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Computational Studies of Protein-Inhibitor Interactions

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COMPUTATIONAL STUDIES OF PROTEIN-INHIBITOR INTERACTIONS

BY

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DISSERTATION

Submitted to the University of New Hampshire
in Partial Fulfillment of
the Requirements for the Degree of

Doctor of Philosophy

in

Chemical Engineering

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This dissertation is dedicated to my family.

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NOMENCLATURE

BC — Boundary Conditions.

CV — Collective Variable.

BSA — Buried Surface Area.

DCC — Dynamic Cross Correlation.

DI — Deuterium Incorporation.

GDP — Guanosine Diphosphate.

GPCR — G-protein-coupled Receptor.

GTP — Guanosine Triphosphate.

HDX — Hydrogen-deuterium Exchange.

MD — Molecular Dynamics.

MM/GBSA — Molecular Mechanics Generalized Born Surface Area.

MM/PBSA — Molecular Mechanics Poisson-Boltzmann Surface Area.

TAMD — Temperature Accelerated Molecular Dynamics.

PME — Particle Mesh Ewald.

PMF — Potential of Mean Force.

RCSB — Research Collaboratory for Structural Bioinformatics.

RGS — Regulator of G-protein Signaling.

PDE4 — Phosphodiesterase 4.

RMSD — Root Mean Squared Deviation.

RMSF — Root Mean Squared Fluctuation.

SASA — Solvent Accessible Surface Area.

TDZD — Thiadiazolidinone.

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Abstract

Computational Studies of Protein-Inhibitor Interactions

In this thesis work, I conducted large-scale molecular dynamics (MD) simulation studies of interactions of enzymes and signaling proteins with inhibitory ligands. Specifically, I have studied three classes of proteins: the first part of my thesis reports studies on the hydrogen-producing [FeFe]-hydrogenase enzyme, the second part reports on studies of regulatory proteins from the G-protein coupled receptor (GPCR) family, and the third part reports on studies of the phosphodiesterase (PDE) enzyme family.

In the first part, I studied the problem of [FeFe]-hydrogenase sensitivity to the presence of inhibitory gases oxygen (O_2) and carbon monoxide (CO) that cause irreversible damage to the active site of this enzyme. Therefore, a detailed knowledge of the diffusion pathways of these inhibitory gases is necessary to develop strategies for designing novel enzymes that are tolerant to these gases. Specifically, I studied the diffusion pathways of O_2 and CO in the CpI [FeFe]-hydrogenase from *Clostridium pasteurianum*. I used several enhanced sampling and free-energy simulation methods to reconstruct a three-dimensional free-energy surface for diffusion of each gas which revealed free-energy minima forming an interconnected network of pathways. I discovered multiple pathways of minimal free-energy as diffusion portals for O_2 and CO, and observed that the global minimum in the free-energy surface is located in the vicinity of the active site metal cluster, the H-cluster. Among potential residues that I propose as candidates for future mutagenesis studies to increase the tolerance of this enzyme to both inhibitory gases, 11 residues are shared between O_2 and CO. I hypothesize that these shared candidate residues are potentially useful for designing new variants of this enzyme that are tolerant to both inhibitory gases.

In the second part, I have studied the interplay of protein conformational dynamics and effects of small-molecule inhibitors in a class of signaling proteins, known as the Regulators of G protein signaling (RGS) proteins, that negatively modulate signaling in GPCRs. Recently discovered thiadiazolidinone (TDZD) compounds that target cysteine residues have shown different levels of specificities and potencies for several known RGS proteins, thereby suggesting intrinsic differences in dynamics of these proteins upon binding of these compounds. I characterized the effect of binding of several small-molecule inhibitors on perturbations and dynamical motions in RGS4.

Specifically, I studied two conformational models of RGS4 in which a buried cysteine residue is solvent-exposed due to side-chain motions or due to flexibility in neighboring helices. I found that TDZD compounds with aromatic functional groups perturb the RGS4 structure more than compounds with aliphatic functional groups. Moreover, small-molecules with aromatic functional groups but lacking sulfur atoms only transiently reside within the protein and spontaneously dissociate to the solvent. I further probed the salt-bridges forming across isoforms of RGS proteins, resulting in a hypothesis that differences in salt-bridges between a pair of helices in RGS proteins are responsible for differences in flexibility and potency among isoforms.

In the final part of my thesis, I evaluated differences in binding interactions of phosphodiesterase 4 (PDE4) inhibitors within the PDE4 catalytic domain. From residues within 5 Å of the ligand binding site, five residues revealed significant differences in non-bonded interaction energies that could account for the differential binding affinities of inhibitors. I found one site (Phe506 in human PDE4; Tyr253 in the *C. elegans* PDE4 catalytic domain) that alters the binding conformation of roflumilast and zardaverine (human PDE 4 inhibitors) into a less energetically favorable state. These results support the feasibility of designing the next-generation of anthelmintics/nematicides that could selectively bind to nematode PDEs. Overall, my thesis has resulted in enhancement of detailed mechanistic insights into several protein and inhibitory ligand interactions that are potentially useful in the development of novel inhibitors targeting protein/protein and protein/ligand interactions.

CHAPTER 1

INTRODUCTION

1.1 Protein-Protein Interactions

The past decade has seen many advances in molecular understanding of regulatory systems critical in physiology and pathology, and protein-protein interactions (PPIs) (Fig. 1.1) signify a key objective space for pharmacological intervention. Recent genomics advances found that there are up to 10000 disease related proteins [18], and yet only approximately 400 proteins have been explored for therapeutic development. The majority of these proteins are the G protein-coupled receptors (GPCRs), enzymes, ion channels, and nuclear hormone receptors [19].

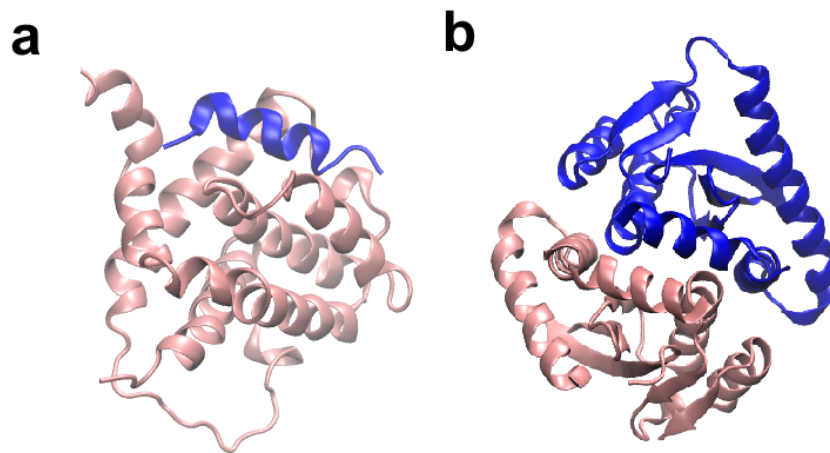


Figure 1.1: The protein-protein interaction (a) between a protein and a peptide (Protein Data Bank; PDB:1BXL), and (b) between two proteins (Protein Data Bank; PDB:2B4J).

Most of these disease related proteins exert their functions via interactions with other proteins (Fig. 1.1), and recent developments demonstrate that PPIs can be modulated with small-molecule drugs and peptides [19]. To enhance development of novel therapeutic compounds, a molecular understanding of the structure and dynamics of interacting proteins is essential.

1.2 Protein-Ligand Interactions

Protein functions are usually coupled to binding of small molecules such as substrates or ligands. These ligands often interact with endogenous ligand-binding sites of their target proteins. The modulation of a target protein by a ligand has been extensively explored for a diverse range of objectives in drug discovery [19]. The interaction of a ligand with a protein is a complex biophysical event, as the surface of a protein constitutes a multitude of polar and hydrophobic sites that represent challenges to understanding the interaction mechanism of a ligand with the target protein.

1.3 Classification of ligands

Ligands modulate proteins and PPIs via three broad mechanisms ([18]): (1) orthosteric inhibition, (2) allosteric regulation, and (3) interfacial binding. The orthosteric ligands bind target proteins at the overlapping areas in PPIs, directly inhibiting the formation of macromolecular complexes. The allosteric inhibitors bind to their target protein at a site distinct from the macromolecular interface. The interfacial inhibitors bind to a pocket at the internal space or macromolecular interface often transiently formed, and cause a nonproductive conformation of the protein/protein complex. In this thesis, I studied three categories of ligands, small-molecule interfacial inhibitors, orthosteric inhibitors and allosteric inhibitors.

1.3.1 Small-Molecules Inhibitors

Small-molecule inhibitors have been extensively explored in modulation of many diseases related to proteins [20]. These inhibitors could be as simple as small gas molecules O_2 and CO , that are known to inhibit enzymes with metallic clusters. Most proteins exert their functions through interactions with other proteins, and they often lack obvious druggable pockets for small molecules.

Although PPIs are essential for many aspects of cellular functions and provide vast potential for drug development, targeting PPIs with small molecules is challenging for two main reasons. Firstly, the interacting areas between proteins typically contain numerous polar and hydrophobic interactions distributed across a large interface. Unless the interaction hotspot can be identified, a small molecule with a reduced contact area with a protein has difficulty in achieving effective binding due to lower interactions. Secondly, the protein-protein interfaces are typically featureless. The lack of pockets limits the site of binding to only limited areas of a small molecule; such binding is much less effective than binding with multiple areas in a deeper pocket. Despite these difficulties, there has been a growing interest in identifying small-molecule inhibitors against PPIs [21].

1.3.2 Allosteric Inhibitors

These small-molecule ligands bind to target proteins at sites distinct from the macromolecular interface. Ligand binding induces changes in either the static conformation or the dynamic properties of the target protein and hinders the macromolecular interaction in an allosteric (distant) manner. In this thesis, I studied thiadiazolidinone (TDZD) compounds that have been successfully exploited to modulate the activity of their target proteins allosterically.

1.4 Background of Major Proteins Studied

In this thesis, I have studied three essential classes of proteins that are affected by ligands: (i) [FeFe]-hydrogenase, (ii) Regulators of G-protein signaling (RGS) proteins, and (iii) Phosphodiesterase 4 (PDE4) enzyme. The hydrogen-producing [FeFe]-hydrogenase is a promising source of renewable H₂. RGS proteins are clinically significant drug targets. PDE4 enzymes are drug targets for the development of nematocides. In this thesis, I explored these three classes of proteins and their interactions with inhibitory compounds.

1.4.1 FeFe-hydrogenase

Biological H₂ production from [FeFe]-hydrogenase, found in photosynthetic microorganisms, is an attractive alternative source of renewable energy in future [22, 23, 24]. Among such microorganisms, green algae and cyanobacteria are potential sources of H₂ production because they have a maximum solar to hydrogen (STH) conversion efficiency [11, 25, 26]. The hydrogenase family is broadly classified into [NiFe]-hydrogenase and [FeFe]-hydrogenase, based upon the metal-cluster present in the active-site [27, 28]. In comparison to [NiFe]-hydrogenase, [FeFe]-hydrogenase is primarily responsible for H₂ production, but is significantly inhibited by O₂ [29, 30]. In this work, I have focused on CpI [FeFe]-hydrogenase from *Clostridium pasteurianum*. Fig. 1.2 shows the crystal structure (PDB code 1FEH) of the CpI [FeFe]-hydrogenase containing 574 amino-acid residues, and six metal clusters: four [Fe₄S₄] metal clusters, three of which are ligated to four cysteines, distal metal cluster that is ligated to three cysteines and one histidine, and one surface cluster [Fe₂S₂] that is ligated to four cysteines.

The central metal cluster, which is known as the H-cluster, consists of [2Fe]-subclusters with ligation to carbonyl and cyanide groups [31]. Hydrogen production in this enzyme takes place at the H-cluster buried inside the protein [32]. In [FeFe]-hydrogenase, the H-cluster is a complex structure consisting of Fe₂(CO)₃(CN)₂(dithiomethylamine), which is covalently

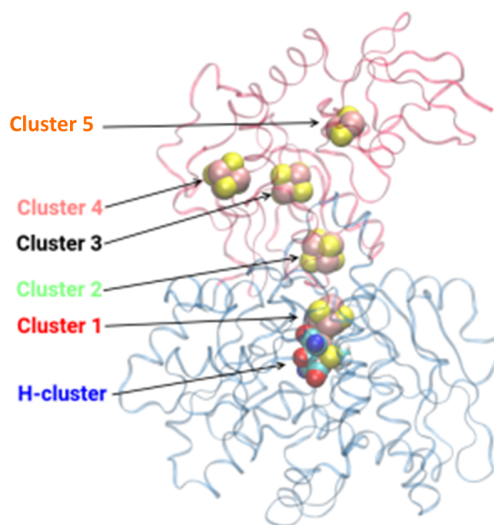


Figure 1.2: Locations of metal clusters in [FeFe]-hydrogenase are highlighted. The protein-backbone is shown as ribbons, and metallic clusters are shown as space-filling. The active-site domain of the protein is shown in blue ribbons, and the non-active-site domain is shown in red ribbons.

bound to an accessory metal cluster $[\text{Fe}_4\text{S}_4]$ [33, 34, 35]. The side of the protein that harbors the H-cluster (blue ribbons in Fig. 1.2), is designated as the active-site domain, and the side of the protein that harbors other metal clusters, is designated as the non-active-site domain (red ribbons in Fig. 1.2).

A major obstacle to achieving the goal of sustained H_2 production is the inactivation of this hydrogenase by oxygen [25, 26]. Although both NiFe and FeFe hydrogenases are sensitive to inactivation by O_2 , [NiFe]-hydrogenases capable of naturally resisting the inactivation by O_2 with different metallic clusters were recently found [36, 37]. Recently, success has also been achieved in improving the O_2 sensitivity of [NiFe]-hydrogenase [38, 39, 40], but no similar progress has been made on H_2 -producing [FeFe]-hydrogenase. It is thought that the inactivation of [FeFe]-hydrogenase by O_2 consists of diffusion of O_2 from the solvent into the enzyme interior, getting to the H-cluster buried in the core of the protein-matrix, and subsequently binding to the H-cluster [5, 41, 42].

Although the exact mechanism of inactivation of the H-cluster after binding of O_2 is not well understood, X-ray absorption measurements show that binding of O_2 to the distal Fe (Fe_d) in the H-cluster can subsequently damage the $[Fe_4S_4]$ accessory cluster next to the H-cluster [43, 44, 45, 46, 47]. It has been proposed that possible formation of reactive oxygen upon O_2 binding to Fe_d of the H-cluster and its diffusion toward the $[Fe_4S_4]$ accessory-cluster damages this cluster (cluster 1 in Fig. 1.2) [43, 48]. However, Swanson et al. [49] describe a rather complex mechanism for degradation of the H-cluster, including the formation of a reversible state before its full degradation. Aside from the H-cluster, Liebgott et al. [50] have shown that by altering the structure of [NiFe]-hydrogenase using a site-directed mutagenesis approach, the diffusion rate of O_2 can be slowed by orders of magnitude, thereby slowing the inactivation rate. This is accomplished by replacing the amino acids along known gas diffusion channels with bulky amino acids, likely blocking the possible O_2 diffusion path within the protein [50]. Similarly, Nienhaus et al. [51, 52] show that CO diffusion in myoglobin is altered by mutations of amino acids along CO-diffusion paths.

The presence of gas channels in [FeFe]-hydrogenase for O_2 and H_2 has been investigated both computationally and experimentally. The existence of two main gas channels in the active-site domain of the [FeFe]-hydrogenase has been proposed [5, 11, 53, 25]. Kubas et al. [10] investigated the rate of O_2 diffusion along these two main channels [10]. A major limitation of these investigations has been a narrow understanding of O_2 diffusion pathways within the other parts of [FeFe]-hydrogenase, and interactions of these diffusion channels with each other. Moreover, computational studies for gaining insights into ligand diffusion mechanisms, such as O_2 diffusion in [FeFe]-hydrogenase, is still an active area of research [54]. In fact, Bingham et al. [55] have reported that the mutations in an area away from the two main gas channels affect O_2 diffusion noticeably, indicating the potential presence of other gas channels in [FeFe]-hydrogenase or an allosteric communication between O_2 binding sites. Although, both Lautier et al.'s [6] and Ghirardi et al.'s [11] mutagenesis investigations suggest that O_2 is possibly using the two main channels, the likelihood of the presence of other pathways

for O₂ within the protein was highlighted [6]. Indeed, a network of interconnected diffusion pathways have been found for ligand diffusion in various proteins [56, 57, 58], clearly showing the presence of key residues or reaction sites along the predicted pathways, and elucidating the interconnected nature of diffusion pathways of ligands in proteins. Therefore, a better understanding of gas diffusion in [FeFe]-hydrogenase is needed to develop strategies for its functional modification. Thus, a detailed mapping of diffusion pathways of CO and O₂ in the FeFe-hydrogenase is not only needed for resolving these questions, but also for developing approaches to enhance the tolerance of the FeFe-hydrogenase to inhibitory gases.

1.4.2 Regulators of G-protein signaling (RGS)

GPCRs are membrane proteins of profound clinical relevance [59, 60, 61, 62, 63], as they mediate key roles in many cellular reactions to neurotransmitters and extracellular ligands [64]. Signaling by GPCRs is negatively modulated by a family of proteins known as RGS proteins that serve as a critical node in controlling various cellular responses [65, 66, 67, 68]. The mechanism of action of RGS proteins is to bind to activated (GTP-bound) G protein α -subunits ($G\alpha$) and accelerate the rate of GTP-hydrolysis, resulting in conversion of GTP to GDP and deactivation of $G\alpha$ -subunits. By this mechanism, RGS proteins rapidly dampen signaling by GPCRs, and therefore inhibitor candidates targeting RGS/ $G\alpha$ protein-protein interaction are potentially useful to enhance signaling by GPCRs [69]. Inhibiting protein-protein interactions, such as the one between an RGS protein and a $G\alpha$ subunit, is particularly challenging [70, 71, 72, 73], because of the lack of suitable binding pockets for small-molecules at the RGS/ $G\alpha$ protein-protein interface. Although progress has been made recently in the development of small molecules as covalent allosteric protein modifiers [74, 75, 76, 77], application of small-molecule ligands for the inhibition of RGS proteins and protein-protein interactions is an active area of research [78, 65, 79, 80, 81, 82, 70, 83, 84, 85, 4].

Specifically, high-throughput screens have revealed several classes of small molecule inhibitors acting through covalent modification of cysteine thiols on the RGS4 protein [85, 84, 86, 87], which is a well-characterized member of the R4 subfamily [66]. Among them, the thiadiazolidinone (TDZD) inhibitor CCG-50014 and its congeners (Fig. 1.3a) have shown high potency (measure of the concentrations of a drug at which it is effective) and specificity (the degree to which the effects of a drug are due to the one pharmacological action) [4]. CCG-203769 is one of the congeners of CCG-50014 with two aliphatic side-chains, but with substantially improved solubility [4]. RGS4 is highly expressed in the striatum [88, 89], regulating synaptic plasticity in response to dopamine signaling [90, 91], and has been associated with Parkinson’s disease [92]. The effectiveness of CCG-203769 on RGS4 has been demonstrated through in-vivo studies, by reducing bradykinesia in a raclopride model of certain Parkinson’s-like motor deficits in mice [93]. Furthermore, RGS4 has also been implicated in cancer [94, 95, 96], heart rate control in the sinoatrial node [97], suppression of the anticonvulsant action of adenosine [98], and opioid receptor signaling [99].

Structurally, RGS proteins are defined by a conserved ~ 120 residue-long box-domain with nine α -helices [100]. The crystal structure of RGS4 bound to $G\alpha_{i1}$ (PDB code 1AGR) is shown in Fig. 1.3b [101], highlighting the RGS domain with each helix distinctly colored. The crystal structure of RGS4 reveals the location of four cysteine residues (highlighted by orange spheres in Fig. 1.3b), three of which are solvent-exposed and one cysteine residue (labeled C95 in Fig. 1.3b) is buried. The C95-residue is also a conserved residue among many RGS proteins, thereby making it an attractive target for covalent modification. Although other cysteines in RGS4 play an important role in potent inhibition of RGS4 by covalent inhibitors [84], these residues are not conserved among all inhibited RGS proteins. Therefore, an understanding of the actions of TDZDs at C95 will be broadly applicable to other RGS proteins. Importantly, TDZD compounds can allosterically inhibit the RGS/ $G\alpha$ protein-protein interaction via the conserved and buried cysteine residue C95 [84, 85, 102]. Moreover, the activity of RGS4 is regulated allosterically by endogenous lipid modulators

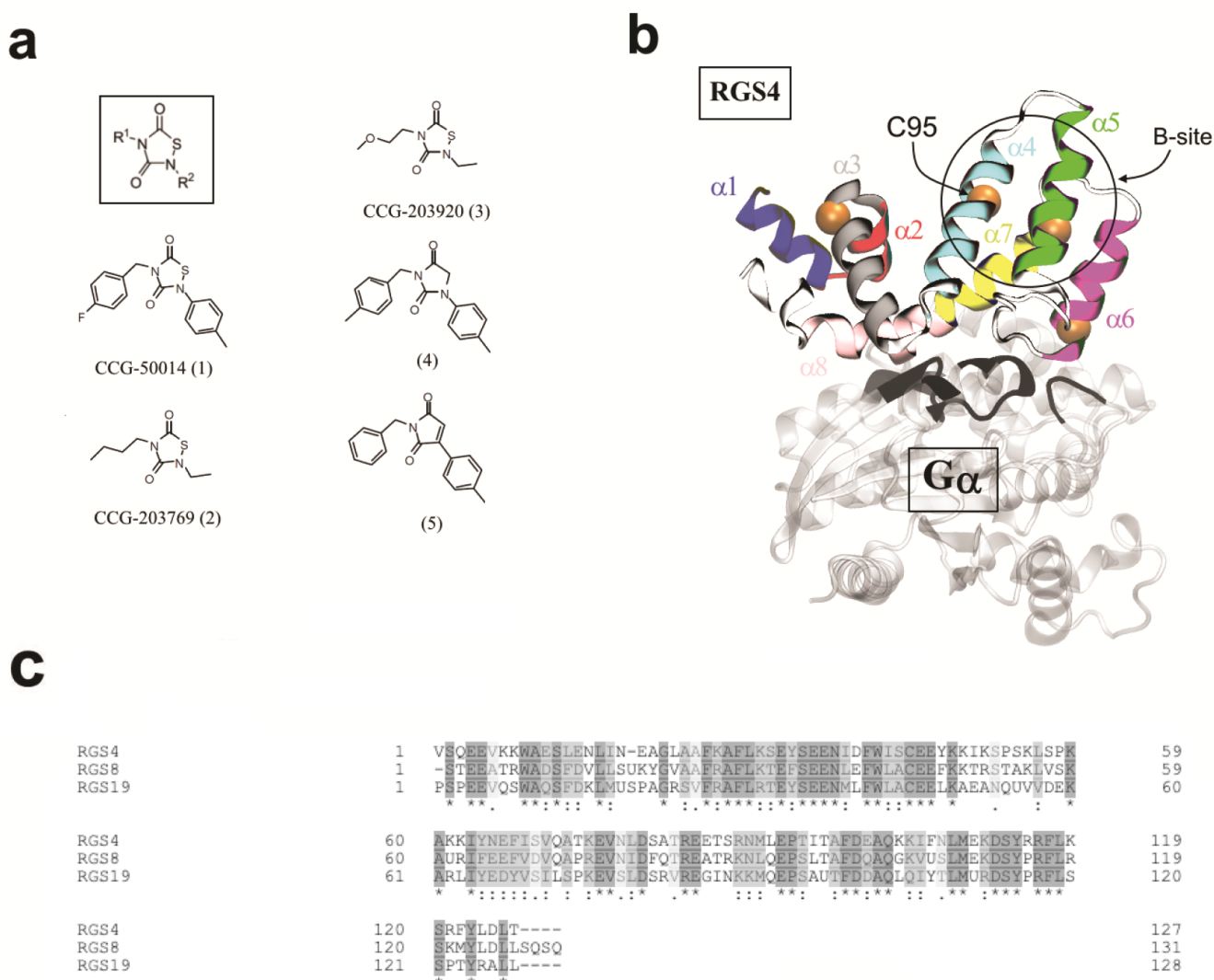


Figure 1.3. (a) Small-molecule structure with R¹ and R² functional groups (inset in box) along with the chemical structures of all small-molecules studied. [4] (b) A cartoon representation of the RGS4-Gα_{i1} complex (PDB code 1AGR) is shown. Each of the 9 α-helices of RGS4 is colored and labeled, and the location of four cysteine residues are shown by their C_α-atoms as orange spheres with the C95 residue labeled. The Gα_{i1}-subunit is shown in transparent white ribbons, and the loops of Gα_{i1} in the proximity of RGS4 are highlighted in black ribbons. (c) Sequence alignment of RGS4, RGS8, and RGS19.

at a site far from the $G\alpha$ interaction interface but near the C95 residue, also known as the “B-site” (the region marked by a circle in Fig. 1.3b) [65]. This allosteric site is also a promising drug target as many physiological regulators of RGS4 directly interact with the B-site [103, 104].

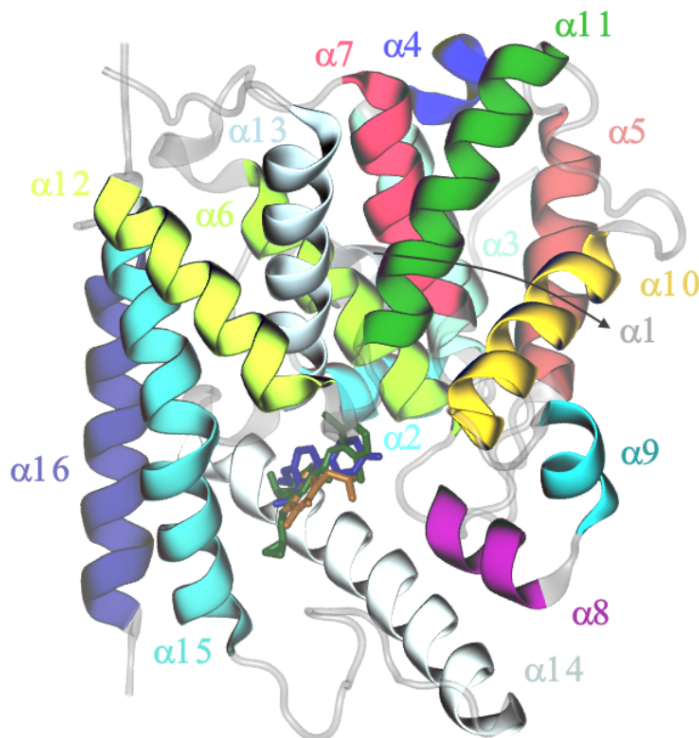


Figure 1.4. The Structure of Phosphodiesterase 4 (PDE4), and studied inhibitors (roflumilast, zardaverine, and IBMX are shown in green, yellow, and blue, respectively) overlaid in the catalytic domain of PDE4. The helices of PDE4 are labeled and colored.

1.4.3 Phosphodiesterase 4 (PDE4)

The efficacy of and resistance to anthelmintic/nematicidal compounds for controlling parasitic nematodes is a growing concern in the fields of medicine, veterinary medicine, and agriculture [105, 106]. I have studied the human and nematode PDE4 enzyme for targeting

nematode PDEs for the development of nematicides. The class I PDE superfamily in vertebrates consists of eleven PDE families that share a conserved signature motif in the catalytic domain [107]. The crystal structures of the catalytic domains of almost all the PDE families have been solved, providing atomic-level details on the enzymatic and pharmacological properties of enzymes in this superfamily [108].

The catalytic domain is made up of ~ 330 amino acids whose secondary structure consists of 16 α -helices (Fig. 1.4). These α -helices create three subdomains [109] which form a deep catalytic pocket at their center. The active site is composed of two sub-pockets, which bind two divalent metal ions and the substrate, respectively [108]. Zinc and magnesium ions are stabilized by conserved His and Asp residues in the metal binding pocket [108]. For example, the structure of the human PDE4 catalytic domain is stabilized by interactions with divalent cations and Asp and His residues of the metal binding pocket as well as with conserved Gln and Phe residues in the hydrophobic pocket. The invariant Gln residue of PDEs have been shown to be critical for substrate and inhibitor binding [108].

1.5 Thesis Outline

In Chapter 2, I introduce details of all computational methods and software packages used in my thesis work. These include the basics of molecular dynamics (MD) simulations, and other related computational techniques. Chapter 3 describes studies of inhibitory gases O_2 and CO in FeFe-hydrogenase. Chapter 4 contains the results of studies of allosteric inhibitors for the RGS4 protein. Chapter 5 contains the results of studies on coupling of protein dynamics and salt-bridging interactions among RGS proteins. Chapter 6 provides structural studies of inhibitors on human and nematode PDE4. Each chapter begins with a brief introduction describing the background information. The introduction of each chapter is followed by results and discussion. The usage of “we” in this thesis refers to work done in collaboration

with other researchers and their contributions have been acknowledged throughout the thesis. Appendices A through D contain samples of computer scripts and codes used in this thesis, appendix E through H provides supporting information for chapters 3 to 6, respectively, and appendix I highlights a cover image related to the work described in chapter 4. My brief curriculum vitae is included in Appendix J.

CHAPTER 2

METHODS

2.1 Introduction

In this chapter, I describe specific methods in detail including classical MD and enhanced sampling MD techniques. The software used are: VMD (Visual Molecular Dynamics) [110], NAMD (NAnoscale Molecular Dynamics) [111], MD-TASK [112], ICM [113], and CHARMMGUI (CHARMM is an acronym for Chemistry at HARvard Molecular Mechanics) [114]. The analysis scripts were implemented in MATLAB (stands for MATrix LABoratory), TCL (Tool Command Language), and Linux bash scripting. The computations were performed on local supercomputing resources at UNH (Premise and Trillian), and Comet (San Diego Supercomputer Center).

2.2 Molecular Dynamics Simulation

Statistical mechanics bridges macroscopic observations with microscopic details using statistical theories combined with classical physics concepts rooted in thermodynamics and Newton's laws [115, 116, 117]. In this context, an ensemble is the group of all microscopic states that correspond to one macroscopic state. In other words, macroscopic properties are statistical averages over multiple microstates of the system, also known as ensemble averages. In statistical mechanics, different kinds of ensembles exist which are characterized by fixed values of thermodynamic variables such as total number of particles N , volume V , temperature T , total energy E , or chemical potential μ [118]. In my thesis, all simulations

were carried out either in the NVT ensemble or in the NPT ensemble. These two ensembles are more commonly compared to experimental settings.

In my thesis work, I used MD simulations as a main tool to study complex interactions between biomolecules and small-molecule inhibitors. Given the expensive nature of quantum mechanical calculations, MD simulations provide a more practical way to study complex biomolecules while retaining classical molecular models based on quantum calculations [119, 120, 121].

MD simulations are rooted in numerical techniques for solving equations of motion for a large number of atoms. The basics of underlying equations of motion originate from classical physics. The source of a force calculation in classical physics is a potential energy function that plays a key role in describing the behaviour of a many-body system. The potential functions account for both bonded and non-bonded interactions. The typical form of the potential energy function is given by the following equation [122]:

$$\begin{aligned}
U = & \sum_{bonds} k_b(b - b_0)^2 + \sum_{angles} k_\theta(\theta - \theta_0)^2 + \sum_{dihedrals} k_\phi(1 + \cos(n\phi - \delta)) \\
& + \sum_{impropers} k_\omega(\omega - \omega_0)^2 + \sum_{Urey-Bradley} k_u(u - u_0)^2 \\
& + \sum_i \sum_{j>i} 4\epsilon_{ij} \left[\left(\frac{\sigma_{ij}}{r_{ij}} \right)^{12} - \left(\frac{\sigma_{ij}}{r_{ij}} \right)^6 \right] + \sum_i \sum_{j>i} \frac{q_i q_j}{4\pi\epsilon_0 r_{ij}}
\end{aligned} \tag{2.1}$$

The first term in the energy function accounts for bond stretches where k_b is the bond force constant and $(b - b_0)$ is the net displacement from equilibrium. The second term in the equation accounts for angles where k_θ is the angle force-constant and $(\theta - \theta_0)$ is the angular deviation from equilibrium for three bonded atoms. The third term is for dihedrals (also known as torsion angles) where k_ϕ is the dihedral force-constant, n is the multiplicity of the function, ϕ is the dihedral angle and δ is the phase shift. The fourth term accounts for impropers, that is out of plane bending, where k_ω is the force constant and $(\omega - \omega_0)$ is the out of plane angle deviation. The Urey-Bradley component (cross-term accounting for

angle bending using 1,3 nonbonded interactions) comprises the fifth term, where k_u is the respective force constant and $u - u_0$ is the distance between the 1,3 atoms in the harmonic potential.

Nonbonded interactions between pairs of atoms (i, j) are represented by the last two terms. The electrostatic interactions are represented by a Coulombic potential where q_i represents the charge of an atom, ϵ_0 is the permittivity of free space, and r_{ij} is the distance between the centers of mass of two atoms. The van der Waals forces are represented by a Lennard-Jones potential function. In this equation, ϵ_{ij} represents the energy well depth, σ_{ij} is the hard sphere radius [122]. There are no special terms to represent hydrogen bonding, as they are treated as a sum of van der Waals and electrostatic interactions. For a potential function used in an MD simulation, the above parameters are computed based on quantum-mechanical calculations and optimized against experimental data [111].

In this thesis, I have used the CHARMM force-field with the CMAP correction [122, 123] for all the MD simulations. CHARMM is the most common and a widely used force-field for proteins, peptides, lipids, and small molecule ligands [124]. All small-molecules used in this thesis and their force-fields were parameterized using the Multipurpose Atom-Typer for CHARMM (MATCH) tool [125].

2.2.1 Components of an MD Simulation

2.2.1.1 Langevin Equation

To improve the dynamical stability of the system in an MD simulation, the equation of motion is modified to a partially stochastic Langevin equation, which is given by:

$$M\dot{v} = F(r) - \gamma v + \sqrt{\frac{2\gamma k_B T}{M}} R(t) \tag{2.2}$$

$$F(r) = -\nabla U(r)$$

where M is the mass of a particle, $v = \dot{r}$ is the velocity, \dot{v} is the acceleration, F is the force, U is the potential energy, r is the position vector, γ is the friction coefficient, k_B is the Boltzmann constant, T is the temperature, and $R(t)$ is a univariate Gaussian random process. The typical values of friction coefficient are 5 ps^{-1} or 10 ps^{-1} . The second term is for dissipative forces, and the last term adds fluctuations for stochasticity which crudely mimics molecular collisions and viscosity in the realistic cellular environment.

2.2.1.2 Numerical Integration

For evolving the system in time, the equation of motion for each atom is numerically integrated. The complexity and stochastic nature of biological systems work against the convergence and stability of an integration algorithm.

The time-step ($\Delta t \sim 1-2 \text{ fs}$) also increases the cost of numerical integration significantly. The numerical integration algorithm also needs to maintain the accuracy of the system properties, e.g., temperature and pressure, while being efficient even with the existence of fluctuations in the system [111]. The Brünger-Brooks-Karplus (BBK) method [126], the extension of velocity-Verlet algorithm [121], is the common scheme used in MD simulation packages. The BBK method obtains the position and velocity of each particle through the following recurrence relation:

$$r_{n+1} = r_n + \frac{1 - \gamma\Delta t/2}{1 + \gamma\Delta t/2}(r_n - r_{n-1}) + \frac{1}{1 + \gamma\Delta t/2}\Delta t^2[M^{-1}F(r_n) + \sqrt{\frac{2\gamma k_B T}{\Delta M}}Z_n] \quad (2.3)$$

where r_n is the position vector, γ is the friction coefficient, Δt is the time-step, M is the mass of a particle, F is the force, k_B is the Boltzmann constant, T is the temperature, and Z_n is a group of Gaussian random variables with a mean of zero and a variance of one. In the BBK method, only one random variable is needed for each degree of freedom. This method has a global error proportional to Δt^2 [127].

2.2.1.3 Choice of integration time-step Δt

The time step is an important parameter in an MD simulation for numerical integration of differential equations for progress of the system. It is important because it determines the extent of accuracy and convergence in MD simulations. On one hand, larger values of Δt will lead to instabilities in the integration scheme, but on the other hand, smaller values of Δt will lead to limited exploration of the phase space and will increase the computational cost. The choice of time-step needs to be adjusted to satisfy the numerical stability of simulation, and convergence of the numerical integration scheme [128]. A suitable value for Δt in MD simulations of proteins for which the bonds to hydrogen atoms are fixed is 2 fs, and if the bonds to hydrogen atoms are not fixed, the suitable value for Δt is 1 fs. The implementation of a multiple-time-stepping algorithm could double the computational efficiency [111, 128]. Multiple-time-stepping algorithms compute slow-varying-forces such as long range electrostatics forces less frequently than fast-varying-forces such as Lennard-Jones and short range electrostatic [128]. The limitation of this algorithm is in conserving energy within a simulation system, which makes this scheme inapplicable to the NVE ensemble. In this thesis, I have not used the NVE ensemble and used only the NPT or NVT ensembles in MD simulations.

2.2.1.4 Initial conditions, minimization, temperature control, and pressure control

In an MD simulation, the initial coordinates of each atom in the system must be specified. As in a finite time-scale MD simulation, trajectories are highly dependent on initial configurations. As expected, the initial configuration of proteins and other biomolecules should be near a stable state of the system. The main repository for the initial configurations of different proteins used in this thesis was the Protein Data Bank (PDB)¹, which is

¹<http://www.rcsb.org/pdb/home/home.do>

the data repository for 3D structures of biomolecules determined by X-ray crystallography, Nuclear Magnetic Resonance (NMR), and other approaches. In addition to initial coordinates of the system, for solving the differential equations of motion, the initial values of all velocities should also be assigned. In an MD simulation, initial velocities to each atom from a Maxwell-Boltzmann distribution given by:

$$P(v) = \sqrt{\left(\frac{m}{2\pi k_B T}\right)^3} 4\pi v^2 e^{-\frac{mv^2}{2k_B T}} \quad (2.4)$$

where v is the velocity, m is the particle mass and $k_B T$ is the product of Boltzmann's constant and the temperature. Based on the ergodic hypothesis in MD runs.

In MD simulations, boundary conditions of physical systems could be defined in three possible ways: (i) vacuum, (ii) a reflecting wall, and (iii) periodic boundary conditions (PBC). The simplest case for BC is vacuum, but the global properties computed will not be representative of the condensed phase [129]. In the case of reflecting wall BC, for keeping the atomic particles inside the simulation box, solid boundaries with a potential are applied [130]. In PBC, a common practice in MD simulations and which is used in this thesis, particles are isolated in a simulation box that is infinitely replicated by periodic translations in all dimensions [129]. In two-dimensions, each unit cell has eight nearest neighbors, while in three-dimensions each unit cell would have 26 nearest neighbors. An atom or a particle can leave the simulation box from one side, and consequently, it is replaced by an image of the particle which enters from the opposite side of the unit cell, in doing so conserving the number of particles in a simulation box, which is requirement of the NVT and NPT ensembles used in this thesis.

Prior to an MD simulation, the initial configuration of the system obtained from PDB may need preparation to add often missing hydrogen atoms or missing atoms in residues. In most cases, the preparation of the system could include incorporating mutations, solvation,

and ionization of the system. Upon preparation of the desired system for an MD simulation, by using steepest descent or conjugate gradient schemes [131] the system arrangement in space energy-minimized. In this thesis, for all pre-simulation systems, I used the conjugate gradient method for minimization.

The temperature control for an MD simulation can be applied by a Langevin thermostat where additional damping and random forces are introduced in the system. In this method, the temperature of the system is maintained at the chosen value by frequently adjusting the momenta of all atoms in the system. Other techniques used in MD schemes are the Andersen thermostat [132] and the Berendsen thermostat. For pressure control, the Nosé-Hoover barostat algorithms were used [133, 134, 135].

2.3 Software Packages for Biomolecular Simulations

2.3.1 MD Simulation

In my thesis, I have used NAMD (versions 2.11-2.12) as the primary software tool to carry out MD simulations. NAMD is a parallel MD simulation code designed for large-scale simulations of biomolecular systems [111]. NAMD is capable of effectively using hundreds of processors on computational nodes in supercomputing clusters, and can be further deployed on GPU architectures with a significant reduction in computation time. MD simulations of very large systems of up to ~ 64 million atoms are possible with NAMD [136]. The NAMD package contains several efficient techniques to accurately control temperature and pressure, and numerically integrating Newton's equations of motion. NAMD is an open source code package that is available² to academic researchers, and its features can be modified by Tcl/Tk scripting (Tk is the standard GUI for Tcl).

²<http://www.ks.uiuc.edu/Research/namd/>

2.3.2 Modeling and Analysis

In this thesis, I primarily used NAMD’s companion software VMD (version 1.9) [110] for visualization and analysis. VMD was used for preparing input files for NAMD, and for extracting different properties from MD trajectories. VMD is a versatile software that can run from a simple laptop up to GPU clusters, and it is able to perform parallel computations for computationally intensive tasks.

In this thesis, I used MATLAB (versions 2016-2018) for various calculations, optimizations, and creating data plots. The numerous built-in functions, various toolboxes, and libraries make MATLAB a unique tool for computational science researchers. I have provided all MATLAB codes in this thesis in Appendix A.

I used the MD-TASK package [112] for calculating dynamic cross correlation (DCC) maps. I calculated the DCC maps of each system based on the C_α atoms of residues using the MD-TASK package³ [112]. These maps are reported in Fig. 5.4 (chapter 5), and Fig. 6.5 and Fig. 6.6 (chapter 6). Each cell value (C_{ij}) in the matrix of the DCC map was calculated using the following formula:

$$C_{ij} = \frac{\langle \Delta r_i \Delta r_j \rangle}{\sqrt{\langle \Delta r_i^2 \rangle} \sqrt{\langle \Delta r_j^2 \rangle}} \quad (2.5)$$

Where Δr_i represents the displacement from the mean position of atom i , and $\langle \rangle$ denotes the time average over the whole trajectory. The positive values of C_{ij} show correlated motion between residues i and j , moving in the same direction, whereas negative values of C_{ij} show anti-correlated motion between residues i and j , moving in the opposite direction.

³<https://md-task.readthedocs.io/en/latest/home.html>

I also used Linux-based operating systems (openSUSE and Ubuntu) for working with all local software packages, managing and deploying simulations on clusters, shell scripting for analyzing data, and scripting. In Linux, bash shell is the terminal language for various command line operations from simple tasks to analyzing data and for parallel calculations. I have provided all useful bash codes in Appendix B and C. I used MobaXterm (version 9.4) and Cygwin64 software for remotely accessed Linux-based workstations and supercomputing clusters.

I used CHARMM-GUI⁴ and the MATCH web server⁵ to produce both the topology and parameter files for small molecules. An important utility program termed catdcd (version 4.0) which is a built-in feature in VMD, was used for concatenating trajectory files and selecting the desired segments of trajectories (Appendix D).

2.4 Advanced Sampling Methods

In a conventional MD simulation, due to a large number of degrees of freedom, the explored conformational space of biomolecules is limited [137]. Therefore, several enhanced sampling methods have been developed to overcome the limitations of conventional MD simulations. Enhanced sampling methods such as temperature accelerated MD (TAMD) and metadynamics are methods to accelerate sampling and overcome free-energy barriers for enhancing conformational exploration [111, 138, 139, 137]. In this thesis, I used metadynamics and TAMD for enhanced sampling MD simulations in conjunction with a suite of methods, called the single-sweep, for the free-energy reconstruction, and for finding the interconnected network of migration pathways of ligands within proteins [140, 141, 137]. The details of these

⁴<http://www.charmm-gui.org/>

⁵<http://brooks.chem.lsa.umich.edu/index.php?matchserver=submit>

methods are described in the following.

2.4.1 TAMD Algorithm

TAMD is a method of enhanced sampling to explore the physical free-energy landscape [140], and has been successfully applied to a number of systems [142, 143, 144, 15]. TAMD algorithm begins by defining an extended version of the physical system in which new variables, z , are coupled to the physical variables via collective variables (CV), $\theta(x)$ [145, 140]. The new total potential (U) contains, in addition to the physical energy ($V(x)$), a term that couples CVs ($\theta(x)$) to new z -variables, thus

$$U_\kappa(x, z) = V(x) + \frac{\kappa}{2} \sum_{j=1}^m (\theta_j(x) - z_j)^2 \quad (2.6)$$

Here, m is the number of CVs, κ is a spring constant and is an adjustable parameter. The new variables z are assigned a fictitious mass (M) and a fictitious temperature (\bar{T}) which is different from that of the physical system. The evolution of the extended system is governed by the following dynamical equations:

$$M\ddot{x} = -\nabla_x V(x) - \kappa \sum_{j=1}^m (\theta_j(x) - z_j) \nabla_x \theta_j(x) + \text{thermostat at } \beta^{-1} \quad (2.7)$$

$$\bar{\gamma}\dot{z} = \kappa(\theta(x) - z) + \sqrt{2\bar{\beta}^{-1}\bar{\gamma}}\eta^z \quad (2.8)$$

where $\bar{\gamma}$ is the artificial friction coefficient, $\bar{\beta}$ is the inverse of the artificial temperature ($\bar{\beta} = 1/(k_B\bar{T})$, where k_B is the Boltzmann's constant), β is the inverse of the physical temperature ($\beta = 1/(k_B T)$), and η^z is the white-noise associated with the Langevin evolution of the new variables.

2.4.1.1 Free-energy Reconstruction

The exploration of the protein interior space via TAMD results in a dense cloud of ligand positions (Fig. 2.1). In this thesis, I have used the single-sweep method [141] for

reconstructing the free-energy landscape in CVs that are TAMMD-explored positions of the center-of-mass of O_2 and CO . This method has been successfully used as an efficient way of computing multidimensional free-energies [141]. Out of a large number of explored positions, it is necessary to choose a unique set of positions, namely centers, covering various areas of the protein. For estimating the mean-force required to keep the ligand at those positions, these centers are chosen along the trajectory based on a distance criterion: beginning with the value z_1 as the first center, a new point is added to the set when its distance from all other members of the set exceeds a prescribed cut-off distance. In the end, a large number of centers (~ 300 - 700) are harvested (Fig. 2.1a). Each chosen location is referred to as a “center”, and the k^{th} center is indexed as z_k .

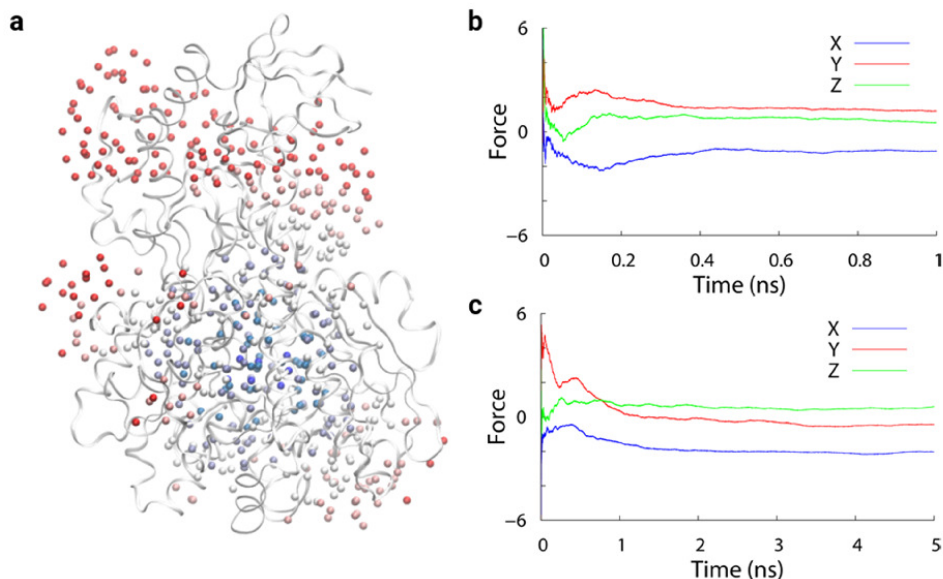


Figure 2.1: (a) The locations of 497 unique centers (the positions of the center-of-mass of O_2) chosen after TAMMD sampling are highlighted. The protein-backbone is shown as white ribbons and the extracted centers are shown as spheres. The centers are colored from blue to red, based on their distance to the H-cluster. (b) The running-average of mean forces (kcal/mol.Å) for three Cartesian directions for one representative center (blue, X-direction; red, Y-direction; green, Z-direction) are shown from a 1 ns (panel b) and 5 ns (panel c) mean-force calculation.

The mean-force at each of the obtained k th center, $f(z_k)$, is calculated using the following time average:

$$f(z_k) = \frac{1}{t} \sum_{j=1}^t \kappa(\theta(x(t_j)) - z_k) \quad (2.9)$$

Here, κ is the spring constant, z_k is the position vector of the k th center, t is the number of time increments in the trajectory and θ is the center-of-mass of the single molecule as a CV [58, 146]. The mean-force for each center was computed by using an MD simulation with CVs restrained at the center using a coupling constant of 10 kcal/mol·Å². The C_α atoms of 3 residues were restrained using a spring constant of 2.0 kcal/mol·Å² to orientally restrain the protein during each force calculation, as has been employed in other applications [57, 58]. The solvent centers around protein were chosen and (Fig. 2.1a) for which mean-forces were nearly negligible. The simulation for all centers are independent of each other, and therefore are independently distributed among different computing nodes [56]. Each mean-force calculation lasts for 1 ns, after which the convergence of mean-force values was observed (Fig. 2.1b). For some selected centers that needed more than 1 ns for force convergence, calculations were continued up to 5 ns (Fig. 2.1c). The histogram of standard deviation computed over the last 10% of each force calculation represent the convergence of mean-forces by delta function at origin (Fig. E.4).

Given the harvested centers and their computed respective mean-forces, the global reconstruction of the free-energy surface (FES), $A(z)$, was carried out as a linear combination of Gaussian radial basis functions [141]. Therefore, the analytical free-energy can be written as:

$$A(z) = \sum_{k=1}^K a_k \varphi_{\sigma}(|z - z_k|) + C \quad (2.10)$$

Here φ_σ is a Gaussian with a width of σ , K is the total number of centers, a_k 's are coefficients in expansion, and C is a constant for adjusting the height of $A(z)$. The optimization method based on the least-square fitting was used for obtaining the optimal values of σ and a_k [147, 141, 148]. Briefly, optimal values of σ and a_k are obtained by minimizing the following error function (E):

$$E(a, \sigma) = \sum_{k=1}^K \left| \sum_{k'=1}^K a_{k'} \nabla_z \varphi_\sigma(|z_k - z_{k'}|) + f(z_k) \right|^2 \quad (2.11)$$

2.4.1.2 Minimum free-energy pathways (MFEP)

Following the reconstruction of the free-energy surface, the minimum free-energy pathways (MFEP), as pathways of the single molecule diffusion on the FES, can be determined using the zero-temperature string method (ZTS) [149]. These MFEPs are identified as the diffusion pathways of ligands inside the protein [150]. The diffusion pathways are the network of MFEPs that interconnect local minima and the solvent portals. Given an analytical approximation of the free-energy landscape, obtained from the PMF reconstruction step, the ZTS method gives MFEPs as the steepest-descent paths from saddle points on the reconstructed FES [151, 149]. This optimization method starts from an initial guess for a curve on the FES, and in a two-step iterative algorithm finds the optimal pathways; in the first step, the points that were obtained by discretizing the line were allowed to move based on the steepest descent on the FES landscape, while the second reparametrization step kept the distances between the points uniform [151, 152]. MFEPs are the most probable pathways that one can take on the FES between two local minima [151, 152]. Since the three-dimensional FES has been reconstructed and it is available analytically, the ZTS method was employed without requiring further MD simulations [56].

2.4.2 Metadynamics Simulations

Metadynamics is an enhanced sampling method for faster exploration of conformational space in a specified set of collective variables (CVs) by augmenting the force-field with a history-dependent biasing potential (V_{meta}) of the following form [153, 154]:

$$V_{\text{meta}}(\xi) = \sum_{t'=\tau_G, 2\tau_G, \dots}^{t'<t} W \prod_{i=1}^{N_{\text{cv}}} \exp\left(-\frac{[\xi_i - \xi_i(t')]^2}{2\delta_{\xi_i}^2}\right) \quad (2.12)$$

where ξ_i is the current value of the CV and $\xi_i(t')$ is the value of the CV at time t' . V_{meta} is constructed as a sum of N_{cv} -dimensional repulsive Gaussian functions with a chosen height (W) and width (δ). The Gaussian functions can be added at a desired frequency τ_G . The three main parameters in metadynamics that control the efficiency and accuracy of the free energy reconstruction from converged metadynamics potential (V_{meta}) [139] are W , δ , and τ_G . The metadynamics method has been successfully applied to study many biophysical problems including computational drug design [155, 138, 156, 157, 158, 159, 160].

2.5 Procedures for MD Simulations

Three steps for carrying out MD simulations in NAMD are (i) solvation and ionization of the initial configuration, (ii) simulating the system using the MD software, and (iii) analyzing output files. These steps are described below.

All biomolecular systems are found in an aqueous environment with ions, and hence the prepared systems of biomolecules need to be solvated with water and neutralized by adding ions to mimic the physiological environment. The number of added water molecules need to be adjusted, as more water molecules are added the size of the system will increase, thereby increasing the computational cost. The solvate plugin in VMD provides the tool for solvation of the system prior to an MD simulation. The autoionize plugin in VMD provides a tool for adding ions to the solvated system. In this thesis, MD simulations of all systems were carried out using NAMD. The generated trajectories were visualized and analyzed by the

VMD software. In this thesis, I also used MATLAB and python based MD-TASK package for futher analyzing trajectory data.

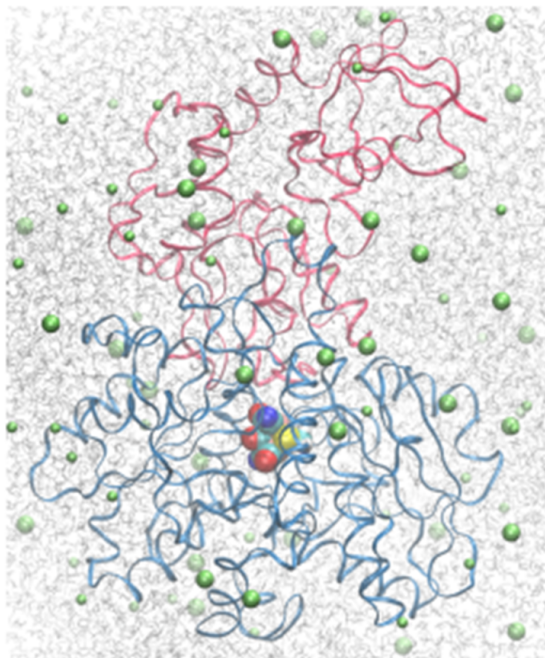


Figure 2.2: A snapshot of an MD simulation domain is shown. The H-cluster of [FeFe]-hydrogenase and ions (green) are shown as space-filling, and water molecules are shown as gray wireframe.

2.6 Computational Resources

The computational resources used in this thesis are as the following: (i) Lennard is a local linux workstation assigned to me in our laboratory at the University of New Hampshire equipped with 20 processors (model name : Intel(R) Xeon(R) CPU E5-2630 v4 @ 2.20GHz), 2 GB RAM, and a 2 TB internal hard-drive; (ii) the Trillian cluster, which was acquired through an NSF Major Research Instrumentation (MRI) grant, with cost sharing from the UNH Space Science Center ,Senior Vice Provost for Research, the College of Engineering and Physical Sciences (CEPS), and the Research Computing Center (RCC). This supercomputer

has 132 compute nodes, each with 2 AMD 16 core 'Abu Dhabi' 2.4 GHz CPUs, for a total of 4,224 cores; (iii) the Premise cluster which is made up of 37 compute nodes connected together using 56 Gb FDR Infiniband networking, each node has two 12-core CPUs, all nodes have at least 128GB of main memory. Ten nodes have NVIDIA K80 GPUs which I used mainly in MD simulations because of the higher performance compared to non-GPU compute nodes. The entire cluster shares 225TB of usable Lustre storage; and (iv) the Extreme Science and Engineering Discovery Environment (XSEDE) which is a powerful collection of integrated digital resources and services such as supercomputers, visualization and storage systems, collections of data, software, networks, and expert support that scientists, engineers around the world use to advance solving complex problems.

Comet is a dedicated XSEDE cluster, which was designed by Dell using Intel's Xeon Processor E5-2600 v3 family, with two processors per node and 12 cores per processor running at 2.5GHz. Each compute node has 128 GB (gigabytes) of traditional DRAM and 320 GB of local flash memory. Since Comet is designed to optimize capacity for modest-scale jobs, each rack of 72 nodes (1,728 cores) has a full bisection InfiniBand FDR interconnect from Mellanox, with a 4:1 over-subscription across the racks. There are 27 racks of these compute nodes, totaling 1,944 nodes or 46,656 cores. In addition, Comet has four large-memory nodes, each with four 16-core sockets and 1.5 TB of memory, as well as 36 GPU nodes, each with four NVIDIA GPUs (graphic processing units). The GPUs and large-memory nodes are for specific applications such as visualization, and MD simulations.

CHAPTER 3

STUDIES ON DIFFUSION PATHWAYS OF INHIBITORY GASES IN [FeFe]-HYDROGENASE

3.1 Abstract

The H₂ production potential of [FeFe]-hydrogenase, a hydrogen-producing enzyme from green algae, is reported to be promising for economical and large-scale production of H₂ as an alternative source of renewable energy. The production of hydrogen takes place at the catalytic center buried in the enzyme core. Unfortunately, binding of O₂ and CO to the catalytic center of the enzyme irreversibly inactivates it, essentially blocking hydrogen production. Therefore, a better understanding of the mechanism of O₂ and CO entry/exit is necessary to develop strategies for designing oxygen-tolerant enzymes. In this work, I have investigated the pathways and diffusion channels of O₂ and CO gases in [FeFe]-hydrogenase. Through exhaustive mapping of O₂ and CO diffusion channels, I computed a full thermodynamic map of preferred binding locations of O₂ and CO gases within the enzyme interior, which showed that O₂ and CO can enter and exit the enzyme through multiple pathways along which are key residues that are known to perturb binding rates of O₂ and CO binding. The global minimum for each gas in the free-energy landscape is located near the H-cluster, a key metallic center within the enzyme. Along O₂ and CO diffusion channels, I further identified several residues that could be potential candidates for mutations to increase the tolerance of [FeFe]-hydrogenase to both inhibitory gases.

3.2 Background

Enzymes use their active sites to exquisitely perform essential processes in living organisms by catalyzing chemical transformations. Given that the active sites are buried within the protein matrices, ligand migration between the solvent and the active site is an indispensable process for chemical reactions to occur. This observation hints at the existence of ligand entry/exit channels through which the activity of enzymes can be modulated by altering ligand diffusion (e.g. limiting the diffusion of inhibitory gases) [161, 162]. Although permanent diffusion channels may not exist in flexible structures of enzymes, transient thermal fluctuations throughout the protein matrix can potentially lead to the formation of a network of metastable pockets that can serve as conduits for site-to-site hopping of ligands.

Described in this chapter are studies on the thermodynamics of ligand diffusion pathways in the FeFe-hydrogenase, an enzyme of the hydrogenase superfamily, that is reported to be a key source of biological hydrogen production due to its ability to catalyze proton reduction and/or H₂-oxidation under optimal conditions [163, 164, 165, 166, 167, 168, 169]. The crystal structure of the CpI FeFe-hydrogenase from *Clostridium pasteurianum* revealed a deeply-buried active site (the H-cluster) with three 4Fe4S and one 2Fe2S accessory metal-clusters [170, 171]. The H-cluster is comprised of one Fe₄S₄ subcluster and one 2Fe subcluster with a vacant site on Fe_d (distal) which can bind different ligands [167] including inhibitory gases CO and O₂ that can severely hamper the enzymatic activity of the FeFe-hydrogenase. Specifically, it has been suggested that O₂ attacks and irreversibly deactivates the H-cluster, while CO can compete with O₂ for binding to the H-cluster and prevent inhibition by O₂ when the enzyme is oxidizing H₂ or is at the resting state (no reaction) [172, 173, 174]. Regardless, CO can still irreversibly deactivate the enzyme during H₂-production [174]. Importantly, both CO and O₂ attack the same site, the Fe_d of the 2Fe subcluster of the H-cluster [175, 176, 177]. Moreover, with CO binding to the active site, O₂ cannot access the active site, and previous studies have suggested a faster binding rate for CO than O₂ [178, 175].

Since accessing the active site requires channels connecting to the solvent, several possible scenarios can be hypothesized to explain the phenomenon of the protective effect and inhibition by CO: (a) CO diffuses faster than O₂ if both gases compete for the same diffusion pathway; (b) CO migrates along shorter diffusion pathways if each gas has an independent pathway; and/or (c) CO has a higher binding-affinity for Fe_d than O₂. Therefore, a detailed mapping of diffusion pathways of CO and O₂ in the FeFe-hydrogenase is not only needed for resolving these questions, but also for developing approaches to enhance the tolerance of the FeFe-hydrogenase to inhibitory gases.

