

Computational Methods for Modifying Enzyme Specificity:  
from Molding the Active Site to Allosteric Considerations

by

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## ABSTRACT

Enzymes keep life nicely humming along by catalyzing important reactions at relevant timescales. Despite their immediate importance, how enzymes recognize and bind their substrate in a sea of cytosolic small molecules, carry out the reaction, and release their product in microseconds is still relatively opaque. Methods to elucidate enzyme substrate specificity indicate that the shape of the active site and the amino acid residues therein play a major role. However, lessons from Directed Evolution experiments reveal the importance of residues far from the active site in modulating substrate specificity. Enzymes are dynamic macromolecules composed of networks of interactions integrating the active site, where the chemistry occurs, to the rest of the protein. The objective of this work is to develop computational methods to modify enzyme ligand specificity, either through molding the active site to accommodate a novel ligand, or by identifying distal mutations that can allosterically alter specificity.

To this end, two homologues in the  $\beta$ -lactamase family of enzymes, TEM-1, and an ancestrally reconstructed variant, GNCA, were studied to identify whether the modulation of position-specific distal-residue flexibility could modify ligand specificity. RosettaDesign was used to create TEM-1 variants with altered dynamic patterns. Experimental characterization of ten designed proteins indicated that mutations to residues surrounding rigid, highly coupled residues substantially affected both enzymatic activity and stability. In contrast, native-like activities and stabilities were maintained when flexible, uncoupled residues, were targeted. Five of the TEM-1 variants were crystallized to see if the changes in function observed were due to architectural changes to the active site. In a second project, a computational platform using RosettaDesign was

developed to remodel the firefly luciferase active site to accommodate novel luciferins. This platform resulted in the development of five luciferin-luciferase pairs with red-shifted emission maxima, ready for multicomponent bioluminescent imaging applications in tissues. Although the projects from this work focus on two classes of proteins, they provide insight into the structure-function relationship of ligand specificity in enzymes and are broadly applicable to other systems.

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## CHAPTER 1

### INTRODUCTION

#### 1.1 A Brief History of Enzymes

The word *enzyme* was first used in a biochemical context by Wilhelm Kühne in 1878 to describe the catalytic activity present in yeast extracts (Sumner & Somers, 2014). Prior to this nomenclature assignment, microorganisms had been used by humans for thousands of years for fermentation and cheese-making, but the nature of the molecule performing the chemical reaction was unknown and would remain unknown until 1946. Nobel Prize winning work by James B. Sumner (prize share  $\frac{1}{2}$ ), Wendell M. Stanley (prize share  $\frac{1}{4}$ ), and John B. Northrop (prize share  $\frac{1}{4}$ ) on the crystallization of urease, pepsin, trypsin and chymotrypsin, revealed that enzymes were made of amino acids (Nobel Foundation, 1964). We now know that enzymes are protein catalysts that play an integral role in metabolism, cell signaling, immunity, and gene expression, to name a few. Without enzymes, reactions that keep life nicely humming in a cell would take years or potentially even longer to occur and life on the timescale that we are used to would cease to exist.

Nature has naturally evolved enzymes to perform chemistries important to life in the organism housing them. These enzymes vary in catalytic ability from one of the slowest, but most abundant enzymes, rubisco that turns over 3 molecules of carbon dioxide every second (Bar-On & Milo, 2019); to one of the fastest enzymes, carbonic anhydrase that can turn over  $10^6$  carbon dioxide molecules a second (Lindskog & Colemant, 1973). Most enzyme efficiencies fall in between these two extremes. Since nature only optimizes enzymes to be as stable and catalytically efficient as natural

selection dictates, most native proteins do not naturally possess functions important to human applications. Thus, the field of protein engineering is built around the desire to modify or create proteins with functions important to human lives.

## **1.2 Methods in Protein Engineering**

Protein engineering, the process of altering protein sequences to provide variants with new or enhanced properties, was born in the 1960s when Christian Anfinsen (Nobel Prize in Chemistry in 1972) demonstrated that a protein's amino acid sequence alone determines its structure which in turn determines its function (Forsen, 1993). Initially, changes to protein sequence were performed on an organismal scale through the process of strain optimization (Heckmann & Paradisi, 2020). In this technique, entire organisms were placed in mutation-inducing conditions like radiation or chemical agents. The resulting strains carrying random mutations were then screened for beneficial phenotypes. Although effective in specific cases like in the development of penicillin acylase by Bayer, (Buchholz, 2016) strain optimization was time-intensive and only applicable to organisms with short replication cycles. In the 1970-1980s improvements in DNA cloning technologies, the judicious use of restriction enzymes, and an increased understanding of protein function reduced the scale of protein engineering from the cellular to the macromolecular level (Mullis et al., 1994). Researchers were able to directly and specifically target genes encoding proteins leading to changes in function (Hughes, 2011). In 1993, Frances Arnold employed these advances to successfully use error-prone PCR to create large libraries of variants of subtilisin E with the goal of identifying mutants that could tolerate high concentrations of dimethylformamide (K. Chen et al., 1991; K. Chen & Arnold, 1993). From this point forward, the field of protein

engineering has experienced explosive growth. Its importance in medicine, agriculture, and industry was publicly acknowledged when Frances Arnold (prize share 1/3), was awarded the 2018 Nobel Prize in Chemistry for her work on engineering enzymes with novel functions (Garcia, 2018).

In the field of protein engineering, there are two principal approaches: perturbation-based strategies known as “directed evolution,” and design-centric strategies based on rational, knowledge-based modifications of protein structure (Clark & Pazdernik, 2008; Tiwari et al., 2012). Directed evolution, the experimental method Frances Arnold perfected, is an approach in which a target gene is randomly mutated and partnered with a screening or selection protocol that identifies variants possessing the desired new function (F. H. Arnold et al., 2001). Rational design, the second main approach, is a site-specific, structure-based method in which portions of the protein are targeted for mutagenesis (Song et al., 2023). Novel functionality is identified using screening or genetic selections. As these approaches are not mutually exclusive, many researchers rely on an iterative combination of the two to optimize the final product (Chica et al., 2005).

### **1.3 Computational Design of Proteins using the Rosetta Software Suite**

Despite the successes achieved by incorporating directed evolution approaches, large selections and screening procedures are cumbersome and expensive. To circumvent the pitfalls of this methodology, rational designers incorporate computational methods into the design process including the groundbreaking Rosetta Software suite for macromolecular modeling developed in the David Baker lab at the University of Washington (Bender et al., 2016; Richter et al., 2011). Rosetta is used to quickly test

hypotheses based on initial structural models that are labor intensive to perform experimentally. Algorithms for a variety of protein modeling design tasks have been developed within the Rosetta software, and the majority rely on a Monte Carlo simulated annealing search which is an iterative sampling of the sequence and/or conformational space of a protein (Rohl et al., 2004). The total energy of the modified protein is assessed using a physics and knowledge-based score function composed of 19 different weighted terms derived largely from information gleaned from the more than 200,000 crystal structures in the Protein Data Bank (PDB) (Alford et al., 2017a). Many of the score terms were developed using statistical occurrences of pair-wise interactions from crystal structures including electrostatics, the 6-12 Lennard-Jones potential for van der Waals forces, a geometric-centric hydrogen bonding potential, solvation effects, and amino acid side-chain conformation energetics (Dunbrack, Roland L., Jr. and Karplus, 1993; Kortemme et al., 2003; Lazaridis, 2003). To avoid a progress-halting descent into a local minimum on the conformational landscape, a Metropolis criterion is also applied that randomly accepts certain mutations that increase the energy (Kaufmann et al., 2010). The solutions proposed by Rosetta are predicted to be designs that will fold properly into functional, soluble proteins. However, experimental characterization of designed proteins is an integral component to a comprehensive research strategy.

Significant advances in protein engineering made possible by Rosetta include the design of proteins with the ability to bind metals and small molecules (Mills et al., 2013, 2016; Tinberg & Khare., 2017) and *de novo* design of proteins with folds that do not exist in nature (Huang et al., 2016) to name a few. Another important application for which Rosetta has been used is enzyme design (Richter et al., 2011). Novel enzymes with their

unparalleled ability to catalyze reactions under mild conditions with stereo-selectivity make them profoundly important to sustainable manufacturing processes, the creation of new small molecule therapeutics, and even toxic waste clean-up (Singh et al., 2016).

#### **1.4 Computational Design of Enzymes: Successes and Current Challenges**

Traditionally, computational design of enzymes follows the Linus Pauling view of enzymatic catalysis which is that the enzyme binds with higher affinity to the transition state (Pauling, 1946). Therefore, the design methodology is heavily weighted on the principle of transition state stabilization (Bolon & Mayo, 2001; Privett et al., 2012; Tantillo et al., 1998; Zhu & Lai, 2009). For modeling, a theoretical transition state structure, a “theozyme,” consisting of the catalytic residues and a bound substrate molecule, is generated *in silico*. Ideally, the theozyme includes activation-barrier-reducing interactions between the catalytic residues and the substrate. The next step is to search a library of structures to identify a protein backbone that can accommodate the theozyme. This scaffold set can consist of crystal structure models of thermostable proteins in the PDB or, more recently, can be a *de novo* protein backbone generated by machine learning techniques like RFDiffusion (Watson et al., 2023; Yeh et al., 2023). When the theozyme is modeled into the chosen scaffold, further structural modifications to stabilize the active site by buttressing the catalytic residues are undertaken. The goal of the final computational optimizations is to design a well-packed, well-organized catalytic pocket (Smith et al., 2008). The goal of this three-step procedure is a protein that will fold into a soluble enzyme with the ability to carry out the desired reaction. This protocol has been implemented to successfully design a Kemp eliminase, a retroaldolase, ester

hydrolase, a Diels-Alderase, and more recently a luciferase (Jiang et al., 2008; Richter et al., 2012; Röthlisberger et al., 2008; Siegel et al., 2010; Yeh et al., 2023).

Despite significant advances in successfully engineering proteins with novel function, computational design methods are plagued by enzymes with very low efficiencies when compared to naturally evolved enzymes (Mak & Siegel, 2014). Hypotheses regarding the observed discrepancy in catalytic efficiencies between naturally occurring and designed enzymes include suboptimal transition state modeling, suboptimal active site access to solvent, and the inability to model productive protein dynamics into the scaffold (Chovancova et al., 2012; Kiss et al., 2010; Leaver-Fay, Jacak, et al., 2011; Mak & Siegel, 2014; Radzika & Wolfenden, 1995). Work is ongoing to tease out what exactly current enzyme design methods are missing. However, the hypothesis of greatest interest to this work is that enzyme design is not currently able to recapitulate integral dynamic movements that are key to productive enzymatic function.

The relationship between enzyme catalysis and dynamics has been a fiercely debated topic for many years (Kamerlin & Warshel, 2010; Schneider et al., 2021; Schnell et al., 2004; Warshel et al., 2006; Warshel & Bora, 2016). Understanding the relationship between structure and dynamics in enzymes is fraught with challenges because even small changes in enzyme structure can lead to changes in dynamics (Schnell et al., 2004). However, new insights into enzyme function have given weight to the importance of dynamics in enzyme function (Baker, 2010; K. Henzler-Wildman & Kern, 2007; Korendovych, Ivan V., DeGrado, 2014; Markin et al., 2021; Vaissier Welborn & Head-Gordon, 2019; Wolf-Watz et al., 2004a).

## **1.5 Protein Dynamics and Their Role in Enzyme Function**

Early on, proteins were perceived as static entities with fixed structures. Initial models of enzymes followed in this tradition with work in 1894 by Emil Fischer on the specificity of enzymes for their substrates leading to the “lock and key” hypothesis. This model provided the first mental image of molecular recognition as a stiff interaction between well-organized sidechains in the active site and a ligand (Lichtenthaler, 1995). However, the realization that proteins possess intrinsic flexibility emerged with advancements in techniques like NMR spectroscopy, molecular simulations, and advanced imaging methods which allowed for the visualization of protein structures at higher resolutions, capturing subtle changes in atomic positions. As these technologies evolved, discrepancies between static models and experimental data emerged, leading to the recognition that proteins, and more specifically, enzymes, exist in an equilibrium of conformations. In 1958, Koshland’s “induced-fit” model (Koshland, 1958). suggested a more dynamic molecule with interactions that develop in real time as the ligand binds in the active site inducing a conformational change in the enzyme. This model is very similar to the conformational selection model that posits that an enzyme exists in a conformation of substates, and the substrate binds to one of these states thereby stabilizing it in a particular conformation. The difference between these two models is in the order of the conformational change upon ligand binding. The induced-fit model asserts that the ligand binding event predates the conformational change while the conformational selection model asserts that the ligand binding is the result of a conformational change. Regardless of the ordering, both models view enzymes as dynamic entities.

Current thought is that enzymatic activity requires synergy between flexibility and structural stability. Enzyme structure is not just a scaffold to correctly position critical catalytic residues in the active site, it also allows for sampling of functionally productive conformational states (Ramanathan et al., 2011; Ramanathan & Agarwal, 2011). To function well, enzymes must maintain their three-dimensional structure, but be flexible enough to properly bind the substrate, accommodate the transition state and any intermediates, and release the product (Bar-Even et al., 2015; Dellus-Gur et al., 2015). What is known is that these dynamical motions are not random events; they play a pivotal role in enabling enzymes to perform their diverse chemistries. It has also been noted that these dynamic fluctuations span over a range of timescales that depend on the complexity of the motion or the number of atoms involved. The range, from the simplest motion, the vibration of atoms, takes place on the femtosecond timescale; while the movement of side chain methyl groups takes place on the picosecond timescale; the movement of active site residues occurs on the picosecond-nanosecond timescale; and the largest domain movements occur on the microsecond-millisecond timescale (Ojeda-May et al., 2021; Schramm, 2011; Tugarinov et al., 2021; Zeymer et al., 2016).

The Dorothee Kern lab at Brandeis University discovered that the rate of large domain movements taking place on the  $\mu$ s-ms timescale, like the opening and shutting of the lid on adenylate kinase, is the catalytic rate-limiting step. Moreover, in adenylate kinase homologues, differences in their catalytic efficiencies were directly tied to their rates of lid opening and closing (Boehr et al., 2006; K. A. Henzler-Wildman et al., 2007; Wolf-Watz et al., 2004b). How motions on smaller timescales relate to these large



domain movements and in turn enzyme function and structural stability is the topic for ongoing research efforts.

## **1.6 Modulating Protein Dynamics Through Allostery**

In the 1960s, Monod and Jacob coined the term “allosteric” inhibition as a rebuttal to the widely held belief that only steric analogues of the substrate could inhibit enzyme activity. At this time, it was well-documented that many biosynthetic pathways involving a series of enzymes were regulated by the end-product. This allowed for exquisite control in the cell. At the time, it was believed that the enzyme active site was the only place where a substrate could bind and have an impact on the catalytic rate. Therefore, the end-product was regulating enzyme function by competing with the substrate for binding in the active site. Monod and Jacob disagreed. They argued using data on threonine deaminase from *Changeux* that competitive inhibition in this system was not due to a mutually exclusive binding event, but that there were two distinct binding sites on threonine deaminase. In addition, these two sites were interacting with each other (Monod & Jacob, 1978). Allosteric regulation, or the “communication” between locations distal to each other in geometric space, is an important concept to drug development and signaling. However, an allosteric interaction is not limited to a small molecule-protein binding event, it can be applied in a broader context. For the purposes of this work, the term allostery is defined as any perturbation in a protein that affects the function of a distant site. This broader definition encompasses site-specific perturbations like amino acid substitutions and protein-protein binding events (Ma et al., 1999; Paul & Weikl, 2016).

There is now ample evidence suggesting that changes to locations far from the active site can have a significant impact on the chemistry taking place in the active site (Amor et al., 2016; Boehr et al., 2006; Goldsmith & Tawfik, 2017; Guarnera & Berezovsky, 2016; K. A. Henzler-Wildman et al., 2007; Kamerlin & Warshel, 2010; Leferink et al., 2014; Morley & Kazlauskas, 2005; P. Singh et al., 2015; Warshel & Bora, 2016). Frances Arnold described this phenomenon in her 2018 Nobel lecture. “Mutations happened all over the protein: twenty to thirty angstroms away from the active site where no one could explain them, much less predict them” (F. Arnold, 2019). Unfortunately, Arnold was correct when she stated that a robust method to predict which distal mutations would have an effect on enzyme function does not exist. Currently, the only way to dial into mutations distal to the active site that can fine-tune dynamics and improve catalysis is by random mutagenesis.

