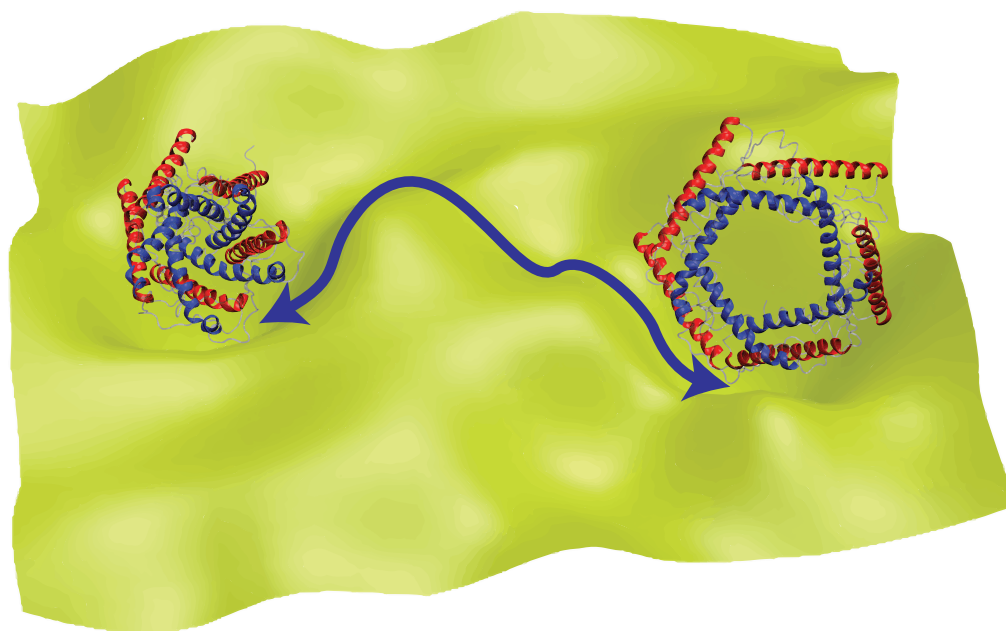


Exploring Complex Reaction Pathways

Mahmoud Moradi

Department of Chemistry and Biochemistry

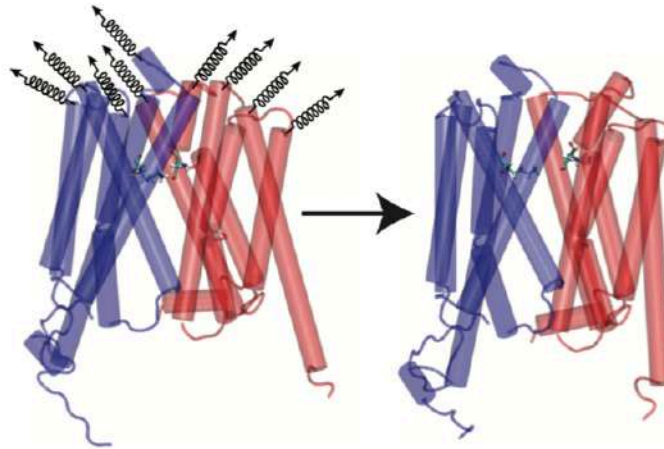
University of Arkansas



"Hands-on" Workshop on Enhanced Sampling and
Free-Energy Calculation at Urbana, IL

September 14, 2018

Exploring Complex Conformational Transition Pathways



Tutorial by Mahmoud Moradi

September 18 2017

Outline

- **Introduction**
 - How to study large-scale conformational changes?
- **Methodology**
 - Empirical search for good pulling protocols
 - Iterative combination of free energy calculation methods and path-finding algorithms

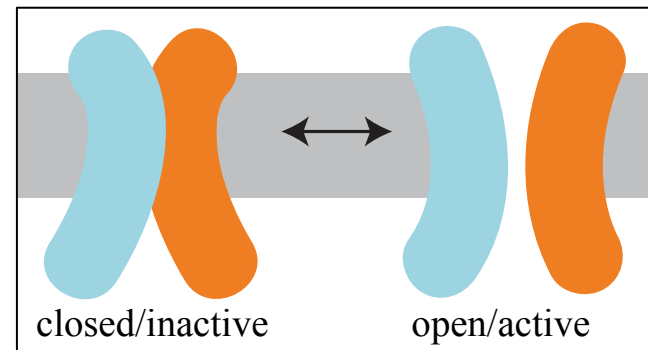
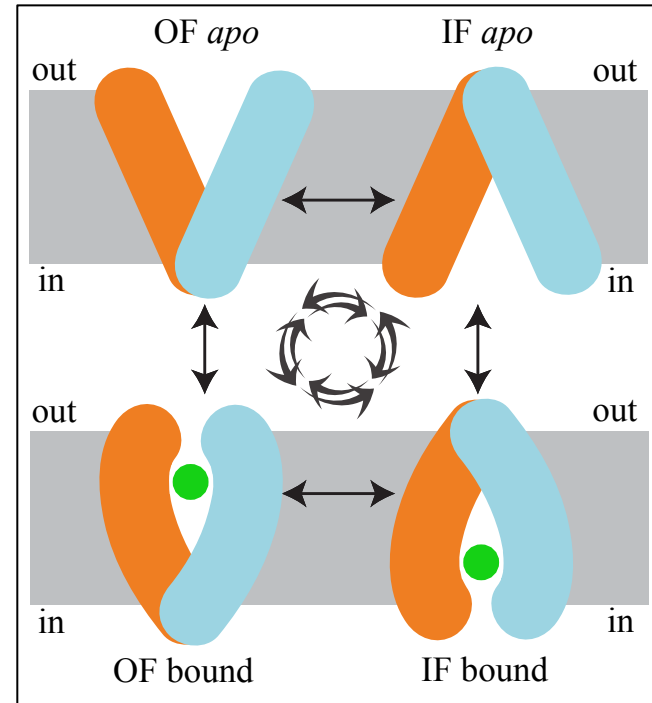
Outline

- **Introduction**

- How to study large-scale conformational changes?

Large-Scale Conformational Changes in Membrane Transport Proteins

- **Membrane transporters** rely on large-scale conformational changes between **inward-facing (IF)** and **outward-facing (OF)** states (**alternating access mechanism**)
- **Channels** may require large-scale conformational changes between their **open/active** and **closed/inactive** states.



A Case Study: Proton-coupled Oligopeptide Transporters (POTs)

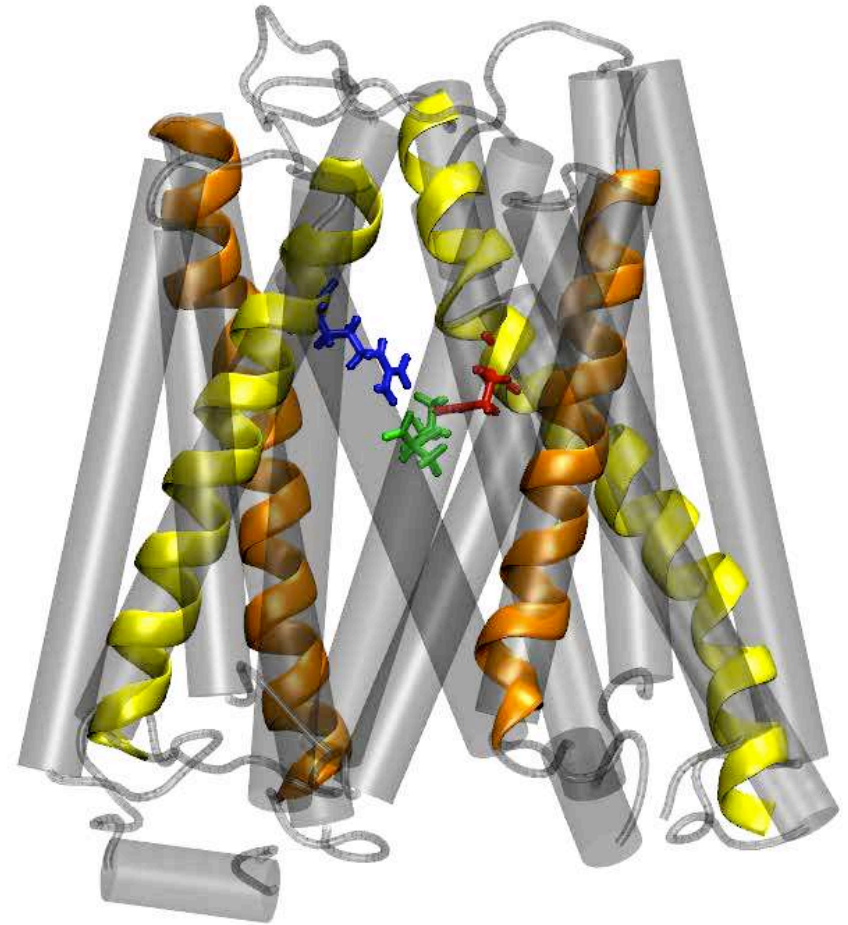
GkPOT (4IKV, 1.9 Å)

~100,000 atoms

Conventional unbiased
simulations performed:

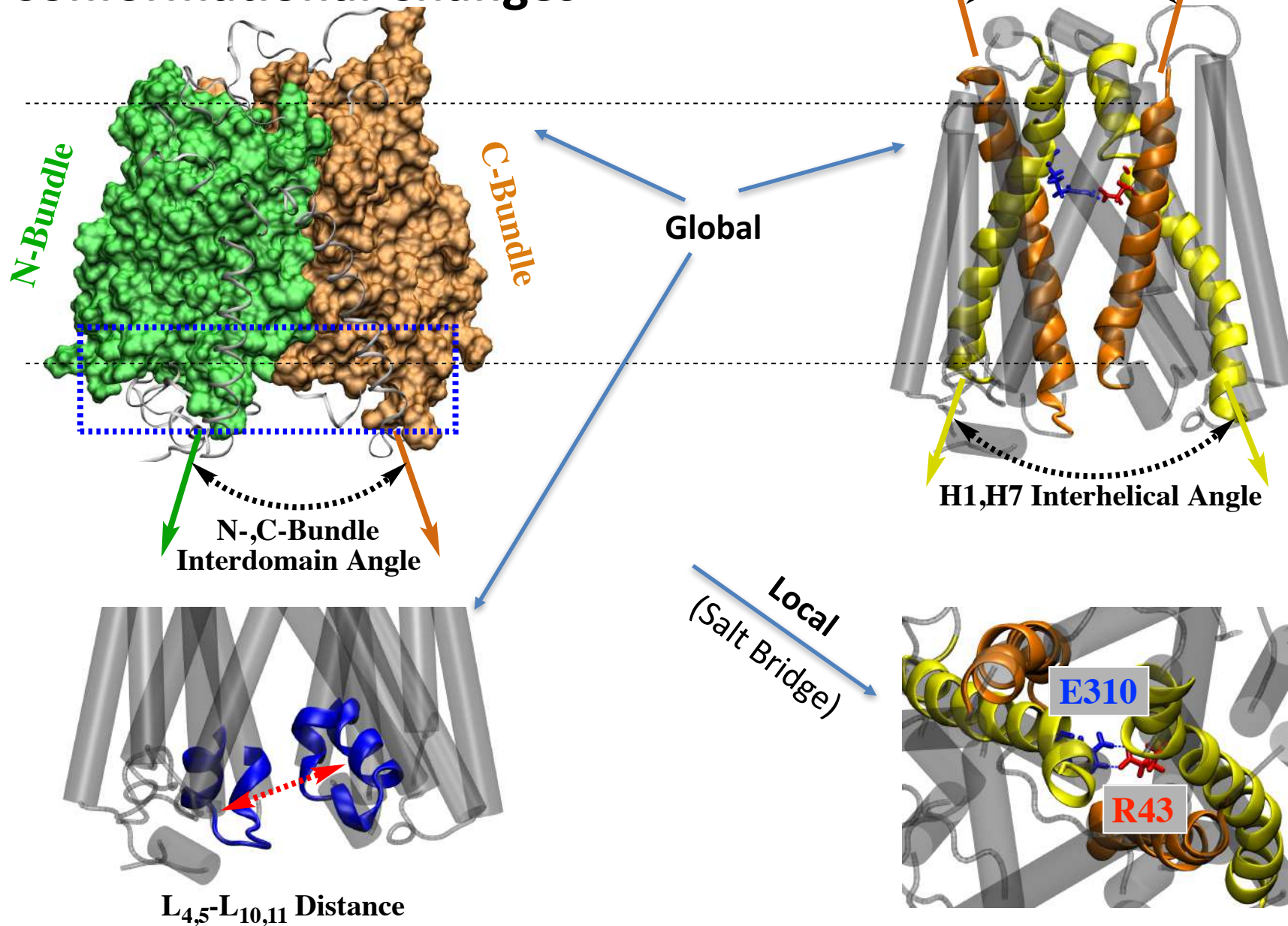
8 conditions × 400 ns

× 2 repeats = 6.4 μs

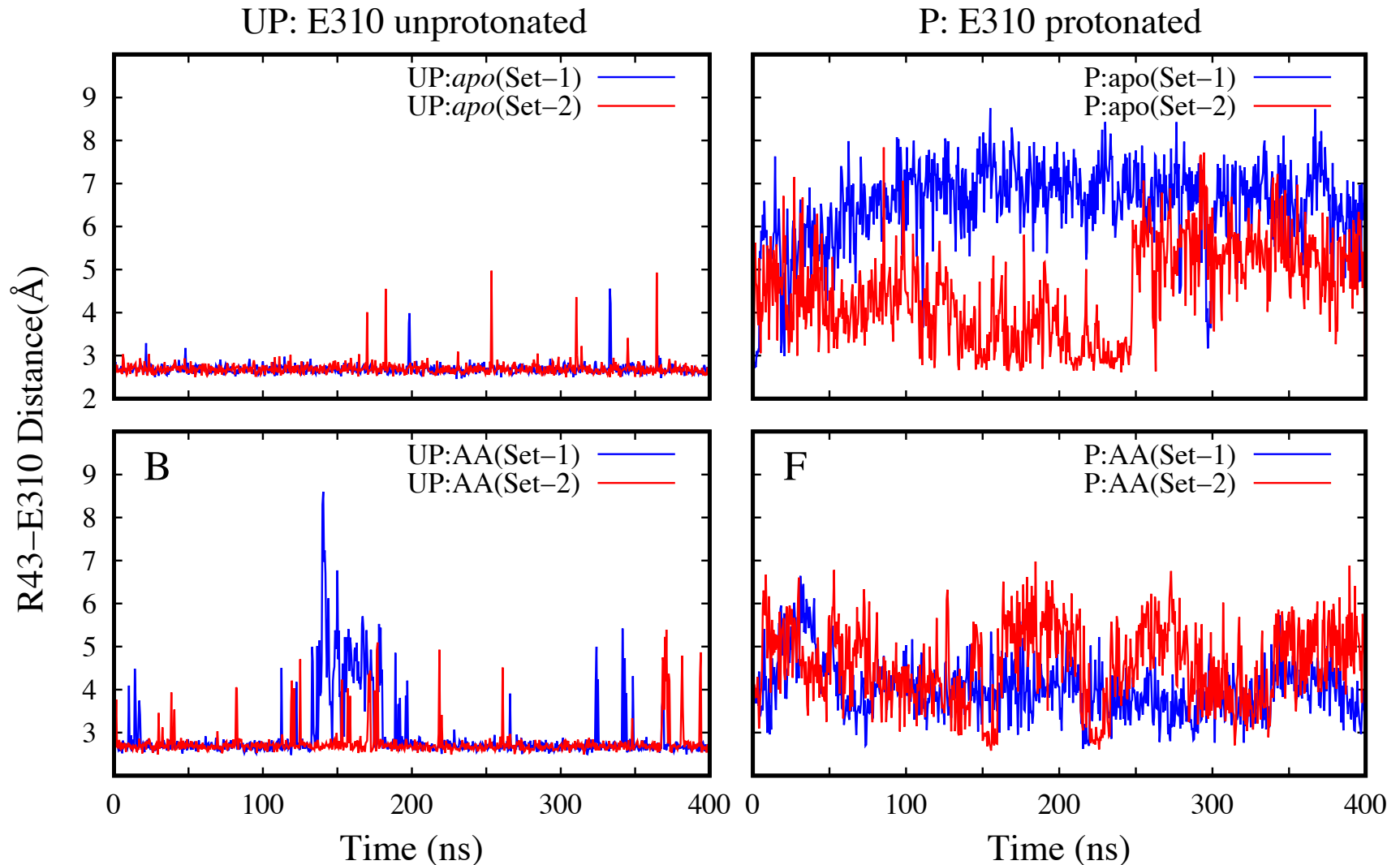


K Immadisetty, J Hettige, and M Moradi, *What Can and Cannot Be Learned from Molecular Dynamics Simulations of Bacterial Proton-Coupled Oligopeptide Transporter GkPOT?* *J. Phys. Chem. B*, **121**:3644-3656, 2017.

