

The accelerated weight histogram method, AWH

for PMF calculations in GROMACS

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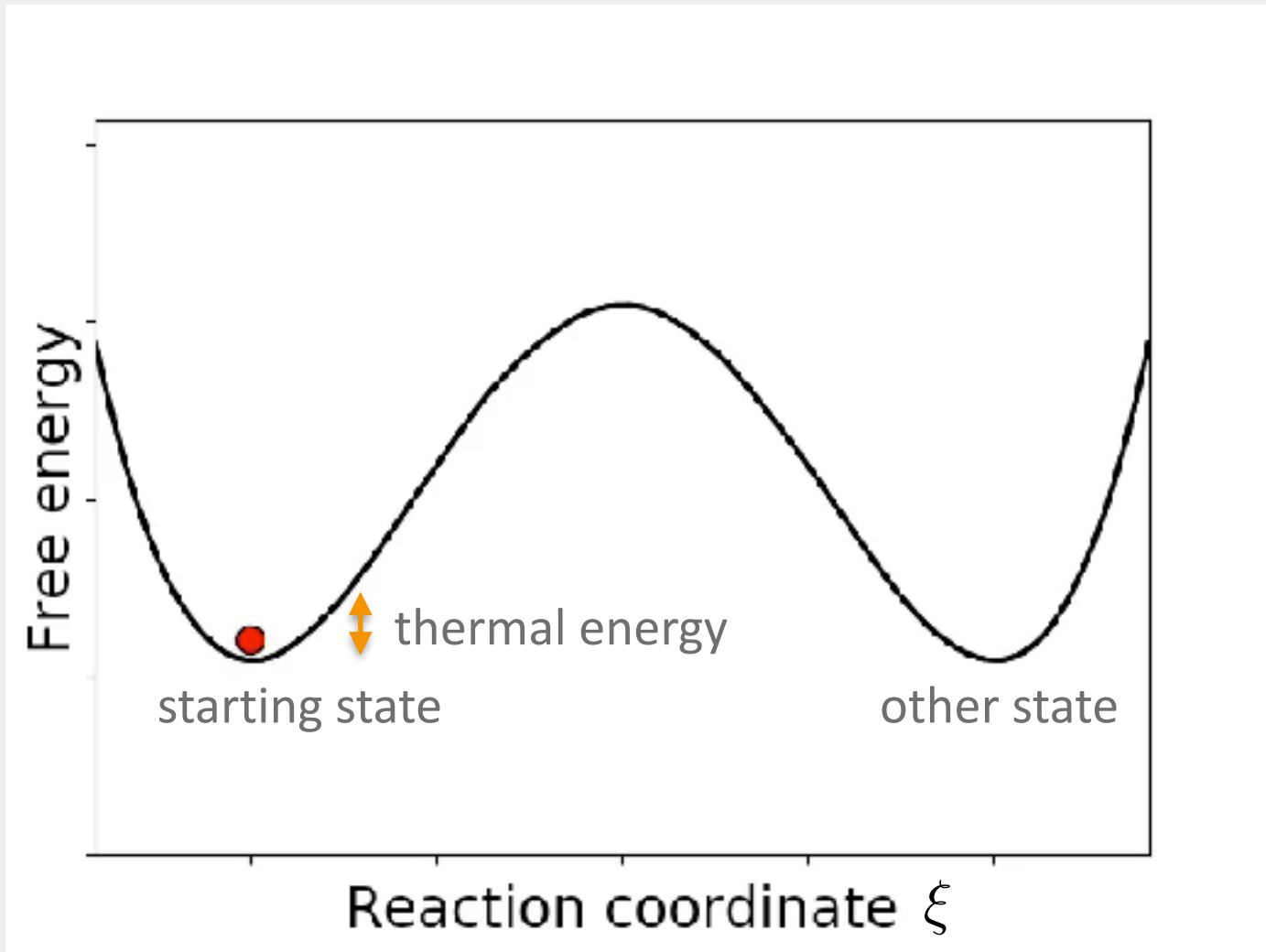
