The Accelerated Weight Histogram method

References

- The GROMACS reference manual
- V. Lindahl, J. Lidmar and B. Hess, Accelerated weight histogram method for exploring free energy landscapes, J. Chem. Phys. 141, 044110 (2014)
- V. Lindahl, J. Lidmar and B. Hess, Sequence dependency of canonical base pair opening in the DNA double helix, PLOS Computational Biology, 13 (4), e1005463 (2017)
- V. Lindahl, P. Gourdon, M. Andersson and B. Hess, Permeability and ammonia selectivity in aquaporin TIP2; 1: linking structure to function, Scientific Reports, 8, 2995 (2018)
- V. Lindahl, J. Lidmar and B. Hess, Riemann metric approach to optimal sampling of multidimensional free-energy landscapes, Phys. Rev. E, 98, 023312 (2018)

most of the work done by Viveca Lindahl



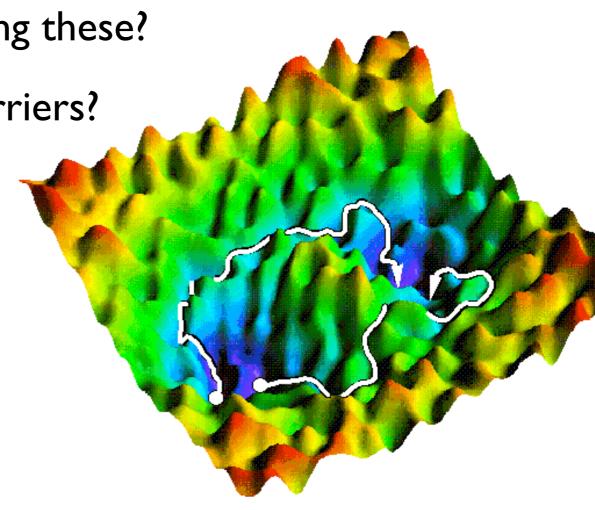
Sampling free-energy landscapes

- Energy landscapes of biomolecules are high dimensional and often rough
- Typical situations:
 - You know the beginning and the end state

• What are the paths connecting these?

• What are the free-energy barriers?

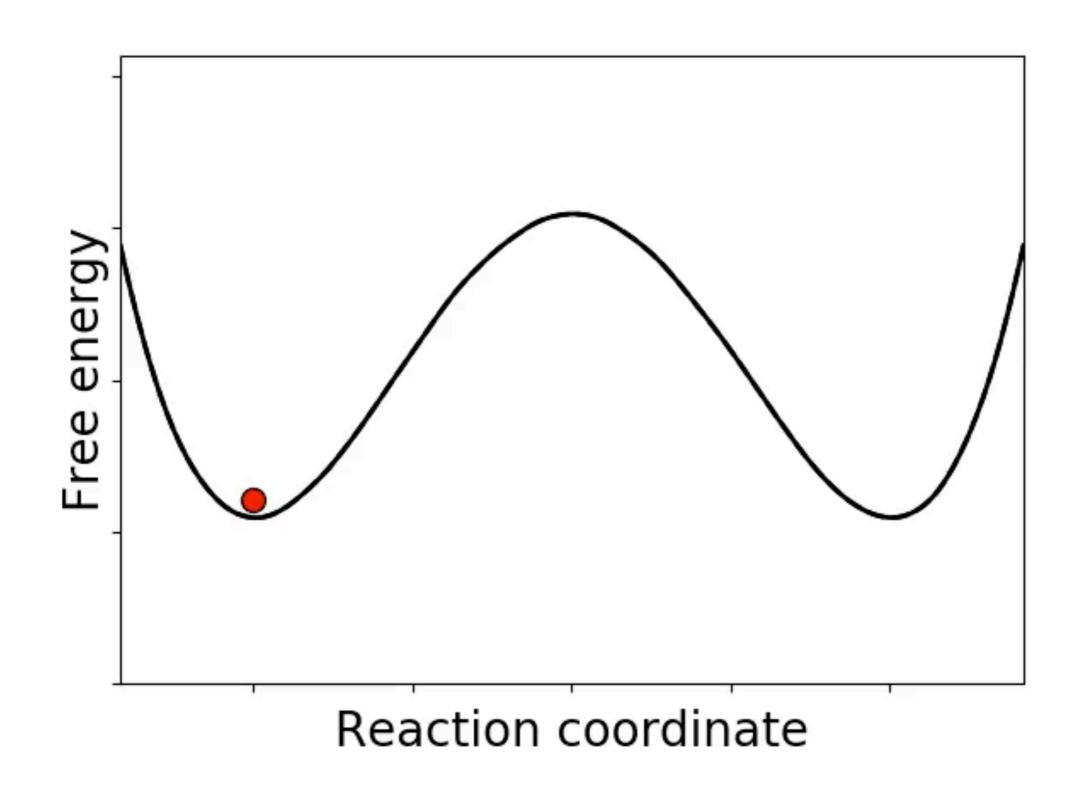
- You know the beginning state
 - ...
- Can you come up with a reaction coordinate?



Typical issue in molecular dynamics

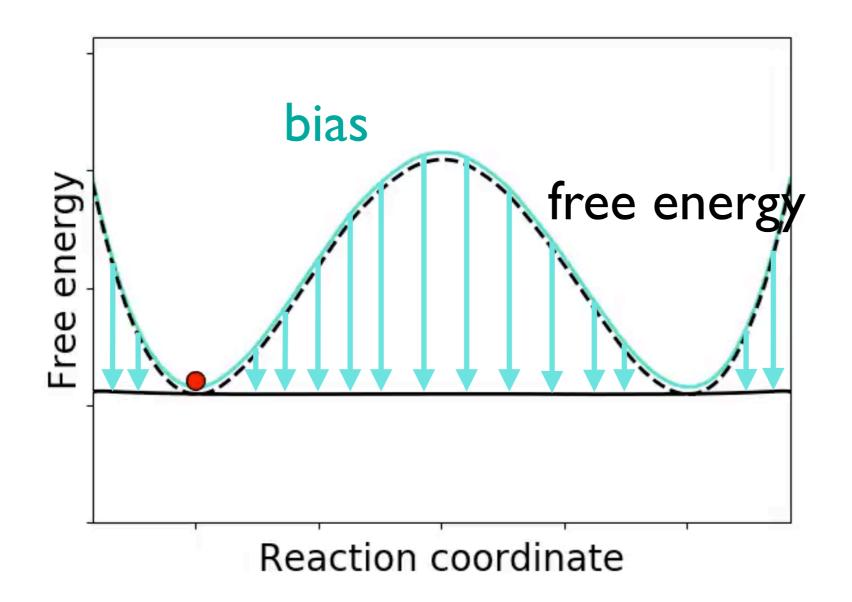
- Interesting events happen on a microsecond timescales
 - thus we needs billions of time steps of 2 fs
- But the event itself is often fast
 - We can (smartly) use many independent simulations
 - We can smartly bias simulations to get more events
- Thus
 - More efficient use of compute time
 - Shorter time to solution

The sampling problem



The solution

Trick: add a bias potential to make the effective potential flat Issue: the potential (or free-energy) is what we are after!