Monitoring Global and Local Conformational Changes

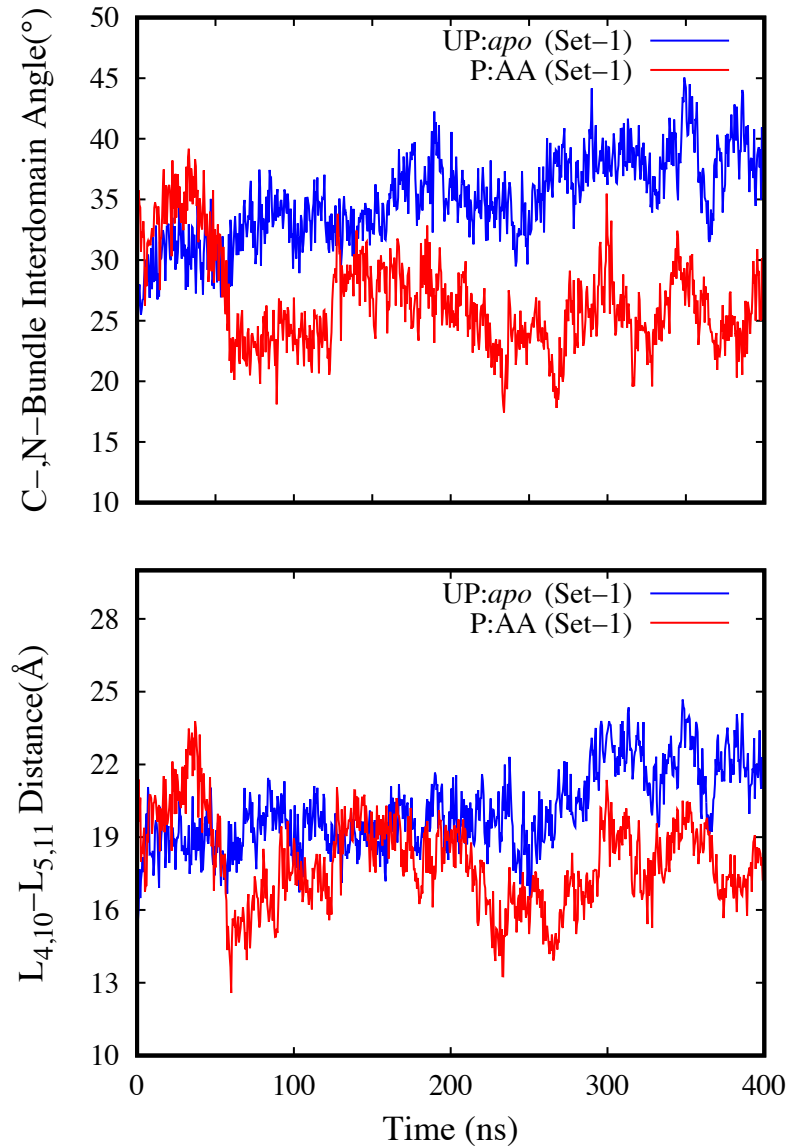


Local Conformational Changes



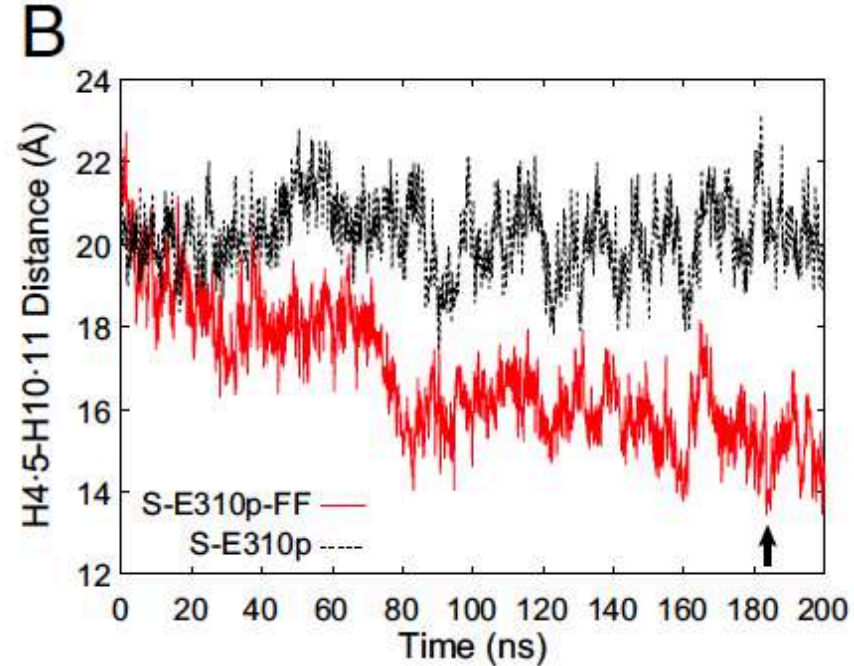
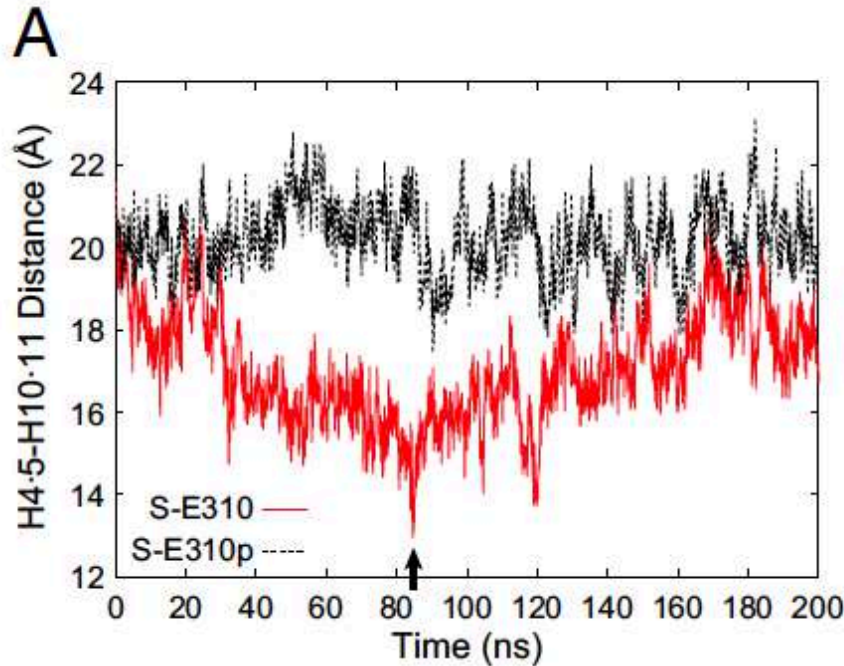
There is a clear distinction between different conditions.

Global Conformational Changes



There is no clear distinction between different conditions.

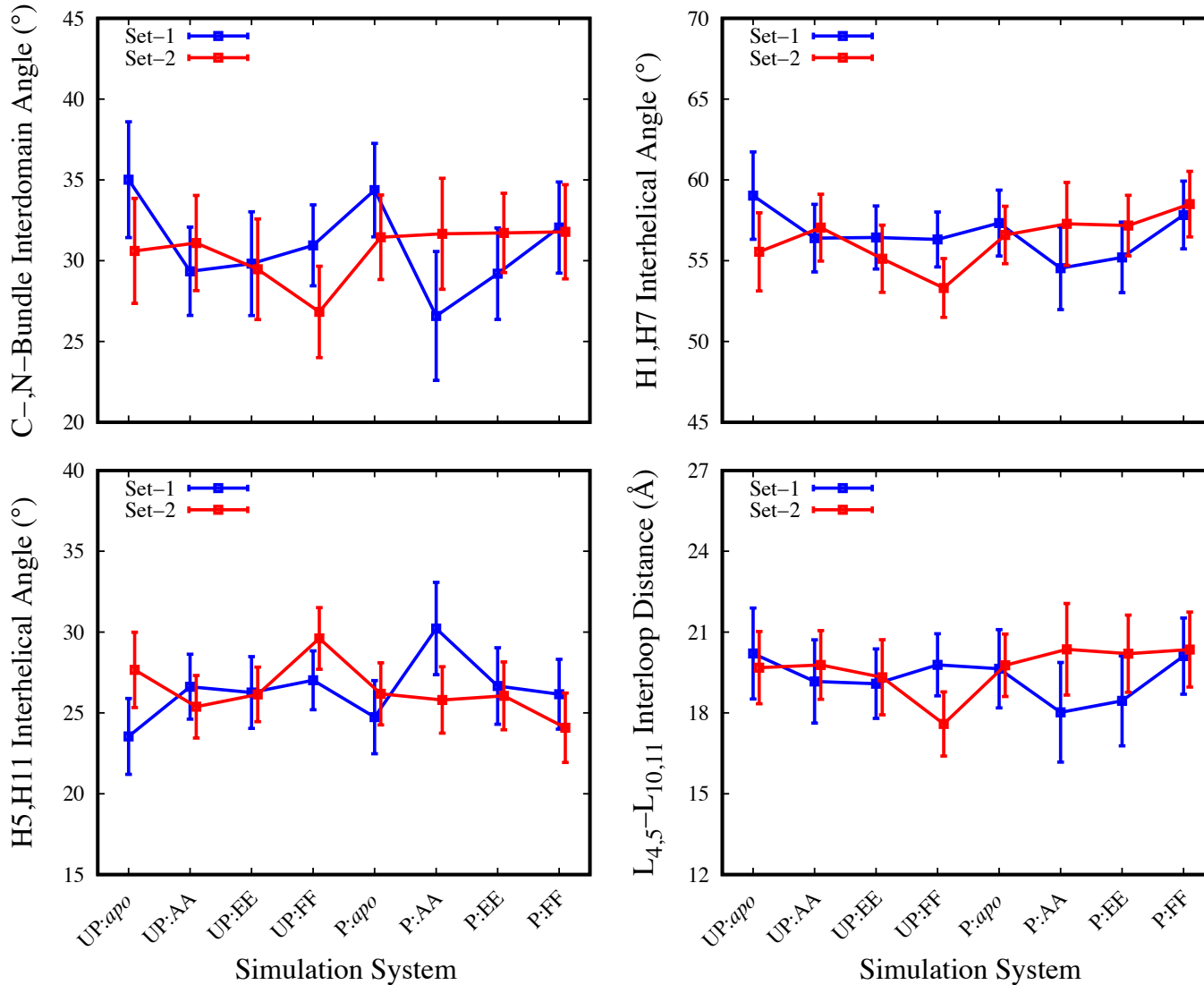
Global Conformational Changes



“Structural basis for dynamic mechanism of proton-coupled symport by the peptide transporter POT.” **PNAS 2013** | vol. 110 | no. 28 | 11343–11348.

Although a common practice, statements made about millisecond-level biomolecular events based on sub-microsecond level simulations are not reliable.

Global Conformational Changes

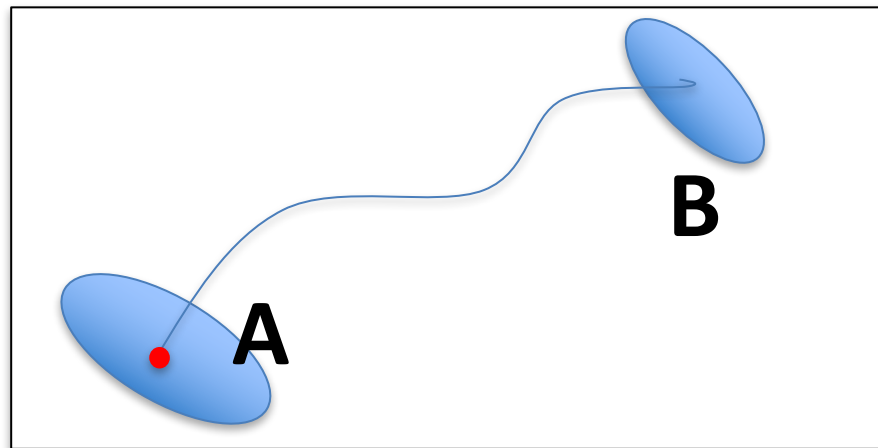


There is no statistically significant distinction between different conditions.

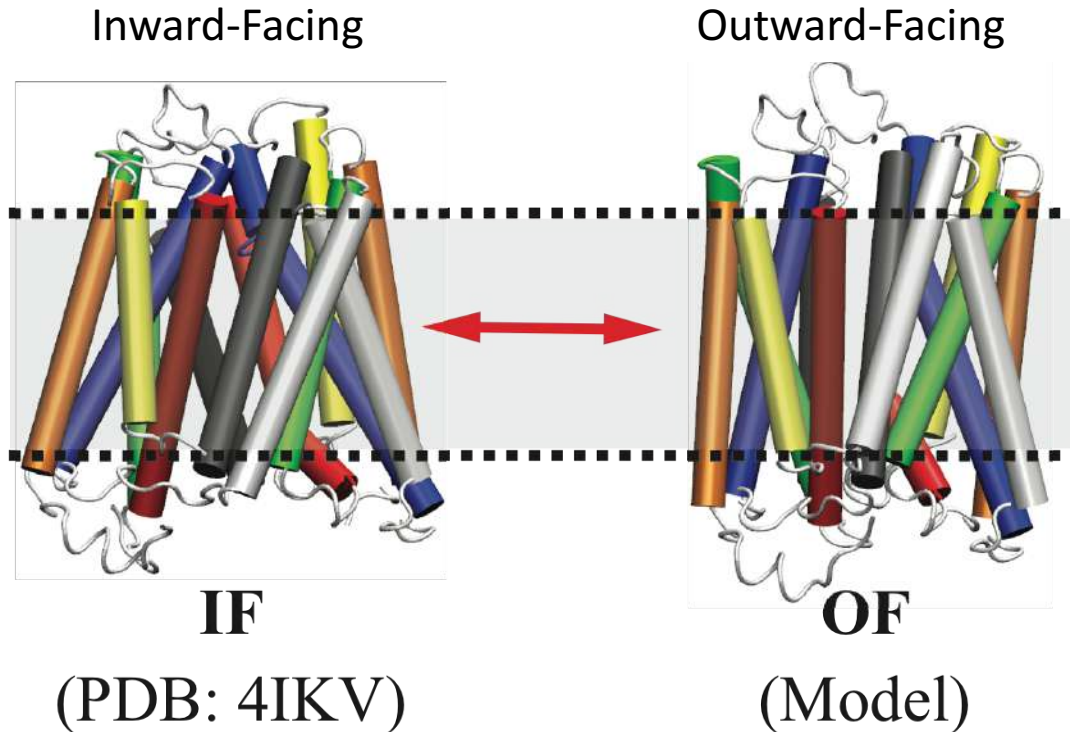
- **Introduction**

- How to study large-scale conformational changes?

It is not reasonable to speculate about the conformational transition between two states based on fluctuations around one of the end points.

