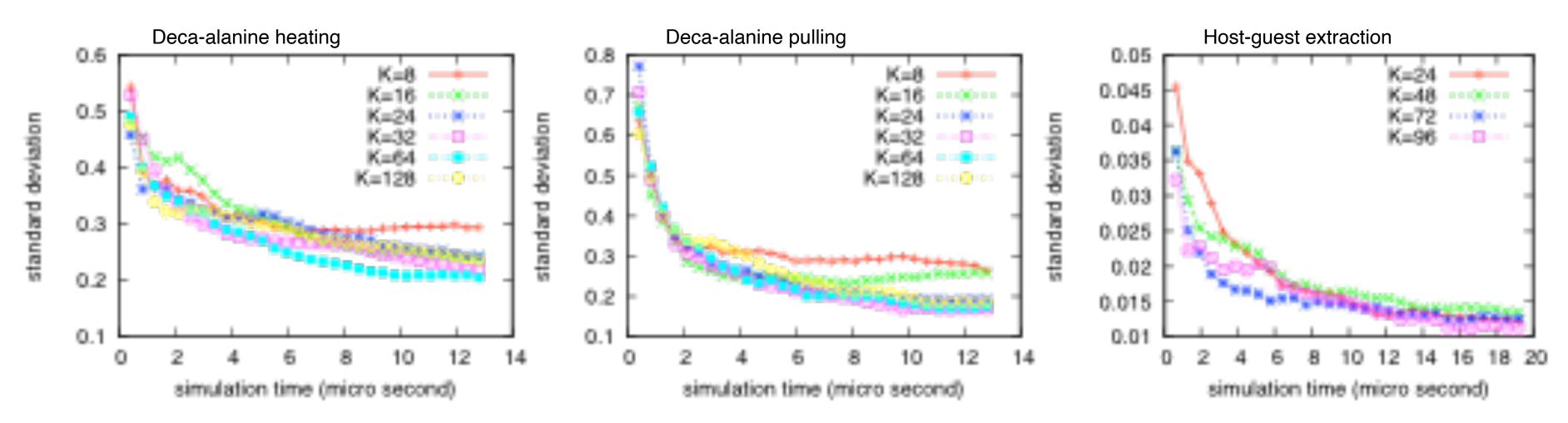


## Why perform replica exchange?

- Improve mixing of MCMC chains; higher-entropy states help sample configurations in lower-entropy states
  - At high temperature, energetic barriers are crossed more quickly than at low temperature
  - Decoupled ligands move freely compared to bound ligands. See <a href="http://mypages.iit.edu/~dminh/images/pubs/dock\_1hnn.gif">http://mypages.iit.edu/~dminh/images/pubs/dock\_1hnn.gif</a>.
- Minimal added computational expense

## Selecting thermodynamic states

- Thermodynamic state selection has been thought to be a tricky optimization problem
- In Nguyen and Minh (2016) [1]
  - processes were simulated 100x each for each number of states, K
  - the standard deviation of the free energy was evaluated as a function of the total simulation time



- If there are not enough states, the convergence curve levels off
- If there are enough states, the standard deviation of free energy estimates depends on the aggregate simulation time and is insensitive to the number of states.

#### Review

- What is a thermodynamic process?
- What are the benefits of umbrella sampling?
- What are the benefits of performing replica exchange?
- How do you know if there are enough states along a thermodynamic process for replica exchange?

# Estimating thermodynamic properties

#### **Estimating Thermodynamic Properties**

- Molecular simulations are used to calculate thermodynamic and kinetic properties
- In general, the thermodynamic properties are
  - expectation values of an observable, including
    - probability of the observable having a certain range of values
    - potential of mean force with respect to the observable
  - free energy differences between thermodynamic states
    - in biomolecular systems,  $\Delta G \sim \Delta A$