Furthermore, to improve the tolerance of hydrogenase enzymes to inhibitory gases, mutations in specific amino-acid residues have been carried out. For example, the rates of CO and H₂ diffusion in another enzyme of the hydrogenase family, *D. fructosovorans* NiFe-hydrogenase, can be decreased after mutations of two residues (V74 and L122) [175, 178]. However, it is highly challenging to rationally choose residues for mutations among a large number of residues (e.g. 574 residues in the CpI FeFe-hydrogenase), especially without the knowledge of diffusion channels and the location of the free-energy barriers along them. This is evident in an experimental study on mutations of seven residues in the CpI FeFe-hydrogenase (C299, F417, V423, A427, A431, I461, and F493), where none of the residues showed any strong effect on the kinetic performance of this hydrogenase although some of these residues (A427, A431, I461, F493) are located in the pathway A [176]. This suggests the existence of additional pathways for gas diffusion in the FeFe-hydrogenase. Indeed, my previous study [1] disclosed an exhaustive mapping of the diffusion network of O₂ in the CpI FeFe-hydrogenase showing several new pathways to access the active site besides those previously suggested. We also suggested several new candidate residues lining diffusion pathways of O₂ whose mutations have the potential to significantly disrupt the O₂ diffusion network, as these residues are located in the vicinity of the free-energy barriers. However, no studies on the exhaustive mapping of the diffusion pathways of the competing inhibitor CO in the FeFe-hydrogenase have been carried out so far although pathways of CO diffusion in structurally-unrelated

NiFe-hydrogenase have been studied [179].

Hypothesis: An interconnected network of migration pathways exists for inhibitory gases throughout CpI [FeFe]-hydrogenase.

Therefore, I have studied the pathways and diffusion channels of O₂ and CO in CpI [FeFe]-hydrogenase by employing a judicious combination of enhanced sampling and free-energy reconstruction methods [56]. By finding the three dimensional free-energy of O₂ and CO binding within the enzyme, and interconnected network of diffusion pathways of both inhibitory gases, unique insights into O₂ and CO diffusion in [FeFe]-hydrogenase have been obtained. The results show that, the global minimum in the free-energy landscape is in the proximity of the H-cluster binding site (location where O₂ bind to the H-cluster), and another local minimum is in the proximity of the Xe-binding site (an experimentally found location for xenon presence [180]). The work shows the presence of numerous interconnected pathways of O₂ and CO diffusion toward the H-cluster beyond the two main channels, providing enhanced information on other areas of protein affecting O₂ diffusion. Moreover, along each pathway, energy barriers and local minima in the free-energy surface can be readily found. The important residues along the obtained pathways are prime candidates for guiding and designing [FeFe]-hydrogenase tolerant to both inhibitory gases.

3.3 Methods

3.3.1 System Setup

I carried out all MD simulations (section 2.2) using the CHARMM force-field for proteins, ions, water, and metal clusters [122, 181, 182]. I prepared a fully-solvated and ionized simulation domain using the initial coordinates for all protein atoms from the crystal structure of the FeFe-hydrogenase (PDB code 1FEH), solvating with explicit water (TIP3P) molecules,

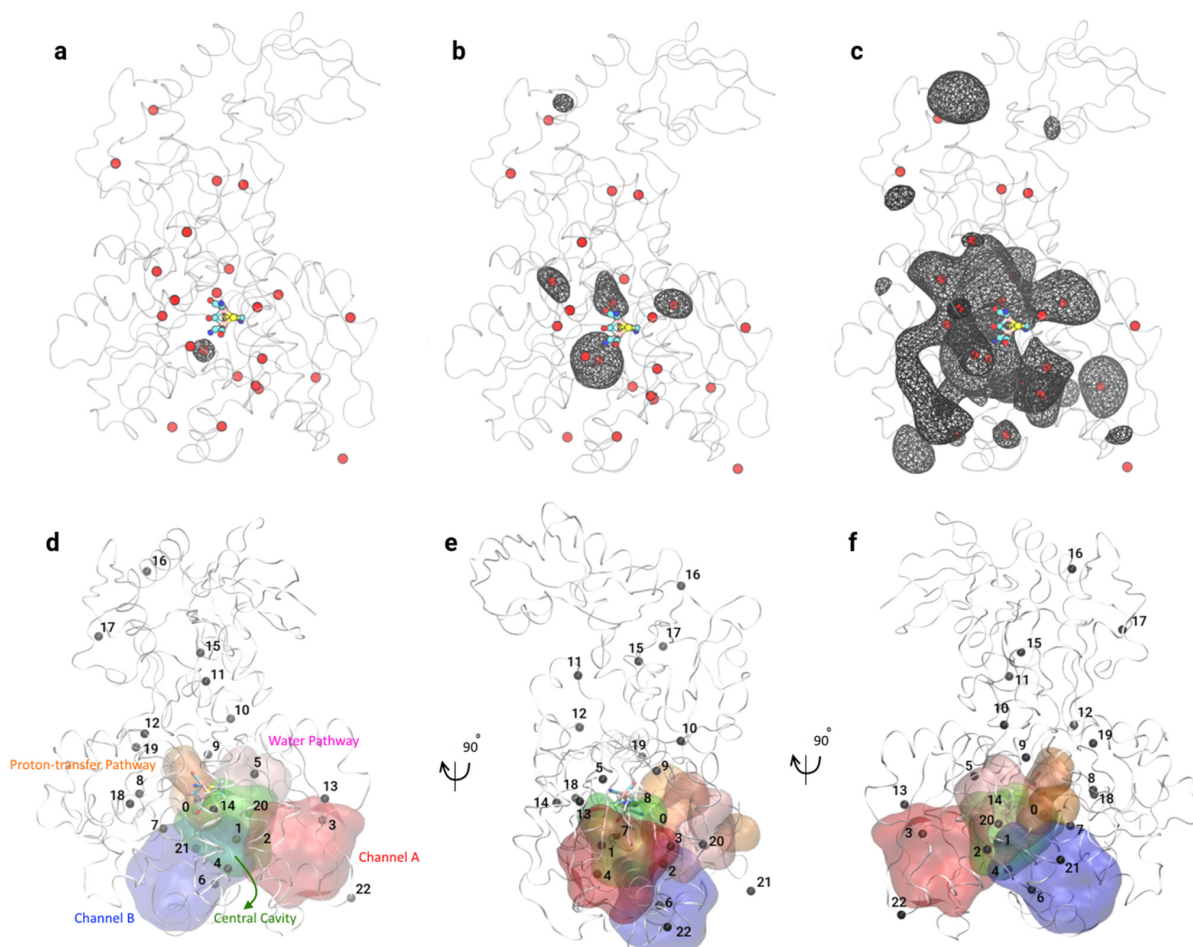


Figure 3.1: Isosurfaces of the 3D-PMF map of O_2 inside [FeFe]-hydrogenase are shown in top panels. Energy levels are represented at (a) 3 kcal/mol, (b) 12 kcal/mol, and (c) 17 kcal/mol with respect to the global free-energy minimum in proximity of the H-cluster. The local minima are represented by red spheres. The backbone of protein is shown as ribbons, the isosurfaces are shown as black mesh, and the H-cluster is shown as sticks. In bottom panels, the proximal area of the central cavity (green surface), channel A (red surface) and B (blue surface) [5], water pathway (pink surface) [6], and the proton transport pathway (orange surface) [7] are shown in surface representations as a front-view (d), a side-view (e), and a back-view (f). The local minima are represented by black spheres and labeled. The protein backbone is rendered as ribbons, and the H-cluster is shown as sticks.

and ionizing with NaCl. After an initial energy minimization of this system for 1000 steps, the box-volume was equilibrated by conducting a 1-ns long molecular dynamics (MD) simulation in the NPT ensemble, where the temperature was controlled at 310 K using the

Langevin thermostat and the pressure was controlled using the Nosé-Hoover barostat. An integration time-step of 1 fs, periodic boundary conditions, the particle-mesh Ewald summation (with a grid spacing of 1 Å) for long-range electrostatics, and a cutoff distance of 10 Å for van der Waals interactions was used. For exhaustive exploration of gas diffusion using enhanced sampling methods, I created several initial systems, each differing only in the initial position of gas molecule, which was chosen to span various locations throughout the enzyme matrix. Each of these systems was MD equilibrated for 1 ns in the NPT ensemble using the protocol described above. The coordinates from the end of the MD trajectories of these equilibrated systems were used as initial conditions for TAMD simulations (*vide infra*).

3.3.2 TAMD Simulations

I applied TAMD (section 2.4.1) for exhaustively exploring the gas-accessible interior volume of the FeFe-hydrogenase by using the Cartesian coordinates of the center-of-mass of each gas as the three-dimensional CV-space. In TAMD simulations, $\bar{\beta}^{-1}$ values ranging between 2 kcal/mol and 7 kcal/mol, a spring constant κ of 200 kcal/mol·Å², and a fictitious friction coefficient $\bar{\gamma}$ of 50 ps⁻¹ were used. Starting from each of the initial locations of each gas, 30 independent 1-ns long TAMD simulations were performed resulting in a total of $\sim 300 - 600$ ns of simulation time for exhaustive sampling of each gas diffusion.

3.3.3 Free-energy Surface Reconstruction and Minimum Free Energy Pathways

To reconstruct the 3D-FES of O₂ and CO diffusion in the FeFe-hydrogenase, we used the single-sweep method [141, 56] in which mean-forces are computed at CV positions sampled by TAMD (section 2.4.1). Specifically, it is necessary to choose distinct CV positions, also termed unique centers, throughout the protein matrix. Due to the rotation and translation of the protein, all TAMD trajectories were aligned to a common reference structure for extracting centers. Out of all gas positions sampled by TAMD, we chose several unique centers

(Fig. 2.1a). The centers were chosen by beginning with the first center (z_1) and adding new centers such that the distance of each new center from all previous centers exceeds a prescribed cutoff distance.

To compute the mean-force at each of the k centers, f_k , we launched a set of new MD simulations in which the center-of-mass of O_2 and CO is harmonically restrained about the CV-value at the chosen center. We conducted restrained MD simulations by using a spring constant κ of 10 kcal/mol·Å². Each of these simulations generally lasted for 1 ns within which we observed the convergence of the mean-force (Fig. 2.1b), but some centers required simulations up to 5-ns long for the convergence of the mean-force (Fig. 2.1c). We show the convergence of all mean-force calculations by plotting the histogram of the standard deviation of each mean-force calculation computed over the last 10% of each restrained MD trajectory. This histogram reveals the behavior of a delta function centered at the origin indicating convergence (Fig. E.4).

Using converged mean-forces at all centers, we globally reconstructed the free-energy surface $A(z)$ (section 2.4.1.1) as a linear combination of Gaussian radial basis functions by obtaining the optimized parameters (coefficients, a_k ; and the Gaussian width, σ) via a least square fitting procedure per equation [141]. We further studied the 3D-FES using the zero-temperature string (ZTS) method [183, 184, 185, 186] (section 2.4.1.2) and found MFEPs between all pairs of minima as the steepest descent paths from saddle points on the analytically reconstructed free-energy hypersurface, $A(z)$. Therefore, MFEPs are the most probable pathways between pairs of minima on $A(z)$ and are identified as the diffusion pathways for O_2 and CO inside the enzyme. We also note that the application of the ZTS method does not require any additional MD simulations since the analytical functional form of the 3D-FES is available from equation.

3.4 Results

3.4.1 Mean Force Estimation and PMF Map Reconstruction for O₂

The centers were selected as described in Chapter 2 (Section 2.4, Fig 2.1a). The converged mean-force values for 497 centers (chapter 2) were used for the reconstruction of the free-energy landscape of O₂ diffusion (inside the [FeFe]-hydrogenase), as a linear combination of radial basis functions using the single-sweep method [56]. In Figures 3.1a to 3.1c, I show isosurfaces of the 3D-PMFs, obtained from single-sweep calculations for O₂, at three energy levels of 3, 12, and 17 kcal/mol with respect to the global energy minimum in the free-energy landscape. The obtained 3D-PMF of O₂ binding sites on scale of the entire protein elucidates the detailed physical insight of O₂ transport in this protein. The global minimum in the PMF is located in the vicinity of the active-site pocket of the [FeFe]-hydrogenase (local minimum 0 in Figures 3.1d to 3.1f), and I found a free-energy difference of ~ 18.5 kcal/mol between the global minimum and the solvent space.

In addition to global minimum, we identified 22 other local minima within the protein matrix (Figures 3.1d to 3.1f). I observed a local minimum in the proximity of the only known Xe-binding site [187] that is near the central cavity [5] and next to the H-cluster (local minimum 1 in Figures 3.1d to 3.1f). The relative free-energy difference for this local minimum is ~ 13.31 kcal/mol with respect to the global minimum. I further observe local minima 2 and 20 (Figures 3.1d to 3.1f) within a previously identified water pathway [6] (pink surface in Figures 3.1d to 3.1f) that extends from the solvent to the central cavity [6] (green surface in Figures 3.1d to 3.1f). The local minimum 2 is in the proximity of S320 (with a free-energy of 11.72 kcal/mol), and the local minimum 20 is in the proximity of the surface-exposed K571 (with a free-energy of 13.76 kcal/mol). Several other local minima are located throughout the protein matrix with a large number of them in the active-site domain (Figure 3.1).

3.4.2 Diffusion Pathways of O₂ in [FeFe]-hydrogenase

I characterized diffusion pathways of O₂ in [FeFe]-hydrogenase by computing MFEPs between pairs of local minima located in the 3D-PMF (see methods section for details). In Figures 3.1a to 3.1c, I show the locations of all local minima by red spheres. The MFEPs are relatively dense in the vicinity of the H-cluster in the active-site domain of the protein, and in the non-active-site domain, I find MFEPs only near the protein surface (Figure 3.2a). The pathways to the Geminant-site (G-site, a location near the global minimum indicated by 0/0-G in Figure 3.2a,b) are largely originating in the lower side of the active-site domain via the central cavity (green surface in Figures 3.1d to 3.1f), in the water pathway [6] (pink surface in Figures 3.1d to 3.1f), and in the upper-part of the active-site domain.

In Figure 3.2a, I show the diffusion pathways of O₂ within the [FeFe]-hydrogenase with color-coded values of the free-energy. The global minimum, located next to the H-cluster, is shown by a blue sphere, on the side of the distal Fe-site of the H-cluster [49]. The local minimum corresponding to the Xe-binding site inside [FeFe]-hydrogenase is designated as minimum 1, and the MFEP that connects the Xe-site to the G-site is shown by a pathway between minima 0 and 1 (Figure 3.2a), and the barrier of this MFEP is in the proximity of A272. The local minimum 2 and 20 are in the proximity of S320 and K571, respectively; both have MFEPs to the G-site (Figure 3.2a). The energy barrier of the MFEP between the local minimum 2 and the G-site is in the proximity of A321.

The local minimum 9 is in the proximity of cluster 1, and the MFEP between the G-site and local minimum 9 shows a pathway for O₂ with access both to the H-cluster and to the adjacent [Fe₄S₄] cluster. The local minima 7, 8, and 19 with free-energy values of 11.86, 12.18, and 8.88 kcal/mol are located within the active-site domain, and the pathway between them has access to the H-cluster from the local minimum 7, and has access to cluster 1 from the local minimum 19. The access of O₂ to the local minima 7, 8, and 19 from pathways originating in the active-site domain is more likely due to lower energy barriers (Figure 3.2a). The Xe-site has access to the local minima 2 and 4 (Figure 3.2a) with the energy barriers

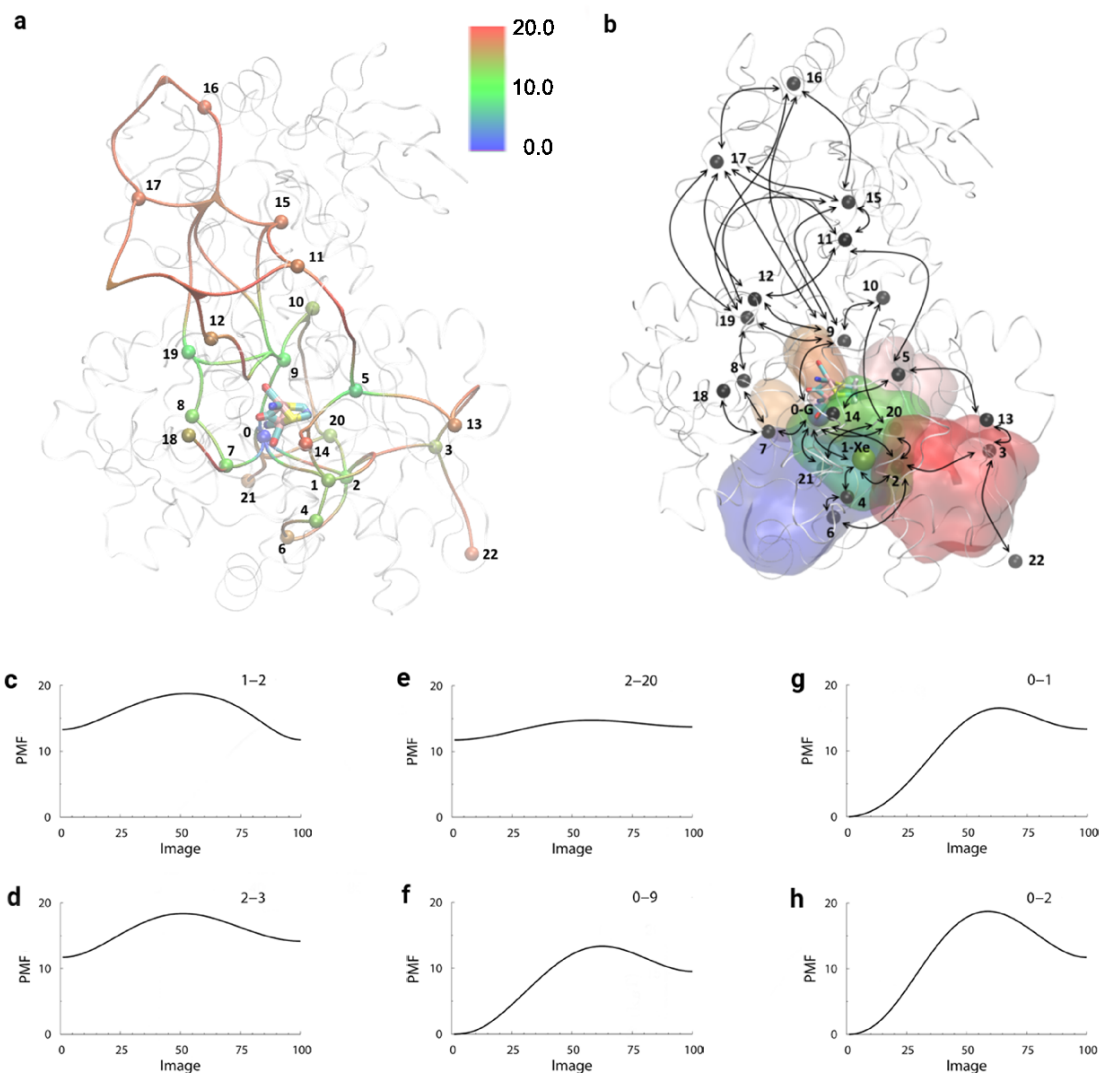


Figure 3.2: (a) Pathways of O₂ migration inside [FeFe]-hydrogenase, values of the PMF along each MFEP and local minima, are indicated by color. The pathways are represented as curved lines and the local minima as spheres. (b) The connection between local minima along pathways are shown by black arrows. The local minimum 0 corresponding to the G-site is shown as a blue sphere, and the local minimum 1 corresponding to the Xe-site is shown as a green sphere. The bottom panels show the PMF profiles along MFEPs between local minima 1 and 2 (panel c), local minima 2 and 3 (panel d), local minima 2 and 20 (panel e), local minima 0 and 9 (panel f), local mlocal minima 0 and 1 (panel g), and local minima 0 and 2 (panel h).

of 7.72 and 0.76 kcal/mol, indicating that it is easier for O₂ to enter the Xe-site from the local minimum 4 than from the local minimum 2. The local minimum 6 is located within the channel B (blue surface in Figures 3.1d to 3.1f), as described by Cohen et al. [5], and it is connected to local minima 2 and 4 with energy barriers along MFEPs of 1.57 and 1 kcal/mol, suggesting that the energy barriers are lower along the pathway 6-4-1 than the pathway 6-2-1.

The energy barrier for O₂ entering the G-site from the Xe-site (pathway 1-0) is 3.6 kcal/mol, whereas the energy barrier from the local minimum 2 to the G-site (pathway 2-0) is 7.29 kcal/mol, suggesting that it is easier for O₂ to access the H-cluster from the Xe-site than from the local minimum 2. The sum of energy barriers along three pathways to the H-cluster starting from local minimum 6, pathways 6-4-1-0, 6-2-0, and 6-2-1-0 are 5.36, 8.86, and 12.89 kcal/mol, suggesting that it is unlikely for O₂ to take pathway 6-2-1-0 in comparison to pathways 6-2-0 and 6-4-1-0 (Figure 3.2b). The local minimum 3 is located within the channel A (red surface in Figures 3.1d to 3.1f), as identified by Cohen et al.[5], and it is connected to the local minimum 2 with an energy barrier of 4.33 kcal/mol along the MFEP. The two pathways to the H-cluster starting from the local minimum 3, pathways 3-2-0 and 3-2-1-0, have energy barriers of 11.62 and 15.65 kcal/mol, suggesting that O₂ diffusion is less likely along the pathway 3-2-1-0 than the pathway 3-2-0 to the H-cluster (Figure 3.2b). A table of energy barrier values between local minima is reported in Table E.5.

The free-energy profiles for MFEPs between local minima 0 and 1, local minima 1 and 2, local minima 2 and 3, local minima 0 and 2, local minima 0 and 9, and local minima 2 and 20 (Figure 3.2c to 3.2h) are showing the change in PMF values as O₂ is migrating along the pathway between them, and all PMFs are shown with respect to the global minimum at the G-site (indicated by 0 in Figure 3.2b). The pathways on the upper side of the H-cluster, e.g., pathway 7-8-19-9, are suggesting the existence of a O₂ circulation network with relatively low-energy barriers around the H-cluster; however O₂ migration is less likely to direct to

pathways in the non-active-site domain due to higher energy barriers, e.g., energy barrier of the pathway 9-12 is 11.96 kcal/mol, suggesting that it is unlikely for O₂, after entering the protein from the active-site domain, to access the non-active-site domain versus entering the protein from the active-site domain to access the H-cluster.

The local minima 21 and 22 with free-energy of 16.91 and 18.21 kcal/mol, respectively, are located in the solvent. These PMF values are showing an estimate of the free-energy values in the solvent and suggest an oxygen diffusion gradient toward the protein interior. The local minimum 22 is connected to the local minimum 3 that has access to the H-cluster via channel A (Figures 3.2b) [5]. The local minimum 21 is connected via an MFEP to the local minimum 0 or the G-site with an energy barrier of 1.8 kcal/mol which is in the proximity of F570, suggesting that this particular pathway potentially provides the easiest access route for O₂ to the H-cluster. This pathway is located between the channel B [5] and the water pathway [6]. The local minimum 20 with an energy of 15.34 kcal/mol is located within the water pathway, and is connected to the local minimum 2 and the G-site with energy barriers of 1.04 and 3.5 kcal/mol, indicating that it is less likely for O₂ to migrate along pathways 20-2-1-0 or 20-2-0 than the pathway 20-0. The pathway 9-0 is in the proximity of residues K358 and E361, defining the water-based proton transfer pathway [188, 7] (orange surface in Figures 3.1d to 3.1f). The free-energy profile of O₂ diffusion along this pathway is shown in Figure 3.2f indicating an energy barrier of 4 kcal/mol along this MFEP. The local minimum 10 with a free-energy value of 14 kcal/mol is connected to the local minimum 9. Via this pathway, O₂ entry from the local minimum 10 could be directed to the proximity of cluster 1 and to the H-cluster, suggesting that the pathway 10-9-0 could be another access route for O₂ to access the H-cluster, and similarly, the pathway 13-5-1-0 shows another possible route to the H-cluster.

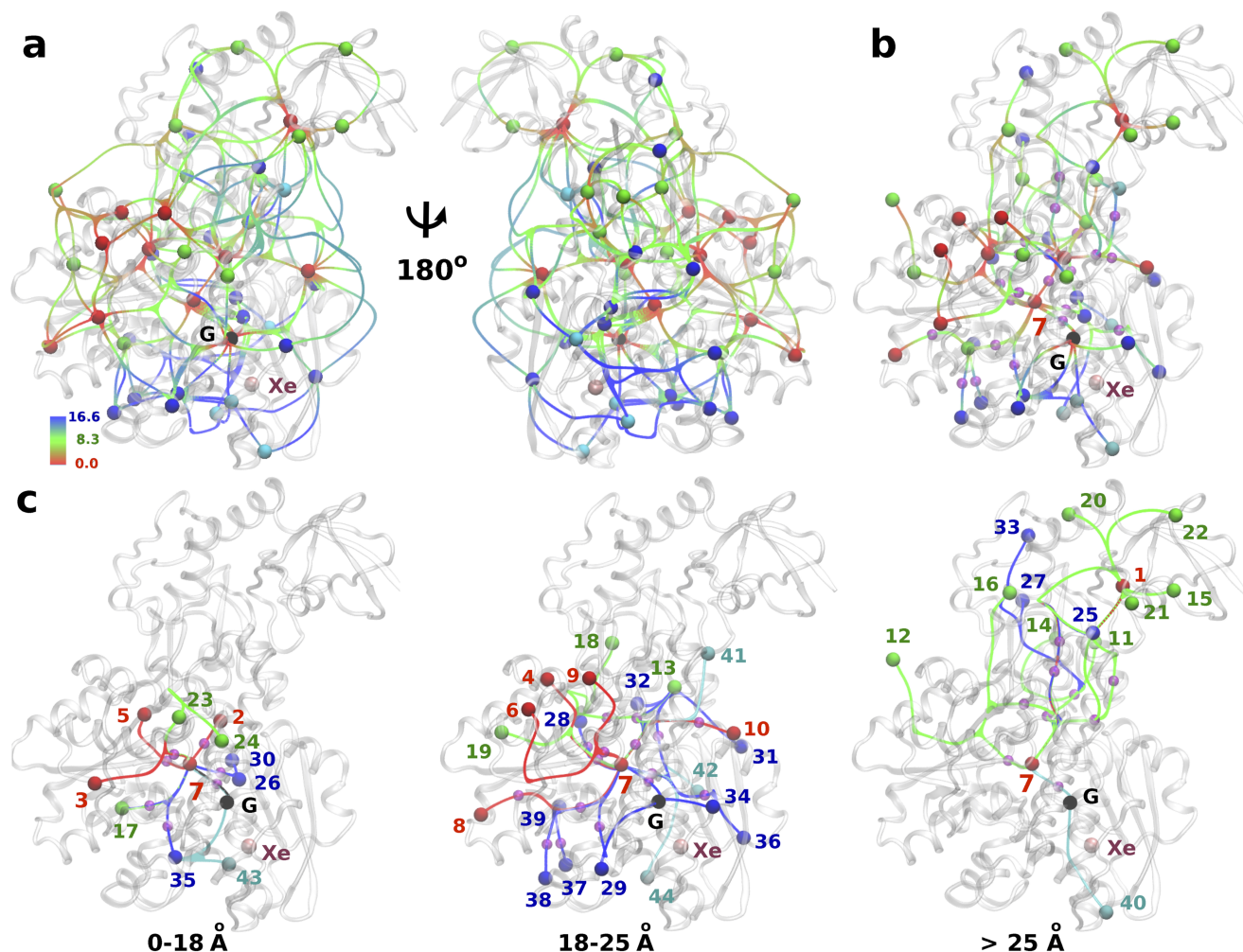


Figure 3.3: (a) Front and back views of MFEPs for CO diffusion are depicted on the structure of the FeFe-hydrogenase along with the positions of all minima (indicated by spheres). (b) The MFEPs for CO diffusion from the other local minima to the local minimum 7 are shown. (c) The MFEPs for CO diffusion within 0-18 Å (leftmost panel), 18-25 Å (middle panel), and over 25 Å (rightmost panel) of the local minimum 7 are shown. The locations of the energy barriers are shown by transparent magenta spheres in panels b and c (This figure was created by my co-author Yong Liu. [3]).

3.4.3 Network of CO Diffusion Pathways

We located an MFEP for each pair of local minima using the zero-temperature string (ZTS) method [183, 184, 185] (section 2.4.1.2). The ZTS-algorithm begins with an initial guess for a curve connecting the chosen minima pair on the analytically-reconstructed 3D-FES. This initial guess is then iteratively optimized via a discretized parameterization scheme

by keeping points along the curve equidistant on each iteration while allowing points to move on the basis of the steepest descent on the 3D-FES. Therefore, MFEPs are the pathways of minimal free-energy for CO diffusion in the FeFe-hydrogenase, which are assumed as the most likely diffusion channels for CO.

We show the network of these pathways in Fig. 3.3a, where one can visualize the color-coded free-energy values along with the MFEPs between local minima pairs (depicted as spheres). Consistent with the 3D-FES, we found that the MFEPs of CO diffusion along previously proposed hydrophobic pathways have high free-energy values (over 8 kcal/mol), which are shown as blue curves. We also observed that the MFEPs interconnect multiple local minima between the solvent-space and the H-cluster. We present a center-by-center matrix of energy barriers between all pairs of local minima in Appendix E. We find that major energy barriers along MFEPs originating from the global minimum to neighboring local minima are ~ 10 kcal/mol, and these barriers are higher than the energy barriers of reverse MFEPs. This suggests a preference for CO diffusion toward the global minimum. Previous studies have suggested that CO binds the Fe_d of the 2Fe subcluster of the H-cluster, thereby competing with the binding of O₂ [174, 178, 175]. To explore the CO diffusion pathways connecting the solvent region and the H-cluster, we present the MFEPs of CO diffusion between the local minimum 7 (which is located in the vicinity of the Fe_d of the 2Fe subcluster of the H-cluster, an area known as the geminate site) and other neighboring minima (Fig. 3.3b and 3.3c). The locations of the major energy barriers along MFEPs are indicated by magenta spheres (Fig 3.3c).

Within 18 Å of the H-cluster, we identified 12 local minima, including the global minimum: 5 minima with the free-energy values lower than 4 kcal/mol (red/black spheres), 3 minima with the free-energy values in the range of 4 to 8 kcal/mol (green spheres), 3 minima with the free-energy values in the range of 8 to 12 kcal/mol (blue spheres), and 1 minimum with a free-energy value of 18.20 kcal/mol (cyan sphere).

The energy barriers for pathways 2-7 and 5-7 are 2.45 kcal/mol and 5.23 kcal/mol, respectively. The pathways 3-7, 23-7, and 24-23-7 have energy barriers of 9.29 kcal/mol, 3.61 kcal/mol, and 7.58 kcal/mol, respectively. The pathways 17-7 and 35-7 share a part of their MFEPs, but the major energy barriers are located at distinct positions, in the vicinity of the residue F328 for the pathway 17-7 *vs.* the residue D273 for the pathway 35-7. The MFEP originating at the minimum 30 passes through the minimum 26 before arriving at the minimum 7. Therefore, the pathways 30-7 and 26-7 share the same major energy barrier of 2.49 kcal/mol. The pathway 43-7 passes through the local minimum 35 as well as the global minimum (G), and both MFEPs 43-7 and G-7 have the same major energy barrier of 14.62 kcal/mol. This suggests that a deep local minimum with high energy barriers of CO escape could trap CO.

Within 18 to 25 Å of the H-cluster, we found 20 additional distinct local minima, most of which are located at or near the surface of the enzyme and likely serve as local reservoirs for CO as it enters from the solvent. The local minima 4, 6, 8, 9, and 10 have free-energy values lower than 4 kcal/mol (relative to the global minimum). The pathways 6-7 and 9-7 share part of their MFEPs as well as the location of the major energy barrier in the vicinity of the residue A230, but the values of the free-energy barriers for each pathway are distinct, 9.29 kcal/mol (pathway 6-7) and 8.15 kcal/mol (pathway 9-7). Also, the pathways 6-7 and 9-7, respectively, pass through the minima 3 and 23, two minima located within 18 Å of the H-cluster (left panel; Fig. 3.3c). Similarly, the pathway 4-7 passes through the local minimum 5 and shares a part of its MFEP with the pathway 19-7 as well as the location of the major energy barrier in the vicinity of the residue V352, but the value of the energy barrier for each pathway is different, 5.23 kcal/mol for the pathway 4-7 (same value as for the pathway 5-7) and 8.15 kcal/mol for the pathway 19-7. Even though the pathways from the local minima 8, 37, 38, and 39 to the minimum 7 merge in the pathway 17-7, the locations of the major energy barriers along the MFEPs of 8-7, 37-7, and 38-7 are different. Specifically, the barriers along pathways 8-7, 37-7, and 38-7 are located near residues F348, I567, and

A331, respectively. The energy barrier for the pathway 39-7 is in the vicinity of residues S323 and Q325. Since the pathway 29-7 passes through the local minimum 35, the location of the energy barrier is same as for the pathway 35-7. Several other pathways show shared as well as unique features: the pathways 10-7, 41-7, 32-13-7 and 18-5-7 merge into the pathway 2-7, but the location of the energy barrier for each pathway is different, and the pathways 31-13-30-26-7, 36-26-7, 42-30-26-7, and 44-26-7 merge into the pathway 26-7, but the energy barriers for 31-13-30-26-7 and 36-26-7 are located in the unshared parts of their MFEPs and the energy barriers for 42-30-26-7 and 44-26-7 are located in the shared part 26-7. The pathway 36-7 passes through the global minimum and has the same value for the major energy barrier with the same location as the pathway G-7.

We observed 13 local minima at a distance over 25 Å from the Fe_d of the H-cluster, most of which are located in the non-active-site domain of the enzyme (rightmost panel in Fig. 3.3c). Among these 13 minima, two (12 and 40) are located in the solvent region near the enzyme surface and show MFEPs reaching to the minimum 7: the pathways 12-19-5-7, 19-5-7, 4-5-7, and 5-7 all arrive at the minimum 7 and along which the major energy barrier of 5.23 kcal/mol is shared, and the pathway 40-7 passes through the local minimum 43 as well as the global minimum to arrive at the local minimum 7. The MFEPs of several other minima in the non-active-site domain merge at the local minimum 2, but each MFEP has a different location and magnitude of the energy barrier (see Tables E.5 and E.6).

3.5 Discussion

In this chapter, we have presented details on characterizing O₂ and CO diffusion channels and their underlying thermodynamics via a combination of several state-of-the-art computational methods [145, 140, 141, 150]. The 3D-PMF for O₂ binding shows a preference for O₂ diffusion into the protein interior relative to the solvent. The preference for protein environment by O₂ has been similarly observed in the earlier work of Cohen et al. [5]. The global minimum

Table 3.1: List of candidate residues for mutagenesis to increase the tolerance of the CpI FeFe-hydrogenase for CO. The symbol \checkmark indicates that the residue has also been previously tested/proposed [1, 2] for disrupting the diffusion network of the competing inhibitor O_2 in this enzyme (This table was created by my co-author Yong Liu [3]).

Residue	MFEP(s)	experimental	computational
L191	13-2		\checkmark
C193	31-26		
I197	25-2	\checkmark	
A228	6-7		\checkmark
A230	9-7, 6-7		
M295	44-26		\checkmark
C300	26-7		
G302	42-26, 31-26		\checkmark
S323	39-7, 25-7		\checkmark
Q325	39-7, 25-7		
V352	19-7, 4-7		\checkmark
K358	5-7, 4-7, 19-7, 28-7		\checkmark
T380	4-7		\checkmark
I416	41-G		
V423	G-7	\checkmark	
A426	29-G		\checkmark
V496	36-26, 44-26		
H500	10-2		
C503	2-7		

in the free-energy is close to the residue P324 and the H-cluster (Figure 3.4a), in the vicinity of an area termed as the **G**-site (Geminate) [10]. The G-site is a location one step away from covalent binding of O_2 to the H-cluster that irreversibly inactivates the protein. The noticeable difference in the free-energy between the global minimum and other local minima would make it difficult for O_2 to leave the global minimum to other areas of the protein matrix, thus making binding to the H-cluster highly likely, because this is the only path from the G-site that leads to a lower free-energy, as also highlighted in previous studies [10, 5]. All local minima with energies close to the global minimum are in the active-site domain of the protein, whereas those local minima in the non-active-site domain of the protein have energy values closer to those in the solvent.

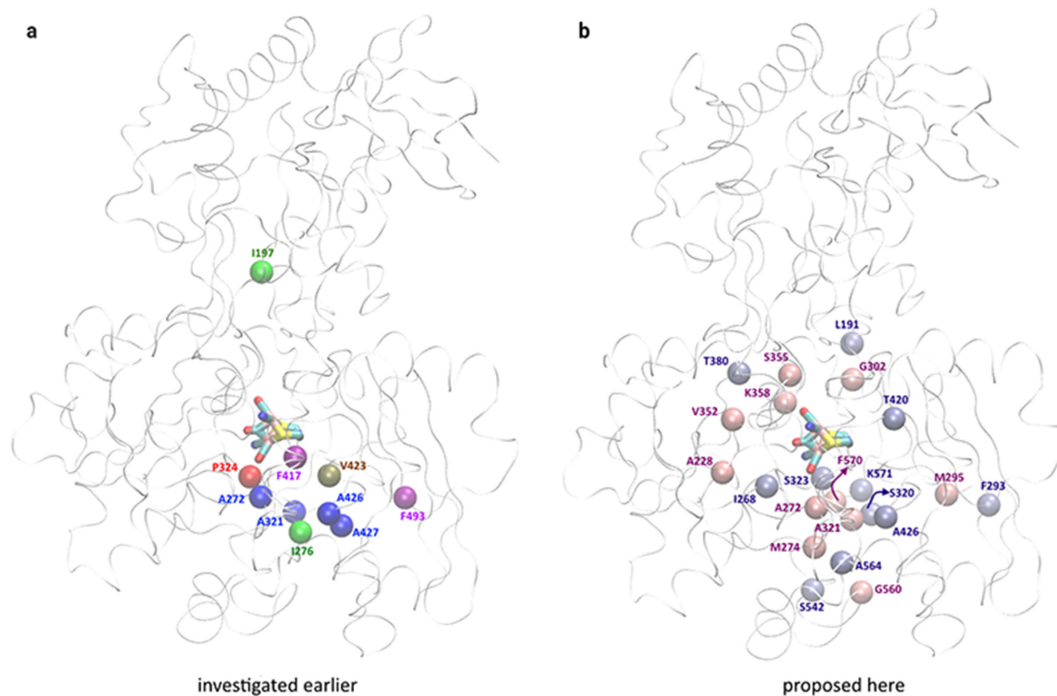


Figure 3.4: The α carbon-atoms of several candidate residues investigated by earlier studies (panel a) and proposed in this work (panel b) are shown. (a) The residue P324 is located near the H-cluster and the G-site. The residues I197 [8, 9], F417 [10], A321 [11], and V423 [11] are experimentally shown to alter O_2 diffusion in [FeFe]-hydrogenase. The hydrophobic residues A426, A427 and V423 are located in the proximity of the Xe-site. The residues F493 and A427 are along the pathway to the Xe-site. The residues I276, A272, V423, F493 and A427 are proposed for mutation in the patent application of King et al. [12] (b) Potential candidate residues for mutations to alter the oxygen-diffusion kinetics in [FeFe]-hydrogenase (A321 has been tested in study of Ghirardi et al. [11]) from our work. The residues at the local minima are shown as blue spheres, and near the energy barriers are shown as pink spheres.

3.5.1 O_2 Pathways in the Vicinity of the Xe Site

In crystallographic studies, Xe as a probe (similar to O_2) has been observed to prefer hydrophobic environments in the protein, but it is more electron rich than O_2 , so it can readily be detected by the crystallographic methods [180]. We observed a local minimum close to the only experimentally known Xe-binding site in this [FeFe]-hydrogenase [187]; as expected, this local minimum is located in a mostly hydrophobic environment of the protein,

e.g., A426, A427, and V423 (Figures 3.4a). The Xe-binding site is reported to be blocked by the side-chains of A427 and F493 in CpI hydrogenase [6]. We found that the computed pathway to this local minimum is well aligned with the side-chains of A427 and F493, along the pathway 1-2. This local minimum is in the proximity of A426 (Figure 3.4a). Importantly, another experimental study on [FeFe]-hydrogenase found that the mutation A426L makes the Xe-binding site smaller [6].

3.5.2 O₂ Diffusion in Water and Proton-Transport Pathways

The location of two local minima (the local minimum 2 and 20) within the proposed water pathway [6], shows that O₂ and water may both use this pathway, as has been suggested by Lautier et al. [6]. The local minimum 2 and 20 are in the proximity of S319 and K571, respectively. In agreement with previous experimental and theoretical work, both residues are found to have contact with stabilized water molecules [170, 33]. Several theoretical and experimental studies show that surprisingly, hydrophobic ligands can occupy and use hydrophilic channels as well. In a previous computational study, MD simulations have revealed that water molecules occupy a Xe-binding site in myoglobin [189], and another study shows that the local minimum of 3D-PMF of water in myoglobin corresponds with the Xe-binding sites [57]. It is therefore interesting to note that Xe or O₂ displace water molecules under pressure in protein crystals [180, 190]. Indeed, another theoretical study on catalase using MD simulations found that O₂ and water share the same channel [191]. The presence of pathways in the vicinity of water pathway [6] and the proton-transport [188, 7, 192] pathway in [FeFe]-hydrogenase suggest the existence of possible common pathways between water and gas molecules to access the active-site, as also observed in [NiFe]-hydrogenase by Sumner et al. [193].

3.5.3 O₂ MFEPs Near the G Site and the Xe Site

The O₂ migration pathways in [FeFe]-hydrogenase show the areas within the protein through which the access of O₂ is most likely. Notably, there is only one pathway found

between the Xe-binding site and the G-site located within the central cavity. The residue A272 is found along the energy barrier of this pathway (Figure 3.4b), and its substitution is proposed in a patent application by King et al. [12] for increasing oxygen tolerance of [FeFe]-hydrogenase. Furthermore, this MFEP is in the proximity of areas of a key transition path to the G-site, as described by Kubas et al. [10], because they found that the mutation F417Y decreases the kinetic rate constant of O₂ transition by 100-fold along this path [10]. We observed an increase in the free-energy barrier of oxygen transition in this area by F417Y mutation (see Figure E.1).

3.5.4 O₂ MFEP to the Accessory Metal Cluster

The MFEPs of O₂ diffusion within the protein matrix core could reveal information about the mechanism of [FeFe]-hydrogenase active-site inhibition by O₂. The pathway 9-0 (Figure 3.2a) shows a migration path of O₂ from the [Fe₄S₄] cluster to the H-cluster. This MFEP shows that the direction of O₂ diffusion path from cluster 1 to the G-site is thermodynamically more favorable, which correlates well with the experimental study by Stripp et al. [43]. It is suggested that O₂ reacting with the H-cluster produces reactive oxygen species (ROS), that either binds, presumably, at the distal Fe-site of the H-cluster [49] or migrates along a short-path to oxidize cluster 1 [43, 194]. Stripp et al. [43] suggested that ROS very likely is a superoxide that is more potent oxidant than O₂ with a higher energy to overcome the energy barriers to leave the H-cluster space for cluster 1 [43], and it is possible that superoxide could potentially migrate along this MFEP for O₂ between the G-site and cluster 1.

3.5.5 MFEPs and Experimental Understanding of O₂ Pathways

Many experimentally tested residues that are known to impact O₂ access to the H-cluster are located along our observed pathways (Figure 3.4a) [55, 11, 6]. Using the site-directed mutagenesis, Stapleton et al. [8, 9] found only residue I197 to measurably decrease O₂ inhibition of Cpl [FeFe]-hydrogenase by a single mutation to V197 [55]. We find that I197

is located along pathways 9-15, 9-16, and 9-17 in the upper-side of the H-cluster that are connected to the local minimum 9 in the proximity of cluster 1, suggesting that the mutation of I197 is impacting the pathways forming a circulation network of O₂ on the upper side of the the H-cluster by limiting the access of O₂ to the active-site of the protein. Ghirardi et al. [195], using an *E. coli* expression system, made mutations (A321I and V423L) in CpI [FeFe]-hydrogenase that were effective in altering O₂-access to the H-cluster [11]. Pathways 3-2-1-0 and 3-2-0 are located in the proximity of the channel A [5] toward the H-cluster, and pathways 6-2-1-0, 6-4-1-0, and 6-2-0 are located in the proximity of the channel B [5] toward the H-cluster (Figures 3.2b). The residue A321 is located along the pathway 2-0 with an energy barrier of 7.29 kcal/mol, consistent with the work of Ghirardi et al. [11], and V423 is located near the pathway 1-0; thus either of these site mutations can impact pathways within the central cavity [5], where pathways from both channels merge. By looking at the full network of pathways, we can find pathways 1-5 and 20-2, that both have accessibility to the solvent other than pathways within the channels A and B. Therefore, the single mutations of V423 and A321 could affect multiple O₂ pathways to the H-cluster. In fact, such an unintended effect of these mutations is hypothesized by Ghirardi et al. [11], but is confirmed by our network of O₂ pathways within the protein.

The interconnected network of MFEPs shows that it is more likely for O₂ to diffuse along channel B than channel A. As there are three possible pathways from the channel B to the H-cluster, and 2 possible pathways from the channel A to the H-cluster, and the energy barriers of these 5 pathways show that the lowest overall energy barriers are along pathways in the channel B. Channel B is a dynamic pathway, first discovered by a theoretical study of O₂ diffusion in [FeFe]-hydrogenase [5], and given the presence of many pathways to the H-cluster from the channel B, residues in the channel B along the MFEPs, such as A321 and A564, are prime candidates for mutations (Figure 3.4b). However, a patent application protecting the redesign of [FeFe]-hydrogenase by incorporating bulky amino-acids suggested residues that are concentrated near the central cavity and the channel A [12], neglecting

residues in the channel B. This disparity is also noted by Lautier et al. [6]. Nevertheless, we find many residues along MFEPs, the mutations of which are suggested in the patent application by King et al. [12]: A272 is along the pathway 0-1, I276 is along the pathway 4-1, V423 is near the Xe-site, and A427 and F493 are along the pathway 2-1 toward the Xe-site (Figure 3.4a).

The presence of pathways 20-2-0 and 20-0 (within the water pathway [6]) to the H-cluster corroborates with the proposal [6] that O_2 is likely using this hydrophilic channel to access the active-site of [FeFe]-hydrogenase (Figures 3.2b). Although the local minimum 20, close to K571, is located within the water channel, we note that only pathway 20-2 is along the suggested water pathway, and the pathway 2-0 is located in the central cavity. The pathway 20-0 is partially located in an area between the channel B [5] and the water pathway [6]. The pathway 21-0 is perhaps the easiest access route to the H-cluster for O_2 located between the channel B [5] and the water pathway [6].

3.5.6 Potential Candidate Residues for Future Mutations along O_2 MFEPs

The proximity of the pathway 21-0 to these channels might cause the mutations of residues (in the channel B and the water channel) to affect this pathway indirectly, i.e., allosterically. The energy barrier of this pathway is located close to the surface-exposed F570, making this residue a good candidate for mutations to disrupt the O_2 diffusion along this pathway. Similarly, O_2 diffusion along pathway 13-5-1-0 could be hindered by mutations of residues along energy barriers of pathways 5-1 and 1-0. In a similar fashion, for pathway 10-9-0, the residues close to energy barriers are suitable choices for mutations. All such candidate residues are shown in Figure 3.4b. Our work suggests the existence of multiple pathways for O_2 to access the H-cluster beyond previously known channels A and B, such as the presence of pathways in the vicinity of the water pathway [6] and the proton transport [188, 7], and pathways in other areas of the active-site domain. The construction of a systematic network of migration pathways of O_2 along MFEPs provides unique physical insight into the complex

nature of ligand diffusion in [FeFe]-hydrogenase, and shows that it is highly informative for the purpose of identifying candidate residues (Figures 3.4b) for future experimental studies by mutagenesis or other methods.

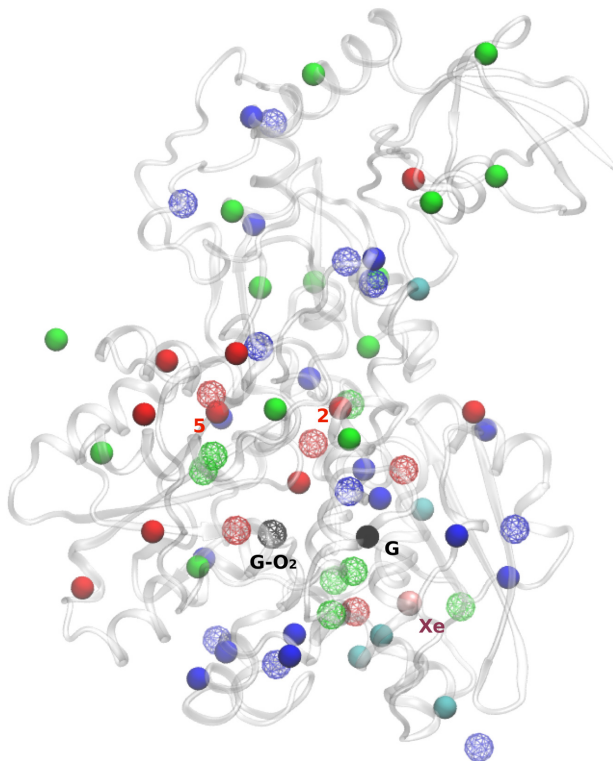


Figure 3.5: Locations of the minima for CO (filled spheres) and O₂ (wireframe spheres) diffusion in the CpI FeFe-hydrogenase are shown. The global minimum for both gases are shown in black filled or wireframe spheres and are labeled (G for CO and G-O₂ for O₂). The Xenon cavity is shown as a pink sphere and labeled as Xe. For CO, colors of other minima are consistent with Fig. 3.3: red (0-4 kcal/mol), green (4-8 kcal/mol), blue (8-12 kcal/mol), and cyan (over 12 kcal/mol). For O₂, colors of minima indicate the following energy ranges: red (0-12 kcal/mol), green (12-16 kcal/mol), and blue (over 16 kcal/mol) (This figure was created by my co-author Yong Liu [3]).

3.5.7 CO MFEPs to the active site

The MFEPs of CO diffusion from other minima to the local minimum 7 could potentially disclose the mechanism of CO access to the active site. In Fig. 3.3, we show the MFEPs from 44 distinct minima to the local minimum 7 as well as the locations of major barriers along MFEPs. The major energy barriers range between 2.45 kcal/mol and 16.68 kcal/mol,

among which the the pathways 2-7 and 41-7 have the lowest free-energy barriers and the pathway 28-7 has the highest free-energy barrier.

Among thermodynamically favored local minima at the surface of the enzyme (those with free-energy values below 4 kcal/mol relative to the global minimum), the pathway 10-2-7 has the lowest energy barrier of 4.15 kcal/mol, although the pathway 4-5-7 also has a comparable energy barrier of 5.23 kcal/mol. However, the exit route for CO from the H-cluster requires overcoming a barrier of at least 5.04 kcal/mol, which is the energy barrier along the reverse MFEP 4-5-7. The lower energy barriers along the MFEP 4-5-7 for CO entry/exit from the enzyme make this pathway a highly likely route for CO diffusion between the solvent and the H-cluster.

3.5.8 Comparisons between CO and O₂ diffusion in the enzyme matrix

In our study of O₂ diffusion in the Cpl FeFe-hydrogenase, we identified 23 minima including the global minimum, which are shown in Fig. 3.5 in a wireframe representation [1]. We observe that two local minima of O₂ partially overlapped with the local minima 2 and 5 of CO diffusion, indicating that CO and O₂ may reside within the enzyme matrix at shared locations. The global minimum for both inhibitory gases CO and O₂ are located near the active site, which suggests a high-affinity for both gases surrounding the active site. The global minimum for CO is located near the residue F417, where F417Y mutation was previously shown to decrease the kinetic rate constant for O₂ diffusion [10, 1].

Although some common features can be found, CO diffusion and O₂ diffusion in the Cpl FeFe-hydrogenase differ in many aspects. For example, no local minimum for O₂ is found in the immediate neighborhood of the H-cluster, but for CO, the local minimum 7 is situated at a distance of 2.75 Å from the H-cluster. This suggests that CO likely has easier and faster access to the active site than O₂, as has been previously suggested [196, 175]. This is further reinforced by the fact that, on comparing the 3D-FES for CO in this work with that of O₂ from our previous work [1], we find overall lower free-energy barriers for CO diffusion (with

respect to the global minimum for CO) than for O₂ diffusion (with respect to the global minimum for O₂), even though the free energy difference (ΔF) between the global minimum and the solvent space for each gas are comparable, 18.81 kcal/mol (CO) and 18.5 kcal/mol (O₂).

However, in the regions of the enzyme with two hydrophobic pathways, we find that O₂ diffusion is thermodynamically favored over CO diffusion, because five local minima (four shown in green wireframe spheres and one shown as a red wireframe sphere near the Xe cavity in Fig. 3.5) with relatively lower free-energy values of O₂ diffusion are located along hydrophobic pathways.

Even though there exists a local minimum in the vicinity of the Xe cavity for both CO and O₂, the Xe cavity is located in the lower free-energy region of the 3D-FES for O₂ while CO diffusion in the vicinity of the Xe cavity requires overcoming higher free-energy barriers. The Xe cavity is surrounded by three hydrophobic residues (F493, A427, and A431) which may stabilize apolar gases over polar gases [176], indicating that O₂ is more likely to be found in the Xe cavity. These differences can also be explained by the observation that CO is a polar molecule with a van der Waals volume of 16.20 cm³/mol, while O₂ is an apolar molecule with a van der Waals volume of 13.00 cm³/mol [197]. The effect of the size of gases on diffusion pathways has been demonstrated in previous MD simulations of H₂ and O₂ diffusion in the CpI FeFe-hydrogenase, where H₂ with a smaller van der Waals radius was shown to diffuse in a broader region of the protein matrix and on shorter timescales [198]. Furthermore, CO was shown as a better ligand for binding to electron-rich metals compared with O₂ [199]. This is consistent with the observation that CO reacts much faster than O₂ with the FeFe-hydrogenase based on electrochemical studies [175, 174, 196, 200].

3.5.9 Candidate residues for mutations

To prevent inhibitors (e.g. CO and O₂) from binding to the Fe_d of the 2Fe subcluster of the H-cluster, mutagenesis of protein residues is an approach to alter diffusion pathways.

Based on the pathways and thermodynamics of CO diffusion, we propose key candidate residues for mutations to disrupt or block the diffusion of CO (Table 3.1). These 19 residues are located in the vicinity of the energy barriers of the targeted MFEPs for CO diffusion. For example, the residue S323 is located near the energy barrier for the pathway 39-7 and is along the previously proposed water pathway [176]. Similarly, the residue K358 is in the proximity of the energy barrier of the pathway 5-7, and is also one of the residues defining the water-based proton-transfer pathway [1, 188, 7]. We hypothesize that replacing this residue with mutants having larger side-chains could disrupt the CO diffusion along multiple MFEPs (5-7, 4-7, 19-7, and 28-7). The presence of MFEPs for CO in the vicinity of the proton-transport pathway [192, 7, 188] as well as the water pathway [176] also suggests the existence of shared pathways between water and gas molecules [193, 57].

Among the proposed 19 residues to disrupt CO diffusion within the enzyme, 11 have been experimentally studied or computationally proposed (marked as ✓ in Table 3.1) to decrease the rate of O₂ diffusion inside the CpI FeFe-hydrogenase [8, 11, 1]. For example, both I197 [9] and V423 [11] have been experimentally studied to alter the diffusion rate of O₂. The other 9 residues have been proposed to block O₂ diffusion [1], among which A228, M295, G302, V352, and K358 are located in the proximity of energy barriers for O₂ diffusion. This result indicates that mutagenesis of common residues may decrease the diffusion rates of CO and O₂ in the CpI FeFe- hydrogenase. We speculate that mutations of I416, V423, and A426 may have a weaker effect on decreasing the rate of CO diffusion because these three residues are located in the enzyme region with higher free-energy values for CO diffusion.

3.6 Conclusion

In this work, [FeFe]-hydrogenase is investigated using the single-sweep/string method of Maragliano et al. [56] to map the migration pathways of O₂ and CO diffusion in this enzyme. In the reconstructed free-energy map, local minima are notably found near the active-site,

the H-cluster, and a Xe-binding site. By assuming that the diffusion of O₂ and CO inside this protein is following the pathways of minimal free-energy, we found that there are multiple interconnected O₂ and CO migration pathways from the solvent to the H-cluster. The overall network of pathways shows that O₂ and CO transport in [FeFe]-hydrogenase is not limited to previously known pathways [5, 6], and O₂ is able to access the H-cluster for inactivation via several alternative routes. Our results provide an enhanced view of gas diffusion in [FeFe]-hydrogenase and suggest new residues (Figure 3.4b), mutations of which could increase oxygen tolerance. Furthermore, 11 of the proposed residues were also found in CO studies. These findings will guide future experimental studies to increase CO and O₂ tolerance of hydrogenases, as recently the more tolerant mutant of [FeFe]-hydrogenase found by mutation of residue found in the neighborhood of proposed residues [201].

3.7 Publications

The work described in this chapter has resulted in the following journal articles:

- Mohammadi, M., Vashisth, H. (2017). Pathways and Thermodynamics of Oxygen Diffusion in [FeFe]-Hydrogenase. *J. Phys. Chem. B*, 121(43), 10007-10017.
- Liu, Y., Mohammadi, M., Vashisth, H. (2018). Diffusion network of CO in FeFe-Hydrogenase. *J. Chem. Phys.*, 149(20), 204108.

CHAPTER 4

STUDIES ON TDZD (ALIPHATIC VS. AROMATIC) INHIBITORS OF RGS4

4.1 Abstract

RGS proteins play a pivotal role in regulation of GPCR signaling and are therefore becoming an increasingly important therapeutic target. Recently discovered thiadiazolidinone (TDZD) compounds that target cysteine residues have shown different levels of specificities and potencies for the RGS4 protein, thereby suggesting intrinsic differences in dynamics of this protein upon binding of these compounds. In this work, we investigated the effect of binding of several small-molecule inhibitors on dynamical motions in RGS4. Specifically, we studied two conformational models of RGS4 in which a buried cysteine residue is solvent-exposed due to side-chain motions or due to flexibility in neighboring helices. We found that TDZD compounds with aromatic functional groups perturb the RGS4 structure more than compounds with aliphatic functional groups. Moreover, small-molecules with aromatic functional groups but lacking sulfur atoms only transiently reside within the protein and spontaneously dissociate to the solvent.

4.2 Background

We have previously reported [102] an open-state model of RGS4 (shown as Model 1 in Fig. 4.1d) which is conformationally different from apo-RGS4 (shown as Model 1 in Fig. 4.1c). In this model, the flexibility in the $\alpha 5$ - $\alpha 6$ helical pair facilitates access to the buried C95 residue for covalent-docking of the inhibitor CCG-50014 (compound **1** in Fig. 4.1a). On

inhibitor binding, we found that the the $\alpha 5$ - $\alpha 6$ helical pair remains perturbed and only partially relaxes toward the closed conformation of these helices in the apo-RGS4 structure. However, in the absence of inhibitor we observed that the protein largely reverts to a confor-

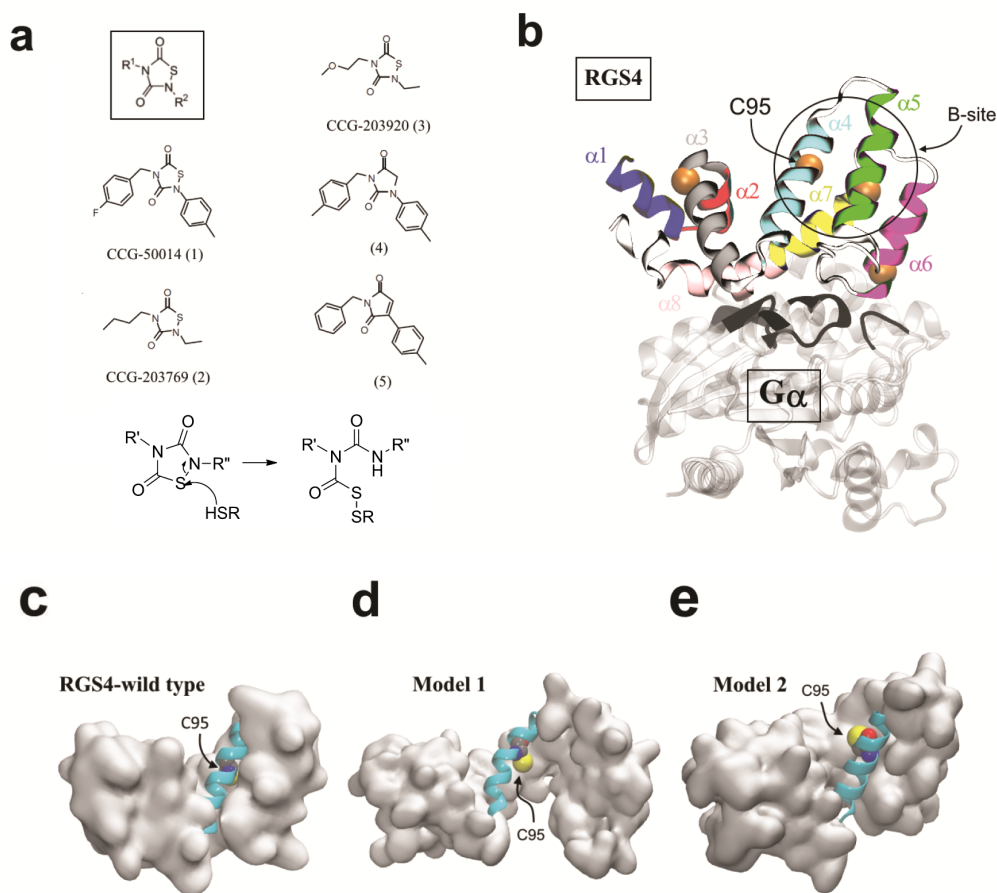


Figure 4.1. (a) Small-molecule structure with R^1 and R^2 functional groups (inset in box) along with the chemical structures of all small-molecules studied and the mechanism of reaction to form an adduct with a cysteine residue on the RGS4 [4]. (b) A cartoon representation of the RGS4- $G\alpha_{i1}$ complex (PDB code 1AGR) is shown. Each of the 9 α -helices of RGS4 is colored and labeled, and the location of four cysteine residues are shown by their C_α -atoms as orange spheres with the C95 residue labeled. The $G\alpha_{i1}$ -subunit is shown in transparent white ribbons, and the loops of $G\alpha_{i1}$ in the proximity of RGS4 are highlighted in black ribbons. (c, d, e) Cartoon representations of RGS4 conformations are shown for the wild-type apo-RGS4, and in its conformationally changed models (Models 1 and 2), respectively. Highlighted in cartoon representations are $\alpha 4$ -helices as cyan cartoons along with the C95 residues as space-filling. The residue C95 is buried in the wild-type RGS4 structure but it is accessible in Models 1 and 2. For each model, the structure of RGS4 (except the $\alpha 4$ -helix) is rendered in a white surface representation.