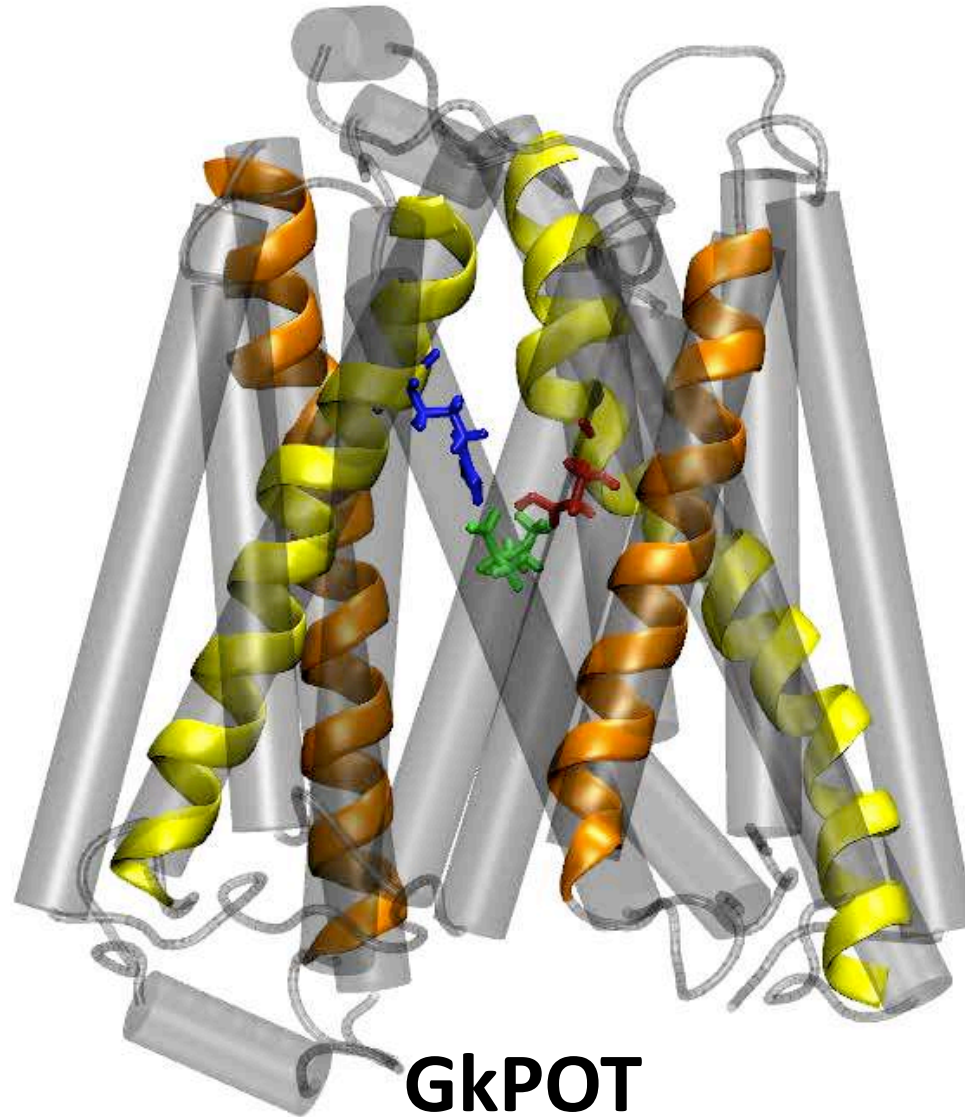


How to study large-scale conformational changes?



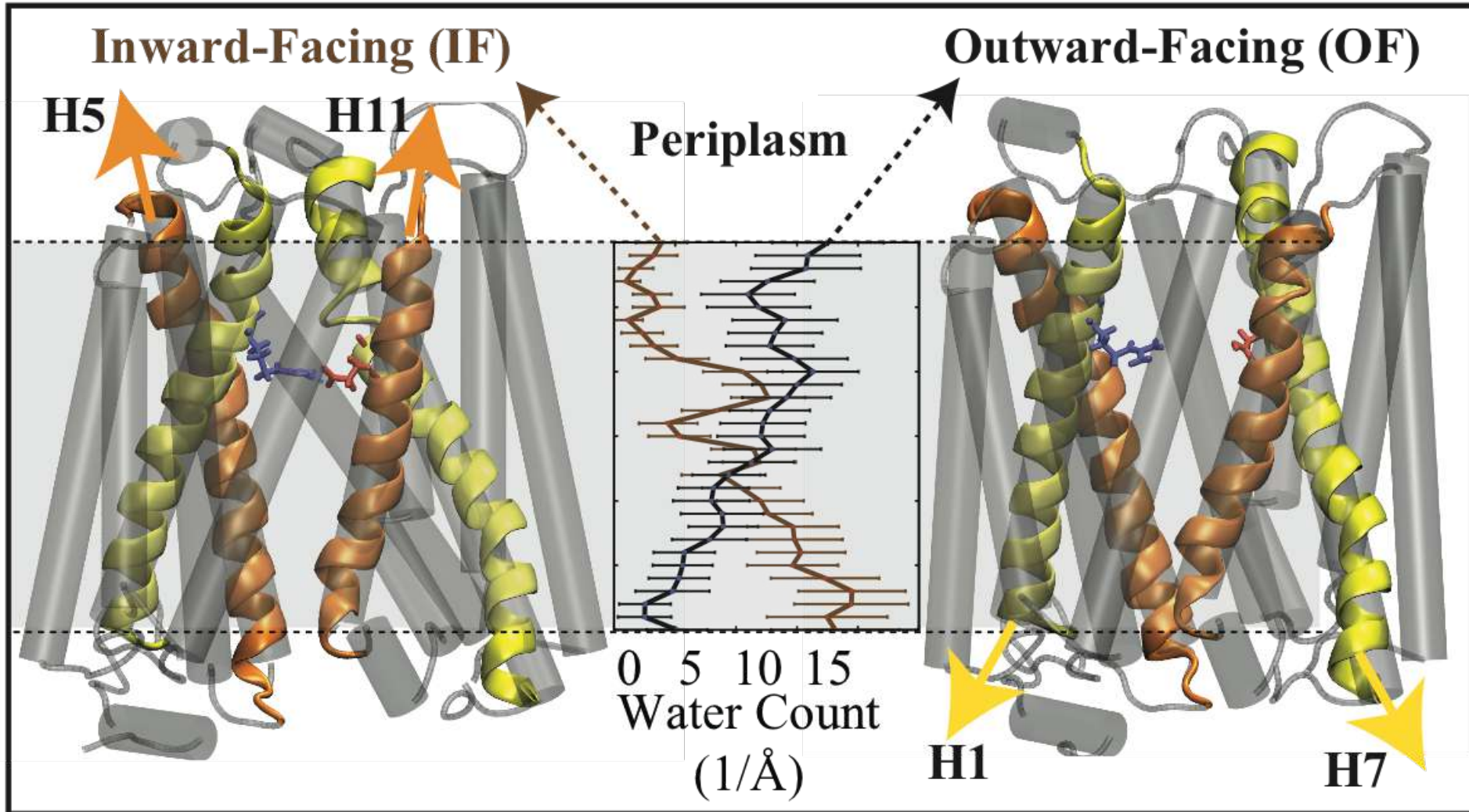
GkPOT

How to study large-scale conformational changes?



If you are not sure how good your collective variable is, start with nonequilibrium pulling.

How to study large-scale conformational changes?



- **Introduction**

- How to study large-scale conformational changes?

- **Methodology**

- Empirical search for good pulling protocols

- Iterative combination of free energy calculation methods and path-finding algorithms

Sampling Ideas

- Free energy calculations require **dimensionality reduction**.
- Traditionally, this is done by designing intuitive, ad-hoc, knowledge-based **collective variables**.
- Another approach is to use **data-driven** collective variables using standard **dimensionality reduction techniques** (PCA, diffusion maps, etc).
- Alternatively (or in combination with the above approaches), one can calculate **free energy along a transition path** (a 1D curve).
- The path can be obtained from **path-finding algorithms**.
- Since sampling is never perfect the procedure could be **iterative** to reach higher accuracies.

Sampling Ideas

- **Reaction coordinates**
 - System-specific collective variables
- **Searching for efficient pulling protocols**
 - An empirical approach to sampling
- **Along-the-curve free energy calculations**
 - Free energy calculations combined with path-finding algorithms
- **Iterative sampling**
 - A posteriori tests of self-consistency

Moradi et al., *Proc Natl Acad Sci* **106** 20746 (2009)

Moradi et al., *Chem Phys Lett* **518** 109 (2011)

Moradi et al., *J Chem Phys* **133** 125104 (2010)

Moradi et al., *Int J Quantum Chem* **110** 2865 (2010)

Moradi et al., *Biophys J* **100** 1083 (2011)

Moradi et al., *J Phys Chem B* **115** 8645 (2011)

Moradi et al., *PLoS Comput Biol* **8** e1002501 (2012)

Moradi et al., *Nucleic Acid Res* **41** 33 (2013)

Moradi et al., *Proc Natl Acad Sci* **110** 18916 (2013)

Moradi et al., *J Phys Chem Lett* **4** 1882 (2013)

Moradi et al., *Methods Mol Biol* **924** 313 (2013)

Moradi et al., *J Chem Phys* **140** 034114 (2014)

Moradi et al., *J Chem Phys* **140** 034115 (2014)

Moradi et al., *J Chem Theory Comput* **10** 2866 (2014)

Moradi et al., *J Phys Conf Ser* **640** 012014 (2015)

Moradi et al., *J Phys Conf Ser* **640** 012020 (2015)

Moradi et al., *Nat Commun* **6** 8393 (2015)

Fakharzadeh & Moradi, *J Phys Chem Lett* **7** 4980 (2016)

Sampling Ideas

- **Reaction coordinates**
 - System-specific collective variables
- **Searching for efficient pulling protocols**
 - An empirical approach to sampling
- **Along-the-curve free energy calculations**
 - Free energy calculations combined with path-finding algorithms
- **Iterative sampling**
 - A posteriori tests of self-consistency

I.1 Defining Practical Collective Variables

Empirical search for practical collective variables for inducing the conformational changes involved in the transition.

I.2 Optimizing the Biasing Protocols

Systematic search for a practical biasing protocol by using different combinations of collective variables.

II. Optimizing the Transition Pathway

Use all of the conformations available to generate the most reliable transition pathway:

1. Bayesian approach for combining the data
2. Post-hoc string method (analysis tool)
3. String method with swarms of trajectories

III.1 Free Energy Calculations

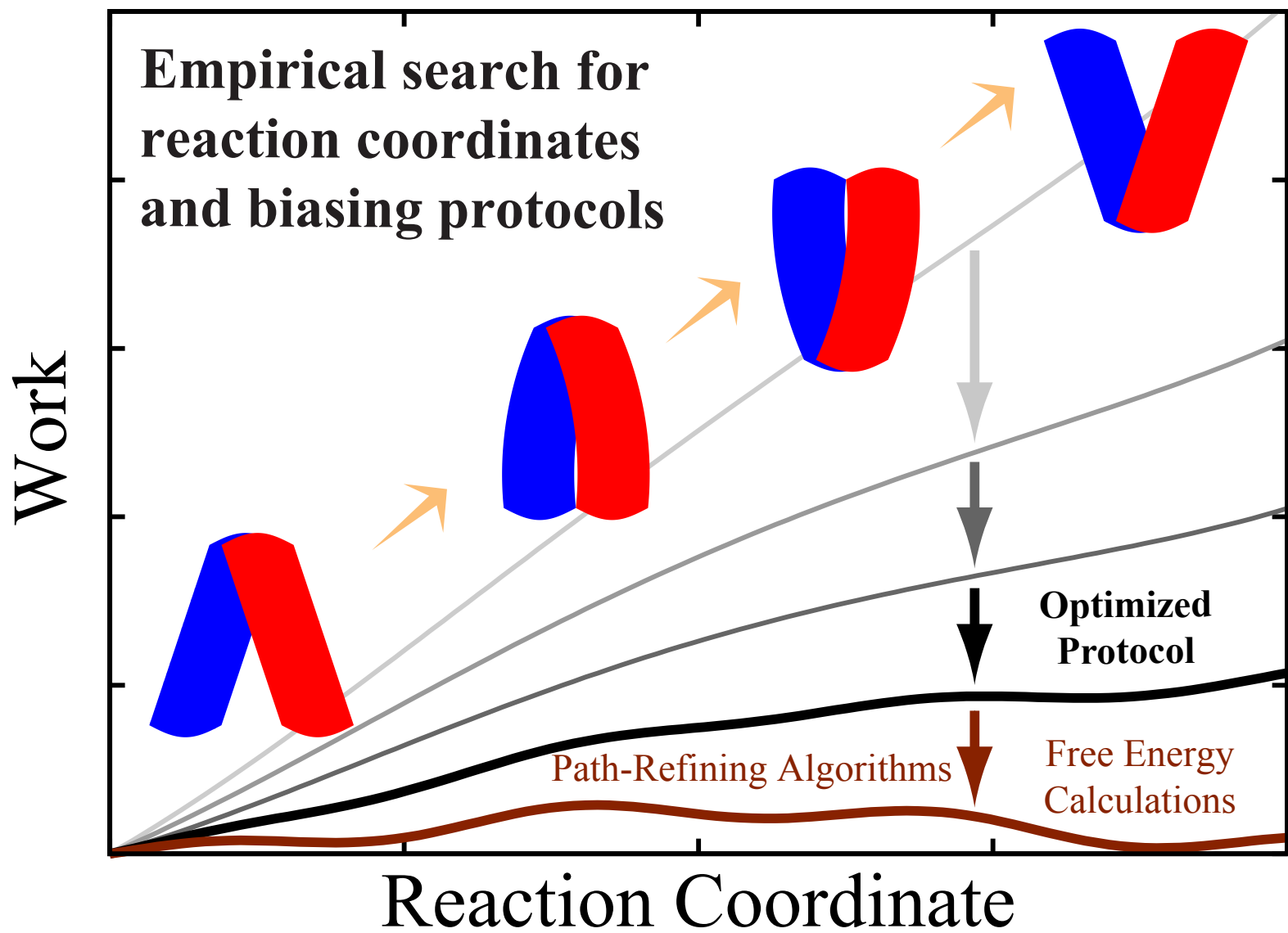
Using the most relevant collective variables (from I.1), biasing protocol (from I.2), and initial conformations (from I.2).

III.2 Assessing the Sampling Efficiency

Detecting the poorly sampled, but potentially important regions, e.g., by using PCA.



Sampling Ideas



- **Introduction**

- How to study large-scale conformational changes?

- **Methodology**

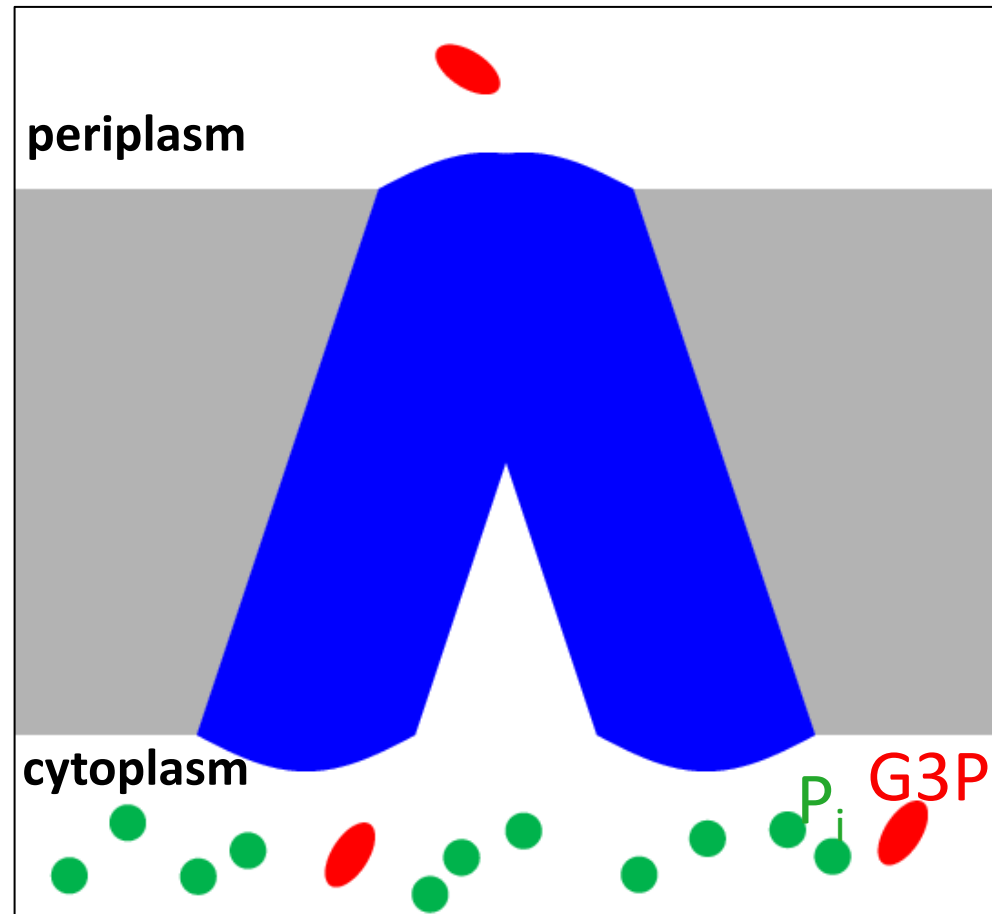
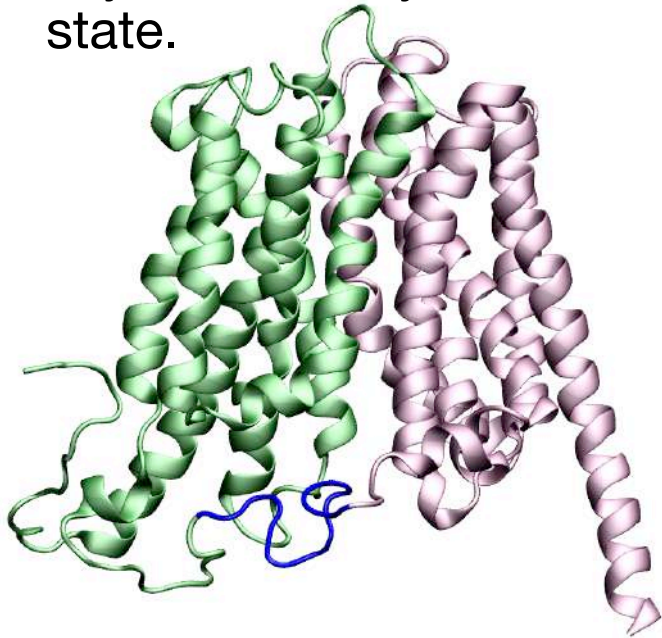
- **Empirical search for good pulling protocols**