## **1.6 Summary of the Work Presented**

The focus of this work was to explore the relationship between enzyme dynamics and substrate specificity. Particularly, elucidating how mutations distal to the active site can influence enzyme catalysis is the focus of chapters 2 and 3. The protein chosen for this work was TEM-1 from the Class A  $\beta$ -lactamase family.  $\beta$ -lactamases are a large family of enzymes that hydrolyze  $\beta$ -lactam antibiotics, conferring resistance to the bacteria producing them. TEM-1 efficiently hydrolyzes antibiotics from the penicillin family but has poor activity against cephalosporins because the added side chain bulk sterically clashes with the active site. In chapter 2, we computationally identified residues that are rigid and allosterically coupled to the active site and we altered the interaction patterns around these important residues. This led to drastic

changes in function and stability. To identify structural alterations that could explain these changes, we crystallized and solved the structures for five TEM-1 variants. Chapter 3 is a structural analysis of the models built with data from x-ray crystallographic experiments. In chapter 4, the focus shifts from allosteric contributes to substrate specificity to the active site. This chapter is a discussion on how to modulate substrate specificity by remodeling the active site of firefly luciferase to identify novel luciferase-luciferin pairs for bioluminescent imaging. Specifically, a high-throughput platform was developed to identify target residues for library development. This workflow is broadly applicable to any study of enzyme-ligand binding.

## CHAPTER 2

### WORK TOWARD MODULATING ENZYME FUNCTION THROUGH ALTERING THE FLEXIBILITY OF RIGID RESIDUES

This chapter is adapted from: “Kolbaba-Kartchner, B.; Kazan, I.C.; Mills, J.H.; Ozkan, S.B. (2021) The Role of Rigid Residues in Modulating TEM-1  $\beta$ -Lactamase Function and Thermostability. *Int. J. Mol. Sci.* 22, 2895.” I Can Kazan shared first-authorship with Bethany Kolbaba-Kartchner. I Can Kazan conducted all computational work related to DFI, DCI, and MD simulations presented while Bethany Kolbaba-Kartchner performed all RosettaDesign methods and experimental characterization.

#### **2.1 Abstract**

The relationship between protein motions (i.e., dynamics) and enzymatic function has begun to be explored in  $\beta$ -lactamases as a way to advance our understanding of these proteins. In a recent study, we analyzed the dynamic profiles of TEM-1 (a ubiquitous class A  $\beta$ -lactamase) and several ancestrally reconstructed homologues. A chief finding of this work was that rigid residues that were allosterically coupled to the active site appeared to have profound effects on enzyme function, even when separated from the active site by many angstroms. In the present work, our aim was to further explore the implications of protein dynamics on  $\beta$ -lactamase function by altering the dynamic profile of TEM-1 using computational protein design methods. The Rosetta software suite was used to mutate amino acids surrounding either rigid residues that are highly coupled to the active site or to flexible residues with no apparent communication with the active site. Experimental characterization of ten designed proteins indicated that alteration of residues surrounding rigid, highly coupled residues, substantially affected both enzymatic

activity and stability; in contrast, native-like activities and stabilities were maintained when flexible, uncoupled residues, were targeted. Our results provide additional insight into the structure-function relationship present in the TEM family of  $\beta$ -lactamases. Furthermore, the integration of computational protein design methods with analyses of protein dynamics represents a general approach that could be used to extend our understanding of the relationship between dynamics and function in other enzyme classes.

## **2.2 Introduction**

Since the 1940s,  $\beta$ -lactam antibiotics, which target a key enzyme in bacterial cell wall biosynthesis, have been the antimicrobial weapon of choice in the war against bacterial infection (Coulson, 1985). The widespread use of  $\beta$ -lactams is likely a consequence of the fact that they are inexpensive to produce and have historically been effective in treating most infections. However, as the use of this class of antibiotics became more widespread, so too did the prevalence of  $\beta$ -lactamase enzymes, which hydrolyze the  $\beta$ -lactam ring and render the antibiotic nonfunctional (Coulson, 1985). Additionally, as new  $\beta$ -lactam antibiotics enter into clinical use, the remarkable adaptivity of  $\beta$ -lactamases complicates efforts to develop novel antibiotics that are resistant to degradation by this class of enzyme (Bush, 2018). The TEM family of  $\beta$ -lactamases has been thoroughly studied to gain insight into the manner in which resistance is achieved (Brandt et al., 2017; Brown et al., 2020; Cortina et al., 2018; Cortina & Kasson, 2018; Gobeil et al., 2019). Despite these efforts, we currently possess an incomplete understanding of the relationship between sequence and function in this enzyme class. A major challenge is that several mutations have been identified that have a significant influence on function, but which

are highly distal from the enzyme active site (M. K. Singh & Dominy, 2012). In addition, even single point mutations (e.g., the well-characterized, M182T substitution), which have minimal effects on enzymatic function can drastically affect the protein's thermostability (Orencia et al., 2001; Wang et al., 2002). Our inability to rationalize the manner in which these thoroughly studied mutations alter enzyme function is suggestive of an incomplete understanding of the sequence-function relationships present in  $\beta$ -lactamases. This in turn limits our ability to develop novel classes of antibiotics that are not substrates for these enzymes (Fair & Tor, 2014).

A possible explanation as to how mutations distal to the active site can still exert influence at a great distance is that they serve to reshape the inherent dynamics of the enzyme (Doucet et al., 2007; Gerek et al., 2009; Gerek & Ozkan, 2011; Kim et al., 2015; Larrimore et al., 2017; Modi et al., 2018; Modi & Banu Ozkan, 2018; Zou et al., 2015). In a recent study, we explored this hypothesis in the TEM-1  $\beta$ -lactamase using two in silico, dynamics-based metrics: the dynamic flexibility index (dfi) (Gerek & Ozkan, 2011; Kumar, Butler, et al., 2015), which measures the mobility of each residue, and the dynamic coupling index (dci) (Campitelli et al., 2018; Larrimore et al., 2017), which assesses the coupling between distant residues (Zou et al., 2015). Using these two metrics, we characterized TEM-1 and a set of ancestrally reconstructed TEM-1 variants that possess vastly distinct physical properties (i.e., thermostabilities) and functions (i.e., substrate specificity) despite having almost identical conformations (Risso et al., 2013; Salverda et al., 2010; Stiffler et al., 2015; Zou et al., 2015). A major finding of our previous study was that TEM-1 and its ancestral homologues possessed distinct dynamic profiles and that these differences in dynamics appeared to have profound effects on

enzyme function. Namely, rigid residues that are distal from, but highly coupled to, residues in the active site appeared to have substantial effects on protein function (Campitelli et al., 2018; Campitelli, Modi, et al., 2020; Modi & Banu Ozkan, 2018; Stiffler et al., 2015). One intriguing hypothesis that might explain these data is that rigid residues can serve as “hubs” of dynamic communication. This notion has also been validated in the context of disease-causing mutations in other proteins, (Modi et al., 2021) in which mutations to rigid residues that are far from the active site are functionally deleterious (Campitelli, Modi, et al., 2020; Kumar, Glembo, et al., 2015; Modi & Banu Ozkan, 2018; Nevin Gerek et al., 2013).

More recently, we used both dfi and dci to analyze members of the TEM family that either arose in the clinic or were generated via directed evolution (Modi & Banu Ozkan, 2018). In this study, we observed that mutations known to confer resistance to non-native substrates 1) often occur at particularly rigid residues as judged by our dfi metric and 2) appear to allosterically modify the flexibility of catalytic residues within the active site as suggested by our dci metric (Modi & Banu Ozkan, 2018). Collectively, these studies support the hypothesis that rigid residues are of particular importance to the overall dynamics of proteins and may have a substantial impact on protein function if they are allosterically coupled to the active site. If our hypothesis is correct, mutations that alter the identity of allosteric rigid residues (or those in their vicinity) could have substantial effects on enzyme activity; however, the ability to thoroughly explore this hypothesis is challenging. Although extensive datasets comprised of clinically derived TEM family variants (NCBI BioProject Database, 2018, February 1) and additional variants generated via directed evolution (Stiffler et al., 2015) exist, the serendipitous identification of

proteins with multiple mutations in the vicinity of known rigid residues would be unlikely. One potential solution is to use computational protein design methods to specifically target mutations to regions of interest. A major benefit of this approach is the ability to “pre-screen” each combination of mutations *in silico* to exclude variants in which protein folding is not predicted to be energetically favorable.

In this work, computational protein design methods were used to alter the environments surrounding two residues that were identified as being rigid and highly coupled to the active site despite being separated from it by a great distance. Dynamic profiles of each designed protein (hereafter referred to as a “design”) were then generated and compared to that of an ancestrally reconstructed variant of TEM-1 (the “Gram-negative common ancestor” or GNCA), which possesses increased thermostability, but reduced activity against ampicillin relative to wild type TEM-1 (Risso et al., 2013). Principal component analysis (PCA) was used to identify designs with dynamic profiles that were predicted to be more similar to GNCA than extant TEM-1, and five designs were characterized in the laboratory. All designs exhibited reduced activity against ampicillin relative to TEM-1, but an increase in thermostability was also observed. Reduced activity against ampicillin and increased thermostability relative to TEM-1 are both features of GNCA. Alternatively, when identical design protocols were applied to flexible residues, that were not coupled to the active site, native-like catalytic abilities and thermostabilities were maintained. Finally, in an effort to further link dynamics to enzyme function, we developed a novel analytical approach termed the “dynamic distance analysis” (dda) that was applied retrospectively to our experimentally characterized proteins. The dda analysis appeared to capture functional differences



between our designed proteins and could be a useful tool for dynamic profile analysis in future studies. Collectively, our results serve to further highlight the importance of allosteric rigid residues in regulating the dynamics of the TEM-1  $\beta$ -lactamase.

## 2.3 Materials and Methods

**2.3.1 Molecular Dynamics (MD).** The AMBER software package was utilized for simulating all  $\beta$ -lactamases in this study. Each system was parameterized with the ff14SB force field and the explicit water model TIP3P (Campitelli, Ozkan, et al., 2020; Salomon-Ferrer et al., 2013). The solvation box was assigned as 16 Å. The system was neutralized by sodium and chloride ions and minimized for 11,000 steps using the steepest descent algorithm. Isothermal, isobaric, and constant number of particles ensemble production trajectories were performed at 300K and 1 bar pressure. For each production, a 1  $\mu$ s simulation was conducted. The residue covariances were calculated using a 50 ns length window shifted by 10 ns (example: 1–50 ns, 10–60 ns, etc..) over the course of the trajectories.

**2.3.2 Dynamic Flexibility Index (dfi).** The dfi metric (Gerek & Ozkan, 2011; Kumar, Butler, et al., 2015; Modi & Banu Ozkan, 2018) calculates the relative flexibility/rigidity of a residue in a protein by incorporating the residue covariances. The protein can be modeled with the Elastic Network Model (ENM) in which harmonic springs connect C $\alpha$ s (Atilgan et al., 2010). Taking the second derivatives of the potential forms a Hessian matrix,  $H$  Equation (1). The inverse of the Hessian matrix is proportional to the covariance matrix. The models based on ENM cannot capture changes in the dynamics of the designed variants based on C $\alpha$  positions alone. Therefore, we substituted the inverse of the Hessian with the covariance matrices from MD trajectories to capture

the effect of mutations on the protein conformations. The covariance matrix,  $G$ , contains the residue covariances, obtained by the MD trajectories Equation (2) and (3) (Bishop, 2006; Kumar, Glembo, et al., 2015; Larrimore et al., 2017; Modi & Banu Ozkan, 2018; Nevin Gerek et al., 2013).

$$[\Delta\mathbf{R}]_{3N \times 1} = [\mathbf{H}]_{3N \times 3N}^{-1} [\mathbf{F}]_{3N \times 1} \quad (1)$$

$$[\Delta\mathbf{R}]_{3N \times 1} = [\mathbf{G}]_{3N \times 3N} [\mathbf{F}]_{3N \times 1} \quad (2)$$

$$dfi_i = \frac{\sum_{j=1}^N |\Delta R^j|_i}{\sum_{i=1}^N \sum_{j=1}^N |\Delta R^j|_i} \quad (3)$$

The residue response vector ( $\Delta\mathbf{R}$ ) is the resultant vector containing the fluctuation responses from multiplying the covariance matrix with the force vector,  $\mathbf{F}$ .  $|\Delta\mathbf{R}^j|_i$  denotes the magnitude of the residue response fluctuation vector of position  $i$ , when  $j$  is exposed to a random force vector.

**2.3.3 Dynamic Coupling Index (dci).** The dynamic coupling index (dci) (Bishop, 2006; Larrimore et al., 2017; Modi & Banu Ozkan, 2018) measures the degree of dynamic coupling between two residues. Namely, it captures the strength of displacement of a residue  $i$  upon perturbation of a distinct residue  $j$ , relative to the average fluctuation response of position  $i$  when all the positions within a structure are perturbed. Generally, this metric is used to establish the communication between a functionally important residue and other residues within the protein that are many angstroms away. The dynamic coupling index of a given residue  $i$  is calculated using the equation below Equation (4):

$$dci_i = \frac{\sum_j^{N_{Functional}} |\Delta R^j|_i / N_{Functional}}{\sum_{j=1}^N |\Delta R^j|_i / N} \quad (4)$$

where  $|\Delta R^j|_i$  corresponds to the magnitude of the residue response vector ( $\Delta R$ ) for residue  $i$  when residue  $j$  is perturbed. The dci score thus provides information on the allosteric behavior of a location associated with active site dynamics. A high dci value implies strong coupling between active sites, inversely, a low scoring position is regarded as weakly coupled to the active site (Bishop, 2006; Larrimore et al., 2017; Modi & Banu Ozkan, 2018).

**2.3.4 Dynamic Distance Calculation.** Principal Component Analysis (PCA) was used to compare and cluster the flexibility profiles of the designed TEM-1 variants with respect to TEM-1 and GNCA. However, because the output of a PCA is dependent on the input data, the calculated distances between any designed protein and TEM-1 or GNCA can change with the inclusion of new or distinct data points (e.g., a different set of designed proteins). To account for this, we employed an iterative, random sampling approach to capture the relative distance of a designed protein from TEM-1 and from GNCA (Figure 2.3.4.1).

For every designed TEM-1 variant, a dataset containing the target design, TEM-1, GNCA and an additional seven randomly chosen designs was constructed and used to generate a PCA. Namely, the dfi profiles of these ten proteins were merged into a matrix,  $X$ , of dimension Equation (5)

$$(m \times n) \quad (5)$$

Here,  $m$  is the total number of datasets that are clustered together, which each have  $n$  number of attributes ( $n = \text{total number of residues}$ ). Singular value decomposition of  $X$  was then carried out as follows Equation (6):

$$[X]_{m \times n} = [U]_{m \times m} [\Sigma]_{m \times n} [V]_{n \times n} \quad (6)$$

Here,  $U$  and  $V$  are unitary matrices with orthonormal columns and are called left singular vectors and right singular vectors, respectively, and  $\Sigma$  is a diagonal matrix with diagonal elements known as singular values of  $X$ .