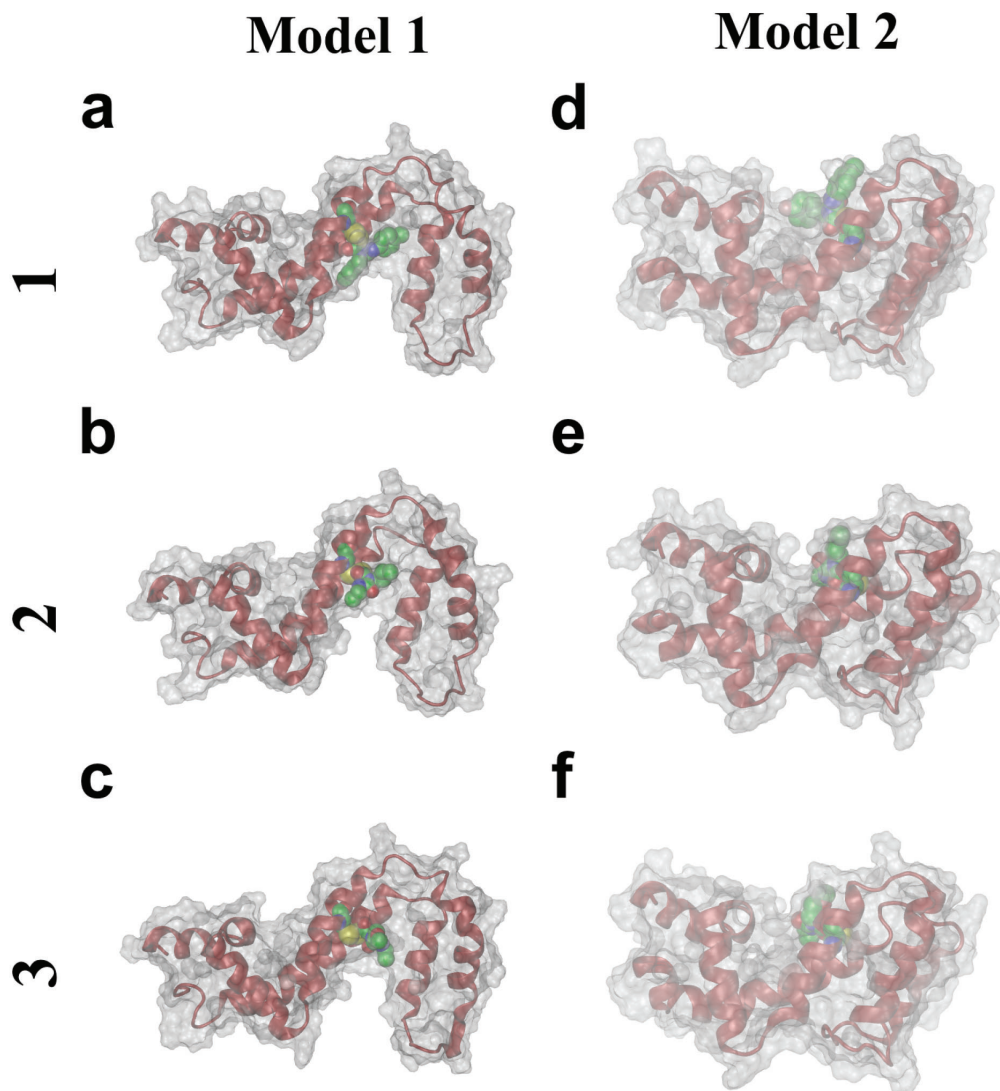


Figure 4.2. Docked initial conformations of TDZD compounds **1**, **2**, and **3** in Model 1 (panels a, b, and c) and Model 2 (panels d, e, and f) are shown. In all snapshots, the protein backbone is shown in red ribbons as well as white transparent surfaces, while compounds, along with the cysteine residue C95, are shown in green space-filling representations.

mation similar to apo-RGS4. While we conducted only short time-scale (~ 40 ns) molecular dynamics (MD) simulations in that work, parallel NMR HSQC results [15] identified several significant perturbations in residues surrounding the inhibitor binding site. Given the short time-scale of our earlier simulations, it is unclear to what extent the $\alpha 5$ - $\alpha 6$ helical pair will remain perturbed by bound ligand in longer time-scale simulations. Moreover, differences

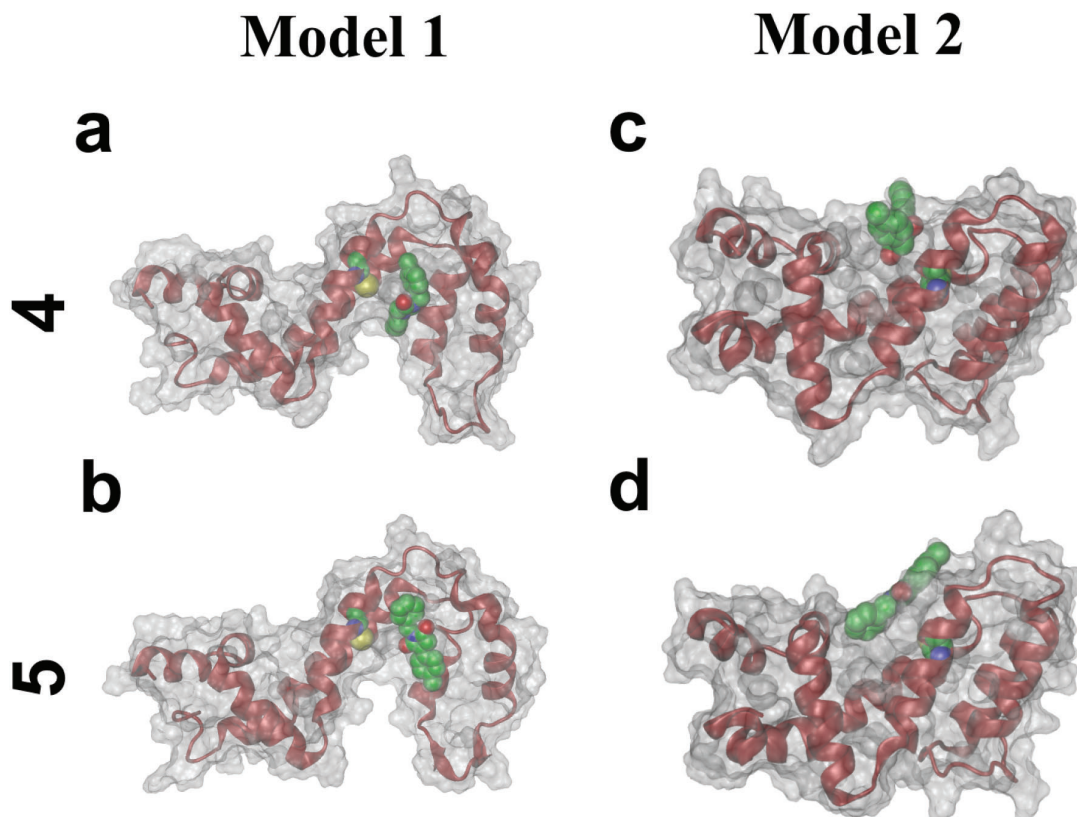


Figure 4.3. Docked initial conformations of non-TDZD compounds **4** and **5** in Model 1 (panels a and b) and Model 2 (panels c and d) are shown. Coloring and labeling schemes are identical to Fig. 4.2.

in the degree of helical perturbations by CCG-50014 and its congeners with smaller side-chains [4] (Fig. 4.1a) remain unknown as those small-molecules were not studied previously in the context of our open-state model (Model 1).

Furthermore, using μ s time-scale MD simulations combined with hydrogen-deuterium exchange (HDX) studies of three RGS proteins (RGS4 and its homologues RGS8 and RGS19) [13, 202], we not only observed signatures of flexibility and partial unfolding in helices, but also found that the side-chains of buried cysteine residues (C95 in RGS4) are transiently exposed to solvent while apo proteins largely maintain a closed conformation. In our previous work [13], we did not carry out simulation studies of RGS proteins with inhibitors, but we

hypothesized that inhibitors could potentially access the otherwise buried C95 residue in the cysteine-exposed closed conformation of RGS4 (Model 2 in Fig. 4.1e).

Hypothesis: Binding of aromatic and aliphatic TDZD analogues to buried and conserved cysteine residue results in allosteric perturbations.

To determine whether ligand access to the buried cysteine residue (C95) was through large conformational changes (Model 1) or through small local fluctuations in the side-chain of C95 (Model 2), we here performed enhanced MD investigations of the mechanisms of interactions of 5 small-molecules (3 TDZD compounds with aromatic/aliphatic functional groups and 2 non-TDZD analogues lacking sulfur atoms; Fig. 4.1a) with the C95 residue of RGS4. We further performed a flow cytometry-based assay [72] to measure concentrations of three TDZD compounds needed to inhibit the RGS4/ $G\alpha$ protein-protein interaction (specifically, the interaction between $G\alpha$ and an RGS4 construct containing only a single-cysteine residue, RGS4 C95). Taken together, these studies highlight the role of cysteine exposure and global protein dynamics in recognition of TDZD compounds by RGS4, and suggest new venues for designing non-covalent small-molecules targeting RGS proteins.

4.3 Methods

MD Simulations

We performed two sets of classical all-atom and explicit-solvent MD simulations for RGS4/small-molecule complexes using the NAMD software [203] and the CHARMM force-field with the CMAP correction [122, 123] (section 2.2). We used VMD for system creation and post-simulation analysis [204]. All small-molecules used in this work are reported in Fig. 4.1a, and their force-fields were parameterized using the Multipurpose Atom-Typer for CHARMM (MATCH) tool [125]. The first set of simulations was for studying small-

molecule analogues in complex with an open-state conformation of RGS4 (Model 1; Fig. 4.1d) [102], and the second set of simulations was for small-molecule analogues in complex with a closed-state conformation of RGS4 (Model 2; Fig. 4.1e) [13].

For simulations of Model 1, the open-state conformation of RGS4 that was reported in our earlier work [102] was used here as an initial state for docking of small-molecules. For specifically understanding the effect of binding of thiadiazolidinone (TDZD) and non-TDZD compounds in a binding pocket in the proximity of the cysteine residue C95 (Fig. 4.1b), a single-cysteine mutant of RGS4 was created, where all cysteine residues (three in total) except C95 were mutated to Ala. Following our previous protocol [102], the open-state con-

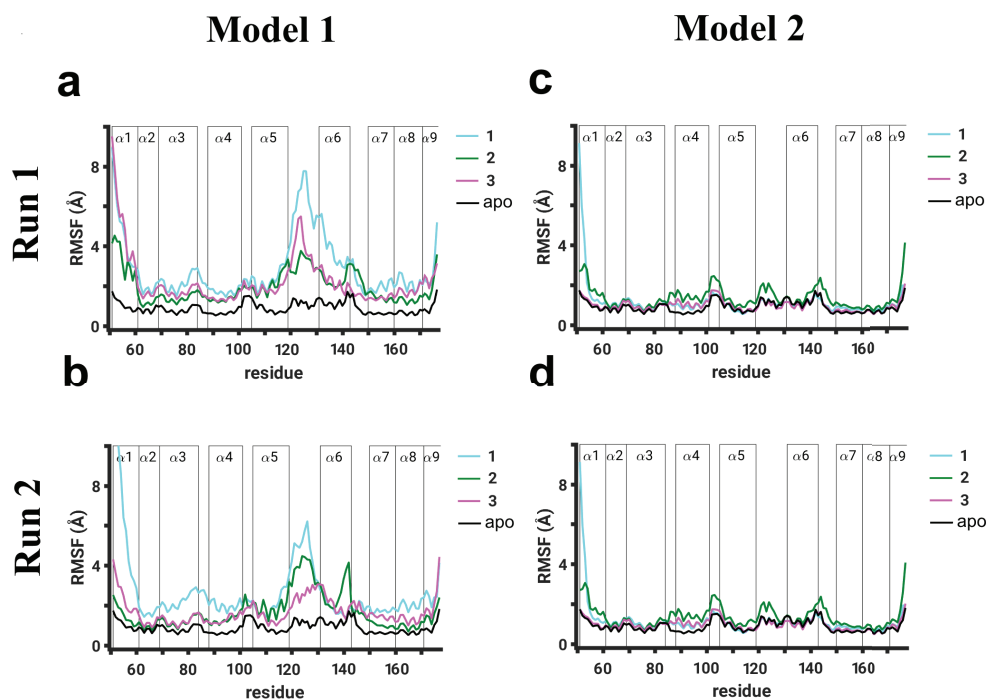


Figure 4.4. Root-mean-squared-fluctuation (RMSF) per residue are shown for Model 1 (panels a and b) and Model 2 (panels c and d) of RGS4. The RMSF values are reported from two independent 1 μ s long simulation runs (Run 1 and Run 2) for each model, where simulations were conducted with TDZD analogues (compounds **1**, **2**, and **3** in Fig. 4.1a) covalently-bound to the C95 residue of RGS4. As a baseline reference, the RMSF values of the RGS4 structure without any compound (apo-form; black traces) are also shown from our previous work [13]. The vertical bars labeled $\alpha 1$ through $\alpha 9$ demarcate the locations of residues in 9 α -helices of RGS4.

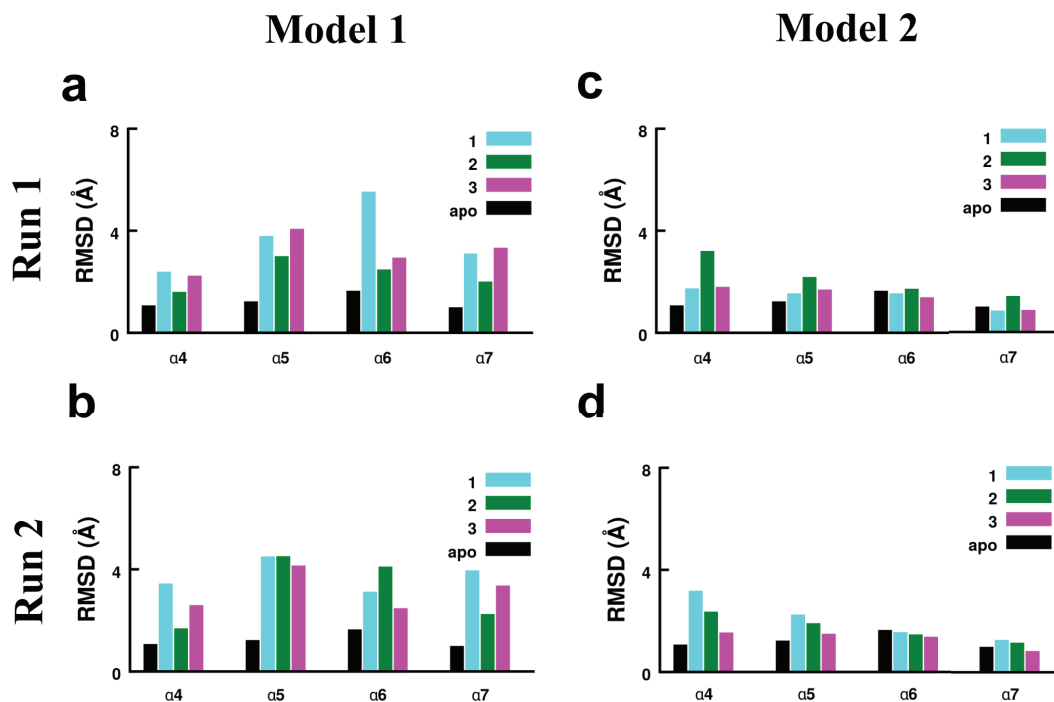


Figure 4.5. The histograms of RMSD-averages computed based upon Models 1 and 2 are shown. Panels a and b show data from two independent runs of Model 1, and panels c and d show data from two independent simulations of Model 2.

formation of mutated RGS4 was then used to create 5 docked complexes with 5 compounds (Fig. 4.1), where each TDZD compound is covalently bound to C95 and non-TDZD compounds are non-covalently docked in the same pocket where TDZD compounds are covalently docked (Fig. 4.2a, b, c, and Fig. 4.3a, b). For simulations of Model 2, a protocol similar to Model 1 was followed where the initial state of RGS4 was a closed-conformation in which the key cysteine residue (C95) is surface-exposed (Fig. 4.1e) per our earlier work [13]. The docking of compounds in Model 2 was further facilitated by the Internal Coordinate Mechanics (ICM) software [113] to obtain their energetically favorable conformations (Fig. 4.2d, e, f, and Fig. 4.3c, d).

We solvated all systems for Model 1 and Model 2 using explicit TIP3P water molecules [205], ionized using NaCl, and added all hydrogen atoms. Each solvated and ionized system was

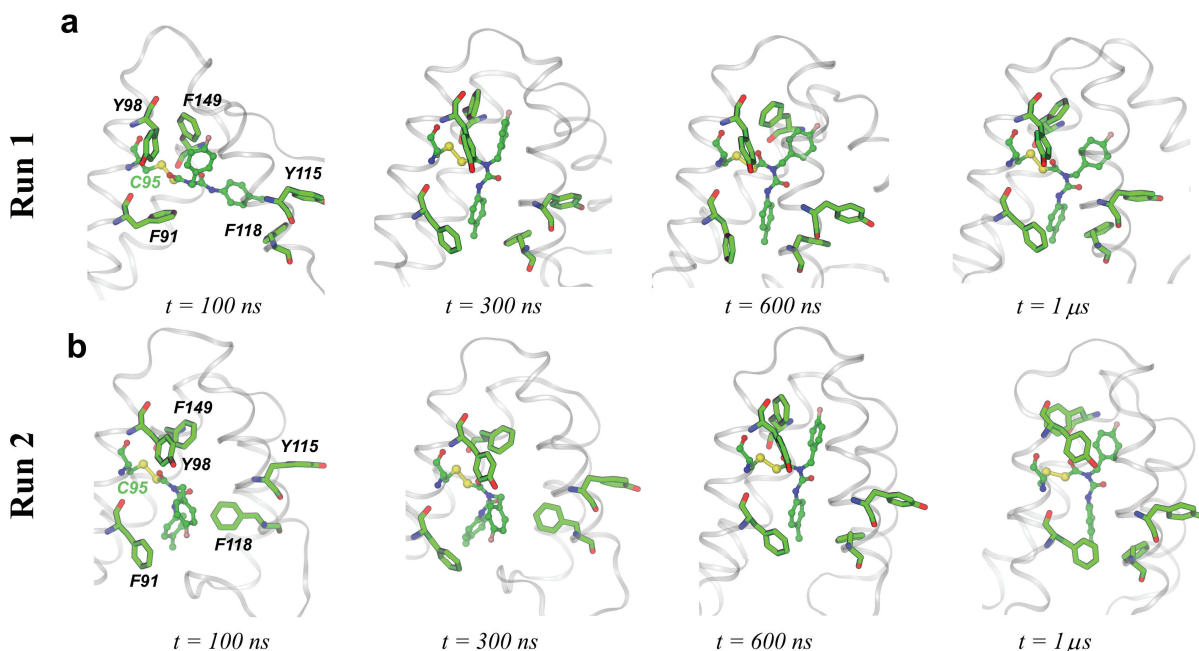


Figure 4.6. The side-chains of aromatic residues in the vicinity of covalently-docked compound **1** are shown at various time-points from two independent simulations of Model 1 (panels a and b). The compound **1** is covalently-linked to residue C95, and neighboring residues are labeled and shown in green sticks. The protein backbone in all snapshots is shown in a white transparent cartoon.

then energy minimized for ~ 500 - 1000 cycles via conjugate-gradient optimization, and equilibrated via MD simulations, conducted with a time-step (Δt) of 2-fs, for $1 \mu\text{s}$ in the NPT ensemble where the Langevin thermostat with a damping coefficient of 5 ps^{-1} was used for temperature control and the Nosé-Hoover barostat was used for pressure control. Periodic boundary conditions were used throughout, non-bonded interactions were accounted with a cut-off of 10 \AA where smooth switching was initiated at 8 \AA , and long-range electrostatic interactions were handled using the Particle Mesh Ewald (PME) method. For each system, two (for TDZD compounds) or three (for non-TDZD compounds) independent MD simulations were carried out. In addition to classical MD simulations, we used metadynamics as an enhanced sampling method (section 2.4.2) to compare thermodynamics of conformational changes between open and closed-states in apo-RGS4 and RGS4/small-molecule complexes.

For metadynamics simulations, we used an eigenvector as a CV [206] that was computed based upon the atomic coordinates of the backbone C_α atoms of all residues in α_4 through α_7 helices. The eigenvector choice of a CV is a projection of the coordinates of a group of atoms (or more precisely, their deviations from the reference coordinates) onto a linear transformation between two end-states: open and closed conformations of RGS4. The vector is normalized, therefore the CV is 0 when the backbone C_α atoms are at the coordinates of the closed conformation of RGS4 and the CV is 1 when they are at the coordinates of the open conformation of RGS4. All metadynamics simulations were carried out with a time-step of 2-fs, and with W , δ , and τ_G values of 0.05 kcal/mol, 0.025 Å, and 2 ps, respectively. Overall, we carried out 7 metadynamics simulations: one simulation for apo-RGS4 ($\sim 0.25 \mu\text{s}$ long), and 3 simulations ($\sim 0.15\text{--}0.45 \mu\text{s}$ long) each for Models 1 and 2 when bound to compounds **1**, **2**, and **3** (Fig. 4.1a), respectively.

4.4 Results and Discussion

To understand the effect of binding of small-molecules in each conformational model of RGS4 (Model 1 and Model 2; Fig. 4.1d,e), we docked TDZD as well as non-TDZD compounds (Fig. 4.2 and Fig. 4.3) in binding pockets created near the C95 residue and conducted several independent μs time-scale MD simulations. The initial conformational state of RGS4 for docking was chosen to either represent flexibility in helices (open-state; Model 1) [102] or cysteine-exposure (closed-state; Model 2) [13]. For each model, we performed two sets of independent simulations (each 1 μs long) for TDZD congeners having aromatic and aliphatic side-chains (compounds **1**, **2**, and **3** in Fig. 4.1a) covalently-docked at the residue C95 on the α_4 helix (highlighted in cyan in Fig. 4.1d,e). We first describe results on various conformational metrics used to characterize the subsequent structural perturbations by three covalently-linking TDZD compounds in each model of RGS4.

Additionally, for two non-covalent TDZD analogues lacking sulfur atoms (compounds **4** and **5** in Fig. 4.1a), we performed three sets of independent simulations (each 1 μ s long) for Model 1 where we hypothesized that non-covalently-docked compounds can transiently reside within the protein due to local interactions. While for Model 2, where compounds **4** and **5** are non-covalently-docked on the protein surface near the exposed cysteine residue C95, our preliminary simulation analyses revealed that compounds **4** and **5** can quickly and spontaneously diffuse into the solvent (*vide infra*). Therefore, we did not perform long time-scale simulations of these two non-TDZD compounds for Model 2. In the following, we describe various conformational metrics used to characterize the subsequent structural perturbations in each model of RGS4.

We performed several μ s-timescale unbiased MD simulations of RGS4 in the open-state conformational model (Model 1) with three covalently-docked (at C95) TDZD congeners (having aliphatic and aromatic side-chains), and two non-TDZD analogues lacking sulfur atoms (Fig. 4.1a). As shown in Fig. 4.1b, the residue C95 is located on the α_4 helix (highlighted in cyan) surrounded by α_5 - α_7 helices (highlighted in green, magenta, and yellow, respectively), making it completely buried in the crystal structure of RGS4, but mostly accessible in the open-state model of RGS4 (Model 1 in Fig. 4.1c).

4.4.1 Simulations of covalently-bound TDZD inhibitors

Root-mean-squared-fluctuation/deviation (RMSF/RMSD) analyses: To resolve residue-level perturbations on binding of TDZD compounds (**1**, **2**, and **3**) in each model of RGS4, we calculated RMSF per residue from two independent sets of simulations for each model (cyan, green, and magenta traces in Fig. 4.4) and compared these values with RMSF per residue values of apo-RGS4 from our previous work (black traces in Fig. 4.4) [13]. In addition to higher fluctuations expected in free terminal helices (α_1 and α_9), we observed in both simulations of Model 1 (panels a and b in Fig. 4.4) that all compounds induced significant perturbations (\sim 4-8 Å higher than in apo-RGS4) in helices α_5 and α_6 , and in

the α_5 - α_6 interhelical loop which directly contacts the $G\alpha$ -subunit in the RGS4/ $G\alpha$ complex (Fig. 4.1b). Moreover, the RMSF values show that compound **1** (with aromatic functional groups) perturbs the α_5 - α_6 interhelical loop more than compounds **2** and **3** (with aliphatic functional groups) (cyan *vs.* green and magenta traces in Fig. 4.4a,b). The perturbations in this interhelical loop also appear to propagate to structural motifs flanking this loop, namely the C-terminus of the α_5 helix and/or the N-terminus of the α_6 helix. For compound **1** in comparison to compounds **2** and **3**, we also observed marginally higher perturbations in the α_3 - α_4 interhelical loop, another structural region of RGS4 that is known to directly contact the $G\alpha$ -subunit in the RGS4/ $G\alpha$ complex (Fig. 4.1b). However, in Model 2 (panels c and d in Fig. 4.4), where all three compounds are covalently-docked on the protein surface (Fig. 4.2), we observed no significant perturbations in any structural motif of RGS4 as the RMSF values are comparable to apo-RGS4.

We further report that the C_α -RMSD traces for helices α_4 through α_7 (Figs. F.1 and F.2) and their average RMSD from both simulations (Fig. 4.5), measured relative to the crystallographic conformation of apo-RGS4 (PDB: 1AGR), highlight that all compounds induce greater perturbations when covalently-docked in Model 1 in comparison to Model 2. Also, compounds with aromatic functional groups (compound **1**) induce larger perturbations than compounds with aliphatic functional groups (compound **2** and **3**). The fluctuations of residues located at the interaction site with the $G\alpha$ subunit was shown in our previous work to result in significant weakening of the RGS/ $G\alpha$ protein-protein interaction in the presence of compound **1** (CCG-50014) [102]. As also observed here, this was largely attributed to structural rearrangements of the α_5 - α_6 helical pair and the loop connecting them. Specifically, we observed significant allosteric perturbations in two residues (T124 and E126) of the α_5 - α_6 interhelical loop in which we previously reported perturbations on binding of compound **1** based upon NMR HSQC data [102]. We also observed ring-ring interactions (Fig. 4.6) between compound **1** and the side-chains of neighboring aromatic residues some

of which (e.g. F91) were reported as highly perturbed on binding of compound **1** in our previous work [102].

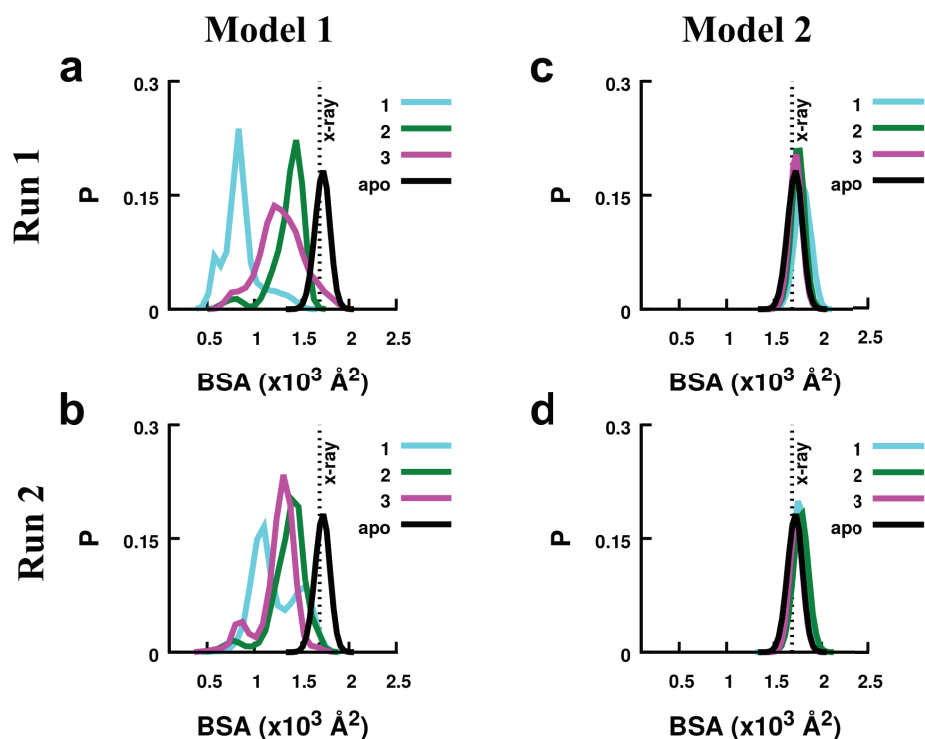


Figure 4.7. The histograms of the buried surface area (BSA) between the α_5 - α_6 helical pair and the rest of RGS4 are shown for Model 1 (panels a and b) and Model 2 (panels c and d). Data are shown for simulations of each model conducted with TDZD congeners (compounds **1**, **2**, and **3**). The vertical dotted lines in panels indicate the values of BSA in the RGS4 crystal structure (PDB: 1AGR). The BSA traces for apo-RGS4 computed from a simulation reported in our previous work [13] are also shown (black traces).

Buried Surface Area (BSA) analysis: To further investigate conformational changes on binding of compounds in each model of RGS4, we measured the BSA between the α_5 - α_6 helical pair and the rest of the RGS4 structure from two independent simulations (Fig. 4.7). In simulations of Model 1 (Fig. 4.7 a, b), we observed that compound **1** produces a greater shift in the peaks of the BSA-histograms than do compounds **2** and **3** compared to the BSA value in the crystal structure (vertical dotted lines labeled x-ray in Fig. 4.7) or in the apo-RGS4 simulation (black traces in Fig. 4.7 a, b). The BSA traces for all compounds in

Model 1 show deviations from the BSA values in the crystallographic or apo-RGS4 simulation indicating that the α_5 and α_6 helices only partially relax toward their closed conformation in the crystal structure. In both simulations of compound **2** with aliphatic functional groups, the BSA traces show a gradual increase in the BSA toward crystallographic values indicating a nearly complete closure of the α_5 - α_6 helical pair. In simulations of Model 2, the peaks of the BSA-histograms (Fig. 4.7 c, d) and the BSA traces show no significant deviation from the crystal structure values, thereby indicating a closed conformation of RGS4 when bound to TDZD compounds. Consistent with RMSD/RMSF trends, these results show that the binding of compounds in Model 1 perturbs the RGS4 structure significantly more than their binding in Model 2. Furthermore, in Model 1, the α_5 - α_6 helical pair, which is critical for RGS/ $G\alpha$ binding, only partially relaxes toward the crystallographic conformation in simulations of compound **1**, but significantly recovers in 1 μ s simulations of compound **2**. This suggests that the compounds with aliphatic functional groups (compounds **2** and **3**), when covalently-docked within the α_4 - α_7 helical bundle, are more easily accommodated than those with aromatic functional groups (compound **1**).

Salt-bridging interactions: The α_4 - α_7 helical bundle in RGS4 has several charged amino-acids (K, R, D, and E) that likely form stable or intermittent salt-bridges due to electrostatic interactions. To understand the ability of TDZD compounds to perturb interactions between charged residues and thereby between helices, we investigated potential perturbations in several salt-bridge forming residue pairs: D90-K125 (α_4 - α_5), E97-K110 (α_4 - α_5), K99-D150 (α_4 - α_7), E126-R134 (α_5 - α_6), and D130-K155 (α_6 - α_7). In Fig. 4.8, we show histograms of average distances between the center-of-mass of these residue pairs from simulations with compounds (cyan, green, magenta bars) and without compounds (black bars; apo-RGS4). On comparing these data for Model 1, we observed that all compounds perturb salt-bridges between the helical pairs α_4 - α_5 (D90-K125/E97-K110), α_5 - α_6 (E126-R134), and α_6 - α_7 (D130-K155), but only compound **1** perturbs the α_4 - α_7 salt-bridge (K99-D150) marginally higher than compounds **2** and **3**; perturbations by compounds **2** and **3** are compa-

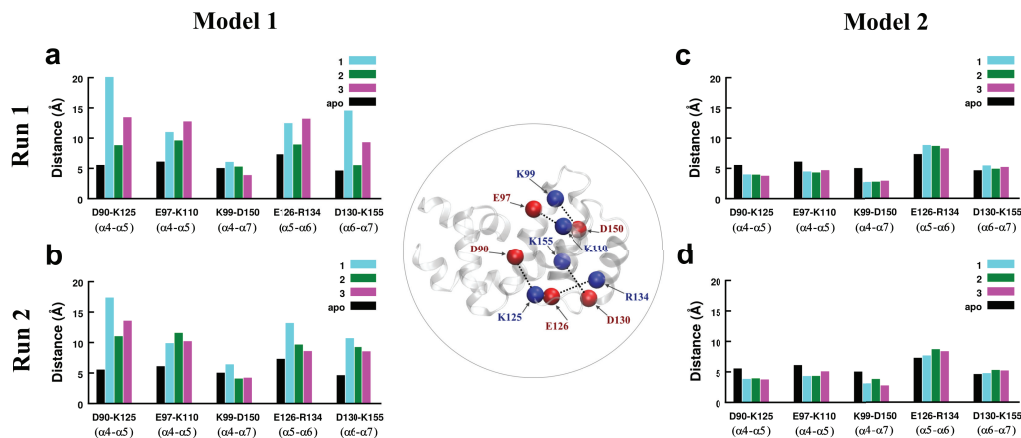


Figure 4.8. The histograms of average distances between the centers of mass of residues involved in five salt-bridge-forming residue pairs are shown from two independent simulations of Model 1 and Model 2 for three TDZD compounds. The data for an apo-RGS4 simulation from our previous work [13] are also shown (black histograms). The C_{α} -atoms of all residues involved in salt-bridges are shown and labeled as red/blue spheres on the RGS4 structure (inset in circle).

able to apo-RGS4. For salt-bridges between the helical pairs α_4 - α_5 (D90-K125) and α_6 - α_7 (D130-K155), compound **1** perturbs salt-bridges more than compounds **2** and **3** in both simulations of Model 1 (Fig. 4.8 a,b). The α_5 - α_6 salt-bridge (E126-R134) is also perturbed more by compound **1** in comparison to compound **2** in both simulations (Fig. 4.8 a,b), and in comparison to compound **3** in the second simulation (Fig. 4.8b). However, compound **2** or compound **3** could perturb one of the α_4 - α_5 salt-bridges (E97-K110) marginally more than compound **1** (Fig. 4.8a,b).

The data from two simulations of Model 2 (Fig. 4.8c,d) reveal no significant perturbations in these salt-bridging interactions although salt-bridges between the helical pairs α_4 - α_5 (D90-K125/E97-K110) and α_4 - α_7 (K99-D150) are marginally stabilized in comparison to apo-RGS4. These data suggest that in Model 1 compounds with aromatic functional groups destabilize interhelical salt-bridging interactions more than the compounds with aliphatic functional groups. Moreover, the perturbations are allosteric since two salt-bridges significantly perturbed by compound **1** (D90-K125 and D130-K155) have residues K125 and D130

that are located in the α_5 - α_6 interhelical loop, away from the docking site residue (C95). As highlighted above, this loop directly participates in the protein-protein interaction between RGS4 and the $G\alpha$ -subunit (Fig. 4.1b). Overall, the mode of binding of TDZD compounds in Model 2 resulted in largely insignificant perturbations in the RGS4 structure in comparison to binding of these compounds in Model 1.

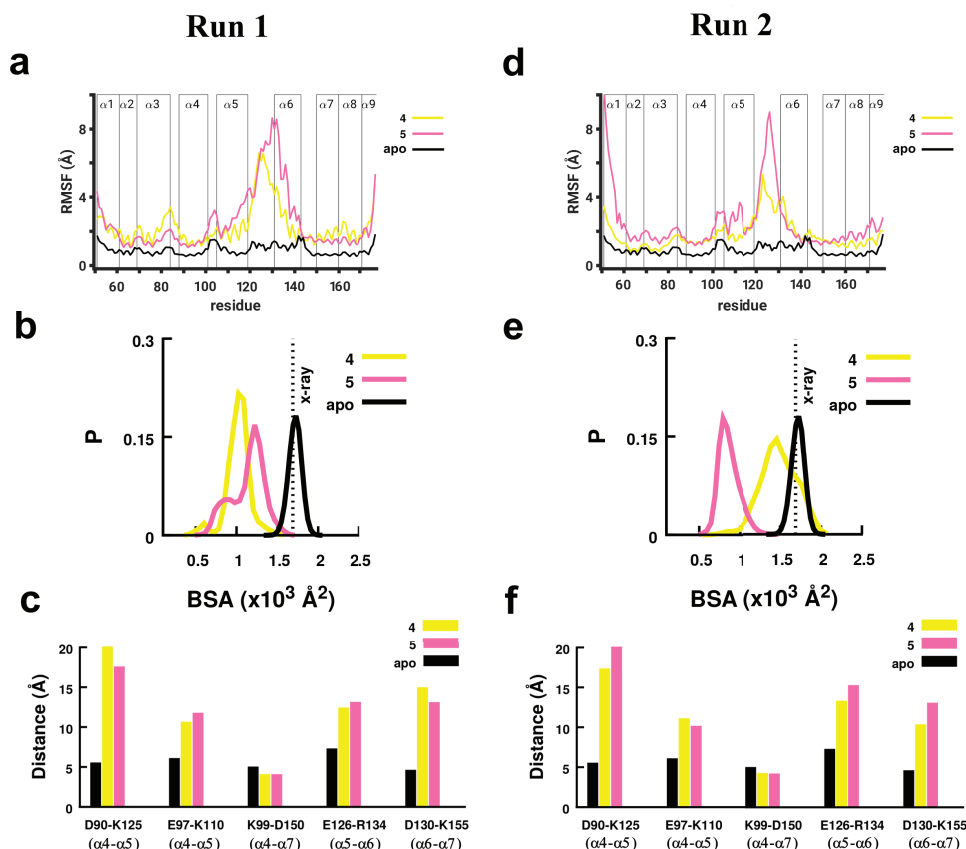


Figure 4.9. The data from RMSF (panel a/d), BSA (panel b/e), and salt-bridging interactions (panel c/f) are shown from two simulations of each non-TDZD compound in Model 1 (compound 4, yellow trace; compound 5, magenta trace). Other details in panels a/d, b/e, and c/f are similar to Figs. 4.4, 4.7, and 4.8, respectively.

4.4.2 Simulations of non-covalent (non-TDZD) ligands 4 and 5

We further studied compounds 4 and 5 that are analogues of compound 1 with aromatic functional groups but lacking sulfur atoms (Fig. 4.1a). While compound 1 is a known

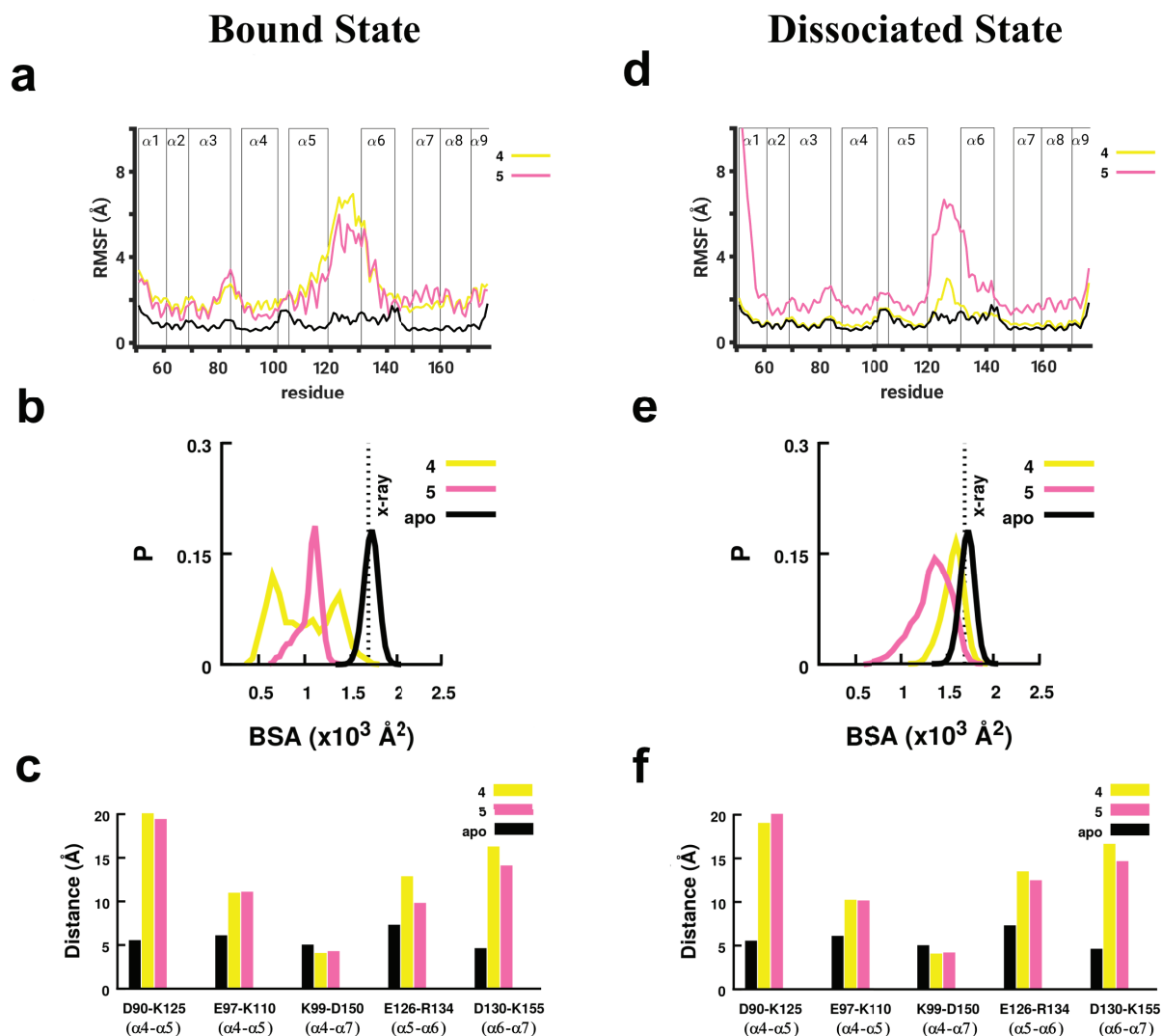


Figure 4.10: Data similar to those presented in Fig. 4.9 are shown from Run3 for compounds **4** and **5** in which diffusion of each compound out of the protein pocket was observed. The left-panels show data for parts of trajectories when compounds still reside within the protein, and the right-panels show data for the remaining parts of trajectories when compounds have diffused out of the pocket.

inhibitor of the wild-type RGS4/ $G\alpha$ protein-protein interaction, compounds **4** and **5** do not inhibit this interaction [4]. The mechanistic basis of this observation remains unknown. To test the ability of compounds **4** and **5** to perturb the RSG4 structure, we conducted MD simulations after docking both compounds in each model of RGS4 (Fig. 4.2 and Fig. 4.3). For

Model 1, we conducted three independent 1 μ s long MD simulations of each compound and observed that in two out of three simulations, these compounds reside within the α_4 - α_7 helical bundle throughout 1 μ s, but in one simulation, each compound diffuses out of the helical bundle into the solvent after transiently residing within the helices. For Model 2 as well, we conducted three independent simulations of each non-TDZD compound. From these simulations, we observed non-specific interactions of compounds (with residues on the protein surface in the vicinity of the docking site) that lead to their rapid dissociation into the solvent. We discontinued these trajectories after the dissociation of each compound. We therefore present analyses from three trajectories of each non-TDZD compound for Model 1 where compounds can transiently reside within the protein.

The data from RMSF, BSA, and salt-bridge measurements from the first two simulations of each compound in Model 1 (where compounds remain bound) are presented in Fig. 4.9 and from the third simulations (where compounds dissociate) are shown in Fig. 4.10. The RMSF data (Fig. 4.9 a,d and Fig. 4.10a) show that both compounds in their transiently bound states perturb the α_5 and α_6 helices as well as the α_5 - α_6 interhelical loop, and the perturbations by compound 5 are marginally higher than compound 4. The peaks of the BSA histograms (Fig. 4.9 b,e and Fig. 4.10b). for each compound are shifted away from the values in the crystal structure or in the apo-RGS4 simulation, thereby indicating an open conformation of these helices. Accordingly, 4 of 5 interhelical salt-bridges (D90-K125, E97-K110, E126-R134, and D130-K155) are significantly perturbed, while the α_4 - α_7 (K99-D150) is marginally stabilized (Fig. 4.9 c,f and Fig. 4.10c). This is consistent with the observation that in two simulations of Model 1, both non-TDZD compounds reside within the α_4 - α_7 helical bundle. On dissociation of compound 4 in the third simulation, we observed decreased fluctuations in the α_5 - α_6 interhelical loop and a shift of the BSA-histogram peak toward the crystallographic or apo-RGS4 values (yellow traces in Fig. 4.10 d,e *vs.* Fig. 4.10 a,b). After compound 5 moved out of helices, it continued to interact with the protein surface thereby perturbing the α_5 - α_6 interhelical loop and neighboring helices (magenta traces in

Fig. 4.10 d,e). The compounds continue to perturb salt-bridging interactions (Fig. 4.10 c,f) so long as they reside within the helices or near the protein surface on dissociation from the pockets.

In each model, simulations of the non-TDZD compounds (**4** and **5**) show spontaneous dissociation, consistent with the hypothesis that these two compounds do not reside inside RGS4 for longer time-scales, as they cannot covalently bind to the key cysteine residue (C95). However, despite the transient residence of non-TDZD compounds within the protein domain and their subsequent spontaneous dissociation to the solvent, their ability to perturb the RGS4 structure while bound suggests the potential of non-covalent compounds (possibly with higher binding affinities) as promising candidates for developing the next generation of RGS inhibitors.

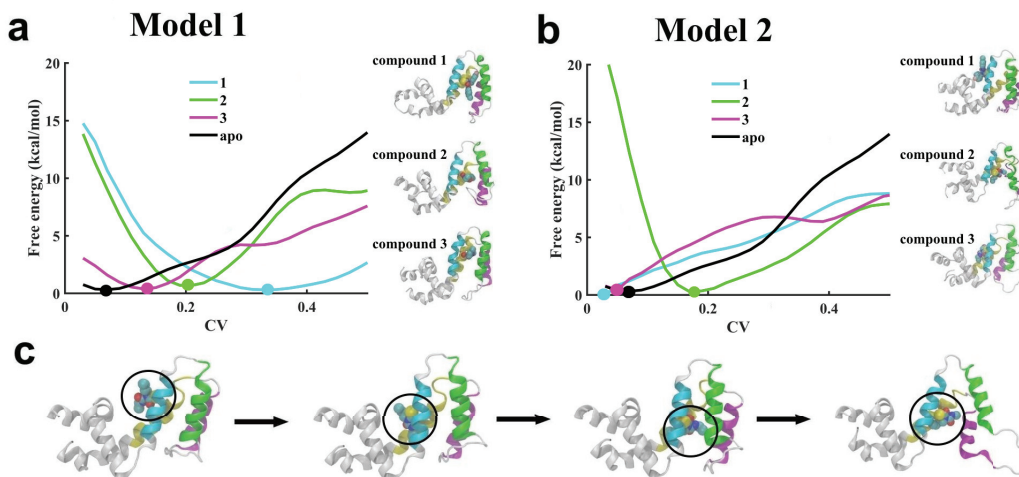


Figure 4.11: (a and b) Free energy profiles are plotted against the collective variable (CV) for structural transitions (between open and closed states) in RGS4 when three TDZD compounds are docked in distinct pockets created in Models 1 and 2. For each model, thermodynamically favorable conformations of RGS4 bound to TDZD compounds are also shown as cartoons in panels a and b. (c) For compound **2** in Model 2, highlighted as cartoons are conformations of RGS4 showing spontaneous diffusion of compound **2** (CCG-203769; Fig. 4.1a) from its initially-docked position on the protein surface to within the α_4 - α_7 helical bundle. The circle in panel c denotes the combined location of covalently-linked residue C95 and compound **2**. For all panels, the protein backbone in snapshots is depicted in white ribbons except helices α_4 through α_7 that are uniquely colored as in Fig. 4.1b (This figure was created by my co-author Hossein Mohammadiarani [14]).

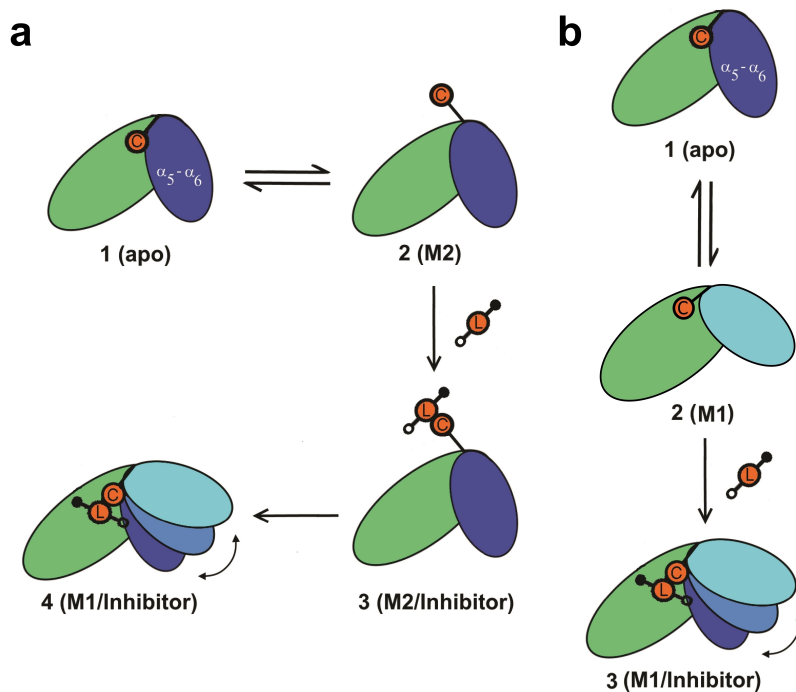


Figure 4.12: (a) A schematic highlighting the proposed mechanism of binding of TDZD small-molecules to RGS4 is shown. In this scheme, the exposure of C95 (orange circle labeled **C**) in the apo-RGS4 conformation (panels 1 and 2) allows initial covalent recognition (panel 3) of small-molecules (orange circle labeled **L** flanked by filled/empty circles indicating R^1 and R^2 functional groups) and a subsequent migration of compounds to the core of the α_4 - α_7 helical bundle causing allosteric structural perturbations in helices (panel 4), especially in residues in the RGS/ $G\alpha$ protein-protein interface. (b) Previously proposed mechanism [15] for the exposure of C95 in the apo-RGS4 conformation (panels 1 and 2) and subsequent binding of compounds causing allosteric structural perturbations (panel 3).

4.4.3 Thermodynamic analyses of apo-RGS4 and RGS4/TDZD complexes

To understand thermodynamics of conformational changes in RGS4 in the apo-form as well as when bound to TDZD small-molecules (compounds **1**, **2**, and **3** in Fig. 4.1a), we carried out seven independent enhanced sampling MD simulations using the metadynamics method [153] (see methods). Specifically, we resolved free-energy profiles for structural transitions between the open and closed-states in apo-RGS4 and in Models 1 and 2 of RGS4 when bound to the three TDZD compounds (Fig. 4.11a and 4.11b). The free-energy profiles were resolved in a multidimensional projection of atomic Cartesian coordinates on an eigenvector reaction coordinate (also referred to as a collective variable, CV) such that the CV spans a range between 0 and 1 for closed and open conformations, respectively. The reference conformational states defining the CV were chosen to represent the apo-RGS4 crystal structure (closed conformation of RGS4) and the conformation reported in our previous work (open conformation of RGS4) [102].

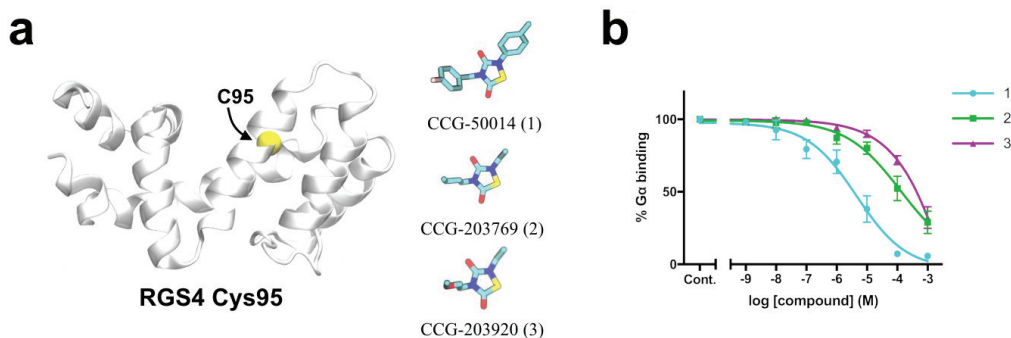


Figure 4.13: (a) The structures of the single-cysteine RGS4 construct (RGS4 C95) and three compounds used in the flow cytometry protein-protein interaction assay are shown. (b) Inhibition of the RGS4 C95/ $G\alpha$ protein-protein interaction by compounds **1**, **2** and **3** over a range of concentrations is shown (This figure is provided by my co-author Vincent Shaw [14]).

We observed that the free-energy profile for a structural transition from a closed to open-state in apo-RGS4 (black trace in Fig. 4.11a,b) has a global minimum (marked by a black filled circle in Fig. 4.11a,b) very close to the initial value of $CV = 0$, which indicates that RGS4 in its apo-state thermodynamically favors a closed-state, as also seen in the crystal structure. On comparing free-energy profiles of protein bound to compound **1** (cyan traces in Fig. 4.11a,b), we observed that the global minimum in Model 1 (marked by a cyan filled circle in Fig. 4.11a) is located at a CV value of ~ 0.35 , while in Model 2 the global minimum (marked by a cyan filled circle in Fig. 4.11b) is located very close to zero. This suggests that open-like RGS4 states are thermodynamically favored when compound **1** is docked in the binding pocket in Model 1 in comparison to Model 2, where thermodynamically favored conformations are similar to the closed-state, as in apo-RGS4.

However, from the free-energy profiles for compound **2** in both models (green traces in Fig. 4.11a,b), we observed that the global minimum (marked by a green filled circle in Fig. 4.11a,b) is located in the vicinity of a CV value of ~ 0.20 . This suggests that open-like RGS4 states likely exist for compound **2** not only in Model 1 but also in Model 2. The ability of compound **2** to stabilize open-like conformations in Model 2 is explained by the observation that the compound **2** spontaneously diffuses during the metadynamics simulation from its initially-docked position on the protein surface to its final position within the α_4 - α_7 helical bundle (snapshots in Fig. 4.11c), thereby acquiring conformations similar to Model 1. Importantly, the diffusion of compound **2** from the protein surface to within this helical bundle is driven both by the flexibilities in helices and burial of the side-chain of key cysteine residue C95.

For compound **3**, the location of the global minimum in each free-energy profile (marked by a magenta filled circle in Fig. 4.11a,b) indicates a conformational stability behavior similar to compound **1** in that the open-like RGS4 states are thermodynamically favored in Model 1 and a closed state is favored in Model 2. However, the global minimum for compound **3** in Model 1 is located at a CV value of ~ 0.14 smaller than the CV value of ~ 0.35 for the global

minimum for compound **1** (magenta *vs.* cyan filled circles in Fig. 4.11a). This suggests that perturbations to the RGS4 structure (relative to the closed state) are smaller for compound **3** (with aliphatic functional groups) than for compound **1** (with aromatic functional groups). Overall, the shift in the free-energy minimum from the higher to lower CV values on binding of TDZD compounds in Model 1 (CV = 0.35, 0.20 and 0.14 for compounds **1**, **2**, and **3**, respectively) is consistent with the perturbation trends observed in classical MD simulations (Fig. 4.4, 4.7, 4.8).

The possibility of binding of TDZD analogues (**1**, **2**, and **3**) to C95 in a closed-state conformation of RGS4 due to a transient exposure of C95 (Model 2) [13] shows that multiple binding mechanisms may exist by which these small-molecules can access the otherwise buried cysteine residue C95. In Model 2, all compounds are docked on the protein surface located near the "B-site" of RGS4 [65], entirely outside of the α_4 - α_7 helical bundle. Therefore, the initial binding of TDZD analogues to C95 is not dependent on a significant opening of the α_5 - α_6 helical pair. This mechanism of small-molecule recognition by C95 is distinct from our previously suggested mechanism [102] that highlighted the flexibility in the α_5 - α_6 helical pair as a potential route for compound binding. However, our enhanced sampling metadynamics simulations showed that compounds covalently-bound to the exposed C95 residue in Model 2 (e.g. compound **2**) could translocate from the protein surface to the core of the α_4 - α_7 helical bundle (Fig.4.11c) and stabilize open-like states similar to Model 1 while remaining covalently-bound, thereby suggesting that Model 2 can evolve toward Model 1 (Fig. 4.12). Therefore, we propose that these mechanisms of binding of compounds may not be mutually exclusive. This substantiates the importance of coupling between local conformational flexibilities in protein side-chains (e.g. exposure of C95) with global protein dynamics (e.g. flexibilities in RGS4 helices) to facilitate small-molecule recognition and allow allosteric inhibition of the protein-protein interface.

4.4.4 Functional analysis of the inhibition of the RGS4-C95/ $G\alpha$ protein-protein interaction by TDZD compounds

To further investigate the effect of binding of TDZD compounds **1**, **2** and **3** on the interaction of single-cysteine RGS4 (RGS4 C95) and $G\alpha$ *in vitro*, we utilized a flow cytometry protein interaction assay (FCPIA) (see methods) [72]. A single-cysteine (RGS4 C95) mutant was used to limit compound binding to C95, such that the results reflect compound action at the same cysteine to which compound was covalently docked in simulations. The concentration-response curves (Fig. 4.13) show that the single-cysteine protein is inhibited by compound **1** with an IC_{50} value of $\sim 4.5 \mu\text{M}$, and less inhibited by compounds **2** and **3** (Fig. 4.13b). The magnitude of structural perturbations and deviations from the native RGS4 conformation (Fig. 4.4, 4.7, 4.8, and 4.11) induced by the compounds in modeling and simulation studies correlate well with the results of inhibition experiments. The use of single-cysteine mutants resulted in lower potencies of inhibition by each compound compared to wild-type proteins [4]. Notably, the difference in potencies between aromatic compound (1) and aliphatic compounds (2 and 3) is more pronounced in single-cysteine RGS4 C95 than in WT RGS4. This may be because in WT proteins, differences in dynamics between compounds bound to C95 are masked by the action of compounds at other cysteines. Compound **1** causes the greatest perturbations in the RGS4 structure and also showed the greatest potency to inhibit RGS4/ $G\alpha$ binding. Compounds **2** and **3** caused smaller perturbations in the RGS4 structure and accordingly showed a reduced ability to inhibit RGS4/ $G\alpha$ binding.

4.5 Conclusion

In this chapter, I have presented modeling and simulation studies predicting structural perturbations in the RGS4 protein on binding to various small-molecules. We found that compounds with aromatic functional groups significantly perturb the protein structure in com-

parison to those with the aliphatic functional groups. Non-covalent compounds only transiently perturb the protein as these compounds spontaneously dissociate to solvent. Thermodynamic analyses of RGS4/small-molecule complexes suggest that two distinct modes of binding can lead to a two-step mechanism in which compounds are initially recognized by the exposed side-chains of conserved and buried cysteine residues (C95 in RGS4) followed by their migration to the helical core of RGS4 that leads to significant allosteric perturbations in those RGS4 residues that are located in the RGS4/G α protein-protein interface. These findings will inform future drug development efforts focused on the discovery of non-covalent compounds capable of inducing similar allosteric perturbations.

4.6 Publications

The work described in this chapter has resulted in the following journal article:

- Mohammadi, M., Mohammadiarani, H., Shaw, V. S., Neubig, R. R., Vashisth, H. (2019). Interplay of cysteine exposure and global protein dynamics in small-molecule recognition by a regulator of G-protein signaling protein. *Proteins: Structure, Function, and Bioinformatics*, 87(2), 146-156.

The work has also appeared on journal cover [17]. The cover image is included in Appendix I.