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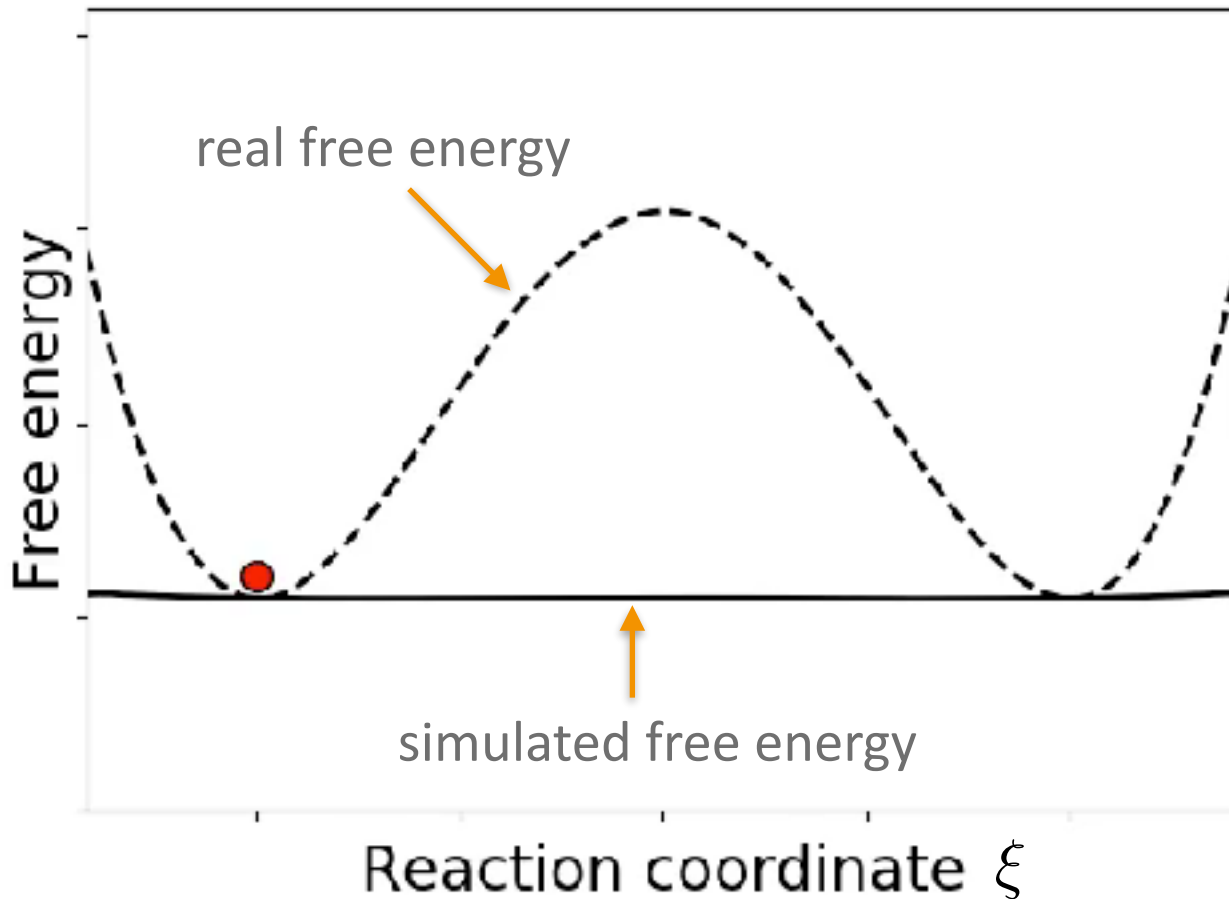
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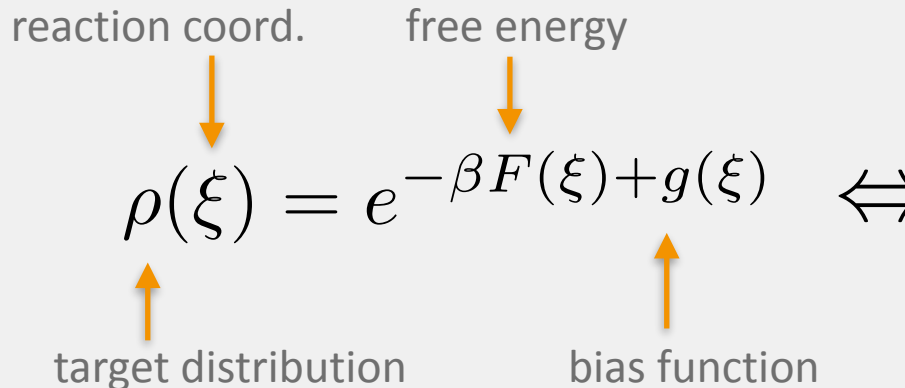
The problem of free energy barriers



