

Day 1 – Lecture 2

Bridging Structure and Function, Experiments and Computations

Ivet Bahar and Pemra Doruker

Department of Computational and Systems Biology
School of Medicine, University of Pittsburgh, Pittsburgh, PA 15260



Summary

1. Theory

- a. Gaussian Network Model (GNM)
- b. Anisotropic Network Model (ANM)
- c. Resources/Servers/Databases (ProDy, DynOmics)

2. Bridging Sequence, Structure and Function

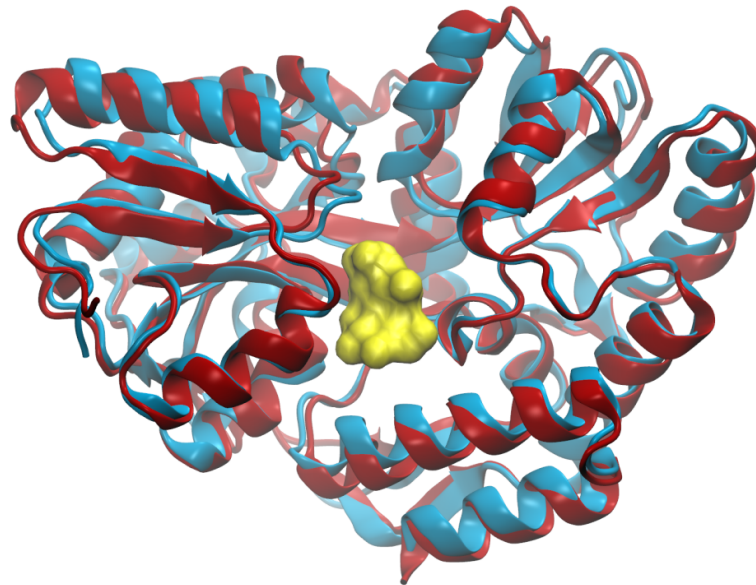
- a. Ensemble analysis and functional modes of motion
- b. Combining sequence and structure analyses – signature dynamics
- c. Modeling membrane proteins and lipid environment with ANM

3. Allostery and druggability

- a. Essential site scanning and allosteric pocket prediction
- b. Druggability simulations

Proteins exploit pre-existing soft modes for their interactions

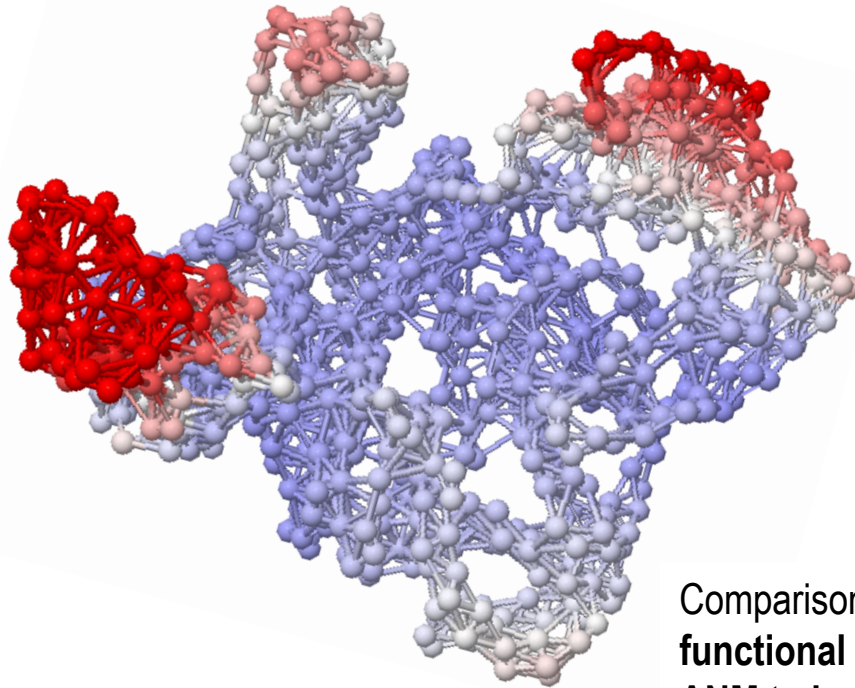
Structural changes involved in protein binding correlate with intrinsic motions in the unbound state



maltodextrin binding protein
Unbound/Bound

Allosteric changes in conformation

Elastic Network Models are particularly useful for exploring the cooperative motions of large multimeric structures



HIV Reverse Transcriptase (RT)

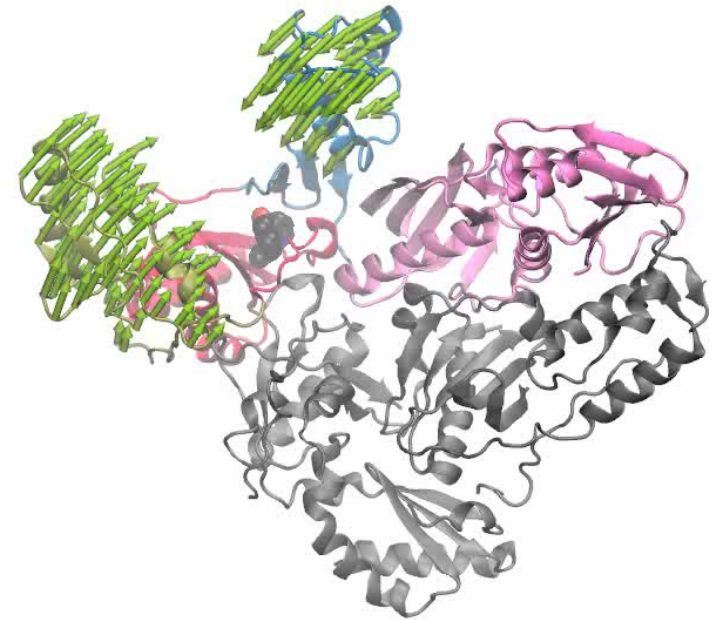
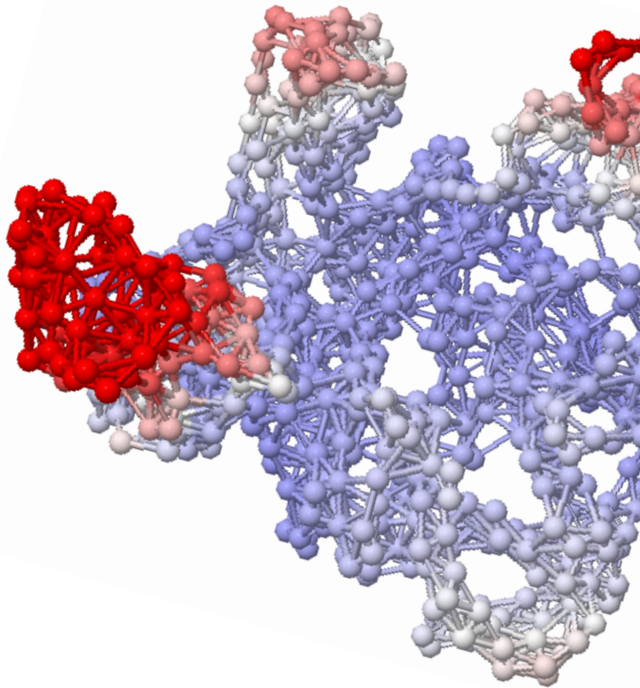
Red: most mobile

Blue: most constrained

Comparison with experimental data shows that **the functional movements are those predicted by the ANM to be intrinsically encoded by the structure**

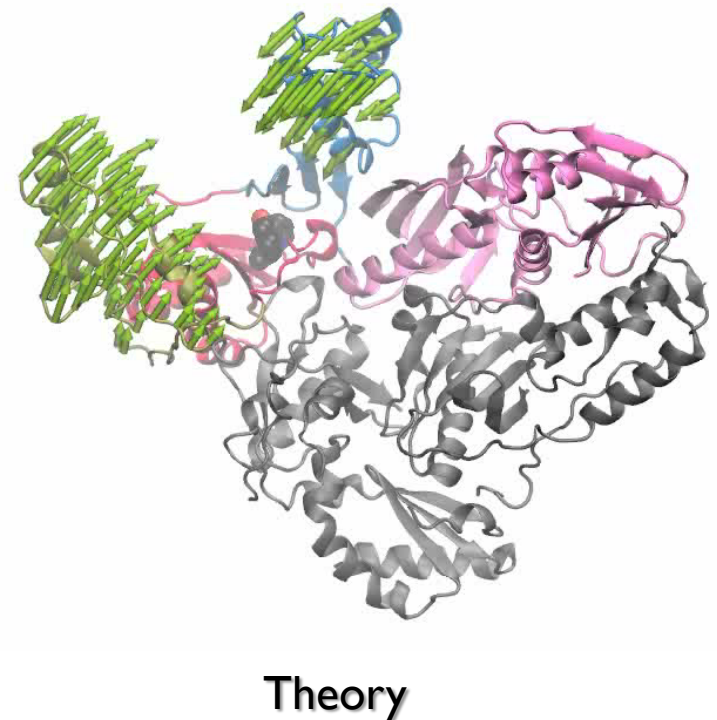
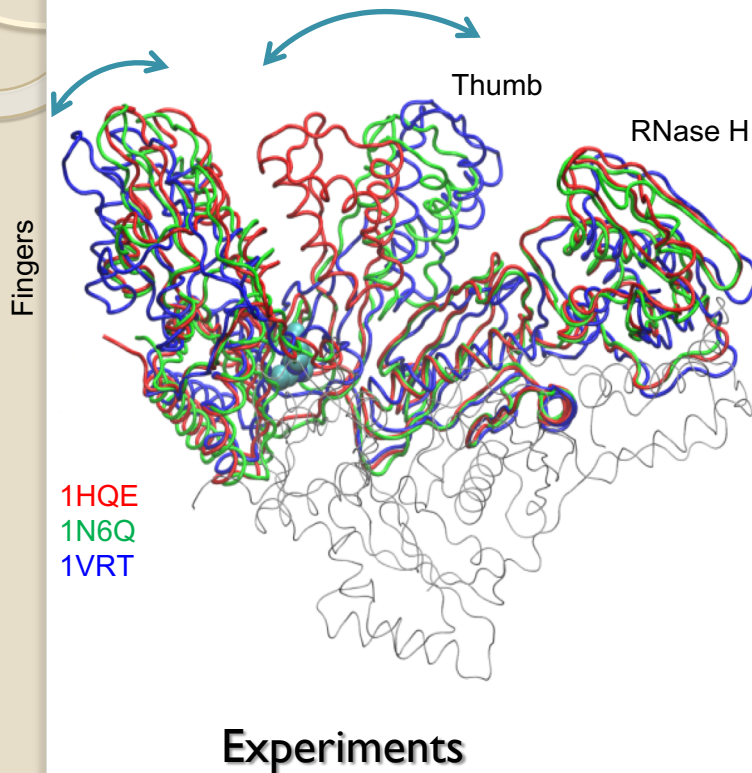
Allosteric changes in conformation

Elastic Network Models are particularly useful for exploring the cooperative motions of large multimeric structures



Comparison with experimental data shows that **the functional movements are those predicted by the ANM to be intrinsically encoded by the structure**

Induced Dynamics or Intrinsic Dynamics?

