3.1.1 Basics: Analyzing molecular simulations

- This module will period will consist of a mini-lecture and jupyter notebook exercises about
 - equilibration versus production
 - aligning trajectories
 - visualizing trajectories in VMD
 - root mean square properties
 - estimating thermodynamic expectations
- After this module, you should be able to
 - answer the following questions
 - what is the purpose of ignoring an equilibration period?
 - what is a potential of mean force?
 - what is true regarding the joint and marginal probabilities of independent events?
 - perform the following tasks
 - load a trajectory into VMD
 - calculate the RSMD and RMSF of a molecular simulation
 - estimate the average of an observable in a molecular simulation

What is MD used to calculate?

- MD simulations may be used to
 - predict events or sequence of events that are physically possible
 - estimate statistical averages of
 - configurational properties, e.g.
 - average distance between two residues
 - histogram of an angle between three domains
 - populations of certain conformations
 - rates
- Statistical estimation is based on the assumption of ergodicity that the time average is equal to the ensemble average

Some python packages analyzing MD simulations

- juypter for interactive coding notebooks
- pandas for data analysis
- mdanalysis for loading and analyzing MD trajectories
- pymbar for calculating free energies. also contains equilibration detection
- For examples of using these packages, see
 - github.com/daveminh/Chem456/static_files/tutorials/ubq_wat-md/3-analysis for a simulation of ubiquitin
 - <u>github.com/CCBatIIT/modelingworkshop/exercises/mpro/RIKEN_trajectory/analysis</u> for a simulation of MPro

What is equilibration?

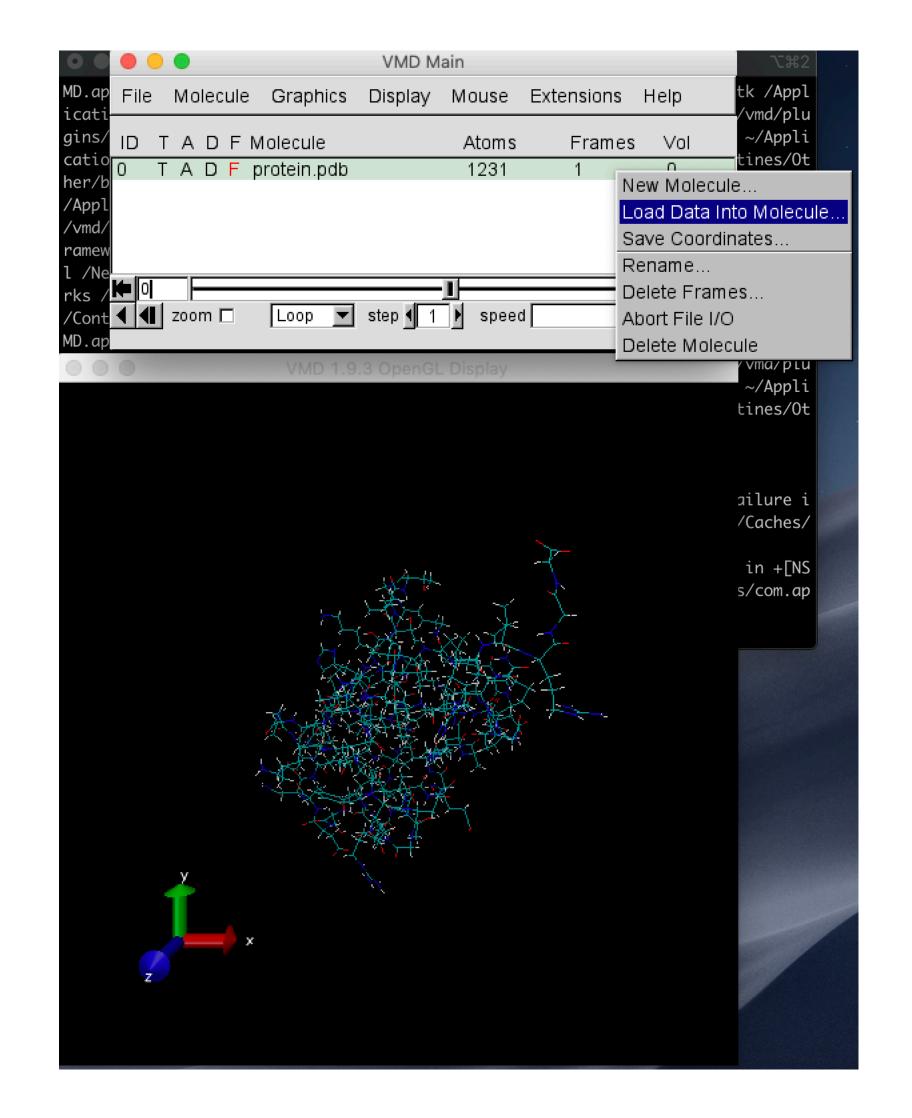
- The early part of a simulation is biased by the initial configuration
 - macromolecule structures start in a local minimum
 - water and ions are placed somewhat arbitrarily
 - box size is probably too large
- Equilibration is the time a system takes to reach a representative configuration
- Generally, simulation results from the equilibration period are ignored
 - The samples actually used to calculate averages are known as the production
 - It is not completely essential to ignore equilibration
- See <u>Equilibration.ipynb</u>, which illustrates these points for a simulation of ubiquitin