CHAPTER 5

STUDIES ON COUPLING OF PROTEIN DYNAMICS AND SALT-BRIDGING INTERACTIONS IN RGS PROTEINS

5.1 Abstract

RGS proteins modulate receptor signaling by binding to activated G-protein α -subunits, accelerating GTP hydrolysis. Selective inhibition of RGS proteins increases G-protein activity and may provide unique tissue specificity. Thiadiazolidinones (TDZDs) are covalent inhibitors that act on cysteine residues to inhibit RGS4, RGS8 and RGS19. There is a correlation between protein flexibility and potency of inhibition by the TDZD CCG-50014. In the context of a single conserved cysteine residue on the α 4 helix, RGS19 is the most flexible and most potently inhibited by CCG-50014, followed by RGS4 and RGS8. We hypothesized that interhelical salt-bridge forming residues are responsible for differences in both flexibility and potency of inhibition among RGS isoforms. RGS19 lacks a charged residue on the α 4 helix that is present in RGS4 and RGS8. Introducing a negative charge at this position (L118D) increased the thermal stability of RGS19 and decreased the potency of inhibition by CCG-50014. Mutations which eliminated salt bridge formation in RGS8 and RGS4 led to decreased thermal stability in RGS8 and increased potency of inhibition of both RGS4 and RGS8. Molecular dynamics (MD) simulations with an added salt bridge in RGS19 (L118D) showed reduced RGS19 flexibility. Hydrogen-deuterium exchange (HDX)* studies showed striking differences in flexibility in the α 4 helix of RGS4, 8, and 19 with salt bridge modifying mutations. These results show that an α 4 salt bridge-forming residue controls flexibility in several RGS isoforms and supports a causal relationship between RGS flexibility and the potency of TDZD inhibitors.

* The experimental works (mutagenesis, HDX, and compound potency studies) performed by my collaborator Vincent Shaw at Michigan State University.

5.2 Background

Drug specificity is often considered to be like a key fitting into a complementary shaped lock. It has become clear recently that protein dynamics can play an important role in drug discovery. RGS proteins bind to activated G_α subunits of G-proteins, thereby accelerating GTP hydrolysis and attenuating G-protein signaling. In regulating G-Protein Coupled Receptor (GPCR) signaling, RGS proteins play a role in the physiology of numerous systems. By inhibiting RGS proteins, signaling via a GPCR may be enhanced. There are twenty RGS isoforms, each with a different tissue distribution. Combination of GPCR agonists with inhibitors specific for a single RGS isoform should limit effects on GPCR signaling to a subset of target tissues that intersects with the distribution of the GPCR. This has the potential to reduce agonist off-target effects, and makes RGS proteins an attractive target for modulation of GPCR signaling.

RGS inhibitors discovered to date are covalent modifiers of cysteine residues and are selective for RGS4 and RGS1 [84, 207]. These have four and three cysteines, respectively, in the RGS homology domain, which is more than in most other RGS proteins. RGS4 has been linked to nervous system related disease states in which RGS4 inhibition may be desirable, including seizures [208] and Parkinson's disease [90, 91, 93]. Continued efforts to seek non-covalent inhibitors are attractive, because the lower risk associated with non-covalent inhibitors is considered safer and may ease further development [209]. In addition, it would be valuable to discover RGS inhibitors with other specificities since other RGS proteins less potently inhibited by covalent modifiers have been implicated as potential targets, including RGS17 in cancer [210, 211] and RGS19 in depression [212]. To identify noncovalent

inhibitors with novel specificities, it will be necessary to understand what factors apart from the number of cysteines in the RGS domain drive selectivity of RGS inhibitors.

The RGS homology domain contains nine helices. A cysteine residue on $\alpha 4$, which faces the interior of the $\alpha 4$ - $\alpha 7$ helical bundle, is conserved among all 20 RGS isoforms with the exception of RGS6 and RGS7 [100]. Interestingly, when RGS proteins are mutated to contain only this single, shared cysteine, there are still dramatic differences in potencies by which different isoforms are inhibited [13]. RGS19, which contains only the shared $\alpha 4$ cysteine, is more potently inhibited than single-cysteine versions of RGS4 and RGS8 [13, 202].

Hypothesis: Inhibitor potency can be modulated by salt-bridging interactions.

Previously, we found using MD simulations that RGS19 is more flexible than RGS4 and RGS8 [13]. In these modeling studies we also found that the extent of perturbations of salt bridge interactions by inhibitor compounds correlated with structural flexibility in RGS4 [14, 17]. In this work, we sought to identify the cause of flexibility differences among these isoforms and hypothesize that mutations aimed at salt bridge interactions that enhance RGS protein flexibility can increase the potency of inhibitors such as CCG-50014.

5.3 Methods

We performed two sets of classical all-atom and explicit-solvent MD simulations (section 2.2) for single-cysteine RGS4/ RGS4 D90L, single-cysteine RGS8/ RGS8 E84L, and RGS19/ RGS19 L118D using the NAMD software [111] and the CHARMM force-field with the CMAP correction [122, 123]. We used VMD for system creation and post-simulation analysis [110]. The initial coordinates were obtained from the protein data bank files with codes 1AGR

(RGS4), 2ODE (RGS8), and 1CMZ (RGS19). Except Cys95 in RGS4 and Cys107¹ RGS8, all cysteines were mutated to alanines. Each protein was then solvated in a simulation box of TIP3P water molecules [205] and charge-neutralized with NaCl. Each solvated and ionized system was energy minimized for ~ 500 -1000 cycles via conjugate-gradient optimization, then equilibrated via $1\mu\text{s}$ MD simulations conducted with a time-step (Δt) of 2 fs. The NPT ensemble with a Langevin thermostat and a damping coefficient of 5 ps^{-1} was used for temperature control and the Nosé-Hoover barostat was used for pressure control. Periodic boundary conditions were used throughout; non-bonded interactions were accounted for with a cut-off of 10 \AA where smooth switching was initiated at 8 \AA . Long-range electrostatic interactions were handled using the Particle Mesh Ewald (PME) method. The dynamic cross-correlation (DCC) maps were created, and salt-bridge interaction analysis were performed as described in section 2.3.2.

5.4 Results and Discussion

Comparison of the structures for RGS19 (PDB 1CMZ[213]), RGS4 (PDB 1AGR[101]), and RGS8 (PDB 2ODE [214]) shows that there are differing numbers of interhelical salt bridges among their $\alpha 4$ - $\alpha 7$ helix bundles. Some of these may contribute to differences in stability and dynamics among the RGS isoforms.

RGS19 has only one interhelical salt bridge in this bundle, between E125 ($\alpha 4$) and K138 ($\alpha 5$) (Fig. 5.1E). This salt bridge is well conserved among all three proteins, however, so it is unlikely to contribute to observed differences in flexibility [13]. RGS4 has this salt bridge (E97-K110, Fig. 5.1B) as well as two additional interhelical salt bridge locations, between D130 ($\alpha 6$) and K155 ($\alpha 7$) (Fig. 5.1D); and a salt bridge network that connects D90 ($\alpha 4$), K125 on the $\alpha 5$ - $\alpha 6$ interhelical loop, and E117 on $\alpha 5$ (Fig. 5.1B). RGS8 has four interhelical

¹Note, amino acid numbering follows that for RGS8, Isoform 1, NCBIRefSeqNP_203131.1

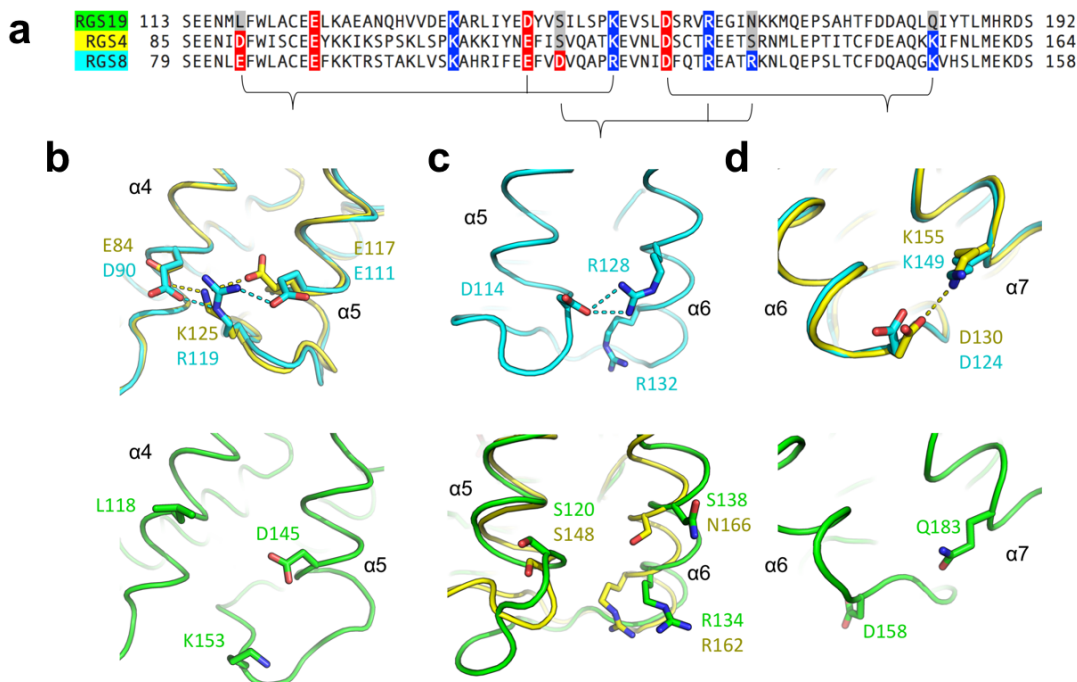


Figure 5.1. (a) Alignment of RGS19, RGS4, and RGS8 sequences in $\alpha4$ - $\alpha7$ helix bundle. Charged residues that make interhelical contacts are indicated in red and blue. Structural alignments of $\alpha4$ - $\alpha5$ (b and e), $\alpha5$ - $\alpha6$ (c and f), and $\alpha6$ - $\alpha7$ (d and g) helix pairs are shown, with highlighted residues in panel a rendered as sticks. RGS19 (PDB 1CMZ) is in green, RGS4 (PDB 1AGR) is in yellow, and RGS8 (PDB 2ODE) is in cyan (This figure was created by my co-author Vincent Shaw [16]).

salt bridges in the $\alpha4$ - $\alpha7$ bundle. One salt bridge (E91-K104, Fig. 5.1B) is shared by both RGS19 and RGS4. Two salt bridges (D124-K149, Fig. 5.1D, and E84-R119-E111, Fig. 5.1B) are shared with RGS4. Finally, a salt bridge between D114 ($\alpha5$) and R132 ($\alpha6$) is unique to RGS8 (Fig. 5.1C).

To estimate the relevance of each of these salt bridges in maintenance of helix bundle rigidity, the time each amino acid in a charged pair spent within a \AA of one another over the course of a long timescale MD simulation was measured [13]. The $\alpha6$ - $\alpha7$ salt bridge, which is present in RGS4 and RGS8 but absent in RGS19, was stably maintained. It occupied a salt bridge-forming distance for 31.5% of the simulation in RGS4 and 36.1% in RGS8. The

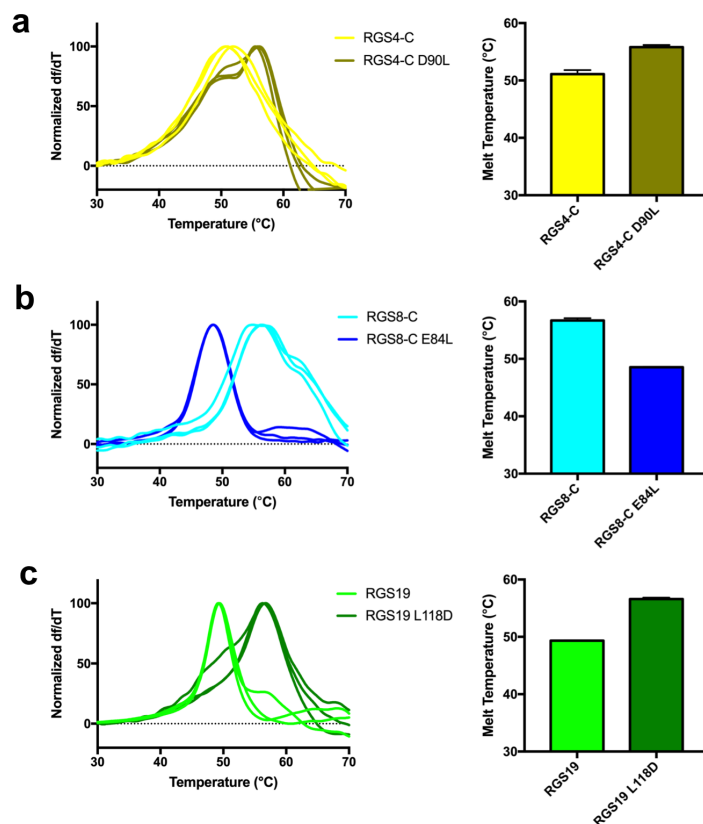


Figure 5.2. Thermal stability was determined by differential scanning fluorimetry. A) The L118D mutation in RGS19 increased melting temperature by 7 °C compared to WT. B) The E84L mutation in RGS8 decreased melting temperature by 8 °C. C) The RGS4 D90L mutation introduced a biphasic melt curve and increased melting temperature by 5 °C. For each pair, derivative melt curves are shown on the left and melt temperatures are shown on the right. Error bars represent SD. n=3. Analyzed by 1-way ANOVA with Sidak's Multiple Comparisons test (This figure is provided by my co-author Vincent Shaw [16]). ****p < 0.0001

salt bridge interaction between residues of $\alpha 4$ and $\alpha 5$ - $\alpha 6$ interhelical loop, also not present in RGS19, was maintained for 58.7% of time in RGS4 and 44.2% in RGS8. The charged pair that is unique to RGS8 between $\alpha 5$ and $\alpha 6$ helices remained in contact for 47.5% of the simulation.

Based on these MD results, we elected to make mutations that altered interhelical ($\alpha 4$ - $\alpha 5$ and $\alpha 6$ - $\alpha 7$) salt bridges to test their functional roles. In helix $\alpha 4$, L118 in RGS19 (Fig. 5.1E)

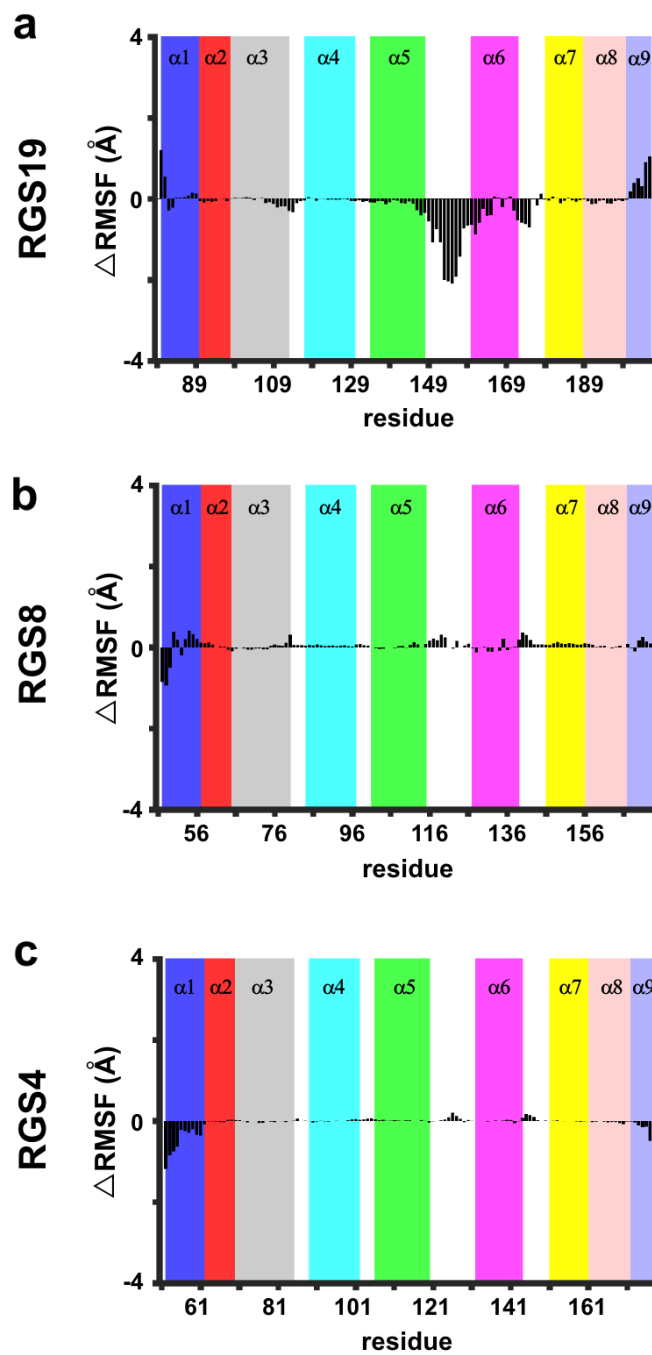


Figure 5.3. Change in RMSF per residue (Δ RMSF) between wild-type RGS proteins and RGS proteins with mutation in the α 4- α 5 salt bridge forming residue. (a) L118D in RGS19 (b) E84L in RGS8 and (c) D90L in RGS4. Data represent differences in RMSF from two independent MD simulations of the mutated and unmutated forms of RGS proteins.

was mutated to an aspartate to introduce the salt bridge found in RGS4 and RGS8 (Fig. 5.1B). In helix $\alpha 7$, Q183 in RGS19 (Fig. 5.1G) was mutated to a lysine, to introduce the $\alpha 6$ - $\alpha 7$ salt bridge found in RGS4 and RGS8 (Fig. 5.1D). In order to eliminate confounding effects due to multiple cysteines in inhibitor potency experiments, all proteins, with and without salt-bridge mutations, used a single-cysteine protein background. Each construct has only the conserved cysteine in helix $\alpha 4$ of the RGS domain.

To determine how disruption of the salt bridges D90-K125 in RGS4 and E84-R131 in RGS8 and addition of a salt bridge by L118D mutation in RGS19 may alter protein structure or dynamics, thermal stability was measured by differential scanning fluorimetry. As expected, addition of a salt bridge in RGS19 by the L118D mutation caused a 7 °C increase in thermal stability compared to WT (Fig 5.2A). Removal of a salt bridge by the E84L mutation in RGS8 caused an 8 °C decrease in thermal stability (Fig 5.2B). Unexpectedly, RGS4 showed a more complex pattern, in which D90L mutation resulted in a biphasic melt curve and a 5 °C increase in melting temperature rather than a decrease (Fig 5.2C).

To probe the molecular details of changes in structural flexibility in the mutated RGS4, RGS8, and RGS19, we conducted microsecond timescale classical MD simulations in explicit-solvent for RGS4 D90L, RGS8 E84L, and RGS19 L118D. To understand the effect of the mutations on the protein structures, particularly in helices in the vicinity of the mutated site, we computed the root-mean-squared-fluctuation (RMSF) per residue from two independent MD simulations of mutated and WT RGS4, RGS8, and RGS19. There were minimal changes in RMSF in RGS4 in comparison to RGS8 and RGS19 (Fig. 5.3). We find a modest increase in fluctuation of residues in mutant (E84L) RGS8 vs. the wild-type structure. These changes are in the loop region connecting helices $\alpha 5$ and $\alpha 6$, the $\alpha 6$ helix, and the loop connecting helices $\alpha 6$ and $\alpha 7$. Similar changes in lesser extent were found in the mutant (D90L) RGS4. The calculated change in RMSF per residue of the mutant (L118D) RGS19

from wild-type RGS19 reveals a strong stabilization and decrease in fluctuations of residues located in helices $\alpha 4$ - $\alpha 7$ and the interhelical loops between these helices. There is a particularly pronounced decrease in motion in the $\alpha 5$ - $\alpha 6$ interhelical loop (Fig. 5.3). Additionally, the RMSF values of residues in helices $\alpha 3$ and $\alpha 8$ of the mutated RGS19 are more stable in compared to wild-type than in mutated RGS4 and RGS8 (Fig. 5.3).

To further investigate whether the mutations D90L in RGS4, E84L in RGS8, and D118L in RGS19 affect residue-residue interactions, we calculated the dynamic cross-correlation matrix for the C α atoms in all MD trajectories. For wild-type RGS8, we find that the motions of residues in the $\alpha 4$ helix (highlighted by the dashed lines in Fig. 5.4) and the $\alpha 5$ helix (highlighted by the solid-lines in Fig. 5.4) are marginally positively correlated. This positive correlation between the $\alpha 4$ and $\alpha 5$ helices remains in the RGS8 E84L mutant, but shows a modest shift in areas of correlation away from the loop connecting $\alpha 4$ - $\alpha 5$ to mid-regions of the $\alpha 4$ and $\alpha 5$ helices (see arrows). For WT RGS4 and RGS19 the correlation matrices show slight positive correlation between the residues of the $\alpha 4$ helix and the residues of the $\alpha 5$ helix. For the RGS19 L118D mutant, we find higher residue-residue correlations between residues in helices $\alpha 4$ and $\alpha 5$ in comparison to unmutated RGS19 (see arrows).

In order to experimentally determine which regions in WT and mutant proteins were affected by the salt bridge mutations, hydrogen-deuterium exchange was performed. After exposure to solvent containing 90% D₂O, proteins were digested with pepsin and deuterium incorporation (DI) was measured by mass spectrometry as previously reported [13]. In RGS4, the fragment surrounding the salt-bridge mutation site (aa 88-91) took up deuterium very slowly in both the WT and D90L mutant constructs, reaching 8.1% and 6.7% DI, respectively. However, the D90L mutation led to a substantial increase in deuterium exchange in the 92-97 fragment surrounding Cys95, from 17.5% to 37.0% DI. The RGS4 D90L mutant also trended toward increased DI across all protein fragments compared to WT RGS4, espe-

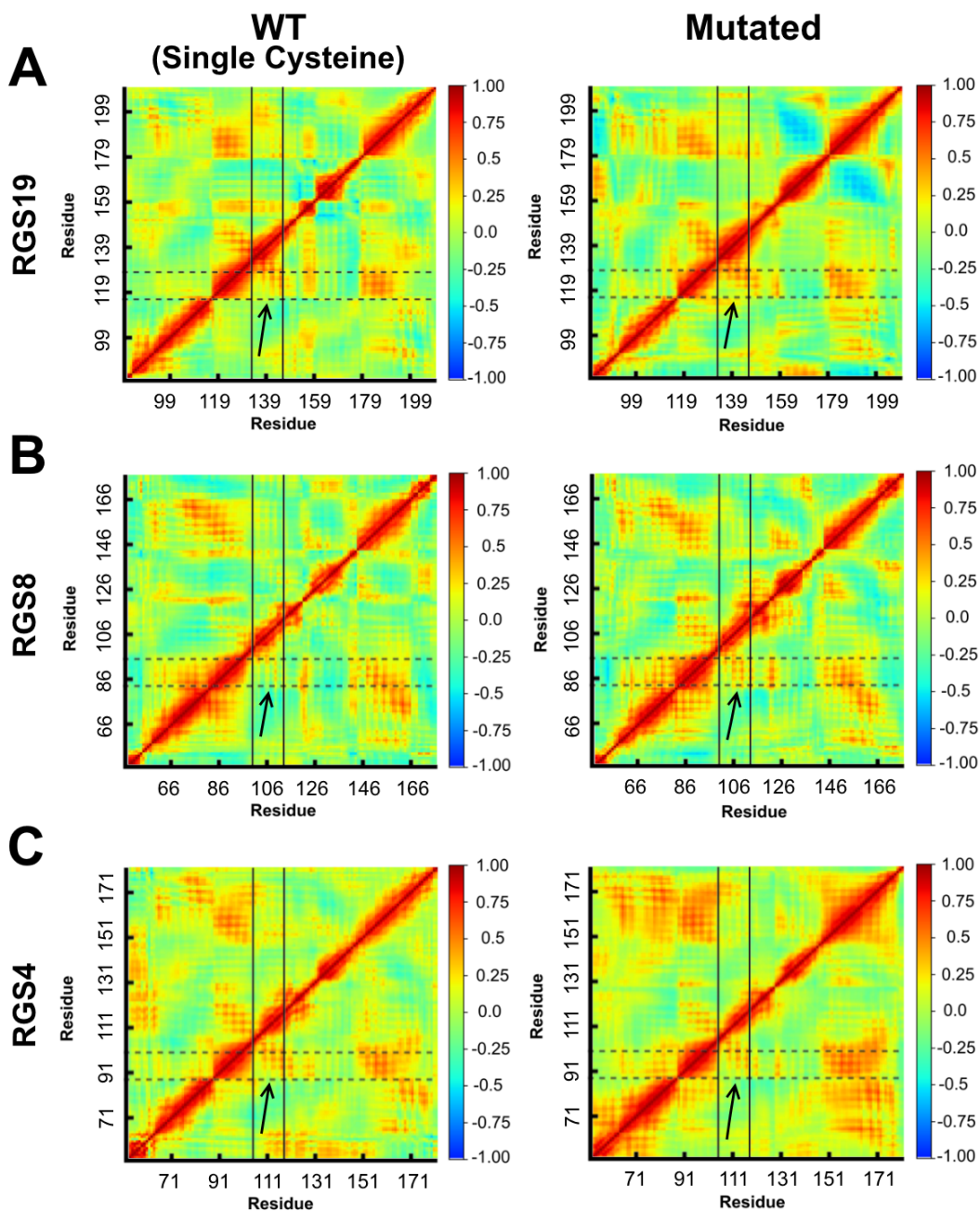


Figure 5.4. (a) Dynamic cross-correlation matrix calculated for the $C\alpha$ atoms of (a) RGS19/RGS19 L118D, (b) RGS8/RGS8 E84L, (c) RGS4/RGS4 D90L. Horizontal dotted lines indicate the regions of the $\alpha 4$ helix, while vertical solid lines indicate the regions of the $\alpha 5$ helix for each protein. The color scheme ranges from anticorrelation (-1.0, blue), no correlation (0, green), and positive correlation (+1.0, red). Values are the average for the two independent simulation runs of each protein.

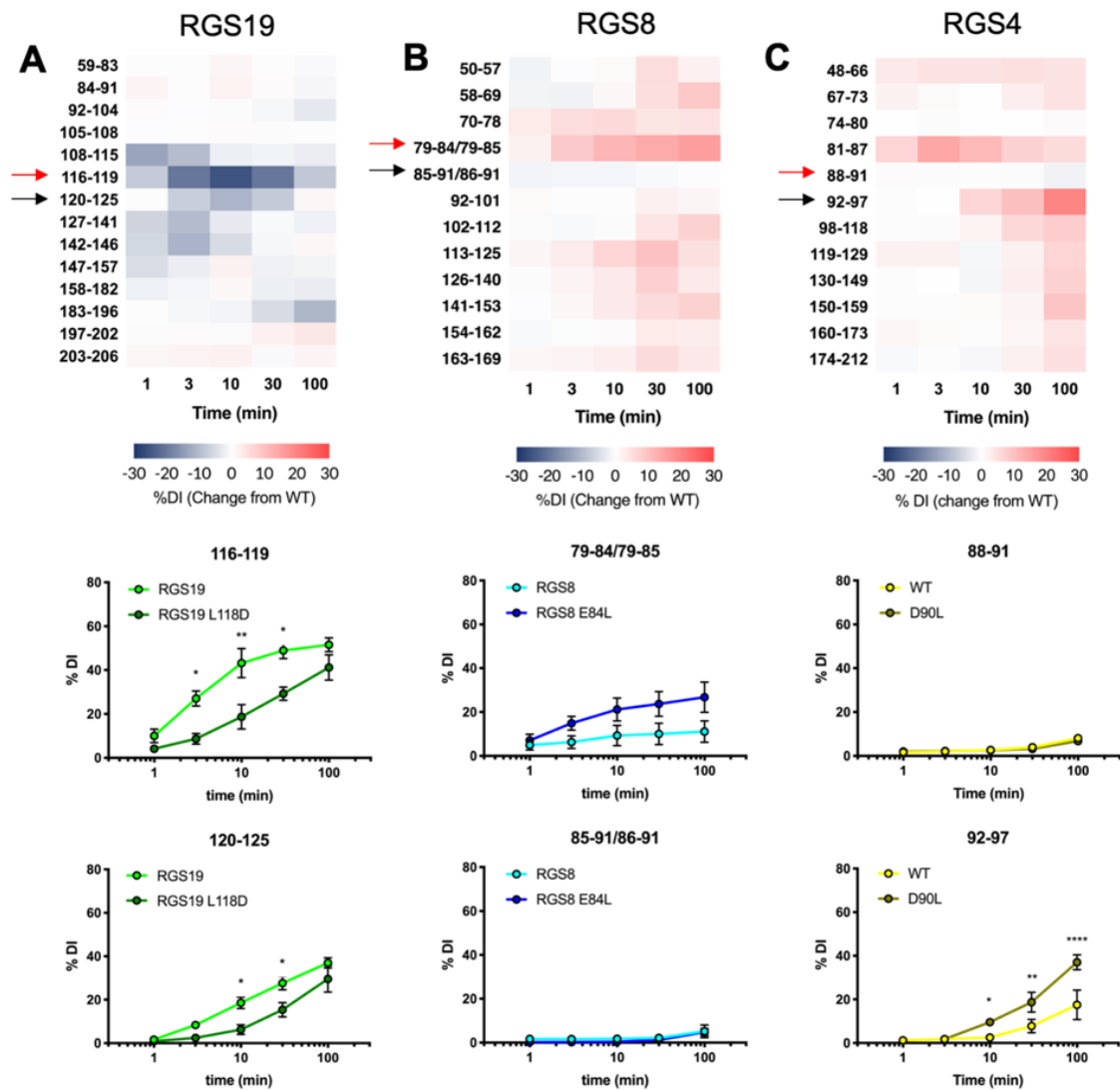


Figure 5.5. Difference in % deuterium incorporation ($\Delta\%DI$) between mutated and unmutated proteins in RGS19 L118D (A), RGS8 E84L (B), and RGS4 D90L (C) fragments, as measured by HDX. Red arrows indicate fragments containing mutated residue, and black arrows indicate fragments containing conserved $\alpha4$ cysteine. Kinetics of deuterium incorporation in these fragments for individual constructs are shown below. $n=3$. Error bars represent SD. Analyzed by 2-way ANOVA with Sidak's multiple comparisons test (This figure is provided by my co-author Vincent Shaw [16]). * $p < 0.05$, ** $p < 0.01$, *** $p < 0.0001$

cially at higher timepoints (Fig. 5.5A). In RGS8, removal of the salt-bridge forming residue by the E84L mutation did not cause a significant change in DI in either of the fragments of the $\alpha4$ helix but trended toward a global increase in DI throughout the protein (Fig.

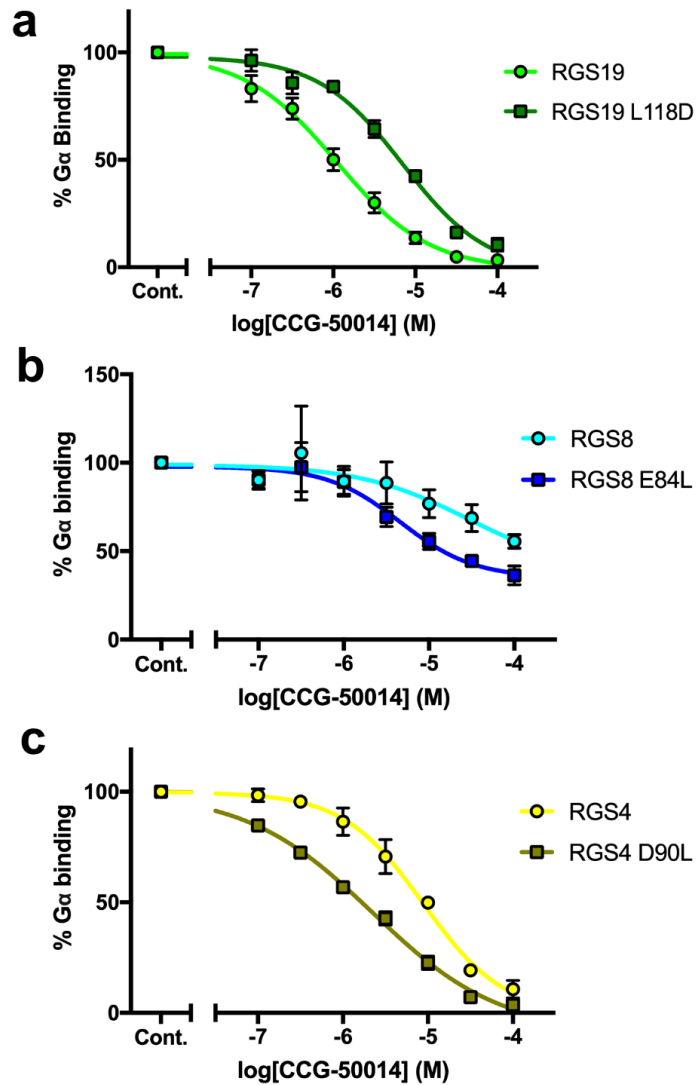


Figure 5.6. Potency of inhibition of CCG-50014 against $\alpha 4$ is altered in salt bridge mutants of RGS proteins. (A) RGS4 IC_{50} : $8.8 \mu\text{M}$, RGS4 D90L IC_{50} : $2.2 \mu\text{M}$. (B) RGS8 IC_{50} : $29 \mu\text{M}$, RGS8 E84L IC_{50} : $4.6 \mu\text{M}$ (C) RGS19 IC_{50} : $7.0 \mu\text{M}$, RGS19 L118D IC_{50} : $1.1 \mu\text{M}$. $n=3$ (This figure is provided by my co-author Vincent Shaw [16]).

5.5B). In RGS19, mutation of L118 to a salt bridge-forming residue, aspartic acid, caused significant decreases in DI in both $\alpha 4$ helical fragments, aa 116-119 and aa 120-125. In the 116-119 fragment, WT RGS19 had reached 43.1% DI by 10 minutes, while the RGS19 L118D mutant showed less than half as much DI (18.7%). In fragment 120-125, WT RGS19 reached 18.5% DI at 10 minutes, while the RGS19 L118D mutant reached only 6.2%. Unlike RGS4

and RGS8, the RGS19 L118D mutant's changes in DI were more restricted to fragments from helices neighboring the mutation site, and were most pronounced in the early (1 to 10 minute) timescale (Fig. 5.5C).

Finally, to see whether differences in the $\alpha 4$ salt-bridge forming residue altered compound potency, we used a flow-cytometry based protein-protein interaction assay (FCPIA) [215, 216] to measure the binding of RGS proteins to $G\alpha o$ and inhibition by CCG-50014. The L118D mutation in RGS19 induced an increase in IC_{50} from 1.1 μM (WT) to 7.0 μM (L118D). Conversely, removal of this charged $\alpha 4$ residue in RGS4 and RGS8 induced a decrease in IC_{50} . CCG-50014 inhibited the RGS- $G\alpha$ interaction with an IC_{50} of 8.8 μM for WT RGS4 and 2.2 μM for the RGS4 D90L mutant. It showed a potency of 29 μM for WT RGS8 and 4.6 μM for the RGS8 E84L mutant. None of the mutations to salt bridge-forming residues on the $\alpha 4$ helix caused notable changes in affinity between $G\alpha o$ and RGS proteins. The L118D mutation in RGS19 shifted the K_d of the $G\alpha o$ interaction from 16.6 nM to 20.2 nM, the E84L mutation in RGS8 shifted the K_d from 5.9 nM to 4.8 nM, and the D90L mutation in RGS4 shifted the K_d from 5.2 nM to 3.9 nM (Fig. 5.6).

5.4.1 The analysis of mutated RGS proteins flexibility

A comparison of the crystal structures of the three RGS proteins studied here revealed several differences in charged residue contacts among the proteins. We first observed that RGS19 has fewer interhelical salt bridges between helices in its $\alpha 4$ - $\alpha 7$ helical bundle than RGS4 or RGS8, which may be responsible for the high flexibility and low melting temperature previously observed in WT RGS19 [13]. RGS8 has four distinct interhelical salt bridge locations within the helical bundle, while RGS4 has three and RGS19 has one (Fig 5.1A), correlating with previously observed flexibility differences. RGS19 is most flexible, followed

by RGS4 and RGS8 [13]. This further suggests a role of salt bridges in RGS protein flexibility.

To determine which of these salt bridge locations was most relevant to the overall stability of the RGS proteins, we analyzed the stability of salt bridges via μ s-timescale MD simulations. A salt bridge between a glutamate on α 4 and lysine on α 5 is well conserved among RGS proteins (Fig. 5.1A, 5.1B, and 5.1E). Therefore, it is unlikely to contribute to differences in flexibility among RGS isoforms. This leaves two positions at which interhelical salt bridges are shared by RGS4 and RGS8 but lack necessary residues in RGS19: L118 (α 4) and Q183 (α 7). Because these positions were most likely to contribute to differences in RGS19 protein flexibility, we chose to mutate each of these positions in RGS19 (L118D and Q183K) in order to replicate the residues in RGS4 and RGS8. The RGS19 L118D mutation significantly increased the melting temperature compared to WT (Fig 5.2A). This increased stability suggests that addition of a D118-K125 salt bridge interaction in RGS19 was successful. Attempts to add a salt bridge at the α 6- α 7 position in RGS19 by Q183K mutation, however, produced no significant change in either thermal stability or potency of inhibition. This correlates with the observation that the α 6- α 7 salt bridge (D130-K155 in RGS4 and D124-K149 in RGS8) was less stably maintained in simulations than was the α 4- α 5 salt bridge (D90-K124 in RGS4 and E84-R119 in RGS8). It is possible that hydrogen bonding contacts between Gln183 on α 7 and Asp158 on α 6 in RGS19 are sufficient to compensate for the lack of a salt bridge at this position. In light of these results, we found it unlikely that the difference between Gln183 in α 6 of RGS19 and the lysines found in RGS4 and RGS8 (K155 and K149 respectively) play a major role in the flexibility differences between these proteins. Rather, the salt bridge-forming residue on α 4 is a stronger driver of differences in protein flexibility.

To verify the importance of differences in the α 4- α 5 salt bridge in protein structure and function, we also tested whether the elimination of this salt bridge in RGS4 and RGS8

could reduce the thermal stability of the proteins. The acidic residues D90 (RGS4) and E84 (RGS8) were each mutated to a leucine, which is the corresponding residue in RGS19. As expected, this significantly reduced thermal stability in RGS8, suggesting a successful removal of this salt bridge (Fig 5.2B). However, while the thermal stability of the RGS4 D90L mutant was significantly altered compared to WT RGS4, the mutation resulted in a biphasic melt curve shape with increased overall thermal stability (Fig 5.2C). This may indicate that the mutation had a more complex effect on the protein structure and its dynamics than a simple disruption of the salt bridge.

To determine the effects of mutations in salt bridge-forming residues on protein dynamics, both an *in silico* approach (all-atom MD simulations) and an experimental approach (hydrogen-deuterium exchange) were employed. MD simulations demonstrated that the addition of a salt bridge by the L118D mutation in RGS19 drastically reduces fluctuations throughout the $\alpha 3$ - $\alpha 8$ helical bundle. This stabilization was particularly dramatic in the $\alpha 6$ helix and the $\alpha 5$ - $\alpha 6$ interhelical loop (Fig 5.3C). Likewise, in HDX analysis, decreases in DI were observed in the RGS19 L118D mutant compared to WT. As expected, the highest magnitude of decrease was observed in the fragment containing the mutation site, suggesting a strong reduction in solvent exposure at the $\alpha 4$ helix. Other decreases in DI were most prominent at 1 to 3 minutes in fragments from the $\alpha 4$ and $\alpha 5$ helices and the $\alpha 5$ - $\alpha 6$ interhelical loop (Fig. 5.5C). The $\alpha 4$ helix changes are of particular interest given the location of the Cys123 as a target of the TDZD compounds. Conversely, mutations that eliminated salt bridges in RGS4 and RGS8 increased deuterium incorporation. Both mutants had increased DI in some fragments from the $\alpha 4$ helix (Fig. 5.5A and 5.5B), but the RGS4 D90L mutant did not have increased DI in the fragment spanning the mutation site (Fig. 5.5A). This fits with the thermal stability data that suggest that the effect of the D90L mutation in RGS4 is more complex than disruption of a single interacting pair.

5.4.2 MD simulations to probe mutated RGS proteins conformations

In MD simulations, the RGS4 D90L and RGS8 E84L mutations did not have as large an effect on the magnitude of residue fluctuations as did mutation of L118D in RGS19 (Fig 5.3A and B). This may be because differences become apparent on shorter timescales in RGS19 than in RGS4 and RGS8, so simulations on μ s timescales may not have captured all of the differences in dynamics caused by mutations in RGS4 (D90L) and RGS8 (E84L). Indeed, in HDX studies, stronger differences in DI were observed between RGS19 and RGS19 L118D at shorter timepoints (1 and 3 minutes) than in RGS4 D90L and RGS8 E84L (Fig 5.5A-C).

The μ s-timescale MD simulations of mutated (D90L) RGS4 and WT RGS4 captured positive residue-residue (C_α - C_α) correlations between the α 4 and α 5 helices. This indicates a complex effect of the D90L mutation on the protein structure and dynamics, especially alongside data indicating increased thermal stability in mutant (D90L) RGS4. We also observed a change in the pattern of positive correlations between the α 4 and α 5 helices in the RGS8 E84L mutant from those of WT RGS8, possibly resulting from the removal of the interhelical salt-bridge. The marked increase in positive correlation between residues in the α 4 and α 5 helices in the RGS19 L118D mutant likely results from the introduced interhelical salt-bridge.

Finally, to determine how changes in protein flexibility affected the potency of inhibition by an RGS inhibitor, we used a flow-cytometry based protein-protein interaction assay (FCPIA) to evaluate the inhibition of $G\alpha$ binding by CCG-50014. Importantly, manipulation of RGS protein flexibility induced the expected changes in the potency of inhibition by TDZD covalent modifiers. The inhibition curve of the RGS19 L118D mutant was shifted rightward in comparison to RGS19, indicating that the more flexible mutant is less potently inhibited. Conversely, the inhibition curves for the RGS4 D90L mutant and the RGS8 E84L mutant were shifted leftward in comparison to those for RGS4 and RGS8, respectively. Thus,

enhancing flexibility by removal of salt bridge-forming residues increased the potency of inhibition by CCG-50014. These results support a causal relationship between the protein flexibility and the potency of inhibition.

5.5 Conclusion

In conclusion, differences in flexibility among RGS isoforms are capable of driving differences in potency of inhibition of a covalent inhibitor, CCG-50014. The differences in isoform flexibility in turn are strongly influenced by the presence or absence of an $\alpha 4$ - $\alpha 5$ salt bridge and manipulation of this salt bridge is sufficient to induce changes in inhibitor potency among single-cysteine RGS proteins. Developing a deeper understanding of these differences in flexibility may enable the development of a new generation of RGS inhibitors with novel specificities.

5.6 Publications

The work described in this chapter has submitted in the following journal article:

- Shaw, V.S., Mohammadi, M., Quinn, J.A., Vashisth, H., and Neubig, R.R. (20XX). An Interhelical Salt Bridge Controls Flexibility and Inhibitor Potency For Regulators of G-protein Signaling (RGS) Proteins 4, 8, and 19. *Molecular Pharmacology*, (Submitted).

CHAPTER 6

OTHER SYSTEMS STUDIED: PHOSPHODIESTERASE 4 (PDE4) ENZYME

6.1 Abstract

Novel chemical controls are needed that selectively target human, animal, and plant parasitic nematodes with reduced adverse effects on the host or the environment. We hypothesize that the phosphodiesterase (PDE) enzyme family represents a potential target for development of novel nematicides and anthelmintics. MD simulations were used to evaluate differences in binding interactions of these inhibitors within the PDE4 catalytic domain. Of 32 residues within 5 Å of the ligand binding site, five revealed significant differences in non-bonded interaction energies that could account for the differential binding affinities of roflumilast and zardaverine. One site (Phe506 in human PDE4, Tyr253 in the *C. elegans* PDE4 catalytic domain) is predicted to alter the binding conformation of roflumilast and zardaverine (but not IBMX) into a less energetically favorable state. These results support the feasibility of designing the next generation of anthelmintics/nematicides that could selectively bind to nematode PDEs.

6.2 Background

Cyclic nucleotide metabolism is of central importance for a wide range of physiological processes in nematodes, as attested by the presence in *C. elegans* of 38 genes that synthesize cAMP or cGMP [217], as well as six PDE genes [218, 219, 220]. The cAMP second messenger has been linked to various behaviors including feeding and locomotion [221, 222]. Neural

pathways responsible for sensory signaling (chemoreceptors, thermotaxis, phototaxis) appear to be regulated through cGMP signaling pathways [218, 223, 224, 225]. Furthermore, RNAi, gene deletion, and pharmacological studies have shown that altered PDE activity can cause lethality, sterility, aberrant locomotion, lethargy, and altered development in *C. elegans* [226, 227, 228, 229, 230]. Some of the observed phenotypes (e.g., lethality, sterility) are clearly relevant to developing effective anthelmintics nematicides that specifically target parasitic nematodes-but not other animal phyla. Indeed, PDEs and their inhibitors are being investigated for their therapeutic potential in combatting various protozoal diseases [231]. Another advantage to targeting phytoparasitic nematode PDEs for the development of nematicides is the greatly reduced likelihood that a PDE inhibitor-based nematicide would have adverse effects on plants, since Class I PDEs have not been identified to date in plants [232].

The Class I PDE superfamily in vertebrates consists of eleven PDE families that have been identified throughout the animal kingdom [233]. The eleven families are distinguished by differences in their substrate specificity, modes of regulation, pharmacological properties, and tissue distribution [234]. However, all Class I PDE enzyme families share a conserved Prosite domain signature (Prosite PS00126¹) in the catalytic domain consisting of the amino acid sequence pattern HD[LIVMFY]xHx[AG]xx[NQ]x[LIVMFY]. The crystal structures of the catalytic domains of almost all the PDE families have been solved, providing atomic-level details on the enzymatic and pharmacological properties of this enzyme superfamily [108]. The catalytic domains of the Class I PDE superfamily are made up of 330 amino acids whose secondary structure consists of 16 α -helices. These α -helices create three subdomains [109], which form a deep catalytic pocket at their center. The active site is composed of two sub-pockets, which bind two divalent metal ions and the substrate, respectively [108]. Zinc

¹<https://prosite.expasy.org/PDOC00116>

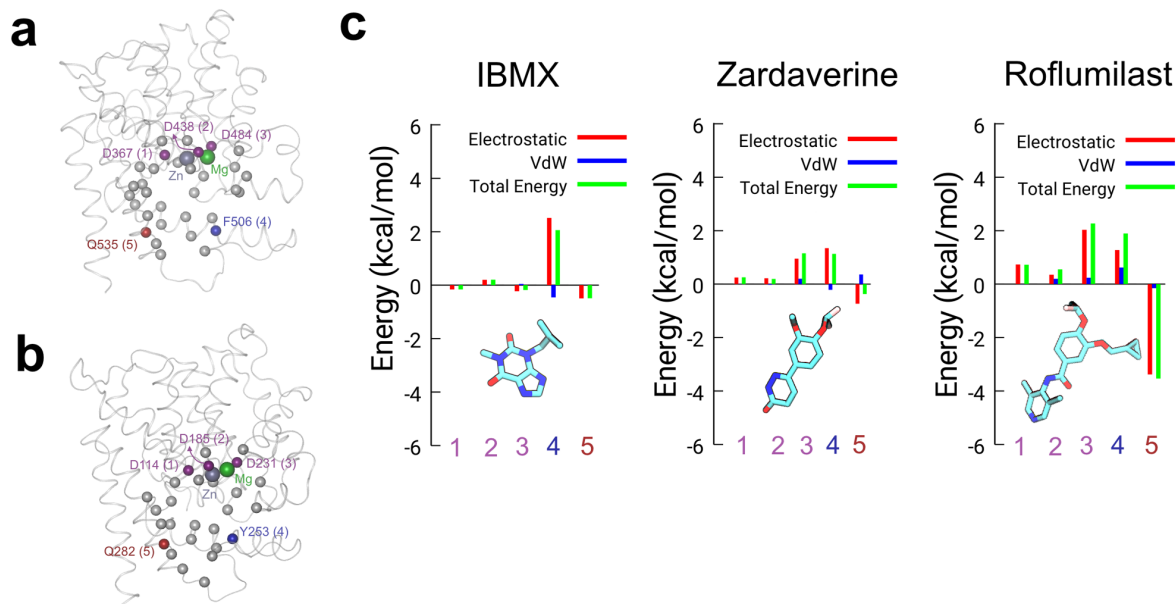


Figure 6.1. Interactions of five key C_{α} -atoms of the 32 residues that interact with ligands in (a) human PDE4D and (b) *C. elegans* PDE4 binding pocket. The Zn and Mg ions are shown as gray and green spheres, respectively. The three Asp residues coordinating with Zn and Mg ions are highlighted by purple spheres. The Gln and Phe/Tyr residues are shown as red and blue spheres, respectively. The protein backbone is represented as ribbons, the C_{α} -atoms of residues of the binding pocket and ions are shown as space-filling. (c) Changes in total non-bonded interaction energy and its components for *C. elegans* PDE4 relative to human PDE4D are shown for selected residues in the binding pocket (labeled 1-5 in panel a and b).

and magnesium ions are stabilized by conserved His and Asp residues in the metal binding pocket [108]. The crystal structure of human PDE4 catalytic domain in a complex with 5'-AMP [235] has revealed that cyclic nucleotides are stabilized by ionic interactions with the bound divalent cations and with Asp and His residues in the metal binding pocket, as well by hydrophobic interactions with conserved Gln and Phe residues in the hydrophobic pocket. The invariant Gln residue of PDEs have been shown to be critical for substrate and inhibitor binding [108].

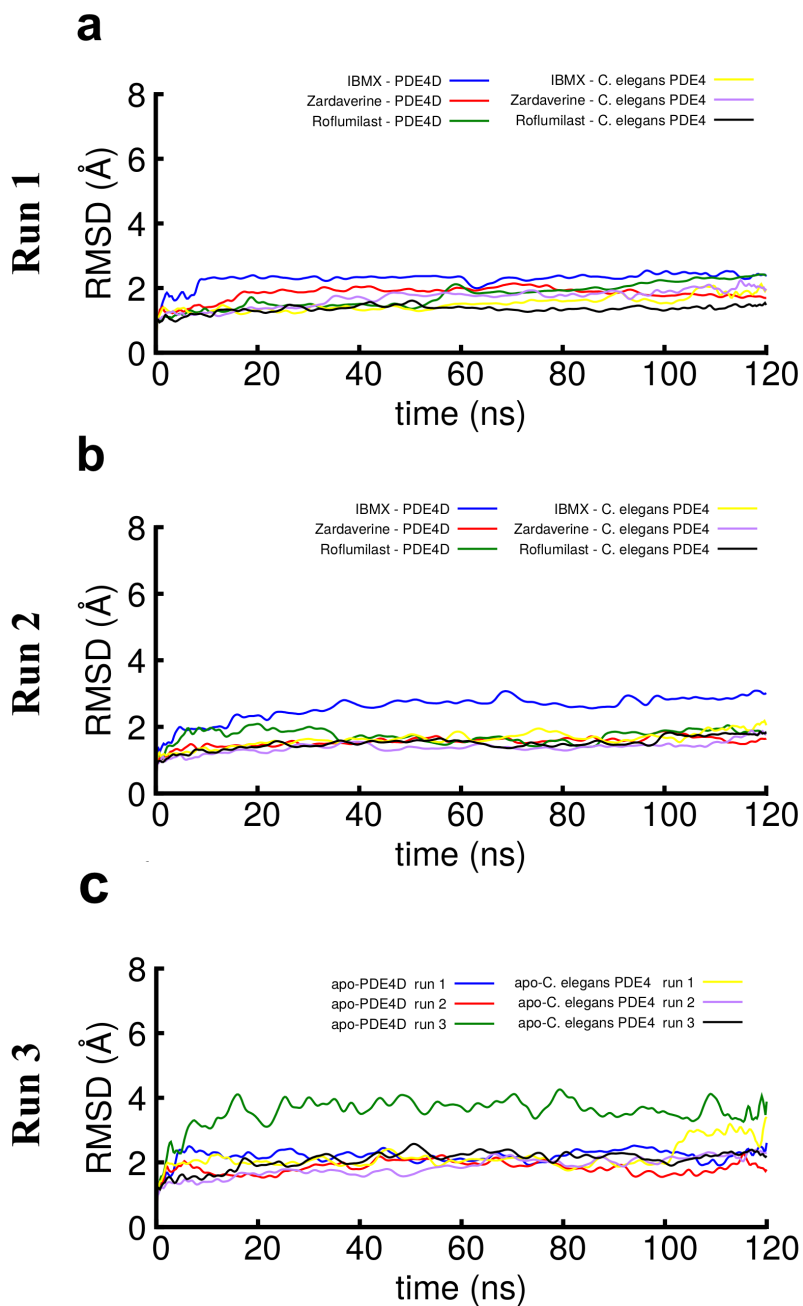


Figure 6.2. The traces of root-mean-squared-deviation (RMSD) vs. simulation time (ns) for PDE4D and *C. elegans* PDE4. (a and b) Two independent simulation runs for complexes of human PDE4D and *C. elegans* PDE4 with IBMX, zardaverine, or roflumilast. (c) RMSD traces of three independent simulation runs of apo-PDE4D and apo-*C. elegans* PDE4.

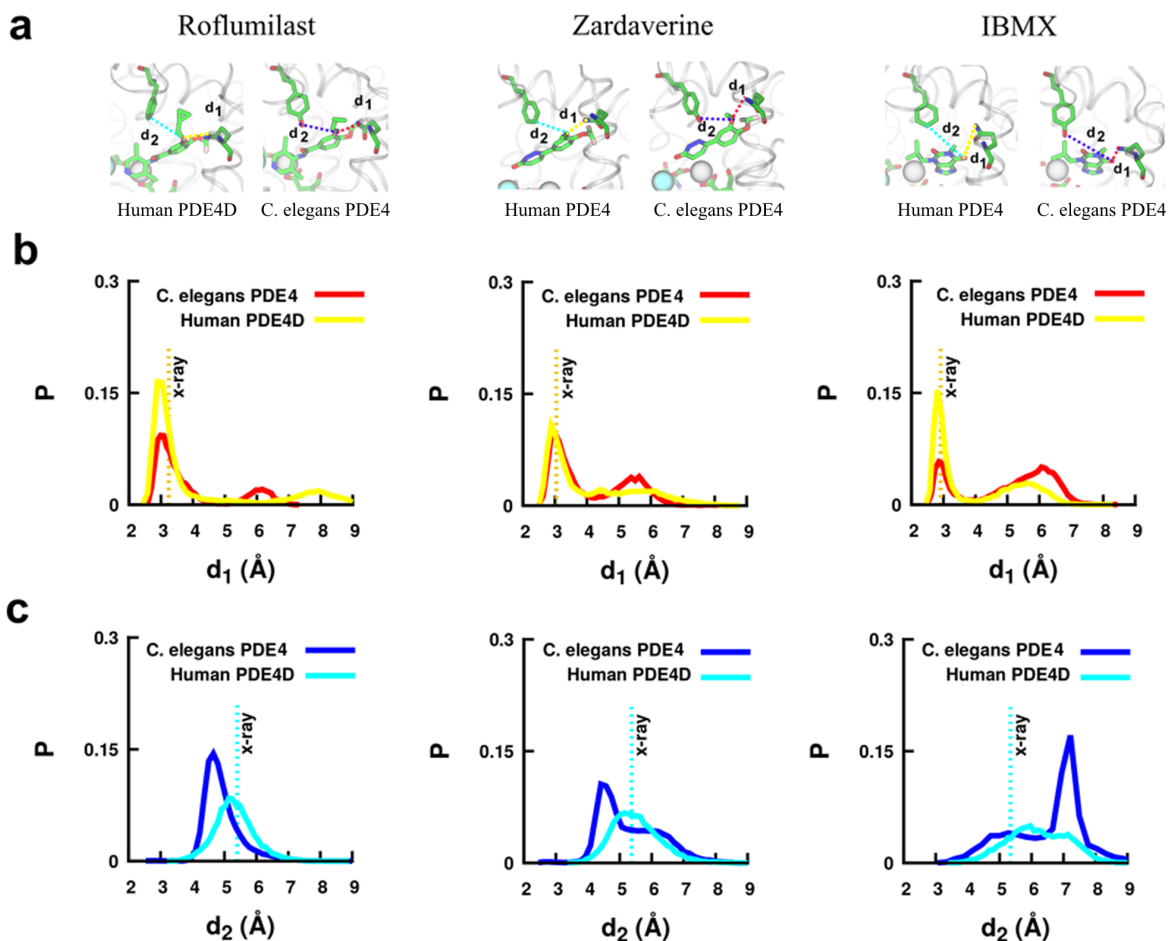


Figure 6.3. Probability (P) distributions of interatomic distances between ligand (O4 atom in roflumilast or zardaverine, or O6 atom in IBMX) and binding pocket residues. a) shows measurements of d_1 [between the oxygen atom on the ligand and the N δ atom on Gln535(human)/Gln282(*C. elegans*)] and of d_2 [between the oxygen atom of the ligand and the C4 atom of Phe506 (human)/O atom on the side chain of Tyr253(*C. elegans*)]. b) illustrates the distributions for distance d_1 for *C. elegans* PDE4 (red) and human PDE4D (yellow). c) shows the distributions for the distance d_2 for *C. elegans* PDE4 (blue) and human PDE4D (cyan). Vertical dotted lines in panels b and c indicate the distances in the crystal structures of inhibitors bound to human PDE4D.

Hypothesis: PDEs in parasitic nematodes represent a viable target for anthelmintic or nematicidal compounds.

The extensive literature on human PDE inhibitor pharmacology [236] and the commercial availability of many types of family-specific PDE inhibitor compounds enabled us to experimentally evaluate the potential of PDE inhibitors to serve as chemical nematicides targeting parasitic nematodes. In this chapter, I have used atomistic MD simulations (an approach used previously to compare the binding of inhibitors to different molecular human PDE4 isoforms [237, 238]) to investigate the role of structural differences in inhibitor interactions in human and nematode PDE4 that underlie the different pharmacological properties of nematode and human PDE4. These results support the idea that differences in the inhibitor binding site of nematode PDEs can be exploited to rationally design nematode-selective PDE inhibitors that act as an anthelmintic or nematicide without adverse effects on vertebrate animals or crops.

6.3 Methods

The initial coordinates for protein structures were obtained from the crystallographic structures of PDE4D bound to ligands with PDB codes: 1ZKN [239], 3G4L [240] [241], and 1MKD; in each instance, only one polypeptide chain of the overall structure was modeled, since the crystal structures of PDE4D consisted of identical PDE catalytic domains that co-crystallized into an oligomeric crystal. Numbering of amino acid residues for human PDE4D sequences used for MD simulations reflect the PDE4D 3 amino acid sequence. For simulation studies of *C. elegans* PDE4 with bound inhibitors, we created structural models using SwissModel² [242] with each of the previously mentioned PDE4D crystal structures as templates.

Upon comparing the *C. elegans* homology models with their templates, we observed that each model had initial mean-squared deviations (relative to their templates) of 0.12 Å,

²<https://swissmodel.expasy.org/>

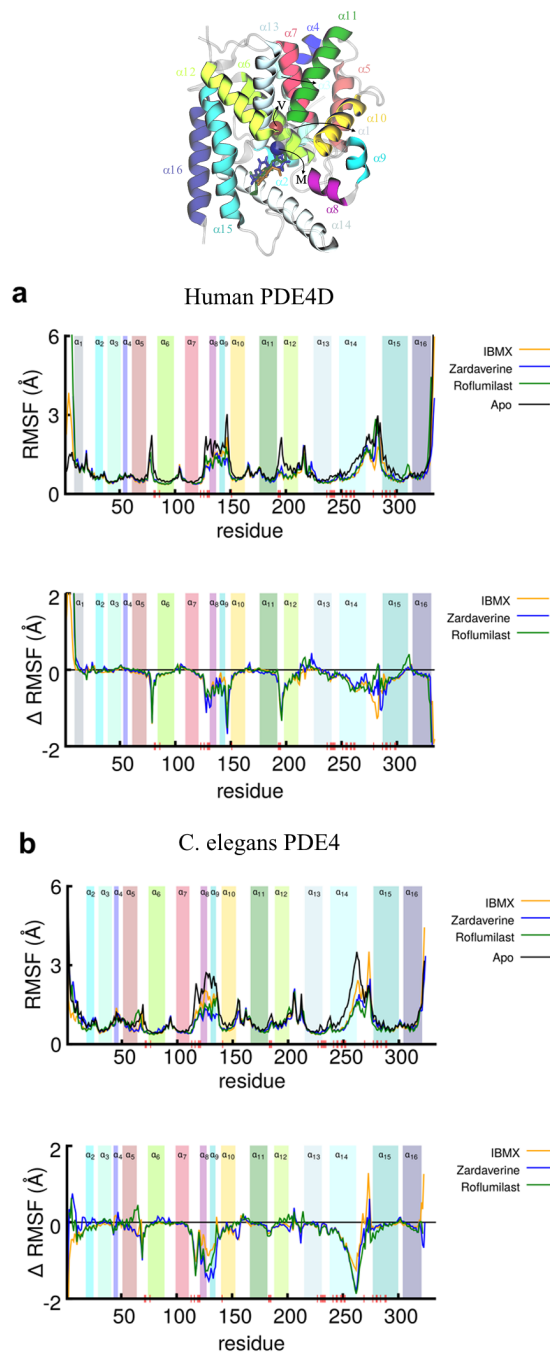


Figure 6.4. Root-mean-squared-fluctuation (RMSF) per residue (top panel) and the change in RMSF (Δ RMSF) per residue (bottom panel) are shown for (a) human PDE4D and (b) *C. elegans* PDE4 complexes with IBMX, zardaverine, and roflumilast. The superimposed structures for human PDE4D/*C. elegans* PDE4 along with superposition of IBMX, zardaverine, and roflumilast ligands (sticks) are shown at the top. The colored helices and vertical bars labeled $\alpha 1$ through $\alpha 16$ highlight the location of residues in the 16 helices in the catalytic domain. The Val334/81 and Met439/186 residues for human/*C. elegans* PDE4 are shown by red and blue spheres, respectively.

indicative of the homology models having a high structural similarity to the human PDE4 crystal structures. The stability of homology models was further tested using all-atom and explicit-solvent MD simulations. The stability of the *C. elegans* structural homology models in our MD simulations [in conjunction with the highly conserved nature of the PDE catalytic domain structure (PDEase_I; Pfam PF00233)] support their usefulness in the absence of experimentally determined structures.

