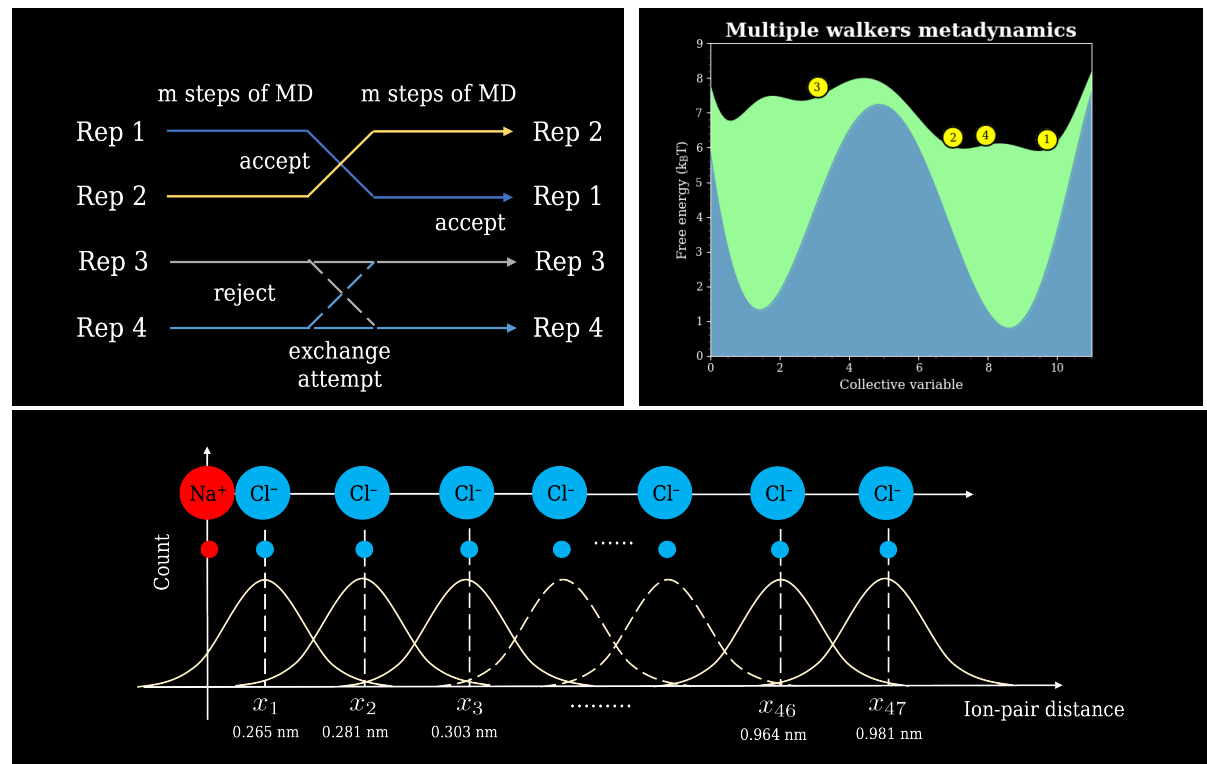


Enhanced sampling molecular dynamics on HPC systems: Ensembles of Simulations

Prof. Michael Shirts
Dr. Wei-Tse Hsu
CU Boulder



i-Co  SE

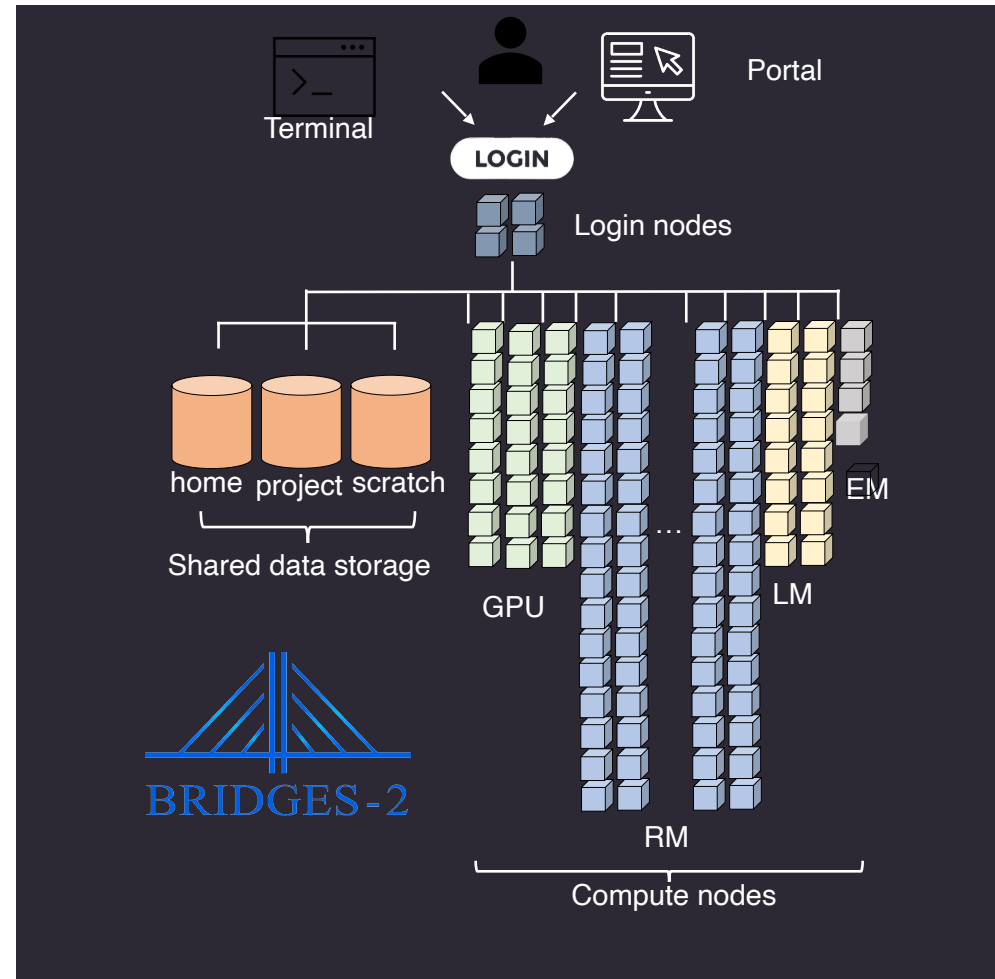
High-performance computing (HPC) systems are powerful for complex and resource-intensive tasks

- Interactive session: `interact -N 1 --ntasks-per-node=128`
