Keep on Moving: Discovering and Perturbing the Conformational Dynamics of Enzymes

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CONSPECTUS: Because living organisms are in constant motion, the word “dynamics” can hold many meanings to biologists. Here we focus specifically on the conformational changes that occur in proteins and how studying these protein dynamics may provide insights into enzymatic catalysis.

Advances in integrating techniques such as X-ray crystallography, nuclear magnetic resonance, and electron cryomicroscopy (cryo EM) allow us to model the dominant structures and exchange rates for many proteins and protein complexes. For proteins amenable to atomic resolution techniques, the major questions shift from simply describing the motions to discovering their role in function. Concurrently, there is an increasing need for using perturbations to test predictive models of dynamics–function relationships. Examples are the catalytic cycles of dihydrofolate reductase (DHFR) and cyclophilin A (CypA).

In DHFR, mutations that alter the ability of the active site to sample productive higher energy states on the millisecond time scale reduce the rate of hydride transfer significantly. Recently identified rescue mutations restore function, but the mechanism by which they do so remains unclear. The exact role of any changes in the dynamics remains an open question. For CypA, a network of side chains that exchange between two conformations is critical for catalysis. Mutations that lock the network in one state also reduce catalytic activity.

A further understanding of enzyme dynamics of well-studied enzymes such as dihydrofolate reductase and cyclophilin A will lead to improvements in the ability to modulate the functions of computationally designed enzymes and large macromolecular machines. In designed enzymes, directed evolution experiments increase catalytic rates. Detailed X-ray studies suggest that these mutations likely limit the conformational space explored by residues in the active site. For proteins where atomic resolution information is currently inaccessible, other techniques such as cryo-EM and high-resolution single molecule microscopy continue to advance.

Understanding the conformational dynamics of larger systems such as protein machines will likely become more accessible and provide new opportunities to rationally modulate protein function.

INTRODUCTION

Even simple enzymes must bind a substrate, catalyze a reaction, and release a product. It is difficult to imagine an enzyme performing all of these actions from a single conformational state. The chemical changes after the substrate has converted to product elicit changes in the conformational ensemble populated by the enzyme (Figure 1). While advances in computational simulations are expanding our ability to see these changes at longer time scales,1 here we focus on determining the conformational ensembles that enzymes populate and how these ensembles change during functional catalytic cycles.

X-ray crystallography and NMR spectroscopy can often be used in concert to answer a question that could not be answered by any single technique.2,3 High-resolution (<1.4 Å) X-ray crystallography, ideally at room temperature, and NMR experiments, including measuring J-coupling constants,4 can be combined to determine the structural basis of excited states critical for transit through catalytic cycles (as in the case of dihydrofolate reductase (DHFR) and cyclophilin A (CypA) discussed below).5,6 However, not all enzymes are amenable to these techniques, and many of the central functional questions of protein dynamics at the domain and complex level can be answered using relatively low-resolution structural data paired with kinetic analyses.

For enzymes amenable to high-resolution techniques, such as DHFR and CypA, these technical advances are catalyzing a shift from asking “can we describe protein dynamics” to “what is the role of protein dynamics in function”. Gaining a deeper understanding of these dynamics has important implications for how we design enzymes for medicinal and biotechnology applications.
understanding of these protein dynamics can be extremely challenging; especially important is resisting the urge to ascribe a function to all motions observed by a specific experimental or computational approach. Structural models of enzyme motions are key to understanding whether sampling of specific conformations is essential for properly orienting the substrate during catalysis, directly addressing the controversy of whether protein motions are coupled to catalytic function.