We prepared six systems for MD simulations: three each for human PDE4D and *C. elegans* PDE4 such that each of the starting structures had a PDE inhibitor bound to it. Each system was then solvated with explicit TIP3P [205] water molecules, and charge-neutralized with counter-ions resulting in various system sizes. We used the software NAMD [111] for all MD simulations, and VMD [110] for system setup and post-processing analysis. CHARMM36 [238] force field was used including the CMAP correction [122, 123] for protein structures, and developed force-fields for all inhibitors using MATCH [125].

We used periodic boundary conditions [243] and computed long range-electrostatics using the particle-mesh Ewald summation [244] with a grid spacing of 1 Å, an integration time-step of 2 fs, and a cutoff-distance of 10 Å for van der Waals interactions; these settings are typically used for conducting MD simulations of solvated systems of proteins using NAMD [111]. We first energy minimized each system, and continued production runs of each system in the NPT-ensemble for 120 ns using a Langevin thermostat and Nosé-Hoover barostat [133, 134, 135]. We also carried out an independent run with the same length of simulation for each system, giving two production runs for each prepared system. Additionally, we carried out simulations of the same length for the apo states of human PDE4D and *C. elegans* PDE4.

6.3.1 Nonbonding interaction energy calculations

To investigate the role of individual amino acids in the binding pocket of each protein/ligand complex, we computed non-bonded interaction energies between all atoms of ligands and those of residues forming the binding pocket (i.e., within 5 Å from each ligand). Interaction energy values were estimated by splitting them into electrostatic and van der Waals interactions, as follows:

$$\Delta E_{non-bonded} = \Delta E_{elec} + \Delta E_{vdW} \quad (6.1)$$

We carried out these calculations by including all frames in each MD trajectory.

6.3.2 Analysis of salt-bridging interactions

The salt-bridging interaction analysis was carried out using VMD based on a distance criterion uniformly applied to determine the existence of salt-bridges for each frame in all trajectories [14]. Specifically, the formation of a salt-bridging interaction was considered if the distance between any of the oxygen atoms of acidic residues and the nitrogen atoms of basic residues were within a cut-off distance of 3.2 Å.

6.4 Results and Discussion

6.4.1 MD simulations to predict inhibitor binding conformations

To investigate the mechanistic details of differences in binding of each inhibitor, we performed two independent sets of MD simulations (120 ns each) of human and *C. elegans* PDE4 with each inhibitor (IBMX, zardaverine and roflumilast) (Fig. 6.1), and also carried out simulations of each enzyme without inhibitors. The root-mean-squared-deviation (RMSD) measured relative to initial structures revealed deviations below 2 Å indicating stable structures for both enzymes (Fig. 6.2). Through visual analyses of these simulations,

we identified 32 residues in the immediate vicinity of bound ligands (defined as within 5 Å of any of the inhibitors; Figs 6.1a-b) as forming a binding pocket and then computed interaction energies of inhibitors with each of these 32 residues. In Figs H.1 and H.2, we present non-bonded interaction energies (van der Waals and electrostatic) for each of the 32 residues where energies were computed based upon all atoms of each residue and of the inhibitor molecule. These analyses for human PDE4D and *C. elegans* PDE4 resulted from two independent sets of simulations. From our interaction energy analyses, we identified five key residues showing differences between the human and *C. elegans* PDE4 (Fig 6.1a-b): (a) three conserved Asp residues (residues 367, 438, and 484 in human PDE4D corresponding to residues 114, 185, and 231 in *C. elegans* PDE4; purple spheres) that are critical for the coordination of the zinc and magnesium ions; (b) a conserved Gln residue (Gln535 in human PDE4D and Gln282 in *C. elegans* PDE4; red spheres) that stabilizes ligand binding via non-covalent interactions; and (c) a Phe residue in human PDE4D (Phe506; blue sphere) and a Tyr residue at the same position in *C. elegans* (Tyr253; blue sphere) that show differences in non-bonded interactions. Fig 6.1c presents the differences in the total non-bonded interaction energy and its components (ΔE) for *C. elegans* PDE4 relative to the human PDE4D at these five sites. A positive value of ΔE indicates a higher non-bonded interaction energy of a given residue with the inhibitor in *C. elegans* in comparison to human PDE4D, and a negative value of ΔE indicates a lower, non-bonded interaction energy. We observed positive ΔE values for the three conserved Asp residues for roflumilast and, to a lesser extent, zardaverine, indicating stronger interactions in *C. elegans* relative to human PDE4D. In contrast, for IBMX the ΔE values between *C. elegans* PDE4 and human PDE4D are comparable. For all three inhibitor complexes with *C. elegans* PDE4, Tyr253 showed higher nonbonded interaction energy with the inhibitors in comparison to the corresponding Phe506 residue in human PDE4D. Based on the interaction energy analysis, we observed a correlation between the change in the interaction energy at the non-conserved and conserved residue sites. Primarily for roflumilast and to a lesser extent for zardaverine and IBMX, we

observed that an increase in the total non-bonded interaction energy at the non-conserved site (F506 in human vs. Y253 in *C. elegans*; labeled as the residue 4 in Fig 6.1) is correlated with a decrease in the total non-bonded interaction energy at the conserved site (Q535 in human vs. Q282 in *C. elegans*; labeled as the residue 5 in Fig 6.1). Similar correlation was observed between an increase in the total non-bonded interaction energy at the conserved site (D484 in human vs. D231 in *C. elegans*; labeled at the residue 3 in Fig 6.1) and a decrease at the conserved site (Q535 in human vs. Q282 in *C. elegans*; labeled as the residue 5 in Fig 6.1).

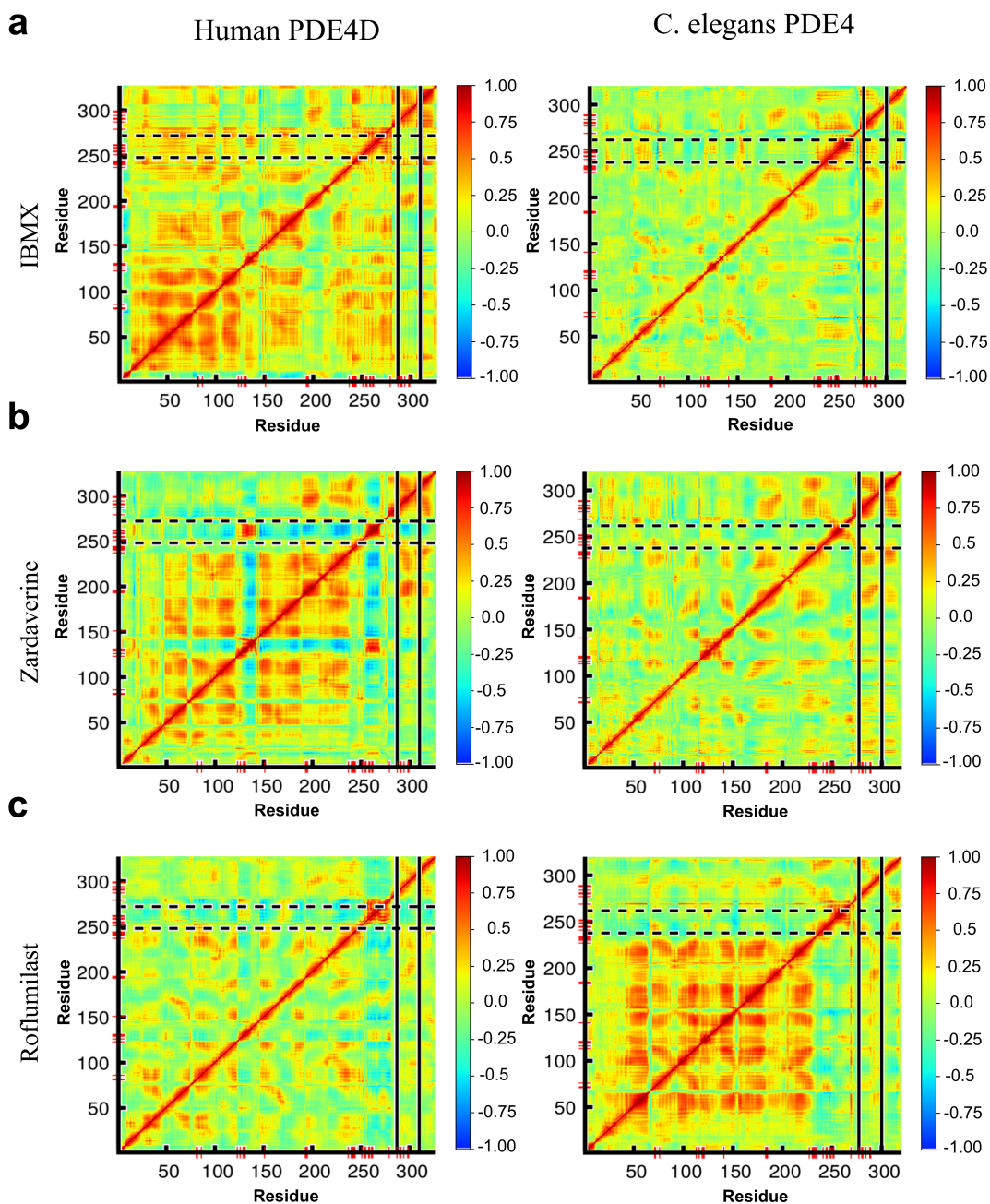


Figure 6.5. Dynamic cross correlation matrices calculated for the $C\alpha$ atoms of human PDE4D and *C. elegans* PDE4 complexed with IBMX (a), zardaverine (b), and roflumilast (c). Residues in the $\alpha 14$ and $\alpha 15$ helices are shown by areas between dashed-lines and solid-lines, respectively. Red tick-marks on the axes represent the 32 residues in the binding site (as depicted in Fig. 6.1a-b). The color scheme ranges from anticorrelation (-1.0, blue), no correlation (0, green), and positive correlation (+1.0, red). Values are the average for the two independent simulation runs.

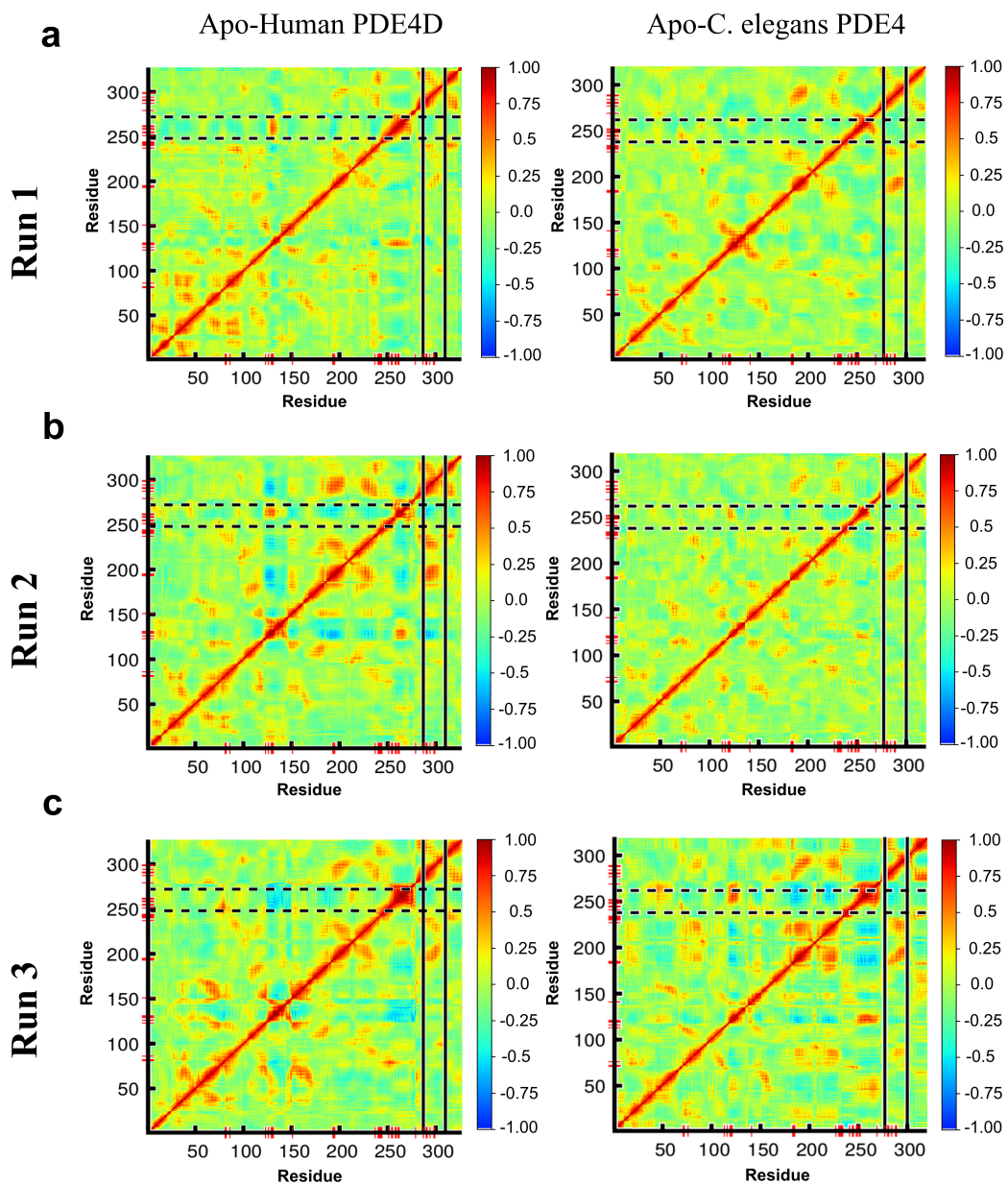


Figure 6.6. Dynamic cross correlation matrices calculated for the $C\alpha$ atoms of human PDE4D and *C. elegans* PDE4 in their apo state. Color scheme is the same as for Fig. 6.5. Panels a-c represent three independent simulations.

To investigate the variation in the docked positions of ligands in the binding pockets of human and *C. elegans* PDE4, I computed the interatomic distances between specific atoms

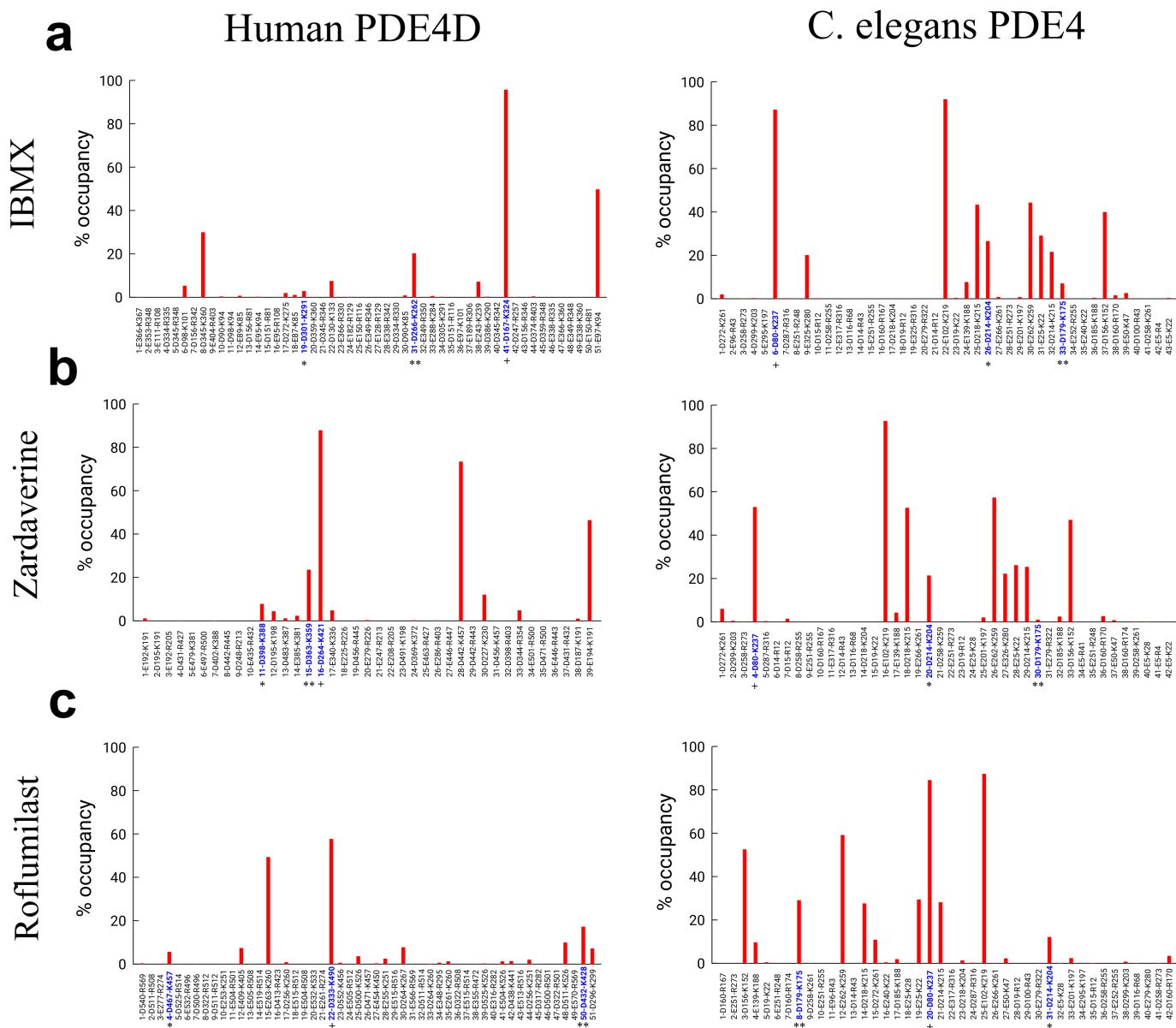


Figure 6.7. Key salt-bridging interactions are shown based upon the first set of MD simulations of human PDE4D and *C. elegans* PDE4 with IBMX (a), zardaverine (b), and roflumilast (c). Three conserved salt-bridges are labeled in blue.

in the ligands and the nearby Gln535(human)/Gln282(*C. elegans*) residues (d1 in Fig. 6.3). For *C. elegans* PDE4, the distributions of d1 are bimodal (red traces in Fig. 6.3b) for all three inhibitors, with roflumilast and zardaverine having a higher probability of being in

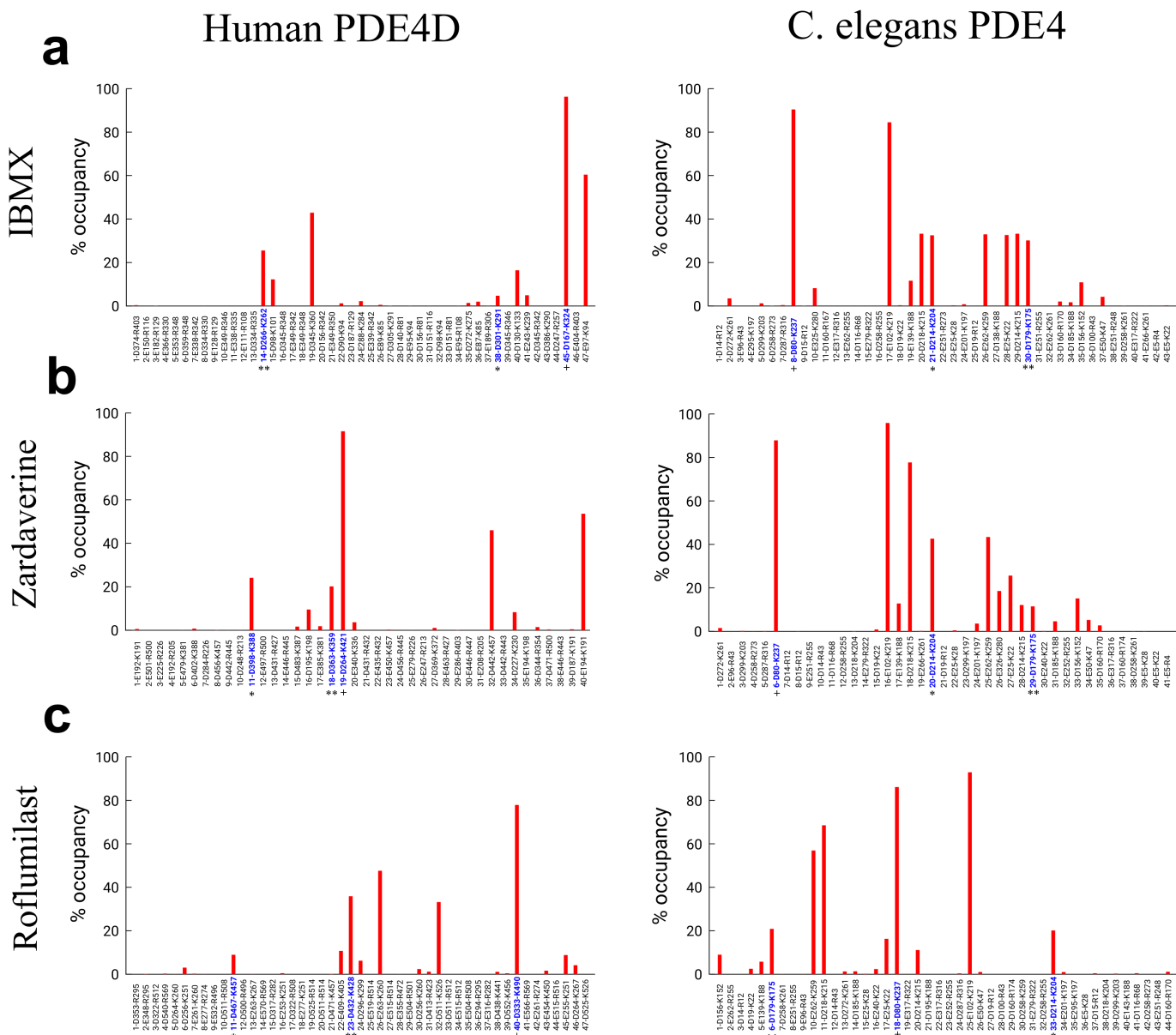


Figure 6.8. Data similar to Fig. 6.7 are shown for a second set of MD simulations with the three inhibitors.

states with $d1 \sim 3\text{\AA}$, while for IBMX both states at $d1 \sim 3\text{\AA}$ and $\sim 6\text{\AA}$ are equally probable. For human PDE4D, all inhibitors show an increased probability of being in states at shorter distances ($\sim 3\text{\AA}$) but bimodal distributions with lower probabilities of states at larger dis-

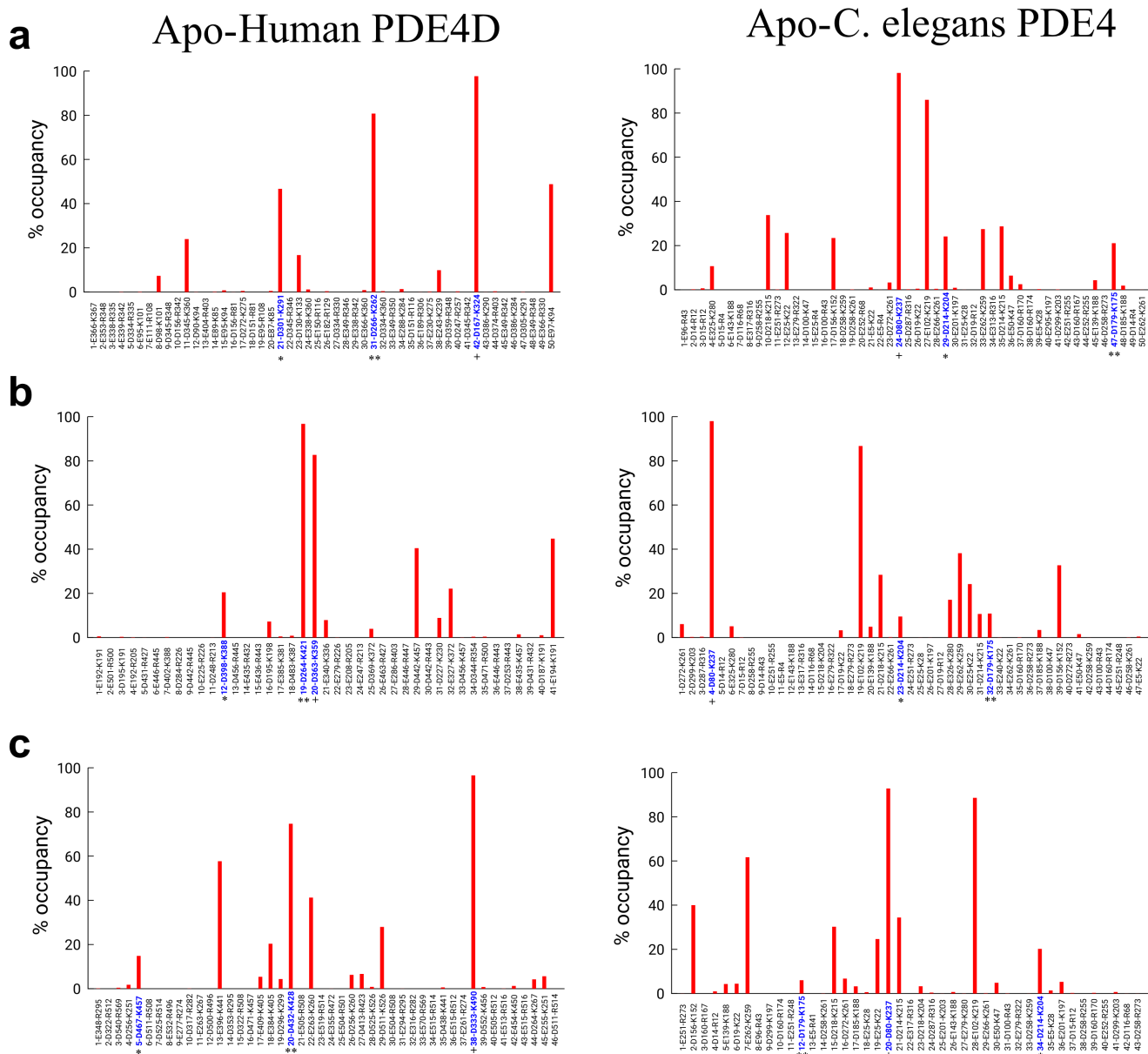


Figure 6.9. Data similar to Fig. 6.7 are shown for three independent sets of MD simulations of apo human PDE4D and apo *C. elegans* PDE4.

tances are observed for zardaverine and IBMX. These observations suggest that zardaverine and IBMX are more likely to transition between two distinct states within the binding pocket

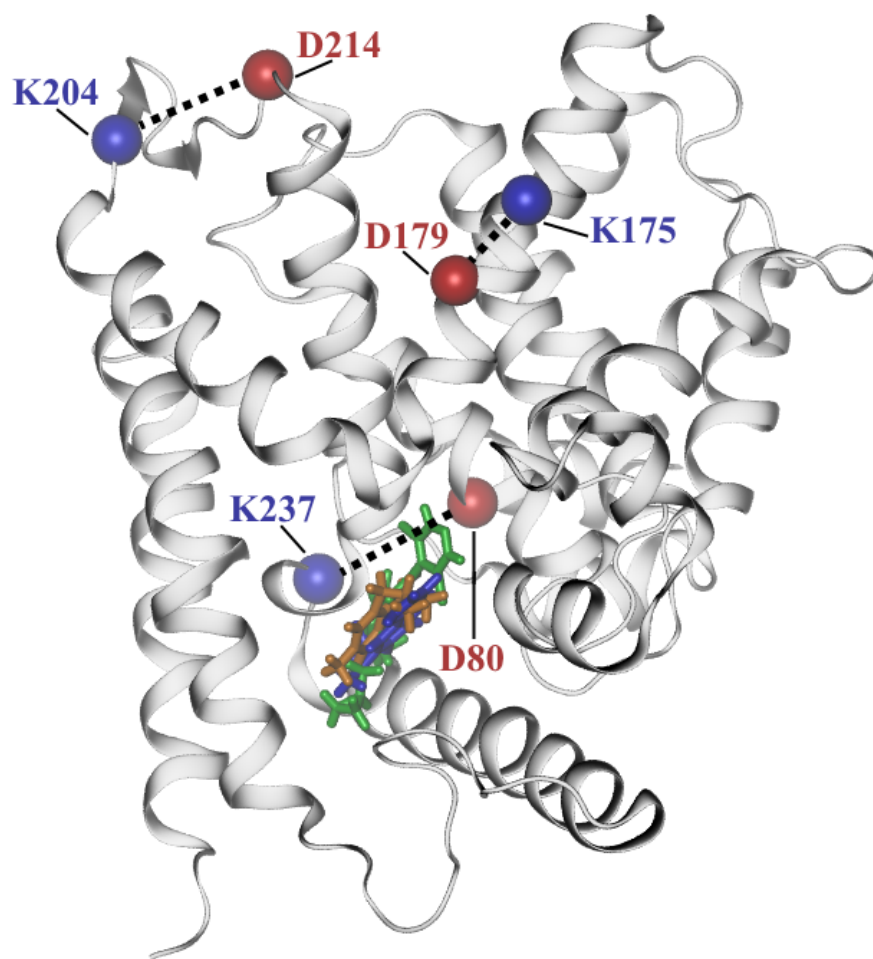


Figure 6.10. *C. elegans* PDE4 catalytic domain illustrating three conserved salt-bridges. Residues participating in each salt-bridge are colored and labeled. The three inhibitors are shown as sticks.

in comparison to roflumilast, which appears to be stably bound, largely in a single state.

In addition, we measured the interatomic distance (d_2 in Fig 6.3) between the oxygen atom of the inhibitors to the side-chains of Tyr253 (*C. elegans*) or Phe506 (human PDE4D). In *C. elegans* PDE4, the distance distributions are bimodal and span a larger distance range

418 ($\sim 3-9$ Å) for zardaverine and IBMX in comparison to a unimodal and narrower ($\sim 4-6$ Å) distribution for roflumilast. In human PDE4D, the distributions are unimodal for roflumilast and zardaverine and bimodal for IBMX, with mean values distinct from those in *C. elegans* PDE4. Overall, the measurements on these distances suggest distinct positioning of inhibitors in the proximity of Phe506 and Gln535 residues in human PDE4D and the corresponding residues Tyr253 and Gln282 in *C. elegans*. Specifically, roflumilast is significantly more stable than other inhibitors in the binding pocket of the human PDE4D and showed a higher non-bonded interaction energy with the Gln535 residue in human PDE4D (Fig 6.1c).

To further probe per-residue perturbations on binding of each inhibitor in both enzymes, we have computed the per-residue root-mean-squared fluctuation (RMSF) of the liganded enzyme structures (top panels in Figs 6.4a-b) and the change in per-residue RMSF relative to their unliganded apo-forms (Δ RMSF) (bottom panels in Figs 6.4a-b). Among binding pocket residues, we observed that ligand binding increased fluctuations in Val334 and Met439 in human PDE4D (corresponding to Val81 and Met186 in *C. elegans* PDE4). However, the residues located in loops connecting $\alpha 5-\alpha 6$, and $\alpha 11-\alpha 12$ helices are more stabilized by the ligands in human PDE4D in comparison to *C. elegans* PDE4. The residues located in the M loop between $\alpha 8$ and $\alpha 9$ helices are more stabilized in *C. elegans* PDE4 by zardaverine and roflumilast and to a lesser extent by IBMX in comparison to human PDE4D. Residue Phe506 (human)/Tyr253 (*C. elegans*) is located in $\alpha 14$ -helix which appear more stabilized by ligands in *C. elegans* PDE4 in comparison to human PDE4D. Residue Gln535 (human)/Gln282 (*C. elegans*) is located in the $\alpha 15$ -helix which is perturbed to a greater extent in human PDE4 than *C. elegans* PDE4 (Fig 6.4). The fluctuations in residues of the binding pocket as observed in the RMSF analyses are correlated with the analyses of non-bonded interaction energies.

To further investigate whether the higher flexibility of the α 14-helix in human PDE4D complexes with bound ligands affects the motion of the residues belonging to the α 15-helix, we calculated the dynamic cross-correlation matrix for the C_α atoms in all MD trajectories. For human PDE4D, the correlation matrices showed neither significant positive correlation nor significant anti-correlation between the residues of the α 14-helix (highlighted by the dashed-lines in Figs 6.5-6.6) and the residues of the α 15-helix (highlighted by the solid-lines in Figs 6.5-6.6). However, we find that the motion of residues in the α 14 and α 15 helices are marginally more correlated in *C. elegans* PDE4 in comparison to human PDE4D. The correlation between the α 14 and α 15 helices is mostly found between neighboring residues of Tyr253 (*C. elegans*) and Gln282 (*C. elegans*). We also observed (Fig 6.5) significantly higher positive residue-residue (C_α - C_α) correlation within human PDE4D complexes with IBMX and zardaverine in comparison to *C. elegans* PDE4, whereas the complexes with roflumilast showed significantly lower positive correlation in human PDE4D in comparison to *C. elegans* PDE4. This indicates that roflumilast induces a different pattern of correlated motions in the protein backbone in comparison to IBMX and zardaverine, and comparable to those of the apo states (Fig 6.6).

To better understand the effect of ligands on the protein structure outside the binding site, we identified all possible salt-bridging interactions within human PDE4D and *C. elegans* PDE4 (Figs 6.7-6.10). Qualitatively, the salt-bridging interactions are observed to occur with a lower frequency in human PDE4D in comparison to *C. elegans* PDE4. We also found a smaller number of salt-bridging interactions in human PDE4D complexed with roflumilast, but these salt-bridging interactions were comparatively stable for longer times during simulations. Furthermore, we identified three salt-bridge pairs conserved between human PDE4D and *C. elegans* PDE4 with higher occupancy number: D179-K175 (*C. elegans*), D204-K214 (*C. elegans*), and D80-K237 (*C. elegans*) (Fig 6.7, and labeled in Figs 6.8 and 6.9). D179-K175 (*C. elegans*) is an intra-helical salt-bridge in the α 11-helix, and

D204-K214 (*C. elegans*) is in the loop connecting the $\alpha 12$ and $\alpha 13$ helices. The D179-K175 (*C. elegans*) interaction pair in the human PDE4D complexes with ligands showed a lower occupancy in comparison to apo human PDE4D. The occupancy of the D204-K214 (*C. elegans*) is higher in the apo-human PDE4D in comparison to *C. elegans* PDE4, whereas it has a higher occupancy in *C. elegans* PDE4 complexes with ligands in comparison to human PDE4D complexes with ligands (Figs 6.7-6.10). The D80-K237 (*C. elegans*) salt-bridge is located near the binding pocket, the D80 residue is in the $\alpha 6$ -helix and the K237 residue is in the loop connecting $\alpha 13$ and $\alpha 14$. I observed that the occupancy of the D80-K237 (*C. elegans*) salt-bridge was significantly suppressed by roflumilast in human PDE4D in comparison to zardaverine and IBMX, while the occupancy of this salt-bridge is not affected by the presence of IBMX and zardaverine (Figs 6.7-6.10).

Both dynamic cross-correlation analysis and salt-bridging interactions revealed allosteric effects of each ligand on the protein structure. Unlike IBMX and zardaverine, roflumilast induced distinct patterns of structural perturbations outside of the binding pocket for human PDE4D compared with *C. elegans* PDE4. Specifically, we observed lower residue-residue correlations for roflumilast in comparison to zardaverine and IBMX in human PDE4D in comparison to *C. elegans* PDE4. While we observed overall a smaller number of salt-bridging interactions in human PDE4D in comparison to *C. elegans* PDE4, the salt-bridging interactions in human PDE4D were significantly more stable for roflumilast. In contrast, roflumilast significantly perturbed some salt-bridging interactions (D88-K237) more than zardaverine and IBMX in *C. elegans* PDE4. Therefore, we suggest that modulation of salt-bridging interactions could be one of the factors that contribute to an altered binding affinity of an inhibitor among different protein isoforms. Taken together, these structural analyses provide a molecular basis for better understanding differential binding of inhibitory compounds in the human PDE4 versus *C. elegans* PDE4 catalytic domain.

6.4.2 Atomistic simulations provide insight into altered pharmacological properties

The substantial differences in the primary sequence of nematode and vertebrate PDEs and pharmacological results revealed significant changes in binding affinities of compounds that were designed for human PDE4D. To gain further insights into differences in the ligand binding sites of *C. elegans* PDE4 and human PDE4D that could explain the reduced affinity of *C. elegans* PDE4 for compounds optimized as human PDE inhibitors, I used homology models and all-atom explicit-solvent MD simulations. The use of homology models has been successfully used in previous studies to identify amino acid residues that are responsible for differences in binding of inhibitors to PDE5 and PDE6 [238]. From the 32 amino acid residues that we defined as constituting the inhibitor-binding site, only five sites differed between the two enzymes and four of those were conservative substitutions that preserved the polar or hydrophobic nature at its position and thus are unlikely to drastically change the binding conformation or energy. However, I observed differences in nonbonded interaction energies due to the movement of inhibitors in the vicinity of residue Phe506 in *H. sapiens* (corresponding to Tyr253 in *C. elegans*). Overall, this Tyr residue contributes significantly more total non-bonded interaction energy than the Phe in the same position for all three inhibitors (Figs H.1 and H.2), likely due to the hydrogen bonding that results from the addition of a hydroxyl group at the 4-C of the aromatic ring.

For IBMX, the Phe to Tyr substitution appears to have little impact on the overall binding of IBMX to either human PDE4D or *C. elegans* PDE4. This is likely a result of IBMX interacting solely with the hydrophobic pocket of PDE4, as suggested by IBIS [245, 246]. Despite the polar Tyr residue coordinating to the ketone at position 6 in the purine ring of IBMX in *C. elegans*, our MD results indicate that the interactions and conformation of IBMX are very similar in the two PDE4 catalytic domains. This is consistent with the ob-

Inhibitor	Human IC ₅₀ (μM)	<i>C. elegans</i> IC ₅₀ (μM)	Fold difference
IBMX	15.8 ± 1.7 (n=4)	34.1 ± 8.7 (n=5)	2
Zardaverine	1.9 ± 0.6 (n=5)	146 ± 34 (n=8)	77
Roflumilast	0.0046 ± 0.0006 (n=3)	0.73 ± 0.13 (n=5)	159

Figure 6.11. Enzyme activity was tested over a range of inhibitor concentrations with 1 μM cAMP substrate 323 concentration. The dose-response relationship was fit to a 3-parameter logistic equation to obtain the 324 IC₅₀ and the standard error of the mean for the indicated number of experiments (This figure was created by my co-author Kevin Schuster).

served similarity in IC₅₀ values for IBMX with the two enzymes (Fig 6.11).

In contrast, the MD simulations suggest that the binding conformation of zardaverine or roflumilast are altered as a result of the substitution of Tyr253 for Phe506 at this site in the binding pocket of the two enzymes, consequently inducing a different pattern of correlated motions in the protein backbone. In *C. elegans*, the hydroxyl group of the Tyr residue coordinates strongly with the methoxyphenyl and cyclopropylmethoxyl group of zardaverine and roflumilast, respectively. This increase in energy contribution appears to result in a displacement of both ligands away from the hydrophobic sub-pocket that further disrupts stabilization by the conserved glutamine residues (Gln535 in human PDE4; Fig 6.1c).

It has been previously reported that Tyr495, Phe506, Gln535 are critical for stabilizing PDE4 inhibitors [237]. This disruption of the desired binding conformation in *C. elegans* could partially explain the reduced IC₅₀ values for these two compounds (Fig 6.11). While per-residue fluctuations are also found to be correlated with non-bonded interaction energy

analyses, other analyses (e.g. interatomic distances and residue-residue correlations) suggest that, among the three inhibitors, roflumilast is more stable in the binding pockets and induces a distinct pattern of correlated motions in comparison to zardaverine and IBMX. In summary, these MD analyses highlight the importance of considering not only differences in residue substitutions (e.g. Tyr253 in *C. elegans* vs. Phe506 in *H. sapiens*) but also allosteric perturbations and overall inhibitor stabilization of catalytic domain conformation in future efforts to design and optimize nematode-specific PDE inhibitor compounds using in silico approaches such as virtual screening and fragment-based drug design [247, 248].

6.5 Conclusion

MD simulations were used to support the hypothesis that the nematode PDE enzyme family differs sufficiently from the vertebrate PDE orthologs to validate the feasibility of developing PDE inhibitor compounds as potent and selective anthelmintics/nematicides. While analysis of the differences in the amino acid sequence or structure-activity relationships for selected PDE inhibitor compounds did not immediately identify which sites of interaction may have been disrupted in *C. elegans* PDE4 for inhibitors designed for human PDE4D, MD simulations revealed the importance of Phe506 (human)/Tyr253 (*C. elegans*) substitution and demonstrated that changes in the conformation of the catalytic domain may collectively lead to inhibitor discrimination in the binding pocket, based on the following analyses: (1) non-bonded interaction energy analysis; (2) changes in the ligand orientation in the binding pocket; (3) RMSF analysis; (4) cross correlation analysis; and (5) salt-bridge interaction analysis.

Collectively, the results indicate that future efforts to discover inhibitor compounds specifically targeting nematode PDE4 must take into consideration not only the molecular architecture of the inhibitor binding site, but also the conformational dynamics of the entire

catalytic domain of the enzyme. Insights gained from this study will advance efforts to rationally design inhibitor compounds that selectively and potently inhibit plant and animal parasitic nematode, PDEs to disrupt their lifecycle, thereby enhancing public health and agricultural productivity.

6.6 Publications

The work described in this chapter has resulted in the following journal article:

- Schuster, K.D., Mohammadi, M., Cahill, K.B., Matte, S.L., Maillet, A.D., Vashisth, H., and Cote, R.H. (2019). Pharmacological and molecular dynamics analyses of differences in inhibitor binding to human and nematode PDE4: implications for management of parasitic nematodes. *PLoS ONE* 14(3): e0214554.

CHAPTER 7

CONCLUSIONS AND FUTURE RESEARCH DIRECTIONS

In this thesis, I used MD simulations described in section 2.2 and enhanced sampling techniques such as TAMM described in section 2.4.1, and single-sweep method [141] described in sections 2.4.1, 2.4.1.1, and 2.4.1.2. Additionally, I used various conformational metrics calculations based on MD simulations (e.g. RMSD, RMSF, DCC, and non-bonded interaction energy). The overall understanding gained from work presented in this thesis could further guide mutagenesis studies in [FeFe]-hydrogenase or drug discovery efforts aimed at RGS proteins and nematode PDEs.

The primary outcome of this thesis work has been an enhanced understanding of the protein-inhibitor interactions in the studied cases. Specifically, the studies of protein conformational changes show transient states (beyond static protein states) that play a key role in inhibitor recognition and migration within protein-cavities. By characterizing the interconnected pathways of inhibitory gases within [FeFe]-hydrogenase, more effective mutations in residues could be performed for increasing the enzyme tolerance to inhibitory gases. The studies of inhibitory compounds of RGS4 protein provide enhanced view of the mechanism of action of TDZD inhibitors among isoforms of RGS proteins, and further elucidate the alternative binding mechanisms for these allosteric inhibitors and correlations between their potencies and the protein dynamics. The studies of human PDE4D inhibitors and *C. elegans* PDE4 provide enhanced insight into effects of inhibitors on the catalytic domain of the enzyme.

In chapter 3, I studied the interconnected network of pathways of inhibitory gases (O_2 and CO) within [FeFe]-hydrogenase. The differences in the free-energy levels of pathways between O_2 and CO show that any residue mutation directed toward one gas might affect the enzyme structure that may affect the interconnected network of the other gases (e.g. hydrogen). In addition to the challenge of increasing tolerance of the enzyme to inhibitory gases, one should consider that the successful mutagenesis studies would require characterization of pathways of other molecules that compete with O_2 and CO such as H_2 and H_2O within the enzyme. Therefore, one future work may be the studies of interconnected network of pathways of H_2 and H_2O within [FeFe]-hydrogenase. The final step in understanding the network of pathways of all of these molecules in the enzyme would be the computation of kinetic rates across each migration pathway for each gas/solvent molecule. One can conduct the calculation for finding the kinetic rates, and further enhance the experimental mutagenesis studies aimed at H_2 production from [FeFe]-hydrogenase.

In chapters 4 and 5, I studied the covalent inhibitors of RGS proteins that inhibit the RGS/ G_α protein-protein interaction. Here, the computational studies were solely directed toward the RGS proteins and not when they are bound to G_α proteins. The differences in potency of the inhibitor molecules across isoforms of RGS proteins suggest similar differences may exist in the interaction of the RGS-inhibitor complex with G_α subunits. The final goal of these inhibitors is to prevent the formation of the RGS- G_α complex. Therefore, one future work may be the MD simulation of the RGS- G_α complex with and without inhibitor molecules. Another future work may be the studies to enhance our view of binding and unbinding rate [249] of the inhibitory compounds to RGS proteins.

In chapter 6, I studied the PDE inhibitor compounds designed for human PDE4D on *C. elegans* PDE4. In this study, I found key sites in the catalytic domain of the enzyme for interaction with the selected PDE inhibitory compounds. The knowledge of differences

in the conformation of the catalytic domain may be supplemented with the calculation of free-energy of binding of inhibitory molecules. The free-energy of binding of compounds can be computed using MM/PBSA and MM/GBSA methods [250]. Another future work may be the computation of network of pathways of inhibitory compounds within the catalytic domain. This study can be performed using the single-sweep method [141] described in sections 2.4.1, 2.4.1.1, and 2.4.1.2.

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APPENDIX A

MATLAB SCRIPTS

A.1 Sample Code for Free-energy reconstruction using Radial Basis Functions

The following script can be used for reconstruction of the free-energy surface (FES) as a linear combination of Gaussian radial basis functions (see section 2.4.1.1) [141].

```
1 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
2 % The sample script is tested for PMF reconstruction %
3 % (c) Luca Maragliano, Istituto Italiano di Tecnologia, Genoa, Italy %
4 % Modified (Lines 97-127)and compiled by M. Mohammadi %
5 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
6 clear all
7 load meanforces_new.dat
8 zkcsa=meanforces_new;
9 n1 = size(zkcsa,1);
10 zc=zkcsa(1:n1,1:3);
11 xc=zc';
12 dz=zkcsa(1:n1,4:6);
13 fc=-dz';
14 %% plot centers & forces
15 plot(.5*(xc(1,:)+xc(2,:)),xc(3,:),'.')
16 figure
17 plot3(xc(1,:),xc(2,:),xc(3,:),'.b')
18 hold on
19 quiver3(xc(1,:),xc(2,:),xc(3,:),fc(1,:),fc(2,:),fc(3,:),'r')
20 grid on
21 axis equal
22 axis vis3d
23 xlabel('z1')
24 ylabel('z2')
25 zlabel('z3')
26 %%
27 d1 = norm(xc(:,1)-xc(:,2))
28 [ac,sigma] = reconstruct2(xc,fc);
29 %% build potential
30 n1 = size(xc,2);
31 bord=.5;
32 sigma0 = sigma(1,1);
33 sigma=sigma0*ones(1,n1);
```

```

34  xmin=min(xc(1,:))-bord;
35  xmax=max(xc(1,:))+bord;
36  ymin=min(xc(2,:))-bord;
37  ymax=max(xc(2,:))+bord;
38  zmin=min(xc(3,:))-bord;
39  zmax=max(xc(3,:))+bord;
40  [xx,yy,zz] = meshgrid(linspace(xmin,xmax,100),...
41      linspace(ymin,ymax,100),linspace(zmin,zmax,100));
42  clear r2;
43  V3 = xx.*0;
44      for j=1:n1
45          r2 = (xx-xc(1,j)).^2+(yy-xc(2,j)).^2 + (zz-xc(3,j)).^2;
46          V3 = V3 + ac(j)*exp(-r2/sigma(j)^2*0.5);
47      end;
48  V3m=V3-min(V3(:));
49  %% plot potential
50  figure
51  t=0.
52  p = patch(isosurface(xx,yy,zz,V3m,t));
53  isonormals(xx,yy,zz,V3m,p)
54  set(p,'FaceColor','red','EdgeColor','none');
55  camlight
56  %axis([-22 18 -12 10 -10 18])
57  view([-156 16])
58  t=1
59  p = patch(isosurface(xx,yy,zz,V3m,t));
60  isonormals(xx,yy,zz,V3m,p)
61  set(p,'FaceColor','yellow','EdgeColor','none');
62  t=6
63  p = patch(isosurface(xx,yy,zz,V3m,t));
64  isonormals(xx,yy,zz,V3m,p)
65  set(p,'FaceColor','blue','EdgeColor','none');
66  alpha(.4)
67  t=7.7
68  p = patch(isosurface(xx,yy,zz,V3m,t));
69  isonormals(xx,yy,zz,V3m,p)
70  set(p,'FaceColor','green','EdgeColor','none');
71  alpha(.4)
72  grid on
73  %axis equal
74  axis vis3d
75  xlabel('z1')
76  ylabel('z2')
77  zlabel('z3')
78  function [ac,sigma] = reconstruct2(xc,fc)
79  % xc is an nd x n1 matrix with the centers coordinate:
80  % nd is the dimensionality of the free energy space,
81  % n1 is the number of centers
82  % fc is a nd x n1 matrix with the forces
83  % the outputs are: ac, the coefficients of the radial-basis fcts, and
84  % sigma, the width of the radial-basis fcts.
85  % the function also plots the relative residual and prints the condition
86  % number of the matrix A at the optimal sigma
87  nd = size(xc,1);

```

```

88 n1 = size(xc,2);
89 % distance between centers
90 d1 = norm(xc(:,1)-xc(:,2));
91 bb = reshape(fc',nd*n1,1);
92 AA = zeros(n1*nd,n1);
93 % minimization loop
94 % sigmav = linspace(0.5*d1,2*d1,le2);
95 % sigmav = linspace(5*pi,6*pi,le1);
96 sigmav = linspace(d1,2*d1,2e2);
97 ss1 = size(sigmav,2);
98 resvrel = Inf(1,ss1);
99     B1 = xc'*xc;
100     r2 = diag(B1)*ones(1,n1)+ones(n1,1)*diag(B1)'-2*B1;
101     for i=1:ss1
102         sigma = sigmav(i);
103         sigma2 = 0.5./sigma.^2;
104         rr = exp(-r2.*sigma2);
105         for j = 1:nd
106             AA((j-1)*n1+1:j*n1,:) = (xc(j,:)'*ones(1,n1)-ones(n1,1)*xc(j
                ,:)).*rr;
107         end
108         AA = AA./sigma.^2;
109         if rank(AA)<n1; break; end;
110         gg = AA\bb;
111         resrelv(i) = norm(AA*gg-bb)/norm(bb);
112         resv(i) = norm(AA*gg-bb);
113         if i>1; if resrelv(i)>resrelv(i-1); break; end; end;
114         fprintf('iteration number = %3d; sigma = %f; residual = %f; cond.
                number = %e\n',...
115                 i,sigmav(i),resv(i)/n1,cond(AA)')
116     end
117 figure(1);clf;
118 plot(sigmav(1:i-1),resv(1:i-1)/n1,'o-')
119 set(gca,'FontSize',16);
120 title('residual per center |AA*gg-bb|/n1','FontSize',16)
121 xlabel('\sigma','FontSize',16)
122 ylabel('residual per center','FontSize',16)
123 [i1 i2] = min(resrelv);
124 sigma = sigmav(i2);
125 %% reconstruction step
126     sigma2 = 0.5/sigma^2;
127     rr = exp(-r2*sigma2);
128     for j = 1:nd
129         AA((j-1)*n1+1:j*n1,:) = (xc(j,:)'*ones(1,n1)-ones(n1,1)*xc(j,:)).*
                rr;
130     end
131     AA = AA/sigma^2;
132 % ac are the coefficients in the radial-basis representation of the free
133 % energy
134 ac = AA\bb;
135 sigma = sigma*ones(1,n1);
136 fprintf('\n')
137 fprintf('# of centers = %3d; optimal sigma = %f; rel. residual = %f; cond.
        number = %e\n',...

```

```

138     n1, sigma(1,1), norm(AA*ac-bb)/norm(bb), cond(AA) )
139 fprintf('# of centers = %3d; optimal sigma = %f; residual / cents = %f; cond.
    number = %e\n', ...
140     n1, sigma(1,1), norm(AA*ac-bb)/n1, cond(AA) )
141 fprintf('\n')
142 end

```

The following is the input file (“meanforces_new.dat”) containing the coordinates of the centers and the corresponding mean-force values.

```

1  1.470000 -2.120000 -7.710000 -9.785590 4.990049 -0.927257
2  1.470000 -2.120000 -8.280000 -9.812816 4.286304 -0.473017
3  1.560000 -2.410000 -9.890000 -6.371151 0.564879 0.349201
4  1.670000 -2.270000 -8.780000 -4.508113 1.930594 -0.806485
5  1.780000 -2.170000 -7.310000 -4.063549 1.879347 3.359805
6  1.800000 -2.210000 -9.440000 -3.680580 3.214869 -2.117216
7  1.840000 -1.670000 -5.830000 -6.623003 5.578274 3.747816
8  1.880000 -2.770000 -9.020000 -0.032609 -4.296424 0.694036
9  1.890000 -2.390000 -10.450000 -1.470600 0.430322 0.394696
10 1.890000 -2.870000 -7.610000 0.880890 -4.438182 3.351721
11 10.160000 0.380000 -5.280000 0.329397 -12.581186 5.083637
12 10.180000 2.250000 -5.310000 -0.890966 7.292744 -0.146103
13 10.210000 1.030000 -5.470000 -1.288463 -0.375553 0.266809
14 10.210000 2.750000 -5.890000 -0.472463 5.437764 -6.729431
15 10.250000 1.750000 -4.570000 -0.497051 2.693496 10.859090
16 10.320000 2.090000 -6.060000 0.531865 7.153510 -7.271102
17 10.330000 1.710000 -5.160000 -1.305893 5.445777 2.823317
18 10.360000 1.310000 -4.320000 -1.538975 -0.944368 11.377527
19 10.380000 0.900000 -6.190000 -0.067118 -1.471345 -5.396713
20 10.400000 0.270000 -6.030000 -1.254845 -9.305528 -0.620820
21 10.400000 3.840000 -5.520000 -1.538517 1.110011 -1.950121
22 10.470000 2.190000 -6.600000 -0.763081 8.472491 -8.433132
23 10.520000 1.420000 -6.180000 -0.929578 5.664027 -6.990566
24 10.530000 3.020000 -6.600000 -0.656510 2.564594 -5.712497
25 10.610000 2.920000 -5.320000 -0.558751 1.545318 0.086698
26 10.700000 1.360000 -5.350000 -0.630059 3.129134 0.481480
27 10.730000 2.560000 -4.840000 -0.295766 5.813913 5.884236
28 10.780000 0.960000 -6.870000 -0.274745 2.583686 -6.032612
29 10.790000 0.510000 -4.670000 -0.604078 -10.895482 12.771118
30 10.790000 1.000000 -5.860000 -1.229582 0.004796 -3.664434
31 10.800000 0.510000 -5.400000 -0.466494 -10.689803 1.947389
32 10.910000 -0.160000 -6.220000 -0.513956 -7.874482 -1.876247
33 10.950000 2.820000 -5.840000 -0.296766 2.573710 -5.011045
34 11.010000 2.040000 -6.080000 -0.665404 7.711624 -7.861041
35 11.020000 1.010000 -4.930000 -0.517248 -1.643624 8.777645
36 11.080000 1.460000 -6.130000 -0.271051 6.087383 -6.318860
37 11.130000 1.840000 -4.860000 -0.373715 5.104568 7.103900
38 11.180000 1.800000 -4.330000 -0.208607 2.583141 8.694338
39 11.190000 1.040000 -4.400000 -0.188699 -3.668468 12.136767
40 11.200000 3.000000 -4.150000 -0.625936 3.359043 4.409662
41 11.220000 0.650000 -6.370000 -0.324947 -3.706987 -5.135758

```

42	11.230000	0.670000	-5.740000	-0.090179	-4.240865	-0.574641
43	11.410000	0.460000	-4.650000	-0.181423	-10.398926	12.095535
44	11.420000	2.350000	-5.390000	-1.269670	7.852689	-1.725379
45	11.430000	1.370000	-4.980000	0.033519	2.122918	6.755236
46	11.500000	1.580000	-7.050000	-0.227878	7.616850	-6.348085
47	11.570000	1.310000	-6.220000	-0.112256	5.913774	-6.734546
48	11.600000	1.580000	-5.550000	-0.216202	5.784586	-2.326657
49	11.600000	1.960000	-4.030000	-0.249035	2.671545	6.284807
50	11.690000	1.880000	-4.560000	-0.120765	3.634639	9.053479
51	11.720000	0.630000	-5.560000	0.308886	-7.141562	0.668203
52	11.810000	0.560000	-6.690000	-0.013099	-4.162477	-6.105688
53	11.860000	2.000000	-6.290000	0.119091	7.982332	-7.276114
54	11.930000	0.930000	-4.470000	-0.032969	-3.756202	12.239510
55	11.930000	1.350000	-6.620000	-0.201110	4.670160	-7.090052
56	11.970000	1.100000	-4.970000	-0.309290	-1.145643	7.833637
57	11.980000	0.460000	-5.080000	-0.350485	-10.708945	6.540088
58	12.020000	2.730000	-5.770000	-0.098955	4.416381	-5.927511
59	12.080000	1.620000	-5.820000	-0.434442	6.149895	-5.374572
60	12.180000	2.170000	-5.330000	-0.993701	7.545303	-0.509524
61	12.190000	0.190000	-4.510000	-0.504289	-10.683225	11.935753
62	12.230000	0.630000	-6.110000	-0.674993	-3.808016	-3.392482
63	12.230000	3.020000	-5.070000	-0.723733	0.707082	3.381953
64	12.250000	0.330000	-5.650000	-0.225927	-12.308530	-0.174033
65	12.260000	3.780000	-5.510000	-0.832179	-0.529532	-1.538337
66	12.310000	1.130000	-6.200000	-0.592559	2.412709	-6.282849
67	12.320000	1.060000	-5.570000	-0.609169	1.395056	-1.438595
68	12.400000	0.200000	-6.610000	-0.288645	-6.167720	-5.226179
69	12.400000	1.530000	-5.140000	-1.124196	3.479462	3.749120
70	12.410000	0.720000	-4.790000	-0.562788	-7.009042	11.317314
71	12.480000	3.250000	-3.740000	-0.503941	0.605541	0.389208
72	2.020000	-1.880000	-8.320000	-2.020616	5.168058	-0.834765
73	2.020000	-2.030000	-9.960000	-1.620296	4.503261	1.162102
74	2.040000	-2.450000	-7.980000	0.073659	-0.876309	-1.257002
75	2.040000	-2.660000	-8.540000	1.127652	-4.122007	0.563482
76	2.050000	-1.690000	-6.490000	-4.047220	7.870610	0.628047
77	2.060000	-2.550000	-9.740000	0.461196	-2.254338	-1.683790
78	2.080000	-2.210000	-6.390000	-0.676476	-5.404862	2.885443
79	2.100000	-1.840000	-7.000000	-2.777419	3.180030	5.175967
80	2.100000	-1.930000	-7.660000	-1.785870	5.148049	-1.229410
81	2.150000	-1.220000	-8.520000	-4.318319	15.486063	-2.222311
82	2.170000	-1.900000	-9.020000	-1.299470	6.323809	-0.809971
83	2.320000	-3.040000	-9.810000	4.911817	-6.365213	1.211002
84	2.350000	-2.420000	-8.860000	2.637871	-3.227060	-0.664330
85	2.420000	-1.790000	-5.860000	-3.144319	1.032389	3.857516
86	2.470000	-2.990000	-7.880000	6.539072	-6.672430	0.550712
87	2.500000	-1.620000	-9.330000	-3.054592	9.481162	-2.412334
88	2.500000	-2.640000	-10.150000	3.527522	-6.034656	1.114888
89	2.540000	-3.000000	-8.740000	6.265618	-9.942799	0.130870
90	2.550000	-1.970000	-6.810000	0.439530	-2.067726	6.908680
91	2.550000	-1.970000	-9.800000	-0.940055	4.023278	0.039894
92	2.560000	-1.910000	-8.680000	-0.381733	3.782025	-0.168624
93	2.560000	-2.240000	-7.460000	0.969988	-3.825991	1.928867
94	2.600000	-2.340000	-8.400000	1.891884	-3.560201	0.438040
95	2.610000	-0.760000	-8.700000	-5.530802	14.634261	-1.594754