`-p RM -p time 00:30:00`
- SLURM job submission script

```
#!/bin/sh
#SBATCH --job-name MD_simulation
#SBATCH -N 1
#SBATCH -p RM
#SBATCH -t 2:00:00
#SBATCH --ntasks-per-node=64
```

```
module load gromacs
module load openmpi/3.1.6-gcc10.2.0
```

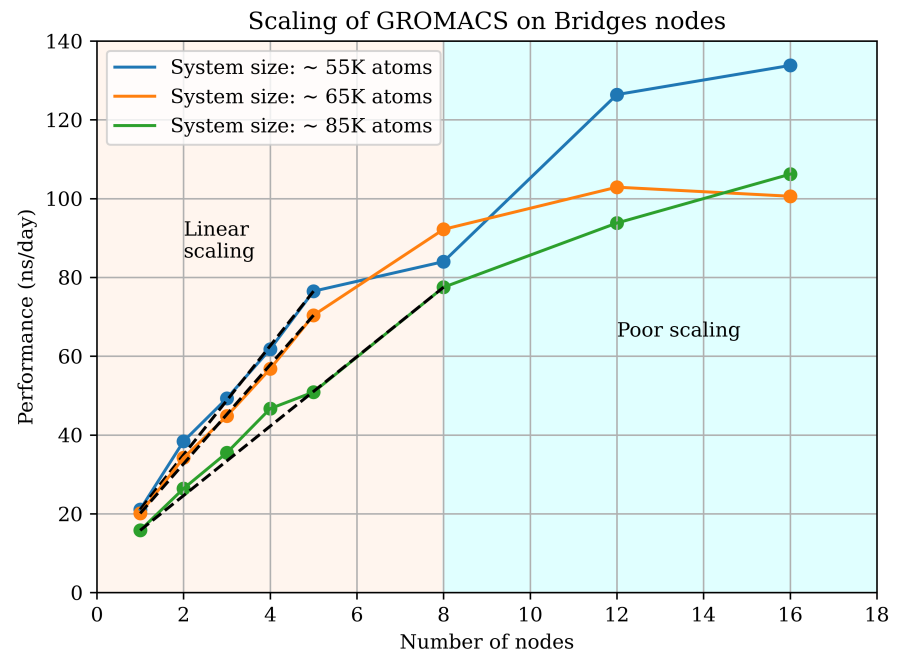
```
mpirun -np 16 gmx_mpi mdrun -deffnm md
```