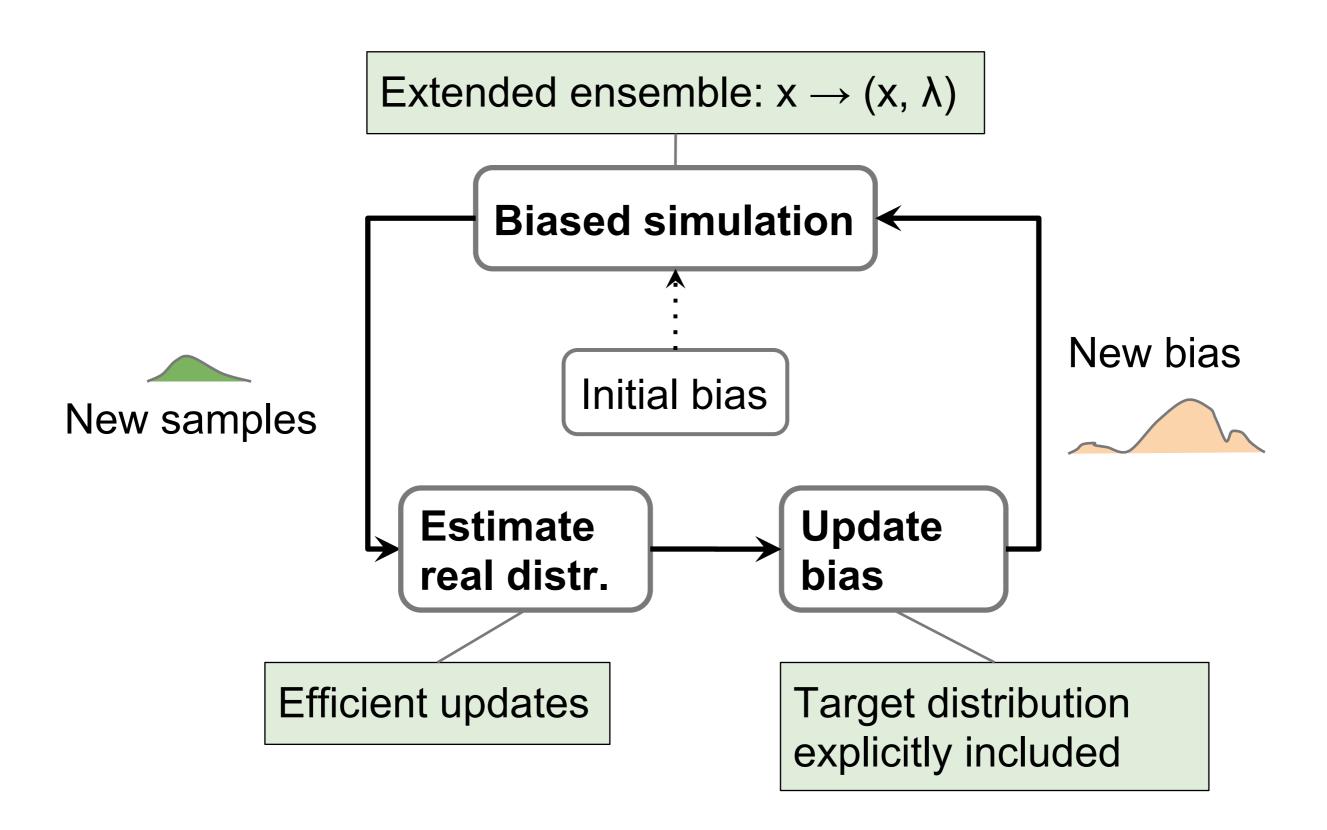
Solving the bias potential problem

- The bias potential is both input and output of the method:
 - an iterative solver is required
- Available solvers:
 - Metadynamics
 - Adaptive biasing force
 - Accelerated weight histogram method
 - ...

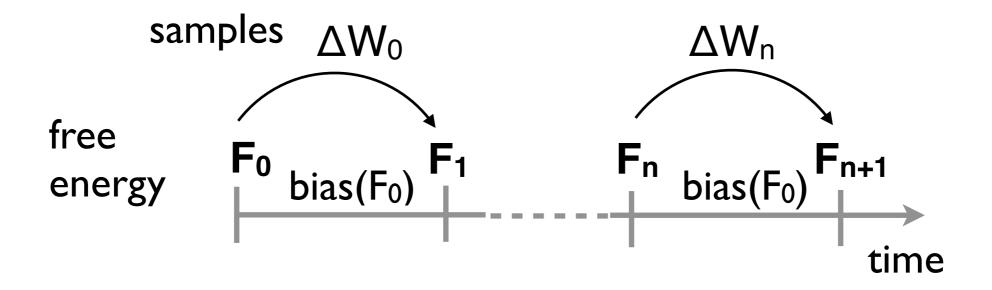
Accelerated weight histogram method

- The target distribution is the central quantity (unlike most other methods)
- The target distribution can be chosen freely and can, but does not have to, depend on the free-energy
- Initial exponential convergence, later sqrt(#samples)
 - automatically controlled
- Only one, uncritical, convergence parameter

AWH schematically



The Accelerated Weight Histogram method



Iterative scheme to solve for the unknown bias / free-energy:

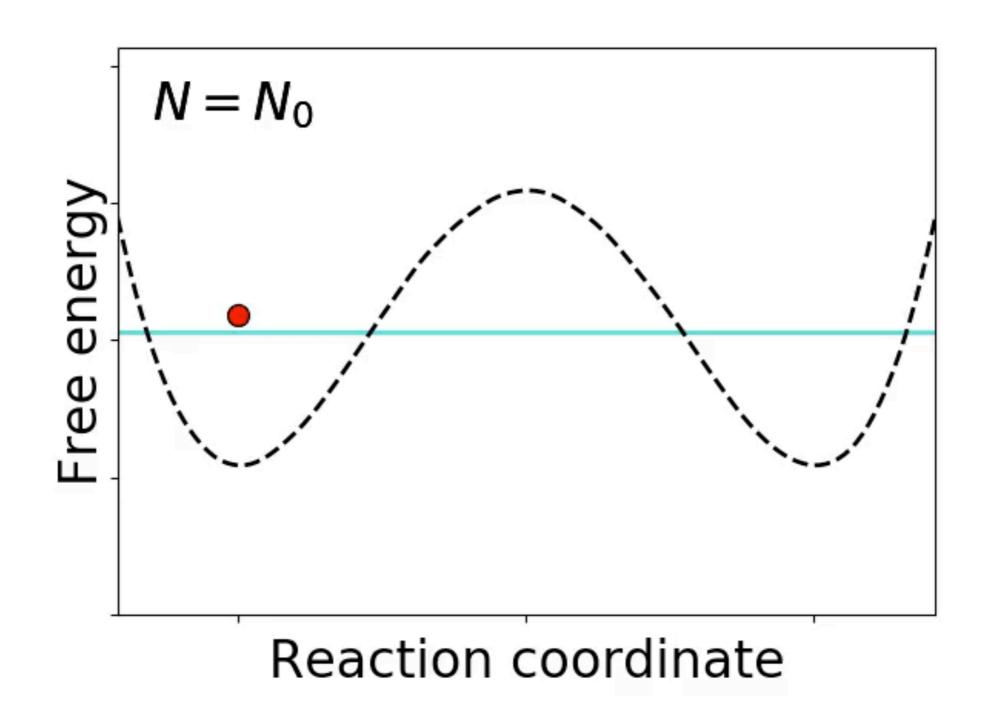
- collect samples (using MD)
- update the free-energy estimate

$$\Delta F_n(\lambda) = -\ln\left(1 + \frac{\Delta W_n(\lambda)}{N_n\pi(\lambda)}\right) \sim \frac{1}{N_n}$$
 prior number of samples (flat)

How does AWH apply the bias?

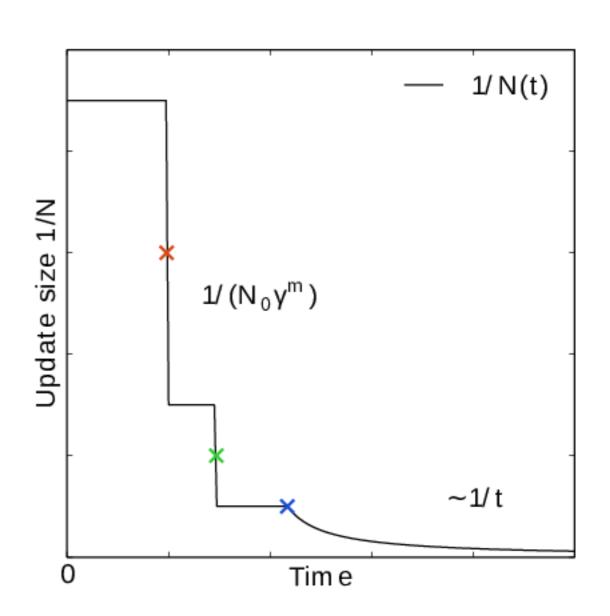
- Initially we used a harmonic umbrella potential which moved using Monte Carlo moves
- Now the default setup is the "Boltzmann inversion" of a convolution of Gaussians, produces by harmonic umbrellas, which creates a smooth bias potential
 - This is done on a regular grid considering only close neighbors
 - Thus no higher cost when using a fine grid / high force constant

AWH in action



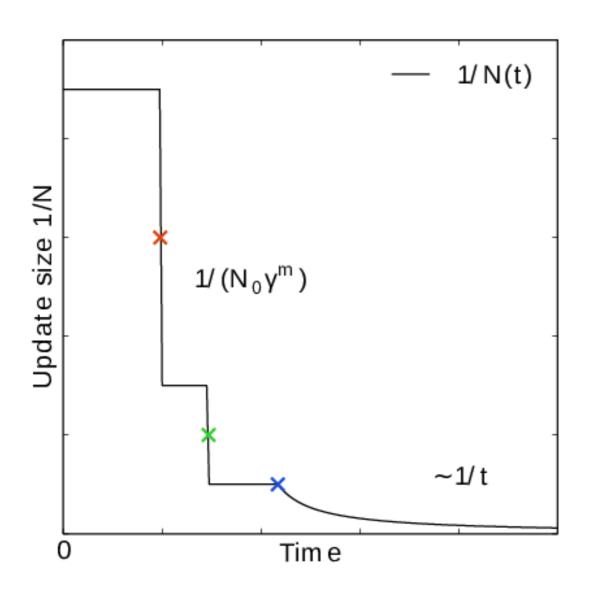
Exponential-linear convergence

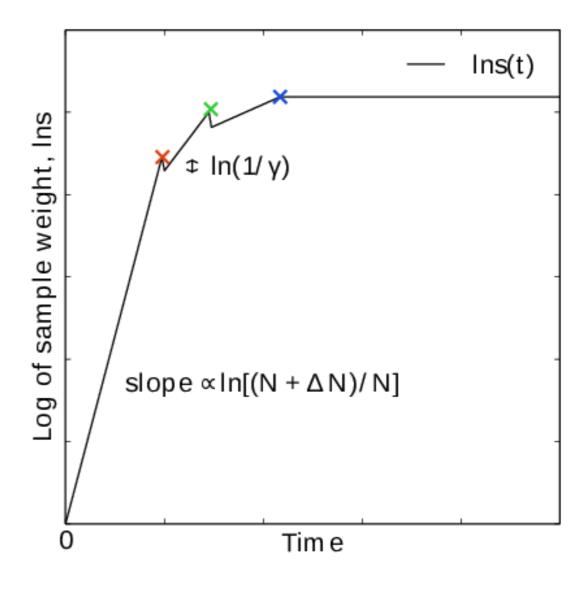
- Initial stage
 - constant update size
 - Divide update size by 3 at every covering (cross
- Final stage
 - constant update weight
- Switch when initial weight would exceed final weight
 - automated equilibration check