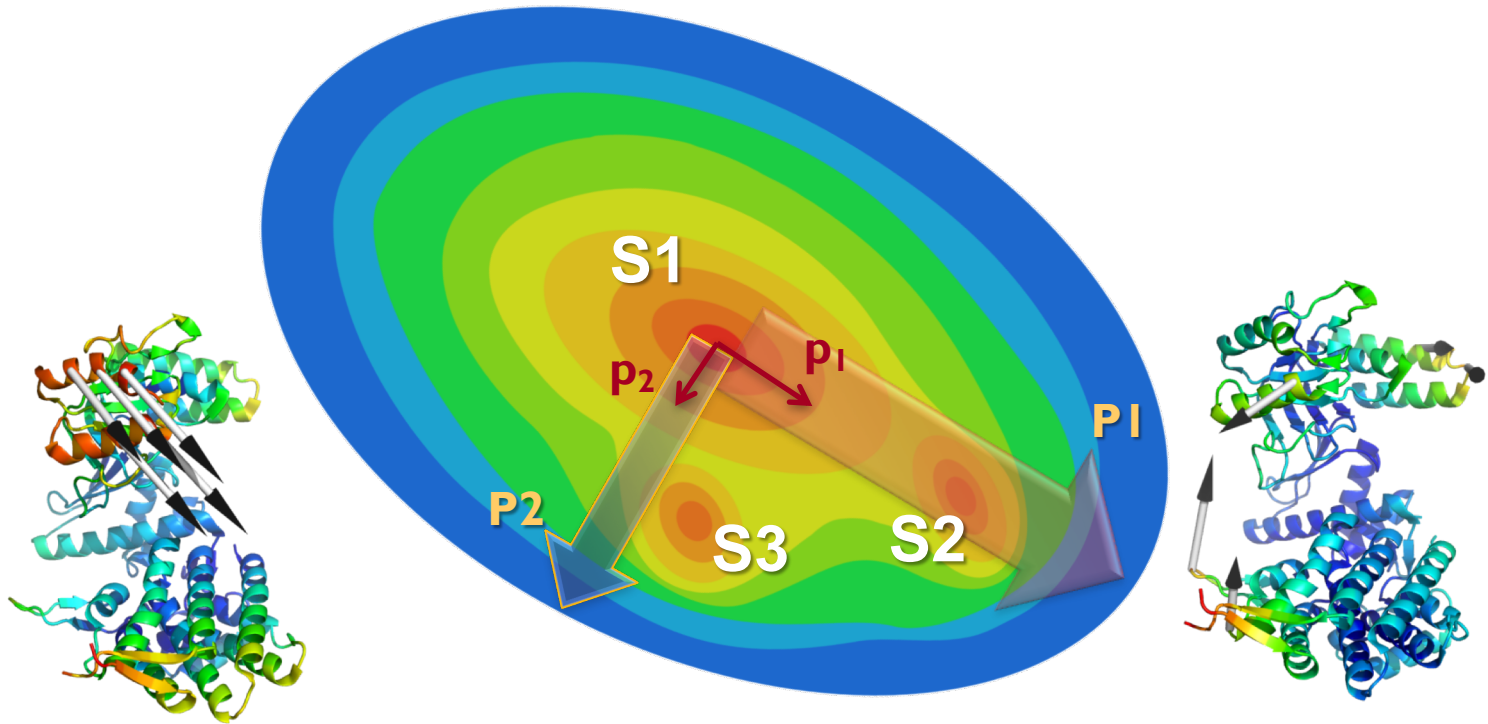


<http://www.youtube.com/watch?v=1OUzdm68YY>

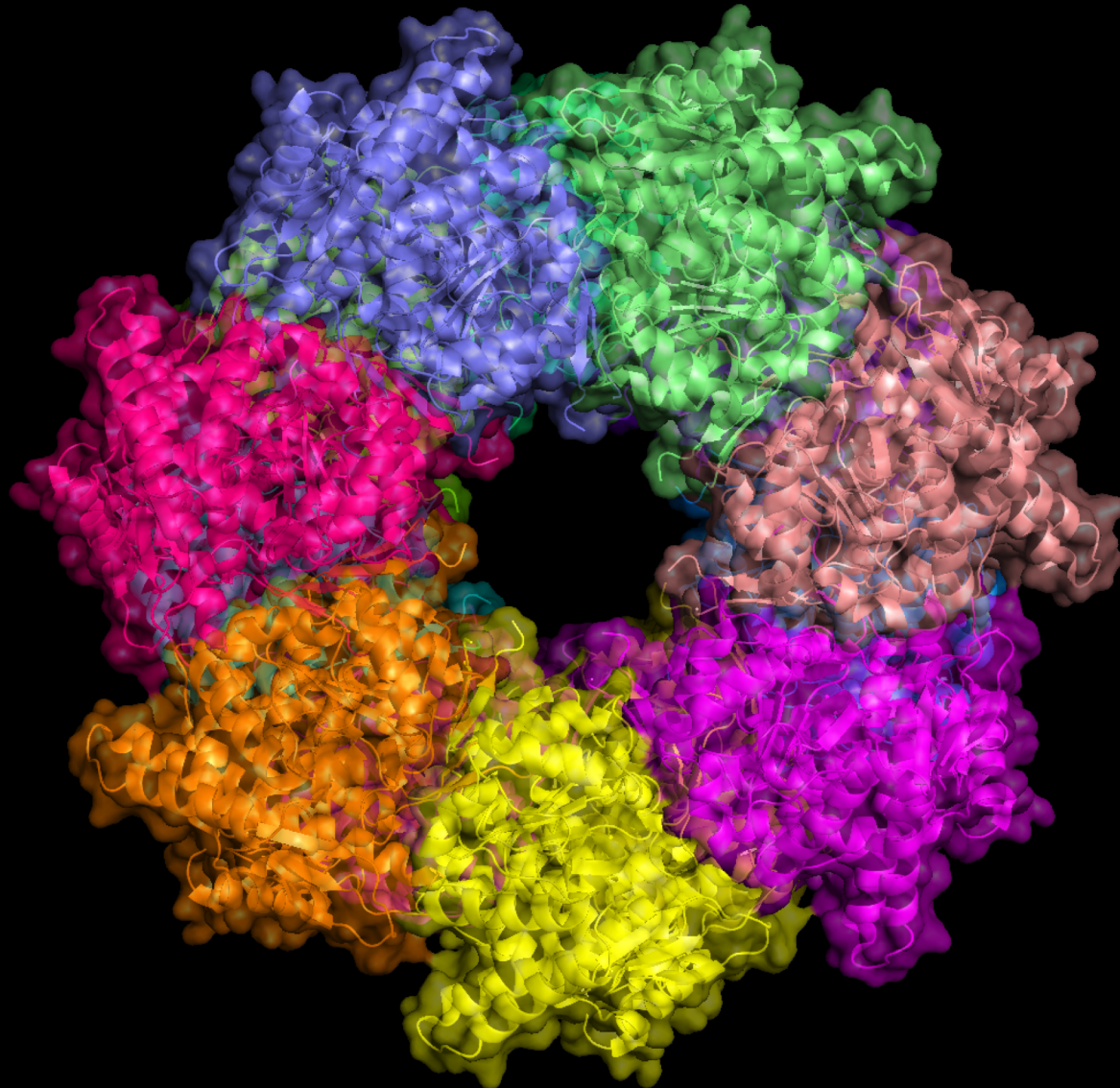
References:

Bakan & Bahar (2009) PNAS 106, 14349-54.

Substates may be identified along soft modes

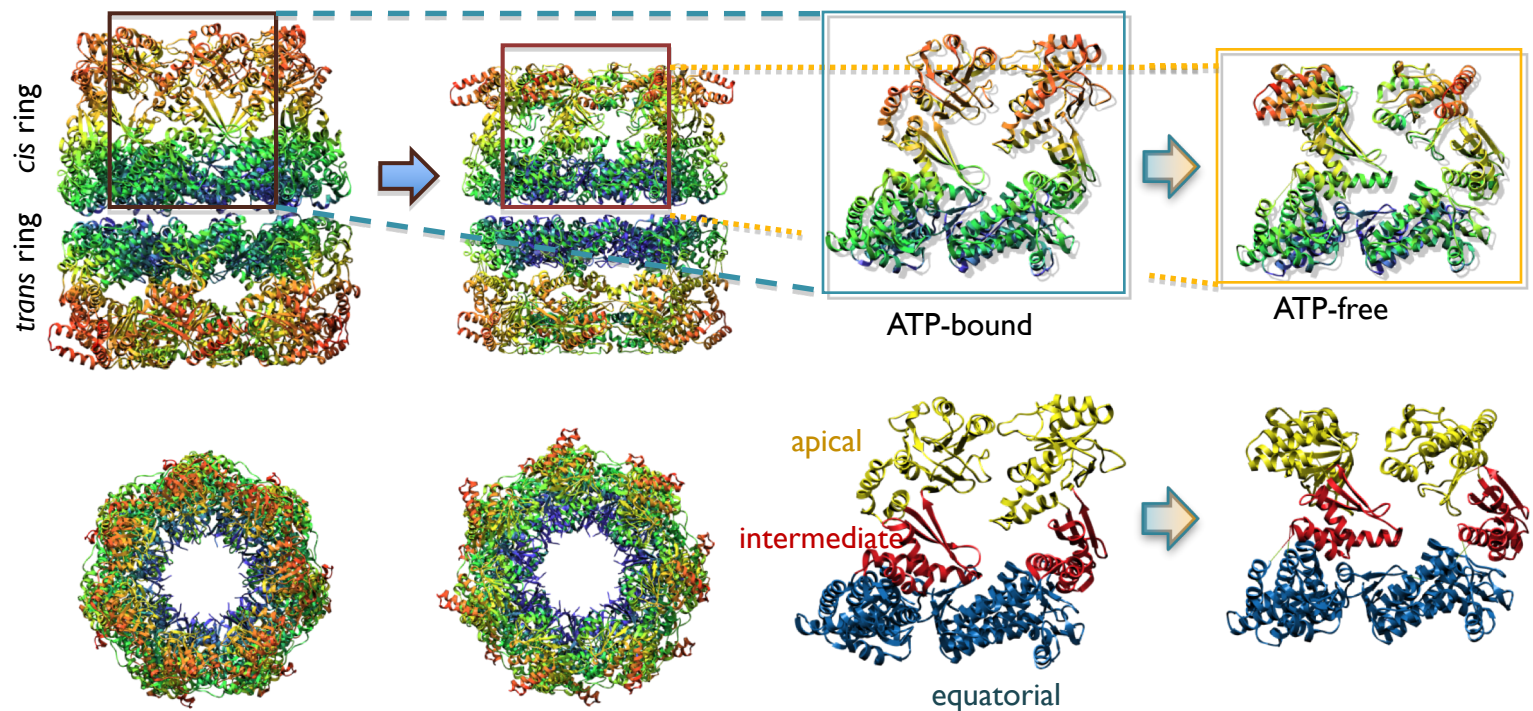


Bacterial chaperonin GroEL: an allosteric machine



GroEL Allosteric Dynamics

Passage between the R and T states



See...

Computations

ANM yields a series of $3N$ dimensional **deformation vectors**

Mode 1 (slowest mode)

Mode 2

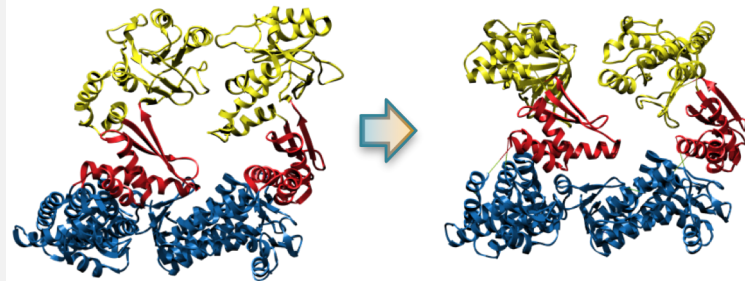
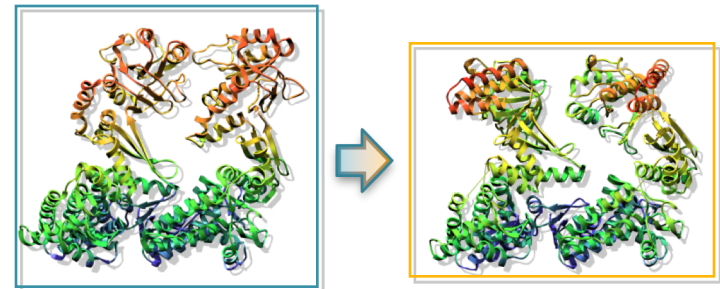
Mode 3

....