- Iterative combination of free energy calculation methods and path-finding algorithms

Example:

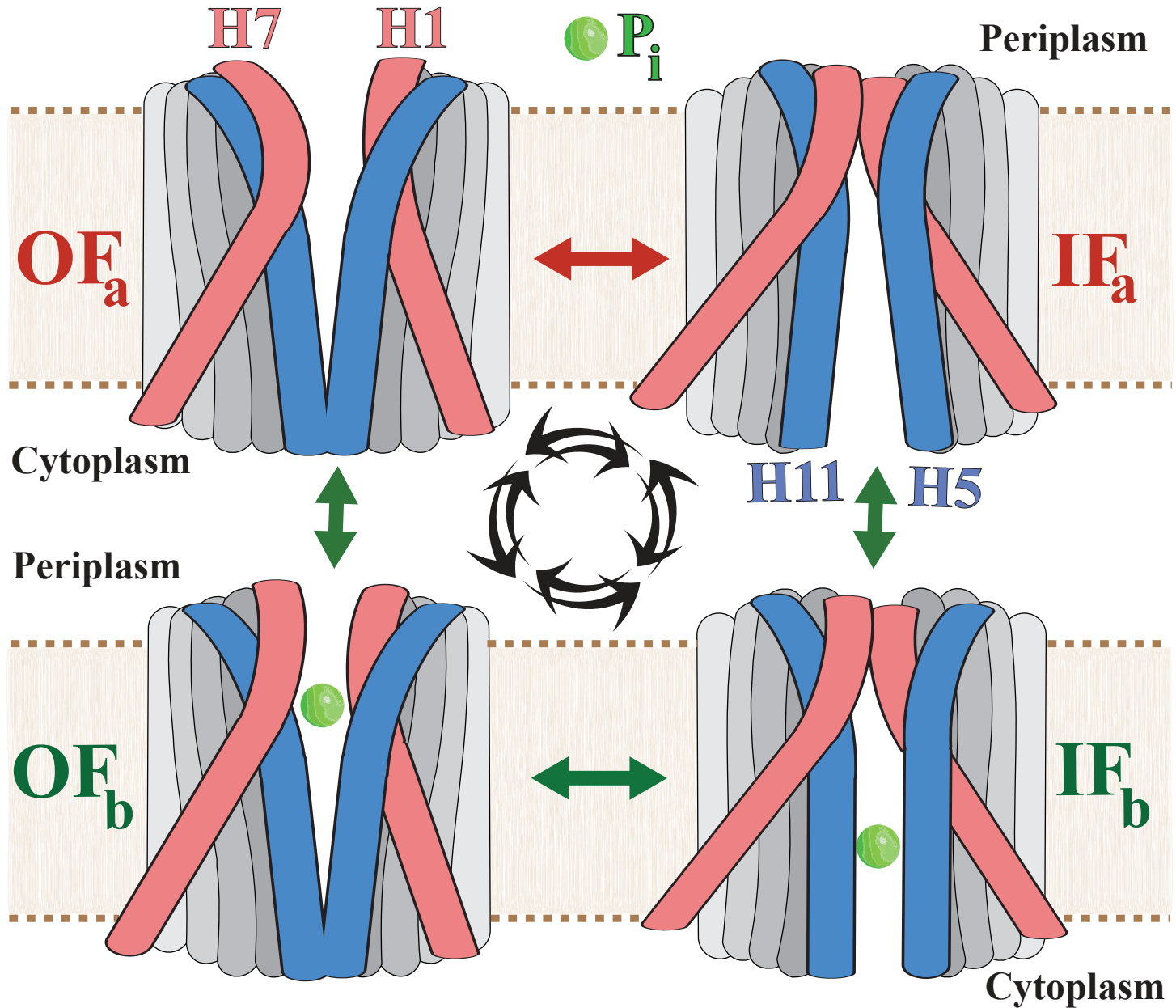
Glycerol-3-Phosphate Transporter (GlpT)

- Major facilitator superfamily (MFS)
- Secondary active transporter
- Crystallized only in the IF state.



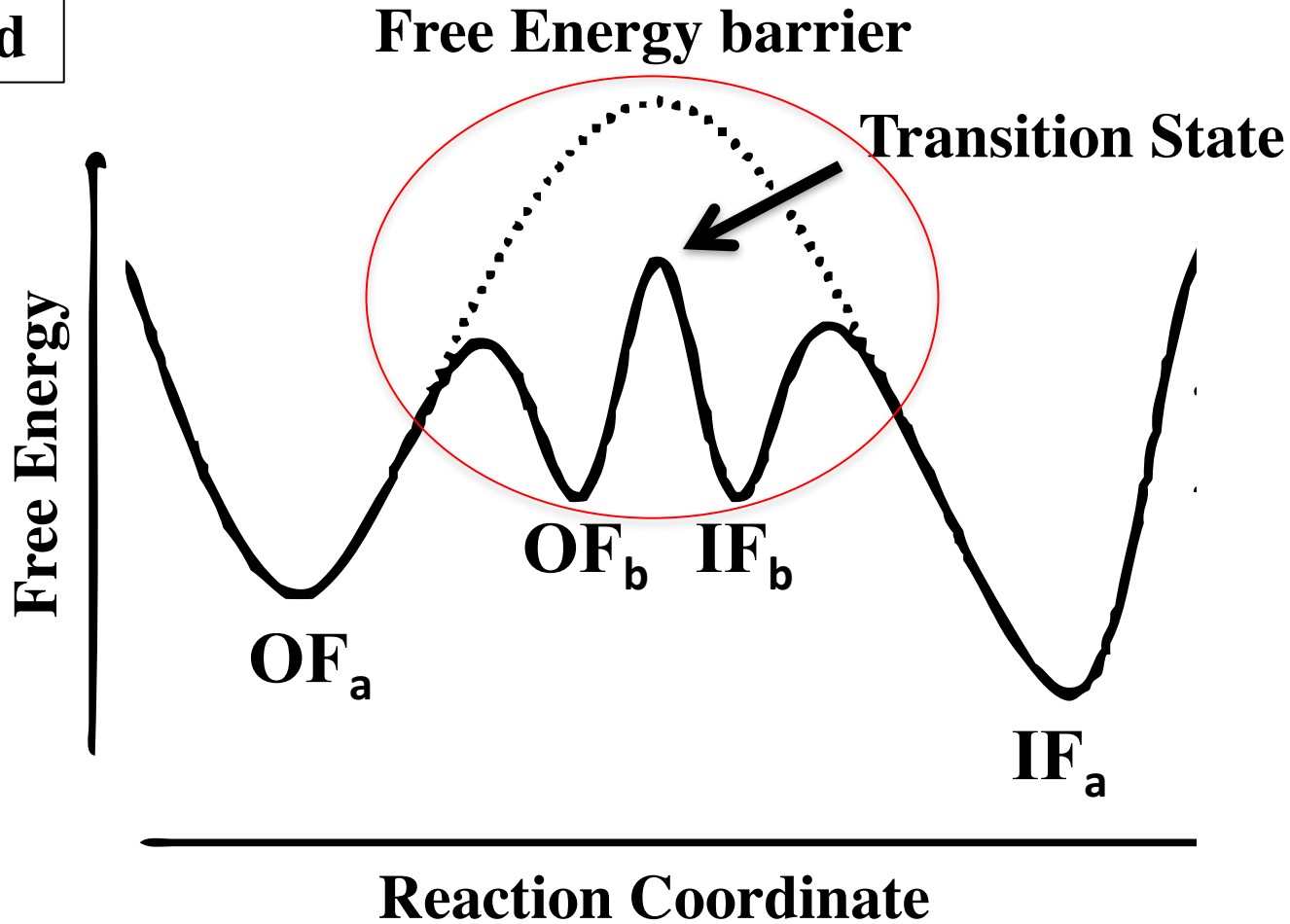
- GlpT transports **G3P** using **P_i** gradient.
- **P_i:P_i** exchanger (in the absence of organic phosphate)

Transport Thermodynamics



Transport Thermodynamics

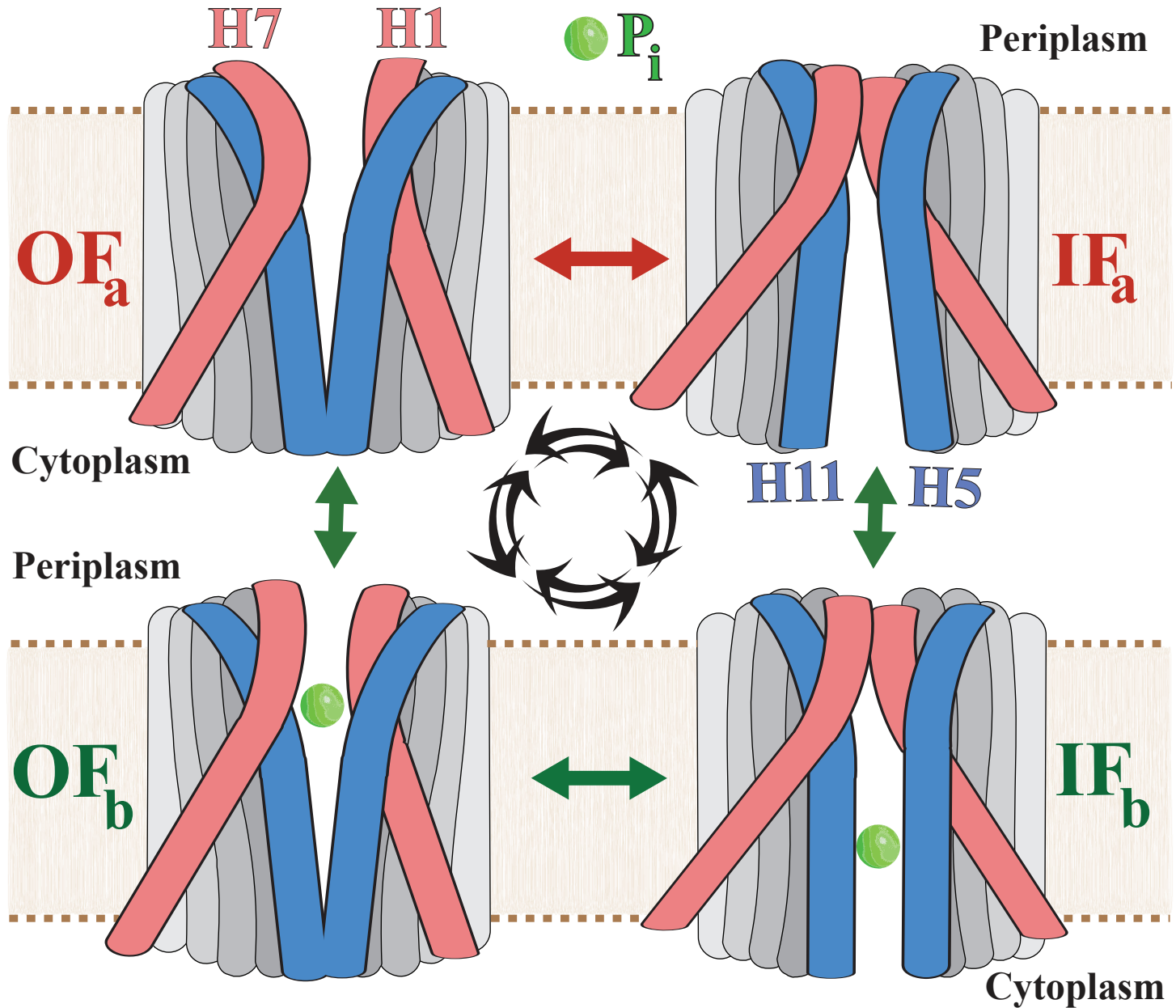
a: *apo*
b: bound



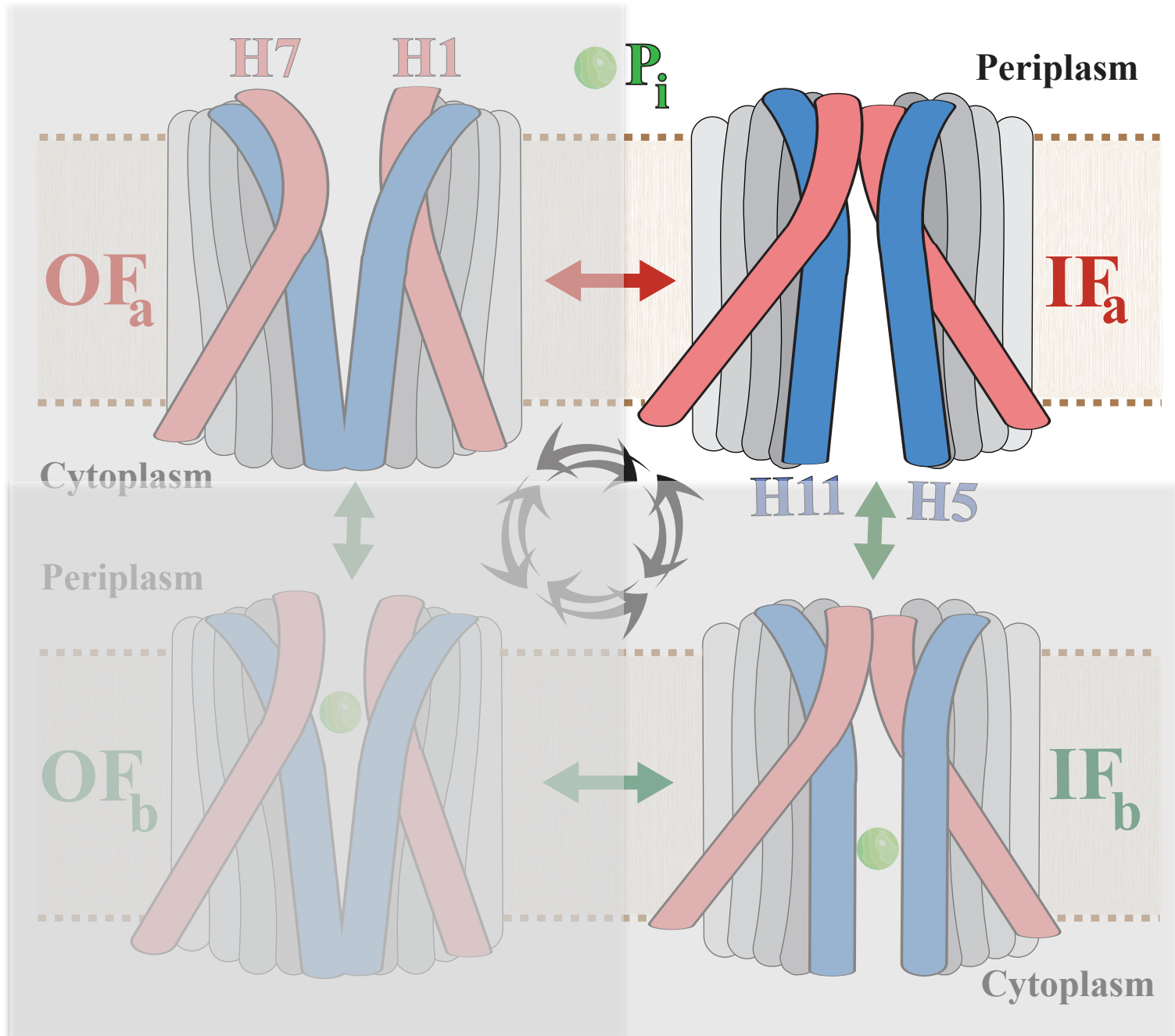
Lemieux, *et al.*, *Curr. Opin. Struct. Biol.* **14**, 405 (2004).

