

Efficient Automated Adaptive Computation of Protein-Ligand Absolute Binding Free Energies



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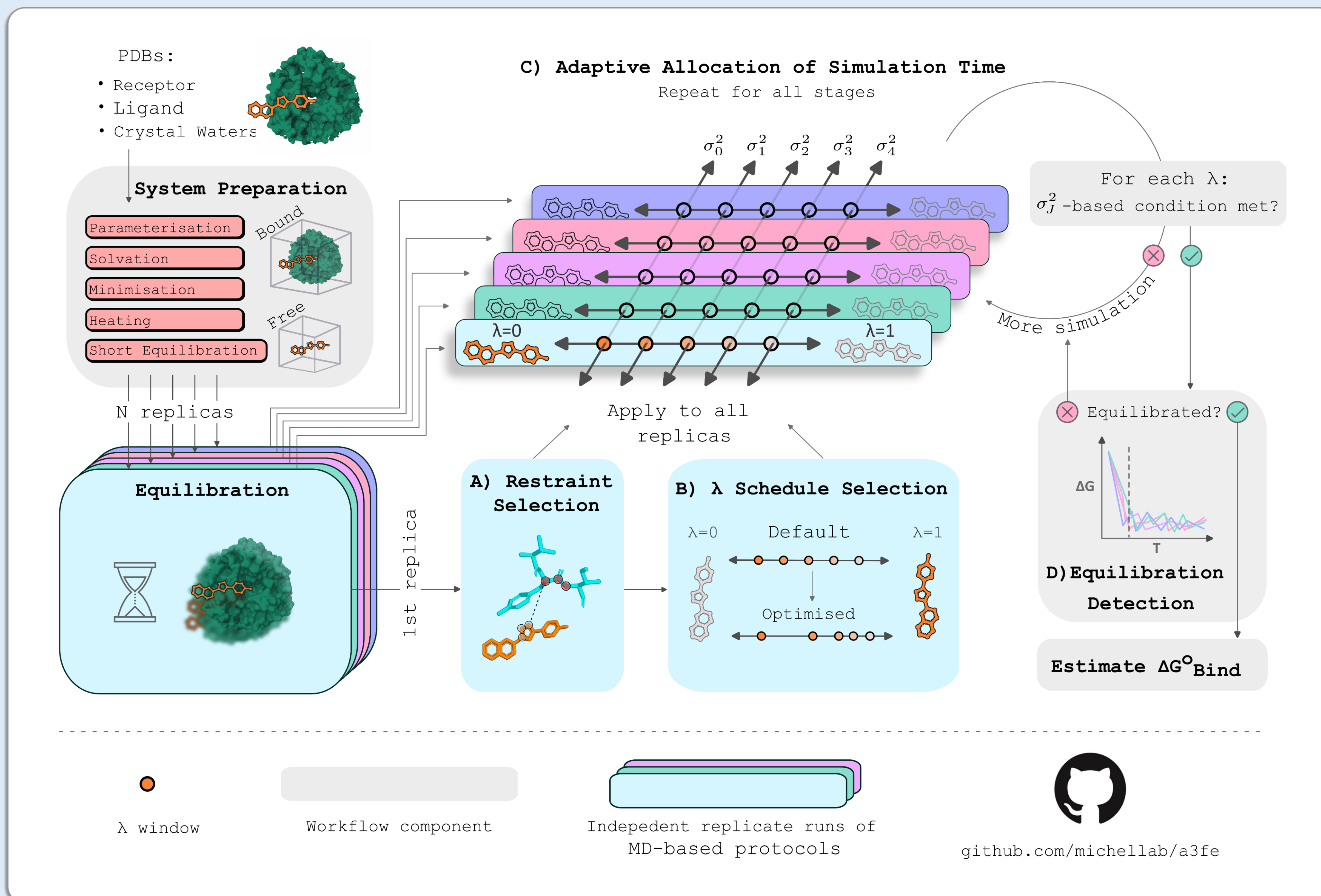
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Alchemical absolute binding free energy calculations must become more efficient