Mode $3N-6$ (fastest mode)

Given by ANM eigenvectors $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3, \dots, \mathbf{v}_{3N-6}$, with respective frequencies proportional to $\kappa_1, \kappa_2, \kappa_3, \dots, \kappa_{3N-6}$

Experiments

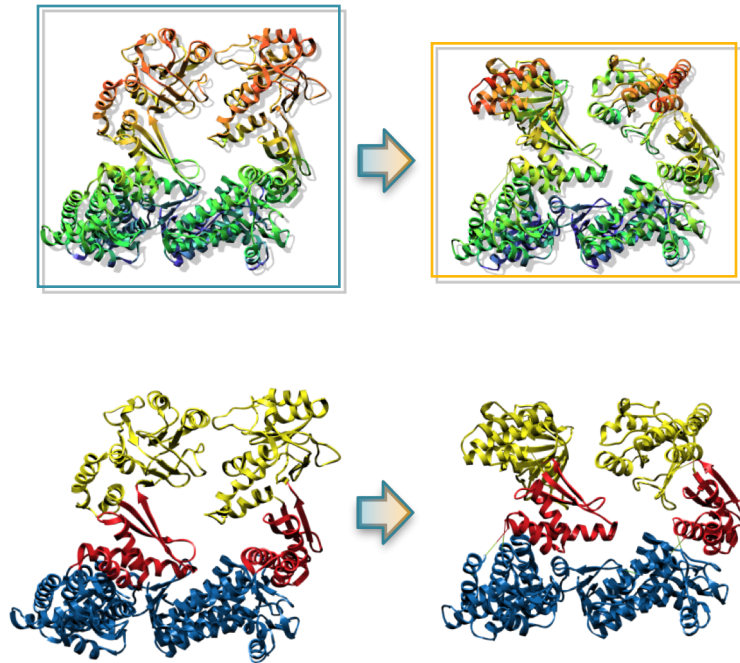
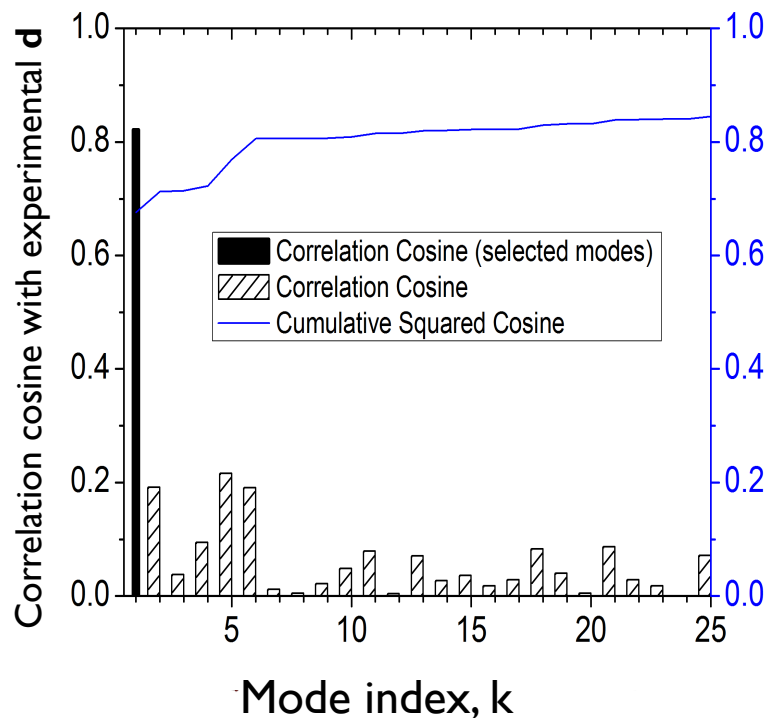


$$\mathbf{d} = [\Delta x_1 \ \Delta y_1 \ \Delta z_1 \ \dots \ \Delta z_N]^T$$

See...

What is the overlap between computations and experiments?

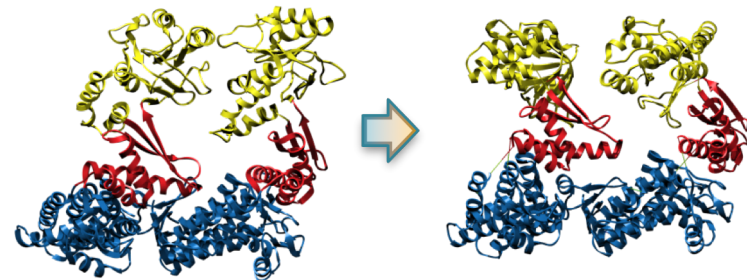
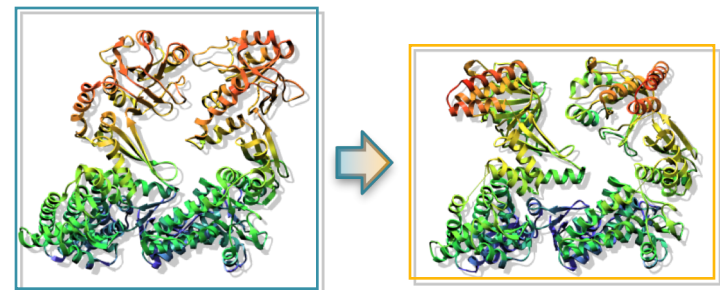
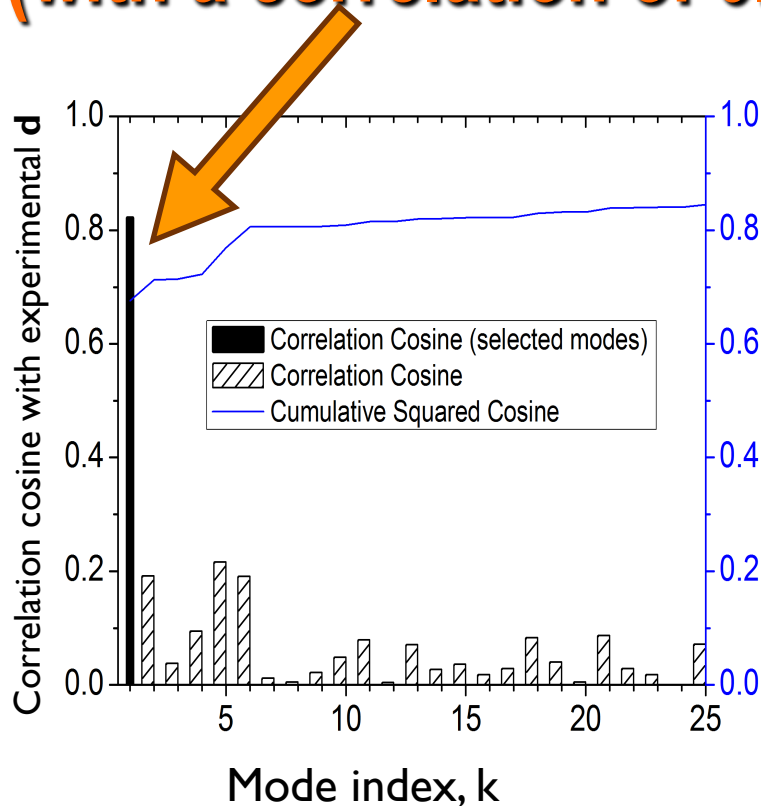
Correlation cosine between \mathbf{v}_k and \mathbf{d}



$$\mathbf{d} = [\Delta x_1 \quad \Delta y_1 \quad \Delta z_1 \quad \dots \quad \Delta z_N]^T$$

See...

The softest mode enables the passage $R \rightarrow T$ (with a correlation of 0.81)



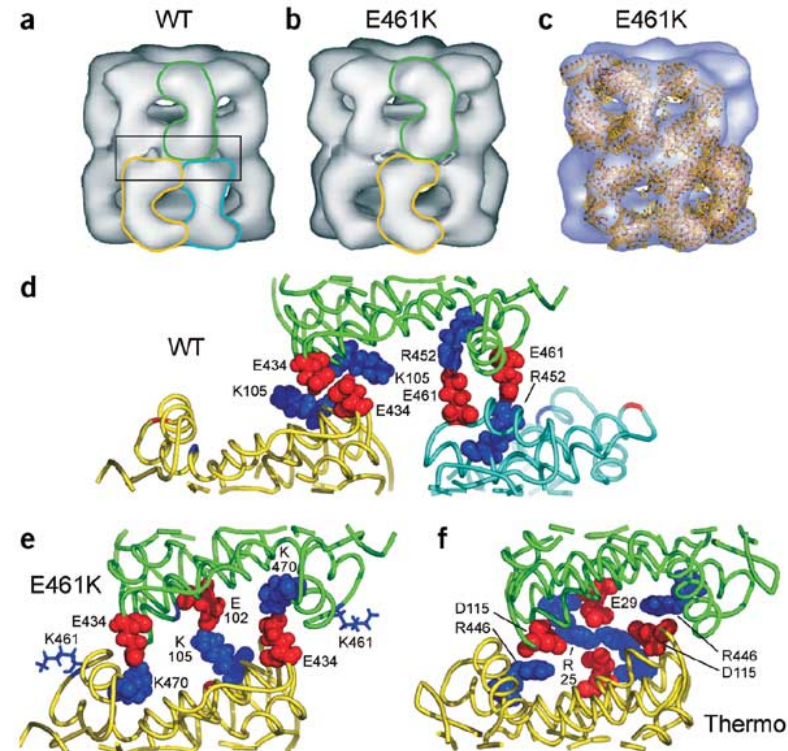
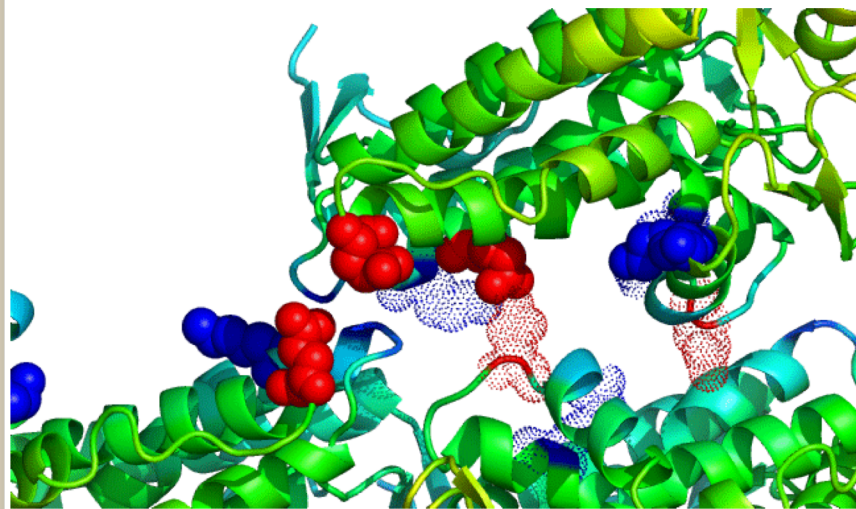
$$CO(m) = \left(\sum_{k=1}^m \left(\mathbf{v}_k \cdot \frac{\mathbf{d}}{|\mathbf{d}|} \right)^2 \right)^{1/2}$$

$$\mathbf{d} = [\Delta x_1 \quad \Delta y_1 \quad \Delta z_1 \quad \dots \quad \Delta z_N]^T$$

See...

Mutations may stabilize conformers along soft modes – which may be impair function

E461 mutant is a deformed structure along mode 1

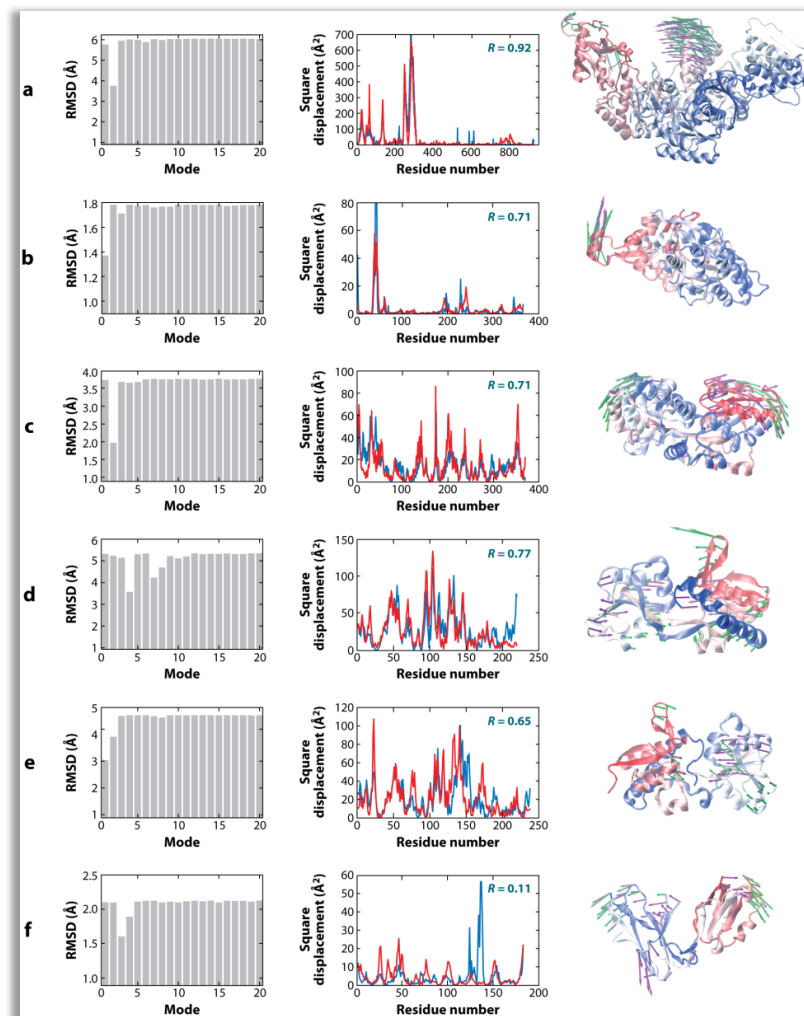


E461K mutation causes disruption of inter-ring transfer of ATP-induced signal (Sewell et al NSB 2004)

Experimentally observed structural changes are usually reconfigurations along soft modes

● Correlation cosine of 0.75 ± 0.15 between one of the softest modes and the experimentally observed change in structure

● Significant decrease in RMSD between the endpoints upon moving along a single soft mode (out of $3N-6$ modes)



Bahar I, et al. 2010.
Annu. Rev. Biophys. 39:23-42

See...