How we'd like to sample



Need the free energy to apply the right bias

reaction coord. free energy


$$\rho(\xi) = e^{-\beta F(\xi) + g(\xi)} \iff g(\xi) = \beta F(\xi) + \ln \rho(\xi)$$

target distribution bias function

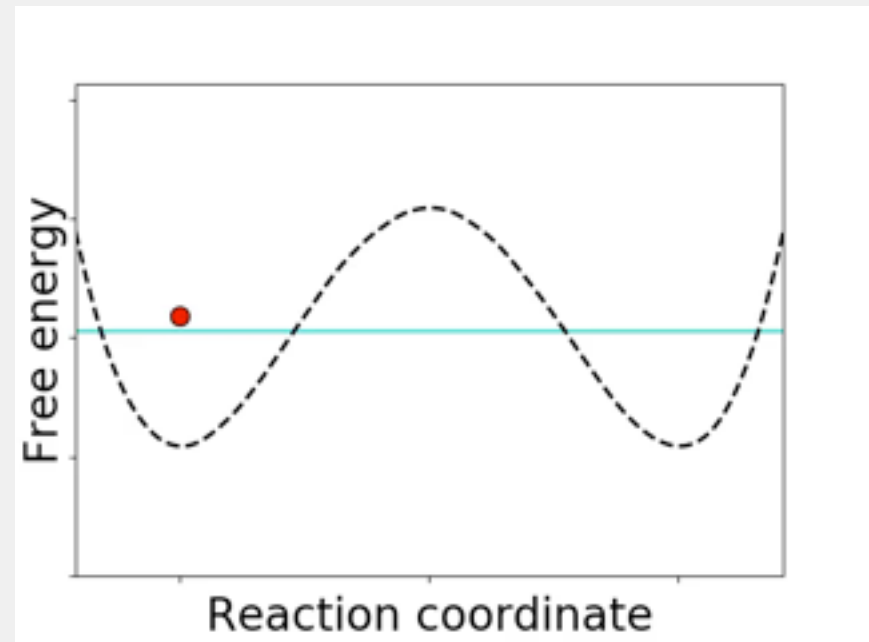
Calculating the free energy requires the bias,
 the bias requires the free energy
 — proceed adaptively!

AWH— Accelerated Weight Histogram method

Adaptively estimates free energy and applies the bias

Basic algorithm:

1. Estimate free energy
2. Set bias
3. Collect (biased) samples
4. Update free energy estimate

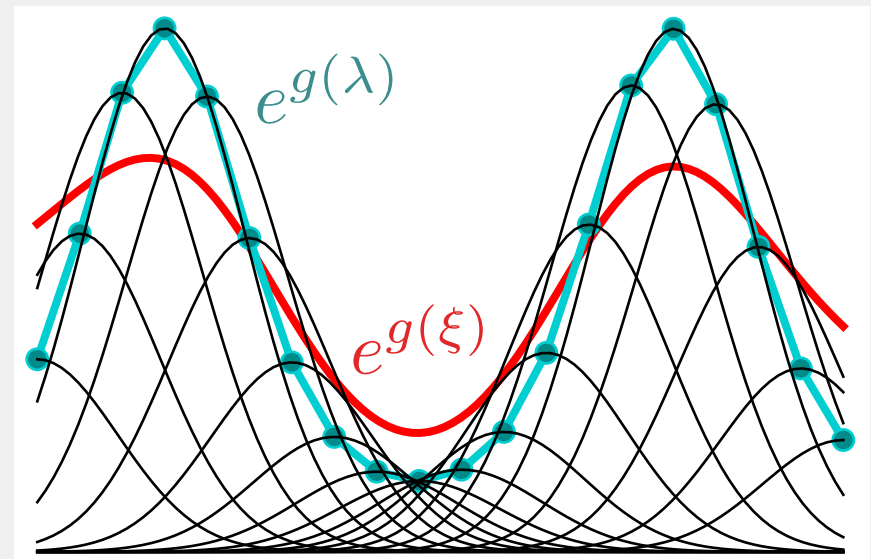


How is the bias represented and applied?

