

# Computing Free Energies with PyBrella Charles Yuan<sup>1</sup> and David Koes<sup>2</sup>

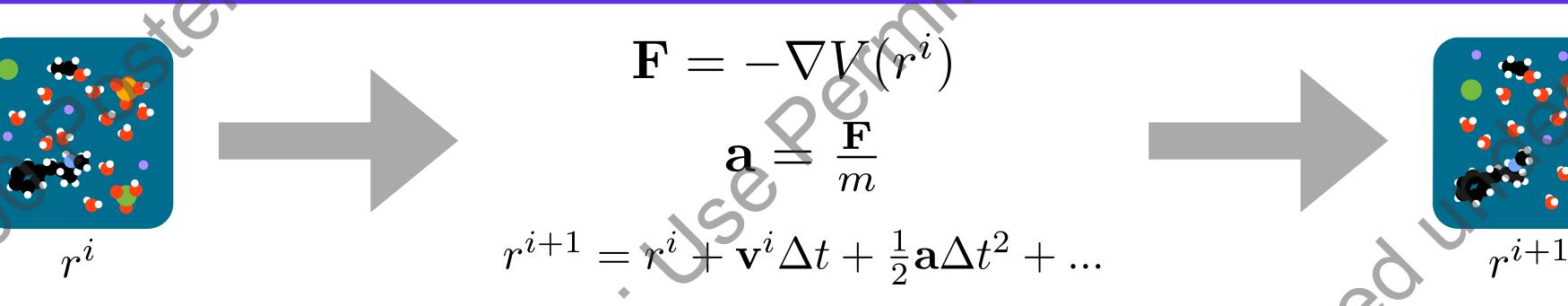
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# Drug Discovery

Modern **drug discovery** is a tedious process that is often limited by the **expense** and **time** of screening a target.

**Computational drug discovery** provides an alternative method for **generating hits** to a target, by allowing:

### Steered Molecular Dynamics



So far, the link between actual efficacy and predicted free energy by PyBrella is insufficient since **more data are necessary** to confirm a correlation, though there is an **expected trend.** 

Conclusions

There are still large **sources of error**, for example in bias inherent in the **SMD trajectory** or **distortion** of the protein or drug **structure**.

**rapid screening**, since computational technology and speed is still advancing;

enormous numbers of compounds, by searching huge databases and using combinatorial chemistry for new molecules;

and **less effort and cost**, by requiring less purchase of compounds and equipment.

 Pharmacophore
 Scoring
 Umbrella

 Matching
 Simulated
 Molecular

 Simulated
 Docking
 Molecular

Computational analysis usually consists of:

spacial and electrostatic docking and matching
of a potential compound;

brief **scoring and ranking** of a compound's predicted efficacy;

and in-depth evaluation of individual molecules

**Molecular dynamics (MD)** is the **simulation of** molecular **motion** that accounts for **positions** and **trajectories**. We used the MD package **AMBER12** to simulate the **removal** of molecules from proteins by **steered molecular dynamics (SMD):** 

Proteins from the CSAR dataset were prepared with various ligands encapsulated in a periodic **20 Å water box,** then were allowed to **minimize** and **equilibrate** for 1 nanosecond of simulation time.

A **pair of atoms** was selected from the protein and the ligand in preparation for applying a pushing force to ensure a **free exit path** for the ligand.

**Constraints** were added to the protein **backbone** to disallow distortion, and a **simulated force** was applied on the protein and ligand until they were **20** Å apart, allowing the ligand to escape into free solution.



Images of the simulation process, from the starting water box to the final exit of the molecule

Umbrella Sampling with PyBrella

There is a **surprisingly large reliance** on the choice for the umbrella sampling **force constant**, suggesting odd behavior by WHAM since the PMF **should ideally be independent** of the force constant.

The availability of **more data** in the longer runs is **not necessarily better,** since several compounds received uncharacteristic PMFs, but this determination is still inconclusive.

## Next Steps

Investigate use of **other force constants** and begin to develop a **general method** to determine force constants for each compound

Test **even larger collections** of shorter runs instead of longer runs to improve speed and reduce error

Expand to more compounds in the CSAR dataset

deemed of sufficient quality.

This last step is particularly **resource and timeintensive**, making it often the **limiting factor** in drug screening.

We seek to develop an **effective and economical** method to reliably correlate the actual **rate of dissociation** with a calculated trajectory of system energy, or **potential of mean force (PMF).** 

Choosing to avoid simple single-point calculations allows us to consider the **full protein-water system** and increase overall prediction accuracy.