Making the link between dynamics and function requires making some perturbation to the system and assessing the effect of this perturbation. These perturbations could include mutations, changes in the environment of the enzyme, or in more complicated systems, changes in other components interacting with the enzyme. It is critical when making mutations to consider all possible effects of the mutation in addition to the effect on protein dynamics, in order to truly understand the role of the mutated residue. This includes understanding the role the mutated residue plays in the structural integrity of the protein and the function of the protein. It is important to note that rigidity and flexibility are simply two extremes of the "protein dynamics" spectrum. Functionally important changes to protein dynamics can act by enhancing rigidity, enhancing flexibility, or altering the correlated motions of residues. For larger and more complicated systems, including protein machines, our level of understanding is often not as detailed compared with that for simple, single-domain, small enzymes. Therefore, the major questions for these machines may still be focused on defining the conformational changes that occur and the relative repositioning of subunits within the folded complex.

### THE ROLE OF DYNAMICS IN DIHYDROFOLATE REDUCTASE

Dihydrofolate reductase (DHFR) catalyzes the stereospecific reduction of dihydrofolate (DFH) to tetrahydrofolate (THF) (Figure 2a). Five stable intermediates are observed in the catalytic cycle of *Escherichia coli* DHFR (ecDHFR): the holoenzyme, ecE:NADPH; the Michaelis complex, ecE:NADPH:DFH; and the three product ternary complexes ecE:NADP':THF, ecE:THF, and ecE:NADPH:THF.10–13

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**Figure 1.** Protein dynamics occur at different time scales. (a) Motions on the picosecond to nanosecond scale involve small changes in backbone or side chain torsion angles. Calcium bound calmodulin (1exr, upper) exhibits conformational heterogeneity on the interface of the peptide binding site. The residual conformational entropy of binding depends on side chains sampling alternative conformations as exemplified by Met36 and Leu39 (lower). Electron density contoured to 2.5 e/Å³ in a dark blue mesh and 0.8 e/Å³ in cyan volume representation. The lessons from calmodulin likely apply to enzymes where the loss of conformational entropy associated with the rigidification of active-site loops or side chains can specifically weaken binding to substrate or product complexes and promote flux through the catalytic cycle. (b) A model of ubiquitin (2k39) derived from RDC data reporting on motions up to microseconds is shown as cartoon, with the other models in the ensemble shown as transparent ribbons (upper). The dynamic β1β2 loop moves between alternative loop conformations, represented as sticks (upper and lower). The population of the up (cyan), mid (blue), and down (purple) β1β2 conformations can be a critical determinant of binding preferences for protein-protein interactions. The rates of transition between these states discriminate between induced fit and conformational selection mechanisms, which can influence catalytic mechanisms and inhibitor discovery. (c) For enzymes, loop motions on the millisecond time scale are often rate limiting for catalytic cycles, with essential roles for governing ligand flux and repositioning key catalytic residues for catalysis. The WPD loop of protein tyrosine phosphatase 1B (PTP1B) moves between the "closed" (1sug, orange) and "open" (1h49, cyan) form on the millisecond time scale, forming the catalytically competent closed active site conformation. Further molecular detail of the two conformations are shown in the lower panel with electron density contoured to 0.3 e/Å³. (d) The archaeal proteasome, a ~700-kDa complex, controls active site access through the dynamic exchange of the N-terminus to block or reveal the central pore on the time scale of seconds. The structure of the proteasome is shown as a homohexameric complex, each subunit in a different color (upper). In the lower panel, the ensemble of structures of the N-terminus of one of the seven subunits is shown in blue (2kui1). (e) Many enzymes enter long-lived states, with distinct catalytic activities, through stochastic fluctuations. Quaternary structure reconstruction of two RAS molecules (yellow) and son of sevenless (SOS) (gray surface) complex. The Cdc25, REM, DH, histone, and PH domains of SOS are colored blue, green, orange, brown, and red, respectively (1xd2 (RAS) and 3ksy (SOS)). This complex exchanges between long-lived states with distinct catalytic rates. The structural basis of this exchange is currently unknown but likely involves rearrangements of protein-protein interfaces shown schematically in equilibrium. (f) The folded crystal structure (1ssx) of DHFR is shown as a ribbon diagram. The β1 loop moves between alternative loop conformations, represented as sticks (upper and lower). The population of the up (cyan), mid (blue), and down (purple) β1β2 conformations can be a critical determinant of binding preferences for protein-protein interactions.
Met20 loop was observed in three dominant conformations: (Met20 loop, residues 9-20) undergoes a conformational change in the active site loop instability of THF. The crystal structures revealed that ecDHFR (E:NADPH:DHF) was used as a stable analog of THF, due to the (E:NADP+:THF, E:THF, and E:NADPH:THF) adopt the conformation, whereas the three product complexes stabilized by hydrogen bonds between Asn23 and the backbone and side chain of Ser148. For binary substrate complexes, the Met20 loop packs tightly against the nicotinamide ring of the cofactor, while in the “occluded” conformation, it projects into the active site and sterically blocks (“occludes”) the nicotinamide-binding pocket, which is compatible with the diffusion of the nicotinamide ring out of the active site (Figure 2a).