- Discretize ξ space with grid points λ
- Parameterize bias using Gaussian “basis” functions*

$$e^{g(\xi)} = \sum_{\lambda} e^{g(\lambda)} e^{-\frac{1}{2} \left(\frac{\xi - \lambda}{\sigma} \right)^2}$$

- Width σ sets resolution
- Bias force: $\nabla g(\xi(x))$



*abuse of notation: two different (but similar) functions g

Physical interpretation of λ

$$e^{g(\xi)} = \sum_{\lambda} e^{g(\lambda)} e^{-\frac{1}{2} \left(\frac{\xi - \lambda}{\sigma} \right)^2}$$

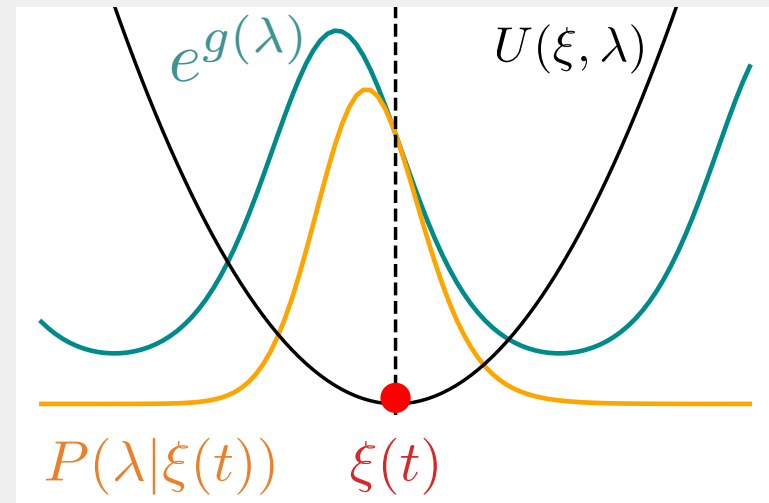
Probabilistically the same as a having a “particle” at λ that experiences an external bias $g(\lambda)$ and interacts with ξ through a harmonic potential

$$U(\xi, \lambda) = \frac{k}{2} (\xi - \lambda)^2, \quad k = \frac{\beta}{\sigma^2}$$

force constant

λ stays close to ξ for large force constants:

$$P(\lambda|\xi(t)) \propto e^{g(\lambda) - \frac{1}{2} \beta k (\xi(t) - \lambda)^2}$$



AWH works on λ rather than ξ

The algorithm again, now with more detail

1. Estimate free energy 2. Set bias 3. Collect samples 4. Update free energy

estimate $F_n(\lambda) \approx F(\lambda)$ exact

$$e^{-\beta F(\lambda)} = \int e^{-\beta F(\xi)} e^{-\frac{1}{2} \left(\frac{\xi - \lambda}{\sigma} \right)^2}$$

“convolved free energy”