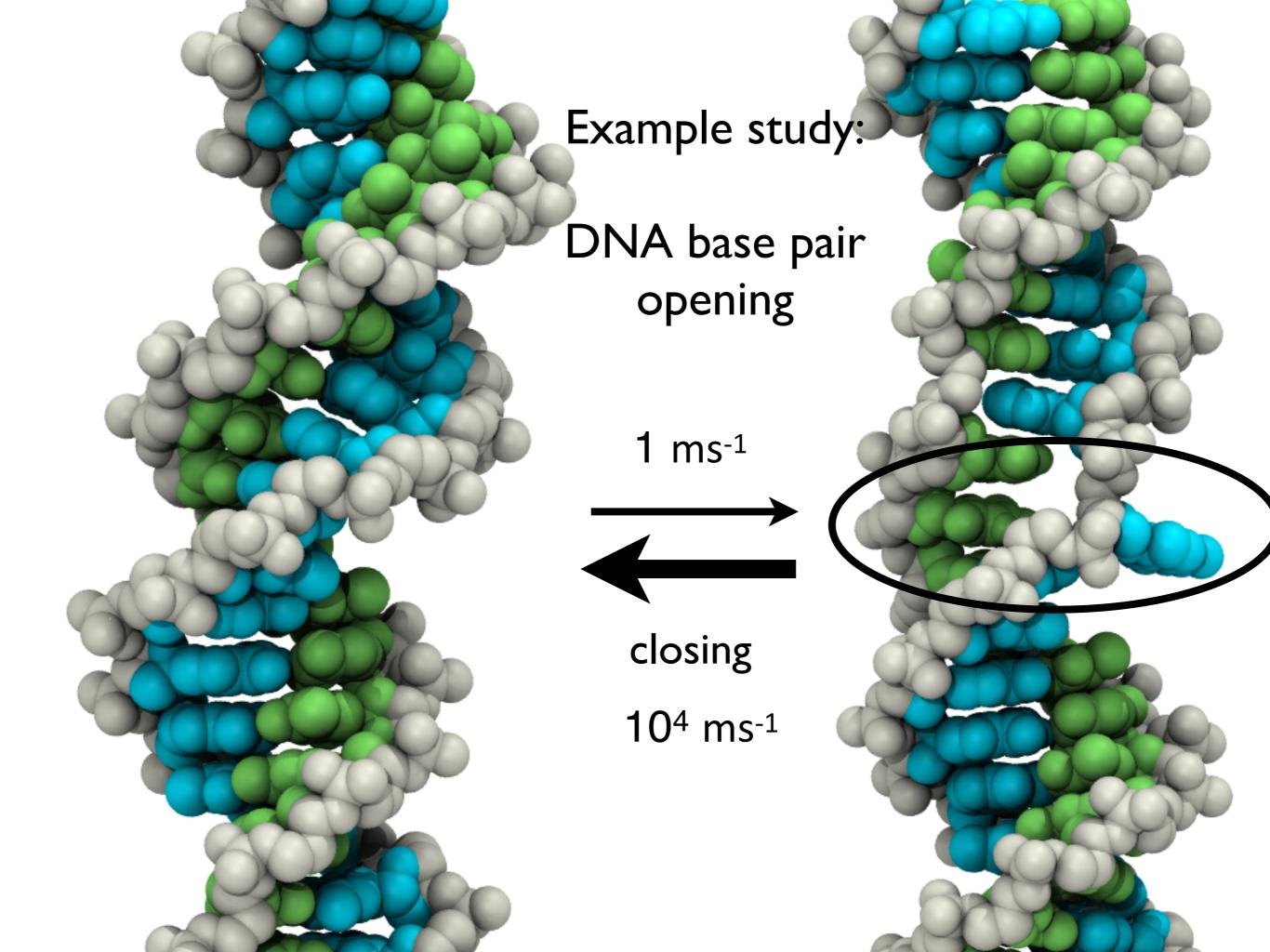


Sample weights

Initial stage, non-equilibrium samples get weighted down

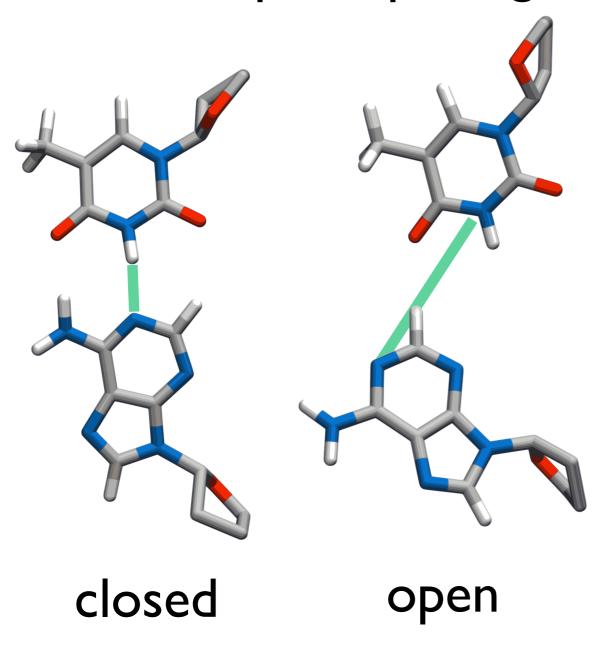


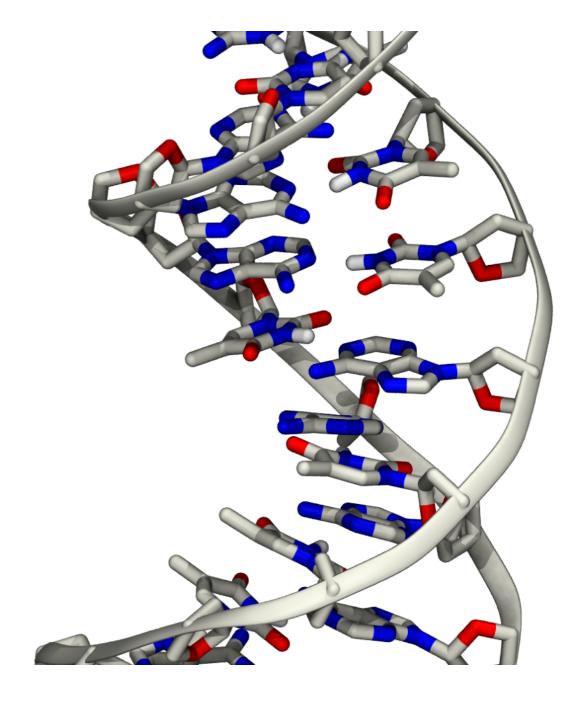




A reaction coordinate

base-pair opening





mdp options

```
awh
                                                   = yes
                     = yeawh-nbias
pull
                         awh1-ndim
                                                   = 1
                     = 2 awh1-dim1-coord-index
pull-ngroups
                     = reswal-start
                                                   = 0.25
pull-group1-name
                                                   = 0.60
                     = reswaladim1-end
pull-group2-name
                         awh1-dim1-force-constant = 128000
                     = 1 awh1-dim1-diffusion
                                                   = 5e-5
pull-ncoords
pull-coord1-geometry = diswadcerror-init
                                                   = 5
pull-coord1-groups
pull-coord1-type = externawhpshaneimultisim
                                                     = yes
pull-coord1-potential-providershagehgroup
                         awh1-equilibrate-histogram = yes
```