#### What is $\Delta G$ ?

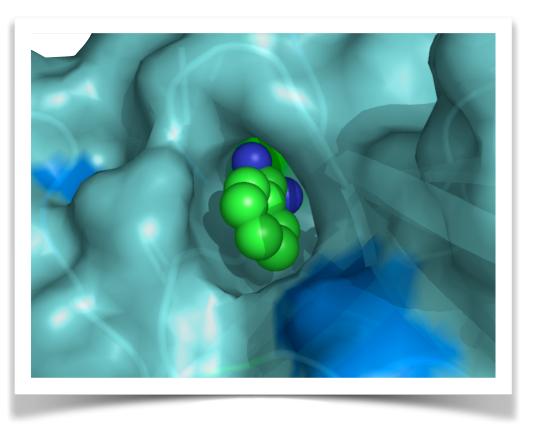
- $\Delta U$  is the change in average internal energy.
  - internal energy can be computed for individual structures
  - in biomolecular simulations, internal energy is modeled by the molecular mechanics force field
- $\Delta H = \Delta U + \Delta(pV)$  is the change in enthalpy
  - in biomolecular simulations, change in pV is usually negligible
- ΔG is the Gibbs free energy
  - at constant pressure and temperature, dictates
    - spontaneity and
    - equilibrium constant of process
  - in biomolecular simulation, interest in free energy differences between
    - conformations of a macromolecule
    - thermodynamic states with different temperature, pressure, volume, or other parameters
  - $\Delta G = \Delta H + T\Delta S$ , but  $\Delta S$  is very challenging to compute

#### What is AA?

- ΔA is the Helmholtz free energy
  - at constant volume and temperature, dictates
    - spontaneity and
    - equilibrium constant of process
  - in biomolecular simulation,  $\Delta A$  and  $\Delta G$  are usually assumed to be equal

#### How are free energy calculations useful?

- Noncovalent binding between molecules (see [2])
  - Design molecules to manipulate protein function
  - Recognize toxins
  - Identify enzyme functions
  - Protein design: design binders to target molecule
  - Aid medicinal chemistry, guide synthesis
- Hydration free energies
  - Part of binding free energy & solubility
- Conformational free energies relevant to
  - biological mechanism
  - binding free energy



$$P + L \stackrel{\Delta G^0}{\longrightarrow} PL$$
 (from [1])

## How are ΔG/ΔA calculated from molecular simulations?

#### **Basic Statistical Mechanics**

• In the Boltzmann distribution, the probability of a configuration  $r^N$  with energy  $U_{\mathfrak{s}}(r^N)$  is,

$$\pi_s(r^N) \propto \exp\left[-\beta U_s(r^N)\right]$$
 (unnormalized) 
$$\rho_s(r^N) = \exp\left[-\beta U_s(r^N)\right]/Q_s \text{ (normalized)}$$

• A partition function is the normalizing constant of the Boltzmann distribution

$$Q_{S} = \int \pi_{S}(r^{N}) dr^{N}$$

• The free energy difference is related to a ratio of partition functions

$$\beta(A_1 - A_0) = -\ln\left(\frac{Q_0}{Q_1}\right)$$

## The Zwanzig Relation: Derivation

- From before,  $\beta(A_1-A_0)=-\ln\left(\frac{Q_0}{Q_1}\right)$ .
- Substituting in partition functions,  $\beta(A_1-A_0)=-\ln\left(\frac{\int e^{-\beta U_1(r^n)}dr^N}{\int e^{-\beta U_0(r^N)}dr^N}\right)$ .
- $\text{Multiplying by one, } \beta(A_1-A_0) = -\ln\left(\frac{\int e^{-\beta U_1(r^N)+\beta U_0(r^N)-\beta U_0(r^N)}dr^N}{\int e^{-\beta U_0(r^N)}dr^N}\right).$
- Defining the potential energy difference  $\Delta U(r^N) = U_1(r^N) U_0(r^N)$ ,  $\int \left\{ e^{-\beta \Delta U(r^N)} e^{-\beta U_0(r^N)} dr^N \right\}$

$$\beta(A_1 - A_0) = -\ln\left(\frac{\int e^{-\beta \Delta U(r^N)} e^{-\beta U_0(r^N)} dr^N}{\int e^{-\beta U_0(r^N)} dr^N}\right)$$