A better comparison:

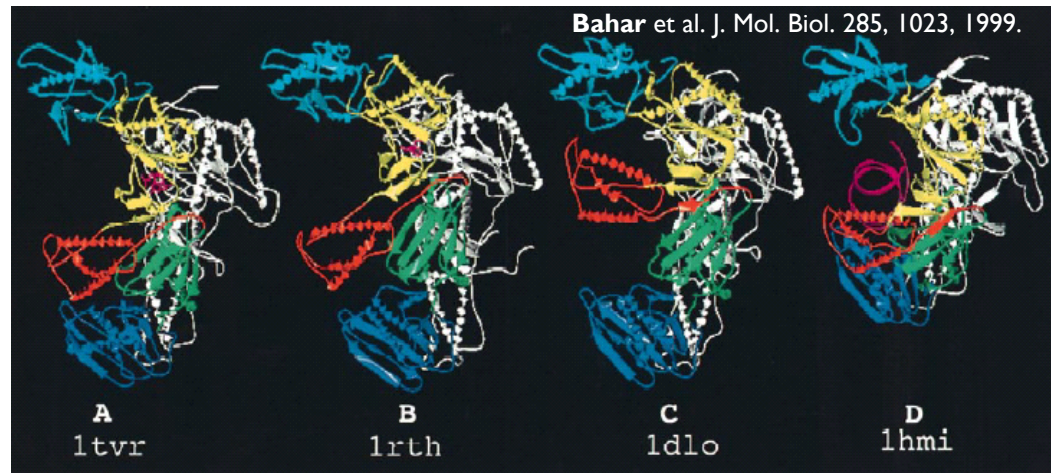
Consider more than 2 end points for a given structure, but all the known structures for a given protein, or the structurally resolved

Ensemble of structures

Bakan A, Bahar I (2009) [The intrinsic dynamics of enzymes plays a dominant role in determining the structural changes induced upon inhibitor binding.](#) *Proc Natl Acad Sci USA* **106**: 14349-14354.

Dynamics inferred from known structures

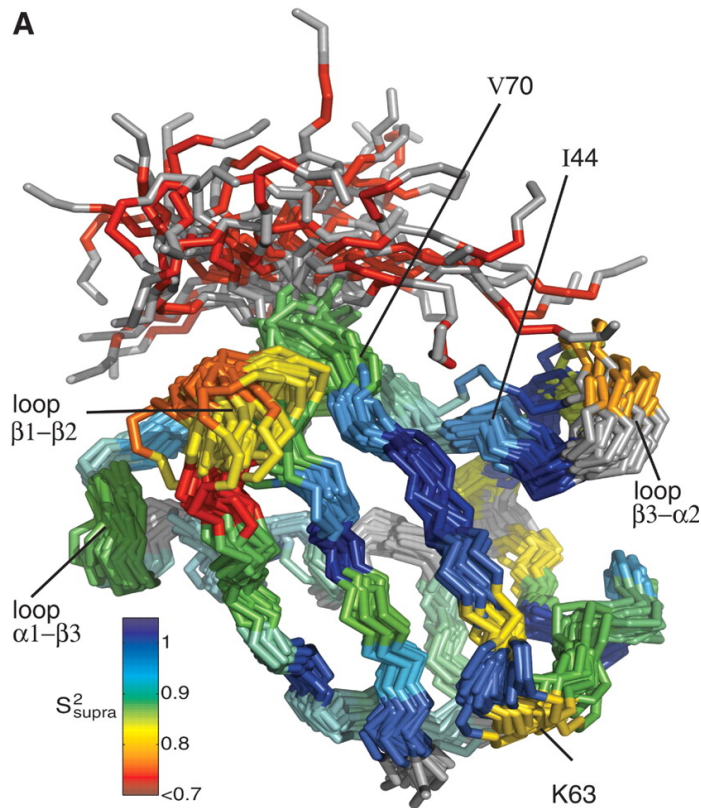
Comparison of static structures available in the PDB for the same protein in different form has been widely used as an indirect method of inferring dynamics.



Different structures resolved for HIV-1 reverse transcriptase (RT)

Recognition Dynamics Up to Microseconds Revealed from an RDC-Derived Ubiquitin Ensemble in Solution

Oliver F. Lange, ..., Jens Meiler, Helmut Grubmüller, Christian Griesinger, Bert L. de Groot



The ensemble covers the complete structural heterogeneity observed in 46 ubiquitin crystal structures, mostly complexes with other proteins.

- **Conformational selection, rather than induced-fit** explains the molecular recognition dynamics of ubiquitin.
- **A concerted mode** accounts for molecular recognition heterogeneity

Ensembles of structures

- Structural changes accompanying substrate (protein) binding
- Structural changes induced by, or stabilized upon, ligand binding



Ubiquitin
140 structures
1732 models

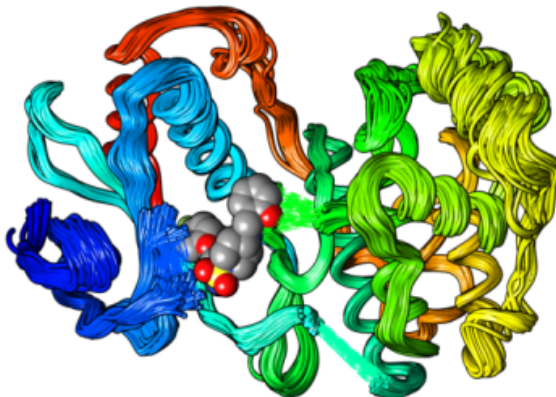
Ensembles of structures

- Structural changes accompanying substrate (protein) binding
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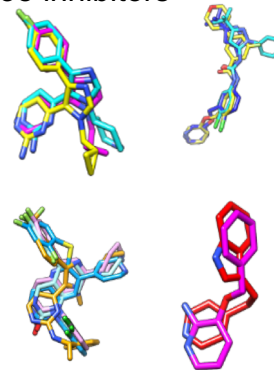


Ubiquitin
140 structures
1732 models

p38 MAP kinase
(182 structures)

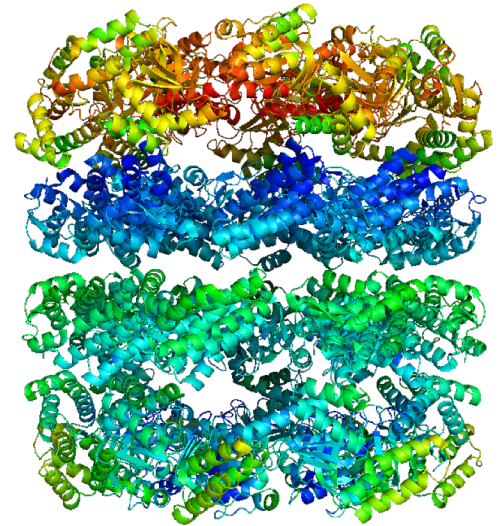


p38 inhibitors



Ensembles of structures

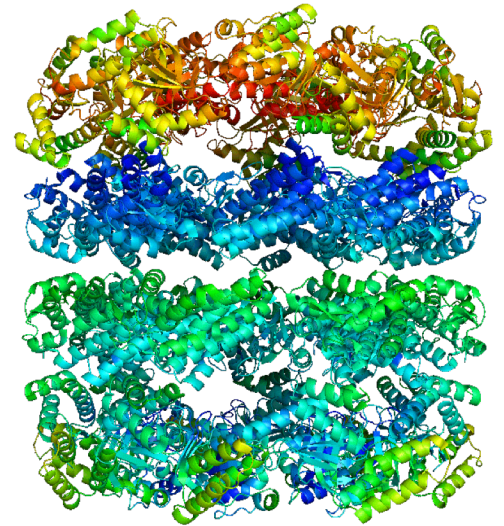
- Structural changes accompanying substrate (protein) binding
- Structural changes induced by, or stabilized upon, ligand binding
- Alternative conformations sampled during allosteric cycles



Yang et al. *PLoS Comp Biol* 2009

Ensembles of structures

- Structural changes accompanying substrate (protein) binding
- Structural changes induced by, or stabilized upon, ligand binding
- Alternative conformations sampled during allosteric cycles



Yang et al. *PLoS Comp Biol* 2009

What is Ensemble Analysis?

Principal component analysis

Input:

An ensemble of structures for a given protein

- NMR models (~40)
- X-ray structures resolved under different conditions (ligand-bound/unbound, different stages of molecular machinery or transport cycle)
- MD snapshots/frames

Output:

Principal modes of conformational changes

- variations/differences between NMR models
- rearrangements/changes under different functional states
- dynamics/fluctuations observed in simulations

What is Ensemble Analysis?

- ANM analysis
- Select a representative structure (e.g. with minimal RMSD from others)

Theoretical

- Decompose either **H** or **C** into a series of modes ($3N-6$ eigenvectors)

Principal component analysis

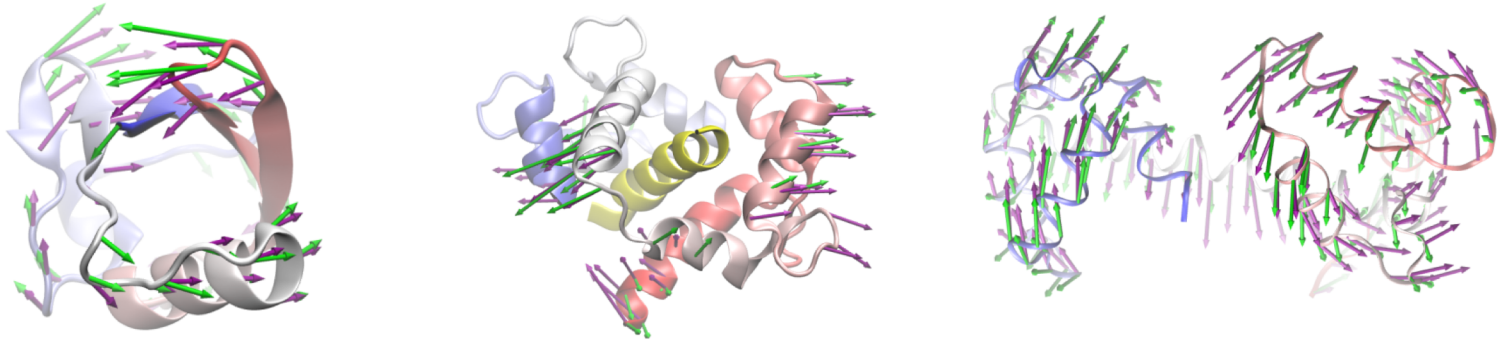
PCA

- Superimpose/align the structures

Experimental

- Decompose it into a series of modes of covariance ($3N-6$ eigenvectors)