It's important to make the best use of compute resources when running parallel simulations

- Parallelization can be performed across
 - Multiple cores in a single node using multiple threads (e.g. OpenMP thread)
 - Multiple nodes using multiple MPI processes by OpenMPI (launched by mpirun or mpiexec)

Communications take time!

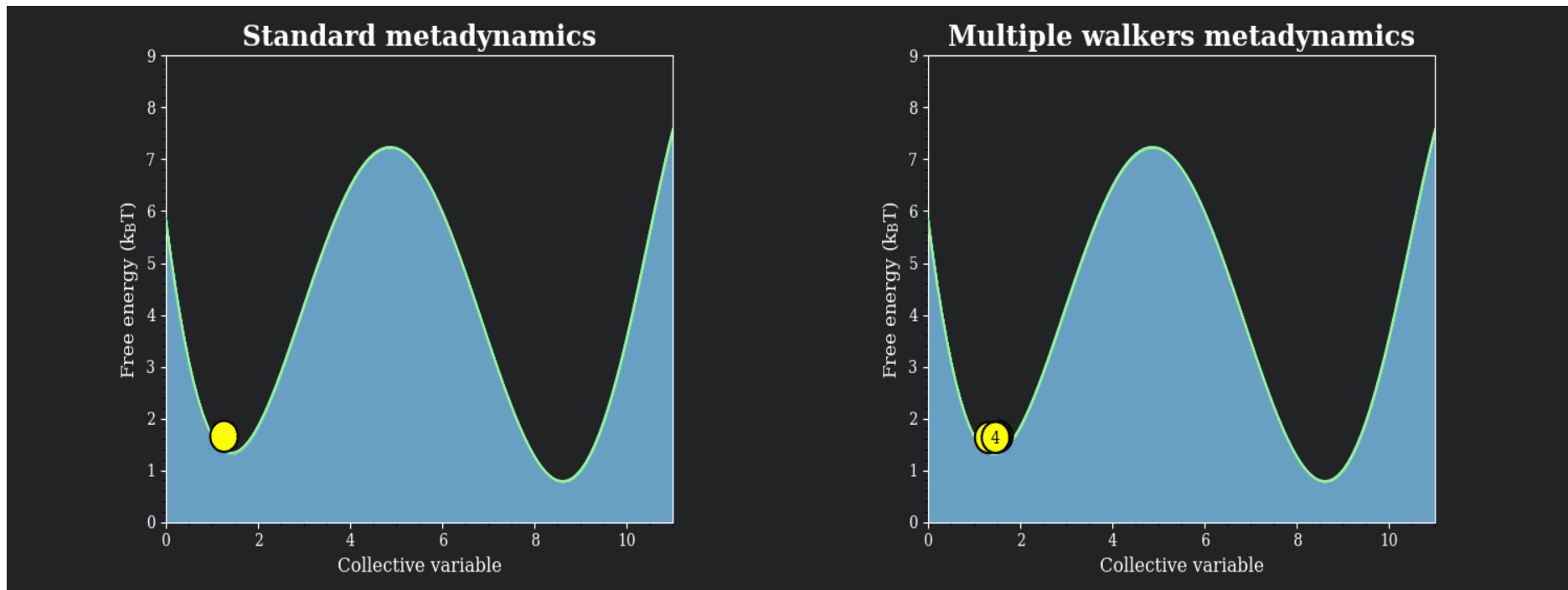


Hands-on exercise 1

Running uncoupled MD simulations in parallel

Metadynamics converge can be sped up by running multiple walkers in the same energy surface