- The PMF $F(\xi)$ is extracted by an on the fly reweighting procedure

AWH works on λ rather than ξ

The algorithm again, now with more detail

1. Estimate free energy
- 2. Set bias**
3. Collect samples
4. Update free energy

$$g_n(\lambda) = \beta F_n(\lambda) + \ln \rho(\lambda)$$

bias

free energy estimate

target

AWH works on λ rather than ξ

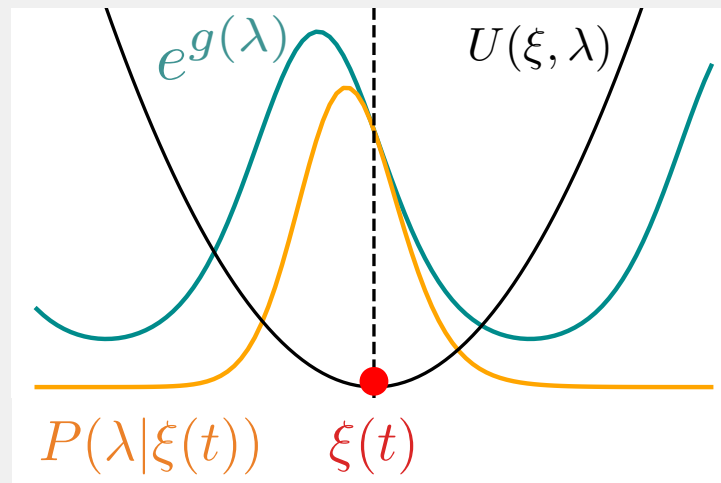
The algorithm again, now with more detail

1. Estimate free energy 2. Set bias 3. **Collect samples** 4. Update free energy

sample at time t , bias g_n

$$\begin{aligned} w_n(\lambda|\xi(t)) &= P_n(\lambda|\xi(t)) \\ &= e^{g_n(\lambda) - \frac{1}{2}\beta k(\xi(t) - \lambda)^2} \end{aligned}$$

a “biased” Gaussian



AWH works on λ rather than ξ

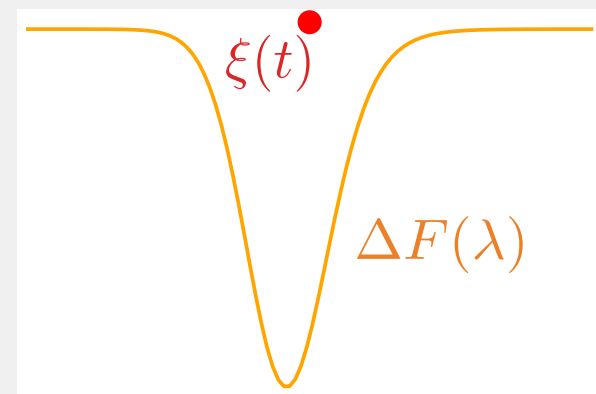
The algorithm again, now with more detail

1. Estimate free energy
2. Set bias
3. Collect samples
4. **Update free energy**

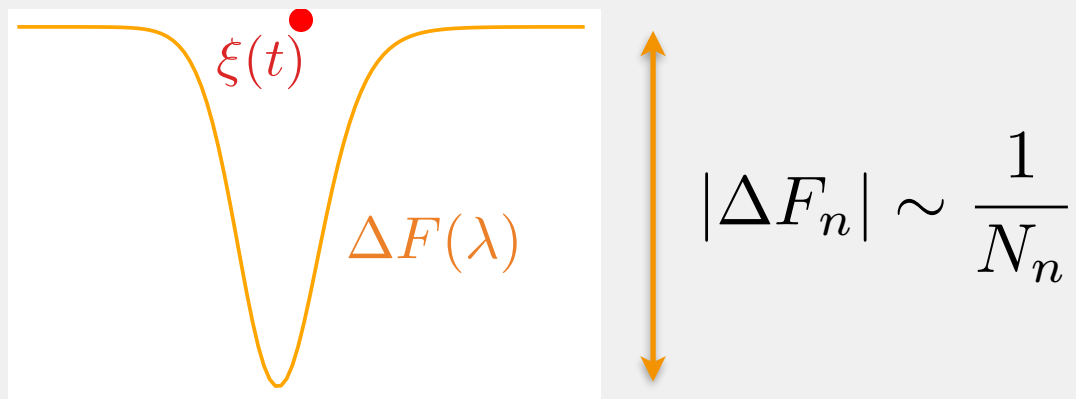
$$\Delta F_n(\lambda) \approx - \frac{\sum_t w_n(\lambda|t)}{N_n \rho(\lambda)}$$