- We implement and test an **open-source fully-automated adaptive workflow for efficient ABFE calculations**
- Based on BioSimSpace (system preparation) and SOMD (Sire/ OpenMM, free energy calculations)



A) Restraints are selected to mimic strong receptor-ligand interactions

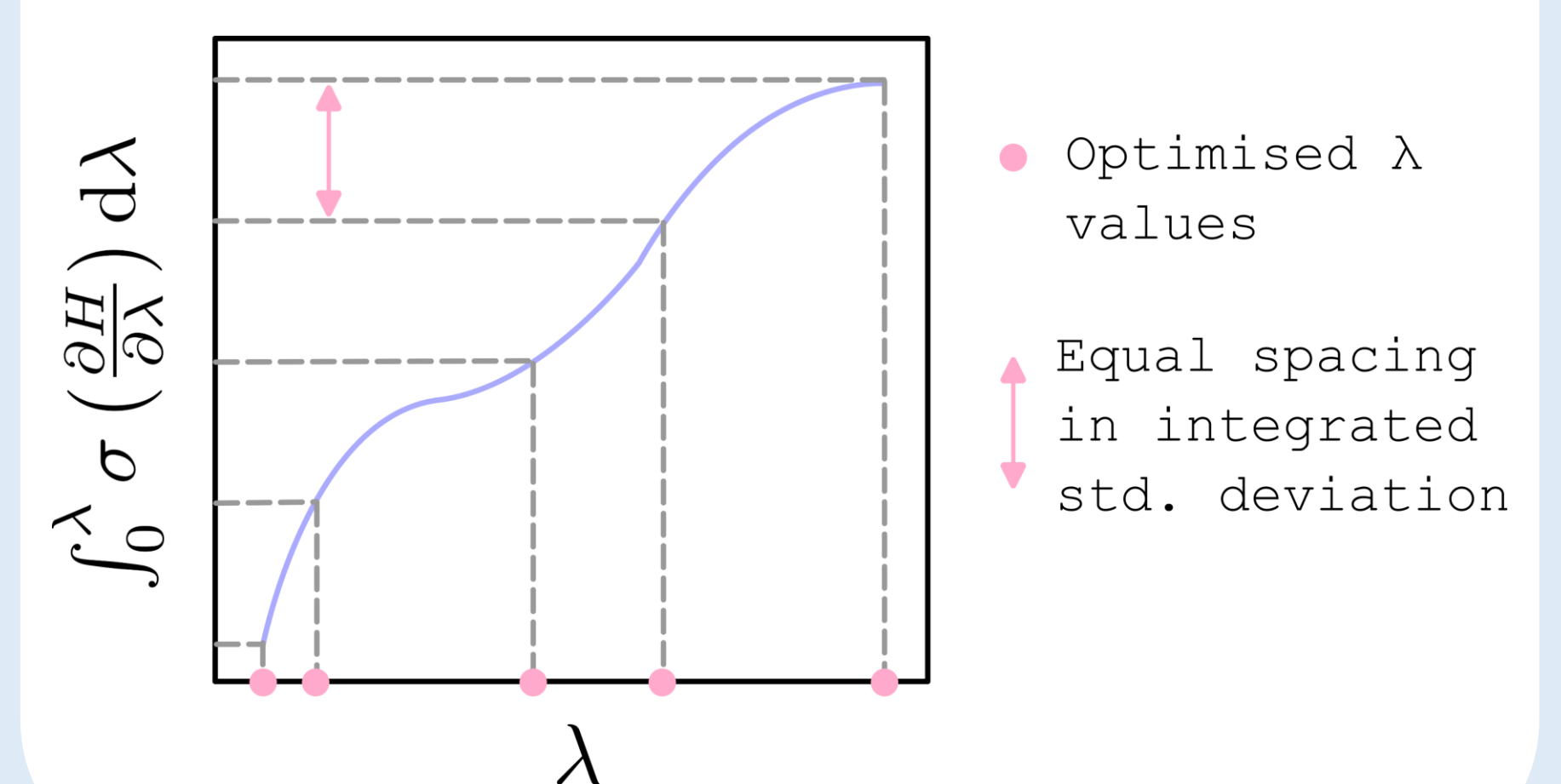
Algorithm

- Candidate sets of anchor points selected from simulation of receptor-ligand complex¹
- Force constants fit to fluctuations observed
- Pick the stable restraints² which most strongly restrict the configurational volume accessible to the ligand once decoupled³