96	2.660000	-1.900000	-7.930000	-1.109495	2.146963	-1.607371
97	2.670000	-3.250000	-8.280000	3.157939	-3.486741	-1.781799
98	2.710000	-2.110000	-10.280000	0.270010	0.184418	1.445464
99	2.740000	-1.930000	-6.300000	-2.207743	-1.718878	1.570019
100	2.780000	-2.420000	-6.710000	3.678854	-4.129190	2.967938
101	2.850000	-1.730000	-7.360000	-2.375362	3.999146	0.111007
102	2.900000	-1.990000	-9.300000	-0.972644	0.563813	-1.047719
103	2.940000	-1.140000	-6.380000	-5.559317	11.024156	-1.825307
104	2.940000	-1.540000	-8.770000	-5.549079	8.218361	-0.677302
105	2.990000	-2.530000	-9.380000	4.276616	-8.264305	-0.789744
106	3.000000	-1.030000	-5.170000	-7.008680	6.814184	5.735588
107	3.010000	-2.820000	-8.790000	0.951670	-5.546646	1.371082
108	3.030000	-1.870000	-9.860000	-2.863721	1.601365	-0.521689
109	3.070000	-2.750000	-7.780000	1.635657	-4.038255	0.695380
110	3.080000	-2.260000	-7.680000	1.876295	-7.781997	0.635802
111	3.120000	-1.260000	-7.090000	-4.728191	9.068522	0.430987
112	3.130000	-1.680000	-6.790000	-0.885649	-0.628200	5.988530
113	3.130000	-2.610000	-7.260000	0.925797	-10.283233	0.693989
114	3.210000	-1.440000	-7.960000	-4.838812	6.758490	-2.426404
115	3.220000	-1.800000	-5.430000	-3.362104	-7.704335	7.602412
116	3.230000	-1.300000	-9.980000	-6.704521	9.601720	1.268847
117	3.230000	-1.550000	-10.460000	-4.596837	6.005295	0.493376
118	3.230000	-2.210000	-5.820000	-5.736266	-10.563909	7.563982
119	3.230000	-2.540000	-8.260000	3.307465	-10.315886	-0.258160
120	3.270000	-4.240000	-7.850000	-1.445826	-0.675387	4.983644
121	3.290000	-2.000000	-8.920000	0.503675	-3.282220	-0.698471
122	3.300000	-1.480000	-9.140000	-6.800216	6.508543	-0.800016
123	3.310000	-1.840000	-7.530000	-1.662242	-3.074127	2.088493
124	3.330000	-2.060000	-10.490000	0.113140	-4.688570	0.718155
125	3.380000	-2.040000	-8.160000	-0.152428	-6.648301	-0.529211
126	3.400000	-3.430000	-8.030000	-1.692701	-2.792933	-1.202766
127	3.430000	-2.630000	-10.140000	-0.738600	-6.175346	1.232766
128	3.450000	-1.510000	-5.990000	0.115159	-5.966683	7.742243
129	3.450000	-4.290000	-8.640000	0.279142	-1.377745	-1.265891
130	3.500000	-2.380000	-9.630000	3.154513	-9.911830	-1.664458
131	3.510000	-2.370000	-6.460000	2.927198	-15.818146	2.007969
132	3.510000	-3.070000	-6.260000	-3.161518	-18.433693	11.068323
133	3.520000	-1.020000	-6.770000	-3.424899	8.114120	1.784130
134	3.540000	-0.800000	-7.560000	-7.025334	9.459388	-0.289985
135	3.540000	-2.730000	-8.810000	-3.272266	-5.279181	-1.238291
136	3.550000	-1.580000	-9.640000	-2.130468	2.722263	-2.072569
137	3.550000	-2.970000	-8.140000	-3.163119	-3.457108	-0.964448
138	3.590000	-1.390000	-7.590000	-3.249656	3.947318	1.177240
139	3.610000	-2.290000	-8.580000	2.735607	-9.292341	-0.888280
140	3.630000	-0.150000	-7.410000	-7.766808	9.189951	-0.475645
141	3.640000	-0.810000	-8.380000	-7.267304	11.606289	-2.581550
142	3.650000	-1.570000	-6.470000	1.162952	-4.141665	7.503477
143	3.660000	-0.980000	-8.920000	-4.490235	9.457324	-0.109234
144	3.680000	-1.280000	-10.350000	-2.787161	6.820652	0.874688
145	3.690000	0.120000	-5.780000	-6.372637	2.239877	-0.305453
146	3.700000	-2.130000	-6.940000	1.409731	-7.795042	6.556067
147	3.760000	-2.080000	-10.040000	2.099290	-4.971352	0.903574
148	3.780000	-2.570000	-7.580000	-3.065761	-4.391506	2.295742
149	3.790000	-1.520000	-7.120000	2.604494	-0.294103	2.959327

150	3.790000	-2.820000	-9.520000	-3.012678	-5.863907	-1.780807
151	3.790000	-3.140000	-10.210000	-2.723597	-5.628188	0.808057
152	3.820000	-1.370000	-8.510000	0.117832	3.867569	-2.052579
153	3.820000	-4.460000	-8.210000	0.772448	-5.439717	3.618620
154	3.820000	-4.820000	-10.020000	1.015474	1.153112	0.444472
155	3.830000	-1.730000	-8.100000	2.382660	-3.871209	-1.386598
156	3.870000	-1.950000	-9.050000	-0.778882	-0.102451	-0.631741
157	3.880000	-0.440000	-6.240000	0.349090	-1.640122	-1.758444
158	3.890000	-1.010000	-5.130000	-2.813109	4.148113	5.958722
159	3.970000	-1.710000	-10.310000	4.767743	-3.070539	1.235396
160	4.000000	-1.740000	-7.560000	5.206293	-5.599441	1.658151
161	4.020000	-1.460000	-9.310000	4.442137	0.699896	-1.343906
162	4.040000	-2.440000	-8.220000	-2.390232	-4.325837	0.072488
163	4.060000	-0.830000	-7.240000	-0.403484	5.645580	3.967544
164	4.060000	-1.120000	-6.810000	3.328613	2.868491	4.076771
165	4.080000	-2.310000	-6.500000	-1.381826	-9.075440	1.049330
166	4.100000	-1.250000	-9.850000	3.599599	3.537593	0.336824
167	4.130000	-0.520000	-7.920000	-3.225067	7.669725	-3.988029
168	4.140000	-1.880000	-5.730000	-1.389061	-3.635261	5.614716
169	4.140000	-2.350000	-10.420000	-0.085269	-2.765264	0.968779
170	4.140000	-4.950000	-9.090000	1.041130	2.274613	-0.229843
171	4.150000	-1.090000	-7.750000	3.057731	4.131807	-0.640695
172	4.160000	-1.170000	-6.270000	4.723494	-1.666095	4.575410
173	4.170000	-2.600000	-8.710000	0.294758	-6.096677	-0.550167
174	4.180000	-2.590000	-9.890000	0.494912	-5.032680	0.418716
175	4.180000	-3.370000	-8.830000	1.951023	-5.788995	-1.321839
176	4.190000	-1.180000	-8.880000	4.772582	3.340467	0.007011
177	4.190000	-2.940000	-6.440000	0.285906	-8.209997	7.013270
178	4.190000	-3.600000	-6.990000	0.495233	-8.116226	4.869160
179	4.190000	0.310000	-5.920000	-5.875549	-5.888099	0.831154
180	4.190000	0.730000	-7.170000	-5.480088	8.180393	-3.603162
181	4.200000	-1.830000	-8.590000	3.647515	-1.879768	-1.340319
182	4.210000	-0.080000	-7.580000	-5.382754	7.006465	-2.415755
183	4.210000	-2.030000	-6.990000	2.688249	-3.912597	5.742791
184	4.220000	-0.530000	-8.490000	-2.488143	7.118298	-2.298807
185	4.220000	-2.420000	-9.200000	0.319968	-4.373995	-0.413001
186	4.230000	-5.100000	-10.240000	1.872499	2.316112	1.220353
187	4.240000	-1.040000	-5.610000	4.805478	-6.281026	5.954221
188	4.270000	-2.680000	-7.600000	-0.027437	-11.381655	-2.144056
189	4.270000	-3.020000	-8.060000	2.405375	-5.822452	-0.615311
190	4.290000	-1.440000	-8.250000	6.813849	-0.321762	-1.935248
191	4.290000	-1.780000	-9.630000	3.458713	0.795462	-1.249165
192	4.310000	-0.460000	-6.810000	-0.954586	2.871875	3.820897
193	4.320000	-1.510000	-7.110000	7.691676	-3.497254	4.865637
194	4.320000	0.180000	-6.550000	-1.403350	-5.783934	4.431160
195	4.330000	-2.050000	-7.820000	0.688616	-1.286833	1.513804
196	4.340000	-0.330000	-5.310000	0.268397	-3.452452	1.169911
197	4.340000	-0.960000	-10.370000	4.821383	4.327572	0.824766
198	4.380000	-1.900000	-5.110000	-0.042070	-10.142058	15.852719
199	4.380000	0.730000	-5.330000	-6.830413	0.573644	3.417525
200	4.390000	-0.580000	-5.820000	-0.806802	2.181452	-0.866439
201	4.400000	-2.100000	-6.140000	1.038072	-6.348476	2.449267
202	4.400000	-4.880000	-9.820000	2.368387	2.986602	-0.159066
203	4.420000	0.720000	-6.090000	-6.008680	3.253870	-2.514157

204	4.440000	0.380000	-4.930000	-4.666213	-8.625240	7.486384
205	4.460000	-0.720000	-7.590000	3.086580	3.153877	-0.737242
206	4.480000	-1.750000	-6.590000	1.308740	4.159887	-1.034718
207	4.500000	-1.510000	-10.160000	6.703095	-0.391844	1.595561
208	4.500000	-2.900000	-10.280000	4.377151	-7.758788	0.832771
209	4.510000	-2.480000	-6.970000	3.726110	-1.892715	2.051724
210	4.520000	1.200000	-7.160000	-7.566908	16.874228	-4.073073
211	4.540000	-0.900000	-6.950000	6.342940	0.016932	4.588568
212	4.570000	-2.890000	-8.860000	4.609167	-8.046249	0.561148
213	4.600000	-2.180000	-5.630000	1.735749	-12.507701	8.796243
214	4.630000	-5.300000	-9.440000	4.628955	0.755702	-0.262614
215	4.640000	-2.290000	-8.530000	3.397215	-3.900287	-0.563893
216	4.670000	-1.630000	-5.690000	2.511834	1.990933	3.175886
217	4.670000	-2.460000	-7.800000	4.120627	-6.442432	1.278630
218	4.670000	0.200000	-7.330000	-2.734028	3.678339	1.063417
219	4.680000	-0.420000	-6.270000	-0.037796	1.173538	-2.268326
220	4.680000	-1.140000	-8.700000	7.454345	2.141599	-2.258181
221	4.680000	-1.300000	-6.360000	5.297482	4.457757	-3.508386
222	4.690000	-0.260000	-7.660000	1.360590	2.313850	-2.382844
223	4.700000	-1.520000	-7.760000	1.766637	6.600609	0.940029
224	4.710000	-5.460000	-10.200000	5.508190	-1.893559	1.051645
225	4.720000	-0.910000	-5.050000	0.169857	3.918847	8.182927
226	4.720000	-2.660000	-9.470000	5.272360	-6.445159	-1.442031
227	4.720000	-2.720000	-8.270000	3.769986	-7.816313	-0.133827
228	4.750000	-1.020000	-7.930000	7.276638	1.175270	-2.425706
229	4.760000	-2.060000	-8.970000	3.356487	0.279907	-0.837866
230	4.760000	1.090000	-5.500000	-6.101597	9.359843	0.203263
231	4.800000	0.430000	-6.420000	0.467138	-4.873386	4.792695
232	4.810000	-1.190000	-7.370000	7.241381	-0.728716	3.841719
233	4.820000	0.290000	-5.750000	-1.780848	-8.969543	0.837479
234	4.830000	-1.330000	-9.520000	2.955362	11.655929	-1.018340
235	4.830000	-2.100000	-6.690000	4.741772	-2.847821	5.423159
236	4.850000	-0.320000	-7.010000	3.437222	-1.181741	4.264129
237	4.860000	0.970000	-6.130000	-3.615946	6.265728	-3.571929
238	4.880000	-2.590000	-10.120000	4.951982	-6.331773	1.560828
239	4.920000	-2.550000	-6.420000	4.253723	-14.110398	2.705759
240	4.940000	0.930000	-7.110000	-5.500820	10.785299	3.028568
241	4.950000	-1.180000	-5.420000	1.194126	6.554039	3.443597
242	4.960000	-1.930000	-7.230000	5.394983	-1.395456	3.952844
243	4.970000	-2.920000	-7.750000	6.492949	-7.132185	0.864996
244	4.980000	-2.040000	-9.800000	5.358382	0.239460	-0.313196
245	5.000000	-1.440000	-4.970000	1.933756	0.447480	15.704404
246	5.000000	-1.620000	-8.240000	6.362592	-0.586621	-1.766079
247	5.000000	0.600000	-5.180000	-1.855560	-5.553953	4.787424
248	5.020000	-2.020000	-5.890000	3.516523	-7.076170	2.975502
249	5.030000	1.460000	-7.160000	-5.047748	17.038799	-3.534401
250	5.050000	-1.390000	-8.940000	6.381571	1.575937	0.620047
251	5.070000	-0.890000	-5.920000	0.886655	9.224613	-2.262109
252	5.070000	-2.070000	-10.410000	5.444685	-1.265097	0.907437
253	5.140000	-0.550000	-5.220000	-0.858790	3.126429	2.643051
254	5.150000	0.850000	-4.730000	-1.016898	-2.063311	9.431469
255	5.160000	-0.250000	-6.150000	0.242950	2.200820	-3.348519
256	5.170000	-1.470000	-7.120000	3.543595	5.844263	0.879714
257	5.180000	-2.450000	-7.170000	7.247987	-5.202087	3.189444

258	5.190000	0.380000	-6.860000	-0.025898	1.250865	3.502156
259	5.200000	-0.090000	-5.570000	0.087813	-8.404875	0.473690
260	5.200000	0.520000	-6.110000	0.467435	-3.455571	-2.608300
261	5.210000	-2.150000	-9.350000	5.476206	-2.749068	-1.138429
262	5.260000	-1.230000	-6.280000	2.189622	9.663296	-2.963120
263	5.310000	0.540000	-7.630000	-1.874913	2.550372	-1.471901
264	5.330000	-0.110000	-7.510000	3.376774	-2.620245	-0.574225
265	5.340000	-2.220000	-8.420000	5.083711	-4.508308	-1.000604
266	5.370000	0.170000	-5.090000	0.232935	-9.805058	5.698012
267	5.380000	1.570000	-4.840000	-4.419427	12.692011	6.333938
268	5.400000	-1.930000	-7.850000	5.018345	-2.792842	1.909229
269	5.450000	1.030000	-5.270000	-0.577851	3.179862	2.316042
270	5.460000	0.110000	-6.470000	1.581735	-8.000470	-3.449772
271	5.480000	-1.860000	-6.220000	4.560303	-0.890183	-0.294788
272	5.520000	-0.370000	-7.090000	0.531156	-1.252364	-2.644332
273	5.520000	-2.330000	-10.350000	5.853322	-5.025993	0.854952
274	5.520000	1.510000	-5.440000	-2.230416	12.202447	-0.385406
275	5.530000	-0.960000	-7.550000	3.987823	11.770851	-3.538373
276	5.550000	-1.510000	-6.790000	4.475621	6.599496	-1.931426
277	5.550000	0.960000	-6.110000	0.789043	4.809904	-4.019577
278	5.560000	-1.630000	-8.310000	5.365530	4.559041	-1.996384
279	5.560000	-1.650000	-5.430000	3.212084	-2.608166	6.615633
280	5.570000	-1.950000	-8.850000	4.798322	-0.391977	-2.439367
281	5.640000	1.070000	-6.640000	1.207548	5.886533	-5.675743
282	5.650000	1.330000	-4.310000	-1.457491	4.520629	10.358261
283	5.660000	-2.510000	-8.050000	4.028168	-9.029806	-3.423799
284	5.710000	0.790000	-8.110000	-3.554358	2.904050	-2.298852
285	5.740000	0.240000	-8.080000	-0.721254	-4.413234	-1.459978
286	5.750000	-0.380000	-7.790000	-1.373982	-1.437502	-2.613933
287	5.750000	-1.420000	-6.080000	4.563731	5.754505	-2.253122
288	5.750000	-1.820000	-7.130000	5.698422	-1.387835	3.877744
289	5.750000	1.420000	-7.150000	-0.401374	11.007213	-3.888427
290	5.780000	-0.680000	-6.460000	-0.391152	4.965101	-3.784924
291	5.820000	0.560000	-5.630000	2.890539	-6.396539	0.541520
292	5.830000	-2.210000	-6.060000	4.781170	-11.652579	2.739130
293	5.850000	0.810000	-7.110000	2.555555	1.336511	-4.327075
294	5.860000	2.000000	-5.530000	-3.603476	14.668583	-2.166133
295	5.870000	0.140000	-4.960000	-1.419550	-9.399266	5.992914
296	5.930000	-0.380000	-5.590000	0.528337	0.490034	-1.294186
297	5.940000	1.650000	-6.150000	-1.009981	12.416338	-5.603166
298	5.950000	-1.260000	-5.540000	4.359454	4.469948	1.812648
299	5.990000	0.060000	-6.700000	3.291725	-7.901625	-4.060096
300	6.010000	-1.840000	-6.390000	5.871569	-1.725065	-1.739007
301	6.020000	0.570000	-6.260000	3.493596	-3.525321	-4.344670
302	6.020000	1.440000	-4.940000	-1.268812	7.358745	5.312328
303	6.030000	1.400000	-6.640000	-0.995089	8.623967	1.814714
304	6.060000	-1.730000	-7.900000	5.574838	0.245764	0.463545
305	6.130000	-1.300000	-6.810000	4.986794	8.583773	-3.595565
306	6.130000	-2.670000	-6.520000	4.532879	-16.382965	2.617609
307	6.130000	1.480000	-4.350000	-2.631996	4.853879	8.800497
308	6.140000	-0.570000	-7.490000	-0.952977	3.764521	-0.457009
309	6.140000	0.770000	-5.300000	0.380271	-6.085596	3.282303
310	6.190000	1.100000	-5.900000	-0.359627	3.592734	-4.143387
311	6.190000	1.250000	-8.110000	-7.029285	10.120116	-0.966697

312	6.230000	-1.340000	-5.050000	2.408447	-0.495421	12.519599
313	6.240000	-0.470000	-6.200000	-0.502339	2.151677	-3.715754
314	6.330000	-0.430000	-5.080000	0.280964	-0.235228	3.233716
315	6.330000	0.850000	-4.820000	-1.218289	-5.465954	7.861419
316	6.360000	-1.410000	-6.090000	3.669792	4.471261	-1.351240
317	6.360000	-1.980000	-5.870000	4.831958	-7.806591	2.895534
318	6.380000	0.900000	-7.150000	-1.442137	-0.576524	-5.142725
319	6.420000	-0.560000	-7.030000	3.069367	-2.227903	-3.250582
320	6.420000	0.630000	-5.900000	-1.046445	-5.798025	-2.383686
321	6.420000	1.570000	-5.680000	-2.960701	9.172733	-4.174515
322	6.450000	1.180000	-3.970000	-3.549352	-3.004582	10.652515
323	6.480000	-1.990000	-6.500000	4.043524	-2.847972	-1.121834
324	6.530000	0.710000	-6.570000	-0.588745	-4.027726	-5.542502
325	6.570000	-0.200000	-5.800000	1.504181	-6.073162	-1.061145
326	6.610000	-1.600000	-5.650000	5.287634	-1.337155	2.173533
327	6.620000	1.540000	-4.570000	-2.859815	3.541240	5.403079
328	6.630000	-1.090000	-6.780000	3.664742	8.018772	-3.706543
329	6.630000	1.710000	-5.240000	-3.770365	9.505173	0.621020
330	6.670000	1.100000	-5.110000	-1.049995	0.135023	3.169182
331	6.690000	0.520000	-5.060000	-0.075904	-10.513305	5.914184
332	6.810000	1.180000	-5.990000	-0.662750	2.993023	-5.205589
333	6.850000	0.840000	-4.420000	-0.099106	-7.133138	11.288634
334	6.850000	1.360000	-6.510000	-0.596096	6.408150	-6.173361
335	6.910000	1.340000	-5.520000	-0.534212	4.565864	-1.636322
336	6.910000	2.070000	-5.800000	-1.432949	9.057551	-5.034982
337	6.920000	2.570000	-5.580000	-4.829117	8.977786	-4.277987
338	7.010000	1.260000	-4.150000	0.055078	-1.672152	8.924435
339	7.050000	1.850000	-4.010000	0.046123	4.027859	7.862283
340	7.080000	-1.850000	-5.730000	3.822482	-7.600072	3.217529
341	7.100000	-0.730000	-7.070000	6.218815	8.665473	-2.926175
342	7.110000	-0.730000	-6.270000	3.795763	5.222723	-4.287915
343	7.170000	0.680000	-5.860000	1.859171	-6.265047	-2.351190
344	7.190000	0.350000	-5.350000	1.971966	-11.064725	2.725801
345	7.210000	1.670000	-7.190000	2.547221	9.167685	2.324850
346	7.320000	2.070000	-4.630000	2.390077	5.542002	4.430370
347	7.370000	1.450000	-5.200000	2.721028	5.053285	2.840806
348	7.390000	1.190000	-4.620000	3.878594	-0.887461	8.373271
349	7.440000	1.540000	-5.720000	3.241314	7.036184	-3.927956
350	7.450000	0.150000	-6.820000	5.234042	-11.940313	-5.242416
351	7.510000	1.390000	-6.380000	3.648288	5.388274	-6.766823
352	7.510000	2.470000	-5.500000	2.035588	1.841560	-2.741476
353	7.580000	1.020000	-6.000000	4.524123	-0.519741	-4.711188
354	7.630000	-1.830000	-5.690000	5.568798	-3.779837	4.768283
355	7.710000	0.890000	-4.910000	2.770515	-4.067473	7.296069
356	7.770000	2.660000	-6.110000	0.466204	-0.316426	-4.973325
357	7.810000	0.970000	-5.550000	1.861250	0.435538	-0.650927
358	7.820000	3.010000	-4.740000	-0.869983	2.620221	4.012755
359	7.850000	1.140000	-4.430000	2.293750	-1.102156	9.739676
360	7.870000	3.100000	-5.770000	2.315459	0.595940	-5.309249
361	7.930000	1.550000	-5.060000	0.771882	5.946854	5.398545
362	8.020000	2.270000	-4.920000	1.866762	3.743695	2.417959
363	8.230000	3.660000	-4.830000	-0.685348	7.291451	3.529174
364	8.320000	0.960000	-5.610000	0.484645	-2.857555	-1.676067
365	8.370000	0.620000	-6.460000	-0.238463	-5.273331	-5.725803

366	8.430000	1.950000	-5.410000	-1.539673	8.510531	-1.092240
367	8.510000	1.030000	-4.710000	-0.949979	-3.224230	11.938912
368	8.690000	1.470000	-6.100000	-0.672348	6.303684	-6.792371
369	8.850000	2.640000	-6.260000	-2.609535	6.346974	-8.074174
370	8.940000	0.350000	-4.760000	-1.009582	-12.140244	11.065896
371	8.970000	0.760000	-5.760000	-1.365027	-4.264435	-2.156946
372	8.980000	1.380000	-4.980000	-0.956414	2.911630	6.951530
373	8.990000	0.540000	-5.260000	-2.014932	-9.671615	4.194186
374	9.070000	0.010000	-5.560000	-2.203293	-13.727834	2.322717
375	9.100000	1.170000	-6.610000	-1.846779	3.245957	-6.790586
376	9.100000	1.270000	-5.700000	-1.061001	2.991233	-2.734217
377	9.120000	2.160000	-5.350000	-1.493177	7.751459	-1.027296
378	9.160000	1.780000	-6.050000	-1.327887	9.236950	-5.988685
379	9.180000	0.960000	-7.770000	-1.268384	1.385118	0.208353
380	9.220000	0.430000	-6.300000	-2.510071	-4.167695	-4.276783
381	9.430000	0.820000	-4.780000	-1.322281	-7.393560	11.321846
382	9.440000	0.390000	-7.060000	-1.679670	-4.249693	-6.689306
383	9.490000	1.170000	-6.290000	-2.293362	3.027558	-6.488503
384	9.510000	1.730000	-5.640000	-1.915270	6.374895	-3.513869
385	9.620000	1.010000	-5.320000	-1.155763	-1.444269	2.390249
386	9.730000	0.470000	-5.870000	-0.330852	-8.231715	-2.128090
387	9.840000	1.480000	-5.050000	-0.266062	3.405301	5.359532
388	9.880000	1.720000	-6.640000	0.204631	7.355597	-7.061658
389	9.920000	2.300000	-6.770000	-1.450918	7.423427	-7.131328
390	9.940000	3.490000	-5.280000	-2.341432	2.985128	0.767734
391	9.970000	4.160000	-5.750000	-2.319189	0.357524	-2.823965

The following output figures are generated from the script.

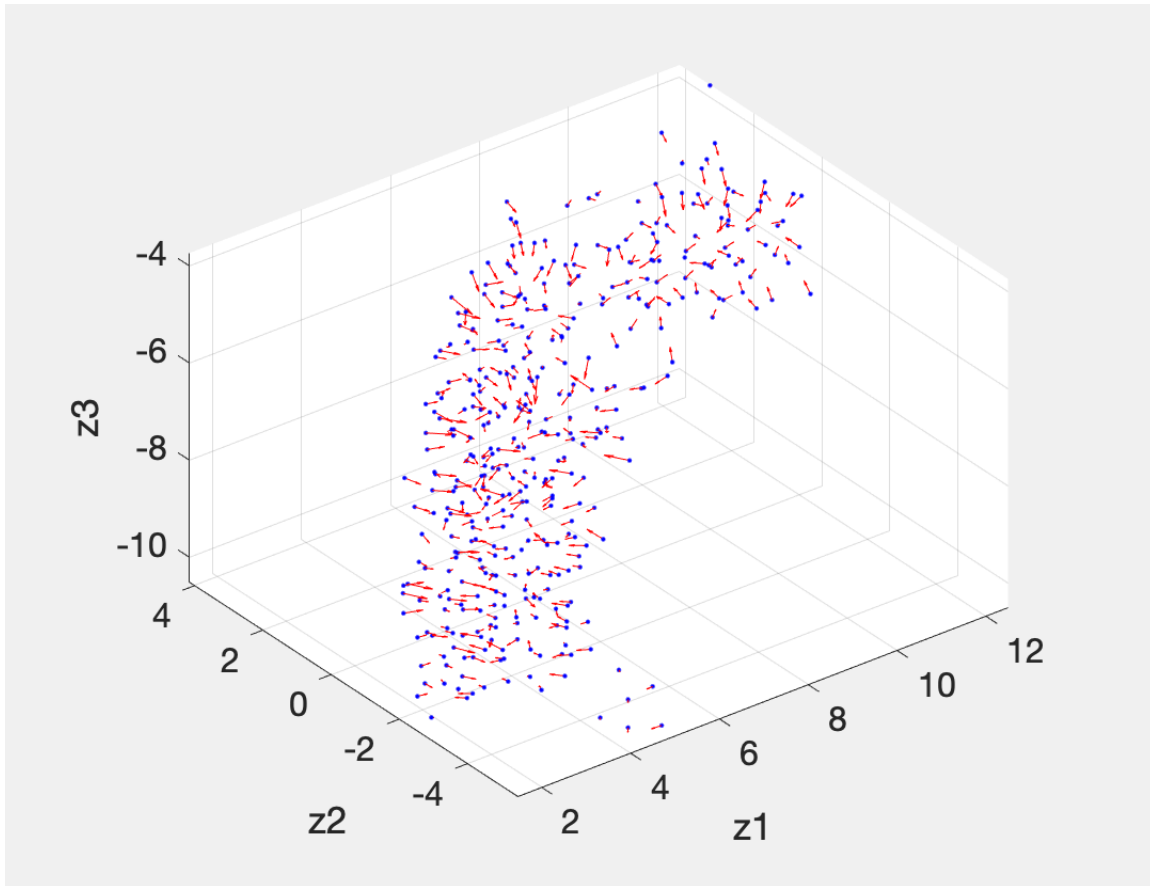


Figure A.1. The locations of centers are shown as blue points, and the mean-forces at each point are shown as red arrow.

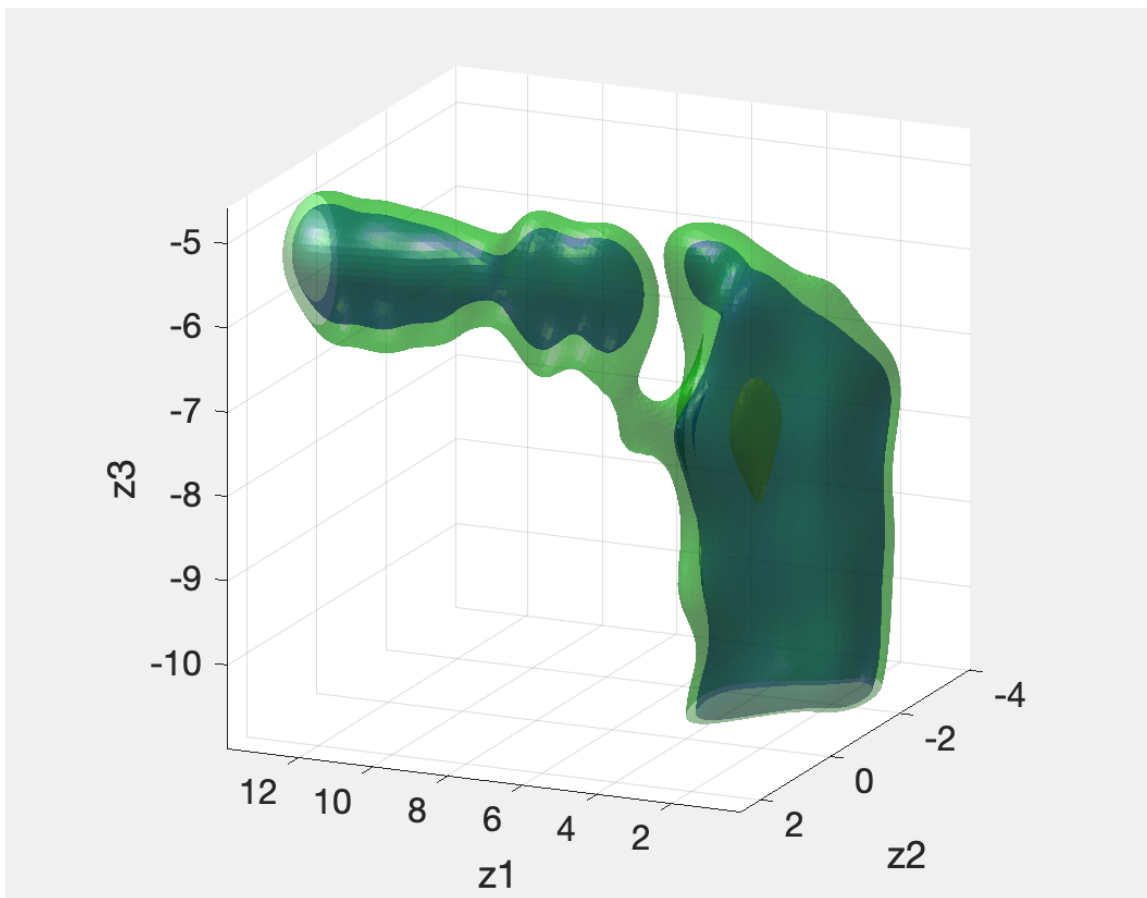


Figure A.2. The reconstructed 3D free energy surfaces are shown. The lower isosurfaces with lower energy values are shown with darker colors.

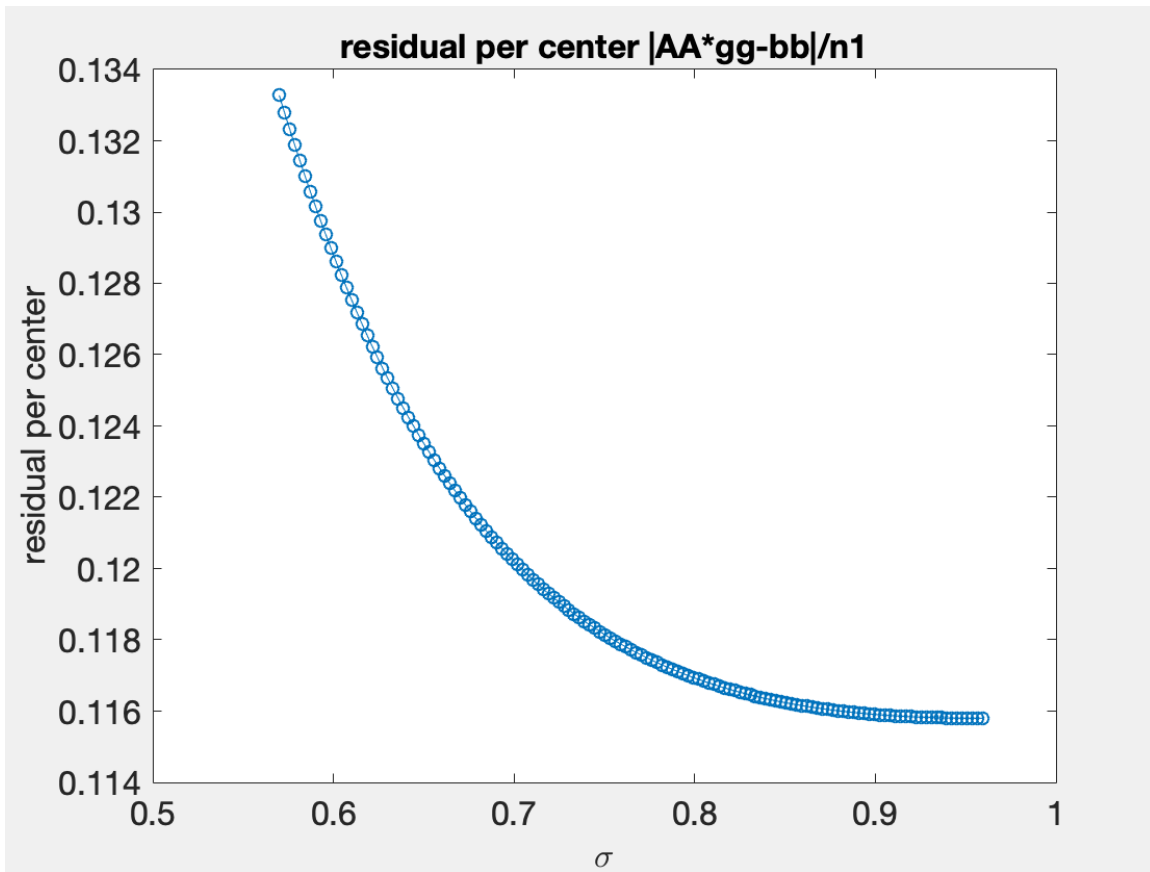


Figure A.3. The optimization profile of σ is shown.

A.2 Sample code for zero temperature string method (ZTS)

The minimum free-energy pathways (MFEP) can be determined using the ZTS method [149] (see section 2.4.1.2).

```
1 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
2 %% Zero-temperature string method (ZTS) code %
3 % The script is only tested for string %
4 % optimization, and nothing else. %
5 % (c) Eric Vanden-Eijnden, NYU, New York, NY %
6 % Modified (Lines 10-30) and compiled by M. Mohammadi %
7 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
8 clear all
9 set(0,'DefaultTextFontName','TimesRoman')
10 set(0,'DefaultAxesFontSize',16)
11 % max number of iterations
12 nstepmax = 2e3;
13 % frequency of plotting
14 nstepplot = 1e1;
15 % plot string every nstepplot if flag1 = 1
16 flag1 = 1;
17 % parameter used as stopping criterion
18 toll = 1e-7;
19 % number of images along the string (try from n1 = 3 up to n1 = 1e4)
20 n1 = 25;
21 % time-step (limited by the ODE step on line 83 & 84 but independent of n1)
22 h = 1e-4;
23 % end points of the initial string
24 % notice that they do NOT have to be at minima of V -- the method finds
25 % those automatically
26 xa = -1;
27 ya = 0.5;
28 xb = 0.7;
29 yb = 0.5;
30 % initialization
31 g1 = linspace(0,1,n1);
32 x = (xb-xa)*g1+xa;
33 y = (x-xa)*(yb-ya)/(xb-xa)+ya;
34 dx = x-circshift(x,[0 1]);
35 dy = y-circshift(y,[0 1]);
36 dx(1) = 0;
37 dy(1) = 0;
38 lxy = cumsum(sqrt(dx.^2+dy.^2));
39 lxy = lxy/lxy(n1);
40 x = interp1(lxy,x,g1);
41 y = interp1(lxy,y,g1);
42 xi = x;
43 yi = y;
44 % parameters in Mueller potential
45 aa = [-1 -1 -6.5 0.7];
46 bb = [0 0 11 0.6];
```

```

47 cc = [-10 -10 -6.5 0.7];
48 AA = [-200 -100 -170 15];
49 XX = [1 0 -0.5 -1];
50 YY = [0 0.5 1.5 1];
51 [xx,yy] = meshgrid(-1.5:0.01:1.2,-0.2:0.01:2);
52 V1 = AA(1)*exp(aa(1)*(xx-XX(1)).^2+bb(1)*(xx-XX(1)).*(yy-YY(1))+cc(1)*(yy-YY
    (1)).^2);
53 for j=2:4
54     V1 = V1 + AA(j)*exp(aa(j)*(xx-XX(j)).^2+bb(j)*(xx-XX(j)).*(yy-YY(j))+cc(j)
        )*(yy-YY(j)).^2);
55 end;
56 figure(1);clf;
57 contourf(xx,yy,min(V1,200),40);
58 hold on
59 plot(xi,yi,'.-w','MarkerSize',14)
60 set(gca,'XTick',-1.5:.5:1,'YTick',0:.5:2);
61 xlabel('x','FontAngle','italic');
62 ylabel('y','FontAngle','italic');
63 title('Initial string');
64 drawnow
65 figure(2);clf;
66 contourf(xx,yy,min(V1,200),40);
67 whitedots = line(x,y,'linestyle','none','marker','.', 'color','w','MarkerSize'
    ,14);
68 set(gca,'XTick',-1.5:.5:1,'YTick',0:.5:2);
69 xlabel('x','FontAngle','italic');
70 ylabel('y','FontAngle','italic');
71 title('String evolution')
72 drawnow
73 %% Main loop
74 for nstep=1:nstepmax
75 % calculation of the x and y-components of the force, dVx and dVy respectively
76     ee = AA(1)*exp(aa(1)*(x-XX(1)).^2+bb(1)*(x-XX(1)).*(y-YY(1))+cc(1)*(y-YY
        (1)).^2);
77     dVx = (2*aa(1)*(x-XX(1))+bb(1)*(y-YY(1))).*ee;
78     dVy = (bb(1)*(x-XX(1))+2*cc(1)*(y-YY(1))).*ee;
79     for j=2:4
80         ee = AA(j)*exp(aa(j)*(x-XX(j)).^2+bb(j)*(x-XX(j)).*(y-YY(j))+cc(j)*(y-
            YY(j)).^2);
81         dVx = dVx + (2*aa(j)*(x-XX(j))+bb(j)*(y-YY(j))).*ee;
82         dVy = dVy + (bb(j)*(x-XX(j))+2*cc(j)*(y-YY(j))).*ee;
83     end;
84     x0 = x;
85     y0 = y;
86 % string steps:
87 % 1. evolve
88     x = x - h*dVx;
89     y = y - h*dVy;
90 % 2. reparametrize
91     dx = x-circshift(x,[0 1]);
92     dy = y-circshift(y,[0 1]);
93     dx(1) = 0;
94     dy(1) = 0;
95     lxy = cumsum(sqrt(dx.^2+dy.^2));

```

```

96     lxy = lxy/lxy(n1);
97     x = interp1(lxy,x,g1);
98     y = interp1(lxy,y,g1);
99     if and(flag1 == 1,mod(nstep,nstepplot) == 0)
100         set(whitedots,'xdata',x,'ydata',y)
101         drawnow
102         pause(0.025)
103     end
104     tol = (norm(x-x0)+norm(y-y0))/n1;
105     if tol <= toll; break; end;
106 end
107 %% Output
108 fprintf('\n')
109     fprintf('\n')
110     fprintf('ZTS calculation with %d images\n',n1)
111 if tol > toll
112     fprintf('The calculation failed to converge after %d iterations\n',nstep)
113 else
114     fprintf('The calculation terminated after %d iterations\n',nstep)
115 end;
116 figure(2);clf;
117 contourf(xx,yy,min(V1,200),40);
118 hold on
119 plot(x,y,'.-w','MarkerSize',14)
120 set(gca,'XTick',-1.5:.5:1,'YTick',0:.5:2);
121 xlabel('x','FontAngle','italic');
122 ylabel('y','FontAngle','italic');
123 title('Final string');
124 drawnow
125 % Energy along MEP
126 tx = circshift(x,[0 -1])-circshift(x,[0 1]);
127 ty = circshift(y,[0 -1])-circshift(y,[0 1]);
128 % potential computed as integral of projection of gradV on string tangent
129 Vz=cumtrapz(tx.*dVx + ty.*dVy);
130 Vz=0.5*Vz;
131 Vz=Vz-min(Vz);
132 ntxy = sqrt(tx.*tx+ty.*ty);
133 tx = tx./ntxy;
134 ty = ty./ntxy;
135 % err is an estimate of the error between discretized MEP and actual one
136 % err scale as 1/n1
137 err = trapz(1-(tx.*dVx+ty.*dVy).^2./(dVx.*dVx+dVy.*dVy))/(n1-1);
138 fprintf('Estimate of difference between discretized MEP and actual one: %f\n',
    err)
139 %
140 % Reinterpolate string with lots of points to get accurate energy along it
141 g0 = linspace(0,1,1e3);
142 x0 = interp1(lxy,x,g0);
143 y0 = interp1(lxy,y,g0);
144 ee = AA(1)*exp(aa(1)*(x0-XX(1)).^2+bb(1)*(x0-XX(1)).*(y0-YY(1))+cc(1)*(y0-YY
    (1)).^2);
145 V0=ee;
146 dVx = (2*aa(1)*(x0-XX(1))+bb(1)*(y0-YY(1))).*ee;
147 dVy = (bb(1)*(x0-XX(1))+2*cc(1)*(y0-YY(1))).*ee;

```

```

148 for j=2:4
149     ee = AA(j)*exp(aa(j)*(x0-XX(j)).^2+bb(j)*(x0-XX(j)).*(y0-YY(j))+cc(j)*(y0-
        YY(j)).^2);
150     V0 = V0 + ee;
151     dVx = dVx + (2*aa(j)*(x0-XX(j))+bb(j)*(y0-YY(j))).*ee;
152     dVy = dVy + (bb(j)*(x0-XX(j))+2*cc(j)*(y0-YY(j))).*ee;
153 end;
154 V0=V0-min(V0);
155 figure(3);clf;
156 hold on
157 title('Energy along MEP');
158 plot(g1,Vz)
159 plot(g0,V0,'r')
160 box on
161 legend('Thermodynamic integration along string','Exact','Location','South')
162 legend('boxoff')

```

The following uotpot figures are generated from the zts script.

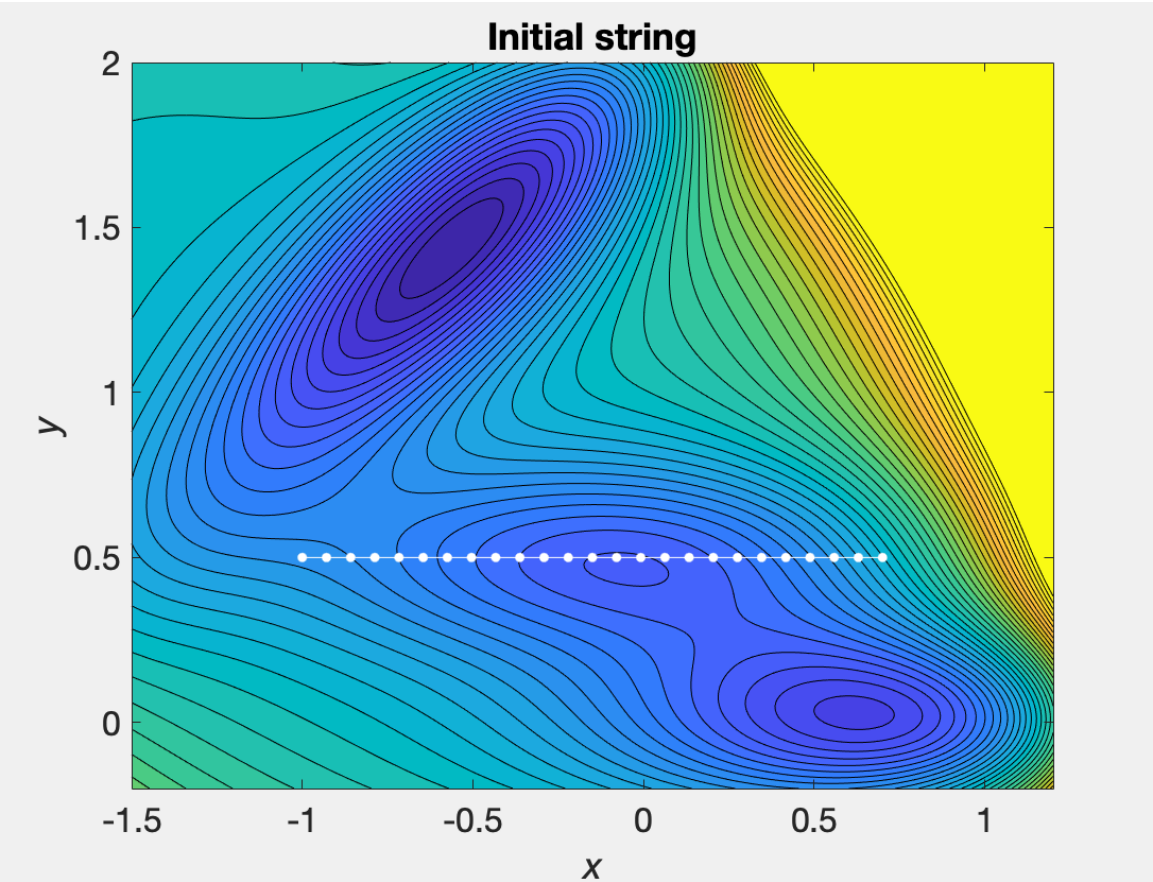


Figure A.4. The initial string before initiation of the script.

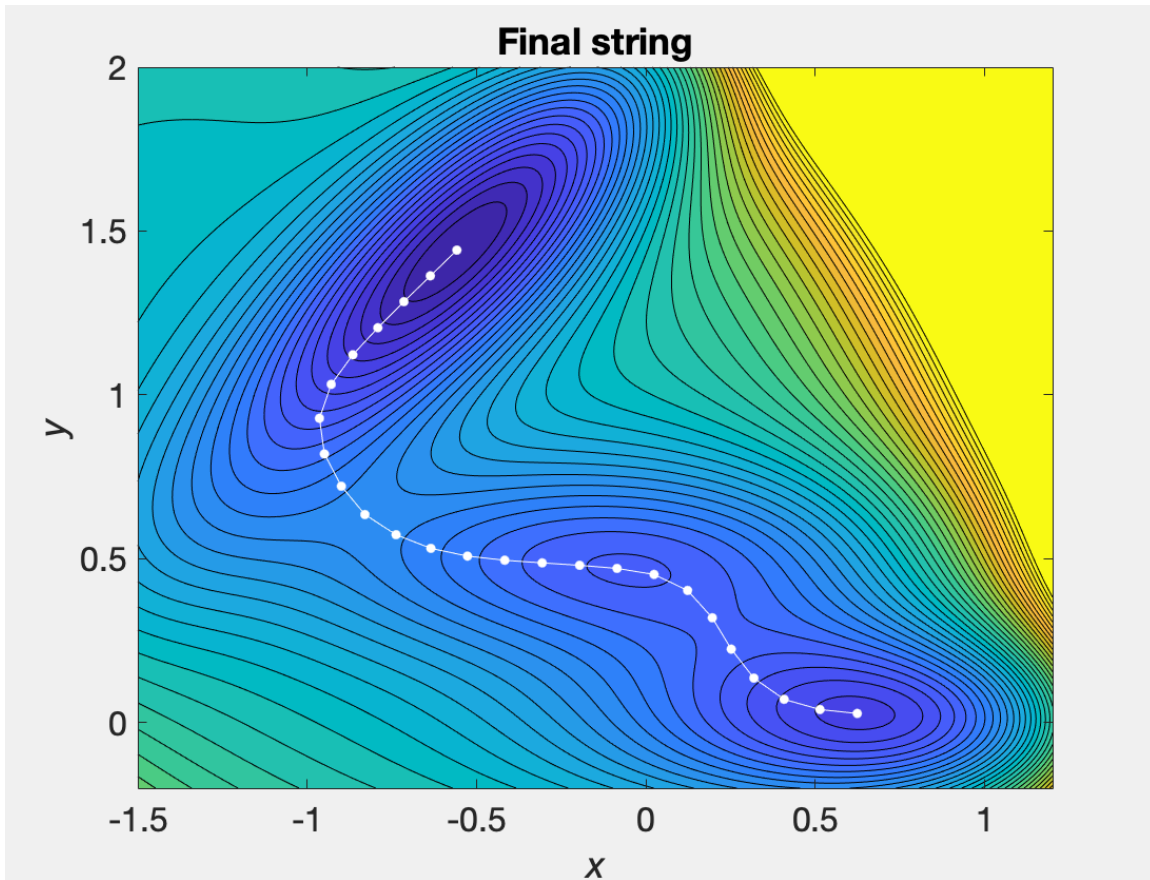


Figure A.5. The final string after completion of the script.

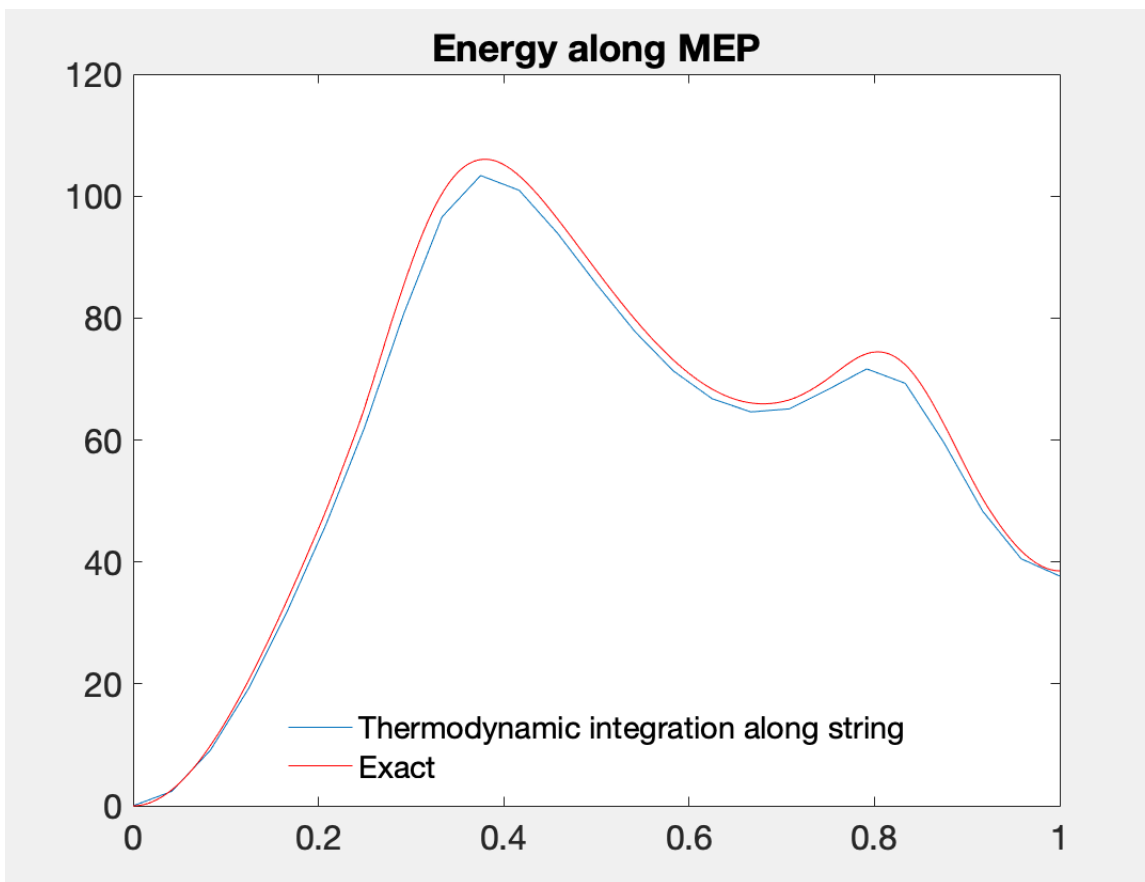


Figure A.6. The 2D PMF, showing the free-energy change across the final string.

APPENDIX B

TCL SCRIPTS

B.1 Sample Code for selecting centers

The exploration of the protein interior space via TAMD results in a dense cloud of ligand coordinates (section 2.4.1.1). I used the following script for selecting the centers within the protein matrix for mean-force calculation.

```
1 #####
2 ## Point Clustering (Version 3.0) ##
3 ## This script has been only tested for selecting positions of ##
4 ## ligand covering various areas of the protein by a distance ##
5 ## cut-off value. These centers could be used for mean-force ##
6 ## calculation described in chapter 3. ##
7 #####
8
9 set outfile [open ox.dat w];
10 set nf [molinfo top get numframes]
11 set sel [atomselect top "segname OX"]
12 set com_list {}
13 set refcom {}
14 set newcom {}
15 set com_list2 {}
16 set chosen_frames {}
17 proc lremove {list match} {
18     set idx_list [lsearch -all $list $match]
19     foreach idx [lreverse $idx_list] {
20         set list [lreplace $list $idx $idx]
21     }
22     return $list
23 }
24 for {set i 0} {$i < $nf} {incr i} {
25     set temp_ox {}
26     $sel frame $i
27     set com_ox [measure center $sel weight mass]
28     lappend temp_ox $com_ox
29     lappend com_list $temp_ox
30 }
31 set com_list2 $com_list
32 set size [llength $com_list]
33 set uni {}
34 set k 1
35 set g 1
```



```

36 puts "Initial total COM: $size "
37 set refcom0 {}
38 set newcom5 {}
39 set new2 {}
40 set refcom0 [lindex $com_list 0]
41 lappend uni $refcom0
42 set ref [lindex $refcom0 0]
43 for {set i 1} { $i < [llength $com_list] } { incr i } {
44     set newcom5 [lindex $com_list $i]
45     set new5 [lindex $newcom5 0]
46     set dd [vecdist $new5 $ref]
47     set g 1
48     if { $dd >= 2.5 } {
49         for {set j 0} { $j < [llength $uni] } { incr j } {
50             set newcom [lindex $uni $j]
51             set s1 [lsearch -all $com_list $newcom]
52             set new2 [lindex $com_list $s1]
53             set new3 [lindex $new2 0]
54             set dddd [vecdist $new5 $new3]
55             if { $dddd > 0 && $dddd < 3.0 } {
56                 set g 0
57             }
58             if { $g == 1 } {
59                 lappend uni $newcom5
60                 set refcom0 $newcom5
61                 set ref [lindex $refcom0 0]
62             }
63         }
64     }
65 set refcom1 {}
66 set newcom1 {}
67 # for {set r 0} { $r < [llength $uni] } { incr r } {
68 # set refcom [lindex $uni $r]
69 # set ref [lindex $refcom 0]
70 # for {set k 0} { $k < [llength $uni] } { incr k } {
71 # set newcom [lindex $uni $k]
72 # set new [lindex $newcom 0]
73 # set ddd [vecdist $new $ref]
74 # if { $ddd < 2.51 && $ddd > 0 } {
75 # puts "$r $k $ddd \ "
76 # }
77 # }
78 #set chosen_frames {}
79 draw delete all
80 for {set i 0} { $i < [llength $uni] } { incr i } {
81     set refcom1 [lindex $uni $i]
82     set s [lsearch -exact $com_list2 $refcom1]
83     lappend chosen_frames $s
84     puts $outfile "$s $refcom1 \ "
85     set ref [lindex $refcom1 0]
86     # puts "$ref"
87     draw color red
88     draw sphere $ref radius 0.5
89 # animate write pdb ${s}.pdb beg $s end $s

```

```

90 }
91 set size [llength $uni]
92 puts "Final total COM: $size "
93 close $outfile
94 puts "frames chosen are: $chosen_frames"

```

The following typical output data file shows how coordinates of the centers are shown in "ox.dat".