Initial update size

```
awh1-dim1-diffusion = 5e-5 (nm<sup>2</sup>/ps)

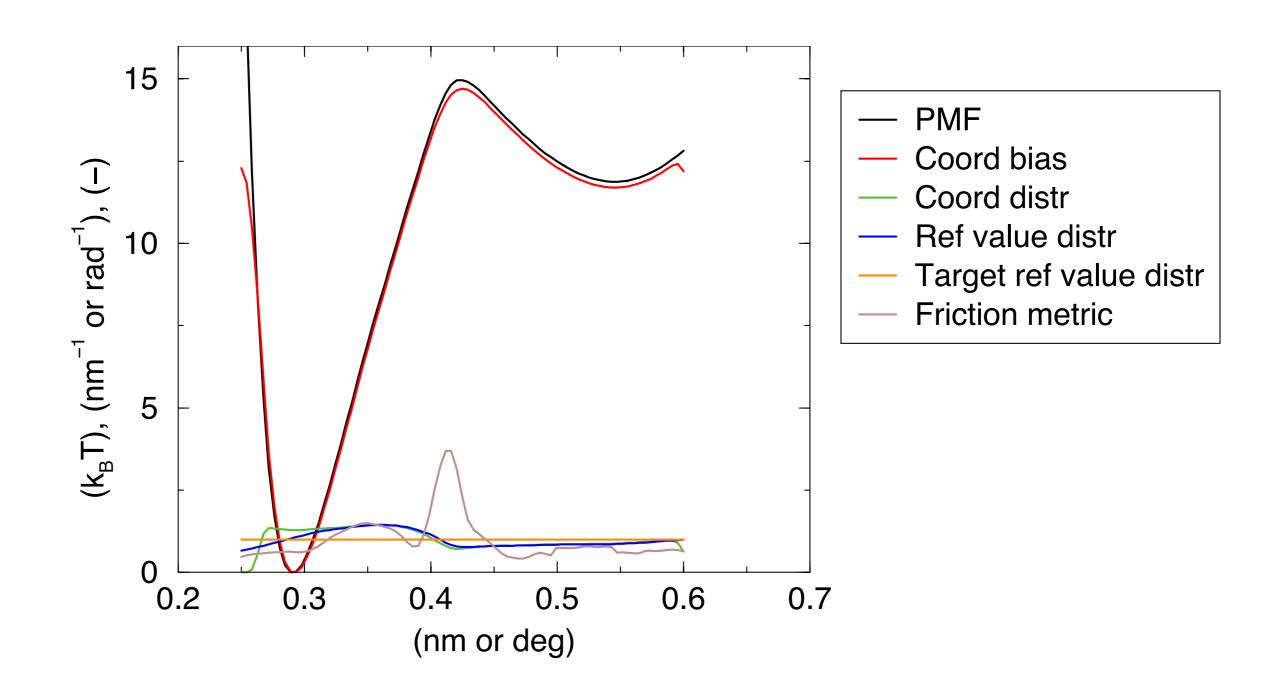
awh1-error-init = 5 (kJ/mol)
```

- AWH has only one tuning parameter: the initial update size
 - This is set using two meaningful parameters:
 - The initial error in the free-energy ε₀
 - The diffusion along the reaction coordinate D

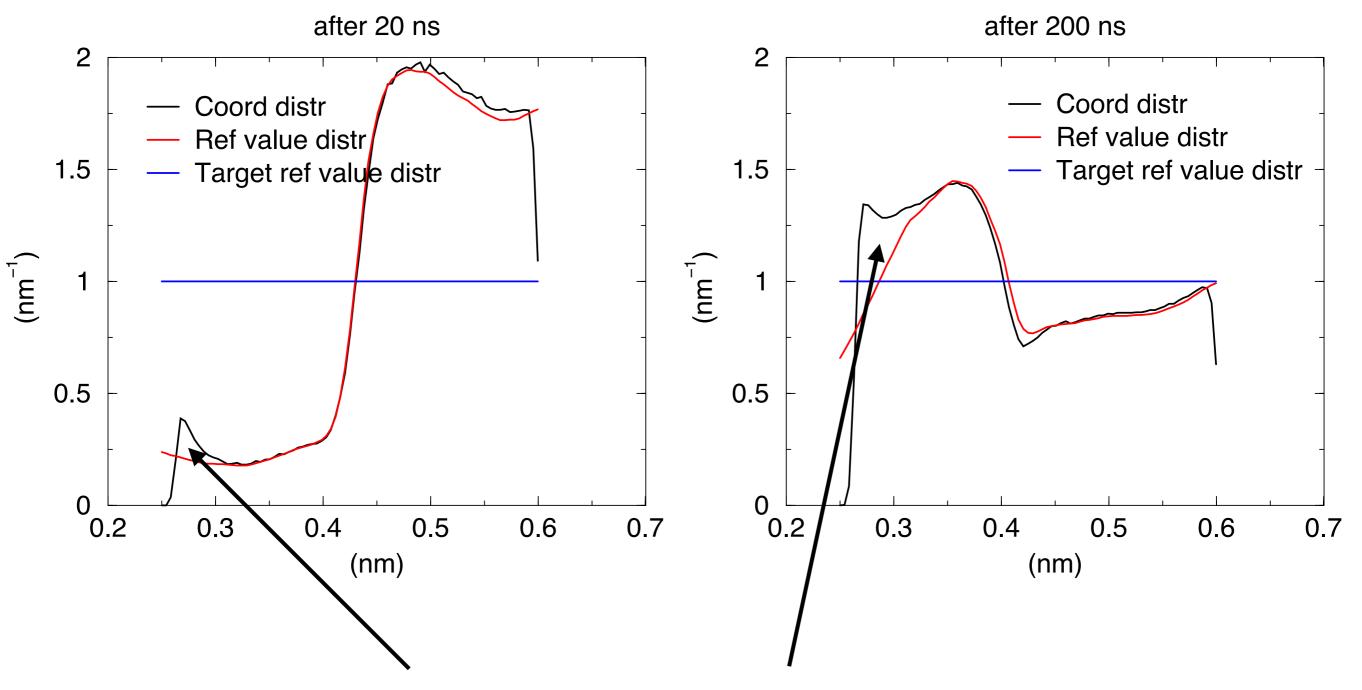
$$\frac{1}{N_0} = \frac{1}{N_0(\epsilon_0, D)} \sim D\epsilon_0^2$$

- Much too large parameters: slow convergence
- Much too small parameters: system might be pulled apart