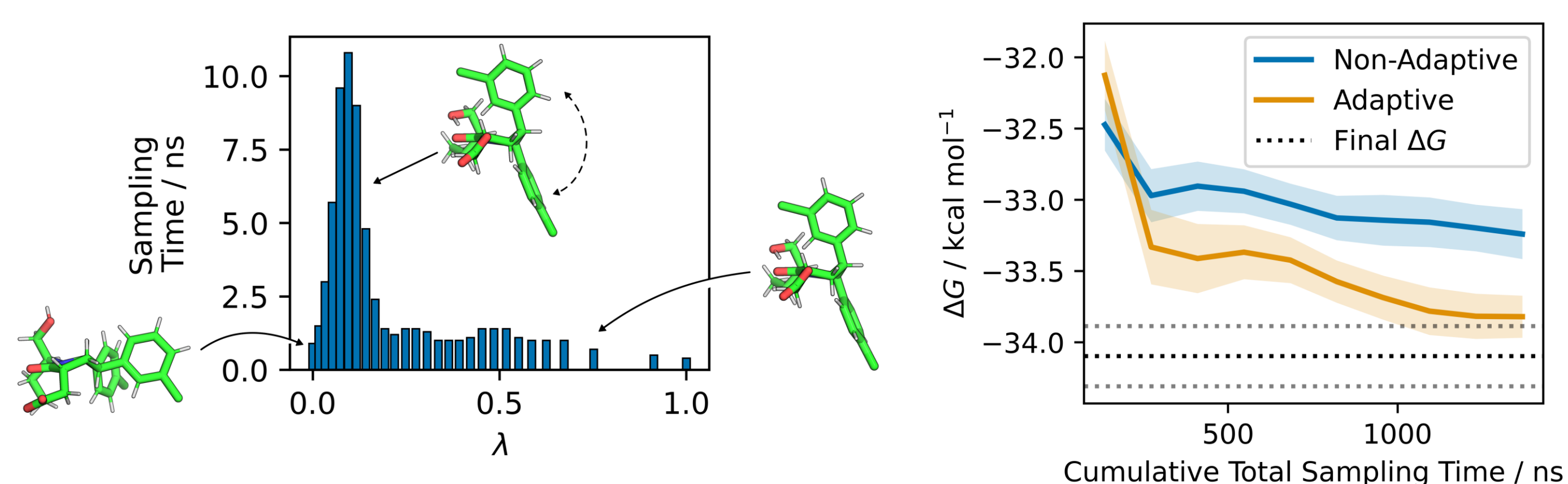
Ensures only two windows required to introduce restraints and free energy cost of introduction is always ~ 1.2 kcal mol⁻¹

B) Lambda windows are spaced based on the standard deviation of the free energy change^{4,5}

Standard deviation of $\partial H / \partial \lambda$ as a function of λ estimated from very short initial simulations



C) Sampling time is allocated to minimise inter-replicate uncertainty



1) Simulation time is allocated according to:

2) This can accelerate convergence to:

$$t_{\text{Optimal}, \lambda} = \sqrt{\frac{t_{\text{Current}, \lambda}}{C}} \sigma(\Delta G)_{\text{Current}, \lambda}$$