1	0	7.1351	85.4152	76.9845
2	156	7.0427	82.6447	75.2778
3	191	4.10029	81.7408	76.387
4	206	1.98574	82.4221	74.2702
5	211	-0.356568	84.6737	74.5849
6	247	2.60047	84.1852	71.0736
7	290	5.08997	82.5053	72.0253
8	338	3.94729	86.4296	74.0891
9	733	7.2174	85.6596	73.1914
10	2580	5.02837	87.7864	76.841
11	2728	-0.823304	81.2919	75.0923
12	2762	-1.12958	82.0141	71.8188
13	5280	-2.86603	82.4115	69.3282
14	5372	3.35803	79.5987	72.587
15	5414	2.97107	79.2217	77.5833
16	6315	7.91289	82.0163	71.0953
17	6363	10.2266	84.9011	76.4793
18	7211	-3.16977	85.1162	73.4462
19	7358	11.4153	85.3845	71.0905
20	7455	9.24379	87.798	71.7588
21	7903	8.47505	88.5709	77.1181
22	7909	6.31513	88.2072	79.6226
23	9779	-0.598473	83.0012	67.4441
24	10410	9.0344	85.1746	79.3655
25	12211	-5.08412	84.4729	70.8897
26	12220	-3.59797	87.3056	70.978
27	12290	-0.088821	86.7477	72.3108
28	12298	-3.06006	89.8068	72.6884
29	12341	-3.01366	89.6928	75.7531
30	12480	-5.77485	88.7539	68.2326
31	12898	12.6081	85.8561	78.2051
32	13693	1.18522	78.7637	75.204
33	14215	-0.810877	87.7092	75.1475
34	14429	-6.34716	88.7802	73.7231
35	14447	-8.36029	89.9232	70.7358
36	14480	-10.123	92.6985	69.7389
37	14486	-7.82316	91.0486	67.4253
38	14912	7.04897	82.8316	79.1575
39	15000	-2.01535	78.1745	82.9957
40	15002	0.43548	77.3959	81.3736
41	15071	-1.03908	76.8404	78.4965
42	15157	2.13227	74.1964	82.2809
43	15195	-0.71501	73.5197	83.2207
44	15225	-4.70946	77.7479	84.7837
45	15326	-1.78592	78.8838	86.0158

46	15343	0.56893	77.6384	88.2306
47	15362	3.55881	77.3868	88.7032
48	15709	-3.18309	78.6372	79.9735
49	15839	1.5584	78.7178	84.0627
50	15860	2.89458	76.0987	86.0084
51	15993	-0.0992651	75.958	85.5896
52	16422	3.42125	78.6341	81.5722
53	16782	-1.61375	76.4536	74.5981
54	16789	1.50726	75.1149	73.5906
55	17207	-5.92617	79.5784	79.0649
56	17214	-7.95139	80.7052	80.9874
57	17268	-8.08941	77.5371	77.6604
58	17340	-5.80022	80.3998	83.165
59	17725	-6.94952	77.566	82.4301
60	19413	-4.72763	78.4518	88.2021
61	19475	-4.35415	80.6275	85.8048
62	19979	-6.52378	83.3175	83.6854
63	20307	-5.43351	82.4185	80.2318
64	20409	-7.71119	80.8883	85.9511
65	20695	-3.38147	73.8707	84.8278
66	20705	-4.76022	75.2638	82.0215
67	20914	0.252021	77.2081	91.2985
68	21256	-5.5409	76.0863	78.6958
69	21746	-0.272053	78.0284	71.8256
70	21756	-1.79839	75.0986	71.1601
71	21776	-0.0915667	77.1357	68.8906
72	21790	2.99189	75.6838	70.3448
73	24403	0.986457	80.5307	87.4241
74	25209	-9.63618	76.6114	80.1423
75	25221	-9.37912	79.2163	83.6128
76	25227	-9.65463	75.566	84.4429
77	25708	-9.59972	80.313	77.5849
78	25711	-12.2069	81.7913	78.0756
79	25718	-10.143	84.0751	78.0835
80	25762	-13.0779	82.0587	74.9921
81	25779	-12.054	84.5249	73.4851
82	25788	-9.05334	83.682	73.6551
83	25807	-7.94901	85.7751	76.29
84	25828	-7.8868	85.6192	79.4377
85	25841	-7.32407	87.5269	82.0868
86	25850	-4.27308	86.5287	81.0078
87	25868	-3.79369	84.1495	78.3871
88	25898	-1.34692	84.6306	80.4013
89	26697	-5.72582	72.2962	83.6206
90	27184	-1.77601	70.1802	82.7375
91	27280	-10.2919	78.37	75.3051
92	27284	-8.37523	75.7658	75.0296
93	27292	-7.24866	73.1177	76.4577
94	27673	-2.45819	74.3541	77.3623
95	27725	-11.6236	75.728	76.2037
96	27746	-8.19481	76.1339	71.4791
97	27756	-9.86193	73.1767	70.6255
98	27775	-9.97434	74.6542	67.6926
99	27783	-8.19527	73.6674	64.7872

100	27790	-5.55023	72.5659	66.2084
101	27807	-6.28999	75.873	66.0208
102	27842	-5.88627	78.481	67.5104
103	27860	-3.21747	75.8665	68.5378
104	28443	0.903283	74.8342	89.3632
105	29289	-4.8102	74.0598	74.455
106	29368	0.854802	73.6352	70.882
107	29711	-12.2702	77.3339	82.1299
108	29915	-1.24518	79.9476	90.7084
109	30000	14.9703	93.7883	94.715
110	30001	13.7146	90.0929	93.2049
111	30094	12.8017	89.2761	90.1279
112	30173	11.2603	87.7012	88.0044
113	30210	8.00348	88.852	88.0841
114	30484	15.8416	90.9543	91.1764
115	30727	8.20999	85.9679	89.5407
116	31210	10.3387	90.5778	91.4926
117	33805	18.1487	93.1588	90.0519
118	33925	16.4352	91.4383	87.9478
119	34363	15.0601	87.0594	89.7112
120	34412	10.7545	86.6302	91.5772
121	35210	5.08066	90.0703	88.5942
122	35228	5.18456	87.7336	90.5426
123	35341	5.45502	90.6981	91.5825
124	35396	11.006	86.8117	94.9752
125	35430	10.7992	83.8352	94.1879
126	35746	13.8828	92.6993	89.1085
127	36407	13.4493	87.3994	96.8475
128	36478	10.896	89.8585	94.5575
129	36846	7.85948	87.7463	92.0737
130	37360	16.2352	90.8196	94.6945
131	38409	5.73247	88.3188	95.3002
132	38764	9.00363	88.541	85.083
133	39764	5.5124	92.4056	86.5484
134	39831	4.29214	93.023	89.9503
135	40256	4.35224	89.4817	85.5505
136	40264	3.62183	92.2582	83.6031
137	40335	3.44229	94.5511	87.4491
138	41144	17.4094	88.5241	88.3844
139	41281	5.85185	94.4184	84.1123
140	41478	3.05991	95.719	83.5125
141	41784	6.77811	92.354	82.1059
142	41841	5.2955	97.198	86.818
143	41861	8.15189	96.1233	87.371
144	41900	9.87323	98.124	89.372
145	41909	7.75953	96.6024	91.6622
146	41957	6.56776	98.4494	89.3902
147	41979	2.83732	98.934	88.2184
148	42363	11.3277	96.8024	86.7179
149	42822	3.7294	96.8739	90.24
150	42841	5.65124	98.6003	93.25
151	42863	9.12351	99.3475	92.9977
152	42909	5.87241	98.6047	96.8971
153	42930	3.9719	96.0604	96.4887

154	42978	2.7325	99.0494	96.0453
155	43114	14.4279	88.3379	87.025
156	43484	4.96169	99.4478	84.0545
157	43498	4.03853	100.659	81.1839
158	44184	10.2113	84.003	87.8207
159	45000	14.74	84.0507	71.6804
160	45043	15.5001	80.96	73.3332
161	45192	12.5429	82.7204	73.2925
162	45715	9.96288	91.3777	71.2315
163	45746	11.4495	91.1158	67.7748
164	45763	11.745	89.9624	64.6253
165	45786	14.1821	91.7683	64.7346
166	45842	16.8772	93.6042	66.9184
167	45850	19.6844	92.9532	65.7277
168	45863	23.0417	92.0531	66.1996
169	45909	20.4208	92.3354	68.8185
170	45925	19.267	90.9382	62.6974
171	45986	19.4953	89.1783	59.7148
172	46144	17.5247	83.8643	72.8341
173	46219	11.9341	86.5122	74.0527
174	46340	11.7277	88.4249	76.5302
175	46746	12.1384	88.1819	69.2111
176	46903	16.7171	86.4872	71.1689
177	46909	15.2958	83.7865	75.7012
178	47838	14.3498	92.3488	68.7978
179	47854	17.0665	95.7922	69.8577
180	47902	23.1192	94.49	68.4883
181	48275	8.21782	89.1538	68.355
182	48328	10.3978	93.2381	73.7507
183	48342	12.5865	94.9603	75.1012
184	48379	10.2766	97.3114	75.1652
185	48398	13.2073	98.7562	73.1421
186	48426	10.3022	98.1995	71.8297
187	48479	9.54532	100.94	70.8073
188	49447	6.80723	92.3623	71.7377
189	49477	6.47318	95.6562	69.599
190	49490	6.53173	93.1849	67.7629
191	50379	8.48999	94.4875	75.961
192	50413	9.67346	96.4316	78.0274
193	51789	14.6155	88.7566	64.1894
194	51933	22.5554	89.5591	64.312
195	51976	20.7023	93.6561	62.5712
196	51982	20.8669	92.2675	59.6872
197	51991	24.056	91.9896	58.2814
198	52937	25.4646	89.6201	63.4884
199	52946	23.3845	89.8324	61.1029
200	53280	8.56636	90.1671	65.3873
201	53404	26.2739	94.7062	67.5974
202	53409	25.4519	93.1284	70.672
203	53418	27.5991	91.6226	68.9346
204	53425	27.3047	91.3423	65.4401
205	53436	30.092	90.6348	66.684
206	53479	24.8038	92.7822	63.0657
207	53486	26.4353	90.0741	60.2316

208	53488	28.9856	91.4444	61.1333
209	53768	4.02202	93.1006	65.746
210	53780	5.22809	94.5092	62.8748
211	53786	8.54911	92.727	62.9972
212	53981	12.162	93.0452	62.5108
213	53990	15.1121	93.2763	61.6849
214	54898	15.766	95.4148	73.7298
215	55225	4.85793	90.9854	74.1641
216	55282	5.23304	91.5737	61.9299
217	55315	9.49886	91.8281	59.9016
218	55324	9.10584	94.7849	60.7624
219	55364	14.395	95.5462	59.7567
220	55368	13.4163	92.4716	58.5636
221	55416	14.0593	96.0322	62.7901
222	55436	17.6941	92.8711	59.3607
223	55455	14.9974	95.8191	56.6137
224	55479	13.0262	98.1597	55.0293
225	55484	15.4991	97.7228	52.1518
226	55901	30.4263	93.1498	68.4496
227	55909	29.3813	92.8719	72.0127
228	55928	29.0693	87.7593	67.7686
229	55936	32.1188	88.4597	67.9656
230	55961	28.8597	88.6311	63.7859
231	56284	13.5151	90.3631	61.86
232	56409	21.7956	92.8813	72.0858
233	56414	22.1254	89.876	72.2056
234	56436	26.3882	88.2905	73.2902
235	56814	14.383	89.6632	67.1647
236	57307	17.5176	95.0115	63.7838
237	57399	26.146	88.6418	66.6097
238	57417	28.2311	89.2433	70.6998
239	57428	29.1795	85.471	65.3813
240	57486	28.141	89.0276	57.6636
241	57491	31.7401	89.1265	57.1754
242	57998	28.0457	94.3463	59.5314
243	58268	6.65527	87.1762	65.82
244	58282	7.1354	87.3422	62.0094
245	58288	10.2792	87.2154	61.7247
246	58658	16.6952	78.4282	72.2514
247	58845	12.1338	91.8763	76.0824
248	58909	11.175	92.7272	79.4656
249	58978	7.36898	95.4803	73.0651
250	59398	17.319	96.574	60.7571
251	59403	15.088	98.6497	59.2278
252	59498	13.8043	99.6462	50.3247
253	59681	15.3074	75.6769	73.0649
254	59705	15.0263	77.2339	69.612
255	59709	10.3717	76.0132	69.1277
256	59719	11.2703	79.2119	70.0091
257	59746	12.6423	77.4116	65.856
258	59756	9.98813	75.282	63.834
259	59780	9.38158	77.8198	62.2754
260	59792	13.017	74.2403	63.6795
261	59860	14.4586	74.6581	66.5257

262	59863	17.9848	75.7387	67.3083
263	59901	20.9957	76.6113	67.212
264	59909	19.9066	75.829	71.1131
265	59923	23.0008	74.4934	68.4037
266	59929	20.7535	72.0209	67.7305
267	59933	23.478	71.2819	69.3446
268	59982	21.2707	74.164	65.1748
269	59988	24.5697	74.6582	65.7611
270	60000	38.6844	93.357	58.0567
271	60051	40.5804	90.8549	59.4269
272	60206	36.4636	94.1357	55.9041
273	60224	35.7699	91.6259	57.8926
274	60335	33.3502	93.5191	59.5877
275	60842	41.1888	93.6283	60.8847
276	61361	42.8916	89.6518	61.5272
277	61898	43.9644	93.3944	62.9925
278	61916	42.7333	92.9752	66.0196
279	61998	44.1118	94.4425	60.1263
280	62185	39.4632	90.513	56.5748
281	62210	33.0591	93.834	56.5127
282	62275	32.1715	94.9969	53.2237
283	62331	31.9768	97.9989	52.5177
284	62342	34.2792	97.4988	54.697
285	63942	45.1992	88.1281	59.5719
286	65184	40.6073	87.745	58.7014
287	65380	41.4359	91.7454	63.3436
288	65411	43.791	89.9484	66.0217
289	65434	46.3555	88.0146	65.0996
290	65496	45.9073	89.4194	62.4605
291	65769	31.1146	95.9421	50.3899
292	65786	33.1677	96.7005	48.1053
293	65807	33.0543	99.5594	49.7603
294	65835	30.7121	101.205	52.7759
295	65862	34.3535	102.983	53.0802
296	65879	32.1505	105.11	54.4152
297	65904	35.169	106.307	52.191
298	65962	33.3459	103.347	50.108
299	65989	33.6168	105.797	47.9588
300	66296	29.2612	98.2719	48.6756
301	66479	31.196	106.126	51.2357
302	66863	43.7688	91.497	59.3224
303	67435	46.751	91.9163	60.9192
304	68226	30.2018	92.3172	56.7756
305	68289	33.6758	94.0451	46.724
306	69682	38.1888	85.831	58.3762
307	69695	36.6053	88.3853	58.8253
308	69726	33.9817	86.5053	58.563
309	69747	36.4678	89.267	55.9564
310	69756	34.3408	87.9183	53.8812
311	69780	34.8667	90.6135	50.9737
312	69868	37.7656	87.2181	53.8966
313	70173	43.2625	86.0259	57.7865
314	70180	41.6479	84.2867	59.6296
315	70185	41.1105	81.6516	58.0926

316	70192	38.539	82.9687	59.9469
317	70403	47.4857	91.7645	63.903
318	70448	44.6488	86.6435	62.995
319	70795	31.3777	97.633	45.8635
320	70814	34.0369	99.5849	46.3521
321	70863	39.2573	104.826	52.8019
322	70868	37.1413	103.235	50.4358
323	70898	37.7274	108.135	53.0657
324	70909	37.8061	108.634	56.3451
325	70957	41.1063	108.549	52.984
326	70980	37.8221	110.59	50.352
327	70984	40.8773	109.738	49.3581
328	70998	41.1812	111.358	46.4168
329	71350	37.7664	98.379	55.0397
330	71363	41.0268	100.018	56.3091
331	71402	41.8071	102.834	57.3825
332	71409	39.5281	100.932	58.8567
333	71447	38.178	101.423	54.2224
334	71490	41.9839	103.074	52.1858
335	71761	32.4447	85.849	55.9095
336	71780	32.9563	86.7426	51.1933
337	71788	35.7298	85.4724	51.8931
338	72727	33.2821	90.1495	60.4076
339	73279	39.9296	93.0622	54.8181
340	73341	41.8858	96.5049	61.4788
341	73783	34.8629	88.6419	48.3954
342	73805	36.9979	91.4334	48.5528
343	73842	37.7461	92.3367	51.4583
344	73863	40.9231	89.998	51.9206
345	73928	38.3432	83.8801	53.7603
346	73982	39.5339	86.1558	50.199
347	74211	30.6099	95.7225	56.5923
348	74481	29.5939	104.194	49.3172
349	74497	31.176	104.972	46.1078
350	75140	31.6289	91.2614	62.8747
351	75163	31.8913	88.234	62.3119
352	75184	30.2776	86.5345	60.1126
353	75209	24.9201	89.0394	57.7541
354	75263	25.1058	88.8183	54.7432
355	75283	26.023	88.6782	51.6782
356	75306	28.7641	89.7452	53.1146
357	75344	27.118	91.2198	55.5535
358	75482	28.017	93.5992	53.2126
359	75680	27.1967	86.7408	61.4738
360	76291	27.791	86.9577	55.4679
361	77268	22.1058	90.08	56.9053
362	78775	23.7425	91.7425	53.4331
363	78878	27.1233	94.1834	56.4755
364	80256	22.2289	88.7297	50.1237
365	80275	23.4285	90.5491	47.7484
366	80287	26.6141	90.1341	45.7716
367	80299	25.8931	92.7881	47.8424
368	80335	26.3547	94.8723	50.3131
369	80364	29.5411	92.3513	50.1328

370	80368	28.3792	89.5369	49.2271
371	80488	29.854	90.7421	45.0148
372	80498	29.2553	92.4854	42.3315
373	80833	26.2475	91.4852	50.5396
374	81256	21.7263	89.1577	53.3164
375	81684	27.6699	83.8427	62.7429
376	81698	24.4008	86.3436	62.5116
377	81727	22.2985	85.8972	59.6308
378	81737	23.4906	86.0971	56.6518
379	81756	22.3084	86.1951	53.8775
380	81787	24.4573	87.2692	48.6541
381	82184	27.9787	84.4605	58.4956
382	82536	28.7468	93.8865	63.0177
383	82860	30.8261	92.1178	53.6098
384	82991	28.1043	95.5177	47.2876
385	83830	27.5138	96.6518	53.3359
386	85656	30.5137	82.9544	59.3211
387	85681	27.7306	81.0812	59.3852
388	85684	28.3489	78.6527	57.3693
389	85692	24.4962	81.004	58.2003
390	85703	22.5304	83.2718	55.9334
391	85710	19.2534	86.5151	54.5126
392	85728	19.5134	84.8532	57.4599
393	85756	22.5055	85.1964	49.706
394	85776	21.2651	86.4879	46.9104
395	85781	23.2734	87.9498	45.036
396	85787	26.1244	86.9024	45.2619
397	86211	18.7137	90.3395	55.4939
398	86248	21.6722	91.7633	50.1848
399	86261	18.7958	89.0472	49.5443
400	86275	19.4712	91.1445	46.6483
401	86789	27.9465	86.0356	47.6391
402	87198	25.0303	84.5905	59.6473
403	87484	28.0525	89.3096	43.0765
404	89216	20.5036	92.6082	53.955
405	89278	23.5655	92.4472	44.774
406	89444	32.0623	93.0243	44.3804
407	89472	30.3518	95.8711	43.0343
408	89486	30.9222	94.878	40.2397
409	89491	33.5377	96.9462	40.2731
410	95673	26.0745	84.3853	55.9713
411	95710	16.4406	87.8914	53.3631
412	95728	17.0207	87.4968	56.4494
413	96755	19.8108	86.2553	50.7753
414	96867	28.255	86.5206	52.0834
415	96934	28.6661	82.782	54.0303
416	96946	26.1825	83.055	52.1309
417	99312	23.0765	93.9322	47.9197
418	100863	33.8043	91.3473	54.4324
419	102715	15.546	90.6606	55.0112
420	102999	26.6686	93.332	44.0585
421	103270	17.4372	87.3008	47.2752
422	103279	18.4729	88.8681	44.6482
423	103990	31.9126	89.9914	50.3358

424	104498	29.9696	101.218	47.708
425	105000	44.2066	102.155	65.9959
426	105139	46.9223	101	67.1848
427	105280	44.9848	104.24	63.2626
428	105305	46.4989	104.231	66.1887
429	105338	48.4888	104.2	68.8257
430	105360	50.259	101.593	70.9061
431	105401	53.5728	101.516	69.9864
432	105484	51.8713	102.77	67.6231
433	105715	43.7239	105.624	66.0976
434	106409	46.0492	102.293	70.0835
435	107410	52.7777	101.977	72.9777
436	107844	45.8206	106.034	68.8815
437	107916	48.8054	103.458	72.9513
438	108710	41.6658	104.541	63.6906
439	109182	42.9229	99.6262	64.6816
440	110403	53.56	104.567	69.5585
441	110498	51.4674	106.214	66.2446
442	110786	47.7289	103.485	62.0441
443	110882	51.3412	105.055	71.9384
444	110907	55.3443	104.288	72.542
445	111435	51.591	99.148	69.6598
446	112595	46.1796	100.609	64.0389
447	113291	49.2123	102.651	65.0806
448	113340	51.7085	107.21	69.7235
449	113363	56.0926	106.964	71.016
450	113413	56.4933	107.746	74.4604
451	113436	58.4697	105.03	72.1628
452	113486	57.6742	105.415	68.7294
453	113497	59.9748	106.569	66.8707
454	115156	44.6561	98.4349	67.1681
455	115215	40.2773	107.724	62.7579
456	115247	40.8529	109.301	59.7641
457	115278	40.1412	111.03	55.7638
458	115286	44.2274	110.141	55.9517
459	115307	45.2747	113.085	56.3029
460	115321	46.9275	110.435	57.4536
461	115335	46.296	114.24	58.902
462	115341	48.9619	114.711	60.617
463	115359	52.1751	114.422	60.0674
464	115363	55.9925	113.145	61.1176
465	115368	54.4061	112.42	58.598
466	115409	56.4225	113.227	64.1421
467	115424	58.6723	112.289	62.1732
468	115489	58.8803	113.656	59.4195
469	115498	58.2396	114.77	56.6617
470	115863	56.5037	102.417	69.5209
471	115936	56.2888	100.337	72.3219
472	115987	54.6659	104.268	65.9115
473	116256	40.458	106.984	56.697
474	116286	43.6556	108.904	51.4399
475	116306	44.8257	111.86	53.4335
476	116331	42.1358	113.592	55.4481
477	116404	49.1075	116.275	57.7059

478	116431	48.8825	112.614	56.3765
479	116480	44.1612	115.262	53.4978
480	116488	47.5714	115.898	53.7411
481	116498	46.105	116.354	50.7057
482	116907	56.4768	102.224	75.5882
483	116917	53.3798	101.624	77.1346
484	116933	55.6788	98.0079	74.7437
485	117291	45.3159	107.445	54.0051
486	117402	55.4575	116.421	61.1546
487	117409	53.7218	116.586	64.2046
488	117475	53.6401	116.47	58.2615
489	117486	55.2147	115.765	55.3992
490	117692	40.9907	102.819	66.2049
491	117781	40.5898	112.339	52.5721
492	117835	44.0065	115.856	56.4589
493	117854	47.5962	118.041	59.7813
494	117974	51.5467	117.787	56.4117
495	117984	52.5634	117.759	53.5008
496	118498	55.8243	117.147	52.3829
497	118913	50.6577	114.306	63.1101
498	118951	52.1204	113.555	56.9154
499	119238	41.9957	106.32	60.373
500	119315	47.7193	110.454	54.411
501	119418	50.7962	111.673	59.2568
502	119424	51.6589	110.281	56.6445
503	119480	50.1562	112.779	53.4196
504	119486	52.7088	111.05	51.8936
505	119912	53.4162	110.944	62.1228
506	119933	55.3974	108.84	58.2424
507	119959	54.362	111.674	55.6412
508	120000	18.953	79.5523	82.011
509	120002	22.0036	78.7428	81.2804
510	120069	20.7699	81.1297	79.2364
511	120211	16.4911	81.7329	80.2903
512	120262	19.3688	79.6948	76.9109
513	120279	20.4945	79.3009	73.6495
514	120291	22.3662	77.039	74.3727
515	120376	18.1417	76.7232	75.4285
516	120413	20.1691	77.3978	78.8793
517	120745	24.1418	80.8811	78.6827
518	120764	24.1967	82.5052	74.7373
519	120787	26.9023	83.7686	73.6383
520	120835	24.9283	86.3467	75.947
521	120863	30.1305	83.5625	75.3522
522	120909	26.9759	83.226	76.7673
523	121247	18.6207	83.9159	77.0771
524	121296	16.3345	85.7534	79.4698
525	121307	18.2223	87.7678	80.7654
526	121363	18.6982	84.5049	81.91
527	121417	13.9186	85.9697	81.5082
528	121478	15.4563	88.902	80.3228
529	121658	24.7281	76.6211	80.3658
530	122141	26.3983	79.405	80.9773
531	122158	27.5543	74.3727	80.4158

532	122166	25.2587	73.7715	82.4485
533	122182	23.7861	70.7096	81.3516
534	122208	22.0043	74.8598	79.6791
535	122264	24.6848	78.0655	77.7256
536	122288	28.9554	80.6778	75.2217
537	122292	28.6257	78.7502	78.3209
538	122338	28.9001	81.2923	81.5429
539	122362	31.9645	81.8943	82.187
540	122368	31.4314	81.0189	79.1994
541	122415	30.2304	82.7532	84.5578
542	122728	21.3261	77.1544	83.7643
543	122979	23.3961	85.3197	73.353
544	123673	25.2833	72.7584	78.9825
545	123762	27.3629	73.5829	76.7484
546	123780	28.9096	74.0266	74.0836
547	123786	32.271	73.2042	73.8527
548	123802	31.5367	74.1317	76.6624
549	123807	34.3934	76.9684	76.02
550	123821	34.7419	73.5767	75.8454
551	123830	35.1875	76.3217	79.1187
552	123841	36.7071	77.5257	82.0445
553	123860	40.0247	75.6943	81.6668
554	123863	42.9344	77.7425	82.8337
555	123882	38.4553	76.3057	84.7259
556	123909	42.9115	76.3875	87.5331
557	123924	44.2058	74.7622	85.13
558	123986	43.7521	76.3962	80.1336
559	124330	24.1933	81.1404	82.9413
560	124680	19.6266	75.5897	81.6119
561	124696	16.6151	74.8446	80.9911
562	124711	13.579	75.8572	79.2193
563	124727	16.9325	72.2923	82.6776
564	124752	19.101	74.1719	78.2375
565	124782	18.4213	72.7819	75.5622
566	124860	25.1572	76.5968	83.72
567	124937	28.2361	75.8027	83.1322
568	125195	22.5192	73.8388	83.887
569	125214	18.5924	79.3364	85.1297
570	125227	20.1638	75.3789	86.5236
571	125290	29.8896	76.735	80.6465
572	125363	35.3671	83.436	83.2209
573	125409	34.4127	84.1936	86.5397
574	125431	36.9498	80.4721	83.5972
575	125436	39.9858	80.5769	84.7838
576	125482	36.3664	80.8938	80.0141
577	125488	39.2943	80.0475	80.7649
578	125863	31.574	76.7378	84.1932
579	125899	29.6274	75.4902	86.2952
580	125910	31.4705	75.4849	89.0238
581	125925	31.3491	72.6293	85.5569
582	125936	34.4816	72.0094	86.588
583	126247	20.8972	82.2915	75.8689
584	126405	31.6317	83.0449	72.5863
585	126411	32.6013	82.2587	76.5508

586	126490	34.9999	82.7893	70.7637
587	126842	19.466	89.532	82.8651
588	126881	17.3827	91.0188	85.0443
589	126980	15.8769	91.2212	82.2961
590	127989	35.2254	83.8848	76.1612
591	128215	16.1023	78.0534	82.8708
592	128228	18.0093	73.6495	85.3294
593	128290	27.0191	73.262	84.9973
594	128306	27.2804	77.4332	86.154
595	128327	27.2788	77.0475	89.2301
596	128335	25.3414	78.6977	90.902
597	128338	28.509	79.2378	90.9366
598	128345	28.0691	76.9534	93.5888
599	128360	31.0288	74.3423	93.508
600	128363	34.8596	76.1697	92.9075
601	128399	33.7218	75.6123	96.1379
602	128410	34.9001	76.4598	99.0427
603	128433	34.5249	72.2692	97.4436
604	128470	33.0866	72.1102	94.0482
605	130184	23.9439	67.4599	80.3345
606	130196	20.5191	70.1337	82.3105
607	130225	18.3102	73.7099	88.8235
608	130236	21.5675	72.7104	87.3349
609	130341	23.4984	79.7731	86.0651
610	130370	31.2862	78.1645	86.9037
611	130410	31.2183	78.2889	92.1407
612	130427	28.3307	73.3692	89.2716
613	130436	31.7829	72.4825	90.599
614	130470	29.0479	70.5954	87.9195
615	130789	30.8362	76.9803	75.9241
616	130892	32.827	84.1408	79.9812
617	130902	35.9437	83.8916	79.9616
618	130999	34.5735	85.0187	73.0541
619	131409	30.2056	75.3662	97.1366
620	131430	29.8995	71.53	94.1349
621	131491	33.2826	74.8288	86.5458
622	131842	28.7526	88.2128	77.2808
623	131863	32.8883	89.4595	77.9321
624	131878	29.828	91.062	78.0509
625	131903	33.1767	93.5581	78.075
626	131909	31.6412	90.2136	81.2517
627	131982	31.6441	92.1283	75.5176
628	131989	33.9699	94.0072	74.7202
629	132403	37.0119	85.9903	85.4499
630	132413	35.8241	82.8704	88.8553
631	132478	33.2338	85.566	83.3277
632	132711	15.5395	78.4889	75.8464
633	132780	19.6392	79.171	70.6692
634	132786	23.2614	77.2184	70.4608
635	132825	24.7563	74.6535	71.2111
636	132842	26.8295	77.4298	71.8677
637	132858	28.6612	73.9001	70.5185
638	132864	32.6014	72.8171	69.7769
639	132876	29.6448	71.5821	72.3498

640	132901	35.0738	71.6998	73.0027
641	132912	33.8484	70.8564	78.6322
642	132928	33.846	69.2363	74.2766
643	132935	36.2367	67.8181	73.0238
644	132963	32.5181	69.6903	71.4375
645	132986	33.3601	71.0312	66.2788
646	132991	36.2441	71.9441	66.5981
647	133490	32.3849	75.0382	81.3756
648	133860	29.6136	85.0134	78.0958
649	133935	31.7834	86.4579	75.9626
650	133997	31.7782	88.4073	72.9392
651	134203	21.1163	69.4115	79.3805
652	134209	17.7349	69.9137	77.0093
653	134233	21.1766	70.9923	74.8239
654	134236	20.119	67.5346	75.4493
655	134246	20.4907	71.618	70.9707
656	134256	18.6703	69.1948	69.2343
657	134276	19.3189	70.4019	65.1386
658	134280	19.5323	72.7602	62.9868
659	134286	23.258	71.6708	62.8961
660	134290	25.2224	69.6034	63.8868
661	134316	26.5619	73.0335	64.0945
662	134329	26.2331	74.3662	68.2771
663	134368	29.7235	72.2161	67.4427
664	134403	33.6404	76.2365	70.2513
665	134481	31.6346	74.6404	66.2478
666	134497	34.7738	74.1346	64.0088
667	134711	15.1197	72.1864	76.9903
668	135000	29.0459	89.6405	88.3355
669	135017	29.7089	89.4147	85.4052
670	135139	32.7202	88.5042	84.8918
671	135205	28.043	90.6592	83.1813
672	135211	24.9868	89.9837	83.7963
673	135228	25.4922	88.3592	86.6472
674	135256	26.4729	87.4144	82.9137
675	135330	31.7874	91.4274	84.2028
676	135407	34.1534	88.9165	87.6539
677	136354	25.9137	91.697	86.5181
678	137848	31.1939	91.7012	88.9097
679	137982	30.8762	93.9244	86.8683
680	138307	35.0116	90.6104	81.8309
681	138338	37.7309	92.0036	82.0862
682	138350	39.5242	91.7616	84.7472
683	138363	43.184	91.5402	85.3436
684	138407	43.5204	92.8336	88.3856
685	138433	44.2816	89.5986	87.8424
686	138835	34.7847	91.8395	84.6473
687	138989	45.5002	93.1503	84.029
688	139335	23.6202	90.1418	88.4485
689	139863	37.5877	89.3288	87.1202
690	139899	36.5508	88.6826	90.3456
691	141355	40.2376	93.3599	87.6434
692	141377	41.1722	90.3503	87.6322
693	141410	44.0243	93.2672	91.3458

694	141498	42.7751	93.6531	82.4144
695	141757	28.7347	88.4752	81.1797
696	141790	33.7175	88.0714	80.6796
697	141862	35.7372	88.9865	84.4535
698	141925	39.7277	88.9495	83.0981
699	142663	30.5177	86.0296	85.0819
700	142928	23.219	86.4031	88.1165
701	143182	24.0513	85.4065	82.8477
702	143209	22.4031	88.9085	81.5483
703	143486	42.0706	90.5368	81.0036
704	143898	46.8342	92.0132	87.1536
705	143913	47.1151	91.2019	90.4712
706	143930	47.3199	89.0836	87.9926
707	143983	47.3622	90.4546	84.2808
708	145215	25.2176	92.9834	81.6979
709	145256	25.7349	90.0135	80.5335
710	145718	26.9247	95.6636	82.1338
711	145732	27.4853	94.4994	85.8176
712	145796	25.755	97.642	86.5393
713	145803	27.8665	98.5886	88.6297
714	145838	30.2117	101.078	88.9476
715	145842	31.9797	101.073	91.4798
716	145860	33.921	97.318	92.6079
717	145863	37.0564	99.8205	92.4365
718	145868	34.9227	99.1155	90.3953
719	145879	34.1388	99.6864	95.04
720	145912	36.8513	99.2215	97.1264
721	145933	37.2424	95.258	95.125
722	146981	37.8177	90.7886	79.0639
723	147211	22.577	92.0831	85.8736
724	147220	25.0907	94.5315	88.0473
725	147260	28.3461	97.4529	85.0337
726	147278	29.0675	99.9565	82.6742
727	147288	32.2775	99.0437	81.693
728	147303	32.9286	100.165	84.7108
729	147328	35.0559	100.55	87.11
730	147341	34.8769	103.375	88.5201
731	147358	38.4842	103.808	88.4414
732	147363	42.315	102.255	89.2497
733	147368	40.8679	102.31	86.4285
734	147403	43.6443	105.568	88.9804
735	147409	42.581	103.882	92.0858
736	147436	45.7118	102.463	90.1809
737	147480	40.4706	105.29	86.2957
738	147487	42.8443	104.313	84.6272
739	147768	24.0576	95.2447	84.9896
740	147845	29.6293	98.7402	91.9923
741	148737	24.1935	89.0342	78.1453
742	149403	37.4235	91.9942	89.381
743	149412	37.3978	89.5144	93.4226
744	149722	21.1081	84.8307	84.5977
745	149791	26.9624	84.7207	84.4126
746	149841	27.9464	86.7248	87.2188
747	150063	0.773091	91.053	82.0184

748	150093	2.78027	88.8576	81.3657
749	150290	5.62595	87.6153	82.5326
750	150694	-1.59979	89.6136	83.316
751	150706	-1.58749	92.386	80.4727
752	150722	-3.79459	94.4117	80.688
753	150746	-3.06832	93.9142	77.109
754	151210	-4.94726	91.2632	81.5575
755	151255	-5.007	91.3878	78.4521
756	151279	-5.51102	93.1721	74.7199
757	151711	-7.35783	93.4426	80.0594
758	151788	-3.56995	92.8891	72.2841
759	151853	-4.84338	96.8489	76.031
760	151898	-2.82231	96.5366	78.6331
761	151915	-3.53869	97.7547	81.6842
762	153756	-2.86001	89.5364	79.9348
763	153873	1.45289	88.9699	84.4974
764	153965	2.90268	86.5006	85.5496
765	155780	-5.32016	96.5822	72.4123
766	155858	-0.278967	98.1172	79.0824
767	155907	0.632117	99.978	81.2631
768	155912	0.124685	99.5026	84.2533
769	155979	-4.19098	100.385	80.3043
770	155998	-2.18842	100.639	78.08
771	157182	-0.324432	87.8975	80.9699
772	157261	-7.98004	92.2618	76.571
773	157278	-7.37256	93.7743	72.3892
774	157382	-5.95177	95.9453	79.1129
775	157480	-7.26707	99.9589	79.0237
776	157980	-8.6012	97.1406	73.1437
777	160455	-2.64491	100.738	83.1272
778	160658	3.53484	85.3781	82.1067
779	160904	1.71756	100.823	78.4633
780	160998	-0.214703	101.841	76.1668
781	161778	-6.92091	96.2931	69.8813
782	161784	-5.34655	93.6329	69.0403
783	161814	-2.60049	95.9501	70.4629
784	161841	-2.7855	98.0858	73.5275
785	161861	0.000899255	96.329	72.6689
786	161903	2.01595	98.0474	71.1514
787	161978	-1.8836	98.9224	70.3448
788	161991	0.512871	100.356	69.2235
789	161998	-2.46085	100.421	67.7863
790	163508	1.26684	93.6631	85.2003
791	163717	-9.09902	95.9062	79.3308
792	163727	-10.1451	93.1266	82.2272
793	163766	-9.73341	95.3182	75.8675
794	164397	4.59494	100.043	72.5609
795	164409	4.42454	99.5809	75.6508
796	164484	3.50253	101.479	67.3089
797	164991	2.71349	102.817	74.909

APPENDIX C

NAMD SAMPLE INPUT FILES

C.1 Sample NAMD configuration file for TAMD simulation

The following script is for a TAMD simulation using the NAMD software.

```
1 #NAMD CONFIG FILE FOR TAMD runs
2 # input
3 # set path /export/users/harishv/abc/tamd_dimer_3kt/
4 set input ionized_lfeh_o2
5 coordinates $input.pdb
6 structure $input.psf
7 bincoordinates npt.restart.coor
8 extendedsystem npt.restart.xsc
9 parameters par_all122_prot_cmap.inp
10 parameters toppar_all122_prot_heme.str
11 #parameters ct800342w_si_003.txt
12 parameters par2.inp
13 paratypecharmm on
14 # output
15 set output tamd77
16 outputname $output
17 dcdfile ${output}.dcd
18 xstFile ${output}.xst
19 dcdfreq 2000
20 xstFreq 10000
21 binaryoutput yes
22 binaryrestart yes
23 outputEnergies 10000
24 restartfreq 10000
25 fixedAtoms off
26 # Basic dynamics
27 exclude scaled1-4
28 1-4scaling 1
29 COMmotion no
30 #dielectric 1.0
31 # Simulation space partitioning
32 switching on
33 switchdist 8.0
34 cutoff 10
35 pairlistdist 12
36 # Multiple timestepping
```

```

37 set dt 1
38 firsttimestep      0
39 timestep           $dt
40 stepspercycle      10
41 nonbondedFreq      1
42 fullElectFrequency 2
43 rigidbonds         none
44 # Periodic Boundary Conditions
45 if {0} {
46 cellBasisVector1    94.549    0.    0.
47 cellBasisVector2    0.    80.798    0.
48 cellBasisVector3    0.    0.    87.958
49 cellOrigin          24.078893661499023 91.66455078125 70.47941589355469
50 }
51 wrapAll            on
52 #PME (for full-system periodic electrostatics)
53 PME                yes
54 PMEGridSizeX       95
55 PMEGridSizeY       82
56 PMEGridSizeZ       90
57 # Temperature control
58 set temperature     310
59 temperature         $temperature
60 # Langevin Dynamics
61 langevin           on
62 langevinDamping     5
63 langevinTemp        $temperature
64 langevinHydrogen    off
65 seed               1125784
66 # external forces
67 if 0 {
68 constraints         on
69 consKCol            B
70 ConsExp             2
71 consRef             fixed.cnst
72 consKFile           fixed.cnst
73 }
74 #set hvcv           /export/users/harishv/hvcv/
75 tclforces           on
76 tclforcesscript     cfacv_tclforces.tcl
77 set labelPDB        label.pdb
78 set cvINP           cv.inp
79 set restrINP        restr.inp
80 set TAMDoF          100
81 # Scripting
82 run 1000000

```

C.2 Sample NAMD configuration files

In this section, I provide general input scripts for running MD simulations in NAMD using the NPT ensemble for RGS proteins and human PDE 4.

C.2.1 RGS4 equilibration

```
1 # equilibration in NPT ensemble run
2 #####
3 ## ADJUSTABLE PARAMETERS ##
4 #####
5 set input      ionized_conjugated_lagr95C
6 structure      $input.psf
7 coordinates    $input.pdb
8 set outputname npt
9 set temperature 310
10 #continuing a run
11 #set inputname      npt.restart      ;# only need to edit this in one place!
12 #binCoordinates    $inputname.coor   ;# coordinates from last run (binary)
13 #extendedSystem    $inputname.xsc    ;# cell dimensions from last run
14 firsttimestep     0
15 #####
16 ## SIMULATION PARAMETERS ##
17 #####
18 # Input
19 paraTypeCharmm     on
20 #parameters        1a.prm
21 parameters         par_all136m_prot.prm
22 parameters         toppar_water_ions_namd.str
23 temperature        $temperature
24 COMmotion          no
25 # Force-Field Parameters
26 exclude            scaled1-4
27 1-4scaling         1.0
28 cutoff             10.
29 switching          on
30 switchdist        8.
31 pairlistdist      12
32 # Integrator Parameters
33 timestep           2.0
34 rigidBonds         all
35 nonbondedFreq     1
36 fullElectFrequency 2
37 stepspercycle     20
38 # Periodic Boundary Conditions
39 if {1} {
40 cellBasisVector1   89.55800247192383    0.    0.
41 cellBasisVector2   0.    66.03299713134766    0.
42 cellBasisVector3   0.    0.    57.72800064086914
43 cellOrigin         -0.7978525757789612 0.15729907155036926 0.8326889276504517
```

```
44 }
45 wrapAll          on
46 #PME (for full-system periodic electrostatics)
47 PME              yes
48 PMEGridSizeX    91
49 PMEGridSizeY    68
50 PMEGridSizeZ    59
51 # Constant Temperature Control
52 langevin         on      ;# do langevin dynamics
53 langevinDamping  5       ;# damping coefficient (gamma) of 5/ps
54 langevinTemp     $temperature
55 langevinHydrogen off     ;# don't couple langevin bath to hydrogens
56 # Constant Pressure Control (variable volume)
57 useGroupPressure no
58 useFlexibleCell  no
59 useConstantArea  no
60 langevinPiston   on
61 langevinPistonTarget 1.01325 ;# in bar -> 1 atm
62 langevinPistonPeriod 100.
63 langevinPistonDecay 50.
64 langevinPistonTemp $temperature
65 # Output
66 outputName       $outputname
67 restartfreq      10000
68 dcdfreq          10000
69 outputEnergies   10000
70 outputPressure   10000
71 xstFreq          10000
72 minimize         1000
73 reinitvels       $temperature
74 run 500000000
```

C.2.2 Human PDE 4 equilibration

```
1 # equilibration in NPT ensemble run
2 #####
3 ## ADJUSTABLE PARAMETERS ##
4 #####
5 set input          ionized
6 structure          ${input}_pde4ibm.psf
7 coordinates       ${input}_pde4ibm.pdb
8 set outputname    npt
9 set temperature    300
10 #continuing a run
11 #set inputname    npt.restart      ;# only need to edit this in one place!
12 #binCoordinates  $inputname.coor  ;# coordinates from last run (binary)
13 #extendedSystem  $inputname.xsc   ;# cell dimensions from last run
14 firsttimestep    0
15 #####
16 ## SIMULATION PARAMETERS ##
17 #####
18 # Input
19 paraTypeCharmm    on
20 parameters        par_all36m_prot.prm
21 parameters        toppar_water_ions_namd.str
22 parameters        ibm.prm
23 temperature       $temperature
24 COMmotion        no
25 # Force-Field Parameters
26 exclude          scaled1-4
27 1-4scaling       1.0
28 cutoff           10.
29 switching         on
30 switchdist       8.
31 pairlistdist     12
32 # Integrator Parameters
33 timestep         2.0
34 rigidBonds       all
35 nonbondedFreq    1
36 fullElectFrequency 2
37 stepspercycle    20
38 # Periodic Boundary Conditions
39 if {1} {
40 cellBasisVector1 80.03099822998047 0 0
41 cellBasisVector2 0 88.6500015258789 0
42 cellBasisVector3 0 0 80.71900177001953
43 cellOrigin       20.97156524658203 6.474047660827637 24.07125473022461
44 }
45 wrapAll          on
46 #PME (for full-system periodic electrostatics)
47 PME              yes
48 #PMEGridSpacing  1
49 PMEGridSizeX     83
50 PMEGridSizeY     91
51 PMEGridSizeZ     83
```

```
52 # Constant Temperature Control
53 langevin          on      ;# do langevin dynamics
54 langevinDamping   5       ;# damping coefficient (gamma) of 5/ps
55 langevinTemp      $temperature
56 langevinHydrogen  off     ;# don't couple langevin bath to hydrogens
57 # Constant Pressure Control (variable volume)
58 if {1} {
59 useGroupPressure  no
60 useFlexibleCell   no
61 useConstantArea  no
62 langevinPiston    on
63 langevinPistonTarget 1.01325 ;# in bar -> 1 atm
64 langevinPistonPeriod 100.
65 langevinPistonDecay 50.
66 langevinPistonTemp $temperature
67 }
68 # Output
69 outputName        $outputname
70 restartfreq       10000
71 dcdfreq           10000
72 outputEnergies    10000
73 outputPressure    10000
74 xstFreq           10000
75 reinitvels        $temperature
76 run 30000000
```

APPENDIX D

LINUX AND BASH SCRIPTS

D.1 Sample code catdcd

Catdcd functions much like the Unix "cat" command: it concatenates DCD files into a single DCD file. You can also use catdcd to write only selected atoms to the final DCD file. Starting with version 4.0, CatDCD is now built as part of the VMD Plugin tree, and shares the same reader/writer with VMD. CatDCD 4.0 can read/write any of the structure/trajectory formats that are supported by VMD by virtue of the plugin interface¹.

```
1 CatDCD 4.0
2   catdcd -o outputfile [-otype <filetype>] [-i indexfile]
3     [-stype <filetype>] [-s structurefile]
4     [-first firstframe] [-last lastframe] [-stride stride]
5     [-<filetype>] inputfile1 [-<filetype>] inputfile2 ...
6 * EXAMPLES *
7 catdcd -num eq01.dcd eq02.dcd
8 Prints the number of frames in the two DCD files, then the total,
9 then exits.
```

D.2 Parallel calculations for per residue non-bonded interaction energy

VMD has an ability for handling analysis with high computational cost using parallel calculations. The following script distributes analysis task on N number of CPUs on the Cluster. In my thesis, I used this feature for calculating per residue non-bonded interaction energy described in chapter 6.

```
1 #####
2 # (c) Hossein Mohammadiarani #
3 # Modified (Lines 7-37) and compiled by Mohammadjavad Mohammadi #
4 #####
5
6 #!/bin/bash
7 #SBATCH --job-name=zarn
```

¹<https://www.ks.uiuc.edu/Development/MDTools/catdcd/>

```

8 #SBATCH --partition=shared
9 #SBATCH --nodelist=node105
10 #SBATCH -N 1
11 #SBATCH --cpus-per-task=24
12 #SBATCH --time 240:00:00
13 #SBATCH --output test.out
14 #SBATCH --get-user-env
15 #SBATCH --mail-type=ALL
16 #SBATCH --mail-user=mm14@wildcats.unh.edu
17 #SBATCH --exclusive
18 #SBATCH --gres=gpu:0
19 /mnt/lustre/chem-eng/mm14/
20 module load mpi/openmpi-x86_64
21 module load vmd/vmd-1.9.2-text
22 psf=/mnt/lustre/chem-eng/mm14/120/zar_nematode_resenergy/ionized_pde4zarc.psf
23 dcd=/mnt/lustre/chem-eng/mm14/120/zar_nematode_resenergy/npt_zc.dcd
24 NumCpus=24; # the number of cpus that deployed for calculations
25 totalnf=6000; # the total length of dcd file to be analyzed
26 FramesCpu=$((totalnf/NumCpus+1)); # the number of frames for each cpu
27 j=0;
28
29 iter=$((NumCpus-1))
30 for j in $(seq 0 $iter )
31 do
32     start=$(( FramesCpu * $j))
33     end=$((FramesCpu * ($j+1)-1))
34     srun --exclusive -n 1 -c 1 -s -v vmd -e intermedcalc.vmd -
        args $psf $dcd $start $end _out_.$j.dat >& analysis_.$j.log
        &
35     echo "last job number is : $j";
36     j=$(expr $j + 1);
37 done
38
39 wait

```


APPENDIX E

SUPPORTING INFORMATION FOR STUDIES REPORTED IN CHAPTER 3

This appendix provides additional analyses related to Chapter 3.

E.1 Supplemental Results: Testing the effect of F417Y mutation

Kubas et al. [10] have previously shown that mutating F417Y results in a decrease in the rate of oxygen diffusion toward the G-site. To demonstrate the effect of this specific mutation (F417Y) on oxygen diffusion along the MFEP between the Xe-site and the G-site (pathway 1-0 in Fig 3.2), I prepared three systems each for the wildtype (WT) and the mutant protein that differed in the location of oxygen along the MFEP.

Specifically, three locations of oxygen were chosen (see Fig E.1): one corresponding to the Xe-site (system 1), another in the proximity of residues F417/Y417 (system 2), and the third away from the F417/Y419, but toward the G-site (system 3). The WT systems contained 62,306 atoms and the mutant systems contained 62,307 atoms. All systems were energy minimized and briefly equilibrated in the NPT ensemble to equilibrate the volumes of simulation domains. I carried out mean-force calculations using 5-ns long restrained MD simulations (following the same protocols as for all centers reported in section 2.4.1.1) for each of the 3 oxygen locations along the same MFEP in the WT and the mutant proteins, and integrated these mean-forces to obtain the PMFs (Fig E.1c) for oxygen diffusion from the Xe-site toward the G-site. These results show an increase in the free-energy barrier of 2.4 kcal/mol for oxygen diffusion along the MFEP from the Xe-site toward the G-site. Importantly, the location of the barrier is near the side-chain of Y417, and the free-energy values after crossing this barrier are significantly lower toward the G-site (located near point 3 along the MFEP; see Fig E.1a,b) with respect to the Xe-site (point 1 along the MFEP; see Fig E.1a,b). The existence of a free-energy barrier for oxygen diffusion along this MFEP in the mutant protein implies a decreased diffusion rate, as suggested by Kubas et al. [10].

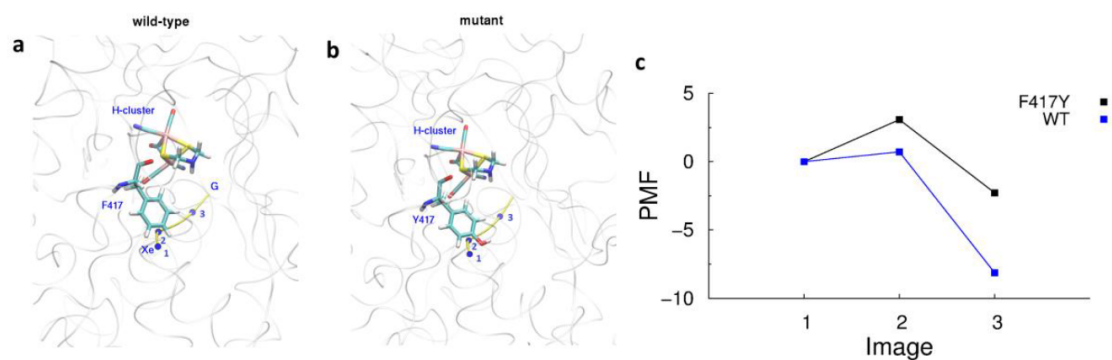


Figure E.1. The effect of F417Y mutation on the PMFs along the pathway 1-0 (see Fig 3.2) in [FeFe]-hydrogenase. (a and b) Snapshots highlighting the location of the wildtype F417 residue and the mutant Y417 residue along the pathway 1-0 (yellow curve). Three points (in the collective variable space) chosen to map the PMF are labeled and depicted as blue spheres in panel a. The approximate locations of the Xe-site and the G-site are marked by Xe and G, respectively. The H-cluster in each panel is labeled and highlighted in a stick representation. (c) The PMF values (kcal/mol) calculated with respect to the first point are shown for the WT protein by blue filled squares, and for the mutated protein by black filled squares. The increase in the free-energy value of the second point closest to the mutated residue is showing a higher energy-barrier of oxygen transition around Y417 toward the H-cluster in the mutated protein.

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22
0		16.68	18.98					13.87		13.51											17.26	18.7	
1	3.6		5.55		0.31	3.91									5.38								
2	7.29	7.72		6.76			6.28															3.07	
3			4.33											4.18									4.09
4		0.76					4.58																
5		9.51									12.6		10.09	10.97									
6			1.57		1.0																		
7	2.01							2.22											9.83				
8							1.9													0.54			
9	4.0									4.57		11.96			8.77	10.95	8.84			8.04			
10																						3.55	
11					2.9							2.22			1.43		1.42						
12									4.71	2.89								2.01					
13			1.14		0.59																		
14																							
15									0.4	0.96						2.57	1.14			3.63			
16									2.1							2.11	0.63			3.17			
17									1.04	1.57	2.17					0.87	0.81			0.79			
18							6.34																
19								3.83	3.78							12.63	12.64	10.06					
20	3.5		1.04								3.79												22.2
21	1.8																				19.05		
22																							

Figure E.2. Energy barriers (kcal/mol) between local minima in the 3D PMF map (Figure 3.2 in the main article). The value on the entry ij is the free energy barrier when going from local minimum i to local minimum j . The empty cells indicate negligible energy barriers or minima pairs that are not connected. The G-site and the Xe-site are shown by local minimum 0, and local minimum 1, respectively (Figure 3.2b).

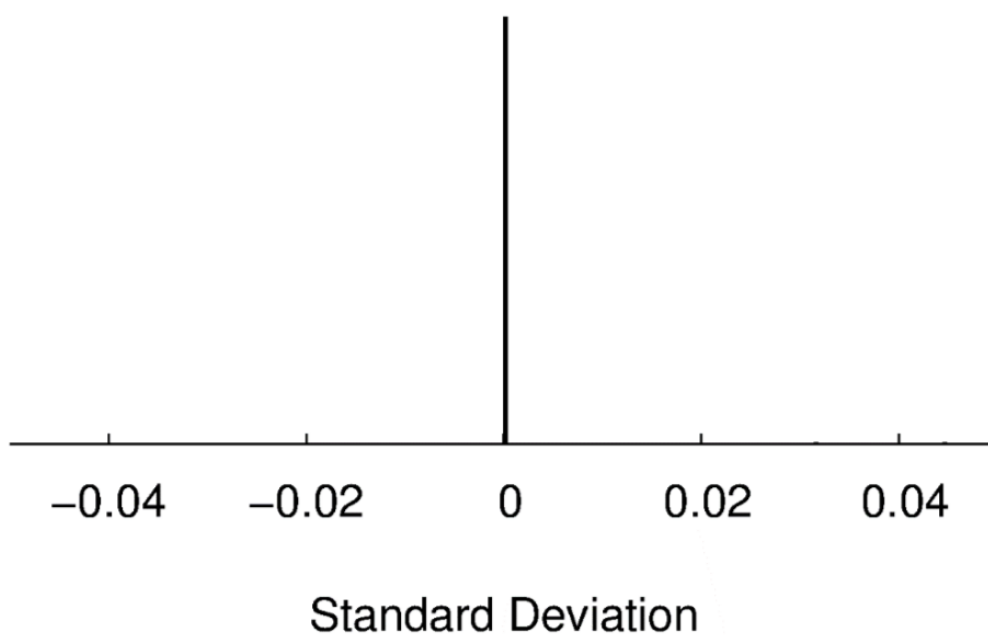


Figure E.3. Histogram of the standard deviation of each mean-force over the last 10% for all centers of O₂ in [FeFe]-hydrogenase. The convergence of mean-force values for all centers are represented as a delta function at the origin.

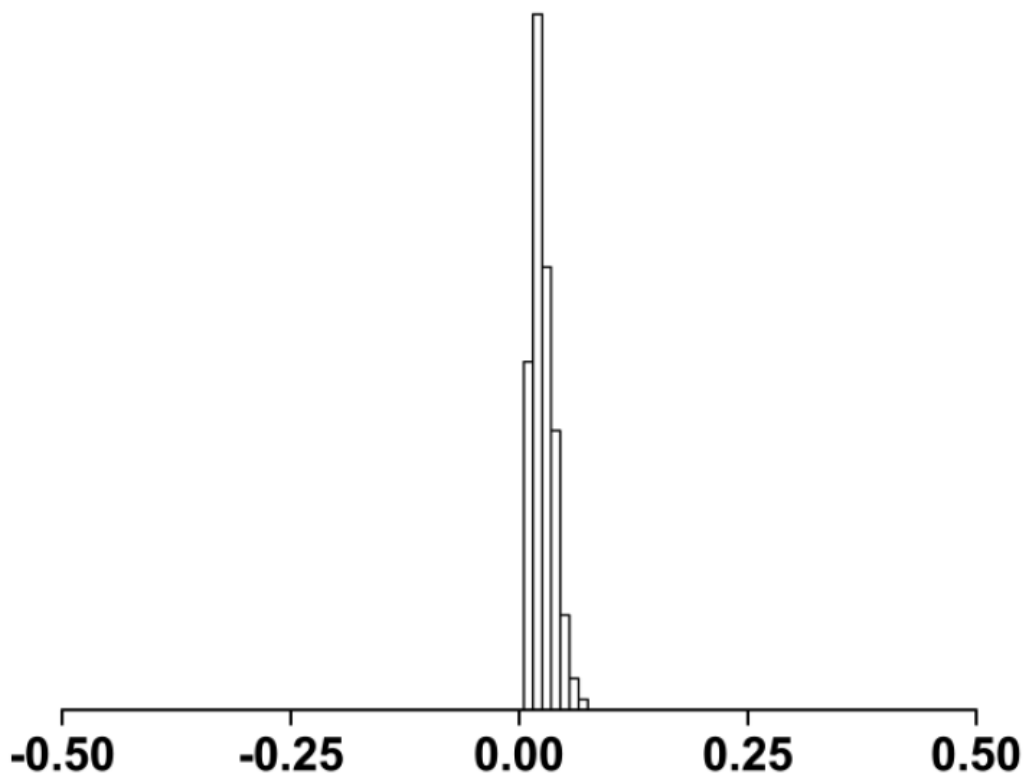


Figure E.4. Histogram of the standard deviation of each mean-force over the last 10% of each mean-force simulation. Data are from mean-force simulations for all 635 unique centers of CO in [FeFe]-hydrogenase (This figure was created by my co-author Yong Liu [3]).

	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22								
6	15.47	14.68	14.69	14.62	16.08	11.4	10.38	8.63	8.46	8.65	8.61	14.58	15.86	14.74	14.61	14.67	14.61	14.62	14.58						
7	12.99	11.35	11.45	11.57	11.85	11.85	11.85	11.85	11.85	11.85	11.85	11.85	11.85	11.85	11.85	11.85	11.85	11.85	11.85						
8	12.42	11.32	9.29	6.3	9.29	5.24	6.3	9.29	8.07	6.05	8.91	8.19	8.25	9.18	8.26	11.34	12.9	13.36	13.36						
9	16.05	10.47	7.77	6	2.87	2.46	5.23	8.7	8.19	7.65	8.15	8.14	8.19	11.43	8.95	7.07	8.94	5.33	8.95	11.21	11.34				
10	11.76	10.16	7.64	5.67	2.13	2.87	3.24	7.62	11.16	3.14	8.14	9.29	11.8	5.71	7.49	9.05	6.39	9.75	11.44	10.77	11.3				
11	13.7	10.86	8.28	4.53	6.04	5.03	8.62	8.68	5.94	8.18	5.57	10.38	13.36	9.53	6.14	9.81	5.12	11.27	12.5	11.3	11.3				
12	10.97	9.68	7.64	5.62	2.73	7.35	2.73	8.41	6.29	8.29	6.25	8.27	8.28	6.05	8.28	12.49	11.57	8.95	8.29	9.19	6.02	8.94	10.8	11.08	
13	6.5	9.78	4.15	9.47	8.26	8.2	4.15	9.16	12.35	-	11.17	11.15	3.36	11.17	10.79	11.34	9.92	13.16	8.63	7.93	5.88	12.78	10.1	10.09	
14	12.53	6.15	6.06	8.16	7.42	8.23	9.05	6.06	8.56	9.03	7.6	7.62	7.93	8.6	7.93	7.93	9.92	13.16	8.63	7.93	5.88	12.78	10.1	10.09	
15	11.86	6.06	7.76	4.01	1.74	6.52	1.74	5.23	4.72	3.54	7.99	8.55	9.16	7.58	8.77	4	5.18	8.84	4.01	6.75	10.77	8.26	8.29	7.83	
16	12.51	4.55	7.63	5.8	2.49	5.83	2.49	7.63	6.32	6.86	9.65	3.87	2.5	5.46	5.55	4.56	-	8.88	13.92	8.88	12.2	7.87	6.88	5.81	
17	11.47	11.08	2.71	4.28	3.39	3.39	6.3	2.71	4.89	6.12	9.55	8.21	3.73	8.22	11.13	13.34	8.94	-	7.07	3.87	5.32	5.25	8.22	7.85	
18	12.49	5.76	5.13	5.31	5.55	5.58	5.55	5.58	5.55	5.58	5.55	5.58	5.55	5.58	5.55	5.58	5.55	5.58	5.55	5.58	5.55	5.58	5.55	5.58	5.55
19	12.38	10.37	7.65	1.03	2.31	1.04	2.31	5.23	2.56	3.56	5.92	9.2	1.34	8.23	8.66	11.16	5.3	6.16	9.19	-	5.29	4.85	7.42	7.84	7.84
20	13.12	2.21	11.93	7.24	3.96	5.73	6.84	11.88	6.3	6.07	13.97	4.13	5.96	8.86	5.64	4.54	5.96	7.06	3.87	4.78	-	7.47	11.86	10.04	10.04
21	12.53	1.17	13.27	11.42	12.25	11.98	7.06	11.02	11.74	6.77	11.83	9.13	11.84	11.94	8.88	4.5	12.3	11.89	11.38	12.65	7.78	-	6.99	2.45	7.85
22	13.09	4.48	5.96	1.8	10.91	13.72	11.11	11.9	12.17	12.25	11.92	8.61	10.27	8.67	9.25	9.76	11.09	8.87	11.53	10.98	10.98	10.98	10.98	10.98	10.98
23	10.95	9.48	7.58	1.88	7.4	6.03	7.58	6.14	6.03	7.58	6.14	6.03	7.58	6.14	6.03	7.58	6.14	6.03	7.58	6.14	6.03	7.58	6.14	6.03	7.58
24	6.5	5.01	6.53	5.81	4.89	4.79	2.73	7.58	6.14	6.03	7.58	6.14	6.03	7.58	6.14	6.03	7.58	6.14	6.03	7.58	6.14	6.03	7.58	6.14	6.03
25	12.59	0.3	6.6	5.69	6.06	8.07	5.34	6.59	8.19	5.34	7.2	8.87	4.75	6.69	5.52	4.5	4.8	6.98	6.76	8.26	7.82	6.65	7.85	6.85	7.85
26	11.66	13.36	2.49	8.61	5.02	5.02	8.58	2.49	8.32	8.14	4.16	2.97	7.99	2.97	7.32	11.5	11.12	5.89	6.1	5.03	11.32	12.51	13.28	13.28	13.28
27	12.43	8.16	8.22	5.56	3.23	3.21	3.31	5.56	6.78	8.26	9.05	3.77	3.23	3.77	3.21	7.97	3.07	6.05	2.83	3.72	4.78	10.63	7.83	7.83	7.83
28	5.73	13.83	14.6	6.87	13.26	13.5	6.25	16.68	6.86	13.12	10.33	14.59	6.85	14.55	9.61	14.33	13.9	7.98	14.43	6.87	14.13	13.68	18.72	18.72	18.72
29	11.67	12.16	2.48	8.62	8.12	6.09	5.25	2.49	8.31	8.23	4.19	2.53	7.96	2.52	7.29	11.57	6.05	6.13	6.04	5.06	7.8	13.37	13.2	13.2	13.2
30	6.54	8.5	4.14	8.58	8.17	8.19	8.21	6.27	8.22	8.2	1.92	3.69	7.95	3.69	5.01	4.89	8.29	6.16	6.12	6.11	7.79	9.96	11.58	11.58	11.58
31	11.62	6.15	5.12	5.31	3.05	3.21	3.01	5.13	7.51	7.8	6.84	13.62	3.86	2.5	3.88	5.62	4.53	0.54	7.07	3.87	4.01	6.05	6.98	7.66	7.66
32	12.57	2.21	7.63	5.82	2.49	5.83	2.49	5.83	2.49	5.83	2.49	5.83	2.49	5.83	2.49	5.83	2.49	5.83	2.49	5.83	2.49	5.83	2.49	5.83	2.49
33	0.61	3.07	14.59	14.43	14.35	14.59	13.97	14.62	16.23	14.26	10.13	11.4	10.55	11.69	9.89	3.74	14.38	14.62	14.56	14.63	9.85	3.3	9.79	9.79	9.79
34	5.48	14.58	6.2	6.42	13.27	5.88	13.82	5.88	7.69	13.82	10.33	8.16	11.51	8	9.02	8.96	11.12	11.33	6.14	6.1	5.01	11.54	10.44	10.39	10.39
35	1.46	9.78	4.12	14.66	14.49	5.19	14.64	3.18	8.24	14.29	1.46	9.08	8	9.02	8.96	11.12	11.33	6.14	6.1	5.01	11.54	10.44	10.39	10.39	10.39
36	11.66	11.81	2.97	4.87	3.42	4.93	6.26	2.92	5.03	8.61	9.5	8.42	3.39	8.36	9.09	13.33	8.94	2.92	9.19	3.39	11.08	12.47	11.26	11.26	
37	6.63	11.16	3.67	4.9	5.72	3.67	3.67	4.9	5.72	3.67	3.67	4.9	5.72	3.67	3.67	4.9	5.72	3.67	3.67	4.9	5.72	3.67	3.67	3.67	
38	11.62	10.68	4.85	3.37	3.36	4.62	6.26	2.71	4.91	6.12	9.55	12.22	8.05	6.19	8.04	10.89	13.36	8.94	3.72	9.19	5.7	8.94	14.36	11.31	11.31
39	6.07	15.42	14.59	6.87	14.63	14.61	14.64	14.62	7.53	14.37	5.28	12.48	6.03	12.51	9.69	10.12	14.58	6.04	14.51	12.86	13.34	13.75	13.68	13.68	
40	8.06	8.56	2.44	8.61	8.22	5.81	7.12	2.49	10.45	8.18	7.83	2.47	8	2.51	7.31	8.46	6.11	10.42	6.12	5.54	7.76	11.56	10.98	10.98	
41	12.59	0.18	2.07	8.09	8.22	8.26	5.23	2.45	7.86	5.24	5.92	8.83	5.82	7.83	8.88	4.54	4.78	7	6.55	8.02	7.81	6.8	10.13	10.13	10.13
42	0.45	15.37	14.6	14.57	14.56	14.62	14.04	14.62	8.28	14.66	10.32	14.52	5.87	14.58	14.55	14.21	14.6	6.06	14.57	8.6	14.53	14.65	14.47	14.47	
43	11.53	13.35	2.49	11.25	4.96	4.89	13.83	2.49	4.96	8.38	3.8	2.98	5	2.97	5.74	30.8	9.51	11.85	9.08	3.5	11.33	14.31	14.31	14.31	

Figure E.5. Major energy barriers (kcal/mol) along MFEPs between minima pairs (MFEPs are shown in Fig. 3.3) . The value on the entry $i j$ is the free energy barrier of the rate-limiting step from the local minima i to the local minima j (This figure was created by my co-author Yong Liu [3]).