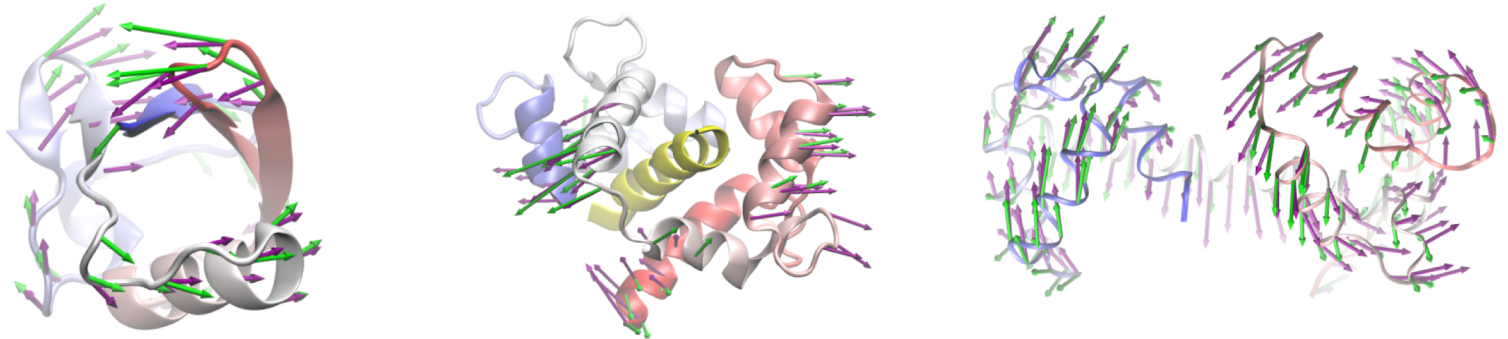
Global motions inferred from theory and experiments



→ PCA of the ensemble of resolved structures

→ ANM analysis of a single structure from the ensemble

Global motions inferred from theory and experiments



The intrinsic dynamics of enzymes plays a dominant role in determining the structural changes induced upon inhibitor binding

Ahmet Bakan and Ivet Bahar¹

Department of Computational Biology, School of Medicine, University of Pittsburgh, 3064 BST3, 3501 Fifth Avenue, Pittsburgh, PA 15213

NAS

Reference:

Bakan & Bahar (2009) PNAS 106, 14349-54

Covariance matrix (NxN)

C =

$\langle \Delta \mathbf{R}_1 \cdot \Delta \mathbf{R}_1 \rangle$	$\langle \Delta \mathbf{R}_1 \cdot \Delta \mathbf{R}_2 \rangle$	$\langle \Delta \mathbf{R}_1 \cdot \Delta \mathbf{R}_N \rangle$
$\langle \Delta \mathbf{R}_2 \cdot \Delta \mathbf{R}_1 \rangle$	$\langle \Delta \mathbf{R}_2 \cdot \Delta \mathbf{R}_2 \rangle$			
...				
...				
$\langle \Delta \mathbf{R}_N \cdot \Delta \mathbf{R}_1 \rangle$				$\langle \Delta \mathbf{R}_N \cdot \Delta \mathbf{R}_N \rangle$

$$= \Delta \mathbf{R} \Delta \mathbf{R}^T$$

$\Delta \mathbf{R}$ = N-dim vector of instantaneous fluctuations $\Delta \mathbf{R}_i$ for all residues ($1 \leq i \leq N$)


$\langle \Delta \mathbf{R}_1 \cdot \Delta \mathbf{R}_1 \rangle$ = ms fluctuation of site 1 averaged over all m snapshots.

Covariance matrix (3Nx3N)

$C_{3N} =$

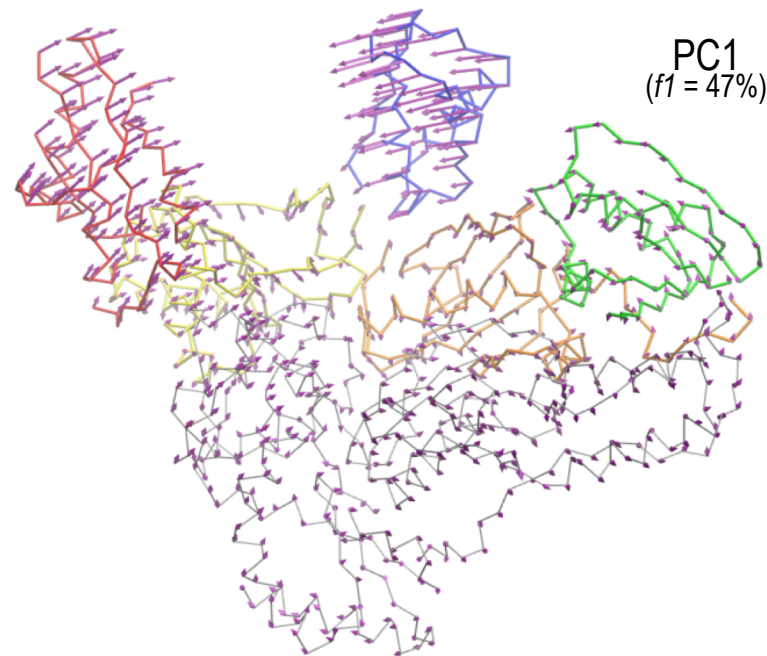
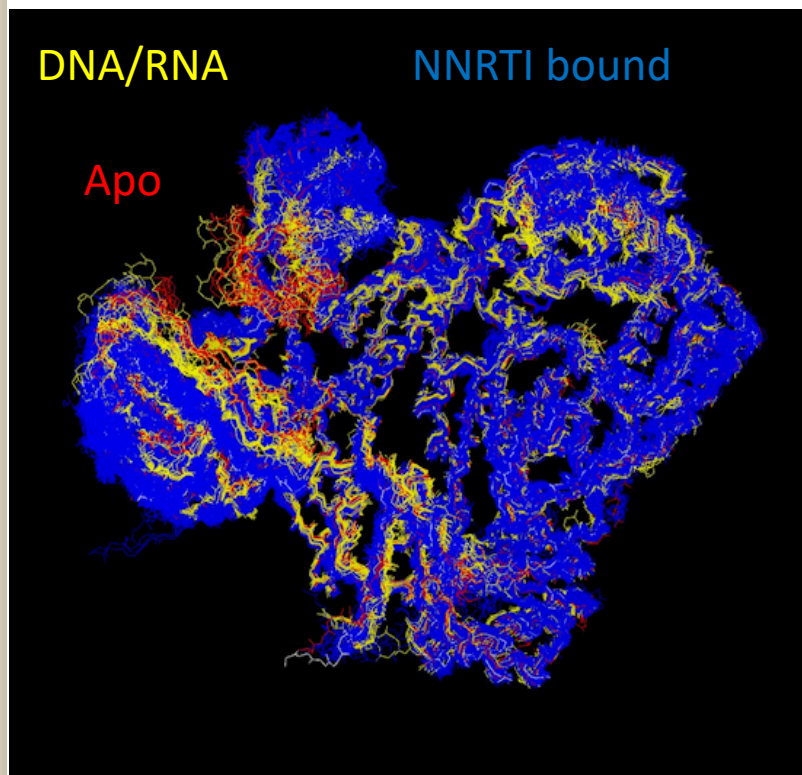
C_{11}	C_{21}	C_{13}		C_{1N}
C_{12}	C_{22}			
C_{N1}				C_{NN}

3N x 3N



$\langle \Delta X_1 \Delta X_2 \rangle$	$\langle \Delta X_1 \Delta Y_2 \rangle$	$\langle \Delta X_1 \Delta Z_2 \rangle$
$\langle \Delta Y_1 \Delta X_2 \rangle$	$\langle \Delta Y_1 \Delta Y_2 \rangle$	$\langle \Delta Y_1 \Delta Z_2 \rangle$
$\langle \Delta Z_1 \Delta X_2 \rangle$	$\langle \Delta Z_1 \Delta Y_2 \rangle$	$\langle \Delta Z_1 \Delta Z_2 \rangle$

Principal Component Analysis (PCA)

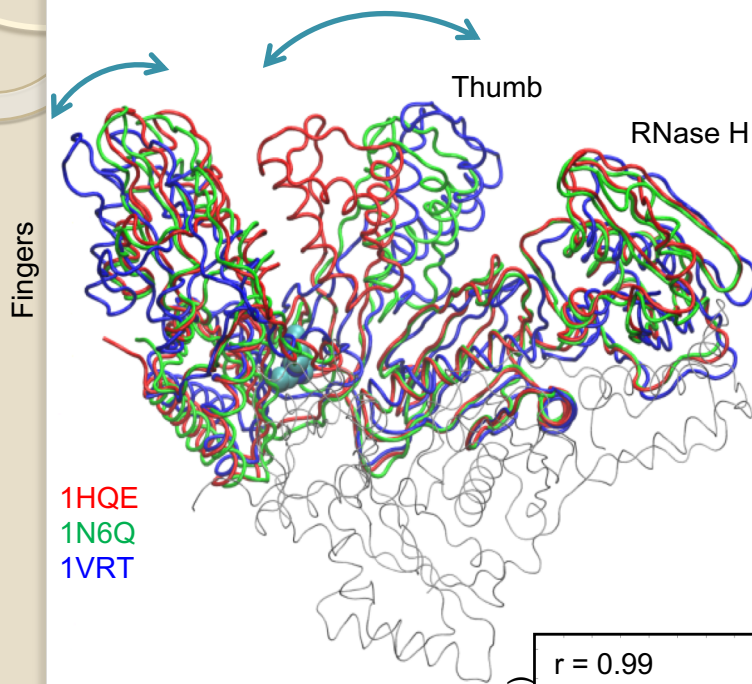


$$\mathbf{C}^{(ij)} = \begin{bmatrix} \langle \Delta x_i \Delta x_j \rangle & \langle \Delta x_i \Delta y_j \rangle & \langle \Delta x_i \Delta z_j \rangle \\ \langle \Delta y_i \Delta x_j \rangle & \langle \Delta y_i \Delta y_j \rangle & \langle \Delta y_i \Delta z_j \rangle \\ \langle \Delta z_i \Delta x_j \rangle & \langle \Delta z_i \Delta y_j \rangle & \langle \Delta z_i \Delta z_j \rangle \end{bmatrix}$$

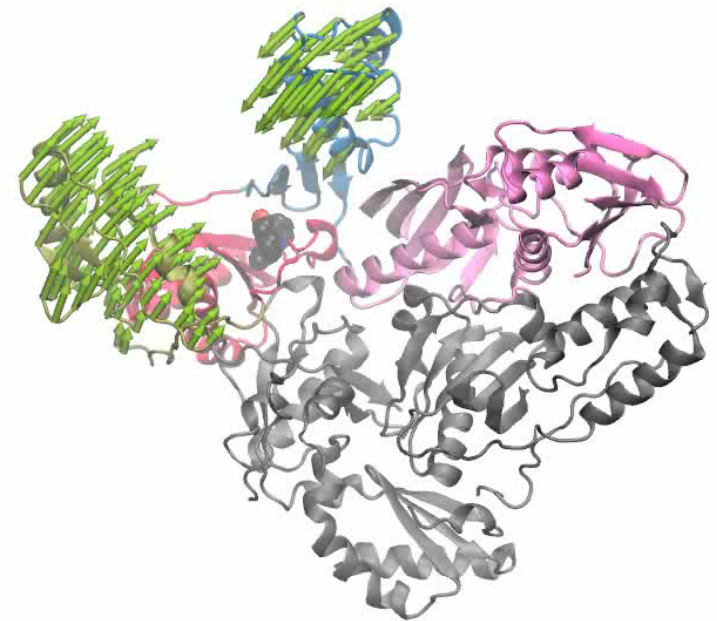
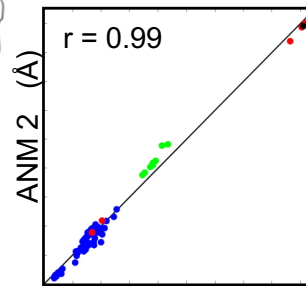


$$\mathbf{C} = \mathbf{PSP}^T = \sum_{i=1}^{3N} \sigma_i \mathbf{p}^i \mathbf{p}^{iT}$$