X-ray structures within the same space group of the holoenzyme (E:NADPH) and the model Michaelis complex (E:NADP+:FOL) show that these complexes adopt the closed conformation, whereas the three product complexes adopt the occluded conformation. The occluded conformation is stabilized by hydrogen bonds between Asn23 and the backbone and side chain of Ser148. For binary substrate complexes, the M20 loop conformation was dependent on space-group, being “occluded”, “open”, or even disordered, with no clear electron density for the majority of the loop. While crystallographic analysis in a variety of space groups showed the Met20 loop to be in three dominant conformations, NMR data including chemical shift analyses and NOEs showed that in solution the enzyme is predominantly either in the “closed” conformation or in the “occluded” conformation and never stably in the “open” conformation.

NMR relaxation dispersion experiments on the five stable intermediates of the catalytic cycle showed that each intermediate samples higher energy “excited states”, whose structural features resemble the preceding or following intermediates.20 Substrate and cofactor exchange likely depends on these excited states, suggesting that the ligand-dependent modulation of these protein conformational dynamics is important as the enzyme proceeds through its catalytic cycle. A mutation in the Met20 loop of ecDHFR (N23PP or N23PP/S148A) inhibits the closed-to-occluded transition and also inhibits the sampling of productive higher-energy states on the millisecond time scale, as assayed by NMR relaxation dispersion.21 Remarkably, enzyme kinetic experiments revealed that hydride transfer, the chemical step of the enzyme reaction, was severely impaired in this mutant. Further analysis using high-resolution room temperature crystallography coupled with multiconformer model building using qFit22 and automated “pathway analysis” using CONTACT23 suggested that the mutation results in an increase in nonproductive, frustrated motions, while the concerted dynamics in the Met20 loop are inhibited. These results led to our current view: the mutation alters millisecond time scale conformational fluctuations, affecting the probability of populating conformations that are conducive to the optimal transition state configuration. As expected, the N23PP and N23PP/S148A mutations also affect ligand flux.

Guided by comparisons to the homologous human DHFR,23 “rescue” mutations in the N23PP background have been identified24,25 (Figure 2b,c). These mutations increase the hydrophobic packing around the substrate molecule. It is currently an open question how these mutations affect conformational dynamics of the protein, if at all, and the mechanism by which they rescue hydride transfer rates remains to be determined. Many computational and experimental studies have also examined flexibility changes introduced by the G121V mutation.26,27 While simulations have provided initial explanations of these effects, a current challenge for computa-

Figure 2. Dynamics in DHFR. (a) Crystal structures of E. coli DHFR show the Met20 loop in the occluded (1rx4, blue) and closed (1rx2, red) conformations. During the catalytic cycle of DHFR, this loop fluctuates between these conformations on the millisecond time scale. The ligands NADPH (left ligand) and folate (right ligand) are shown in orange and yellow, respectively. (b) Mutation of Asn23 to two proline residues (N23PP) shown as sticks in red and Ser148 to alanine (S148A) shown in pink reduce activity of ecDHFR. Mutation of Gly51 to the sequence PEKN (shown in blue) partially recovers the catalytic activity. The activity is increased further by the Leu28Phe (L28F, green) mutation. The structure of N23PP/PEKN (4gh8) is shown with NADPH shown in orange, with substrate mimic methotrexate in tan. (c) pH independent hydride transfer rates of different mutants show the quantitative effects of mutations that alter the dynamics of the Met20 loop and packing around the substrate.
tional approaches is to rationalize both the local and long-range effects across the large activity ranges spanned by these mutations and to make predictions of how other mutations that tune these parameters will manifest experimentally.28–31