Law, *et al.*, *Biochemistry* **46**, 12190 (2007).

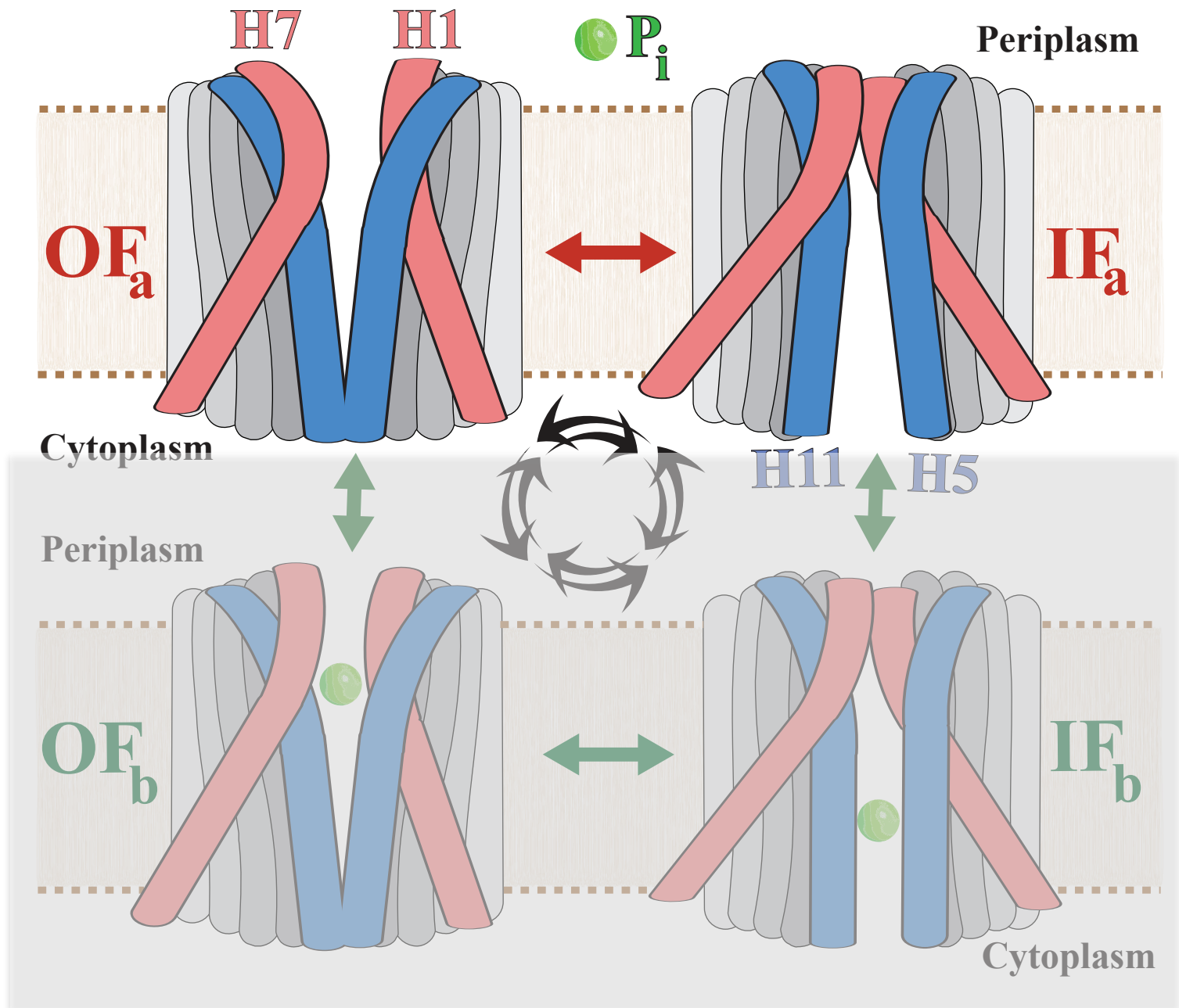
Full Thermodynamic Cycle



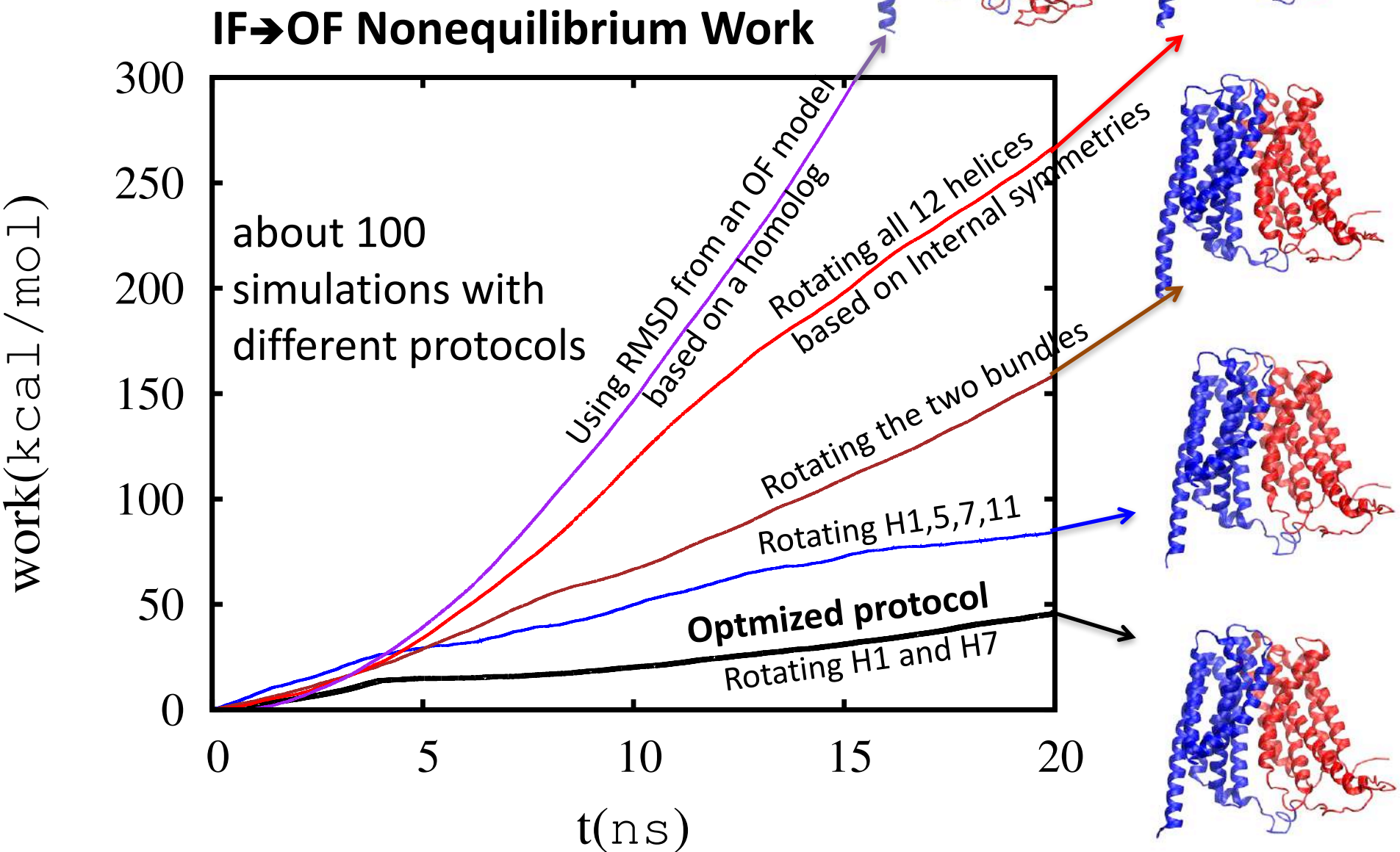
the only available crystal structure



Step 1: $OF_a \leftrightarrow IF_a$



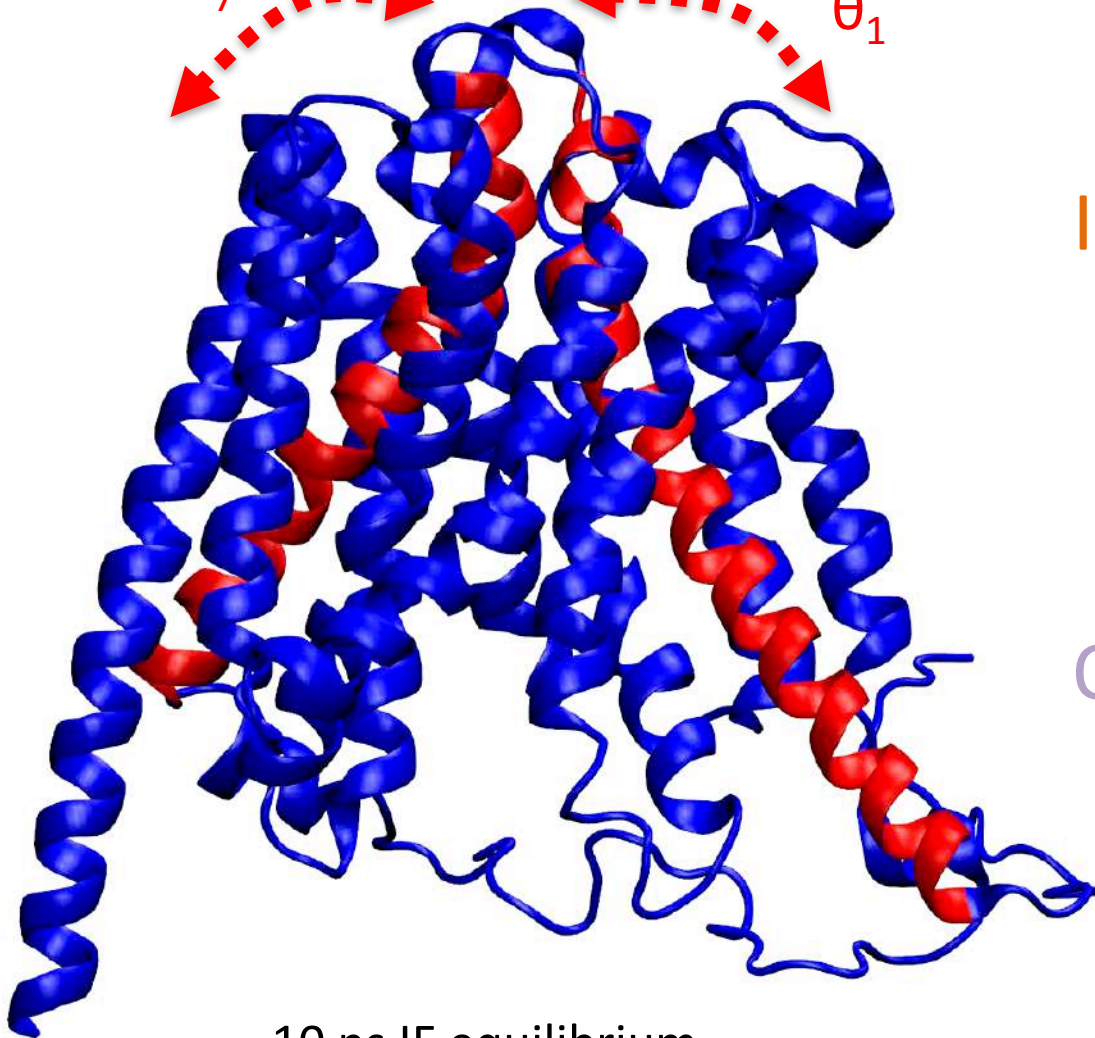
Empirical search for reaction coordinates and biasing protocols:



IF \leftrightarrow OF transition induced by imposing rotational change on

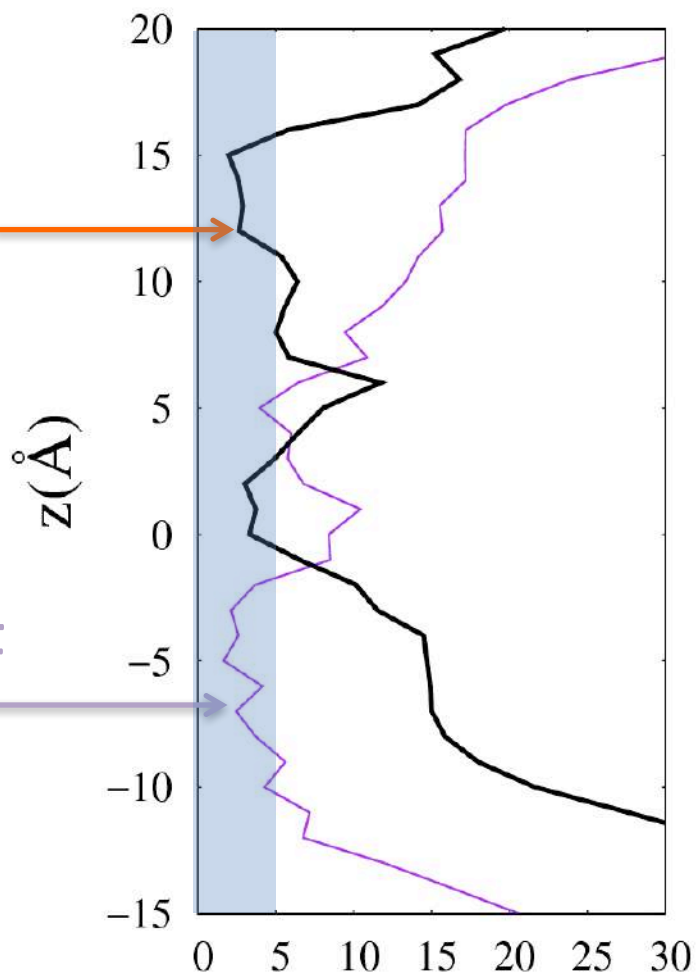
helices **TM1** and **TM7**

θ_7 θ_1



IF

OF



10 ns IF equilibrium

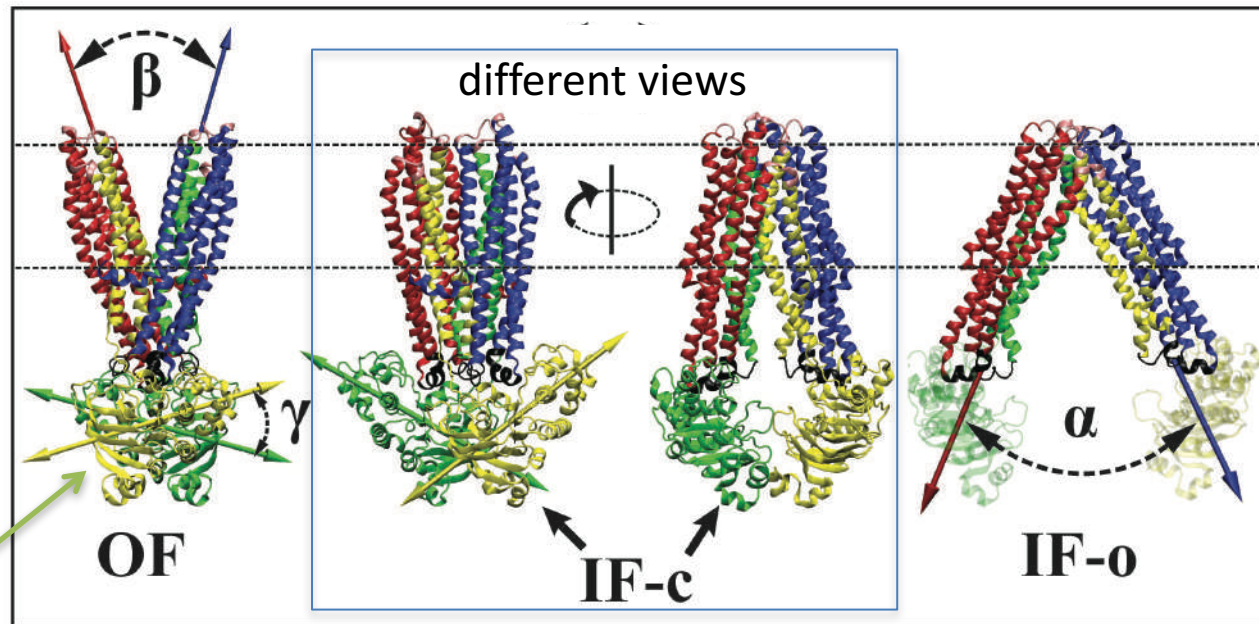
20 ns nonequilibrium (IF \rightarrow OF)