The singular values of  $X$ , by convention, were arranged in a decreasing order of their magnitude,  $\sigma = \{\sigma_i\}$  representing the variances in the corresponding left and right singular vectors. The set of the highest singular values (representing the largest variance in the orthonormal singular vectors) can be interpreted to show the characteristics in the data  $X$  and the right singular vectors create orthonormal basis which spans the vector space representing the data. The left singular vectors contain weights indicating the significance of each attribute in the dataset as Equation (7):

$$w_i = \sum_{k=1}^r \sigma_k |u_{ik}| \quad (7)$$

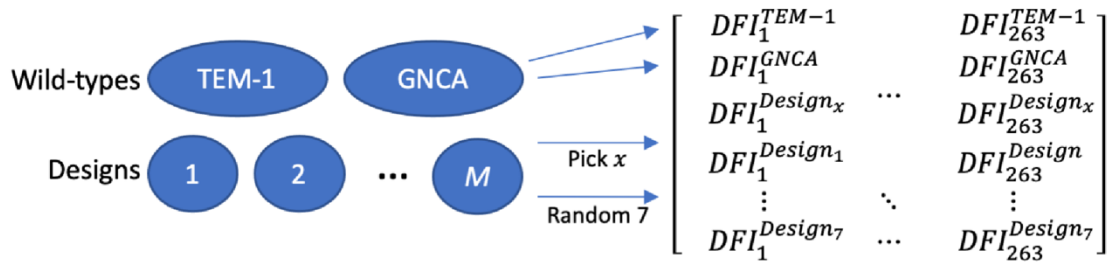
Using these features of the decomposed singular vectors, we created another matrix,  $X^*$  using only the highest three singular values which mimics the basic characteristics of the original dataset. It can be represented as Equation (8):

$$[X^*]_{m \times r} = [V^*]_{m \times r} [\Sigma^*]_{r \times r} \quad (8)$$

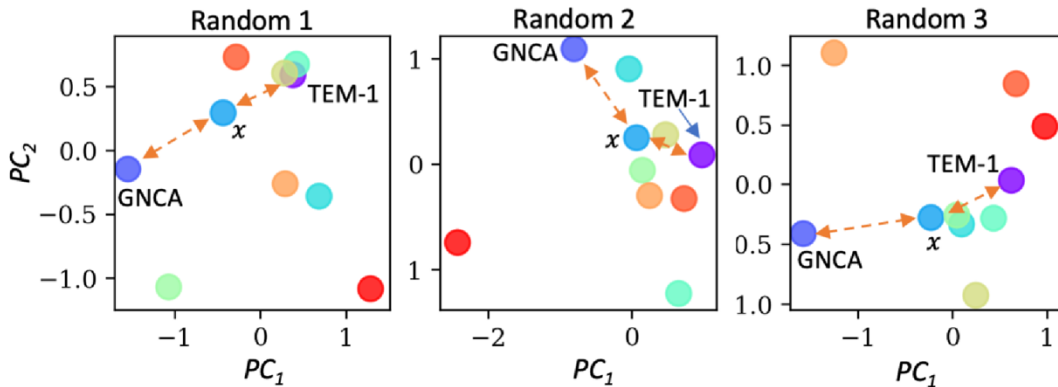
Here,  $\Sigma^*$  contains only the largest 3 singular values and  $V^*$  contains the corresponding right singular vectors. The data were then clustered hierarchically based on the pairwise distance between different proteins in the reconstructed dfi data with reduced dimensions. The distance between designed protein,  $j_1$ , and TEM-1,  $j_2$ , was computed in the reduced dimension using three principal components Equation (9):

$$d_{12} = \sqrt{\sum_{i=1}^3 (X_i^{*j_1} - X_i^{*j_2})^2} \quad (9)$$

We also calculated the distance between each designed TEM-1 variant and GNCA to measure the similarity in their flexibility profiles. The random selection of dataset was repeated a thousand times to create a diverse distance distribution and we called this distance profile analysis dynamic distance analysis (dda). The distributions were fit to a Gaussian mixture model with a Dirichlet prior to estimate the density and the mean of the dynamic distances (Gibson et al., 2009). The distributions and the mean distances were utilized for selecting the designed proteins that cluster close to GNCA and far from TEM-1 (Figure 2.3.4.1).



Calculate the dynamic distance of design  $x$  to TEM-1 and GNCA



**Figure 2.3.4.1.** Schematic of the dynamic distance calculation process. The dynamic profile of each design (using the dfi metric) is clustered using PCA in a set composed of TEM-1, GNCA, and seven randomly chosen designs. The dynamic distance of the design from TEM-1 and GNCA is calculated. Notably, the dynamic distance of the designed protein from TEM-1 and GNCA varies according to the set of proteins incorporated. To capture a statistically accurate distribution, this procedure is iterated a thousand times, each time varying the set of designed proteins.

**2.3.5 Rosetta Design Protocol.** A high-resolution (1.8 Å) structure of TEM-1 (PDB ID: 1btl) was processed to remove waters and non-proteinogenic molecules. The resulting structure was subjected to an energy minimization using the Rosetta relax protocol (Conway et al., 2014a; Nivón et al., 2013). (Detailed descriptions of all computational protocols used in this study can be found in Appendix B). The relaxed 1btl structure was used as an input to the DesignAround protocol within Rosetta using the ref15 score function (Alford et al., 2017a). This algorithm first identifies spheres with user-defined

radii around a defined residue. Residues within these “design spheres” were subjected to *in silico* mutagenesis, conformational sampling and backbone minimization.

**2.3.6 Protein Expression and Purification.** A pET24b plasmid encoding the gene for GNCA was a generous gift from Professor Jose Sanchez-Ruiz (Universidad de Granada). Genes encoding rigid design variants were codon-optimized for expression in *Escherichia coli* cells. The native TEM-1 N-terminal periplasmic localization signal peptide (MSIQHFRVALIPFFAAFCLPVFA) was appended to the beginning of each gene. To facilitate purification, a C-terminal 6xHis affinity tag was added to the end of each gene. Genes encoding each rigid design were synthesized by IDT (Coralville, IA). The gene for wildtype TEM-1 was amplified from a pET21b vector using PCR. Genes encoding the rigid designs and TEM-1 were subcloned into the pET29b vector using the Gibson Assembly (Gibson et al., 2009) at a site that placed them under the control of the T7lac promoter. Genes encoding the uncoupled flexible residue variants were synthesized and cloned into pET29b vectors by GenScript (Piscataway, NJ).

The sequences of all plasmids containing TEM-1, GNCA, rigid or flexible designs were confirmed by Sanger sequencing and were transformed via electroporation into BL21 Star (DE3) *E. coli* cells. Cells containing plasmids encoding GNCA were grown in lysogeny broth (LB) at 37 °C with shaking at 250 rpm until an O.D.<sub>600</sub> of ~0.8 was reached. Isopropyl β-D-1-thiogalactopyranoside (IPTG) was then added to a final concentration of 1 mM to induce expression; cells were grown for 3 h post induction. Cells containing plasmids encoding TEM-1 were grown in LB media at 20 °C with shaking at 220 rpm until an O.D.<sub>600</sub> of ~0.8 was reached. Induction was again carried out with 1 mM IPTG and was allowed to proceed for 8–12 h. Cells containing plasmids

encoding the rigid and flexible design variants were grown in 2xYT media to confluence overnight, and pelleted by centrifugation. After resuspension in fresh 2xYT media, protein expression was induced with 1 mM IPTG and cells were grown for an additional 20 h at 20 °C with shaking at 220 rpm.

After expression, the cells were pelleted via centrifugation at 4,100x g for 15 min and the media was discarded. The cells were resuspended in TBS (50 mM Tris pH 8.0, 500 mM NaCl) and were again centrifuged at 4,100x g for 15 min; the supernatant was discarded. The pellet was incubated at room temperature for 15 min with SET buffer (20% sucrose, 1 mM ethylenediaminetetraacetic acid (EDTA), 30 mM Tris pH 8.0, 1 mg/mL lysozyme). After centrifugation at 4,100x g for 15 min, the supernatant was decanted and saved. The cells were then shocked to release the periplasmic contents with ice cold 100 mM MgCl<sub>2</sub> at a 1:15 ratio of cell pellet weight to solution volume. Cells were vigorously agitated on ice for 15-30 min then centrifuged with the saved soluble fraction from the first stage at 4 °C for 60 min at 12,000x g.

The supernatant was then loaded onto a 5 mL nitrilotriacetic acid agarose (Ni-NTA) (Millipore Sigma, Burlington, MA) column, washed with 5 column volumes of a low imidazole buffer (25 mM Tris pH 8.0, 150 mM NaCl, 15 mM imidazole), and eluted with a high imidazole buffer (25 mM Tris pH 8.0, 150 mM NaCl, 500 mM imidazole). All proteins were then subjected to a second purification step using anion exchange chromatography: Proteins were concentrated to a volume of 0.5–1 mL, diluted into the loading buffer (50 mM Tris, pH 9.0, 50 mM NaCl) and loaded directly onto the 5 mL Hi Trap Q Fast Flow column (Millipore Sigma, Burlington, MA). The column was washed



with 5 column volumes of the loading buffer and eluted with 50 mM Tris, pH 9.0 250 mM NaCl. Protein purity was verified by SDS-PAGE (Figure 2.4.4.1).

**2.3.7 Circular Dichroism Characterization of Protein Folding and Stability.** Far-ultraviolet circular dichroism (CD) measurements were performed in triplicate on a Jasco J-815 spectrophotometer (Jasco, Inc, Easton, MD) equipped with a Peltier temperature controller. Wavelength scans were measured from 300 to 180 nm at room temperature with 1 nm steps using a 1 nm bandwidth, 5 nm/min scan rate; reported data represent an average of three independent scans. Thermal melts were monitored by the absorption signal at 222 nm with a temperature slope of 5 °C/min. For wavelength scans and thermal melts, the purified protein was in a TBS buffer (10mM Tris 50 mM NaCl, pH 7.0) in a cuvette with a 1 mm pathlength. Protein concentrations were calculated in triplicate using the absorbance at 280 nm and absorption coefficients as calculated by the ProtParam tool in the ExPASy software suite (Gasteiger E., Hoogland C., Gattiker A., Duvaud S., Wilkins M.R., Appel R.D., 2005). Protein concentrations ranged between 0.18–0.25 mg/mL for all scans. Thermal melt curves were fitted using nonlinear regression least squares fit with the Hill equation in the GraphPad Prism version 9.0.0 for Windows, GraphPad software, San Diego, California, USA.

**2.3.8 Minimal Inhibitory Assays.** Minimal inhibitory concentrations of ampicillin ( $MIC_{amp}$ ) were performed in triplicate on 96-well plates (Wiegand et al., 2008). For each designed protein, TEM-1 and GNCA, five colonies were picked from a fresh agar plate and used to inoculate a 5 mL culture of LB, which was grown to confluence overnight at 37 °C. Overnight cultures were diluted in LB with 1 mM IPTG to a final working concentration of  $5 \times 10^5$  cfu/mL. Three stock solutions of ampicillin were independently

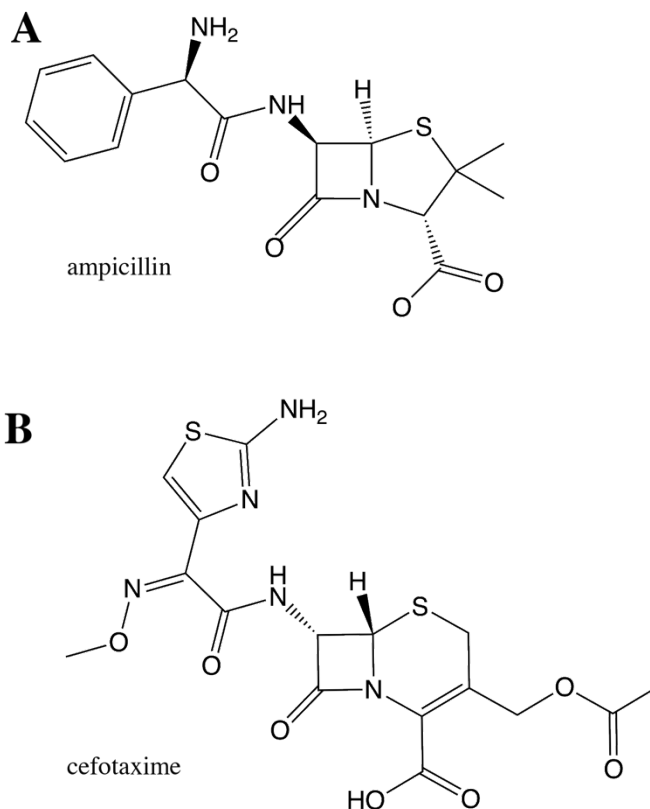
prepared at 6000  $\mu\text{g}/\text{mL}$  in LB with 1 mM IPTG and each solution was subsequently diluted in steps of 0.5 through the addition of LB with 1 mM IPTG to yield a final range of concentrations of 6-3000  $\mu\text{g}/\text{mL}$ . The ampicillin concentrations for GNCA and the rigid designs were prepared at 400  $\mu\text{g}/\text{mL}$  in LB with 1 mM IPTG and each solution was diluted in steps of 0.6 for a final concentration range of 2-200  $\mu\text{g}/\text{mL}$ . The 96-well plates were covered with a fitted lid and incubated at 37 °C for 20 h. All optical density measurements were carried out at 600 nm using a SpectraMax M5 (Molecular Devices, LLC, San Jose, CA); the absorbance of the buffer was subtracted from each measurement. To establish the lowest concentration of antibiotic that inhibited growth, a buffer-subtracted value  $> 0.1$  was used as the threshold for bacterial growth in each well. The MIC<sub>amp</sub> was determined to be the lowest concentration of ampicillin that inhibited growth of the *E. coli* cells.

## **2.4 Results and Discussion**

**2.4.1 Computational Analysis Using dfi and dci.** Our efforts to better understand the relationship between protein dynamics and function began by identifying a TEM-1 variant that could serve as a basis of comparison to the wild type protein. Recently, the putative sequences of ancestral TEM-1 were predicted using Bayesian bioinformatics (Risso et al., 2013). Three ancestral TEM family homologues (the Gram-negative and Gram-positive common ancestor, PNCA; the Gram-negative common ancestor, GNCA, and enterobacteria common ancestor, ENCA) were observed to possess distinct physical and biochemical properties when characterized in the laboratory (Risso et al., 2013). This is likely a consequence of the fact that these proteins are thought to have existed at different times in the evolutionary history of this enzyme (Risso et al.,



Furthermore, GNCA appears to be a “substrate generalist” in that it possesses measurable (but reduced) activity against penam antibiotics (e.g., penicillin and ampicillin) relative to TEM-1, while simultaneously possessing a far greater ability to degrade the bulkier cepham antibiotics (e.g., cefotaxime) (Figure 2.4.1.2) (Risso et al., 2013).



**Figure 2.4.1.2** Chemical structure of ampicillin and cefotaxime. (A) Ampicillin, a member of the penam, or penicillin family antibiotics. (B) Cefotaxime, a member of the cepham, or the third generation cephalosporin family antibiotics.