![Figure 3. Dynamics in CypA.](image)

**THE ROLE OF DYNAMICS IN CYCLOPHILIN A**

Human cyclophilin A (CypA) belongs to the proline isomerase family of enzymes, which play critical roles in protein folding, signaling, and the immune response.32,33 The reaction catalyzed by CypA, cis–trans proline isomerization, does not make or break new chemical bonds, making it possible to detect conformational exchange of both the substrate and the saturated enzyme during catalysis by NMR. Both backbone and side chain NMR dynamics experiments have been performed at various time scales revealing a network of residues that undergo two-state conformational exchange, which are thought to represent the enzyme's conformation when bound to cis- or trans-proline substrates (Figure 3a).34,35 Interestingly, NMR relaxation dispersion experiments show that similar conformational switching is observed in the free enzyme. These rates correspond closely to the sum of the rate constants of cis-to-trans and trans-to-cis isomerization (k_{ex} = 2500 s^{-1}) suggesting that the rate of the catalytic cycle is governed by the intrinsic protein dynamics.

While NMR studies provided numerous insights and hinted that the minor conformation sampled in the free enzyme was relevant to catalysis, a detailed structural picture of the minor conformation was not deciphered from NMR experiments alone. The chemical shift differences between the excited and ground state provided few hints about the structural nature of the "excited" state. Examination of low levels of electron density, derived from new high resolution X-ray data collected at room temperature, uncovered a network of minor side-chain conformations for the residues exhibiting NMR dispersion (Figure 3b). It is worth noting that both visual and automated examination, using Ringer,36 of the electron density on data collected under cryogenic conditions did not uncover the full subset of minor conformations. These results point to several complications for deriving a structural basis for conformational dynamics. First, even though the CPMG experiments conducted here probed changes in the backbone environment, the conformational changes are mostly at the side chain level. Changes in side chain dihedral angles and the proximity of aromatic groups, in particular Phe113, create these changes with little change in backbone torsions. Second, the small energy differences (~1–2 kcal/mol) separating these states render them very susceptible to perturbations like cryocooling or lattice contacts and can have unforeseen consequences, including eliminate evidence for the "excited state" even in high resolution X-ray data.

To further link X-ray and NMR measurements, a mutation (Ser99Thr) was designed to stabilize the network of side-chain rotamers associated with the minor conformation in solution as well as in the crystal (Figure 3c). Like the DHFR N23PP mutant, this mutation ablated detectable exchange on the microsecond to millisecond time scale. Furthermore, enzyme assays showed that shifting the equilibrium toward the side-chain rotamers populated by the minor conformation results in slower enzyme catalysis. Simulations have provided further evidence of a coupling between enzyme and substrate conformations during the catalytic cycle.37 Recent studies incorporating new NMR measurements and simulations point to the importance of considering a conformational ensemble in creating the environment compatible with catalysis and binding both the cis and trans forms of the substrate.38