	23	24	25	26	27	28	29	30	31	32	33	34	35	36	37	38	39	40	41	42	43	44
6	18.9	14.57	14.69	14.61	14.59	14.62	14.43	14.61	11.22	14.61	14.71	11.35	14.3	12.67	14.63	18.38	14.61	18.91	15.45	16.42	18.8	22.55
1	12.02	11.59	7.08	13.97	8.88	11.95	13.54	15.83	12.22	10.97	8.69	12.33	14.39	12.18	13.17	13.92	11.95	18.83	12.09	14.09	18.7	19.65
2	8.26	12.51	11.35	8.76	9.18	9.2	14.29	8.79	9.12	10.38	11.35	12.53	14.91	10.58	12.61	13.34	11.29	18.85	13.37	13.52	18.79	19.05
3	7.56	7.82	11.23	9.29	9.33	11.98	13.3	9.3	10.69	9.62	9.35	12.4	13.3	12.66	12.37	13.13	11.1	13.3	11.27	12.33	18.34	19.41
4	6.25	7.52	10.61	8.26	9.04	9.35	12.89	9.17	10.81	9.32	8.08	11.8	12.96	12.53	12.06	13.12	10.8	18.83	11.43	13.5	18.4	18.55
5	6.13	7.41	11.16	7.88	8.95	9.21	13.13	9.17	10.94	9.21	8.95	11.63	13.99	9.97	11.94	12.74	10.69	18.79	11.34	13.39	18.43	18.34
6	5.08	7.19	10.59	8.65	9.03	9.09	11.08	8.26	9.07	9.3	8.06	11.48	13.47	12.62	12.33	16.16	11.07	18.85	10.63	13.28	17.8	20.34
7	8.62	8.62	11.34	7.9	9.14	11.3	14.15	7.9	10.13	10.27	11.16	11.72	13.73	9.72	11.7	12.46	10.43	18.87	13.37	12.67	18.6	18.18
8	7.55	7.82	11.29	8.29	8.91	12.49	13.28	8.42	10.03	11.25	8.71	13.68	15.51	9.68	11.66	12.43	10.41	15.9	11.09	16.29	15.45	19.56
9	4.57	6.68	10.08	8.71	6.88	9.4	12.72	8.73	9.08	10.21	9.24	10.97	13.48	12.53	11.63	16.25	10.68	18.87	10.12	13.46	18.78	18.17
10	8.26	12.43	12.92	8.64	7.93	9.11	14.3	8.67	7.33	8.58	11.53	11	14.44	8.79	11.69	12.55	11.47	18.85	12.04	11.39	18.75	16.92
11	9.05	12.48	7.07	7.11	4.86	8.24	14.27	7.13	8.76	7.69	6.44	10.82	14.78	8.77	11.69	12.55	11.47	18.85	12.04	11.39	18.75	16.92
12	6.25	7.5	10.43	8.13	5.37	7.61	13.27	8.58	10.22	8.62	6.35	11.23	16.74	9.98	10.32	12.42	9.12	15.62	10.95	12.91	15.44	18.52
13	8.23	12.84	8.72	7.07	4.76	6.51	13.86	7.09	8.7	6.47	6.44	11.09	14.69	8.72	11.7	13.18	10.51	18.87	9.76	10.46	18.75	16
14	8.23	12.84	8.72	7.07	4.76	6.51	13.86	7.09	8.7	6.47	6.44	11.09	14.69	8.72	11.7	13.18	10.51	18.87	9.76	10.46	18.75	16
15	11.99	12.28	7.03	13.94	8.88	11.93	14.23	13.88	11.43	10.95	8.67	11.57	14.23	12.44	13.93	16.56	11.92	15.54	12.09	14.07	18.73	18.94
16	5.32	6.85	7.34	8.7	3.87	5.71	13.54	6.1	7.83	5.71	4.85	12.51	13.98	10.56	11.9	12.73	10.66	18.86	7.73	10.45	18.79	19.04
17	6.13	8.52	11.35	7.77	9.2	11.21	13.54	7.88	10.01	11.25	8.94	11.66	13.53	9.71	8.27	9.11	7.02	12.51	13.36	13.9	15.42	14.75
18	6.11	8.18	8.24	6.16	2.02	4.9	14.03	6.08	7.82	4.9	4.85	12.43	13.95	7.52	11.88	12.71	10.33	18.86	8.26	10.45	18.75	18.31
19	6.13	7.41	11.39	7.88	8.18	7.63	13.28	8.39	10.48	9.21	6.35	12.38	15.73	9.68	9.34	13.1	7.77	17.09	11.25	12.9	16.15	17.46
20	5.81	12.01	7.08	11.94	3.86	5.62	13.79	9.8	11.46	7.48	3.07	12.28	14.46	13.88	12.39	12.72	10.65	18.82	12.08	14.07	18.75	18.99
21	11.79	6.7	6.81	12.94	8.94	11.76	13.86	13.25	12.14	11.94	8.65	6.75	14.21	8.76	12.96	17.98	11.78	18.79	11.96	13.81	18.8	20.9
22	12	12.3	7.08	13.84	8.88	11.91	13.95	13.97	12.23	10.41	8.5	11.58	14.22	11.72	13.1	16.73	11.8	18.67	12.09	14.03	18.68	19.58
23	4.45	5.01	14.61	10.73	9.19	14.29	8.76	8.92	10.34	9.44	11.32	13.65	12.58	11.69	16.34	10.68	18.86	11.4	13.47	18.77	18.17	21.12
24	5.24	4.81	13.37	6.16	6.1	11.54	14.28	6.07	7.83	7.29	6.09	11.02	13.73	4.31	11.67	13.43	11.26	18.79	5.28	13.12	18.77	19.03
25	9.34	8.07	7.77	6.16	6.1	4.9	13.26	6.13	8.75	7.57	4.85	10.89	13.96	7.48	11.8	12.68	10.74	18.63	12.08	11.38	18.75	18.2
26	6.13	8.81	10.07	8.22	2.83	11.12	14.28	6.07	7.83	7.29	6.09	11.02	13.73	4.31	11.67	13.43	11.26	18.79	5.28	13.12	18.77	19.03
27	14.56	14.4	14.2	14.6	13.56	11.12	14.28	6.07	7.83	7.29	6.09	11.02	13.73	4.31	11.67	13.43	11.26	18.79	5.28	13.12	18.77	19.03
28	6.13	8.81	10.07	8.22	2.83	11.12	14.28	6.07	7.83	7.29	6.09	11.02	13.73	4.31	11.67	13.43	11.26	18.79	5.28	13.12	18.77	19.03
29	6.13	8.81	10.07	8.22	2.83	11.12	14.28	6.07	7.83	7.29	6.09	11.02	13.73	4.31	11.67	13.43	11.26	18.79	5.28	13.12	18.77	19.03
30	8.23	12.56	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78
31	8.23	12.56	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78
32	8.23	12.56	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78
33	8.23	12.56	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78
34	8.23	12.56	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78	14.78
35	14.47	14.39	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58
36	14.47	14.39	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58
37	8.67	14.39	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58	14.58
38	6.12	8.61	11.49	7.87	5.22	5.47	10.31	12.49	7.3	9.69	14.55	11.2	6.04	5.29	11.27	7.61	11.73	18.86	11.72	18.79	18.71	21.12
39	6.12	8.61	11.49	7.87	5.22	5.47	10.31	12.49	7.3	9.69	14.55	11.2	6.04	5.29	11.27	7.61	11.73	18.86	11.72	18.79	18.71	21.12
40	14.66	14.39	14.51	12.51	14.43	11.01	14.28	8.12	7.73	7.32	9.153	12.47	14.42	10.02	12.61	13.34	11.23	18.86	11.72	18.79	18.71	21.12
41	8.21	4.8	0.15	8.76	8.87	0.1	14.28	0.75	7.83	7.3	6.42	11.04	13.63	4.31	4.43	10.36	14.19	11.58	15.43	7.69	6.04	9.75
42	8.19	13.83	12.97	0.74	4.86	0.09	14.29	0.75	7.83	7.3	6.42	11.04	13.63	4.31	4.43	10.36	14.19	11.58	15.43	7.69	6.04	9.75
43	14.65	14.41	14.6	14.59	14.52	11.05	1.18	14.67	12.46	14.6	14.48	11.34	0.09	12.37	16.6	7.64	10.31	0.52	15.45	7.18	4.3	4.3
44	8.61	14.4	13.47	0.87	8.98	10.68	2.57	0.88	10.41	10.54	6.04	11.3	2.36	2.61	11.64	11.4	12.04	2.33	13.04	12.79	2.4	2.4

Figure E.6. Continuation of major energy barriers (kcal/mol) along MFEPs between minima pairs from Fig E.5 (This figure was created by my co-author Yong Liu [3]).

APPENDIX F
SUPPORTING INFORMATION FOR STUDIES REPORTED
IN CHAPTER 4

This appendix provides additional figures related to Chapter 4.

Model 1

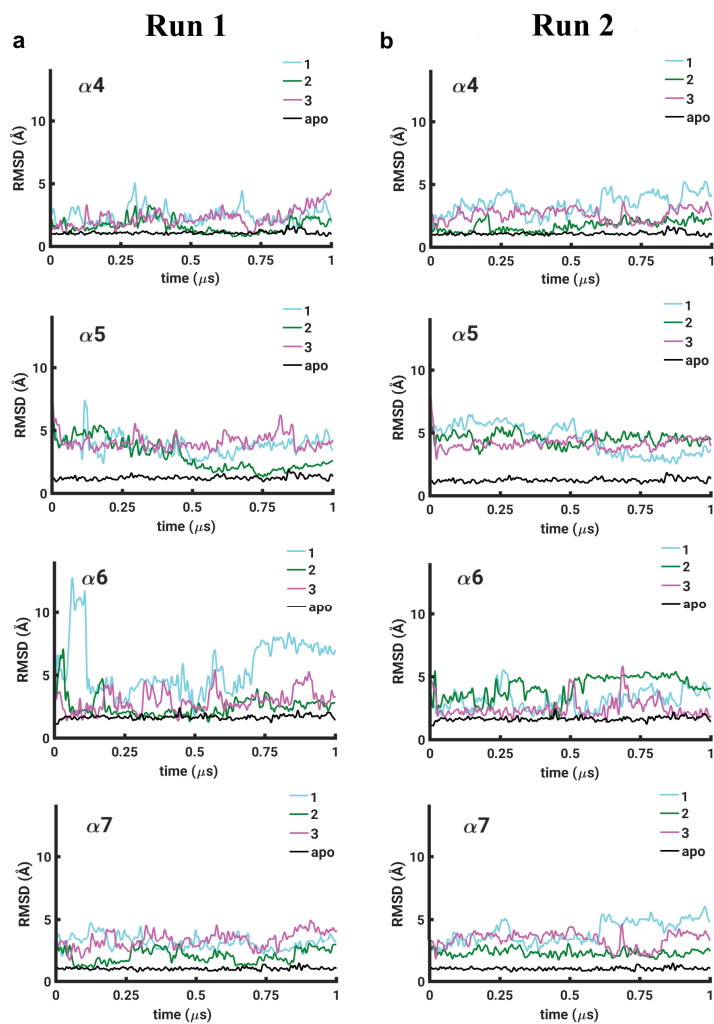


Figure F.1: The traces of root-mean-squared-deviation (RMSD) *vs.* simulation time (μs) for 4 helices in the α_4 - α_7 helical bundle of RGS4 are shown from two independent simulation runs (Run1, panel a; Run2, panel b) for complexes of RGS4 with TDZD compounds **1** (cyan trace), **2** (green trace), and **3** (magenta trace). The black traces show data for an apo-RGS4 simulation from our previous work [13].

Model 2

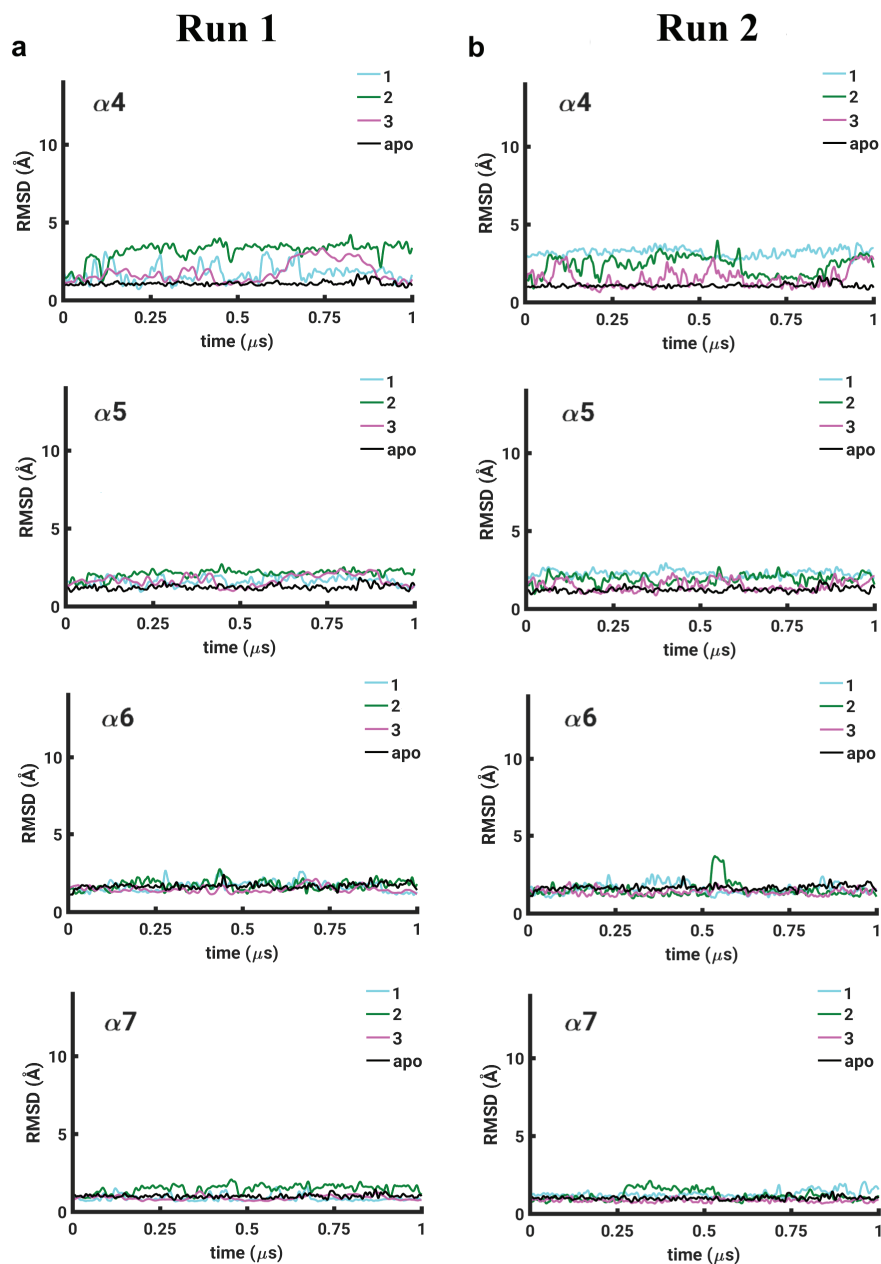


Figure F.2: Same data as in Fig. F.1 are shown for Model 2.

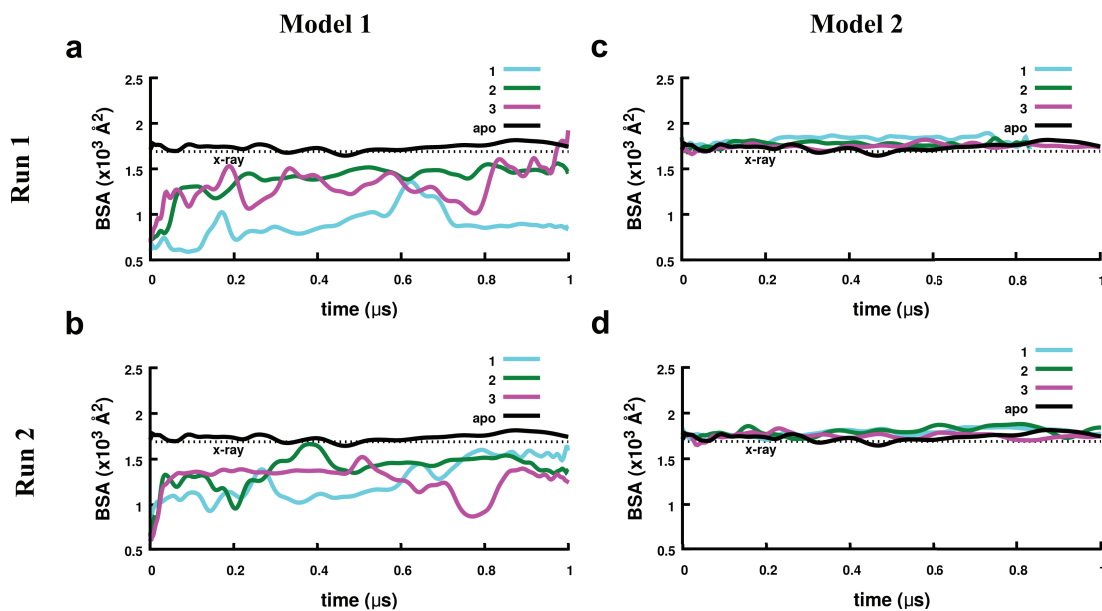


Figure F.3: The traces of buried surface area (BSA) between the α_5 - α_6 helices and the rest of RGS4 *vs.* simulation time (μs) are shown from two independent simulation runs for each Model (Models 1 and 2). The BSA traces are shown for three TDZD compounds (cyan, green, and magenta traces) and from a simulation of apo-RGS4 (black traces). The dotted horizontal line in each panel highlights the BSA-value in the crystal structure of RGS4 (PDB: 1AGR).

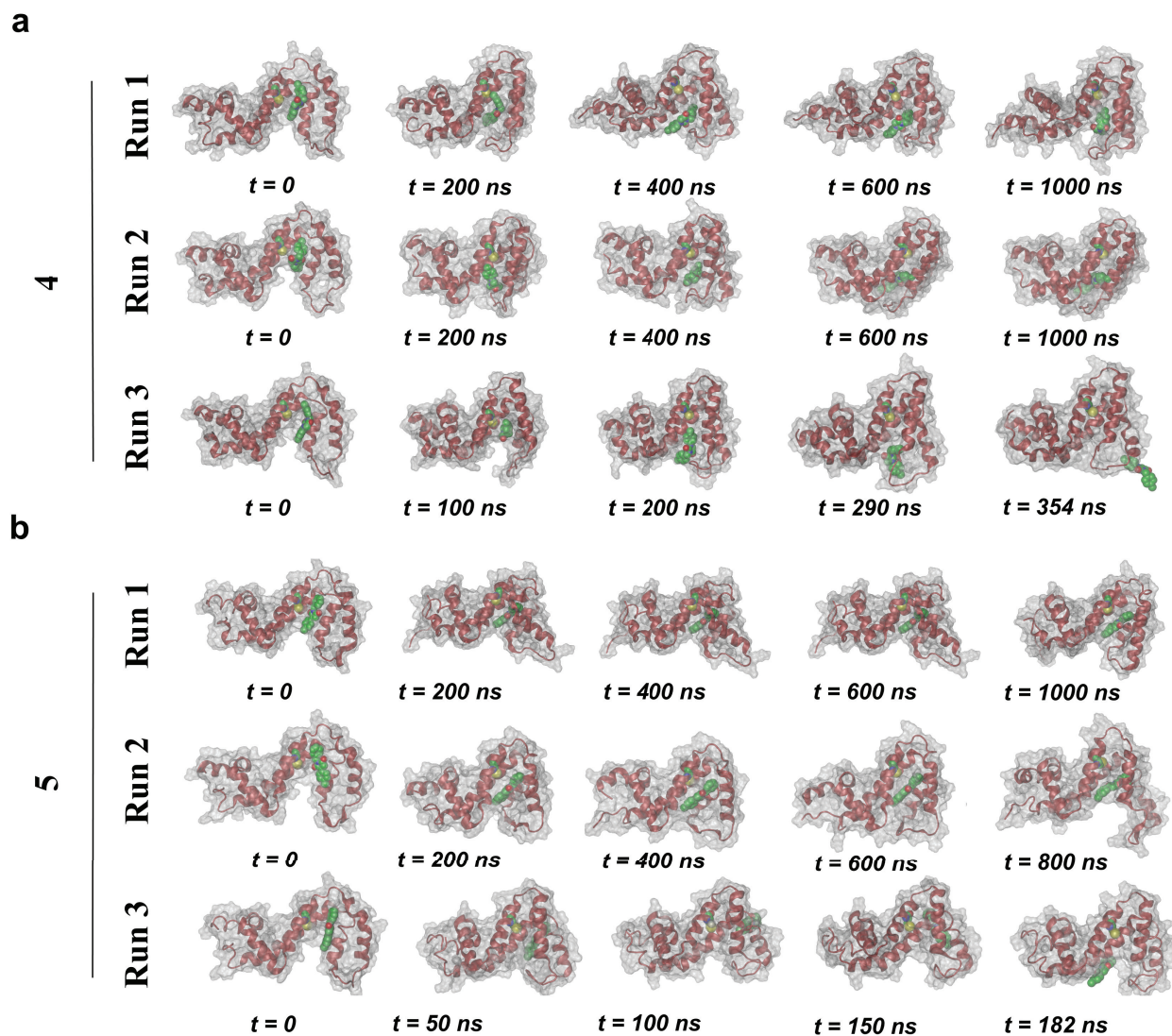


Figure F.4: Snapshots at various time-points for conformational evolution of complexes of non-TDZD compounds **4** (panel a) and **5** (panel b) with RGS4 (Model 1). In each panel, snapshots from three independent simulation runs are shown for each compound. Coloring and labeling schemes are identical to initial states shown in Fig. 4.3.

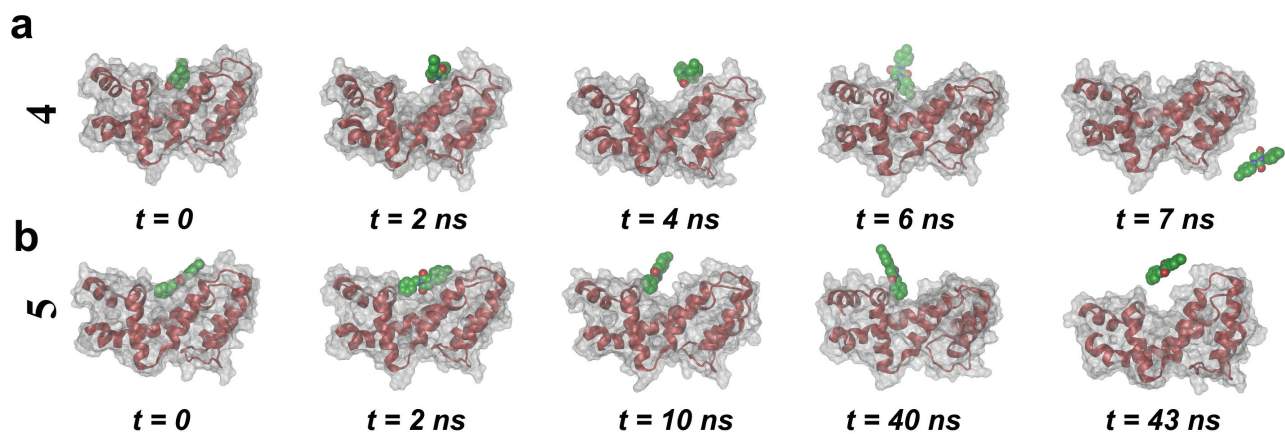


Figure F.5: Snapshots at various time-points for conformational evolution of complexes of non-TDZD compounds **4** (panel a) and **5** (panel b) with RGS4 (Model 2). Coloring and labeling schemes are similar to Fig. F.4.

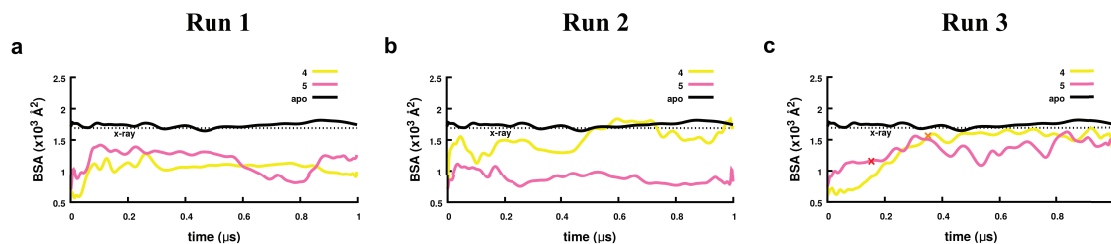


Figure F.6: The traces of buried surface area (BSA) between the α_5 - α_6 helices and the rest of RGS4 *vs.* simulation time (μs) are shown from three independent simulation runs for Model 1. The BSA traces are shown for two non-TDZD compounds (magenta and yellow traces) and from a simulation of apo-RGS4 (black traces). The dotted horizontal line in each panel highlights the BSA-value in the crystal structure of RGS4 (PDB: 1AGR). The symbols (\times) on the BSA traces mark the locations of time-points in Run3 of each compound (panel c) after which compounds diffuse out of the binding pocket.

Model 1

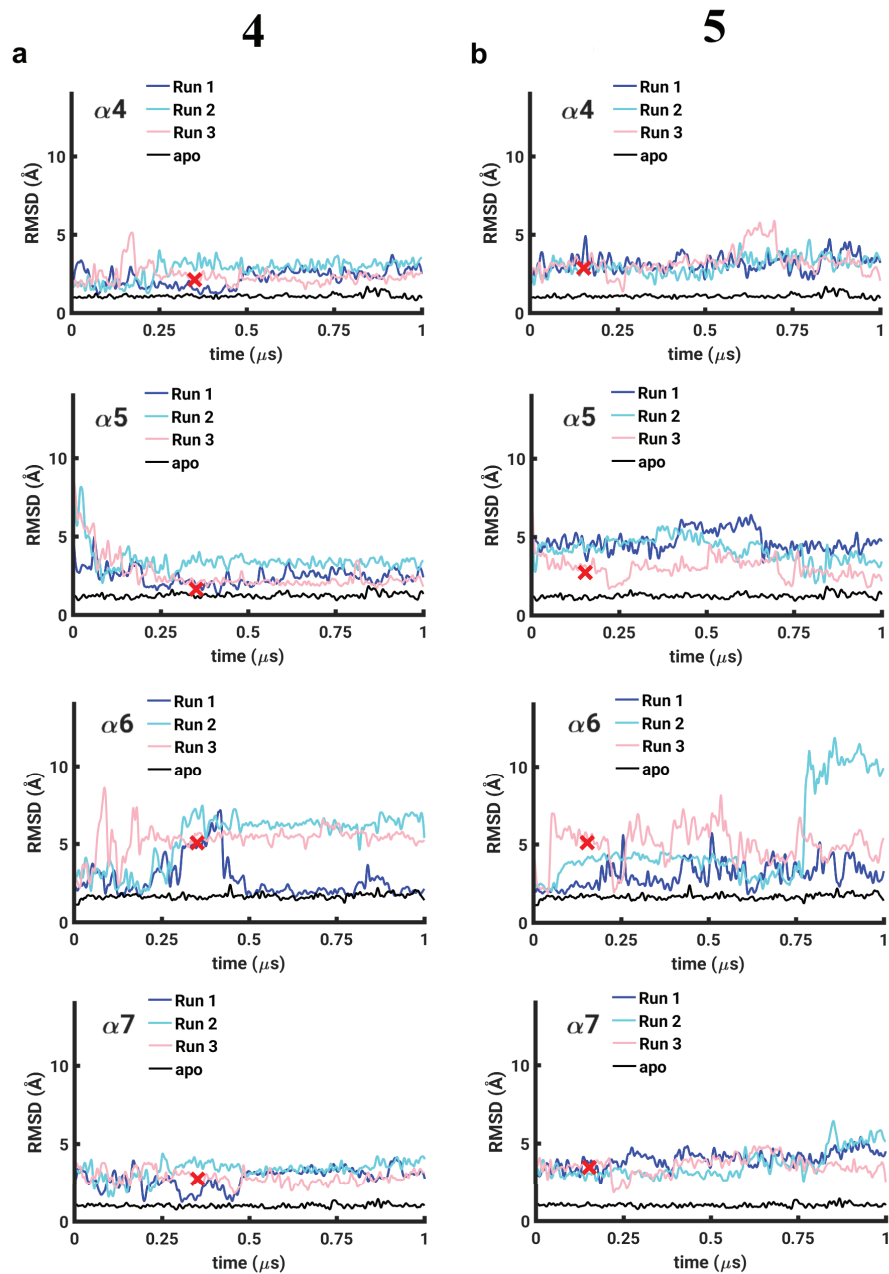


Figure F.7: The RMSD data similar to Fig. F.1 are shown for non-TDZD compounds 4 and 5 from three independent simulations. The red symbol (\times) marks the locations of time-points in Run3 of each compound after which compounds diffuse out of the binding pocket (see snapshots in Fig. F.4).

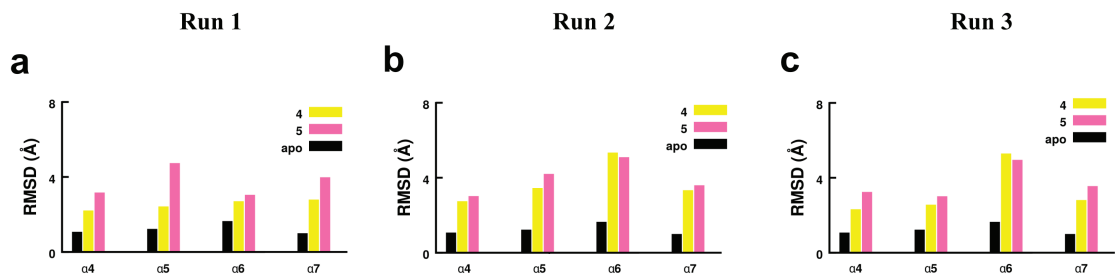


Figure F.8: The histograms of RMSD-averages computed based upon data from each run in Fig. F.7 are shown.

APPENDIX G

SUPPORTING INFORMATION FOR STUDIES REPORTED IN CHAPTER 5

This appendix provides additional figures related to Chapter 5.

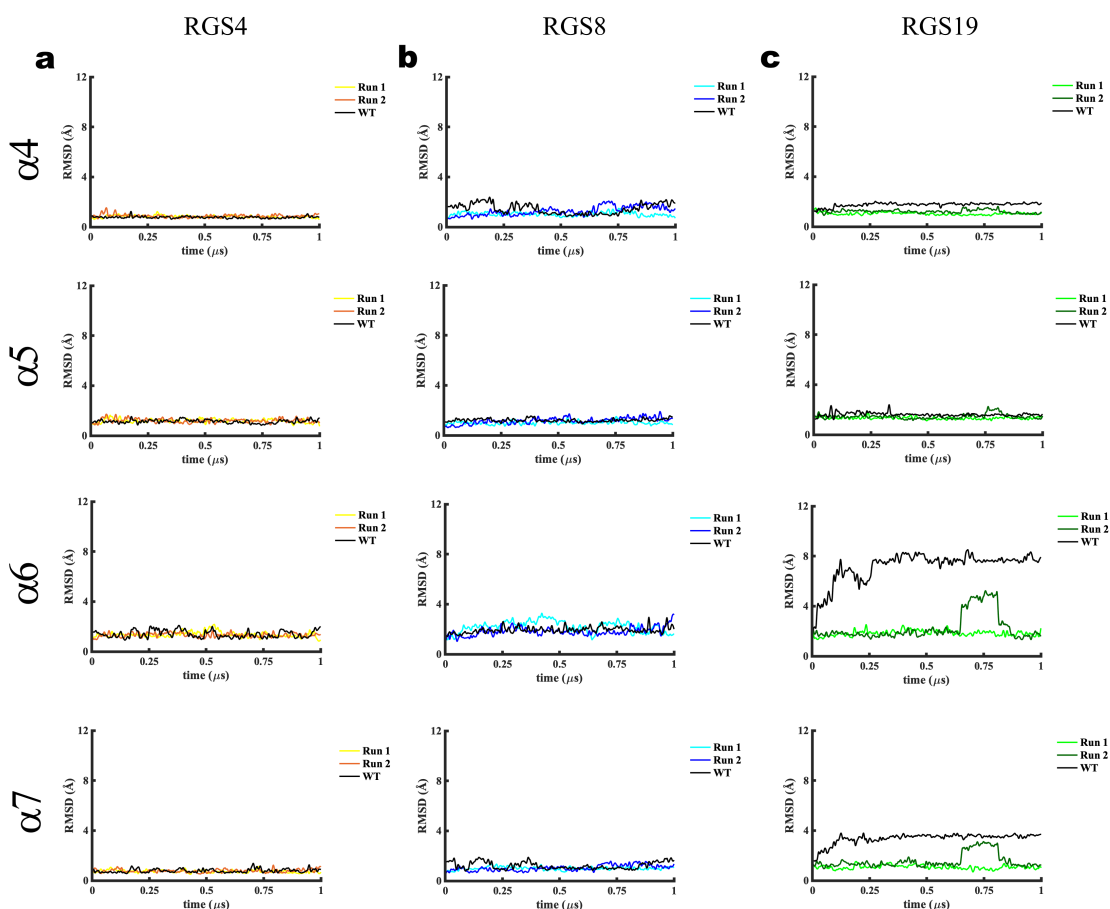


Figure G.1. The traces of root-mean-squared-deviation (RMSD) vs. simulation time (μs) for (a) RGS4 D90L, (b) RGS8 E84L, and (c) RGS19 L118D. Two independent simulation runs for each structure are presented, and the wild-type runs are presented from our previous work [13].

	$\alpha 4$ - $\alpha 5$		% of sim within 4 Å	$\alpha 5$ - $\alpha 6$		% of sim within 4 Å	$\alpha 6$ - $\alpha 7$		% of sim within 4 Å	CCG- 50014 IC ₅₀ (μ M)
RGS4	D90	K125	58.7	S120	S138	-	D130	K155	31.5	8.5
RGS8	E84	R119	44.2	D114	R128	47.5	D124	K149	36.1	>1000
RGS19	L118	K153	-	S148	N166	-	D158	Q183	-	1.1

Figure G.2. The salt-bridge interaction within the $\alpha 4$ - $\alpha 7$ bundle of helices in single-cysteine structures of RGS4, RGS8, and RGS19 from MD simulations and potency of CCG-50014 inhibition of single-cysteine RGS proteins from our previous work [13] (This figure was created by my co-author Vincent Shaw).

APPENDIX H

SUPPORTING INFORMATION FOR STUDIES REPORTED IN CHAPTER 6

This appendix provides additional figures related to Chapter 6.

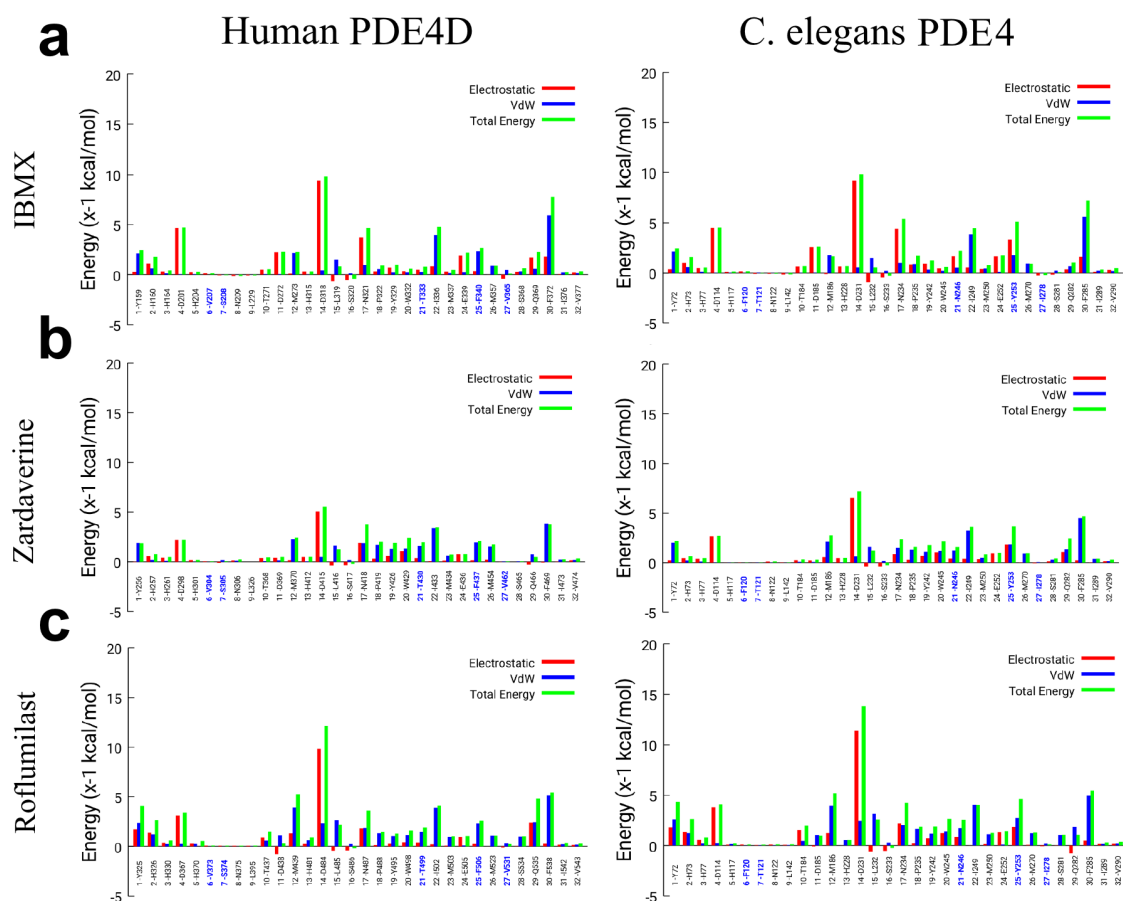


Figure H.1. The nonbonded interaction energy analysis between residues in the inhibitor binding pocket of PDE4D and *C. elegans* PDE4 for the first simulation run. See Fig. 6.1a and 6.1b for depictions of the binding pocket and the 32 residues analyzed with bound (a) IBMX, (b) zardaverine, and (c) roflumilast. Amino acid residues in blue text denote residues that differ between human and *C. elegans* PDE4 sequences.

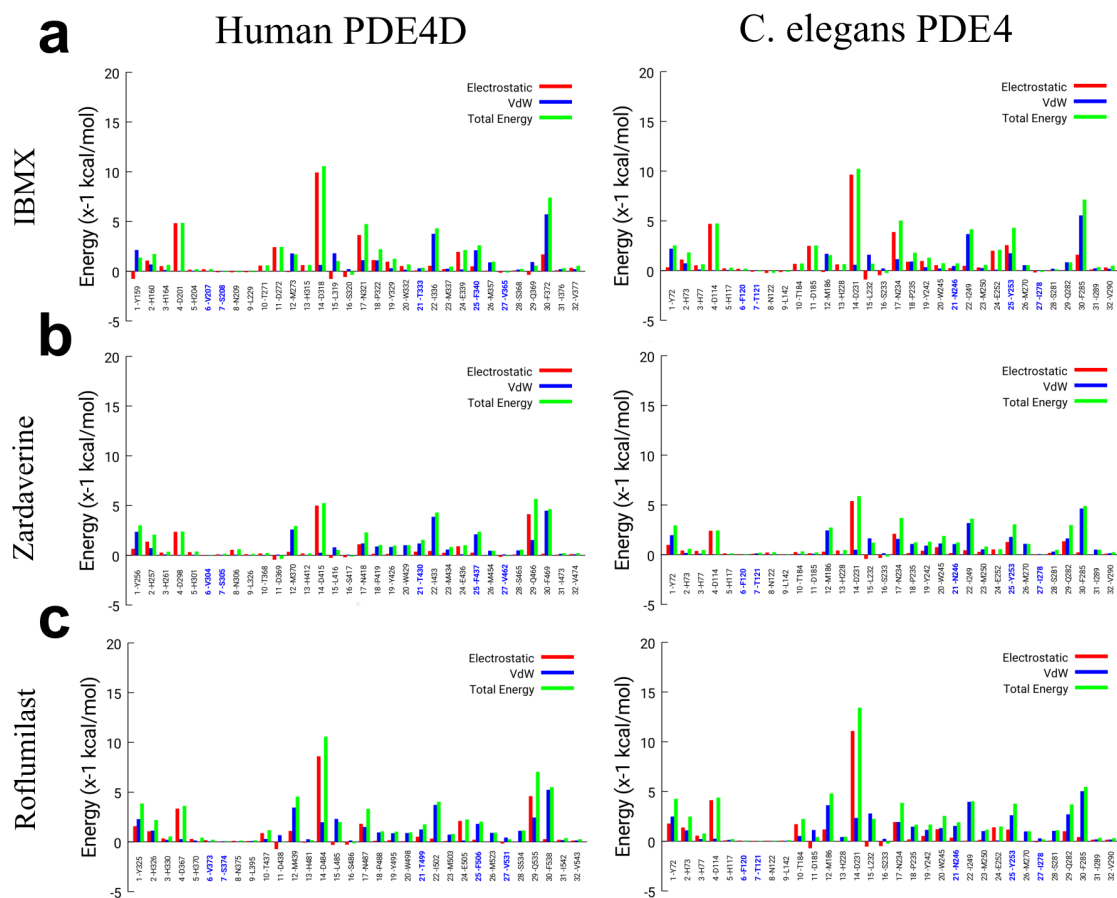


Figure H.2. The nonbonded interaction energy analysis between residues in the inhibitor binding pocket of PDE4D and *C. elegans* PDE4 for the second simulation run. (a) IBMX, (b) zardaverine, and (c) roflumilast. Amino acid residues in blue text denote residues that differ between human and *C. elegans* PDE4.

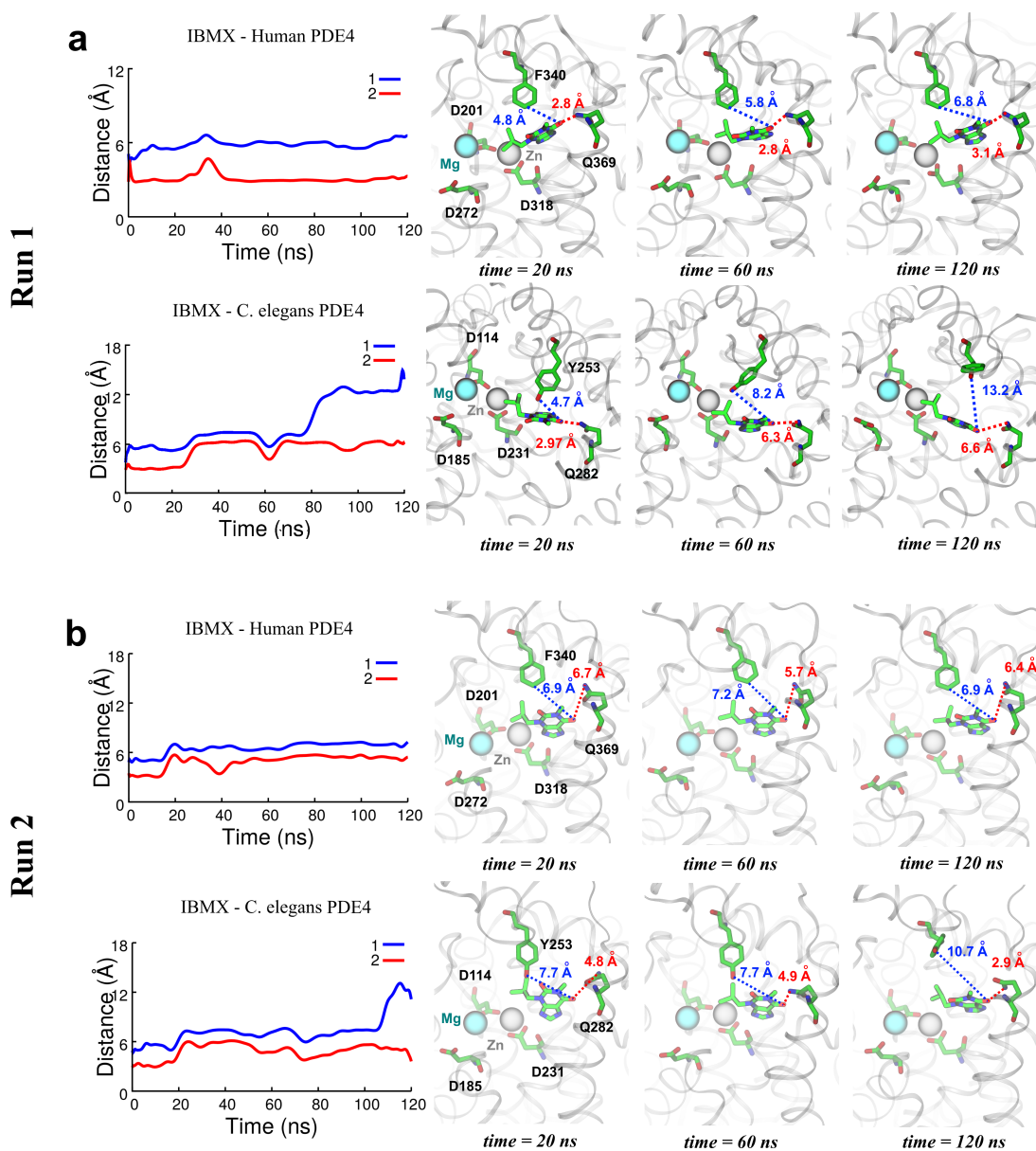


Figure H.3. Interatomic distances between C4 atom of F506(human)/O atom on the side chain of Y253(*C. elegans*) (blue dashed line, labeled 1) or the N δ atom of Q369/Q282 (red dashed line, labeled 2) and the O6 oxygen of IBMX bound to human PDE4D or *C. elegans* PDE4 obtained from two independent MD simulation runs, (a) run 1 and (b) run 2.

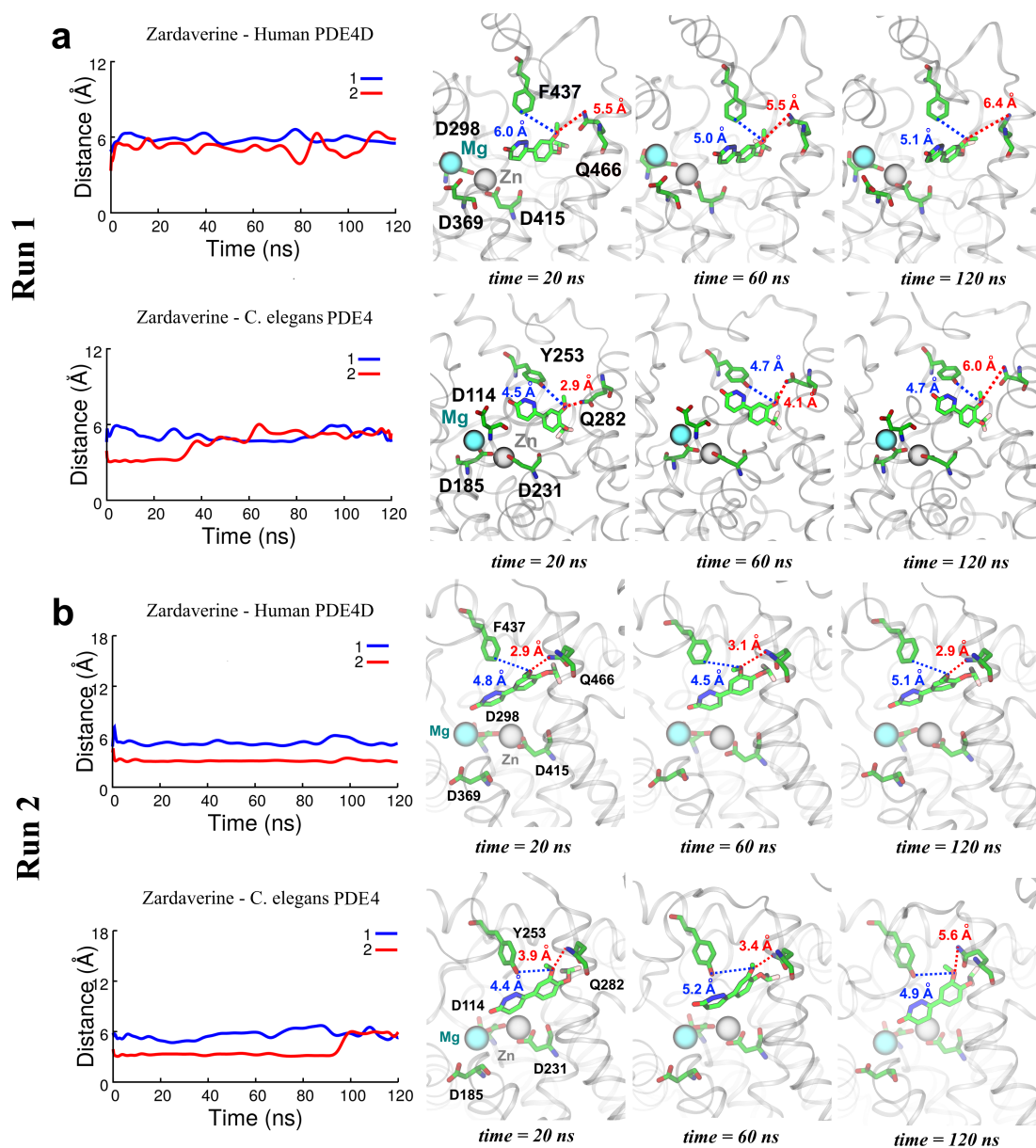


Figure H.4. Interatomic distances between C4 atom of F506(human)/O atom on the side chain of Y253(*C. elegans*) (blue dashed line, labeled 1) or the N δ atom of Q369/Q282 (red dashed line, labeled 2) and the O4 oxygen of zardaverine bound to human PDE4D or *C. elegans* PDE4 obtained from two independent MD simulation runs, (a) run 1 and (b) run 2.

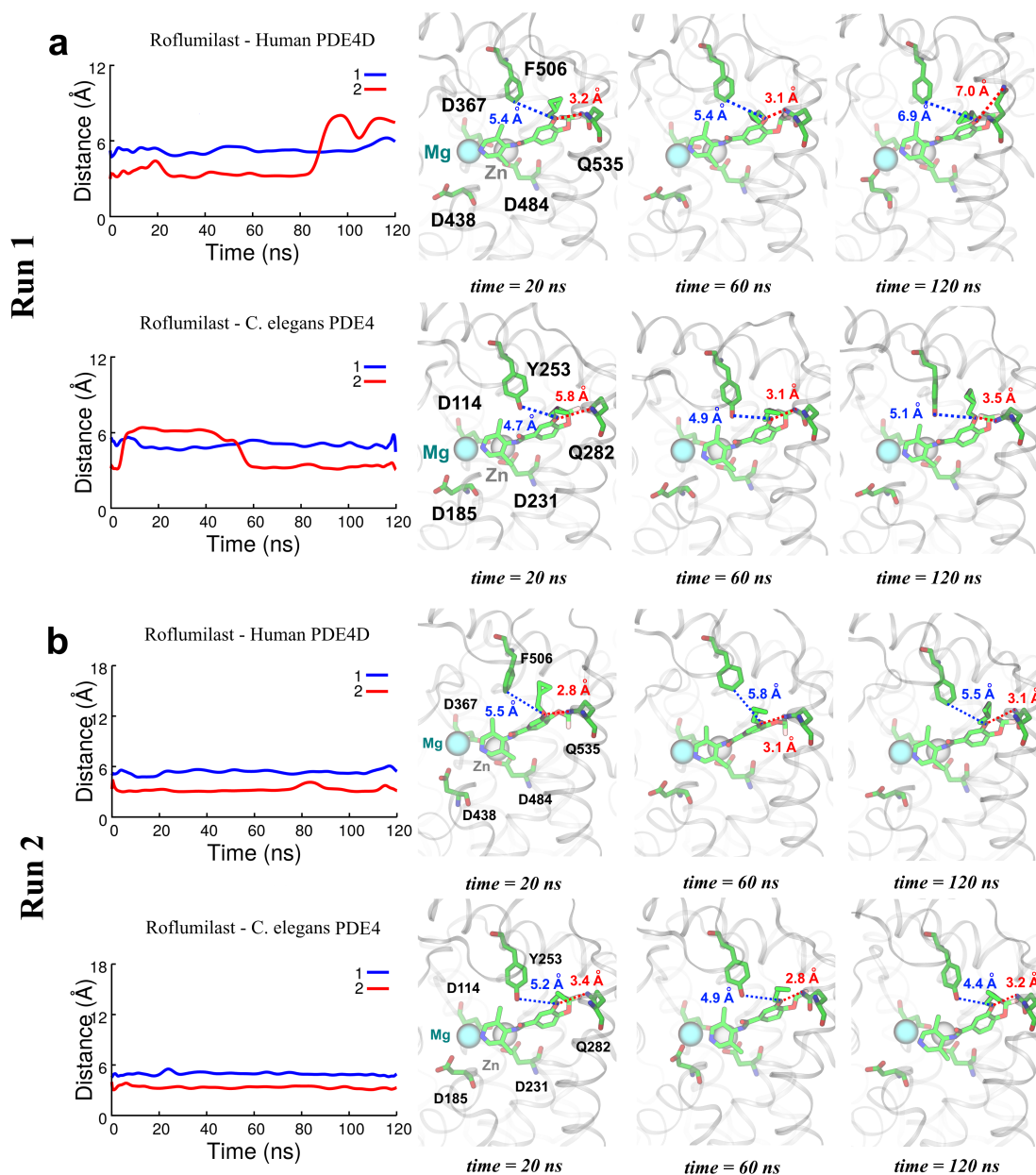


Figure H.5. Interatomic distances between C4 atom of F506(human)/O atom on the side chain of Y253(*C. elegans*) (blue dashed line, labeled 1) or the N δ atom of Q369/Q282 (red dashed line, labeled 2) and the O4 oxygen of roflumilast bound to human PDE4D or *C. elegans* PDE4 obtained from two independent MD simulation runs, (a) run 1 and (b) run 2.

APPENDIX I
JOURNAL COVER IMAGE

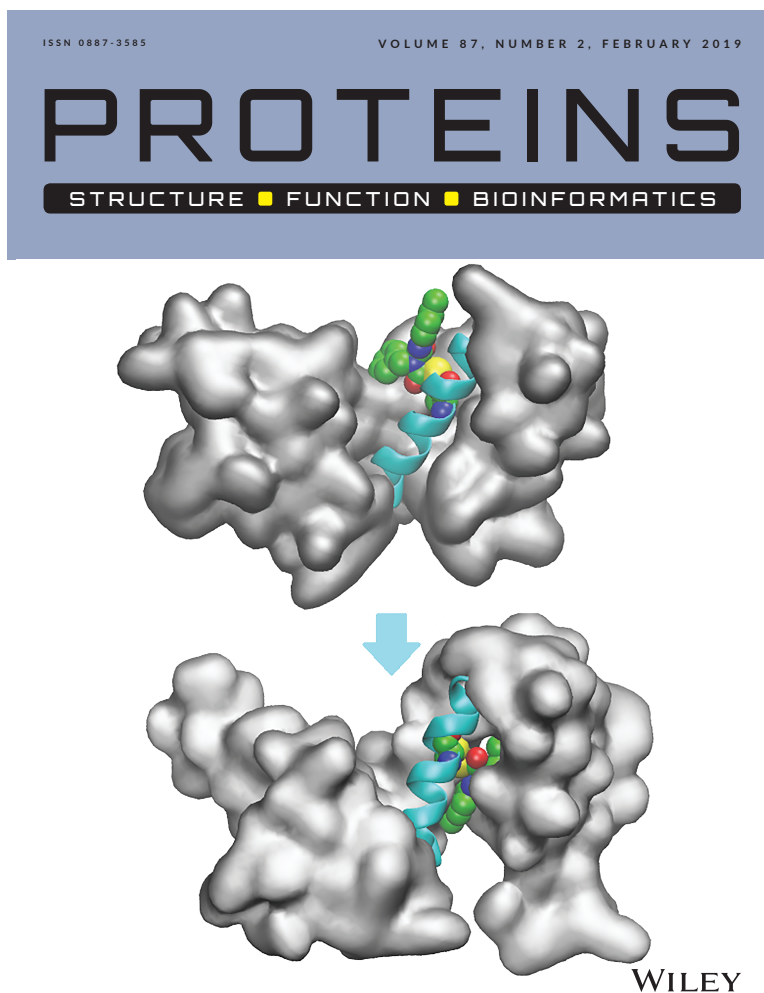


Figure I.1. Shown is the journal cover image [17] related to work described in chapter 4 and in Ref. [14].

APPENDIX J

CURRICULUM VITAE

J.1 Education

University of New Hampshire, Ph.D., Chemical Engineering, May 2019

University of Southern California, M.Sc. in Chemical Engineering, May 2013

Amirkabir University of Technology, B.Sc. in Chemical Engineering, August 2010

J.2 Journal Publications

Shaw, V.S., **Mohammadi, M.**, Quinn, J.A., Vashisth, H., and Neubig, R.R. (20XX). “An interhelical salt bridge controls flexibility and inhibitor potency for regulators of G-protein signaling (RGS) proteins 4, 8, and 19.” *Molecular Pharmacology*, (Submitted).

Schuster, K.D., **Mohammadi, M.**, Cahill, K.B., Matte, S.L., Maillet, A.D., Vashisth, H., and Cote, R.H. (2019). “Pharmacological and molecular dynamics analyses of differences in inhibitor binding to human and nematode PDE4: implications for management of parasitic nematodes.” *PloS ONE* 14(3): e0214554. (PDF)

Mohammadi, M., Mohammadiaran, H., Shaw, V.S., Neubig, R.R., and Vashisth, H. (2019). “Interplay of cysteine exposure and global protein dynamics in small-molecule recognition by a regulator of G-protein signaling protein.” *Proteins: Structure, Function, and Bioinformatics*, 87(2), 146-156 (Featured on the Journal Cover). (PDF)

Liu, Y., **Mohammadi, M.**, and Vashisth, H. (2018). “Diffusion networks of CO in FeFe-Hydrogenase.” *The Journal of Chemical Physics* 149(20), 204108. (PDF)

Mohammadi, M., and Vashisth, H. (2017). “Pathways and thermodynamics of oxygen diffusion in [FeFe]-Hydrogenase.” *The Journal of Physical Chemistry B*, 121(43), 10007-10017. (PDF)

J.3 Conference Publications

Shaw, V.S., Quinn, J.A., **Mohammadi, M.**, Vashisth, H., and Neubig, R.R. (2019). “A salt bridge between $\alpha 4$ and $\alpha 5$ helices drives differences in flexibility and potency of inhibition

among regulator of G-protein signaling (RGS) proteins” *The FASEB Journal 33:1_supplement 784.16*, Experimental Biology 2019 Meeting, Orlando, FL, USA.

Shaw, V.S., Quinn, J.A., **Mohammadi, M.**, Vashisth, H., and Neubig, R.R. (2018). “A Salt Bridge Between $\alpha 4$ and $\alpha 5$ Helices Drives Flexibility and Potency Differences Among Regulator of G-protein Signaling (RGS) Proteins” *The 19th Annual G-Protein Coupled Receptors Retreat*, Detroit, MI, USA.

Mohammadi M. and Vashisth H. (2018). “Computational studies of regulators of G-proteins signaling small-molecule inhibitors” *Abstracts of papers of the American Chemical Society, 256th American Chemical Society National Meeting*, Boston, MA, USA.

Liu Y., **Mohammadi M.**, and Vashisth H. (2018). “Network of inhibitor diffusion pathways in hydrogenase enzymes” *Abstracts of papers of the American Chemical Society, 256th American Chemical Society National Meeting*, Boston, MA, USA.

Liu Y., **Mohammadi M.** and Vashisth H. (2018). “Diffusion Network of CO in [FeFe]-Hydrogenase” *Proc. 78th Annual Physical Electronics Conference*, Durham, NH, USA.

Mohammadi, M. and Vashisth, H. (2018). “Molecular dynamics studies of regulators of signaling proteins inhibitors” *University of New Hampshire Bioengineering Symposium*, Durham, NH, USA.

Mohammadi, M. and Vashisth, H. (2018). “Molecular dynamics studies of allosteric inhibitors for protein-protein interfaces” *20th Annual Northeast Student Chemistry Research Conference*, Boston, MA, USA.

Mohammadi M. and Vashisth H. (2017). “Mapping of gas diffusion pathways in [FeFe]-Hydrogenase” *Proc. 2017 American Institute of Chemical Engineers Annual Meeting*, Minneapolis, MN, USA.

Mohammadi, M. and Vashisth, H. (2017). “Mapping of gas diffusion pathways in hydrogenase enzymes” *19th Annual Northeast Student Chemistry Research Conference*, Boston, MA, USA.

J.4 Honors and Awards

Best Poster Award, University of New Hampshire Bioengineering Symposium	Spring 2018
Anton Supercomputer Workshop Invitee, Pittsburgh Supercomputing Center	Fall 2017
Member, Golden Key International Honor Society	Fall 2015