How is equilibration time determined?

- Arbitrarily
- Once a key property, e.g. the root mean square deviation (RMSD), is stabilized
- By maximizing the effective sample size [Chodera, 2016]
 - A short equilibration leads to a long estimate of the time for the sample to be independent
 - A long equilibration reduces the number of samples
- Not at all
- Equilibration time
 - may look different for different properties
 - if properties are independent, slow equilibration of one may not affect estimation of another

An unaligned trajectory of ubiquitin

- can be visualized by
 - loading a model into VMD
 - loading the trajectory into the model
- molecules, especially water, can be split across a periodic box
- you probably don't need to see all water
- the molecule freely diffuses

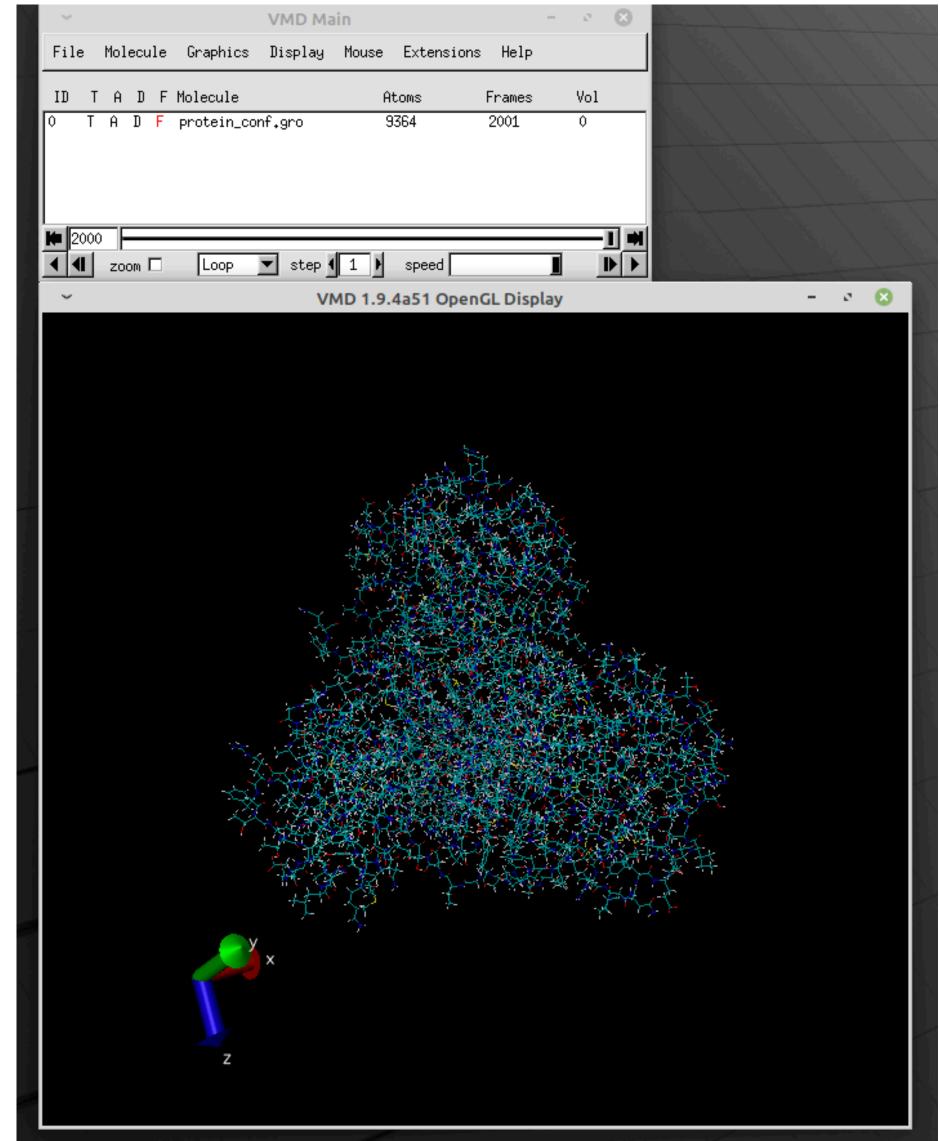


Structural alignment

- Other than identifying equilibration time, structural alignment is usually one of the first tasks of MD analysis
- Why?
 - In most MD simulations, molecules freely diffuse around the box
 - We are usually
 - uninterested in the overall translation and rotation,
 - interested in fluctuations relative to the macromolecule
- Alignment is often based on a rigid-body translation and rotation to minimize the root mean square deviation (RMSD)
- See <u>Alignment.py</u>, which performs a structural alignment for a series of simulations of ubiquitin and outputs a trajectory of the protein by itself.

Visualizing the main protease of SARS-CoV-2

- Let's visualize a trajectory of the main protease of SARS-CoV-2
- You can download it onto your virtual machine with the following script:
 - (base) chemuser@ChemBox:~/Documents/ modelingworkshop/exercises/mpro/ RIKEN_trajectory\$ source get_trajectory.sh
- After starting VMD,
 - create a molecule from sarscov2-10921231structure/protein_conf.gro, a Gromacs GRO file
 - load data into the molecule from traj_protein_snap_every1ns/ protein_snap_every1ns_NtoMus.xtc, a Gromacs XTC Compressed Trajectory



Root mean square analysis

RMSD
$$(v, w) = \sqrt{\frac{1}{n} \sum_{i=1}^{n} ||v_i - w_i||^2}$$

- Root mean square deviation (RMSD)
 - describes the difference between two structures