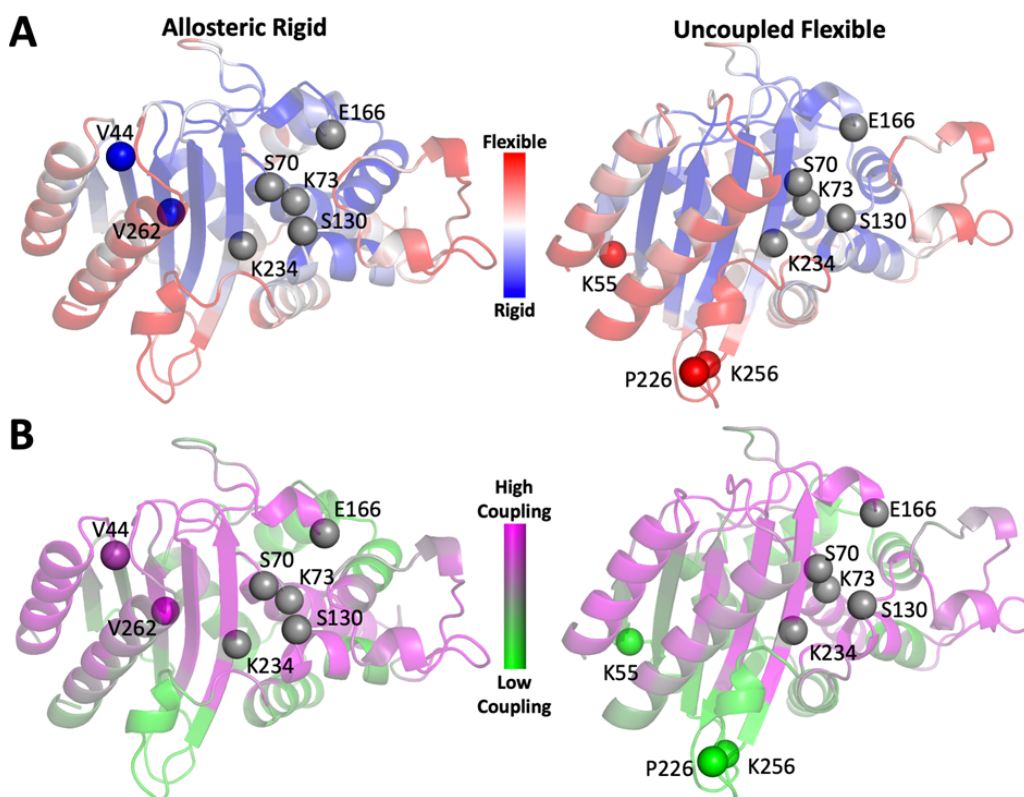
It is difficult to rationalize the substantial differences in function and stabilities that are observed in GNCA and TEM-1 in light of the high sequence identity and structural similarities that exist for these proteins. Previous studies in our laboratory (Modi & Banu Ozkan, 2018; Zou et al., 2015) suggested that the inherent dynamics of both TEM-1 and GNCA might play a role in regulating their functions. To further explore this, we

analyzed the dynamic profiles of both proteins using two metrics developed in our group: The Dynamic Flexibility Index (dfi) and the Dynamic Coupling Index (dci). The dfi method (Butler et al., 2015; Kumar, Glembo, et al., 2015; Nevin Gerek et al., 2013) is based on Linear Response Theory and Perturbation Response Scanning (Atilgan et al., 2010) and calculates the resilience of a given residue to random force perturbations applied to other residues in the protein. A given amino acid's dfi value is therefore related to the relative conformational entropy (i.e., flexibility) of that residue with respect to the rest of the protein. A residue with a high dfi value indicates high flexibility; conversely, a low dfi value indicates rigidity. The dci metric (Larrimore et al., 2017; Modi & Banu Ozkan, 2018) is derived from the same theoretical origin as dfi and is used to quantify the degree to which two residues are dynamically coupled in terms of correlated motions. A high dci value between a pair of residues that do not interact directly indicates allosteric coupling and suggests that a perturbation to one residue will be transmitted to the other even over long distances. A low dci score implies a weak coupling between a residue pair, and no strong communication channel between them is expected.

When we applied the dfi and dci analyses to extant TEM-1 and a set of reconstructed ancestral homologues including GNCA (Modi & Banu Ozkan, 2018; Zou et al., 2015), our analyses indicated that rigid residues (i.e., those with low dfi scores) that are highly coupled to the active site can contribute substantially to protein function. In this study, we hoped to further explore the importance of rigid residues to protein function by altering the identity of amino acids in their vicinity.

We selected two residues in TEM-1 (V44 and V262) as targets for our study. Not only do both residues have low dfi scores (%dfi value < 0.2) (Figure 2.4.1.3.A), but they

are highly coupled to the active site ( $\%dci > 0.7$ ) (Figure 2.4.1.3.B). These two residues were of particular interest to us because they are over 10 Å away from the active site and are located on adjacent  $\beta$ -strands with side chains facing opposite domains. We also identified three distal, flexible residues in TEM-1 (K55, P226, and K256) with high dfi scores ( $\%dfi > 0.8$ ) (Figure 2.4.1.3.A) and low coupling to active site residues as evaluated by the dci metric ( $\%dci < 0.4$ ) (Figure 2.4.1.3.B) and over 10 Å away from the active site to serve as controls. Alteration of the protein environments surrounding allosteric rigid residues would be expected to substantially modify protein function if our hypothesis is correct. Alternatively, modification of amino acids surrounding flexible residues with low dynamic coupling to the active site would be expected to result in proteins with native-like functions. All the allosteric rigid and uncoupled flexible residues we targeted for design are over 10 Å from the nearest catalytic residue, which suggests that mutations in their vicinities should only have an indirect effect on the active site unless other factors (e.g., dynamic coupling) are at play.



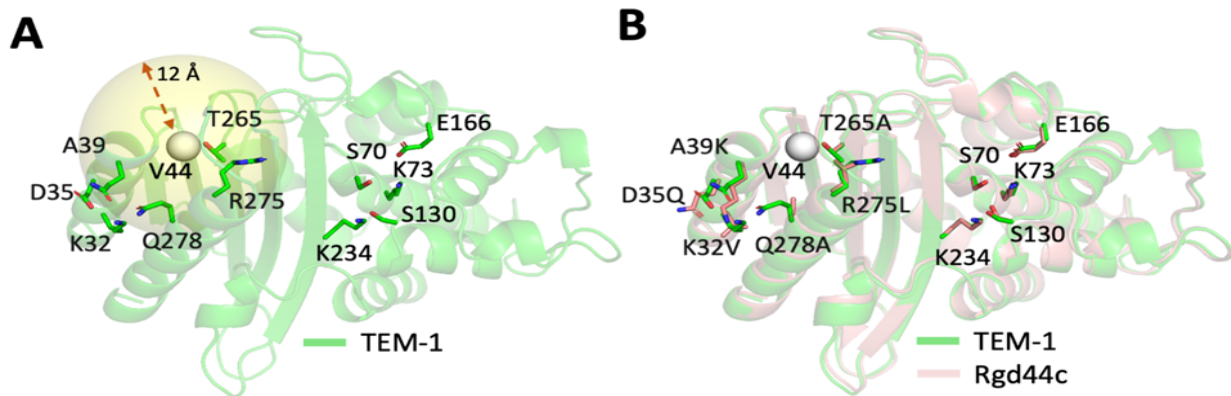
**Figure 2.4.1.3.** The dfi (A) and dci (B) values of each residue in TEM-1 are calculated and mapped onto the structure of TEM-1, which is shown as color coded cartoons. Catalytic residues are shown as grey spheres. Rigid and flexible residues used in this study are shown as spheres that are colored by either their dfi (A) or dci (B) score. Allosteric rigid residues, V44 and V262, have low dfi scores and high allosteric dynamic coupling with the active site residues. Residues K55, P226, and K256 are both highly flexible and exhibit low allosteric dynamic coupling to the active site.

**2.4.2 Computational Design of TEM-1 Variants.** To alter the amino acid compositions surrounding both the rigid and flexible residue positions, we used the Rosetta computational protein design suite (Leaver-Fay, Tyka, et al., 2011). The Rosetta software employs a Monte Carlo sampling protocol to randomize the identity and conformation (rotamer) of a randomly chosen residue; the fitness of the mutated protein is then assessed using the Rosetta energy function (Alford et al., 2017). In the course of a

single design trajectory, the Monte Carlo sampling algorithm is applied iteratively to a set of user-defined residues (see below).

We sought to develop a computational protocol within Rosetta that would substantially alter the chemical properties of the native amino acids without negatively affecting the protein's ability to fold. To do this, the RosettaDesign algorithm (Kuhlman et al., 2003) was used to randomly mutate residues within “design spheres” that had radii from 8-12 Å surrounding each of the target residues (Figure 3A). Slight alterations to the conformation of the peptide backbone were allowed only for residues that fell within the design sphere. A second shell was also defined that extended 4 Å beyond the inner design sphere. Residues in this shell were precluded from mutating but were energetically minimized in the context of adjacent, mutated residues. Independent design trajectories were carried out for all rigid and flexible residues. The two rigid (V44 and V262) and three flexible (K55, P226 and K256) residues that served as targets for our studies were also prohibited from mutating during the design calculations (Figure 2.4.2.1). Finally, catalytic residues (S70, K73, S130, E166, K234) were also maintained as their native identities and conformations during the design process. The designed proteins contained between two and eleven mutations with an average of seven mutations per protein. Ultimately, 64 unique designed proteins were generated using this approach.

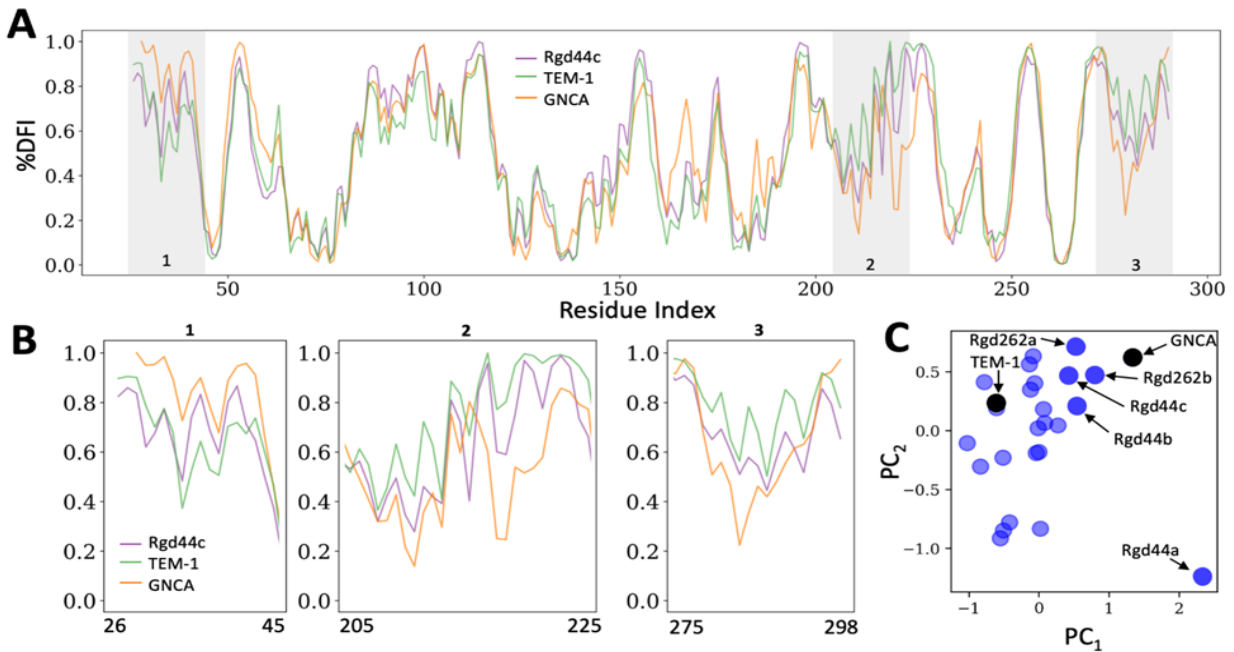




**Figure 2.4.2.1** Our general computational protein design strategy is shown schematically using the designed protein Rgd44c as an example. (A) Residues within an 8–12 Å sphere surrounding a given residue (V44 in this example) are considered as candidates for mutation. (B) A combination of mutations surrounding the target residue are generated using the RosettaDesign algorithm and scored using the Rosetta energy function. An overlay of the Rgd44c design model with TEM-1 (B) indicates that this design protocol creates a diversity of mutations within the design sphere while leaving active site residues untouched. The target rigid residue (V44) is shown as a white sphere in both panels. Both catalytic and designed residues are shown as sticks.

**2.4.3 Selection of the Designed Proteins Using Flexibility Profiles.** To assess how the computationally designed mutations affected TEM-1 dynamics, we subjected all designed proteins to a 1  $\mu$ s molecular dynamics (MD) simulation followed by analysis using the dfi metric (Figure 2.4.3.1.A). In order to rapidly compare the dfi profiles of our designed proteins to those of TEM-1 and GNCA, we used a 2D principal component analysis (PCA). The PCAs both simplified our data and allowed for the facile visualization of relationships between the calculated dynamic profiles of the designed proteins (Figure 2.4.3.1.B). PCAs generated from our rigid designs showed a diverse distribution in both the first and second principal components (Figure 2.4.3.1.C). On the PCA, several designed proteins were positioned relatively closer to GNCA in both components. We chose a subset of five such designs in which the allosteric rigid residues

had been targeted (henceforth referred to as “rigid designs”) for experimental characterization (Figure 2.4.3.1.C).

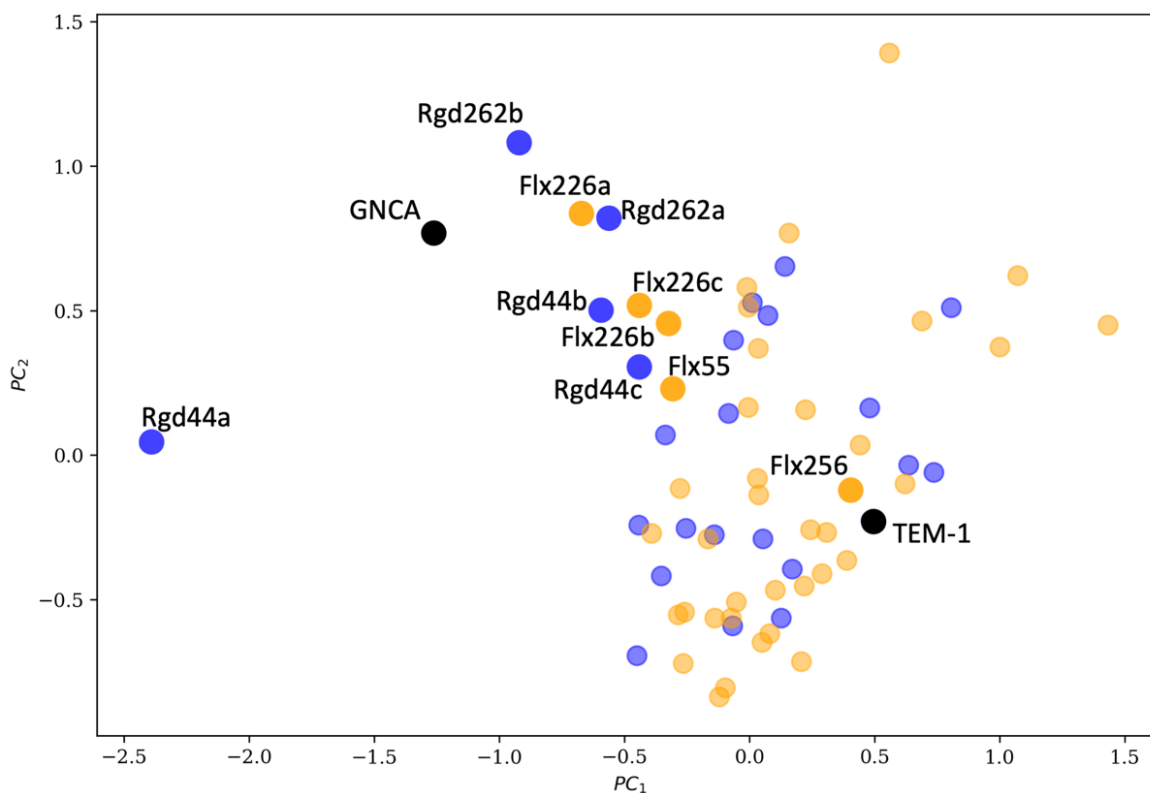


**Figure 2.4.3.1** Dynamic analyses of TEM-1, GNCA, and the rigid design, Rgd44c. (A) Depiction of the dfi profile of TEM-1 (green), GNCA (orange) and variant Rgd44c (purple); Rgd44c is chosen as an example for illustrative purposes. (B) Portions of the full dfi profile of each protein (A) are expanded to highlight dynamic differences between the three proteins. A shift towards a GNCA-like dfi profile is an indication of a change in dynamical characteristics of a protein. (C) Principal Component Analysis (PCA) of the rigid designs. The first (x-axis) and second (y-axis) principal components have weights of 3.5 and 2.7, respectively. Designs chosen for experimental characterization are highlighted using darker colors and labeled with the design name.