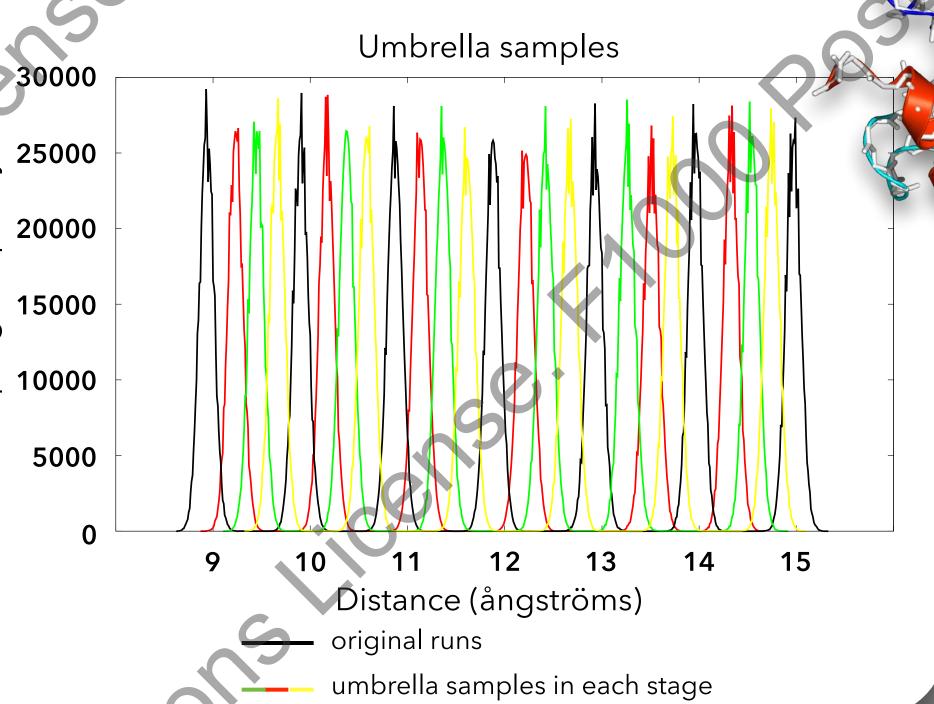
The study uses the dataset **CSAR2012**, which includes numerous **test protein targets** and associated **known active/inactive compounds**.

We used the technique of **umbrella sampling**, which: uses **snapshots** throughout the SMD trajectory as starting points for **new simulations**;

requires **local force restraints,** to weakly hold the compound in place, and **coverage of all distances;** 

and is **slower** if implemented in a naïve fashion.

To implement the method, we developed the program **PyBrella**, which extracts **frames** from SMD output, assigns **constraints**, controls **simulation**, then dynamically adds **new runs** so that all distances receive sampling to a **minimum threshold**, avoiding **extraneous sampling.** Then, it compares the **first 80% to the last 20%** of each run to determine **convergence**, and **extends runs** that have not converged. Finally, we use the **Weighted Histogram Analysis Method (WHAM)** to recover the PMF of the reaction.



A lower dissociation constant, corresponding to a

stronger attraction, does indeed cause a higher PyBrella

Apply constraints on backbone to ensure that the compound exits instead of bending the protein

Structure: PDB 4G6N

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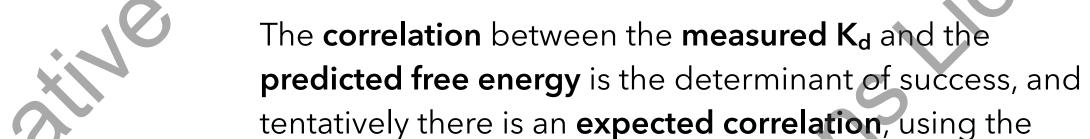
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References



Two main parameters have been under investigation:

the usage of **longer runs** (10 ns) or **collections of shorter runs** (1 ns) for every distance position;



short runs with k=5.

**PMF** across four compounds.



Analysis is shown for a limited set of CSAR. Generally, long runs produce **unsatisfactorily flat and uninformative** PMF curves. Shorter runs yielded higher PMFs and more appropriate energy curves. Likewise, the smaller force constant k=5 showed **drastically improved** results over k=50, with much higher PMFs. Since CS245, CS260, and CS262 all have **high actual affinities**, a **higher PMF** is expected.

Potential of mean force: different force constants Potential of mean force: long and short runs Measured K<sub>d</sub> and PyBrella predicted free energy endpoint (kcal/mol) • CS260 — CS262, k=5 CS262 short CS260, k=5CS260 short ----- CS260, k=50 ----- CS262, k=50 - CS262 long CS260 long . . . . . . . C\$245 ° 6 (kcal/mol) (kcal/mol) R<sup>2</sup>=0.7251 5 5 CS262 o ergy 3 <u>уруза</u> M — CS245, k=5 — CS1 short Ene 2 ----- CS245, k=50 ----- CS1 long Ш 2 50 1E-05 70 1E-08 1E-07 1E-06 1E-04 1E-03 1E-09 30 50 60 20 30 40 40 OctetRed average K<sub>d</sub> Distance (ångströms) Distance (ångströms)

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