 - usually based on a subset of atoms
 - *i* is an index over atoms
- Root mean square fluctuation (RMSF)
 - describes the fluctuations of a specific atom, e.g. alpha carbon, over the course of a simulation
 - usually described per residue, identifying relatively flexible regions of a protein
 - *i* is an index over configurations
- Both require structural alignment

Exercise: RMS calculations with MDAnalysis

- Load and run the jupyter notebook https://github.com/CCBatIIT/

 modelingworkshop/exercises/mpro/RIKEN_trajectory/analysis/RMS.ipynb,
 which shows different types of RMS analysis for the RIKEN simulation of the
 main protease of SARS-CoV-2
- Based on the notebook
 - what is a reasonable number of frames to discard as equilibration?
 - how do the RMSD between alpha carbons and backbone atoms compare?
 - how does the flexibility of the protein vary between monomers
 - which part of the protein is most flexible?
- Based on the included examples and the MDAnalysis <u>atom selection</u>
 <u>language</u>, write your own code snippet to calculate the RMSD to backbone atoms for residues 1-300

Solution to code snippet

```
In [19]: # Evaluate the RMSD of backhone carbons
         # with respect to the reference frame
         from MDAnalysis.analysis import rms
         R = rms.RMSD(sim, ref, select="backbone and resid 1-300")
         R.run()
         rmsd_backbone_to300 = R.rmsd
         plt.plot(rmsd_backbone_to300[:,1]/1000, rmsd_backbone1_70[:,2])
         plt.xlabel('Time (ns)')
         plt.ylabel('RMSD of backbone atoms for residues 1-300 ($\\AA$)');
             3.0
                                              8000
                                                     10000
                                 Time (ns)
```

Thermodynamic Expectations

- One key reason to do a molecular simulation is to estimate expectation values
- The definition of an expectation value of the observable A is,

$$\langle A \rangle = \int A(x)p(x)dx$$
, where p(x) is the probability of x

• For an unbiased MCMC simulation, the typical way to estimate an expectation value is the sample mean, $\bar{A} = \frac{1}{N} \sum_{n=1}^{N} A(x_n)$, over N samples after

equilibration

What A?