Output of: gmx awh

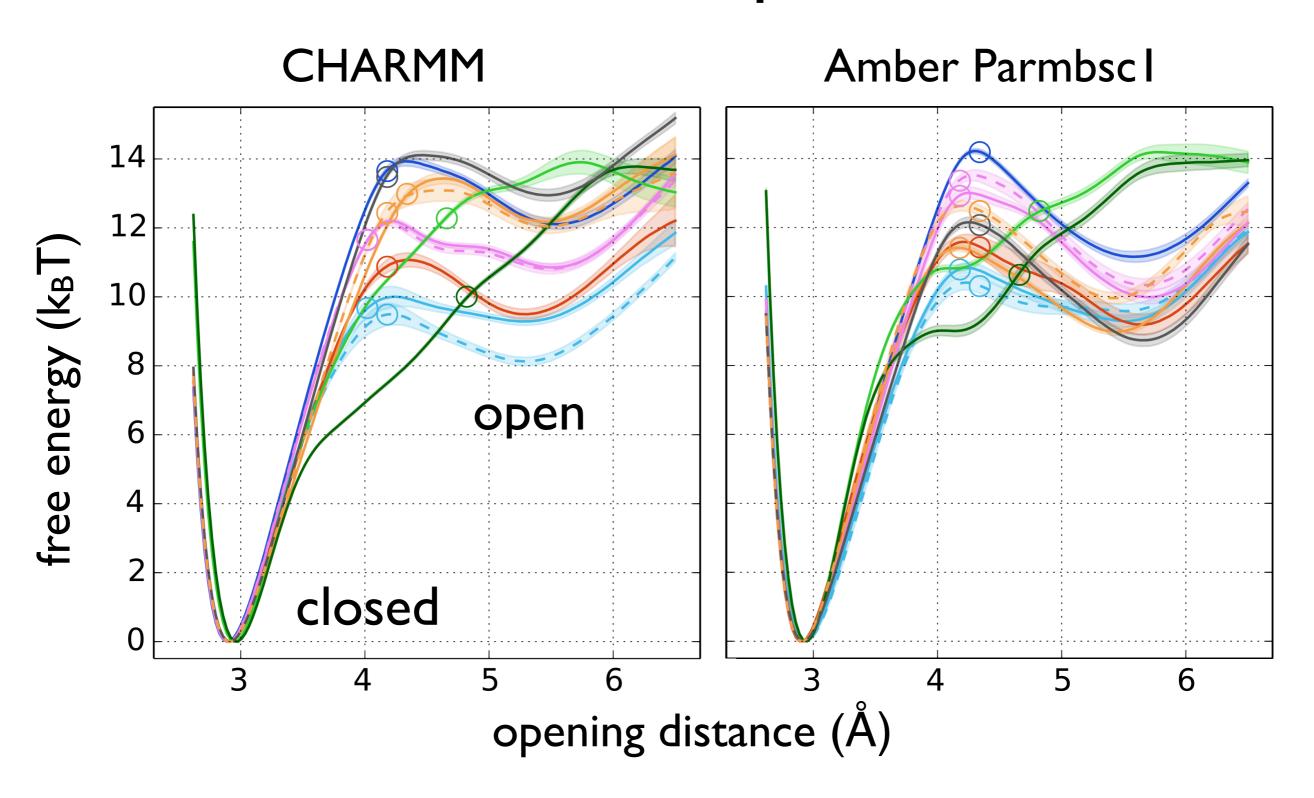


Distributions



typical for the free-energy being steeper than the umbrella potential

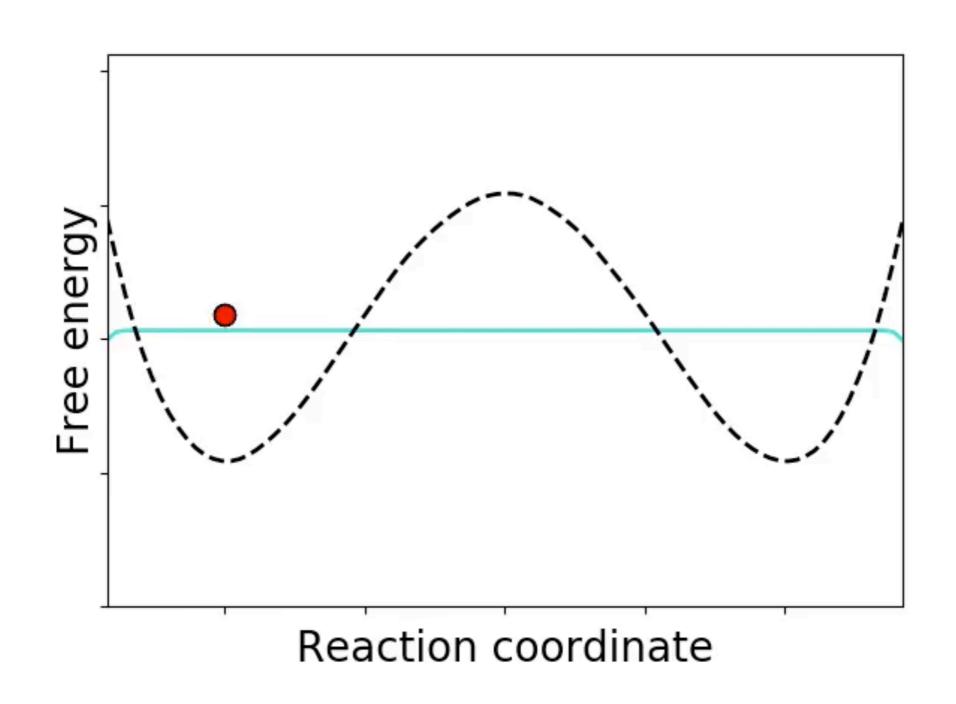
Results for different force fields and base pairs