what was sampled
↓

prior number of samples how we hoped to sample



The free energy update size

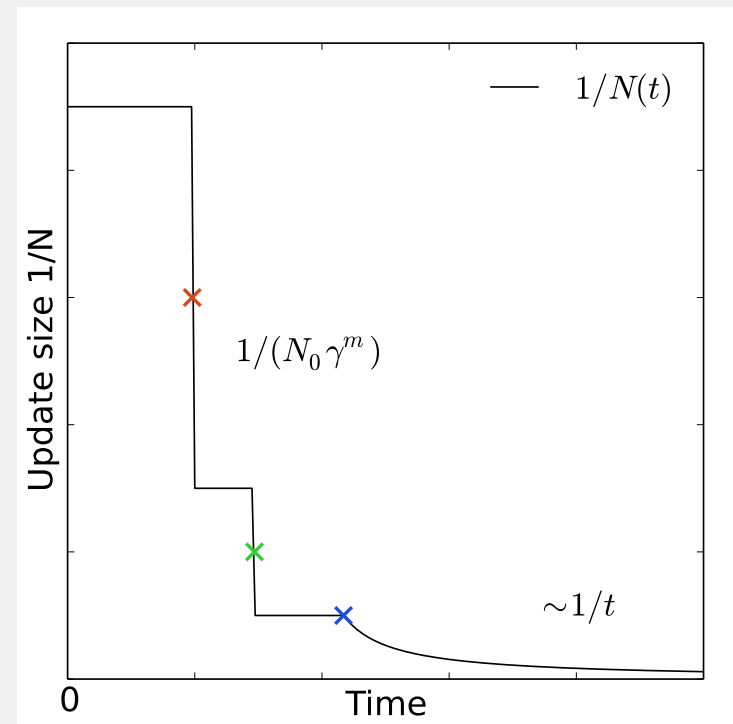
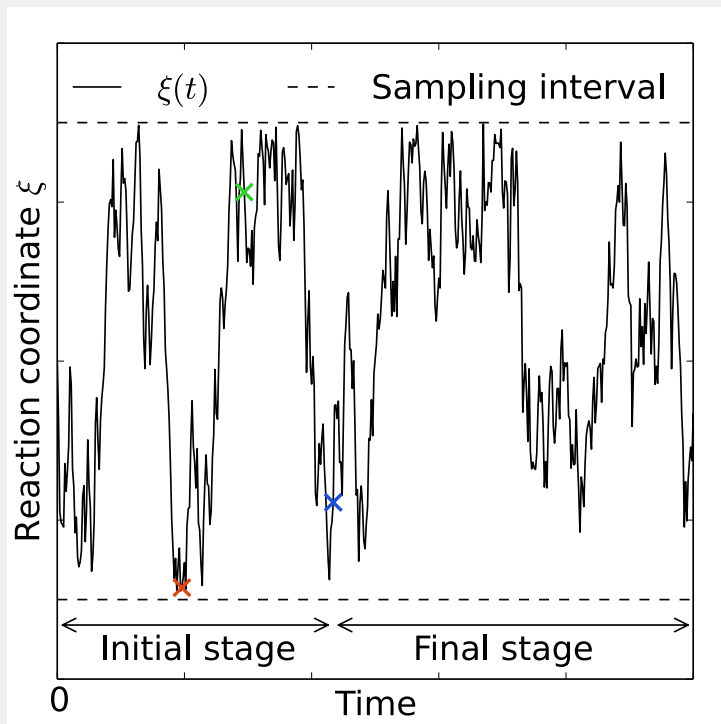


- $N \sim$ total number of prior samples
- reflects the accuracy of the free energy estimate
- should grow at sampling rate