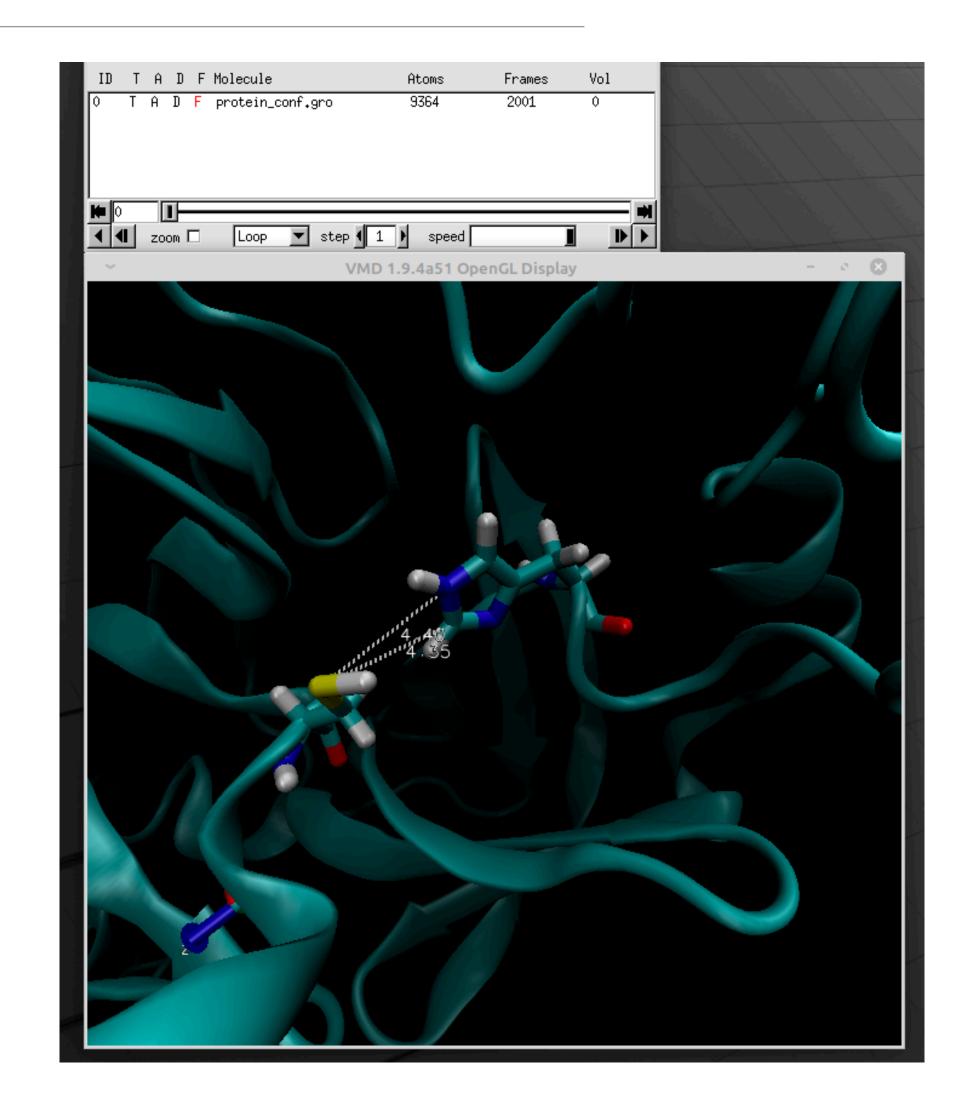
- The observable A can be anything that can be measured from a configuration
 - distance between two atoms
 - an angle between two protein domains
 - an indicator function H(x) that is 1 if $x_1 < x < x_r$ and 0 otherwise
 - this expectation is the probability of x being in this range
 - it is useful for a histogram

Potential of mean force

- A potential of mean force
 - $f(x) = -RT \log \langle H[z(x)] \rangle$
 - H[.] is an indicator function
 - z(.) is an order parameter, a function of the configuration
 - x is the configuration
 - is the log probability of a certain order parameter
 - a potential energy whose gradient would give rise to the mean force on a particle
- A histogram is a simple way to estimate the expectation for a set of bins
- Let's be more concrete and practical...