Soft modes enable **functional** movements



Experiments

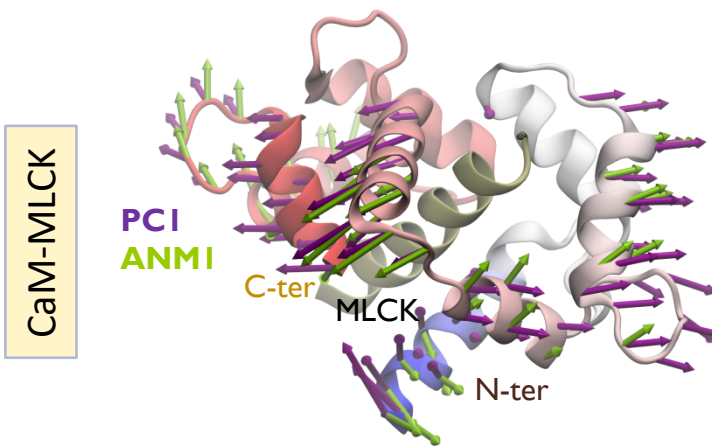


<http://www.youtube.com/watch?v=1OUzdm68YY>

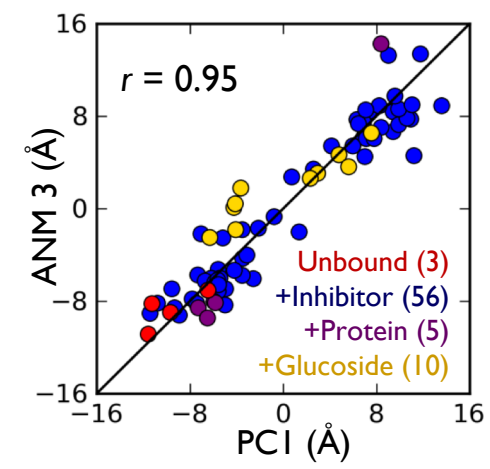
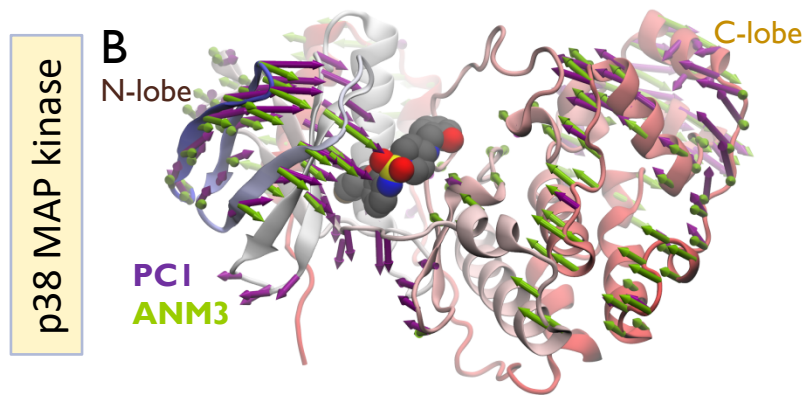
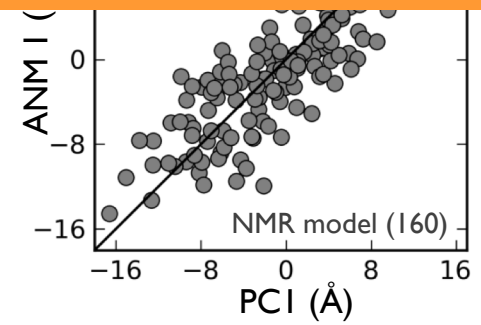
References:

Bakan & Bahar (2009) PNAS 106, 14349-54.

Experimental structures (for a given protein) are mainly variants along soft modes



Pre-existing paths



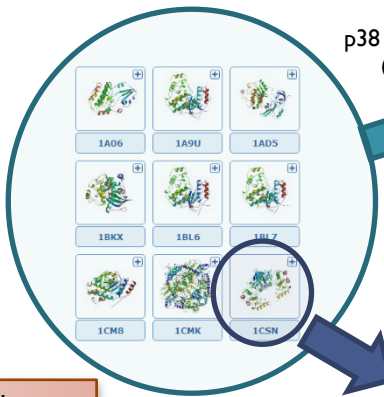
ProDy for exploring conformational space

Protein Dynamics Analysis in Python

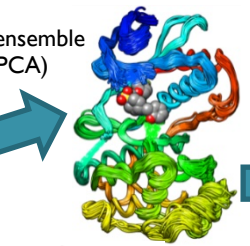
User inputs a protein sequence

```
> 1A9U:A|PDBID|CHAIN
GSSHHHHHHSSGLVPRGSHMSQER
PTFYRQELNKTIWEVPERYQNLSPV
GSGAYGSVCAAFDTKTGLRVAVKK
LSRPFQSIHAKRITYRELRLKHKMKH
ENVIGLLDVFT.....
```

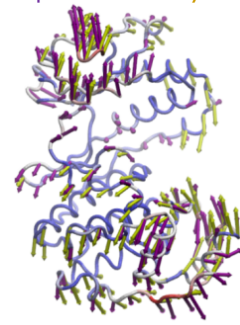
ProDy identifies, retrieves, aligns, and analyzes (PCA) structures that match the input sequence



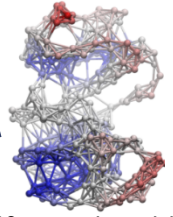
p38 ensemble (PCA)



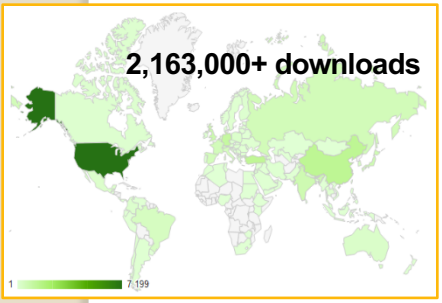
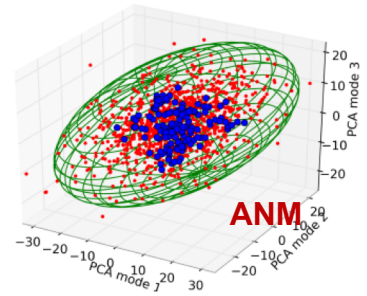
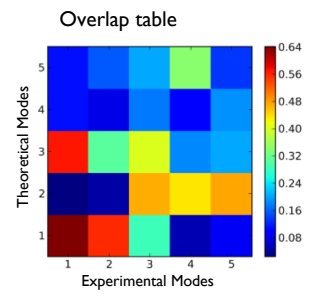
Experiment/Theory



p38 network model (ANM)

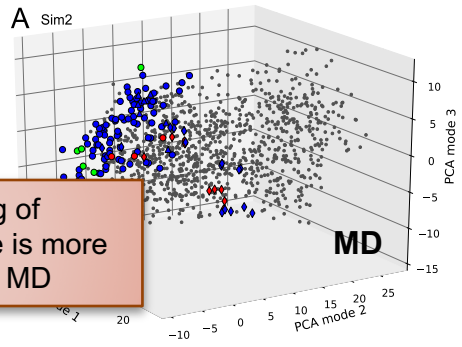


User can compare experimental and theoretical models



Source <http://www.google.com/analytics/>

ProDy-ANM sampling of conformational space is more complete than that of MD



User can sample an ensemble of conformations along ANM modes for docking simulations

ProDy: An Interactive Tool

Languages

Python	80%
C	13%
5 Other	7%

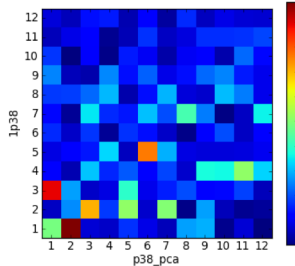
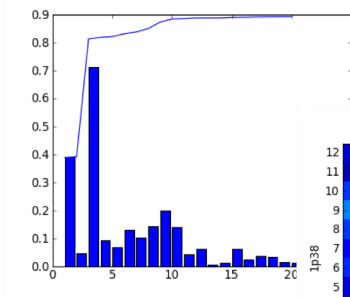
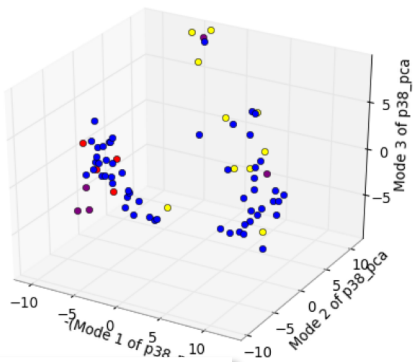


ProDy, updated May 20, 2014

more at Ohloh



IP[y]: IPython
Interactive Computing



```

abakan@orko: ~
578
ENMError: Coordinates are not set. Call select_residues method.

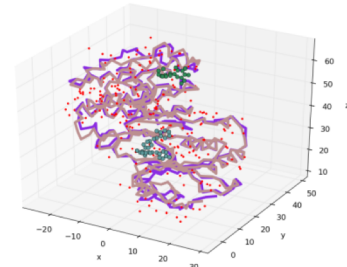
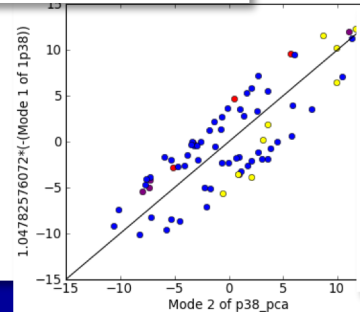
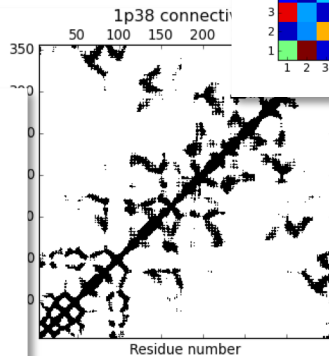
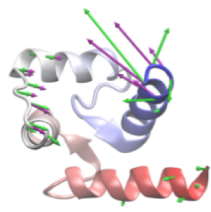
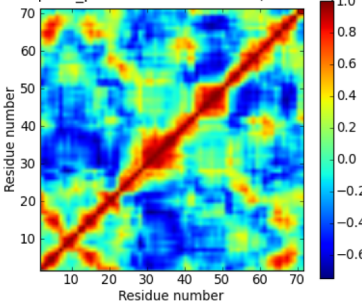
In [4]: ann.se
ann.secondary          ann.set_logger
ann.select_residues   ann.set_sec_str_assignments
ann.set_coordinates   ann.set_workDir
ann.set_hessian

In [4]: ann.select_residues('*')
Which chains and residues do you want to use from lmkp:
Chain A length 144 (Residue ids range from 204 to 347)
You have entered: *
Selection result:
144 residues from chain A

In [5]: ann.perform_analysis ()
@> Hessian matrix is being calculated.
@> Parameter: cutoff = 15
@> Parameter: gamma = 1
@> Hessian is calculated in 0.67s.
@> Normal mode calculation has started.
@> 20 modes will be calculated.
@> Normal modes are calculated in 0.12s.

In [6]:
    
```

ubiquitin_pca cross-correlations (6 modes)



Suite of tools



Elastic Network Model
(ANM/GNM) Analysis
Principal component analysis of
experimentally resolved structures



Multiple Sequence Alignment
Sequence conservation
Correlated Mutations

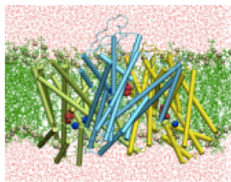


Computational Drug Discovery
Binding Site Prediction
Affinity Estimation



AVMD plugin
Visualization of collective motions
Animations/movies

Suite of tools

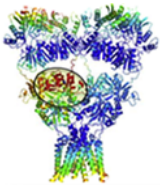


membrANM

Membrane Anisotropic Network Model

Modeling coupled protein-lipid dynamics

Useful for membrane proteins

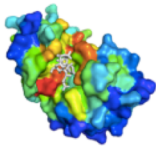


PRS

Perturbation-Response Scanning

Propagation allosteric signals

Effector and sensor residues



ESSA

Essential Site Scanning Analysis

Residues that can alter protein's essential dynamics upon binding

Prediction of allosteric pockets



SignDy

Signature Dynamics of Families

Shared global ENM mode profiles and departures from them, dynamics-based trees

Suite of tools



Unbiased conformational sampling of flexible and large biomolecular systems at atomic resolution and with high efficiency



ENM guided MD simulations
Efficient sampling of energy landscape and transitions



dynamics of protein complexes resolved by cryo-EM

Tutorials: ProDy & Structure Analysis

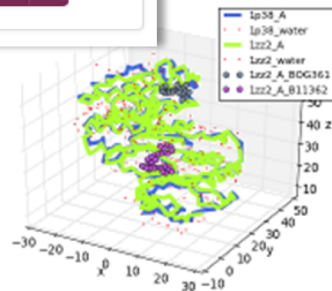


ProDy

Learn how to use ProDy from the introductory ProDy tutorial or from the comprehensive API reference manual.

Tutorial

Manual



Structure Analysis

Learn how to compare and align structures, identify ligand contacts, and extract ligands from PDB files.