10 ns OF equilibrium

Number of water molecules per \AA
(averaged over a 1 ns window)

Example: MsbA Transporter

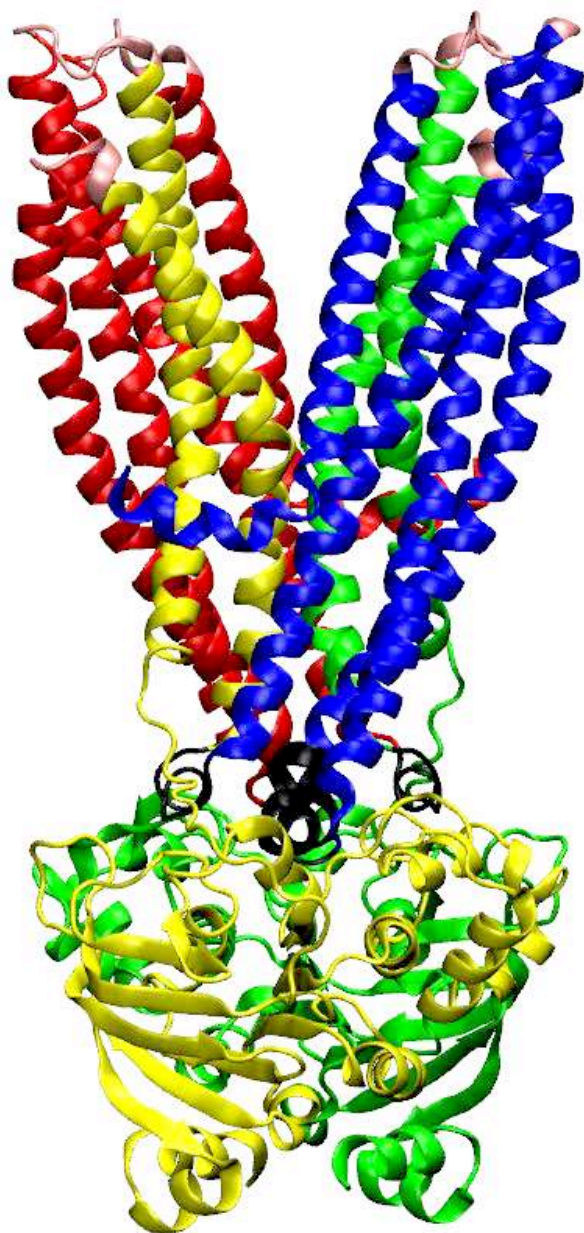
- ATP-binding cassette (**ABC**) exporter with three crystal structures. Ward A., Reyes C. L., Yu J., Roth C. B., Chang G.. PNAS **104** 19005 (2007)
- **Reaction coordinates:**
 α , β , γ (relative orientation of different domains)



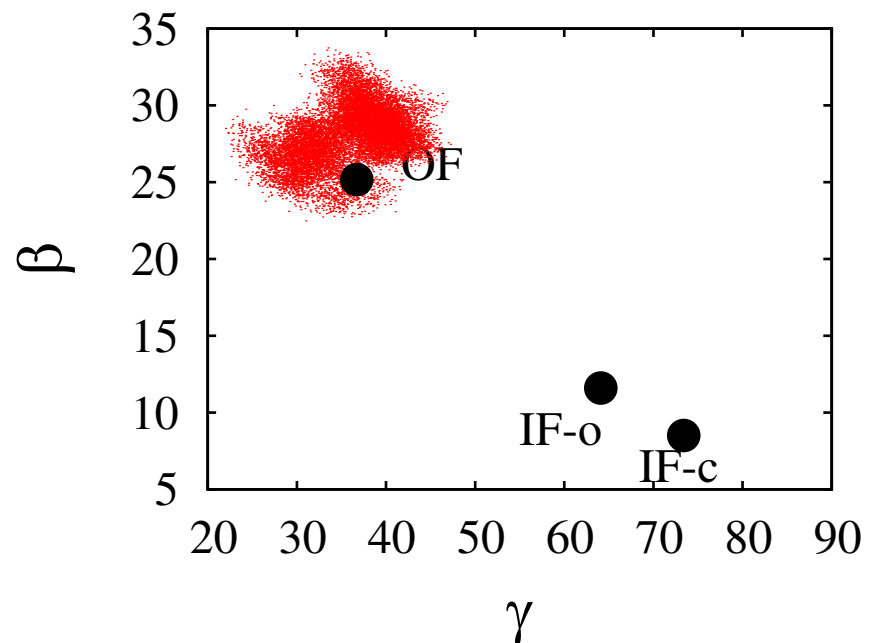
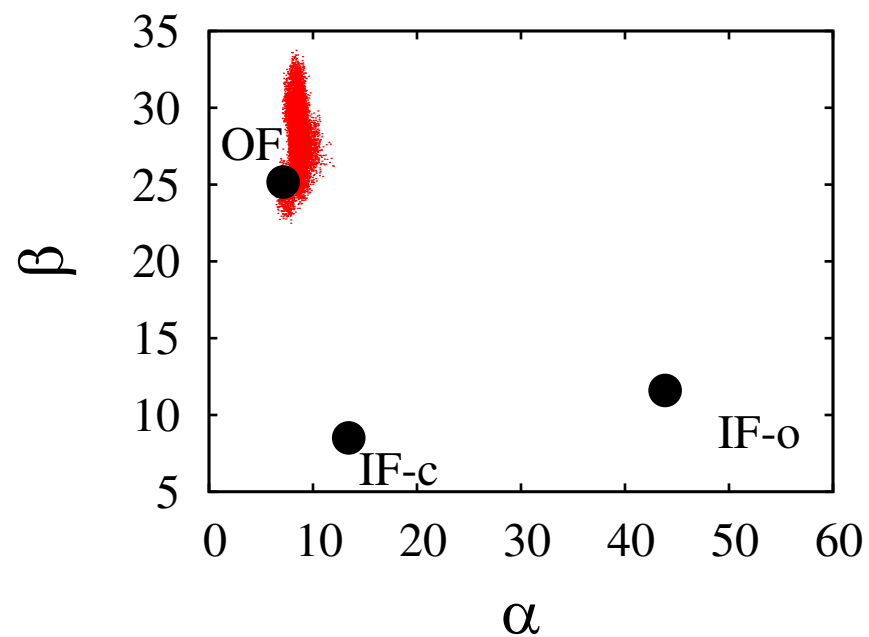
Nucleotide-binding domains (NBD)

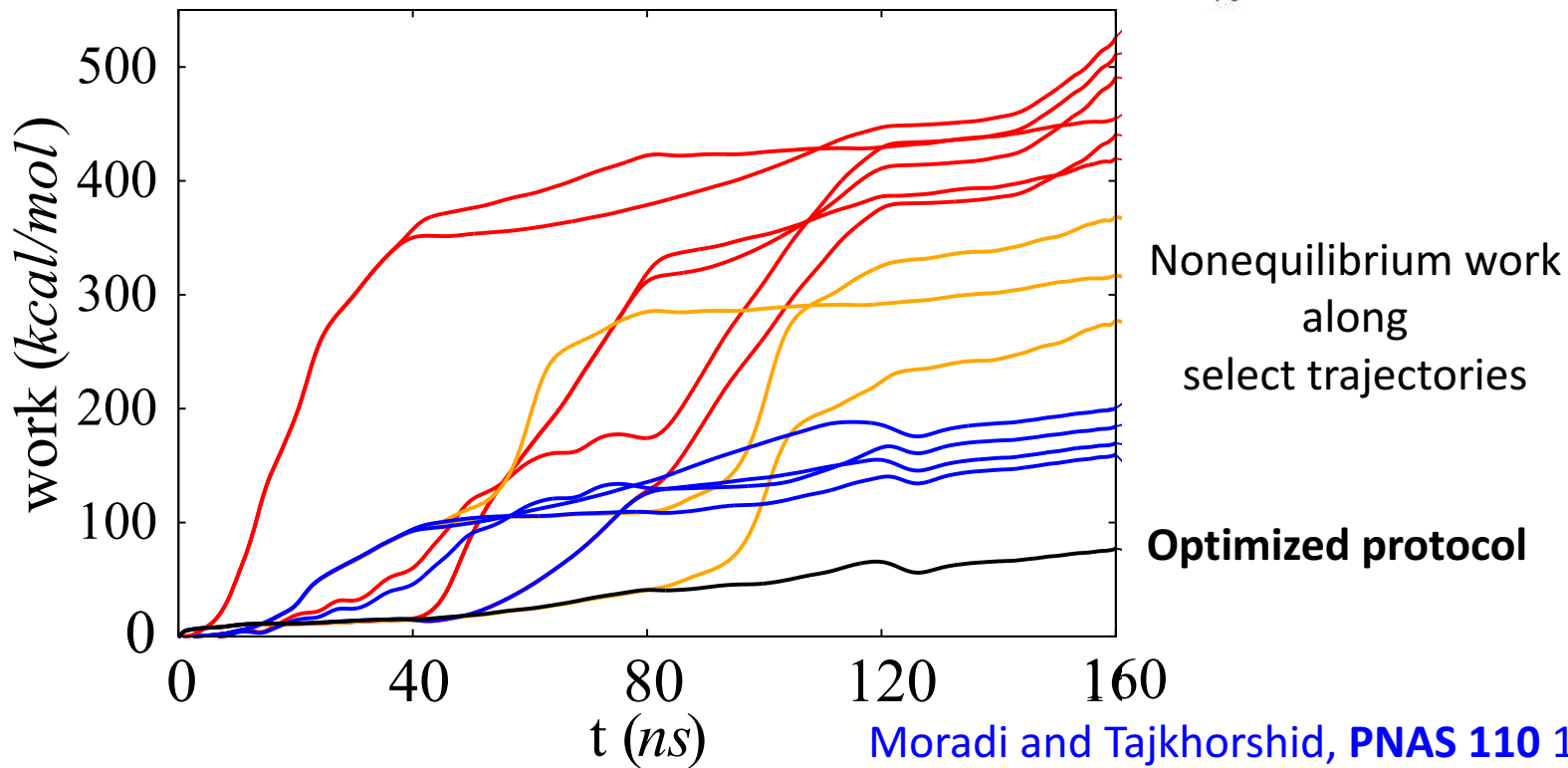
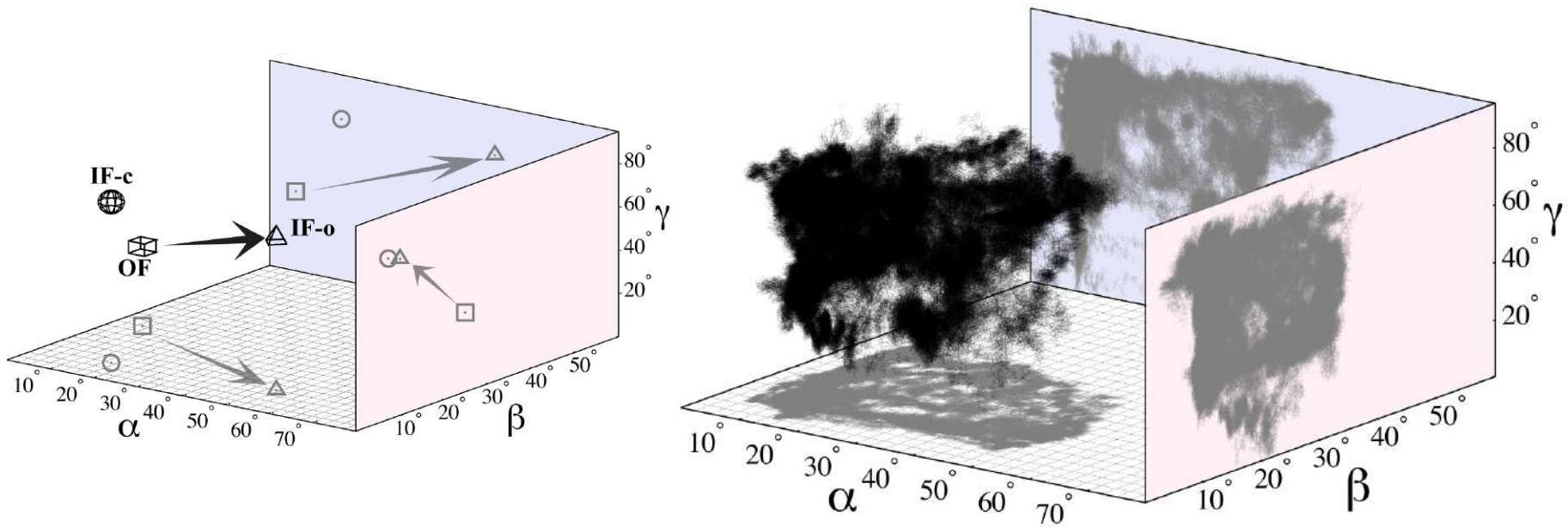
Two IF conformations:
IF-closed (IF-c) and IF-open (IF-o)

Conventional Equilibrium Molecular Dynamics

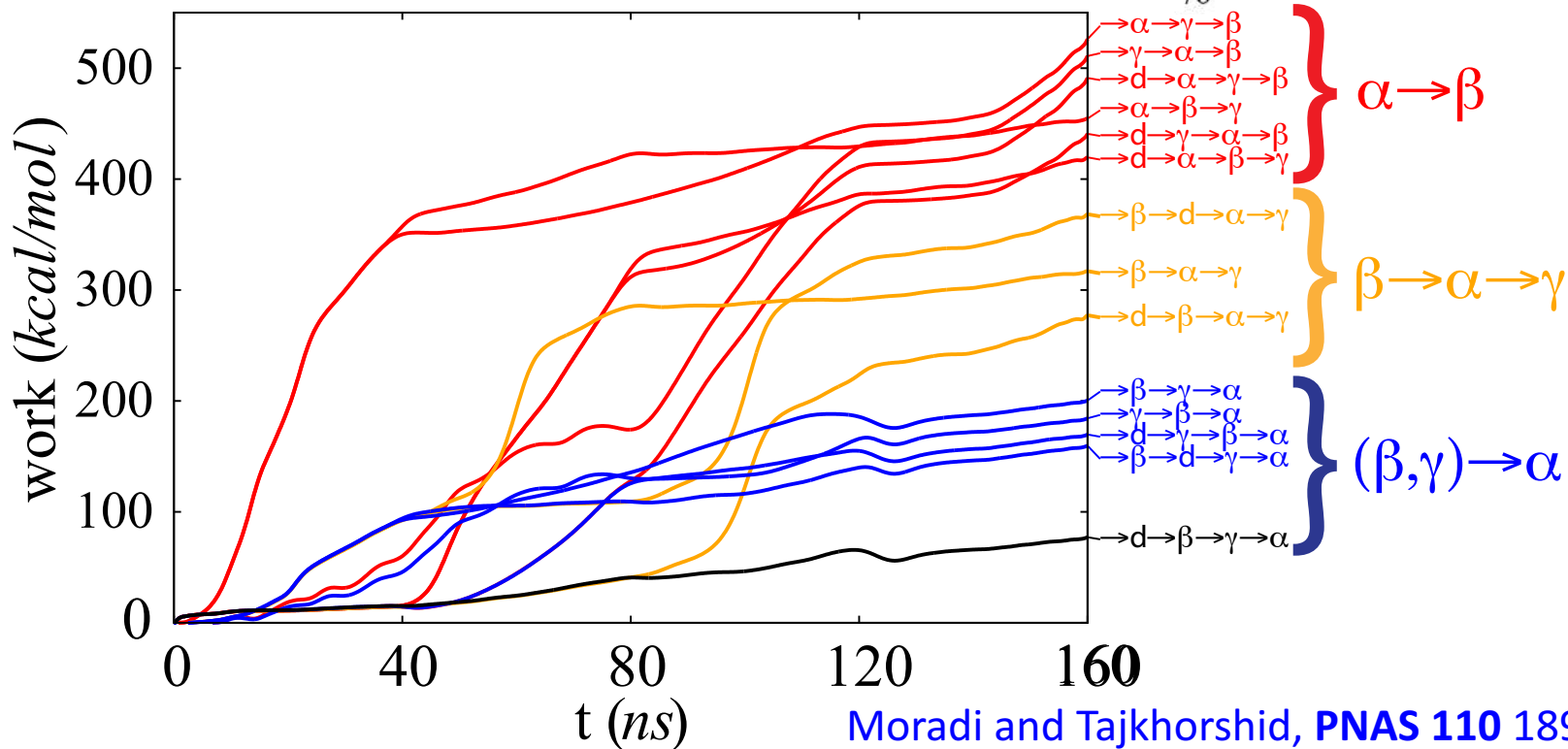
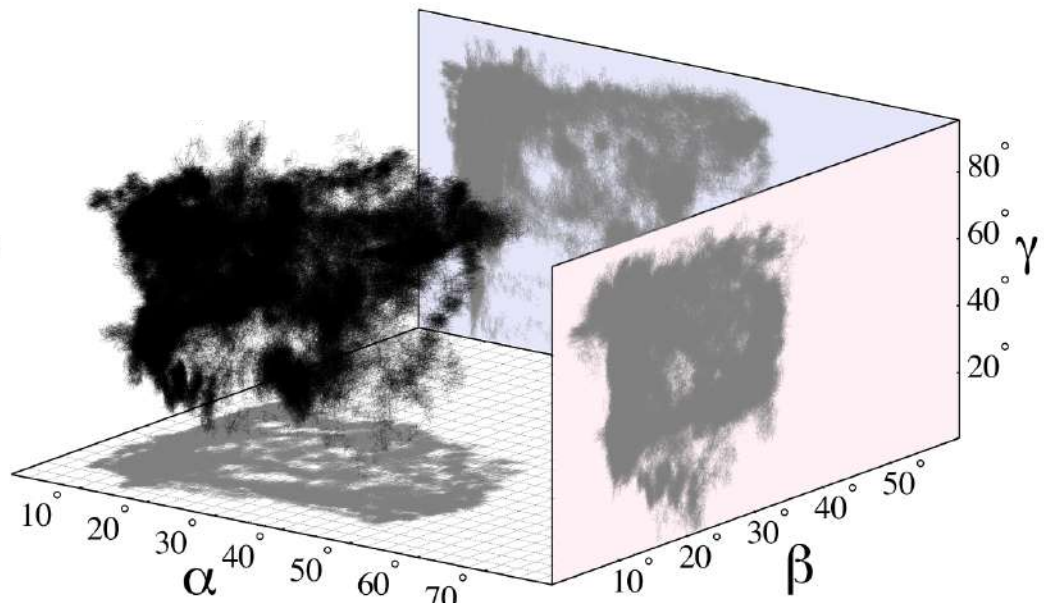
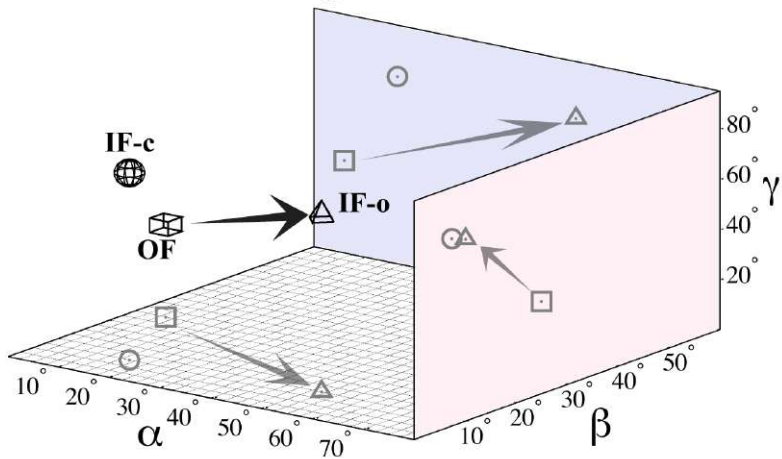