$$N_n \sim t$$

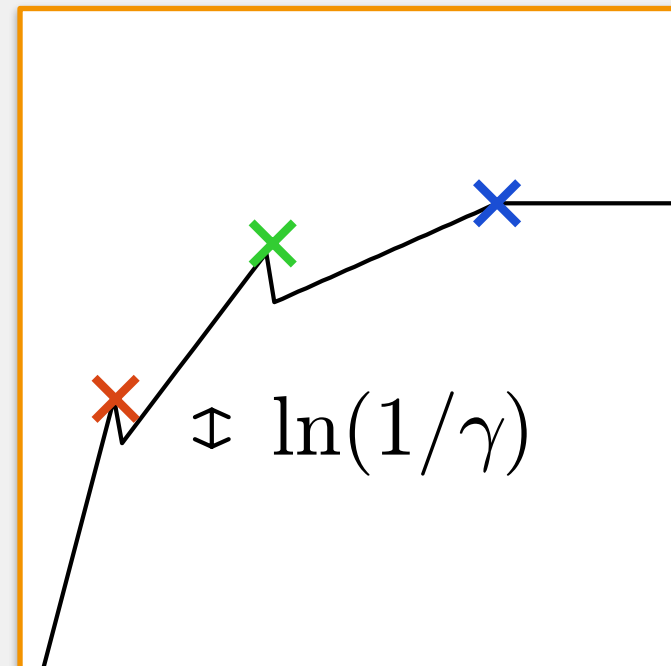
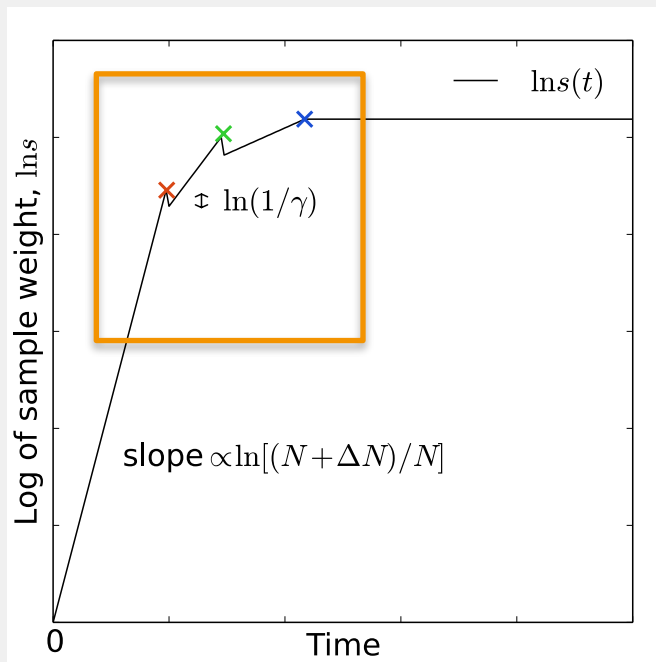
Adding robustness – the initial stage

- Letting N grow at “naturally” decreases the update size too rapidly initially
- *The initial stage*: keep the update size large for the first few transitions.
- After each covering of the sampling interval $N_{n+1} = \gamma N_n$