Go to Tutorial

- Retrieving PDB Files
- BLAST-Searching the PDB
- Constructing Biomolecular Assemblies
- Determining functional motions
- Aligning and Comparing Structures
- Identifying Intermolecular Contacts

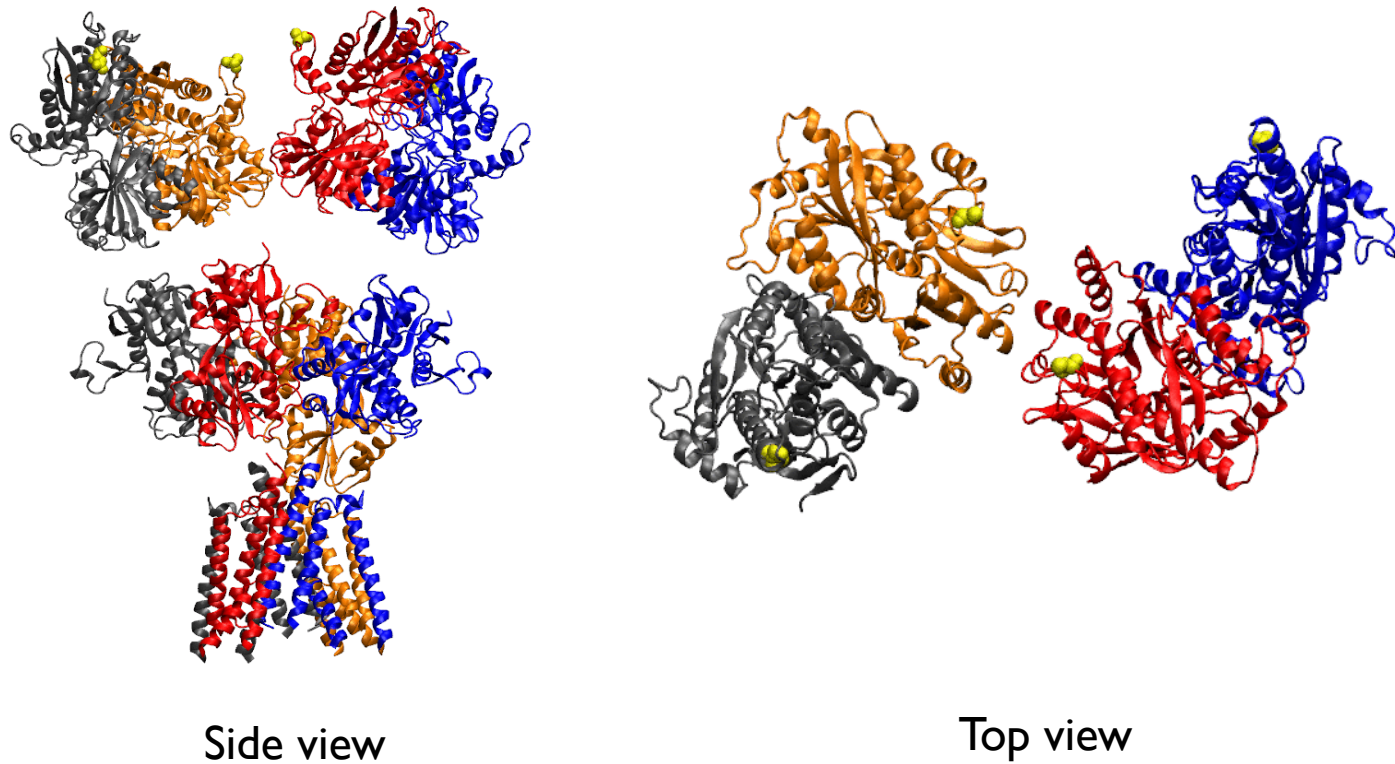
Major advantages of ProDy:

- Simplicity
- Visualizing the global dynamics
- Applicability to large systems
- Assessing cooperative motions
- Efficiency – immediate results
- Relevance to observables, to **functional mechanisms & allostery**

Caveats

- Low resolution approach
- No specific interactions
- Lack of atomic details
- Linear theory – applicable near an energy minimum
- not a tool for structure prediction (could be used for refinement)

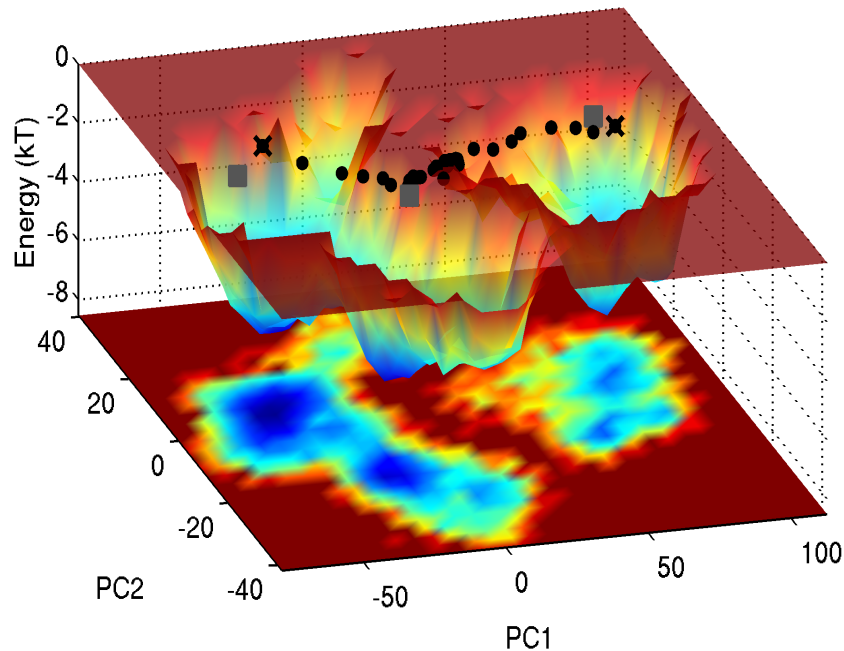
Allosteric transition of AMPAR



The trajectory was generated with adaptive-ANM (aANM) using the first 30 modes
Initial: N-shaped (PDB id: 4uqj) → Target: O-shaped (PDB id: 5ide) AMPAR

Hybrid methods to overcome caveats

ANM-guided atomistic simulations

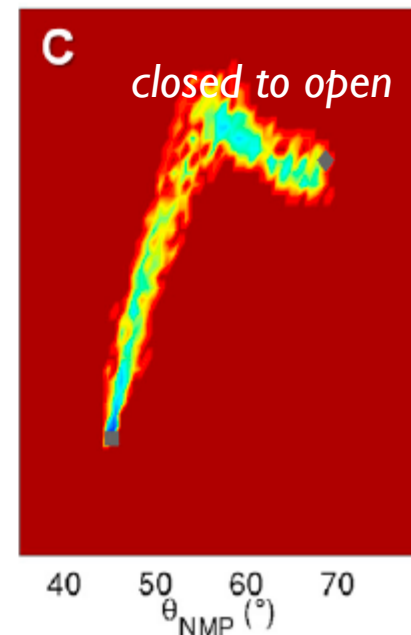
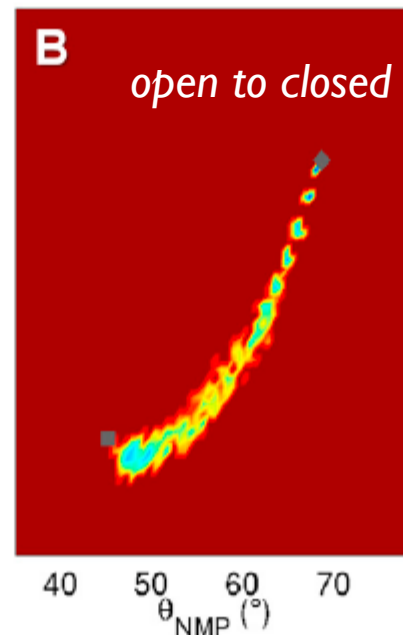
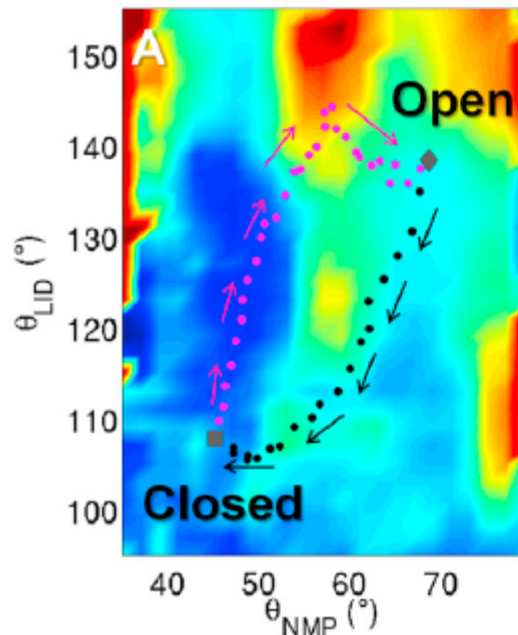
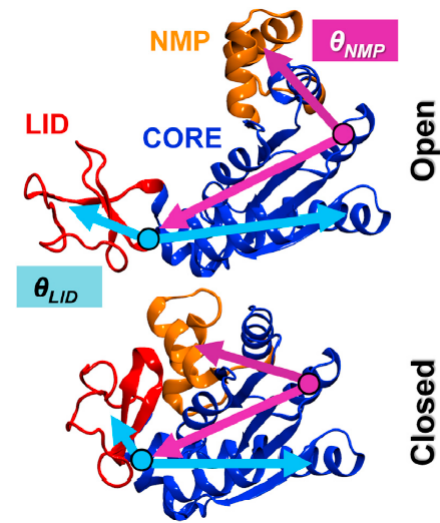


ANM-guided transition pathways

- Isin B, Schulten K, Tajkhorshid E, Bahar I (2008) *Biophysical J* 95: 789-803.
- Yang Z, Májek P, Bahar I (2009) *PLoS Comput Biol* 5: e1000360.
- Gur M, Madura JD, Bahar I (2013) *Biophys J* 105:1643-52
- Das A, Gur M, Cheng MH, Jo S, Bahar I, Roux B (2014) *PLoS Comput Biol* 10: e1003521

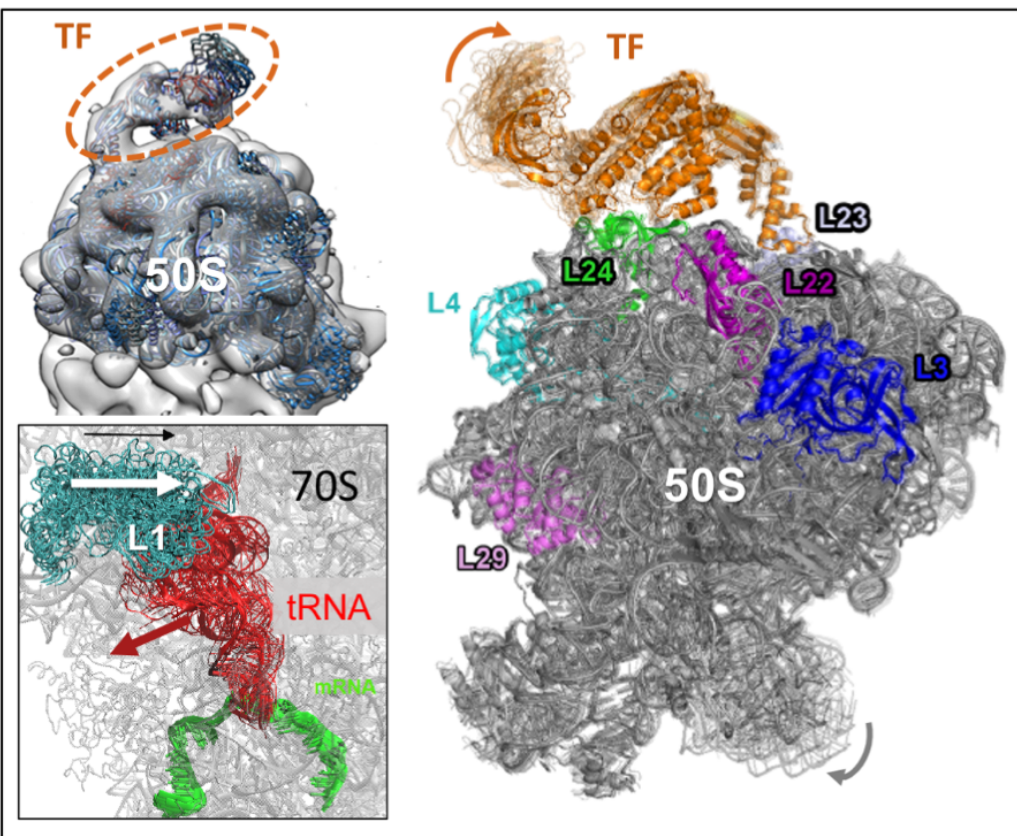
coMD trajectories proceed along the minima of free energy landscape