Four of the five rigid designs (Rdg44b, Rdg44c, Rdg262a, and Rdg262b, where the number in each name corresponds to the rigid residue that was targeted in the design calculations) clustered slightly away from TEM-1 and towards GNCA on both axes of the PCA; alternatively, Rdg44a, clustered near GNCA on the first principal axis but appeared as an outlier on the second axis. We hoped that experimental characterization of Rdg44a might help elucidate the parameters captured in each of the two principal components. It should be mentioned that only four among the five rigid designs that were

chosen for characterization had Rosetta scores that were lower (lower Rosetta scores imply lower energies) than TEM-1. The Rosetta score of Rdg262a was higher than TEM-1, but we selected this design for experimental characterization due to the fact that it clustered near GNCA in both axes of the PCA.

To analyze the designed proteins in which flexible, uncoupled residues were targeted (henceforth referred to as “flexible designs”), we generated a PCA in which all flexible design candidates were compared to TEM-1, GNCA and all the rigid designs including those that were not selected for characterization (Figure 2.4.3.2). Although a wide distribution of flexible designs was observed in this PCA, many of them clustered near TEM-1; a smaller subset clustered near the rigid designs we previously selected for characterization. To avoid biases that might have arisen if we chose only flexible designs that clustered with TEM-1 for analysis, we opted to experimentally characterize four flexible designs (Flx226a, Flx226b, Flx226c and Flx55) that clustered near the rigid designs chosen for experimental characterization and only one (Flx256) that clustered near TEM-1 (Figure 2.4.3.2).



**Figure 2.4.3.2** PCA of a selection of the flexible and rigid designed proteins. The rigid designs with allosteric dynamic coupling to the active site are marked with blue dots. Uncoupled flexible designs are marked with orange dots. TEM-1 and GNCA are shown as black dots. For both rigid and flexible designs, the variants chosen for experimental characterization are named and highlighted with darker colors.

Although clustering in similar locations in the PCA would suggest that the two proteins should have similar properties, it is difficult to infer what feature is represented on each axis of the PCA. We hoped that the diverse selection of proteins chosen for characterization would therefore provide information regarding whether rigid residues serve as hubs of dynamic control and also whether or not the PCA is a useful metric for discriminating between proteins with different activity and thermostabilities.

**2.4.4 Experimental Analysis of the Designed Proteins.** As GNCA and TEM-1 differ substantially with respect to thermostability (90.3 °C and 56.4 °C, respectively) and

activity against penam  $\beta$ -lactam antibiotics (GNCA is  $\sim 2$  orders of magnitude less efficient at degrading ampicillin than TEM-1), we chose to focus our analyses of the designed proteins on these characteristics. To do this, genes encoding each of the selected rigid and flexible designs were first cloned into the pET29b expression plasmid. Sequenced confirmed plasmids were transformed into a BL21 Star (DE3) *E. coli* expression strain in preparation for further analyses.

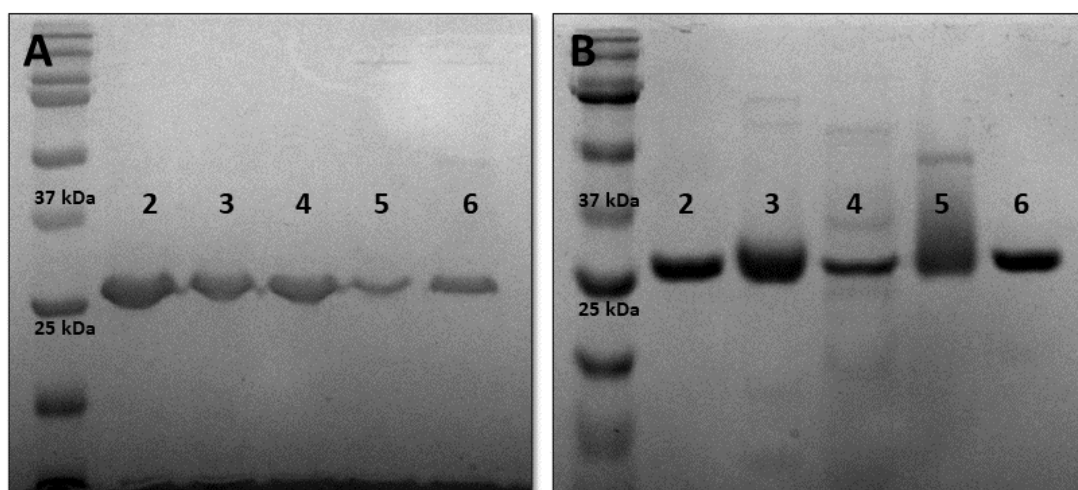
We assessed the resistance of our designed proteins to penam  $\beta$ -lactams by establishing the minimal inhibitory concentration of ampicillin ( $MIC_{amp}$ ) for each of our designed proteins using the protocol of Wiegand et al. (Wiegand et al., 2008). (See Materials and Methods for detailed protocols). Briefly, BL21 Star (DE3) cells harboring a pET29b plasmid that contained a gene encoding one of our variants were grown in a liquid medium containing a range of ampicillin concentrations and 1 mM isopropyl  $\beta$ -D-1-thiogalactopyranoside (IPTG), which induced overexpression of our TEM-1 variants. The ability of cells to grow at each ampicillin concentration was determined by measuring the optical density at 600 nm ( $O.D._{600}$ ); the lowest antibiotic concentration that inhibited growth was recorded. All rigid designs were observed to exhibit either minimal or no activity against ampicillin (Table 1). The two rigid designs that showed the highest activity against ampicillin, Rdg44c and Rdg262b, had  $MIC_{amp}$  values of 26  $\mu\text{g}/\text{mL}$ , which is two orders of magnitude less efficient than wild type TEM-1 ( $MIC_{amp} = 1500 \mu\text{g}/\text{mL}$ ), but is only half that of GNCA ( $MIC_{amp} = 43 \mu\text{g}/\text{mL}$ ). Alternatively, the  $MIC_{amp}$  values of all the flexible designs were in the range of 375-1500  $\mu\text{g}/\text{mL}$  (Table 1) which is on par with wild type TEM-1.

**Table 1.** Minimal Inhibitory Concentrations (MIC<sub>amp</sub>) and melting temperatures of the TEM-1 variants.

Variant	Minimal Inhibitory Concentration of ampicillin	Melting Temperature
	MIC <sub>amp</sub> (µg/mL)	T <sub>m</sub> (°C)
GNCA	43	90.3
TEM-1	1500	56.4
Rdg44a	< 2**	NM
Rdg44b	< 2	63.1
Rdg44c	26	66.4
Rdg262a	< 2**	NM
Rdg262b	26	56.4
Flx226a	1500	57.4
Flx226b	375	53.2
Flx226c	1500	55.6
Flx256	750	58.1
Flx55	750	58.5

Minimal Inhibitory Concentrations for ampicillin (MIC<sub>amp</sub>) values were determined in lysogeny broth. Melting temperatures (T<sub>m</sub>) were determined using circular dichroism. NM indicates that a T<sub>m</sub> was not established for this protein due to aggregation during purification. \*\*Because these variants precipitated out of solution during purification, it is difficult to know whether these values accurately reflect their activities *in cellulo*.

Two possible explanations for the lack of activity against ampicillin observed in our rigid designs are: 1) that only poor protein expression was achieved or 2) that they did not fold into native-like structures; neither of these possibilities are directly examined in MIC assays. We therefore expressed and purified each of the designed proteins and assessed their abilities to adopt native-like structures using circular dichroism (CD) spectroscopy. All designed proteins were observed to express solubly (Figure 2.4.4.1).

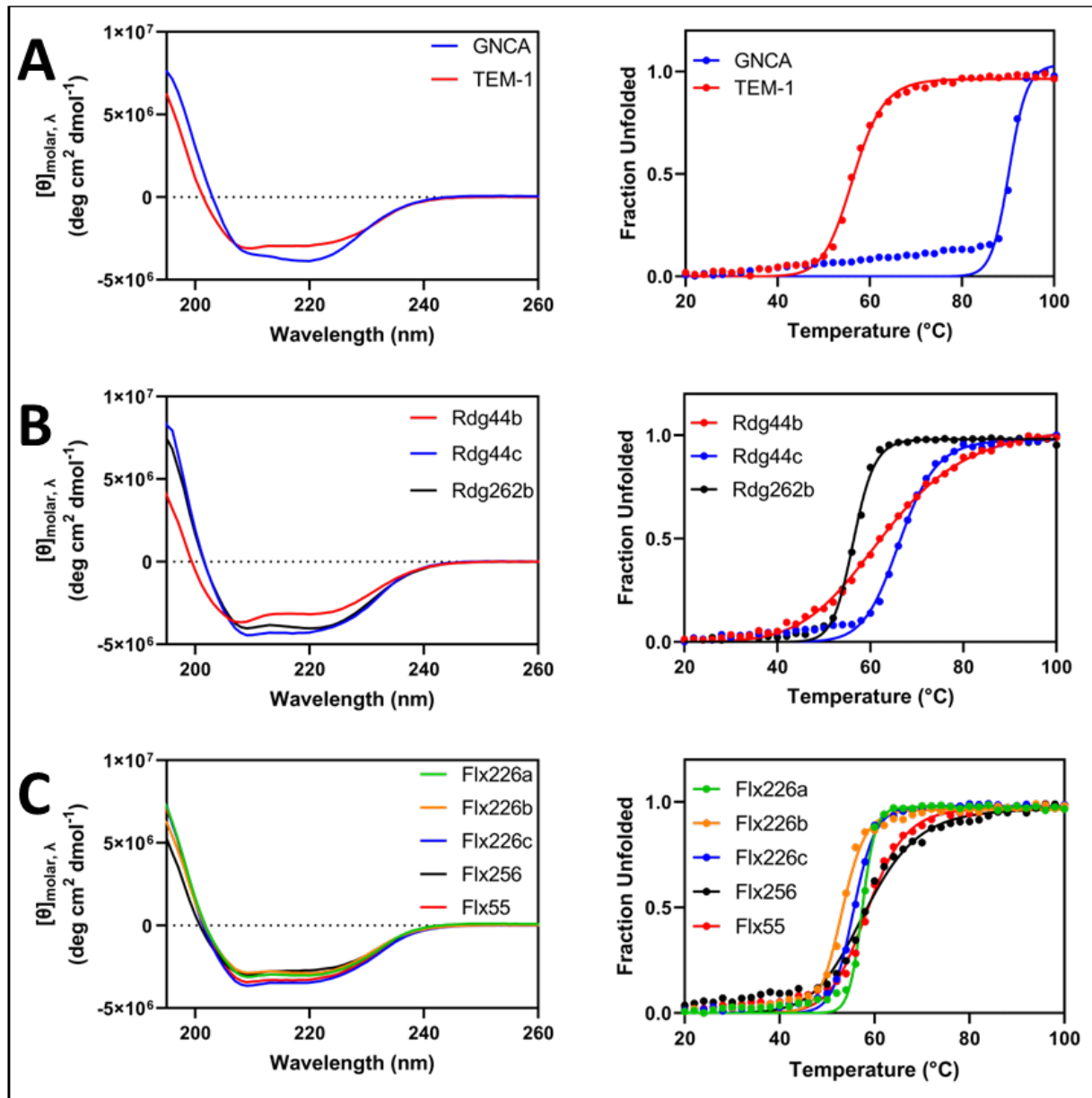


**Figure 2.4.4.1.** 12% SDS PAGE gels of the purified designed proteins. The gels were stained with Coomassie Brilliant Blue G-250. For the gels, proteins were heat denatured. The protein standard (lane 1) is Bio-Rad Precision Plus Protein Kaleidoscope Prestained Protein Standards (A) Flx226a (lane 2) Flx226b (lane 3) Flx226c (lane 4) Flx256 (lane 5) Flx55 (lane 6) (B) TEM-1 (lane 2) GNCA (lane 3) Rdg44b (lane 4) Rdg44c (lane 5) Rdg262b (lane 6).

However, two of the rigid designs, Rdg44a and Rdg262a, had a high propensity to aggregate during the purification process, which precluded further characterization. In contrast, no aggregation of any of the flexible designed proteins was observed throughout the purification process. We subjected all purified proteins to both wavelength scans and thermal melts using CD (see Materials and Methods), which allowed determination of the

melting temperature ( $T_m$ ) of each protein (Figure 2.4.4.2). The  $T_m$ s of all flexible designs fell into a range (53.2 °C to 58.5 °C) that was within ~3 °C of the  $T_m$  of TEM-1 (56.4 °C, Table 1). Alternatively, the  $T_m$ s of the rigid designs varied greatly. Although the least stable of the allosteric rigid designs (Rdg262b) exhibited a  $T_m$  that was on par with TEM-1, two others exhibited marked increases in stability. Namely, Rdg44b and Rdg44c were measured to have  $T_m$ s of 63.1 °C and 66.4 °C, respectively, which correspond to increases of ~6 °C and 10 °C relative to TEM-1.





**Figure 2.4.4.2** Far-ultraviolet circular dichroism wavelength scans and thermal melts with fitted curves of (A) wild type GNCA and TEM-1 (B) protein designs targeting rigid residues and (C) protein designs targeting flexible residues. All measurements were performed in triplicate on a Jasco J-815 spectrophotometer and adjusted for protein concentration. Thermal melts were monitored by the absorption signal at 222 nm with a temperature slope of 5 °C/min. For wavelength scans and thermal melts, the purified protein was in TBS buffer (10mM Tris 50 mM NaCl, pH 7.0) in a cuvette with a 1 mm path length. Protein concentrations were calculated in triplicate using the absorbance at 280 nm and ranged between 0.18-0.25 mg/mL for all scans.

The residues targeted for design in this study exhibit a broad distribution of distances from the active site. For example, the two rigid residues (V44 and V262) are closer to the active site than any flexible residues that were targeted for design with distances of 10.1 Å and 17.3 Å, respectively, while the distance of the flexible residues from a catalytic residue ranged from 17.5 Å–22.1 Å. We therefore sought to assess whether or not a correlation existed with respect to the distance from a targeted residue to the active site and altered enzymatic function. To do this, we calculated the distances between the C $\alpha$ s of all residues mutated during the design process and the C $\alpha$  of the nearest catalytic residue for all experimentally characterized proteins (Table 2) using the PyMOL software (The PyMOL Molecular Graphics System, Version 4.3; Schrödinger, LLC: New York, NY, USA) (*The PyMOL Molecular Graphics System, Version 4.3 Schrödinger, LLC, n.d.*).

The two designed proteins that had the shortest distances between a mutated residue and one of the catalytic residues both targeted residue 262 (Rdg262a and b). Rdg262a carries a mutation at position 233, which is directly adjacent in sequence space to catalytic residue 234. Rdg262b contains the next shortest distance between a mutation and an active site residue at 5.8 Å. Rdg262a showed no activity against ampicillin; it is possible that the observed lack of activity is due to the protein's instability and/or propensity to aggregate as observed during purification. Alternatively, Rdg262b possessed an identical T<sub>m</sub> to TEM-1 but showed minimal activity against ampicillin despite containing a mutation that is only ~6 Å away from a catalytic residue. On the other end of the spectrum, the nearest mutations to any catalytic residue in two of the flexible designs, Flx226a and c, are 18.5 and 17.5 Å away, respectively. Both of these

TEM-1 variants showed near native activity against ampicillin, which is consistent with the fact that mutations that are both distant from and uncoupled to the active site should have little effect on activity.

**Table 2** Mutations present in the computationally designed proteins and the distance of the nearest mutation to a catalytic residue in angstroms.