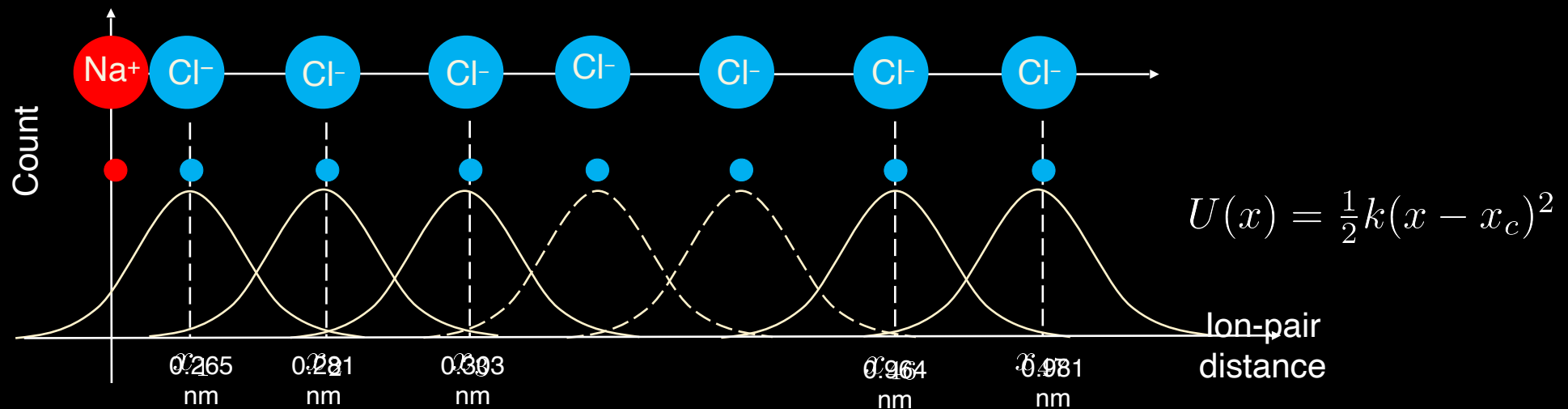
- In multiple walkers metadynamics, the deposited bias is shared between replicas so each replica is aware of the whole history of the bias.



Hands-on exercise 2

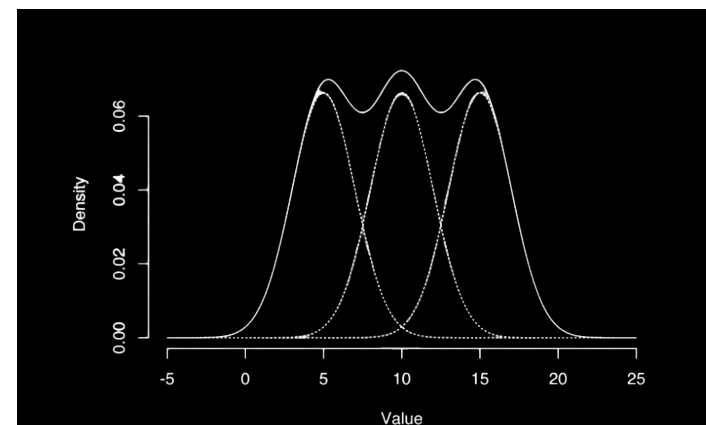
Umbrella sampling runs simulations at different values of some CV

- Pulling simulation: Fix Na^+ and pull Cl^- away to generate a range of initial locations
- Production runs: Fix the whole ion-pair and perform the simulations under harmonic restraint.
- A reasonable force constant should ensure overlap between states.