coMD transition pathways for adenylate kinase



Energy landscape from Beckstein et al, JMB, 2009

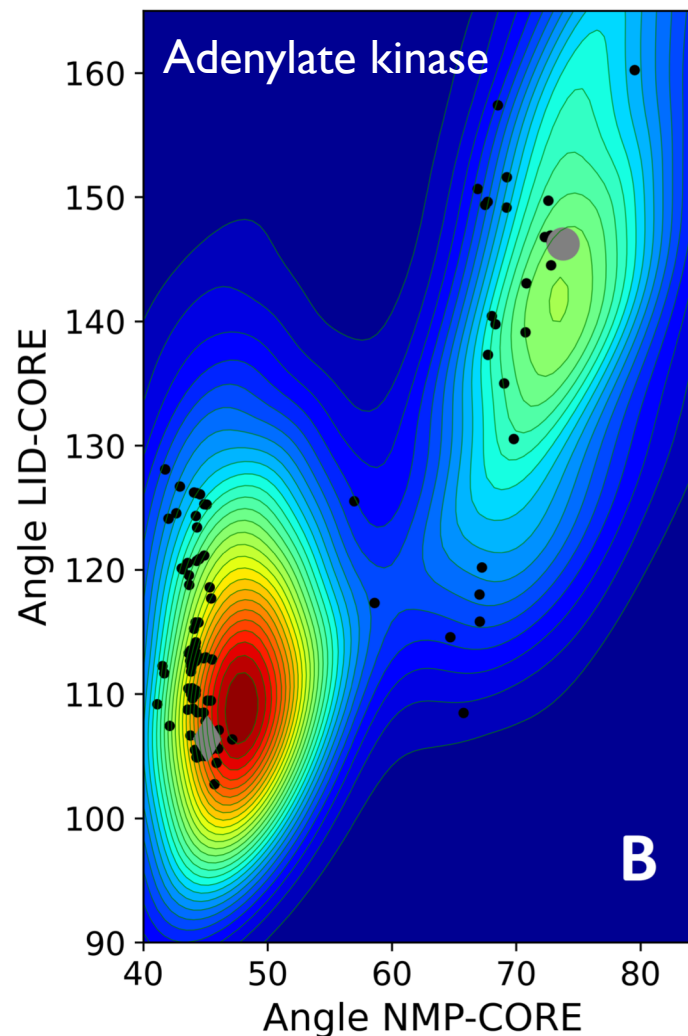
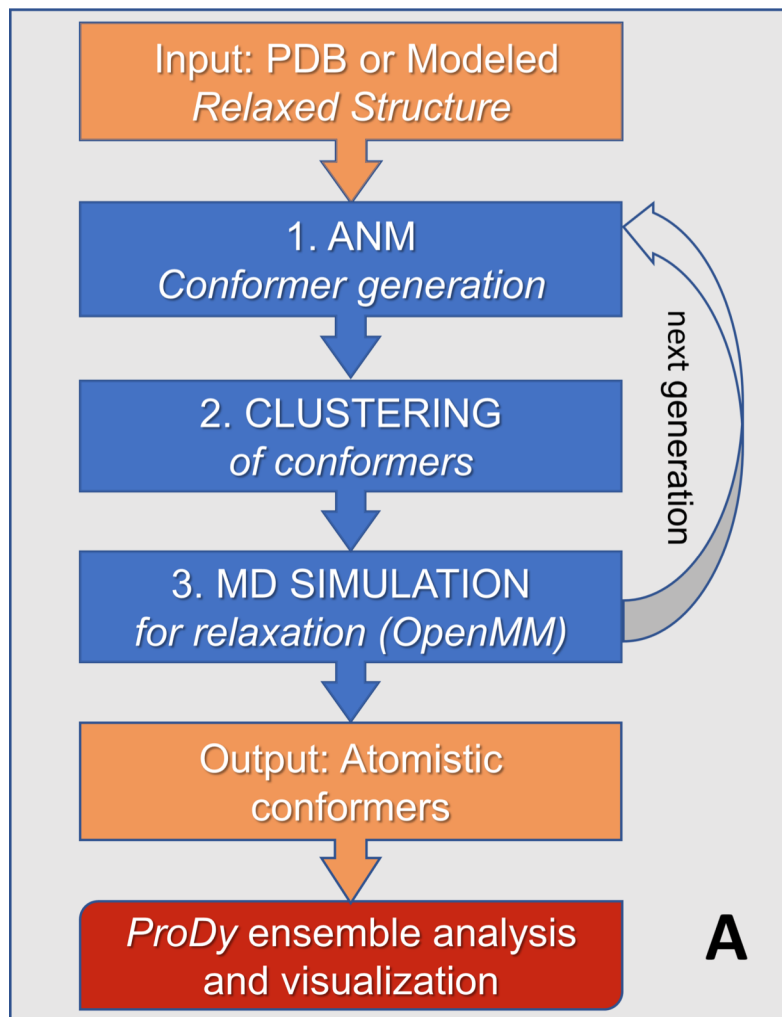
ClustENM for conformational sampling of flexible proteins and supramolecules



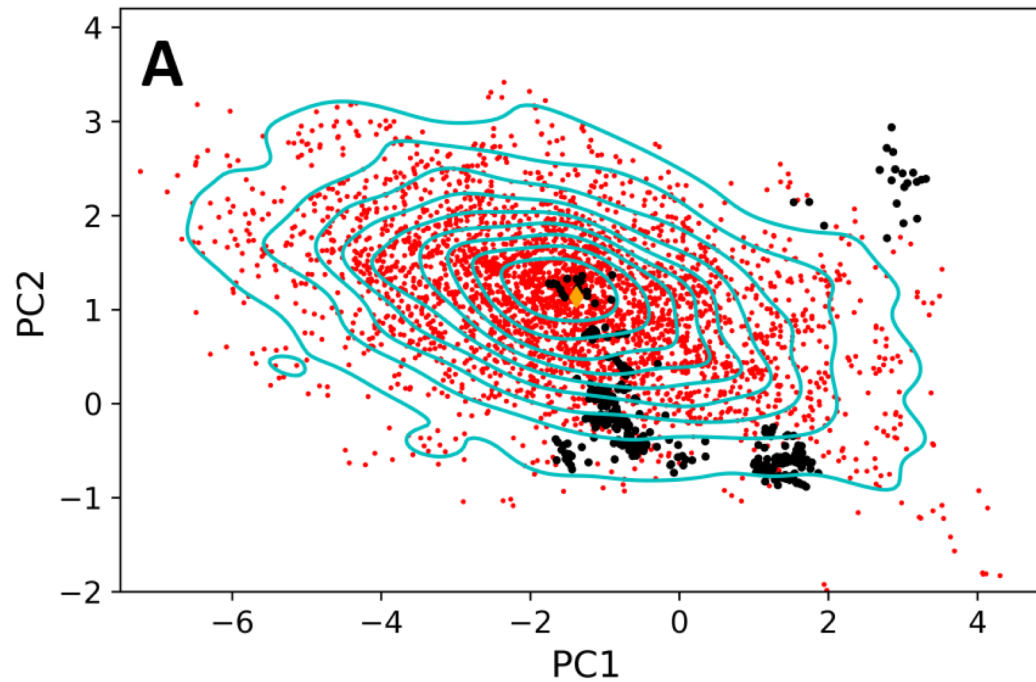
- **Original method:** Kurkcuoglu, Z.; Bahar, I.; Doruker, P., ClustENM: ENM-Based Sampling of Essential Conformational Space at Full Atomic Resolution. *J Chem Theory Comput* (2016) 12: 4549-62
- **Ensemble docking to flexible proteins:** Kurkcuoglu, Z.; Doruker, P., *PLoS One* (2016) 11, e0158063.
- **Trigger factor- ribosome dynamics:** Can, M.T.; Kurkcuoglu, Z.; Ezeroglu, G.; Uyar, A.; Kurkcuoglu, O.; Doruker, P., *PLoS One* (2017) 12, e0176262.
- **Protein-protein/DNA docking with HADDOCK:** Kurkcuoglu, Z.; Bonvin, A., *Proteins* (2020) 88: 292-306

Trigger factor (TF)- ribosome (50S) ClustENM conformers aligned to the cryo-EM map.
Conformers of 70S showing translocation of tRNA.

ClustENMD implementation in ProDy

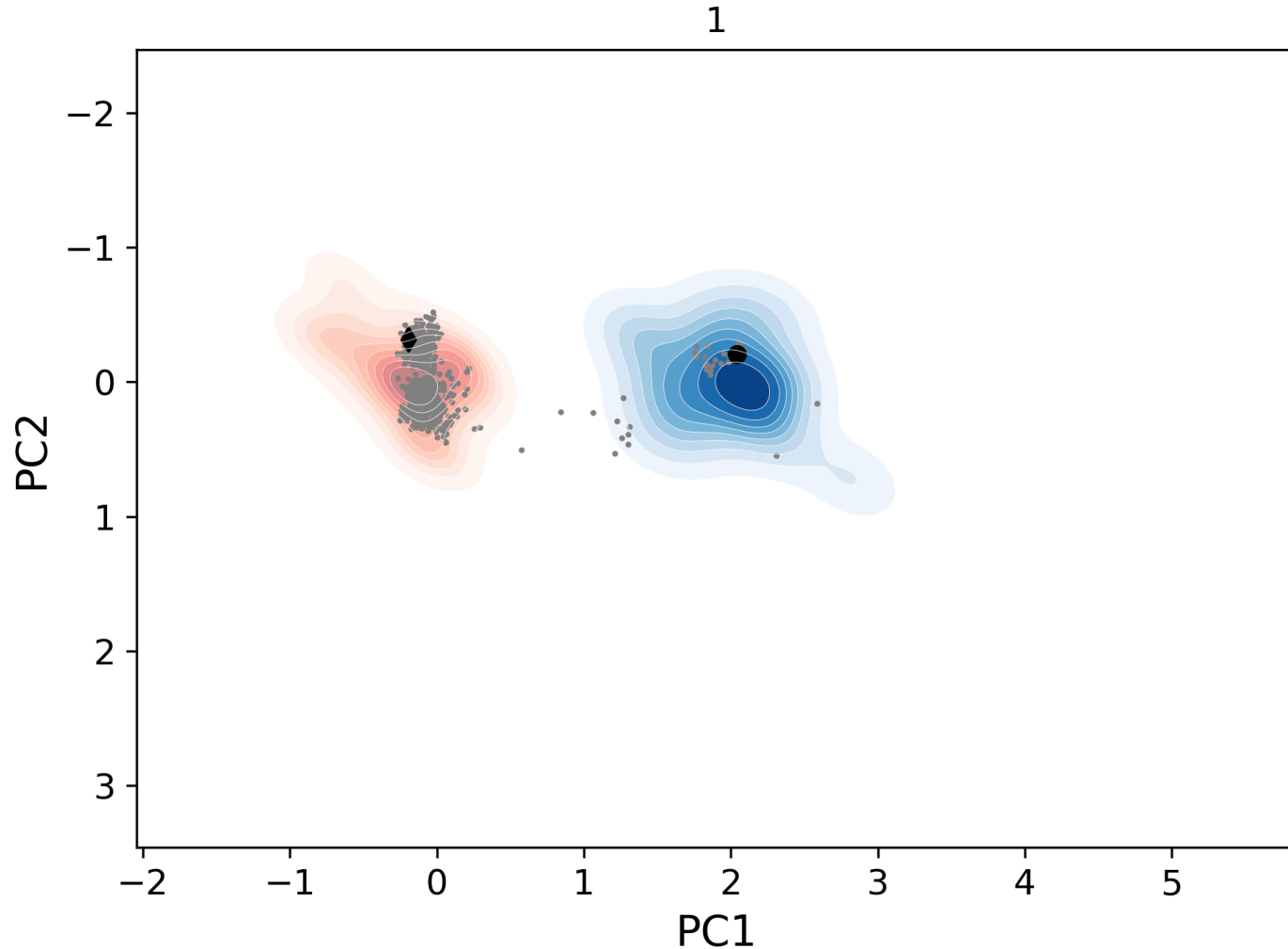


HIV-1 reverse transcriptase



Red dots: ClustENMD conformers
Black dots: experimental structures

ClustENMD for HIV-1 protease





Thank you!