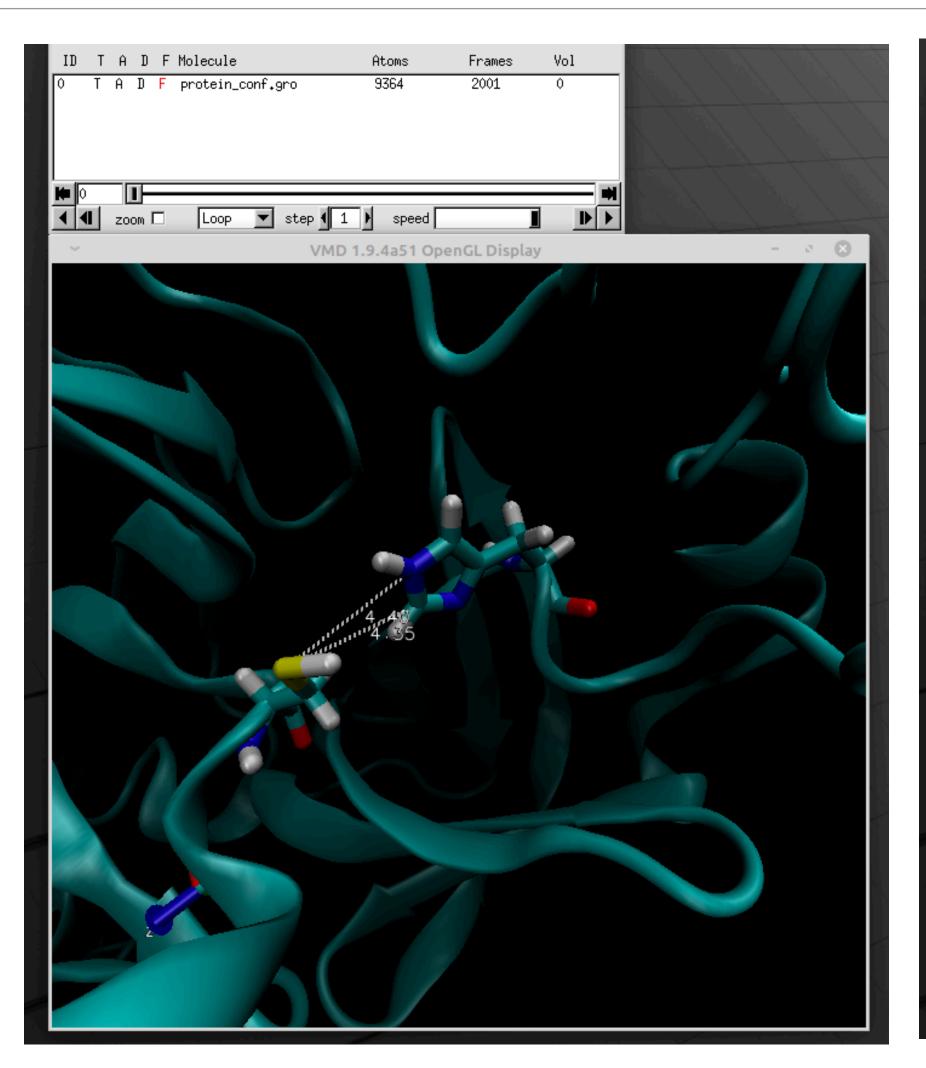
The catalytic dyad of MPro

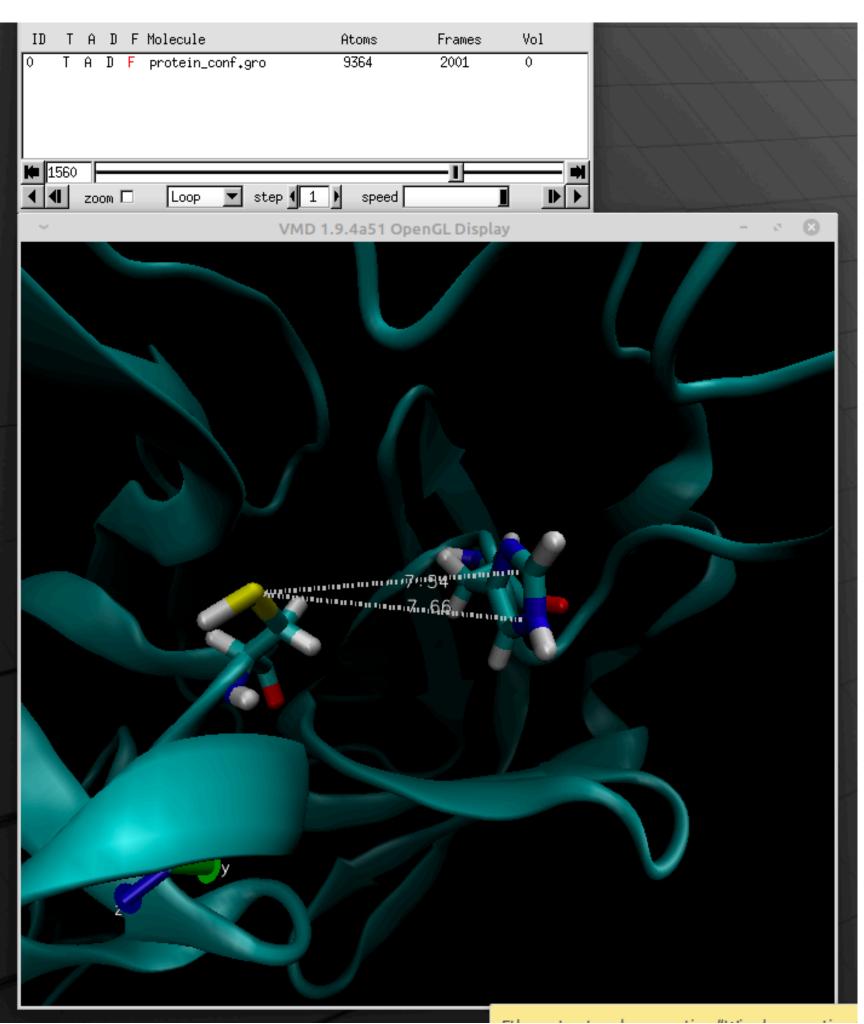
- The main protease of SARS-CoV-2 is
 - a dimer
 - believed to have half-of-the-sites activity
- Chen et al (2006)
 - ran 4 10 ns simulations of MPro from SARS-CoV-1
 - reported the fraction of frames with hydrogen bonds between Cys145 SG and His41 ND1/NE1
 - suggested that during dimerization one subunit is active and the other inactive
- At 10 μs, the RIKEN simulation is ~1000x longer!
- Exercise: Use VMD to label the distance between Cys145
 SG and His41 ND1/NE1