We can use MBAR to recover the free energy profile along the CV of interest from umbrella sampling

- Possible methods to reconstruct the free energy: WHAM, TI, BAR, MBAR and even more!
- Protocol of using MBAR
 - Decorrelate the time series of the collective variables.
 - Reweight from the mixture distribution of all the biases
 - Compute the free energy of each biased simulation
 - Approximate the free energy surface

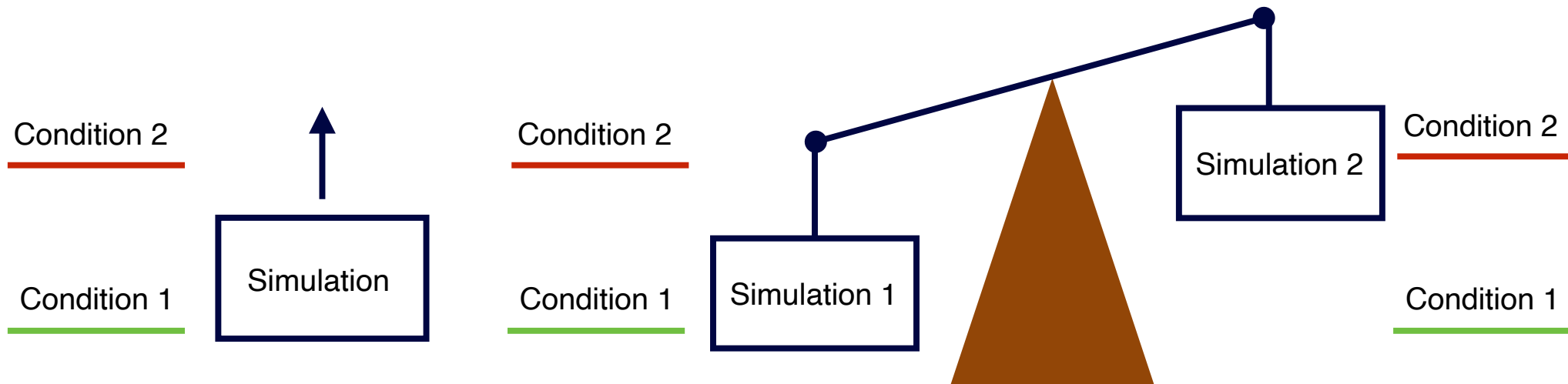


Hands-on exercise 3

Perform and analyze umbrella sampling

Concept of Replica Exchange

- We have an ensemble of simulations, each a different condition (temperature, bias potential)
- Rather than changing conditions independently, we allow them to change in a coupled way.



$$\frac{p_1}{p_2} = \frac{p_{A=1}(x_A)p_{B=2}(x_B)}{p_{A=2}(x_A)p_{B=1}(x_B)}$$

- If we were to switch the configurations, what is the total change in probability?

Considerations of Replica Exchange

- Advantages: Don't have to consider the free energies

$$\begin{aligned}
 \frac{p_1}{p_2} &= \frac{p_{A=1}(x_A)p_{B=2}(x_B)}{p_{A=2}(x_A)p_{B=1}(x_B)} \\
 &= \frac{e^{\cancel{\beta_1 F_1} - \beta_1 U_1(x_A)} e^{\cancel{\beta_2 F_2} - \beta_2 U_2(x_B)}}{e^{\cancel{\beta_1 F_1} - \beta_1 U_1(x_B)} e^{\cancel{\beta_2 F_2} - \beta_2 U_2(x_A)}} \\
 &= \frac{e^{-\beta_1 U_1(x_A)} e^{-\beta_2 U_2(x_B)}}{e^{-\beta_1 U_1(x_B)} e^{-\beta_2 U_2(x_A)}}
 \end{aligned}$$

- Disadvantage:
 - Simulations need to run in lockstep
 - They need to exchange frequently to be useful
 - It gets complicated to exchange when there are many, including multiple dimensions

Temperature replica exchange periodically exchange simulation coordinates at different temperatures

- The sampling alternates between the configurational and temperature space.
- At higher temperature, it is easier to cross the free energy barrier.
- Useful for estimating free energy or heat capacity as a function of temperature

Transition probability:

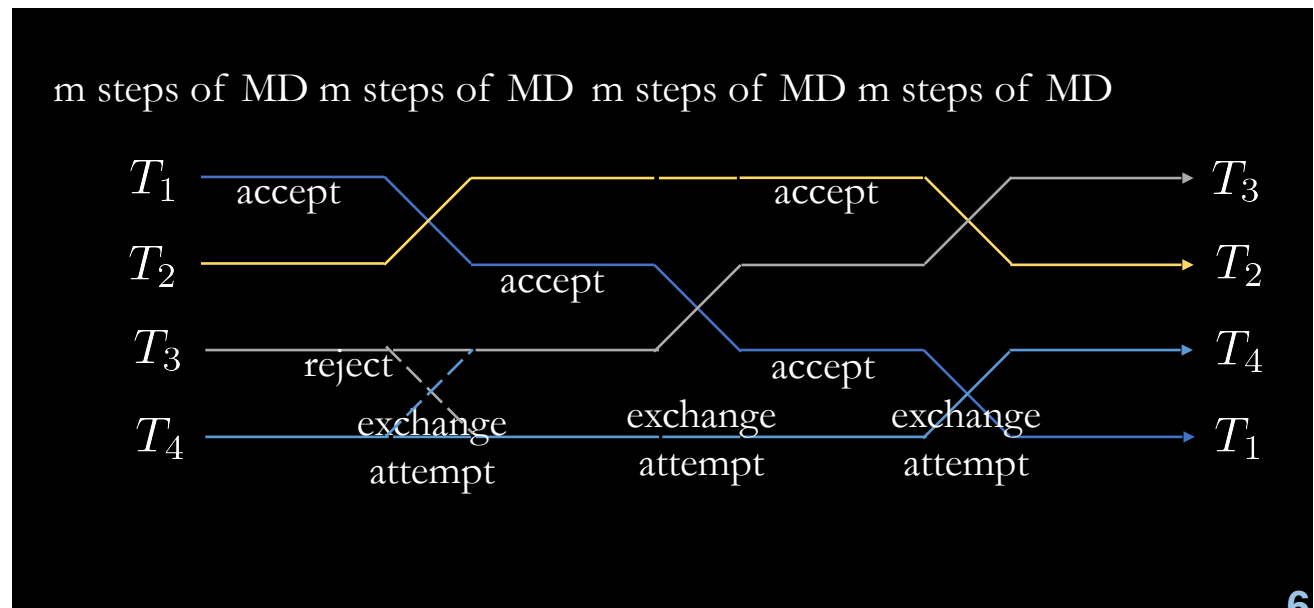
If $\Delta > 1$: Move

If $\Delta < 1$: move with prob Δ

$$\Delta = \frac{p_{T_1}(x_A)p_{T_2}(x_B)}{p_{T_1}(x_B)p_{T_2}(x_A)}$$

$$= \frac{e^{-\beta_1 U(x_A)} e^{-\beta_2 U(x_B)}}{e^{-\beta_2 U(x_A)} e^{-\beta_1 U(x_B)}}$$

$$= e^{(\beta_2 - \beta_1)U(x_A) - (\beta_1 - \beta_2)U(x_B)}$$



Hamiltonian replica exchange

- In Hamiltonian replica exchange, our different conditions are different potentials
- In our case, our different Hamiltonians are different localization biases