Predicted optimal simulation time

Current simulation time

Pre-specified constant

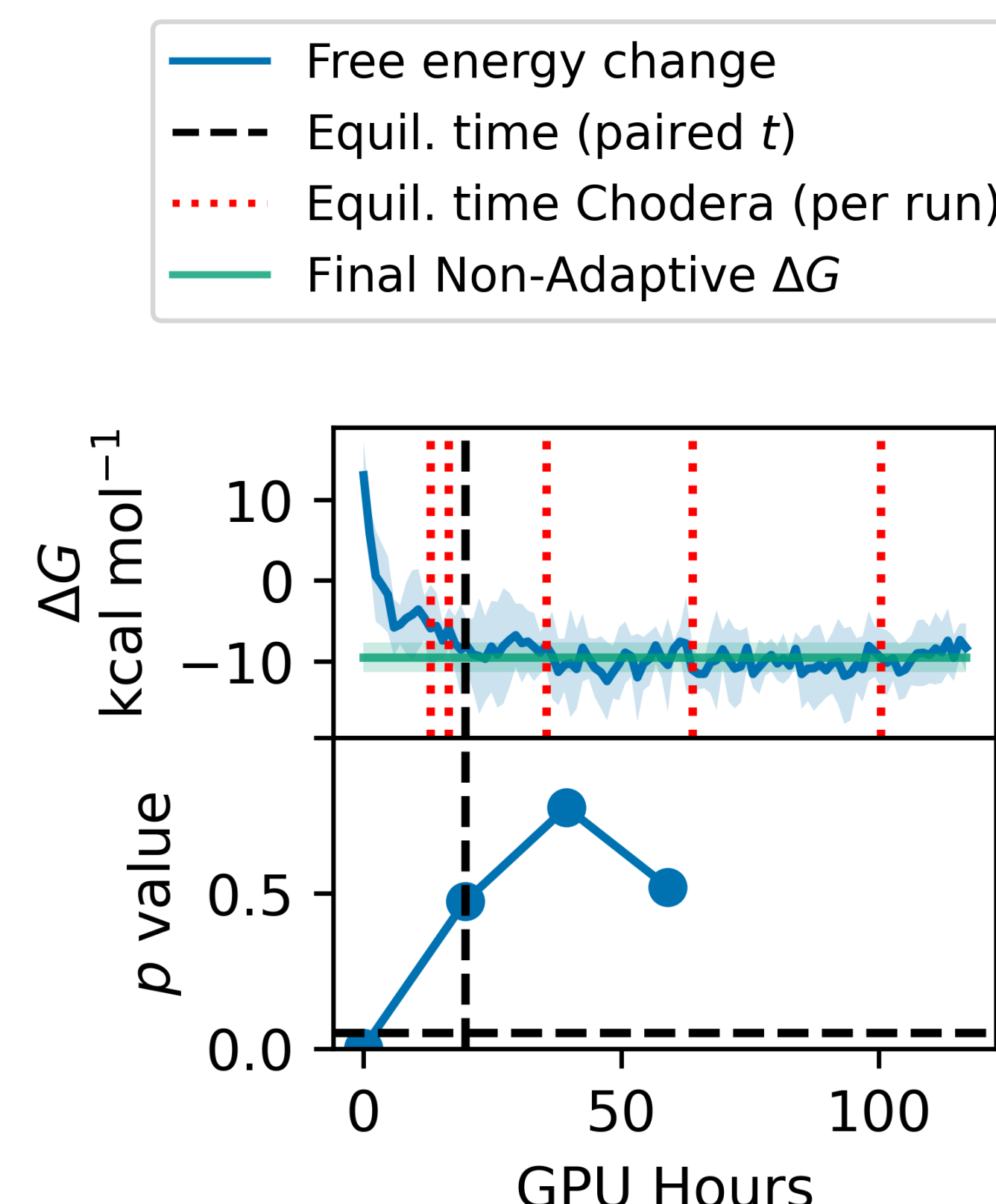
Current uncertainty of per- λ ΔG

Accelerates convergence if the sampling issues occur on a timescale comparable to the simulation duration