apo MsbA in explicit water/membrane (300 ns)



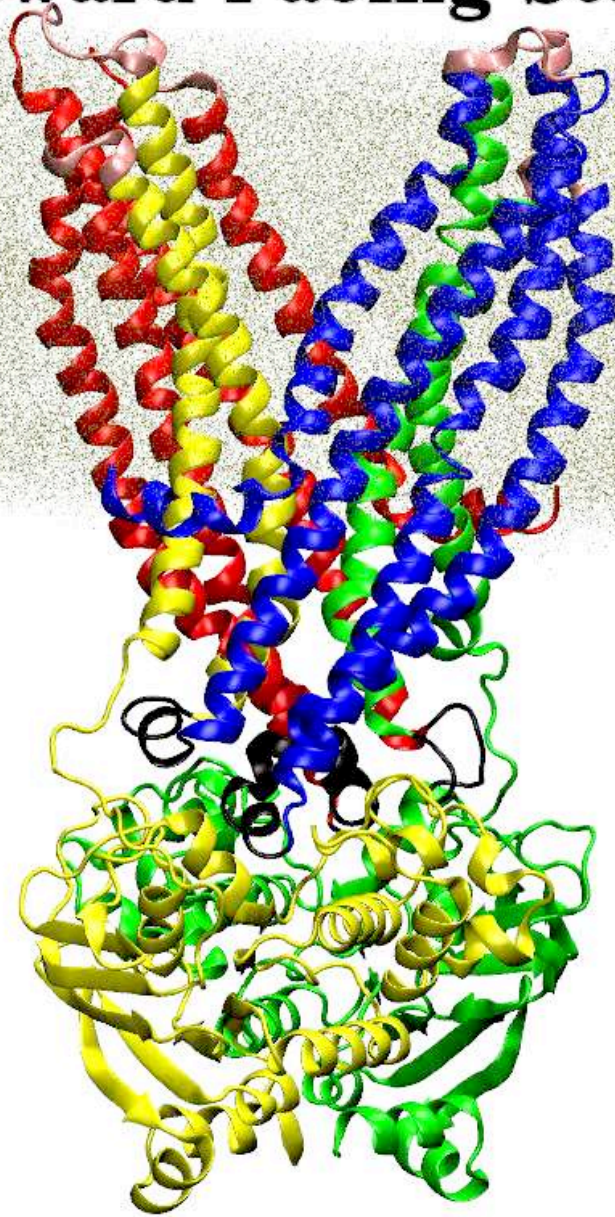


Moradi and Tajkhorshid, **PNAS** 110 18916 (2013)



Moradi and Tajkhorshid, **PNAS** 110 18916 (2013)

Outward-Facing State



OF → IF

NBD Dissociation



Periplasmic Closure



NBD Twist



Cytoplasmic Opening



IF → OF

Cytoplasmic Closure



NBD Twist



Periplasmic Opening



NBD Dimerization



R T R T R T R T R

T Transition

R Relaxation

Steering along orientation quaternion

A quaternion is a 4-component mathematical object:

$$q = (q_0, q_1, q_2, q_3)$$

To find the best rigid body rotation for $\{\mathbf{x}_k\} \rightarrow \{\mathbf{y}_k\}$,
find a unit quaternion q to minimize:

$$\langle |\hat{q}(0, \mathbf{x}_k) \hat{q}^* - (0, \mathbf{y}_k)|^2 \rangle$$

It turns out:

$$q = \left(\cos \frac{\theta}{2}, \sin \frac{\theta}{2} \hat{\mathbf{u}} \right)$$

θ and $\hat{\mathbf{u}}$ are the angle and axis of rotation

Biasing Potential:

$$U_B(q_{ref}(\{\mathbf{x}_k\}), t) = \frac{1}{2} k \Omega^2 (q_{ref}(\{\mathbf{x}_k\}), \overset{\text{target quaternion}}{\text{at time t}} Q(t))$$

$$\cos(\Omega(\hat{p}, \hat{q})) = \hat{p} \cdot \hat{q}$$

Interpolation of orientation quaternion in *colvars*

At each time $t+\Delta t$:

(1) Linear interpolation:

$$(Q'(t + \Delta t) = Q(t) + \frac{Q(T) - Q(t)}{T - t} \Delta t)$$

(2) Normalization:

$$(Q(t + \Delta t) = \frac{Q'(t + \Delta t)}{\|Q'(t + \Delta t)\|})$$

Interpolation of orientation quaternion in NAMD

It turns out:

$$\begin{aligned}\frac{\partial}{\partial t} U_B(q_{ref}, t) &= k \Omega(q_{ref}, Q(t)) \frac{\partial}{\partial t} \Omega(q_{ref}, Q(t)), \\ &= -\frac{\Omega}{\sin(\Omega)} (q - Q(t) \cos(\Omega)) \cdot \frac{Q(T) - Q(t)}{T - t}\end{aligned}$$

that can be used for work measurements using:

$$w^t = \int_0^t \frac{\partial}{\partial t'} U_B(q_{ref}(\{\mathbf{x}_k\}), t') dt'$$

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Non-Parametric Reweighting

$$V_l(\mathbf{x}) = V(\mathbf{x}) + U_l(\mathbf{x})$$

Biased potential

Unbiased potential

Biassing potential

$$g_l(\mathbf{x}) \propto g(\mathbf{x}) q_l(\mathbf{x})$$

Biased density

Known

Unbiased density

Unknown

Biassing factor

Known

Partition function

$$Z_l^{-1} = e^{f_l} \text{ Unknown}$$

$$q_l(\mathbf{x}) = e^{-U_l(\mathbf{x})}$$

$$Z_l = \int d\mathbf{x} g(\mathbf{x}) q_l(\mathbf{x})$$

Conventional WHAM

If the \mathbf{x} space is discrete

or “binable” such that

$$U_l(\mathbf{x}) \approx U_l(\mathbf{x}^k)$$

e.g., U_l is a smooth function of

a 1D coordinate

$$U_l(\mathbf{x}) \approx U_l(\zeta),$$

$$k \equiv \left[\frac{\zeta - \zeta_0}{\Delta\zeta} \right]$$

$$q_l^k \equiv q_l(\mathbf{x}^k)$$

$$g^k = \frac{H^k}{\sum_l N_l Z_l^{-1} q_l^k}$$

$$H^k \equiv \sum_l n_l^k, N_l \equiv \sum_k n_l^k$$

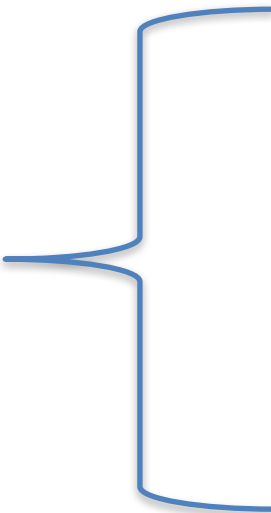
$$Z_l = \sum_k g^k q_l^k$$

S. Kumar, J. M. Rosenberg, D. Bouzida, R. H. Swendsen, P. A. Kollman, “The weighted histogram analysis method for free-energy calculations on biomolecules.”

J. Comput. Chem. **13**, 1011 (1992)

Generalizations

Every conformation sampled is a state.


$$Z_l = \sum_k g^k q_l^k$$

Solved iteratively.

$$g^k = 1 / \sum_l N_l Z_l^{-1} q_l^k$$

C. Bartels, "Analyzing biased Monte Carlo and molecular dynamics simulations."
Chem. Phys. Letters **331**, 446 (2000)

Generalizations

$$Z_l = \sum_k g^k q_l^k$$

$$g^k = 1 / \sum_l N_l Z_l^{-1} q_l^k$$

$$Z_l = \sum_k \frac{q_l^k}{\sum_m N_m Z_m^{-1} q_m^k}$$

$$Z_l = e^{-f_l}$$

$$f_l = -\log \sum_k \frac{q_l^k}{\sum_m N_m e^{f_m} q_m^k}$$

Multi-state BAR (MBAR) equation

M. R. Shirts, J. D. Chodera, "Statistically optimal analysis of samples from multiple equilibrium states."

J. Chem. Phys., **129**, 124105 (2008)

Combining path-finding and free energy methods

- Potential of Mean Force:

$$G(\zeta) = -\beta^{-1} \log \langle \delta(\xi(\mathbf{x}) - \zeta) \rangle$$

$$\langle \delta(\xi(\mathbf{x}) - \zeta) \rangle = \int \delta(\xi(\mathbf{x}) - \zeta) \rho(\mathbf{x}, \mathbf{p}) d^{3N}x d^{3N}p$$

- Perturbed Free Energy:

$$F_i = F(\zeta_i) = -\beta^{-1} \log Z_i$$

$$Z_i = \int e^{-\beta U_i(\xi)} \rho(\mathbf{x}, \mathbf{p}) d^{3N}x d^{3N}p = \int e^{-\beta(G(\xi) + U_i(\xi))} d^n \xi$$

- Non-parametric MLE estimates:

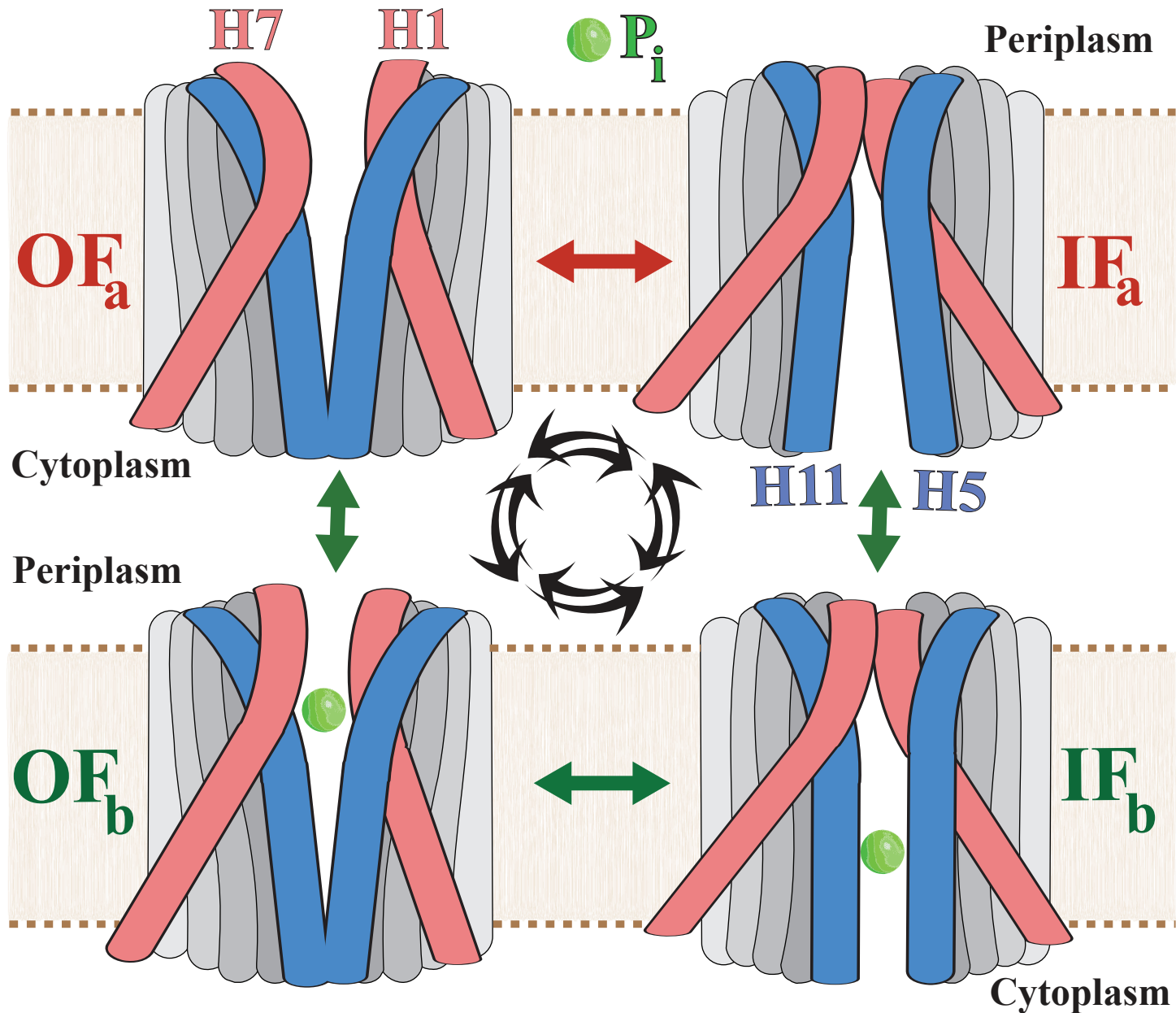
Shirts, Chodera, JCP, **129**, 124105 (2008)