## The Zwanzig Relation: In Practice

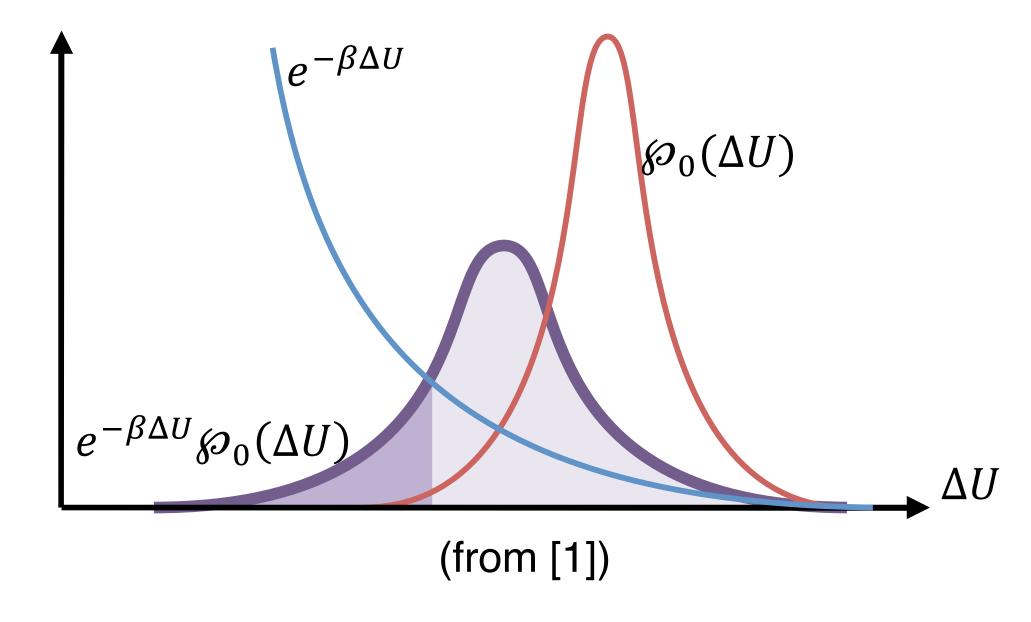
- . Using the definition of  $\rho_{s}(r^{N})$ ,  $\beta(A_{1}-A_{0})=-\ln\int\rho_{0}(r^{N})e^{-\beta\Delta U(r^{N})}dr^{N}$ .
- The Zwanzig relation [3] is
  - $\beta(A_1-A_0)=-\ln\left\langle e^{-\beta\Delta U}\right\rangle_0$  in a simpler notation.
  - $\beta(A_1-A_0)=-\ln\left\langle e^{\beta\Delta U}\right\rangle_1$  can be derived with similar steps
- This shows us that
  - The free energy difference can be computed based on an average over configurations taken from one of the states of interest
  - We can generate these configurations with MC or MD
  - The free energy comes from evaluating the energies of these configurations in both potentials  $U_0$  and  $U_1$ , and taking an appropriate average of the energy difference

## The Zwanzig Relation: Limitations

• In terms of an integral over the distribution of  $\Delta U$  (instead of over  $\rho_o(r^N)$ ) the Zwanzig relation is,

$$\beta(A_1 - A_0) = -\ln \int e^{-\beta \Delta U} \rho_0(\Delta U) d\Delta U.$$

- Sampling is from the red curve
- Accurate estimation requires the purple curve
- The calculation will not be accurate if  $U_0$  and  $U_1$  are very different!
- Potential energies will be different if the states access different parts of configuration space



## Other ways to calculate $\Delta G$

- Multistate Bennett Acceptance Ratio (MBAR) [5]
  - estimates free energies and thermodynamic expectations from a series of states
  - extension of Bennett Acceptance Ratio (BAR) [4], which uses data from two states
  - Proven to be statistically optimal
- Thermodynamic integration is based on the fundamental theorem of calculus, integrating one the derivative of the free energy with respect to a parameter
- <u>All</u> of the methods require thermodynamic states with configuration space overlap, meaning that
  - similar configurations have similar energies
  - the most relevant configuration space is similar

#### References

- [1] Nguyen, T. H.; Minh, D. D. L. Intermediate Thermodynamic States Contribute Equally to Free Energy Convergence: A Demonstration with Replica Exchange. Journal of Chemical Theory and Computation 2016, 12 (5), 2154–2161. https://doi.org/10.1021/acs.jctc.6b00060.
- [2] Many parts of this module were adapted from a lecture by David Mobley (<a href="https://github.com/MobleyLab/drug-computing/tree/master/uci-pharmsci/lectures/">https://github.com/MobleyLab/drug-computing/tree/master/uci-pharmsci/lectures/</a> free energy basics) under the CC BY 4.0 license. The lecture is part of the Drug Discovery Computing Techniques course (PharmSci 175/275) at UC Irvine.
- [3] Zwanzig, R. High-Temperature Equation of State by a Perturbation Method. I. Nonpolar Gases. Journal of Chemical Physics 1954, 22 (8), 1420.
- [4] Bennett, C. H. Efficient Estimation of Free-Energy Differences from Monte Carlo Data. Journal of Computational Physics 1976, 22 (2), 245–268.
- [5] Shirts, M. R.; Chodera, J. D. Statistically Optimal Analysis of Samples from Multiple Equilibrium States. Journal of Chemical Physics 2008, 129 (12), 124105.