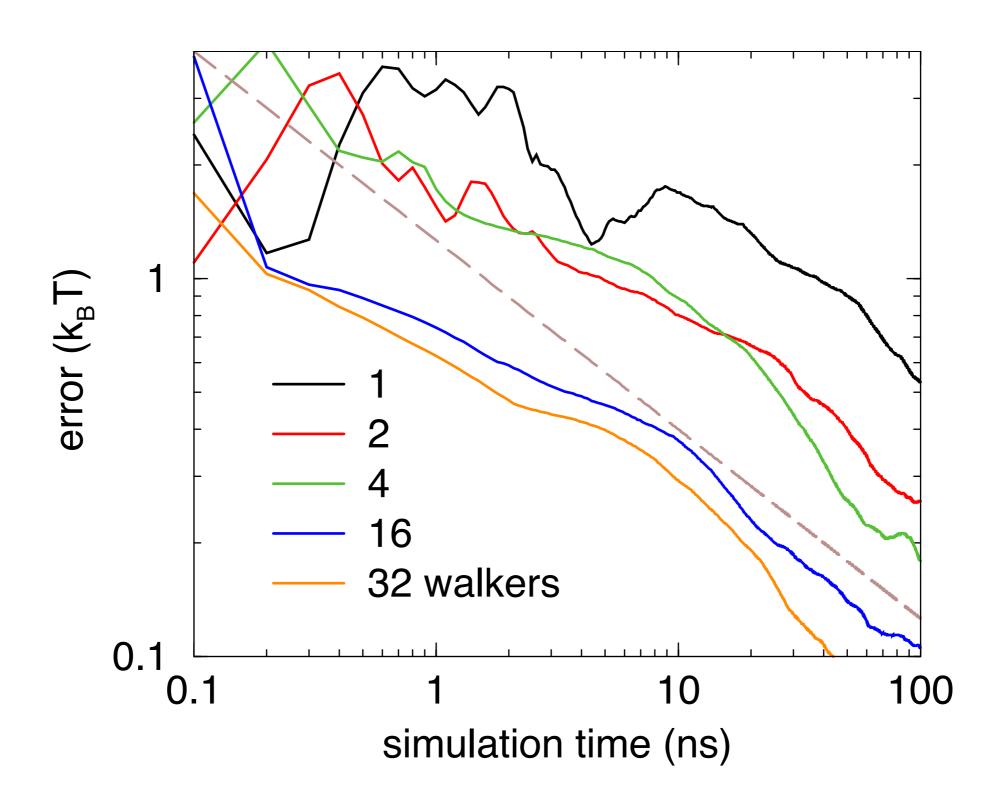
AWH with ensemble parallelism

- There is no reason to limit the sampling phase in AWH to a single simulation
- We can run an ensemble of multiple simulations
- Simulations sample independently and share the bias
 - How many copies can we run?
 - Or: how does the convergence depend on the number of copies?

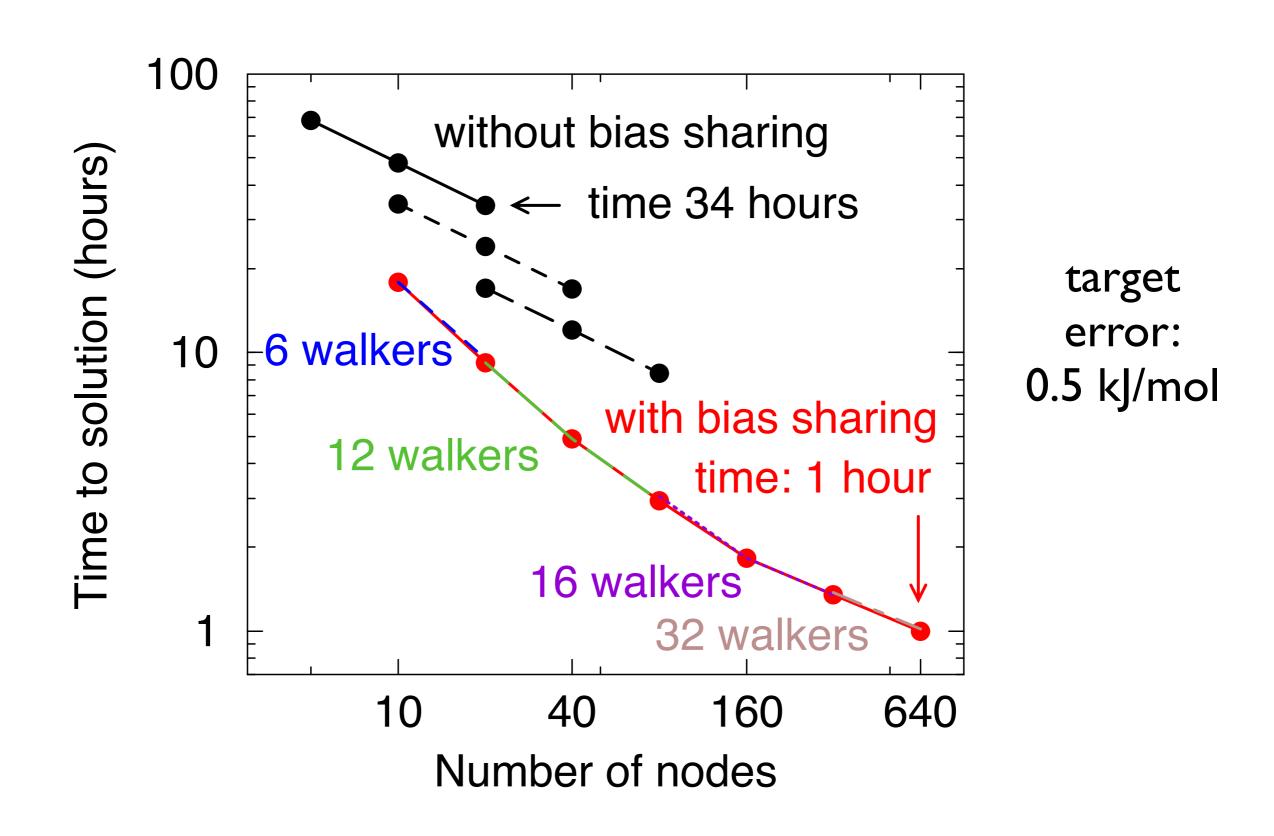
AWH with ensemble parallelism



Converge for #walkers



Time to solution for opening 20 base pairs on CSCS Piz Daint CPUs + P100 GPUs



DNA with 2D AWH with FE cut-off of 20 k_BT

0.6

0.3

0.4

d(N1-N3) (nm)