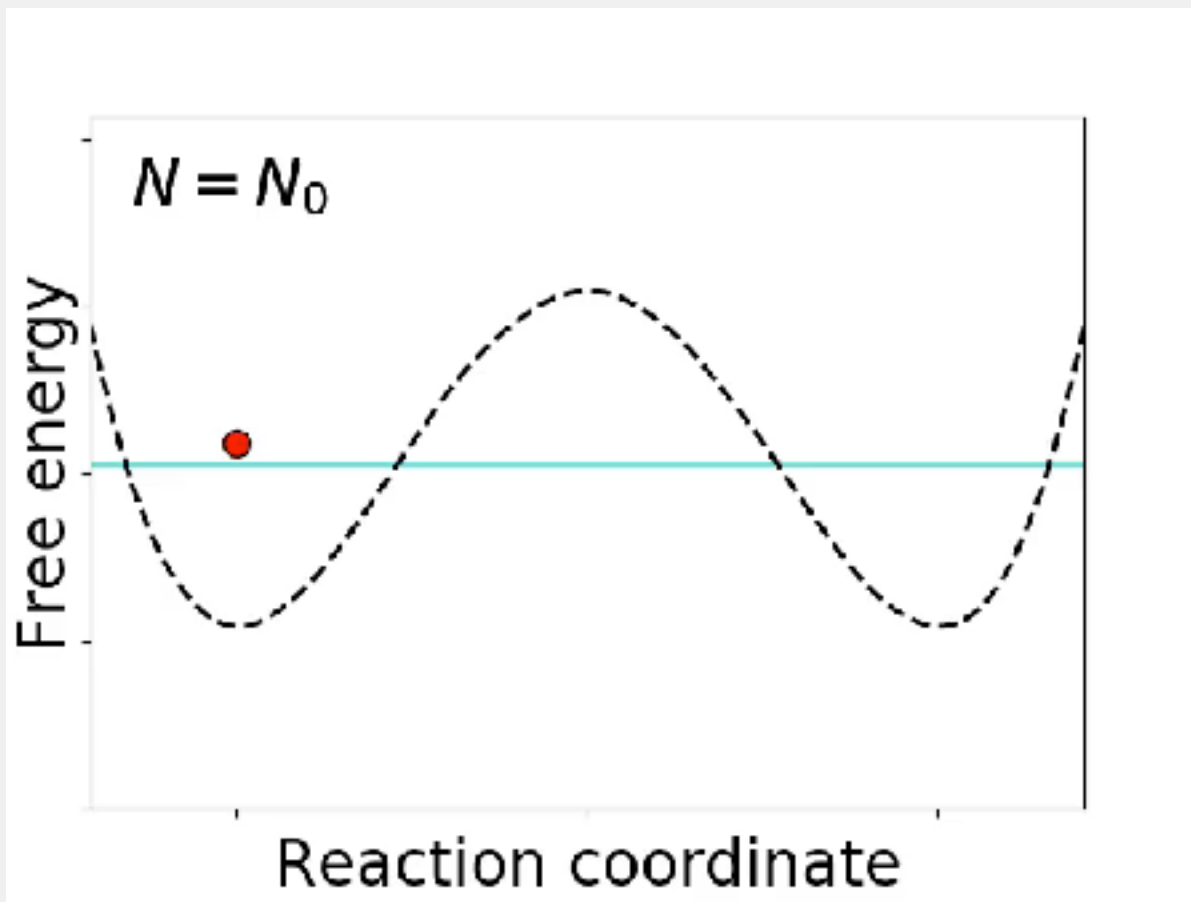


Exiting the initial stage

- Scaling N corresponds to rescaling the current sample weight
- Scaling up, $N_{n+1} = \gamma N_n$, *decreases* sample weight
- Scaling down, $N_n = \text{const.}$, *increases* sample weight
- Exit initial stage when sample weight is no longer increasing



AWH in action

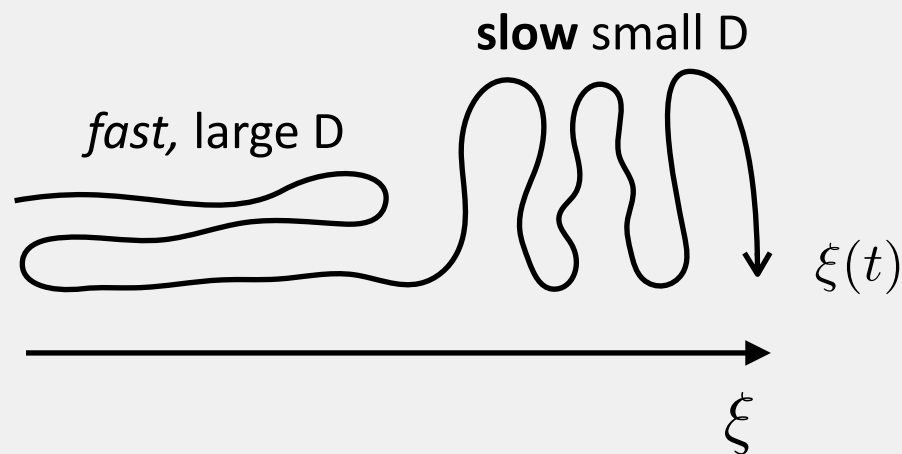


The initial update size $|\Delta F| \sim \frac{1}{N}$

- Sets fluctuation of the free energy and the bias.
- Slow regions get larger fluctuations than fast ones (given N)
- Parameterized in AWH by a *diffusion constant* D

$$\frac{1}{N_0} \sim D$$

- Slow system — small D
- Fast system — large D



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