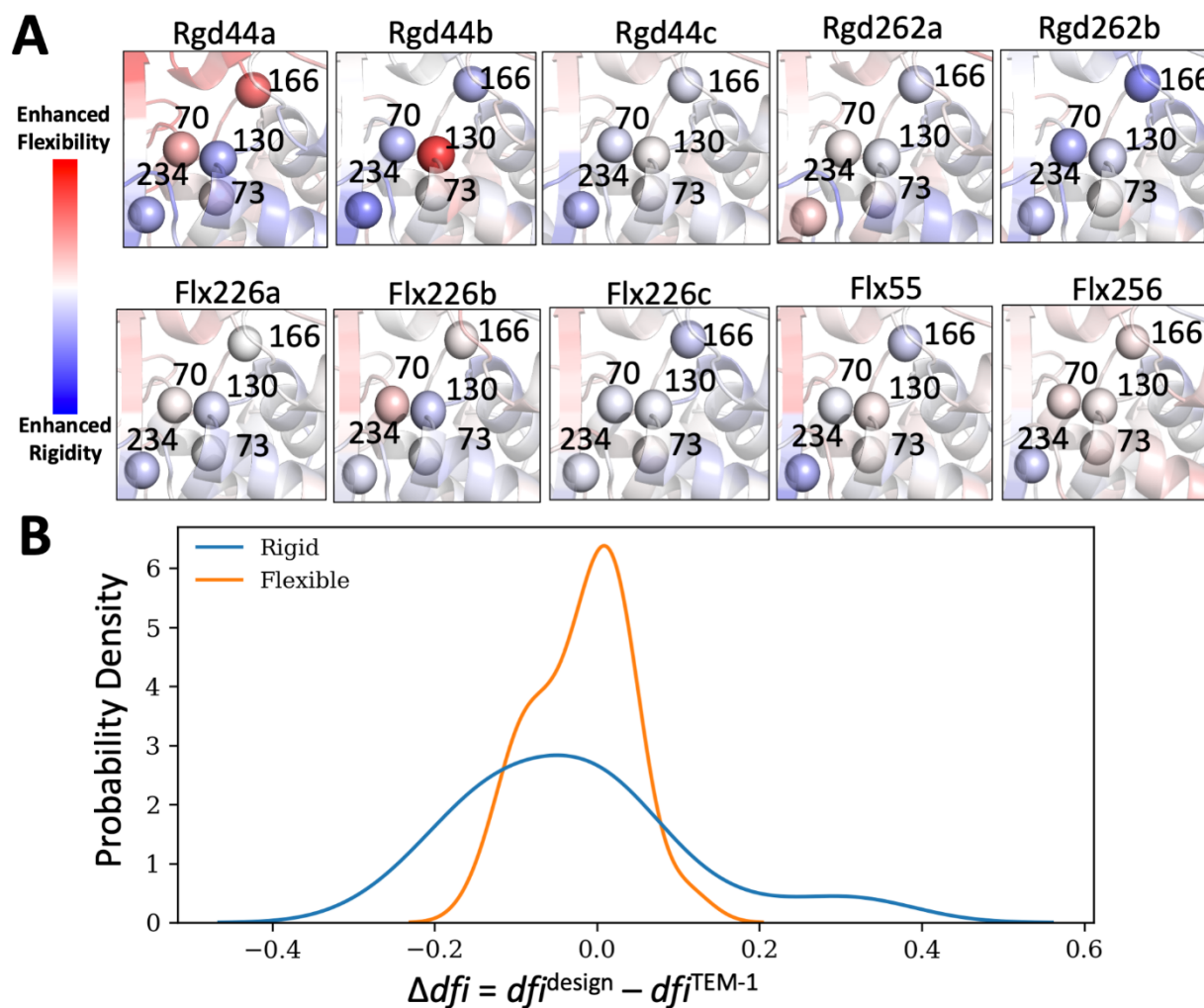
Designed Protein	Mutations	Distance from closest catalytic residue (Å)
Rdg44a	E37V, R43P, Y46M, F60Y, R61N, Y264M, T265M, Q278A	11.8
Rdg44b	D35K, E37V, Q39D, R43P, Y46F, F60Y, R61K, I263T, Y264M, T266S, Q278A	11.1
Rdg44c	K32V, D35Q, Q39K, R244I, T265A, R275L, Q278A	9.7
Rdg262a	Y46F, I47Q, E48L, L49T, S59Y, P62A, M182Q, V184K, T188Q, D233C, Y264F	3.9
Rdg262b	Y46F, E48L, P62A, M182T, V184R, I246Q, Y264M	5.8
Flx226a	D254N, I287F	21
Flx226b	A227P, L250F, D254N, K256V, S258T, I287F, W290Y	12.1
Flx226c	A227P, D254N, K256V, I287F	17.5
Flx256	E212A, A227P, A249S, D254N	9
Flx55	E48L, L51A, N52D, S59A, V184R, T188E, T195L, A249M	9.8

In the remaining designs, the distribution of distances between the nearest catalytic residue and a designed mutation are much more similar irrespective of whether rigid or flexible residues were targeted. For example, Rdg44a and Flx226b both have mutations

that are 12.1 Å from a catalytic residue and Rdg44c and Flx55 have mutations that are 9.7 Å and 9.8 Å away from the catalytic residues, respectively. As these pairs of proteins contain one rigid and one flexible design and also exhibit similar distances between the nearest mutation and any catalytic residue, they appear to provide a direct test of the implications of targeting mutations to flexible vs. rigid residues. Interestingly, Rdg44a was highly unstable and aggregation prone despite only having mutations over 10 Å away from the catalytic residues. In contrast, Rdg44c had activity against ampicillin that was three orders of magnitude less than the wild type protein, but also showed a 10 °C increase in  $T_m$  relative to TEM-1. Alternatively, both flexible designs (Flx226b and Flx55) maintained substantial activity against ampicillin and exhibited  $T_m$ s that were within 3 °C of wild type TEM-1 (Table 1). These data further support the notion that rigid, highly coupled residues play a large role in determining both the activity and physical properties of TEM-1. Furthermore, the fact that the rigid designs that adopted a native-like fold showed a substantial decrease in activity supports the notion that our dc metric can provide meaningful information regarding residues that may be able to affect protein function via allosteric dynamic coupling to the active site.

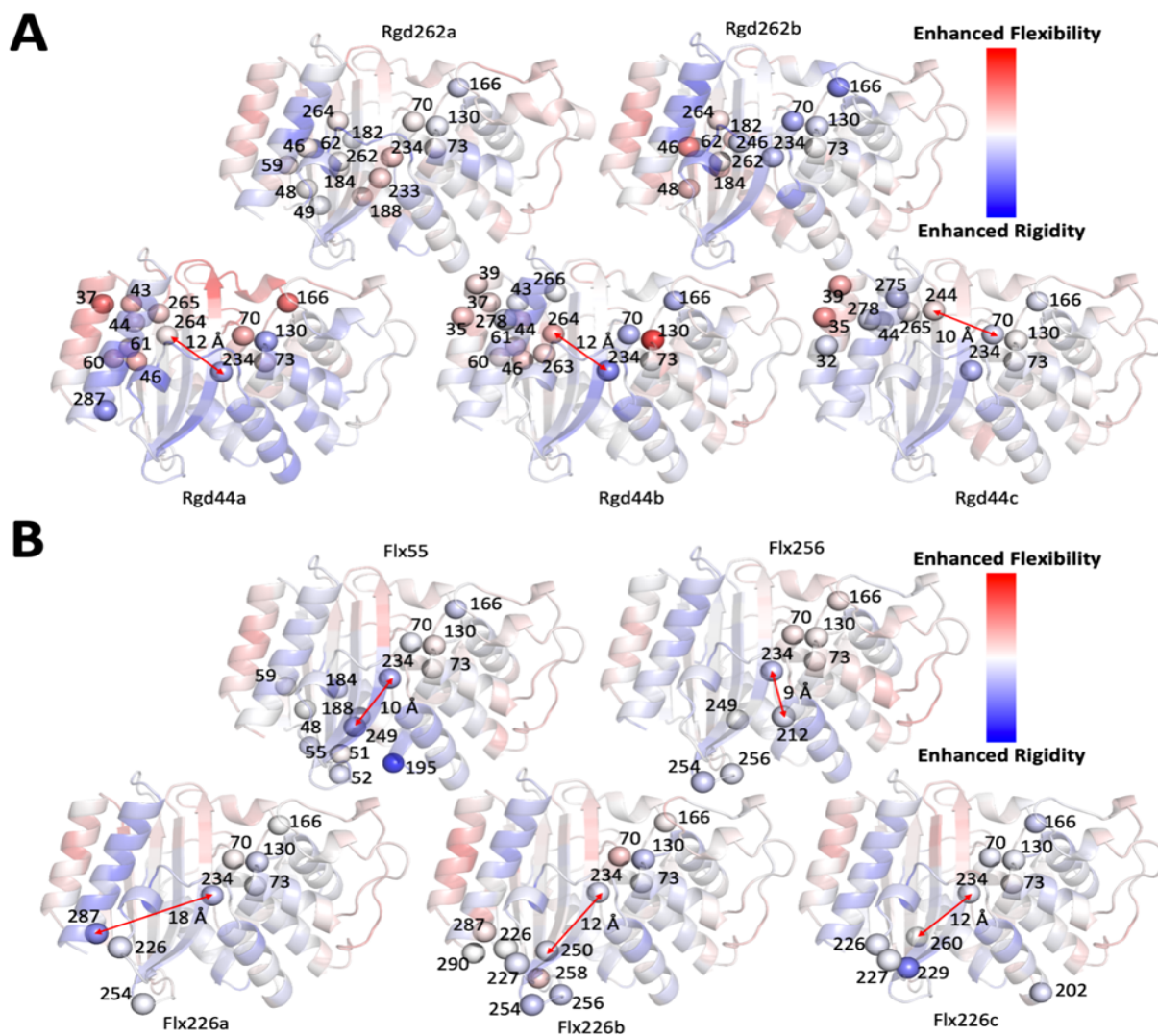
**2.4.5 Dynamics Analysis of the Designed Proteins.** Experimental characterization of our designed proteins demonstrated that the  $MIC_{amp}$  values of the rigid designs were significantly reduced relative to both TEM-1 and the flexible designs irrespective of the distances between the nearest mutations and the catalytic residues. This suggests that changes in the local network of interactions surrounding rigid residues that exhibit long-range dynamic coupling with the active site may allosterically alter the flexibility of active site residues. To further analyze this possibility using our computational metrics,

we calculated the flexibility of the active site residues in both sets of designed proteins using the dfi metric. The dfi values of each catalytic residue in our experimentally characterized proteins were subtracted from those of TEM-1 to generate a  $\Delta dfi$  profile (Figure 2.4.5.1.A).



**Figure 2.4.5.1.** The change in dynamics as measured by the  $\Delta dfi$  mapped onto the catalytic residues of each experimentally characterized protein. A) Catalytic residues are modeled as spheres and color coded by their change in dfi score relative to TEM-1. B) The  $\Delta dfi$  distribution of active site residues in the flexible and rigid designs. The flexible design distribution shows a low variance compared to that of the rigid designs. A change in dfi score of + 0.2 is noteworthy as it is indicative of a shift in flexibility. This analysis suggests that designing new interactions around a rigid residue that is dynamically coupled to the active site can allosterically modulate the flexibility/rigidity of the amino acids in the active site.

A clear difference between the  $\Delta df_i$  values of the catalytic residues of the rigid and flexible designs was observed (Figure 2.4.5.1). Namely, the catalytic residues in the rigid designs underwent a greater change in relative flexibility (both increases and decreases) compared to the flexible designs. Alternatively, the relative flexibilities of the catalytic residues in the flexible designs exhibited a narrower distribution centered at zero (Figure 2.4.5.2). These data support the notion that the rigid residues we chose are highly coupled to the active site (as suggested by our original dci analysis) and also that targeting the local interaction of allosteric rigid residues can indeed alter the flexibilities of residues, even if they are separated by substantial distances.



**Figure 2.4.5.2.** The change in the dynamics profiles of experimentally characterized rigid (A) and flexible (B) designs ( $\Delta dfi$  values) are mapped onto the TEM-1 structure. Point mutations around the residues targeted for design and the catalytic residues in TEM-1 are shown as spheres and labeled with their residue indices. The distance between the mutations closest to the catalytic residues are marked with red arrows and labeled with the corresponding distance in angstroms. The minimum distance in most designs is larger than 10 Å (Rgd262a and b and Flx256 are exceptions), which suggests that the changes in dynamics of catalytic residues is due to distal allosteric communication with the active site in many instances.

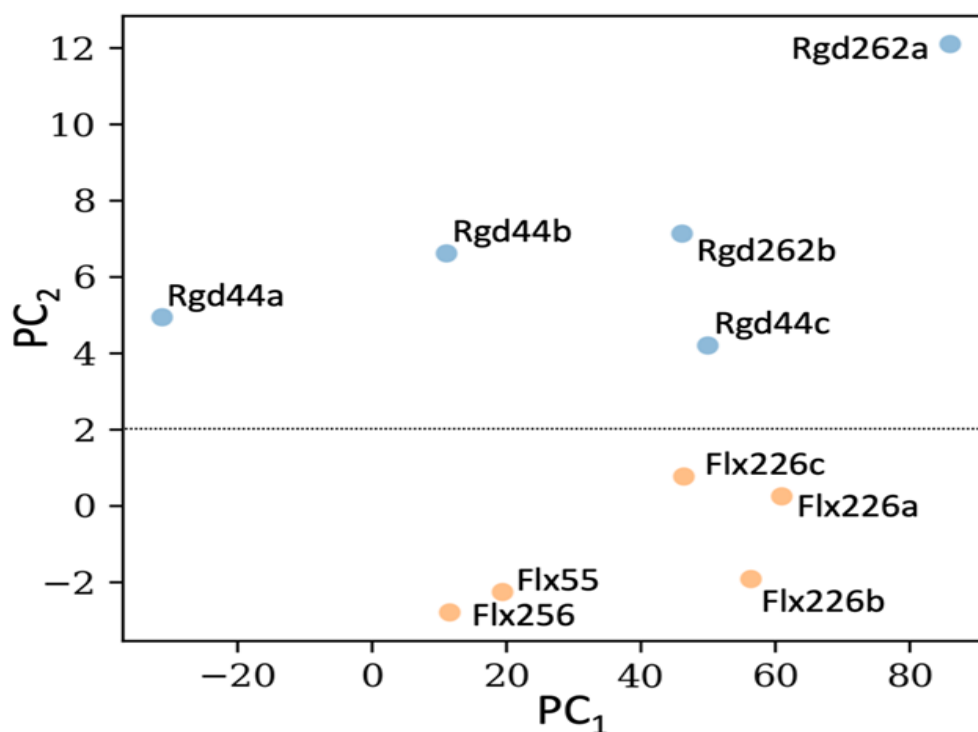
Our experimental results and the detailed dfi profiling of the experimentally characterized designs brought to light the fact that our initial PCA analysis did not appear to adequately discriminate between the activities of the designed proteins. Although

designs in which rigid, coupled residues were targeted often possessed vastly different properties than those in which flexible, uncoupled residues were targeted, many of these designs clustered in similar areas of the PCA (Figure 2.4.3.2). Therefore, we sought to develop a new metric that might have a greater discriminatory ability than the PCA alone. We therefore developed an iterative method that we have termed the Dynamic Distance Analysis (dda) in which the “dynamic distance” of a designed protein to either TEM-1 or GNCA is computed relative to those of randomly selected groups of designed proteins. As the distance between any two proteins in a PCA (based on their three principal eigenvectors, see Methods and Materials) depends on the component proteins used to generate that PCA, randomly selected sets of designed proteins should yield a much better picture of the true relationship between a given designed protein and a target protein (TEM-1 and GNCA).