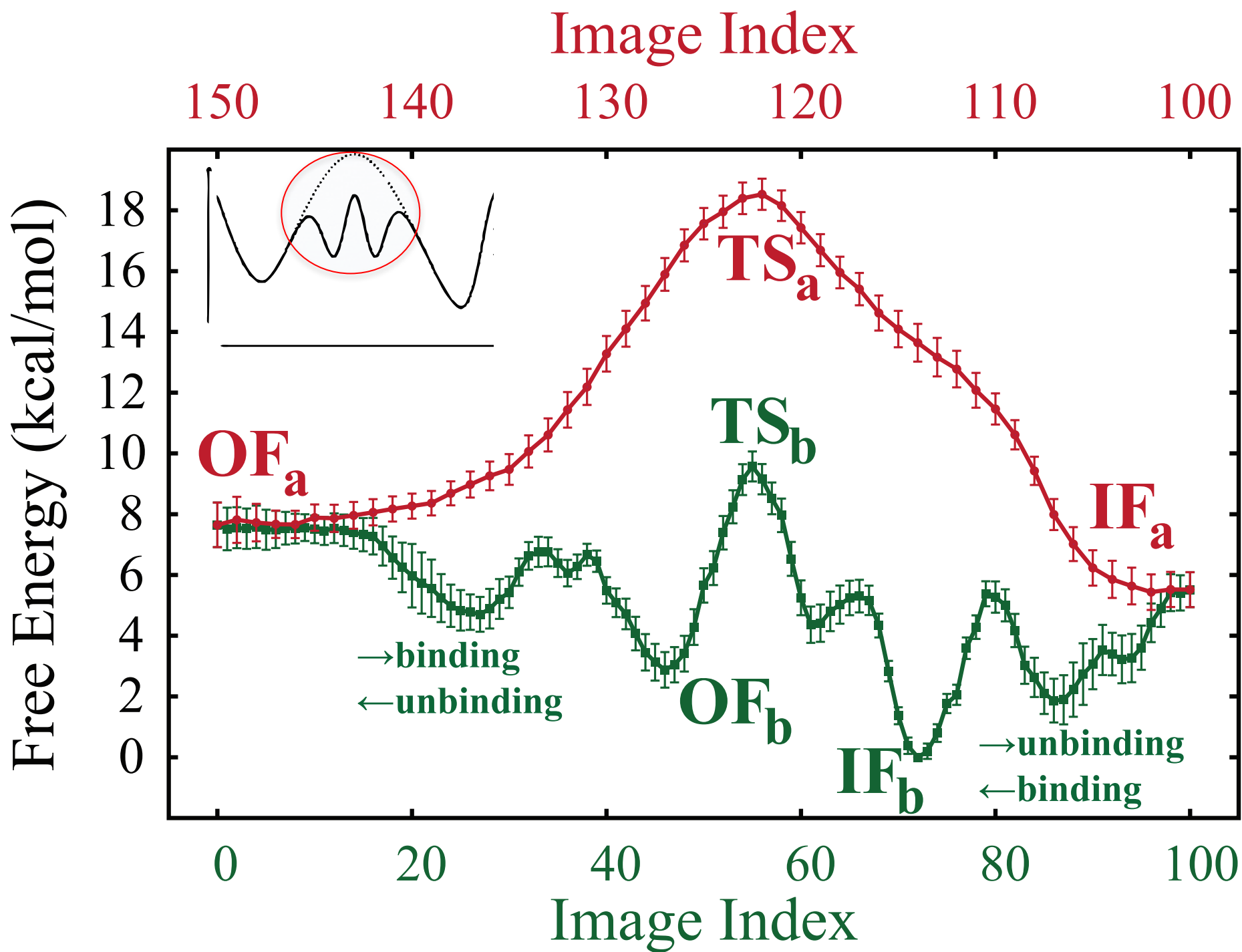
$$e^{-\beta F_i} = \left\langle \frac{e^{-\beta U_i(\xi^t)}}{\sum_j n_j e^{-\beta(U_j(\xi^t) - F_j)}} \right\rangle_{\text{all samples}}$$

Bartels, CPL, **331**, 446 (2000)

$$e^{\beta F_i} = \frac{n_i}{\sum_t w^t e^{-\beta U_i(\xi^t)}}$$

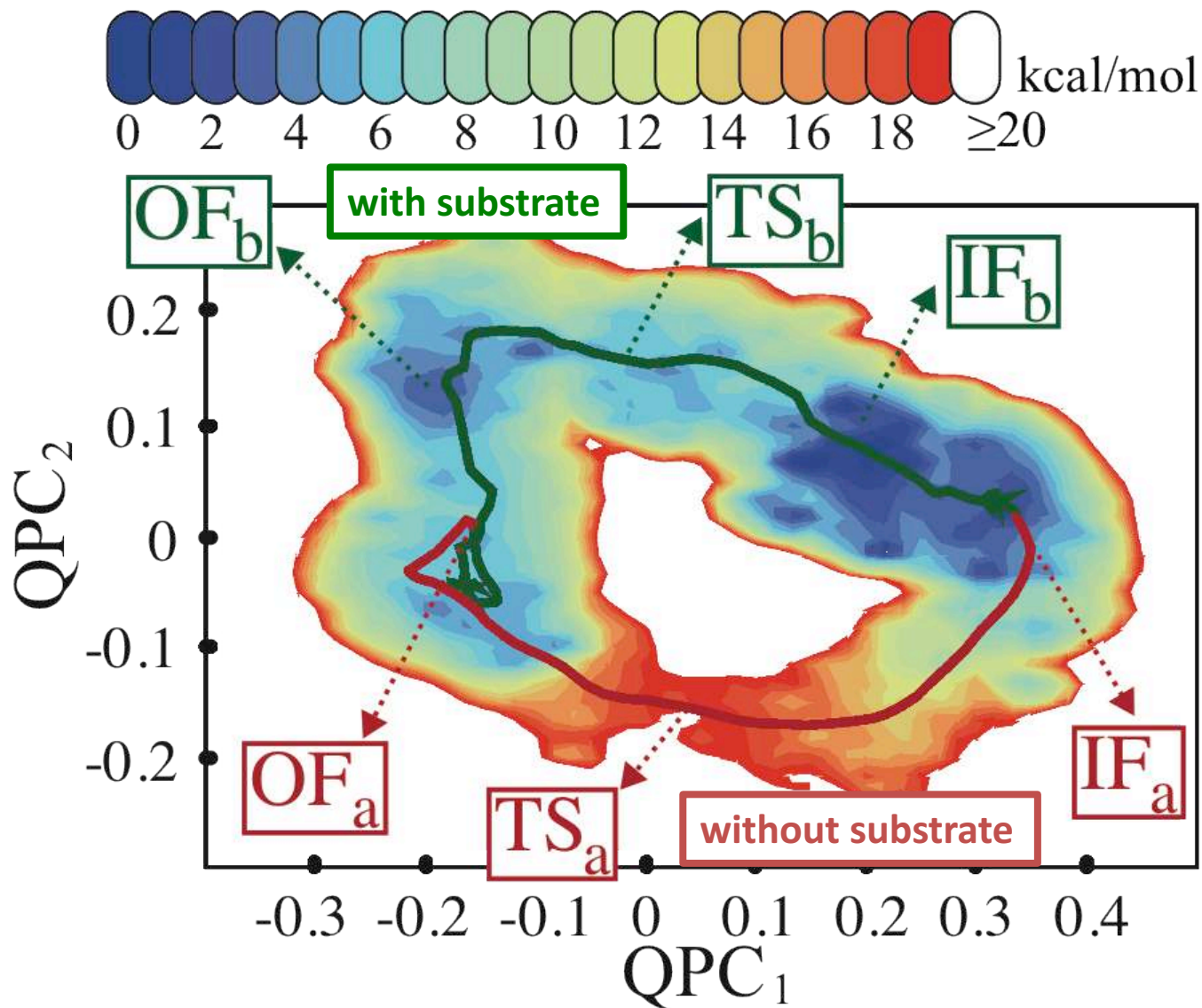
$$w^t = \frac{1}{\sum_i e^{-\beta(U_i(\xi^t) - F_i)}}$$



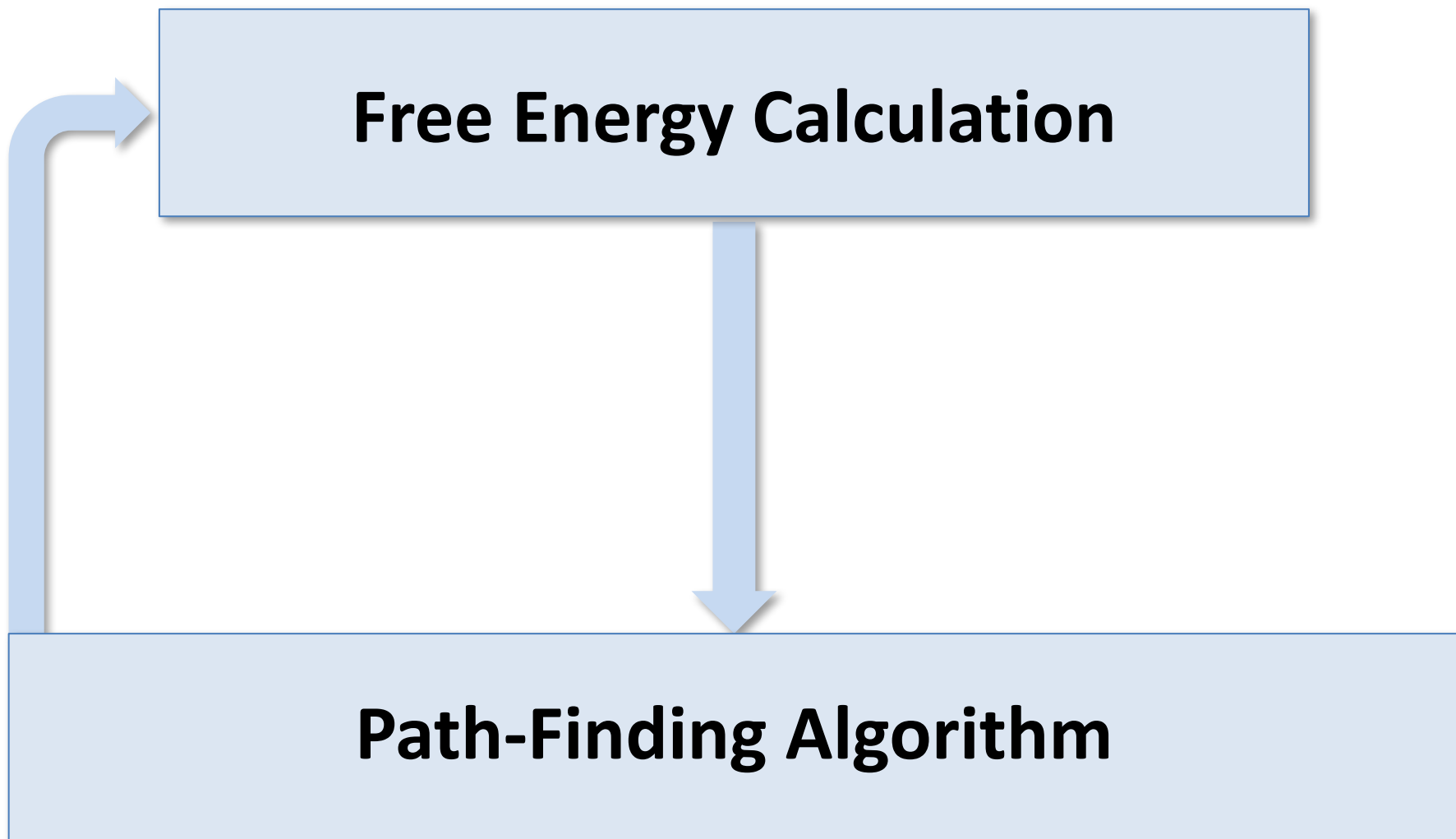


Distinct conformational transition pathways

Quaternion-based principal components (**QPCs**) represent different modes of concerted motions of transmembrane helices.



Iterative path-finding algorithms and free energy calculations



Iterative path-refining algorithms and free energy calculations

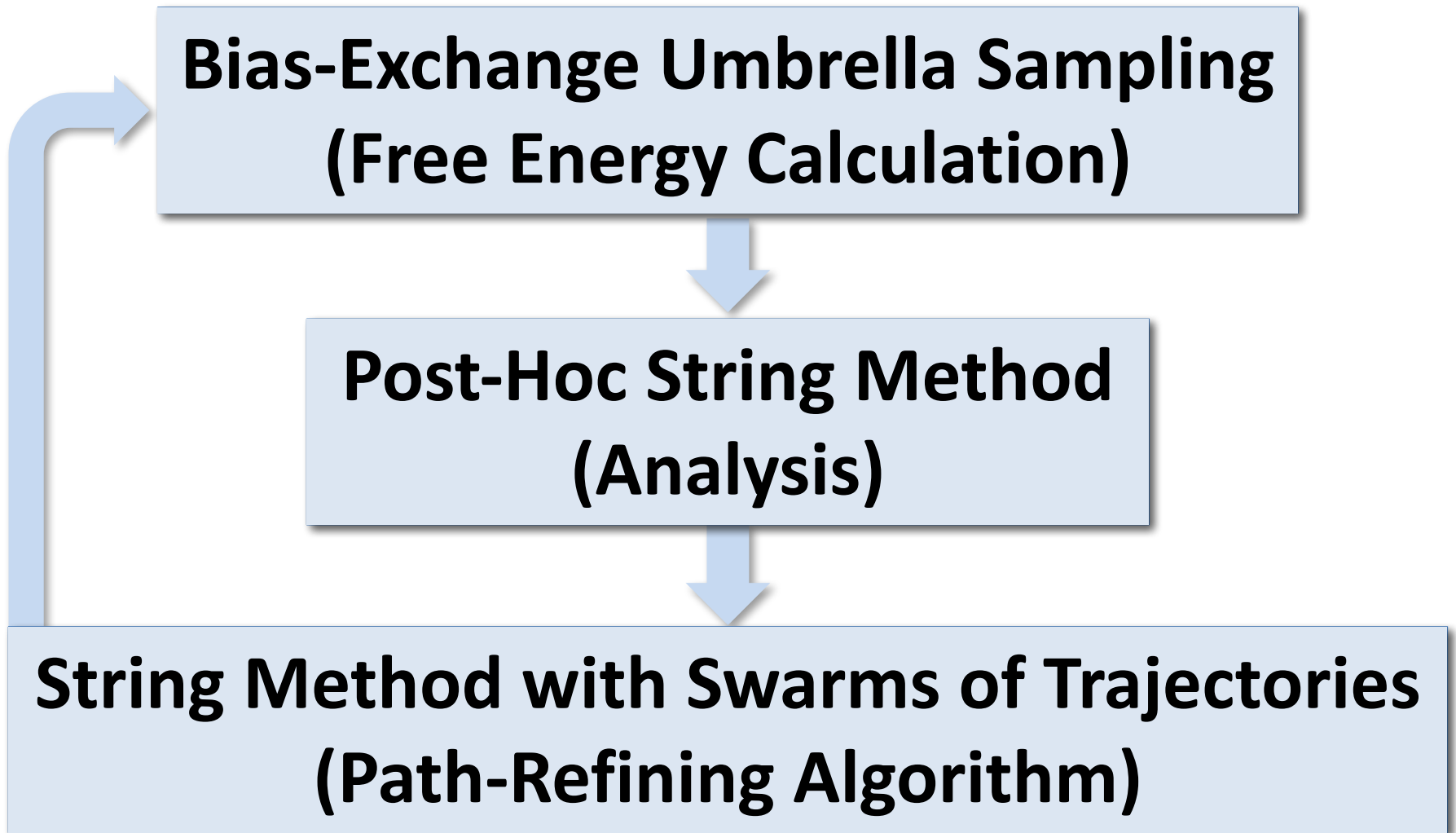
**Bias-Exchange Umbrella Sampling
(Free Energy Calculation)**

```
graph TD; A["Bias-Exchange Umbrella Sampling  
(Free Energy Calculation)"] --> B["String Method with Swarms of Trajectories  
(Path-Refining Algorithm)"]; B --> A;
```

The diagram consists of two light blue rectangular boxes with black text. The top box contains the text 'Bias-Exchange Umbrella Sampling (Free Energy Calculation)'. A thick blue arrow points downwards from the bottom center of this box to the top center of the bottom box. The bottom box contains the text 'String Method with Swarms of Trajectories (Path-Refining Algorithm)'. A thick blue arrow starts from the left side of the bottom box, goes down, then turns left, and finally points up and right into the left side of the top box, forming a loop.

**String Method with Swarms of Trajectories
(Path-Refining Algorithm)**

Iterative path-refining algorithms and free energy calculations



Post-hoc string method (PHSM)

- Suppose we have already sampled a particular “**continuous**” region of configuration space (or some multi-dimensional collective variable space $\{\xi_i\}$) and estimated the **weight** of each sample $\{w^t\}$.
- PHSM finds the **principal curve** in the ξ space using available samples as an **approximate minimum free energy pathway**.