An alternate His41

- In some MD simulation frames, the histidine ring seems to flip away from Cys 145. This could be important!
- Did you notice anything else that I didn't?





Exercises: Distances between the catalytic dyad

- Load and run the jupyter notebook https://github.com/CCBatIIT/modelingworkshop/exercises/mpro/RIKEN_trajectory/analysis/dyad-stability.ipynb
- Answer these questions:
 - Are probability densities of the dyad distance the same in each monomer?
 - Is there any apparent correlation between subunits in the dimer?
 - Are probabilities of forming a short contact in each subunit independent? Write a code snippet based on these hints
 - Use `d_cutoff = 4.25`
 - `pA = np.sum(d_monomerA<d_cutoff)/len(d_monomerA)` is the fraction of frames with a short contact in monomer A
 - `pAB = np.sum(np.logical_and(d_monomerA<d_cutoff, d_monomerB<d_cutoff))/ len(d_monomerB)` is the fraction of frames with a short contact in both monomer A and monomer B
 - If two events are independent, p(A, B) = p(A)p(B).

Solution to the code snippet

Joint probabilities of short dyad distances

```
In [18]: d_cutoff = 4.25
    pA = np.sum(d_monomerA<d_cutoff)/len(d_monomerA)
    pB = np.sum(d_monomerB<d_cutoff)/len(d_monomerB)
    pAB = np.sum(np.logical_and(d_monomerA<d_cutoff, d_monomerB<d_cutoff))/len(d_monomerB)
    print('The fraction of frames with a short dyad distance')
    print('in monomer A is', pA)
    print('in monomer B is', pB)
    print('in both monomers is', pAB)
    print('If the monomers were indepedent, the probabilities would be', pA*pB)

The fraction of frames with a short dyad distance
    in monomer A is 0.0475
    in monomer B is 0.0824
    in both monomers is 0.005
    If the monomers were indepedent, the probabilities would be 0.003914</pre>
```

How conclusive are these results?

Review

- what is the purpose of ignoring an equilibration period?
- what is a potential of mean force?
- what is true regarding the joint and marginal probabilities of independent events?

References

- Chodera, J. D. A Simple Method for Automated Equilibration Detection in Molecular Simulations. Journal of Chemical Theory and Computation 2016, 12 (4), 1799–1805. https://doi.org/10.1021/acs.jctc.5b00784.
- Chen, H.; Wei, P.; Huang, C.; Tan, L.; Liu, Y.; Lai, L. Only One Protomer Is
 Active in the Dimer of SARS 3C-like Proteinase. J. Biol. Chem. 2006, 281 (20),
 13894–13898. https://doi.org/10.1074/jbc.M510745200.

Some other software

- MDTraj: http://mdtraj.org/1.9.3/index.html
- ProDy: http://prody.csb.pitt.edu/tutorials/trajectory_analysis/trajectory.html