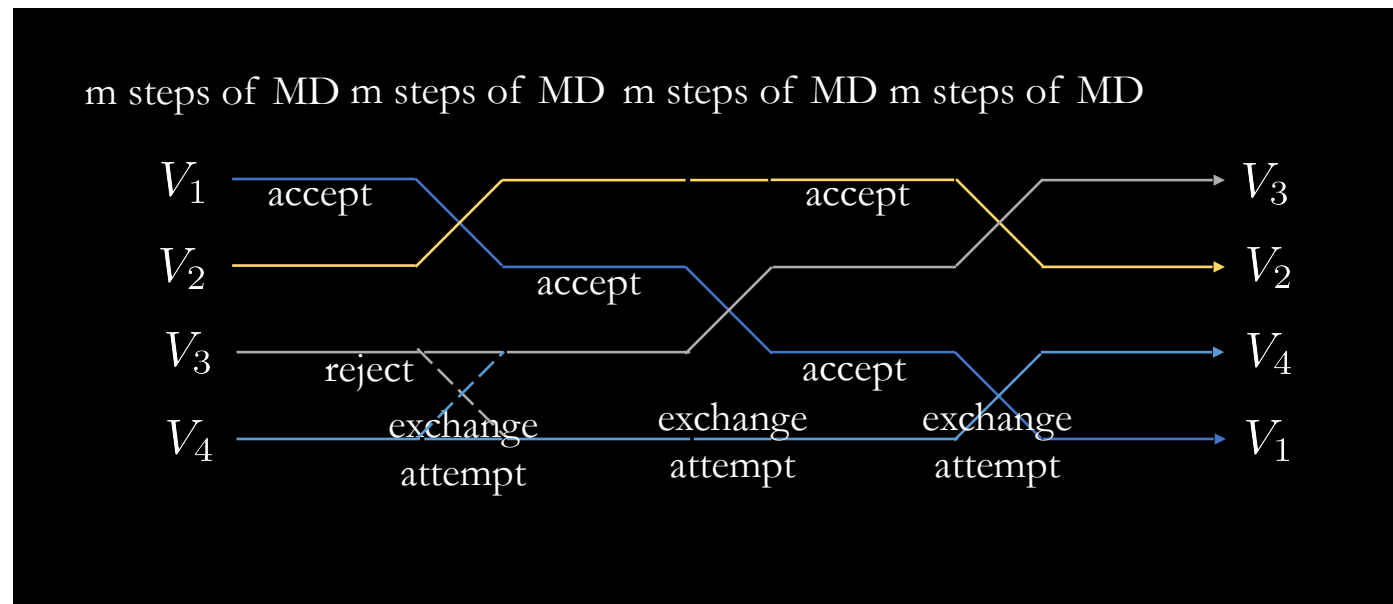
If $\Delta > 1$: Move

If $\Delta < 1$: move with prob Δ

$$\Delta = \frac{p_{T_1}(x_A)p_{T_2}(x_B)}{p_{T_1}(x_B)p_{T_2}(x_A)}$$

$$= \frac{e^{-\beta U_1(x_A)}e^{-\beta U_2(x_B)}}{e^{-\beta U_2(x_A)}e^{-\beta U_1(x_B)}}$$

$$= e^{\beta[(U_1(x_B)-U_1(x_A))+(U_2(x_A)-U_2(x_B))]}$$



Hands-on exercise 4

Perform and analyze REUS

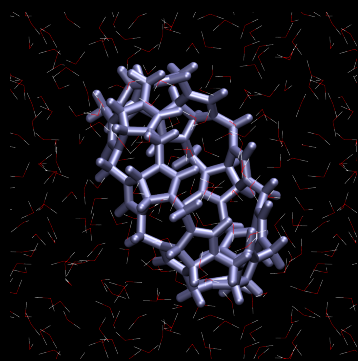
Alchemical intermediate states are useful for calculating solvation, binding or mutation ΔG

- An alchemical intermediate state is defined by a vector of coupling parameters.

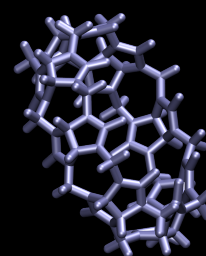
; lambda-states	= 1	2	3	4	5	6	7	8	9
coul-lambdas	= 0.00	0.25	0.50	0.75	1.00	1.00	1.00	1.00	1.00
vdw-lambdas	= 0.00	0.00	0.00	0.00	0.00	0.25	0.50	0.75	1.00
$\vec{\lambda}_i = (\lambda_{vdW,i}, \lambda_{coul,i})$	λ_1	λ_2	λ_3	λ_4	λ_5	λ_6	λ_7	λ_8	λ_9

$$\vec{\lambda}_1 = (0.0, 0.0)$$

Fully coupled
(interactions on)



$$\Delta G_{\text{solv}}$$



$$\vec{\lambda}_9 = (1.0, 1.0)$$

Fully decoupled
(interactions off)

Alchemical replica exchange generalizes the ensemble with the alchemical direction

- Hamiltonian replica exchange (HREX) alternates the sampling between the coordinate and space of Hamiltonians

 Hands-on exercise 5 