D) Equilibration is detected based on an ensemble of replicates

Algorithm

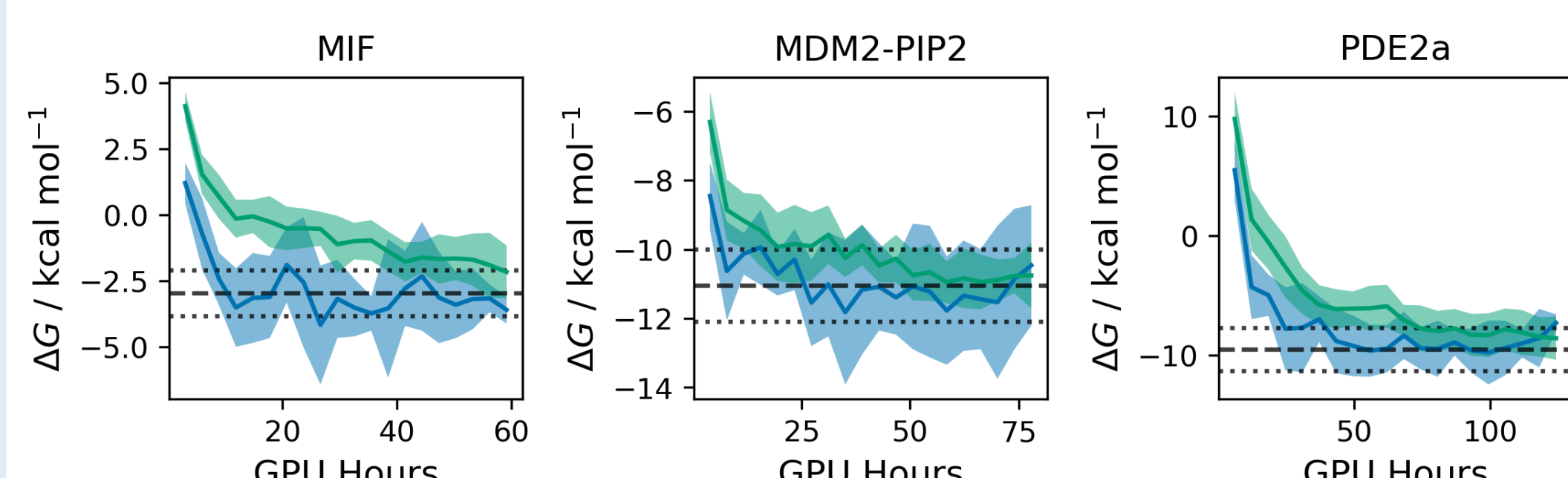
- For each replicate run:
 - Calculate free energy change using first 10 % and last 50 % of data
- Perform a paired t -test on the paired differences between the first and last portions of data
- If $p < 0.05$:
 - Not equilibrated. Discard data from start of runs and repeat
- Else:
 - Assume equilibrated



Paired nature of test increases sensitivity to ensemble trend by ignoring systematic differences between replicates