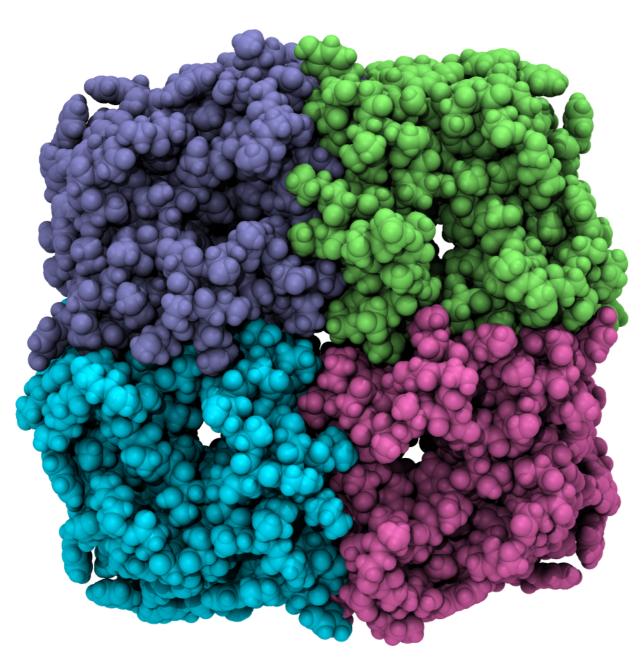
0.5

0.6

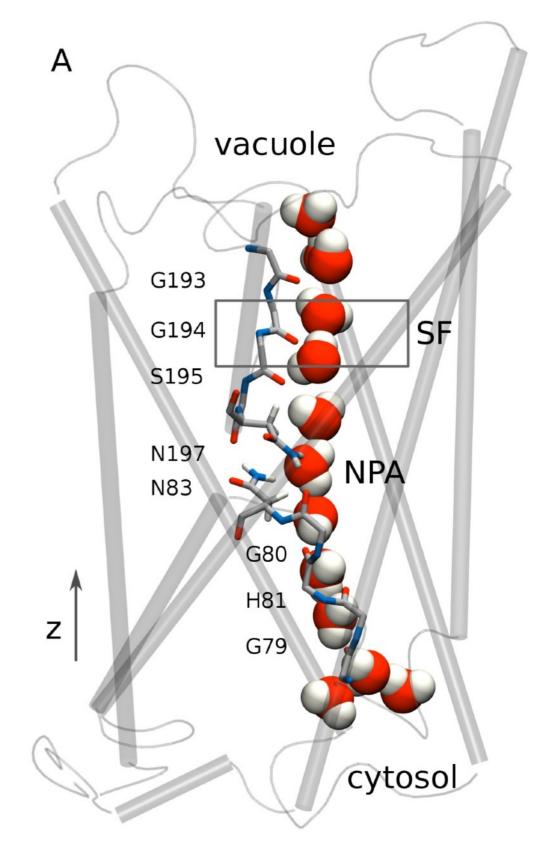
20 0.5 d(O4-C2) (nm) closéo energy 1 HB distance distance used in ID case in open state open 5 0.3 **PMF** 0.3 0.4 0.5 0.6 d(N1-N3) d(N1-N3) (nm) d(O4-C2) d(04-C2) (nm) 6 closed friction open 2 metric 0.3 Riemann metric approach to optimal sampling of multidimensional

free-energy landscapes, Phys. Rev. E, 98, 023312 (2018)

Aquaporin H₂O vs NH₃

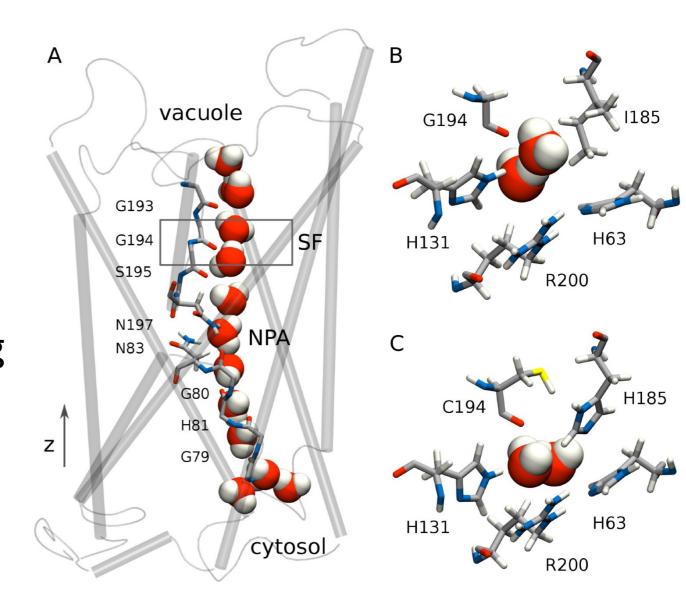


we run 4 independent AWH biases, I for each monomer

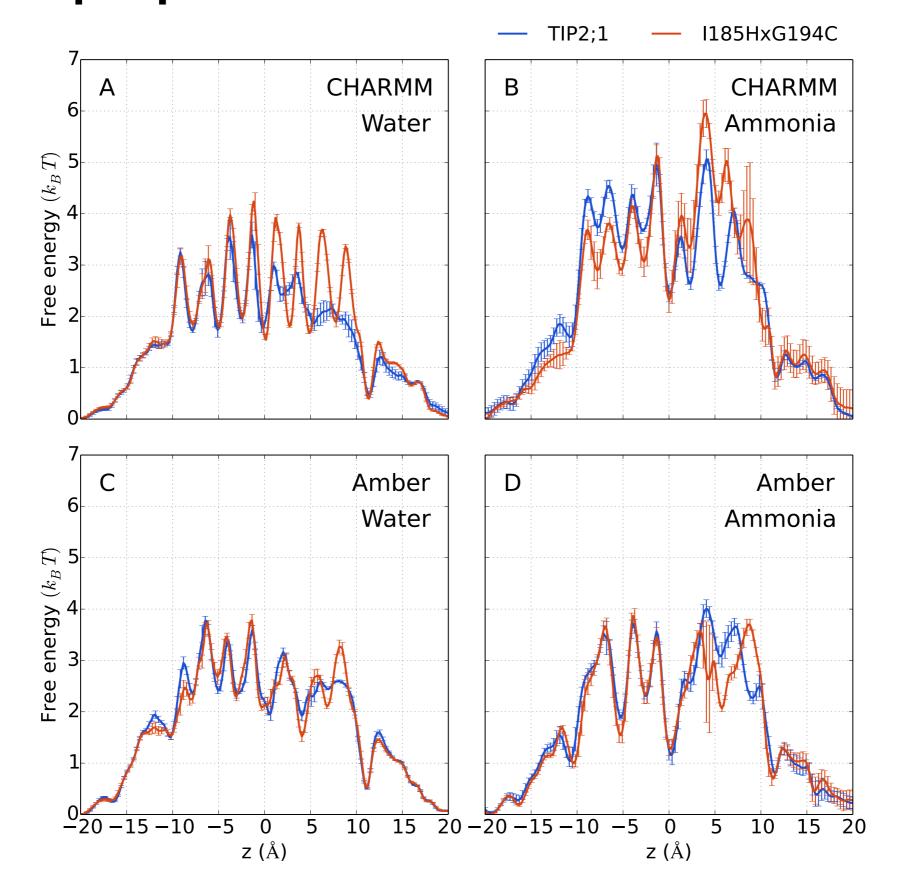


Aquaporin H₂O vs NH₃: methods

- For H₂O permeability we can use equilibrium simulations
- For NH₃ we use AWH
 - Umbrella sampling can block rearrangements of pore lining side chains
 - Forced motion can lead to non-equilibrium artefacts



Aquaporin H₂O vs NH₃: PMFs

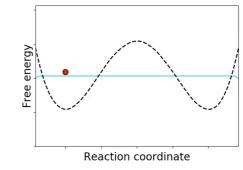


AWH vs PLUMED

- AWH is very robust and easy to use
- AWH can currently only act on COM pull coordinates
- AWH is fully integrated into GROMACS
- AWH and COM pulling works efficiently in parallel
- We will extend the pull code with atomic contact and combinations of pull coordinates
 - Suggestions for extensions are welcome
- We will not cover all options range PLUMED supports

AWH conclusions

- Very robust accelerated sampling method
- Target distribution can be chosen freely
- One only parameter that controls convergence
 - and this parameter is not sensitive
- Supports ID, 2D, 3D and 4D reaction coordinates
- Supports multiple walkers
- Works efficiently in parallel



• Note: The challenge is (still) is coming up with good RCs!