To generate the dda profiles of our designed proteins, we used a bootstrapping approach in which we first generated multiple PCAs using small, randomly chosen subsets of designed proteins and then iteratively measured the distances between the dda profiles of each designed protein and both GNCA and TEM-1 (Figure 2.4.5.3). When we clustered the dda profiles of the rigid and flexible designs using a new PCA; a clear separation between the two emerges (Figure 2.4.5.3), which correlates well with their biophysical characterization. For example, flexible designs Flx55 and Flx256 cluster together in our dda analysis and also possess similar MIC<sub>amp</sub> values (750 µg/mL). Similarly, Flx226a and Flx226c, whose MIC<sub>amp</sub> values are the same as TEM-1 (1500 µg/mL), also appear in very similar regions of the dda PCA. The two rigid designs, Rgd44a and Rgd262a, which exhibited aggregation during purification, are both found as



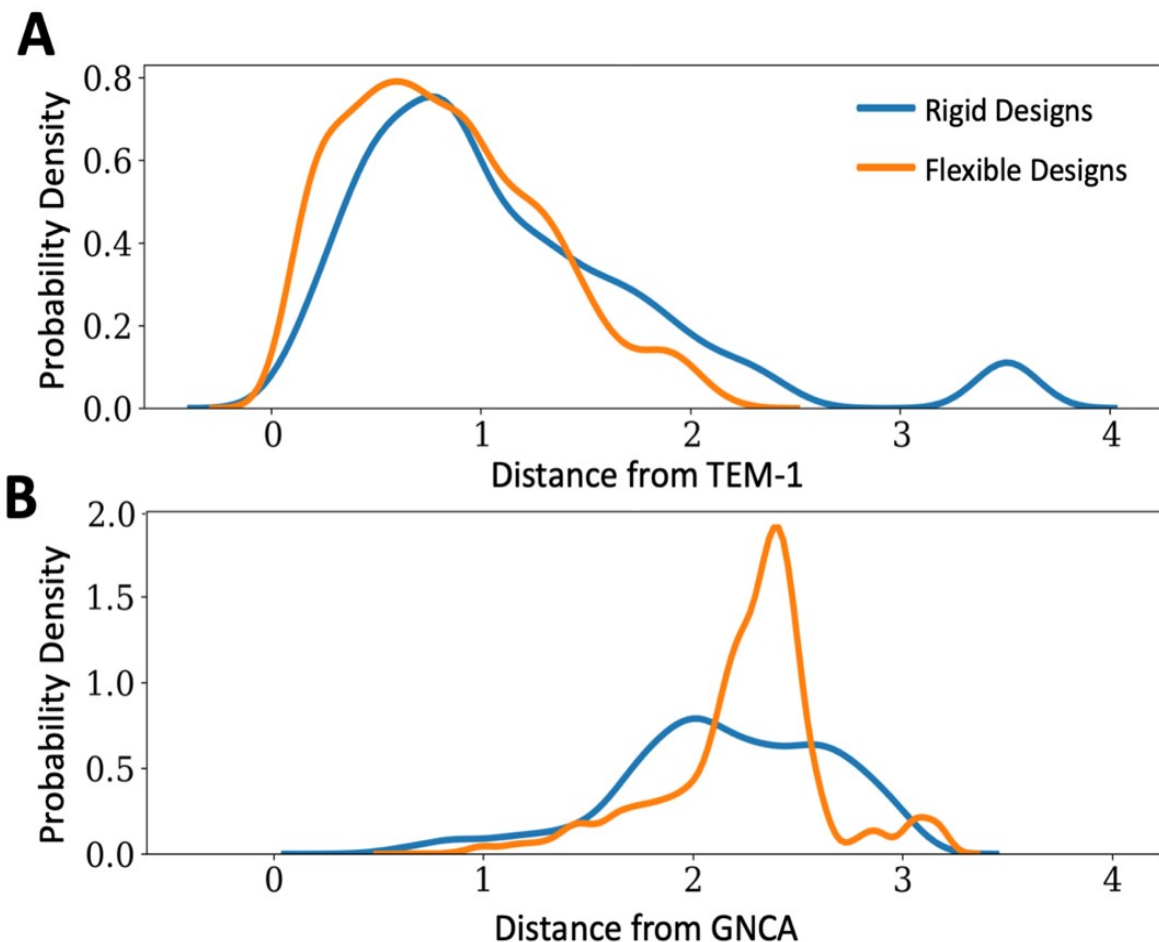
outliers in the dda clustering. Notably, Rgd44c and Rgd262b, which exhibit higher thermostabilities and similar MIC<sub>amp</sub> values to TEM-1, are also clustered in the same vicinity.



**Figure 2.4.5.3** The dynamic distances are clustered for all characterized allosteric rigid (blue) and uncoupled flexible (orange) designs. The weights of PC<sub>1</sub> and PC<sub>2</sub> are 250 and 30, respectively. The rigid designs and the flexible designs cluster separately. Designed proteins with similar MIC<sub>amp</sub> values, (Flx55 and Flx256), (Flx226c and Flx226a), (Rdg262b and Rdg44c) cluster in the same vicinity.

In an effort to assess whether or not the trends observed in the dda analyses of experimentally characterized proteins were universal, we applied dda to all the designed proteins, even those not chosen for characterization. Interestingly, the dynamic distances of the rigid designs are biased away from TEM-1 relative to their flexible design counterparts (Figure 2.4.5.4); conversely, the flexible designs form a narrower distribution that is closer to TEM-1. This suggests that flexible residues that are not

coupled to the active site do not likely contribute to the collective motion of the protein as substantially as do rigid residues. When the distances of our designed proteins to GNCA are considered, the uncoupled flexible designs display a sharp, narrow distribution that is distant from GNCA (Figure 2.4.5.4). Alternatively, the distribution of the rigid designs is broad and contains proteins with dynamic profiles that are more like that of GNCA. These data suggest that the re-design of the environment surrounding rigid residues appears to alter the dynamics of TEM-1 more substantially than when the environment surrounding uncoupled flexible residues is targeted.



**Figure 2.4.5.4.** Dynamic distance distribution from (A) TEM-1 and (B) GNCA for all experimentally characterized rigid (blue) and flexible designed proteins (orange). The distribution of the rigid designs shows a displacement moving away from TEM-1 and

closer to GNCA. Inversely, the uncoupled flexible designs form a narrow distribution close to TEM-1 and further away from GNCA.

## 2.5 Conclusions

The goal of this work was to better understand the relationship between structure and function in the TEM family of  $\beta$ -lactamases. Building on previous evolutionary studies on the  $\beta$ -lactamase enzyme TEM-1 (Zou et al., 2015), we explored the hypothesis that rigid residues can serve to both establish the global dynamic profile of the enzyme and exert substantial influence over physical properties (e.g., substrate specificities) so long as long-range coupling exists between the rigid residues and the active site. To explore this, we used the Rosetta computational protein design software to re-design the local network of interactions surrounding residues that fit the aforementioned criteria. Our designed proteins were analyzed using computational metrics that assessed both the global dynamic profile and the allosteric coupling of each residue to the active site. Based on these metrics, a subset of our designed proteins was selected for experimental characterization.

Ten designed TEM-1 variants were characterized with respect to the minimal inhibitory concentration of ampicillin as well as thermostability. These data suggested that targeting mutations to environments surrounding rigid residues that were highly coupled to the active site often resulted in a substantial shift in protein stability and function; alternatively, targeting flexible, uncoupled residues resulted in protein variants with more native-like activities and thermostabilities. Namely, when mutations were targeted to the vicinity of two rigid residues that do not directly interact with the active site, but which are highly coupled to it, a substantial reduction in TEM-1's ability to

degrade its native substrate was observed in all cases even though native-like folds were maintained in many cases. Alternatively, thermostabilities and activities against TEM-1's native substrate were maintained in a set of designed proteins in which residues that were neither rigid nor predicted to be coupled to the active site were targeted for mutagenesis. These results are consistent with our computational analyses of the designed proteins' dynamics. Namely, it appears that altering the local interactions surrounding rigid residues that are highly coupled to the active site can allosterically alter the flexibility profiles of active site residues at a distance, which can in turn alter the biophysical properties of the enzyme. In an effort to identify an analytical method that was more informative as to the activities that designed proteins might possess, we developed a novel metric that measures the “dynamic distance” between two proteins. Many of our designed proteins with similar functional properties were observed to cluster together when analyzed by this algorithm. These results not only further support the potential importance of mutations in the vicinity of rigid residues, but also support the fact that coupling between distal residues and the active site can have profound effects on enzyme activities.

The relationship between protein dynamics and function is highly complex and studying it represents an exceedingly difficult challenge (Ma et al., 2011; Maier et al., 2015; Orenca et al., 2001; Salverda et al., 2010; M. K. Singh & Dominy, 2012; Zhang et al., 2020). Our approach represents a new method for exploring this subject in a highly directed manner. We hope that additional application of these methods to distinct residues in TEM-1 will ultimately provide a more complete understanding of the complex dynamic landscape present in this class of proteins. This could not only facilitate a rapid

prediction of the biochemical properties of new clinical isolates but could also pave the way for the development of new antibiotics that specifically target new protein conformations accessible only through alterations of the global dynamic profile. Finally, the methods reported here could also find use in understanding the dynamic profiles of other enzyme classes, which could have profound implications from the perspective of understanding and treating diseases.

## **2.6 Acknowledgements**

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## CHAPTER 3

### STRUCTURAL INSIGHTS INTO TEM-1 VARIANTS

#### 3.1 Abstract

To identify if the functional and stability changes observed in the TEM-1 variants from the previous study were due to structural changes introduced by the mutations, Flx226a, Flx226b, Flx226c, Rdg44c and Rdg262b were chosen for crystallization. These variants were chosen because they represent a wide range of function against ampicillin as well as a wide range of melting temperatures as measured by CD spectroscopy.

Models of the variants, solved using molecular replacement, were compared to wild type TEM-1 to see if any of the variants had conformational changes in the active site architecture. Aside from the presence of a pH dependent acetate adduct on Ser70 in all the models, the main catalytic residue, the active site architecture was unremarkable. One residue, Tyr105 that delineates one of the sides of the active site was observed in two conformations, an “inward” and an “outward” conformation that was independent of catalytic efficiency. Overall, the models did not provide obvious reasons for differences in function amongst the variants. However, the observation of the pH dependent serine acetate adduct, has to my knowledge, never been reported and is potentially of interest to industrial applications where serine hydrolases are used in low pH conditions.

#### 3.2 Introduction

In Chapter 2, we explored how changing networks of interactions around residues calculated to be rigid and highly coupled to the active site would affect enzyme function in TEM-1. To this end, we employed RosettaDesign to make mutations around target residues and characterized chosen designs for changes in function using MIC<sub>amp</sub> assays

and for changes in stability using thermal melts. When we designed around rigid residues, enzyme function and thermostability was substantially affected. However, when we designed around the flexible residues, native function was basically maintained and melting temperatures were similar to wild type.

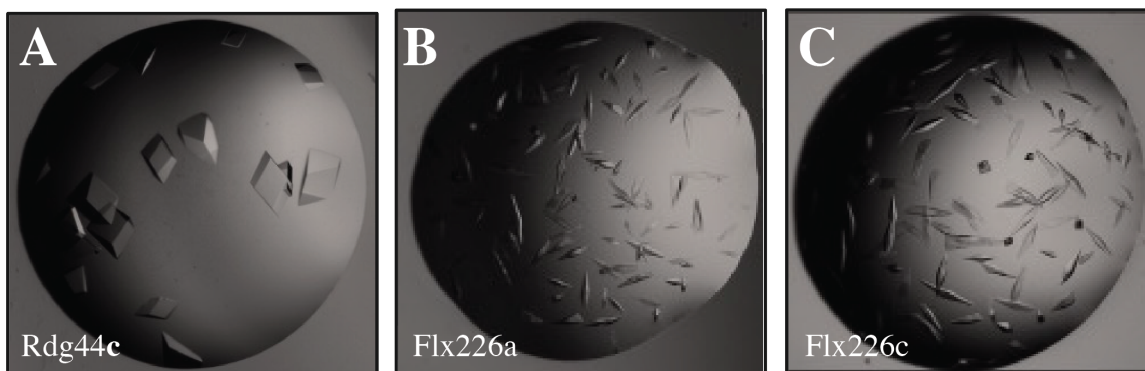
To identify if the mutations made to the variants caused structural changes, Flx226a, Flx226b, Flx226c, Rdg44c and Rdg262b were chosen for crystallization attempts. These variants were chosen because they represent a wide range of function against ampicillin as well as a wide range of melting temperatures (Table 1). Flx226a and Flx226c maintained wild type catalytic efficiency and thermostability as measured by MIC<sub>amp</sub> assays (1500 µg/mL) and CD spectroscopy. However, the third flexible variant, Flx226b, was an order of magnitude less efficient against ampicillin (375 µg/mL) when compared to wild type TEM-1. In addition, the melting temperature of Flx226c was 3 °C lower than wild type which could have been a factor in the loss of function. Rdg44c and Rdg262b, the two designs where rigid residues were targeted, had a significant loss of function, maintaining only ~2% of wild type activity (26 µg/mL). However, the thermostability of Rdg262b was unaffected by the mutations while the melting temperature of Rdg44c went up by 10 °C. It was these differences in both function and thermostability that lead me to believe that there might be structural changes in the variants.

### **3.3 Materials and Methods**

**3.3.1 X-ray Crystallography.** Variants were expressed and purified following the procedure outlined in Chapter 2 Materials and Methods. After verifying purity using a 12% SDS PAGE gel, a solution of Flx226a (10mM Tris 50 mM NaCl, pH 7.0) was

divided in half to determine the crystallization conditions that would lead to diffraction quality crystals. One sample was concentrated to  $A_{280} \sim 9$  mg/mL and the other was concentrated to  $A_{280} \sim 20$  mg/mL for high-throughput crystal condition screenings. Protein concentrations were obtained using a NanoDrop visible spectrophotometer (Thermo Scientific, Waltham, Massachusetts). A Mosquito crystallization robot (SPT Labtech, Melbourn (Cambridge), UK) performed primary screening of crystallization conditions against three hanging drop vapor diffusion Hampton Research crystal screening libraries (Index HT, SaltRx HT, and PEGRx HT). Each screen contained 96 conditions and each condition was tested three times with different v/v ratios of protein to reservoir drop for a total of 864 conditions for each protein concentration in 100, 200, and 300 nL drop sizes. The plates of conditions around the proteins at 9 mg/mL were incubated at room temperature while the plates of conditions around the proteins at 20 mg/mL were incubated at 4 °C. The condition that produced crystals after 7 days was screened using larger-volume hanging drop vapor diffusion to identify the optimal pH, buffer-to-protein ratio, and concentration of the cryoprotectant, polyethylene glycol (PEG) 300, to grow large, single crystals. Diffraction-quality crystals were grown in similar conditions at room temperature for the five variants at a concentration  $\sim 9$  mg/mL. Flx226a, Flx226c and Rdg44c crystals were grown in 0.1 M sodium acetate pH 4.9, 45% PEG 300. Flx226b and Rdg262b crystals were grown in 0.1 M sodium acetate pH 4.9, 50% PEG 300. Each drop contained 2  $\mu$ L of protein solution mixed with 2  $\mu$ L of reservoir solution. Crystals in the shape of tetragonal bipyramids were grown until no new growth was apparent which was  $\sim 1.5$  weeks (Figure 3.2.1).





**Figure 3.3.1** Photographs of three hanging drops of protein crystals. (A) Crystals of Rdg44c. (B) Crystals of Flx226a. (C) Crystals of Flx226c. All of the TEM-1 variants crystallized as tetragonal bipyramids.

Initially, crystals for Rdg262b grew in a starburst geometry from a single point which was suboptimal for diffraction. Therefore, one “starburst” crystal was crushed and used to seed fresh solutions of a 1:1 ratio of protein to mother liquor using a streak seeding technique with a cat whisker. New diffraction quality crystals for Rdg262b grew in ~2 days.

To obtain a dataset of Flx226b at physiological pH, left-over crystals from successful initial crystallization attempts were harvested and soaked for 30-60 minutes in increasing (v/v) concentrations of 100 mM Tris-HCl pH 7.5, 50% PEG.

All crystals were harvested, and flash frozen in liquid nitrogen prior to data collection at 100 K. Data was collected at the Stanford Synchrotron Radiation Lightsource (SSRL) operated for the U.S. Department of Energy by Stanford University. Data were indexed, refined, integrated, and scaled using the XDS software package (Kabsch, 2010a, 2010b). All structures were solved by molecular replacement with the Phaser 2.7.17 software using PDB ID: 1btl as the search model (McCoy et al., 2007). Model building took place using the program Coot (Emsley et al., 2010). All models

were refined using Refmac5 which is part of the CCP4 software package (Kovalevskiy et al., 2018; Murshudov et al., 1997, 2011; Nicholls et al., 2018; Winn et al., 2011).