Overall protocol accelerates equilibration

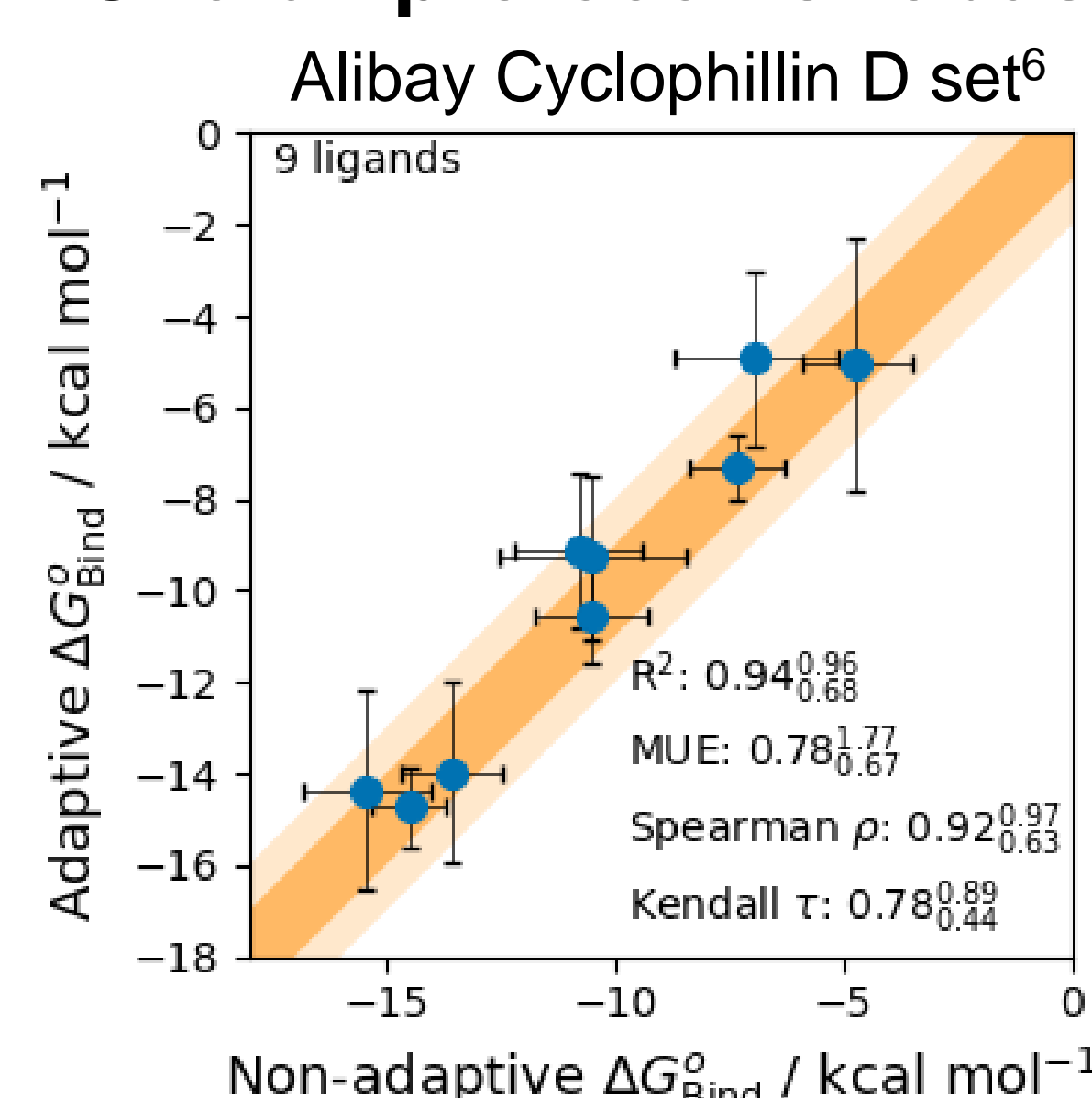
Free energy changes shown for the Lennard-Jones term removal stage, which dominates overall cost and equilibration:



Adaptive protocol:

- Produces equivalent results to non-adaptive protocol
- Often produces faster convergence to long-time result, likely mainly due to wide spacing of λ windows

Overall protocol is robust



- Adaptive protocol: 1/3 less compute time
- Adaptive r^2 to experiment: 0.75 [0.37, 0.91]
- Non-adaptive r^2 : 0.81 [0.46, 0.94]

Conclusions

- Restraint selection, window spacing, and equilibration detection algorithms are simple but robust
- Allocating sampling time according to inter-replicate uncertainty only rarely provides an advantage
- Overall workflow is **fully automated, robust and can accelerate equilibration**

References

- [1] I. Alibay, IAlibay/MDRestrainsGenerator (version 0.1.0) Zenodo 2021.
- [2] S. Boresch et al., *J. Phys. Chem. B*, 2003, **107**, 9535–9551.
- [3] F. Clark, G. Robb, D. J. Cole and J. Michel, *J. Chem. Theory Comput.*, 2023, **19**, 3686–3704.
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- [5] A. Rizzi, Ph.D. Thesis, Weill Medical College of Cornell University, 2020.
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