**DYNAMICS IN DESIGNED ENZYMES**

One of the most exciting recent developments in enzymology is the application of protein design methodologies to create enzymes that catalyze reactions not found in nature.39 Several simple reactions have now been created based on the designs that place catalytic groups in specific orientations relative to a computational model of the transition state.40–42 These initial designs are generally weak catalysts, and directed evolution strategies have been implemented to improve them by several orders of magnitude.43–45 While these evolved enzymes have greatly increased catalytic rates, structural and enzymological studies have revealed that some of the principles implemented in the designs are subverted over the course of multiple rounds of selection. For example, a detailed study of a designed retroaldolase revealed only a modest shift in the reactivity of the lysine and no role for a designed water-mediated interaction...
in catalysis. The design was successful in creating strong hydrophobic binding interactions with the substrate, which parallels lessons from simple micelle systems. A separate study on a distinct designed retroaldolase revealed that a second lysine residue acquired during directed evolution was responsible for greater rate enhancement than the original "catalytic" lysine and that the substrate occupied an entirely new position in the catalytic cavity. Similarly, positioning of the substrate changed dramatically during the course of directed evolution studies of a designed Kemp eliminase (Figure 4a,b). Many rounds of selection resulted in catalysis with efficiency approaching that of natural enzymes and an intricate series of interactions in the active site (Figure 4c). However, examination of low levels of electron density in this high-resolution data set suggests there are still interactions available for optimization (Figure 4d,e). Therefore, even in the most proficient synthetic catalysts, an abundance of highly flexible residues in the active site may be limiting the formation of precise and stable interactions required for catalysis.

A current focus of computational studies on designed enzymes is to use molecular dynamics simulations to screen for candidates with overly flexible active sites or excess solvent penetration. However, collectively these studies suggest that increased active-site conformational flexibility, while detrimental to the chemical step, are likely key to the evolvability of these enzymes: the imprecision of the design process and of nonspecific hydrophobic interactions allows for the eventual tuning of catalytic residues during the course of directed evolution procedures. As the reactions performed by designed enzymes become more demanding, the role of directed evolution in rigidifying the active site will become more important, however, it is important to note that flexibility of second shell residues may still be desirable to ensure efficient transit through the catalytic cycle. Characterizing the energy landscapes of designed enzymes and using this information to improve the design process will be key to realizing the dream of designing synthetic catalysts for many important, yet currently difficult, reactions.

EXTENDING PROTEIN DYNAMICS TO LARGE “PROTEIN MACHINES”

For larger enzymes that often undergo large-scale conformational changes, NMR spectroscopy and high-resolution X-ray crystallography may not be feasible (Figure 5). In these cases, electron cryomicroscopy (cryo-EM) and single-molecule Förster resonance energy transfer (smFRET) or single molecule two-color localization experiments can yield important insights into protein dynamics. For example, smFRET experiments have revealed excited states of the ribosome. Our understanding of protein dynamics using these methods is less atomically detailed and tends to focus on reorientation of large subdomains. However, these experiments will provide key information in guiding future studies on macromolecular machines and complexes. Advances such as recently developed direct electron detectors and sophisticated methods of 3D classification have catapulted cryo-EM to the forefront of understanding protein conformational landscapes in larger systems, at resolutions overlapping with the low end of those...
Figure 5. The challenge of larger protein machines. The motor domain of dynein, a microtubule-based motor protein belonging to the AAA family of enzymes, is shown, colored by domain (3VKH). The entire dynein heavy chain is considerably larger and is difficult to produce in quantities required for structural biology studies. For comparison, DHFR (1RX2) and CypA (2CPL) are shown to scale.

achieved by X-ray crystallography. While the resolutions achievable are still lower than what is required to understand enzyme catalysis at as detailed a level as available for DHFR or CypA, a fairly thorough picture of larger scale conformational changes can be developed through comparison of density maps and de novo models. Using these tools to probe the conformational landscape of larger protein machines will provide exciting new insights into their function.

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Notes
The authors declare no competing financial interest.

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Justin Biel received his bachelor's in biochemistry and biophysics at Oregon State University, working on ultra high resolution X-ray crystallography analysis in the lab of P. Andrew Karplus. In 2013, Justin began graduate school at University of California, San Francisco, where he is an NSF Graduate Research Fellow.

James Fraser was a graduate student at UC Berkeley with Tom Alber, where he developed new experimental and computational methods to investigate protein conformational dynamics. In 2011, James moved to UCSF, where his lab focuses on the role of protein motions in catalysis, ligand binding, and allostery.

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