**3.3.2 Mass Spectrometry of Flx226a and Rdg44c.** Protein solutions left-over from previous crystallization attempts of Flx226a and Rdg44c were centrifuged at 13,000x g for 10 minutes at 4 °C to remove any precipitated protein from the solution. Solutions were then divided in half. One sample was dialyzed overnight at 4 °C into 10 mM sodium acetate pH 4.5, 50 mM NaCl. The second sample was dialyzed overnight at 4 °C into 10 mM Tris pH 7.0, 50 mM NaCl. Samples were diluted to ~50 µg/mL in 18 MΩ water. Data were collected via direct injection on an Agilent 6530 Accurate-Mass Q-TOF LC/MS instrument. Deconvolution of the spectra was done with the Agilent Masshunter Bioconfirm software package.

### **3.4 Results and Discussion**

Crystal structures for the TEM-1 variants were solved to the following resolutions. Flx226a and Flx226c were solved to 1.52 Å and 1.75 Å, respectively. Flx226b (pH 4.9) and Flx226b (pH 7.5) were solved to 1.25 Å and 1.75 Å, respectively. Rdg44c and Rdg262b were solved to 1.04 Å and 1.53 Å, respectively (Table 3).

**3.4.1 Electron Density Around Ser70** The active site architecture was of immediate interest when comparing the models built from crystal structure data of the variants. I hypothesized that when we designed around the highly coupled rigid residues that changes in structure may have been propagated to the active site. These changes may have been as subtle as a change in catalytic residue side chain conformation or as grand

**Table 3** Crystallization statistics for the TEM-1 variants

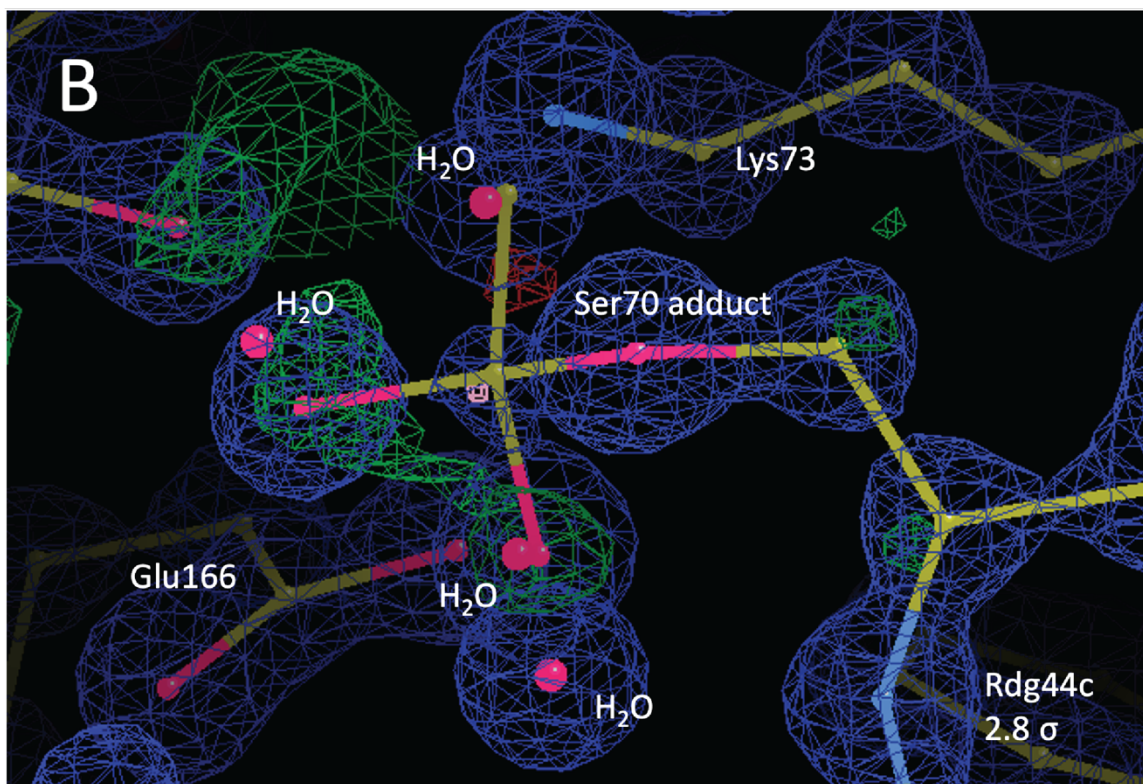
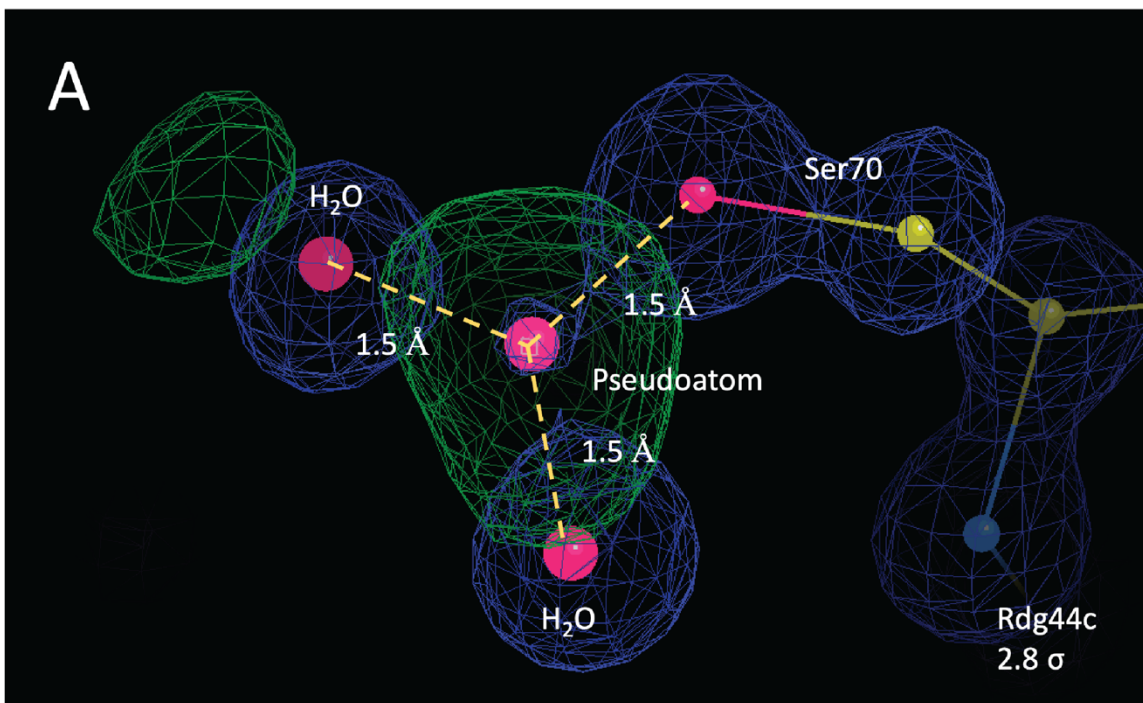
Structure	Rdg44c	Rdg262b	Flx226a	Flx226b (pH 4.9)	Flx226b (pH 7.5)	Flx226c
<b>Experimental</b>	0.1 M sodium acetate pH 4.9 45% PEG 300	0.1 M sodium acetate pH 4.9 50% PEG 300	0.1 M sodium acetate pH 4.9 45% PEG 300	0.1 M sodium acetate pH 4.9 50% PEG 300	0.1 M Tris-HCl pH 7.5 50% PEG 300	0.1 M sodium acetate pH 4.9 45% PEG 300
<b>Data collection</b>						
<b>Space group</b>	P 32 2 1	P 41 21 2	P 4 2 2	P 4 2 2	P 4 2 2	P 4 2 2
<b>Cell dimensions</b> a, b, c (Å) $\alpha, \beta, \gamma$ (deg)	63.612 63.612 126.928 90 90 120	46.552 46.552 257.558 90 90 90	46.565 46.565 257.7 90 90 90	46.67, 46.67, 257.434, 90, 90, 90	47.078, 47.078, 257.168, 90, 90, 90	46.508 46.508 257.099 90 90 90
<b>Total Reflections</b>	1624602	581225	637440	1071646	182556	971156
<b>Unique Reflections</b>	128981	43723	44236	70855	30291	57640
<b>Resolution (Å)</b>	33.5552 - 1.0407	50.000 - 1.529	46.565 - 1.52463	46.67 - 1.25	47.078 1.74807	46.508 - 1.32735
<b>I/<math>\sigma</math>(I)</b>	26.1	12.1	9.2	31.3	13.5	10.3
<b>R<sub>meas</sub></b>	0.028	0.097	0.123	0.026	0.033	0.12
<b>R<sub>pin</sub></b>	0.008	0.026	0.033	0.007	0.014	0.029
<b>CC1/2</b>	0.836	1.321	1.774	0.607	1.348	1.812
<b>Completeness (%)</b>	90.29	98.1	98.28	87.84	98.46	85.87
<b>Redundancy</b>	12.6	13.3	14.4	15.1	6	16.8

as an alternative backbone placement. Notwithstanding, our observations of the electron density maps of the variant active sites were originally befuddling. In the electron density maps of every variant, we observed a strong electron density peak centered around the main catalytic residue, Ser70. This electron density was tetrahedral in shape extended from the side chain of Ser70 to a lobe  $\sim 1.5$  Å from Ser70 O $\gamma$ . Because the wild type TEM-1 model has conserved two waters in the active site 2.7 and 2.9 Å away from Ser70 O $\gamma$ , I reasoned that the density was due to the presence of extra waters. Therefore, first attempts at building the active site model focused on the addition of 2-3 waters  $\sim 2.6$  Å from Ser70 O $\gamma$ . After refinement with Refmac5, there was still a strong positive difference peak extending from Ser70 O $\gamma$  to a sphere 1.5 Å away and centered between the Ser70 O $\gamma$  and two modeled water molecules (Figure 3.4.1.1.A). It was evident that the electron density near Ser70 was not simply extra waters. The distance between the density peak and Ser70 O $\gamma$  prompted us to explore the possibility that a small molecule either in the cytosol, purification buffers, or in the mother liquor must have reacted with

Ser70 to form a tetrahedral adduct. The simplest explanation was that the small molecule was present in the crystallization conditions.

The crystallization buffer that uniformly gave rise to diffraction quality crystals was 100 mM sodium acetate (pH 4.9) with PEG 300, but it seemed unlikely that Ser70 would react with an acetate ion. However, at low pH, an activated Ser could make a nucleophilic attack on a neutral acetic acid molecule. We compared the differences in  $R_{\text{values}}$  post refinement when we solely modeled waters and when we modeled waters with an acetate adduct in the active site. When three waters were modeled the  $R_{\text{values}}$  were 0.1254 and 0.1405 for  $R_{\text{factor}}$  and  $R_{\text{free}}$ , respectively. When Ser70 was modeled as an acetate adduct at 60% occupancy and the three waters at 40% occupancy, the  $R_{\text{values}}$  were 0.1228, 0.1384 for  $R_{\text{factor}}$  and  $R_{\text{free}}$ , respectively (Figure 3.4.1.1.B).

This observation gave us confidence in the presence of an acetate adduct. However, we decided to investigate further with wet lab experiments.



**Figure 3.4.1.1** Electron density and stick models of the Rdg44c active site at 2.8  $\sigma$ . (A) When three water molecules are modeled in the strong electron density surrounding Ser70, a strong difference peak appears 1.5 Å away from Ser70 O $\gamma$  indicating that there is

a covalent bond. In the image, a pseudoatom was placed in the center of the green density for distance measurements. (B) The difference peak is not present when a Ser70 adduct is modeled at 60% occupancy. Three water molecules are modeled at 40% occupancy for the lowest  $R_{\text{values}}$  after refinement with Refmac5. Other catalytic residues are modeled as sticks.

To further explore the presence of a Ser70 adduct, we took older purified protein samples of Flx226a and Rgd44c that were left over from crystallization studies and prepared them for mass spectrometry analysis. We divided both samples into two aliquots. The first sample was dialyzed overnight at 4 °C into 10 mM sodium acetate pH 4.5, 50 mM NaCl. The second sample was dialyzed overnight at 4 °C into 10 mM Tris pH 7.0, 50 mM NaCl. Both samples were then diluted in 18 M $\Omega$  water for input into the mass spectrometer. We reasoned that we could obtain the difference in mass to back calculate the molecular weight of the addition to Ser70. Unfortunately, data from the mass spectrometer did not show any evidence of an adduct and the masses were spot-on for both proteins regardless of dialysis buffer.

Because the crystallization conditions were at low (pH 4.9) we reasoned that the adduct could be pH dependent which is why it was not observed in mass spectrometry experimental data taken at physiological pH. Therefore, our next step to solving the adduct mystery was to collect data on one of the leftover crystals in a buffer at higher pH. We progressively soaked a selection of Flx226b crystals in increasingly higher pH buffers until it was stable in 100 mM Tris-HCl pH 7.5, 50% PEG. This buffer exchange not only raised the pH of the buffer, but also removed the acetate ions from the solution. We collected diffraction data on this crystal from the SSRL and solved the crystal structure for Flx226b at pH 7.5. To our delight, the electron density in the active site of

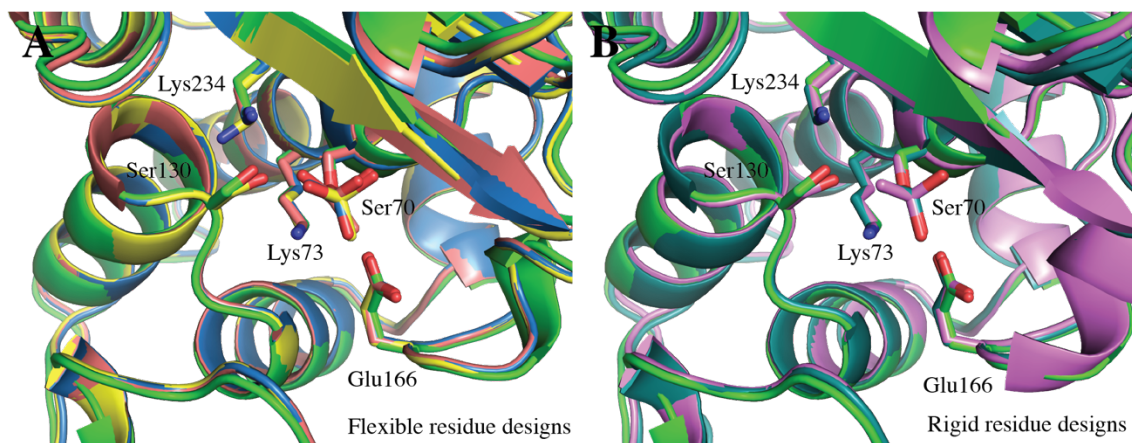
this variant at pH 7.5 was vastly different than that observed at pH 4.9. The electron density clearly showed evidence of the two conserved waters 3.2 and 2.8 Å from Ser70 O $\gamma$ . There was no evidence of an adduct around Ser70, lending weight to our hypothesis that the Ser70 acetate adduct was pH dependent.

**3.4.2 Analysis of the Variant Active Sites** Once the mystery of the adduct was solved, analysis focused on potential changes in the active site architecture that might lead to changes in function against ampicillin. Superimposing the models of the crystallized variants with PDB ID: 1btl showed no obvious differences in backbone conformation or side chain conformations between the variants with serious losses of function to variants that maintained function. All catalytic residues were in the same basic conformation as those observed in the 1btl model (Figure 3.4.2.1).

The only observed differences in catalytic residue side chain conformation was in the models of Flx226b. At pH 7.5, the torsion angle around Ser130 C $\beta$  was observed to be rotated 66 degrees toward Lys234. No other side chains in the active site were affected by this slight rotation (Figure 3.4.2.2). At pH 4.9, with the Ser70 adduct present, the torsion angle around Ser130 C $\beta$  adopts the same basic conformation as the wildtype (1btl) model (within ~15 degrees). However, the 4.9. the torsion angle around Lys234 C $\epsilon$  was observed to be rotated ~75 degrees toward the backbone of I127 to form a sidechain-backbone hydrogen bond (Figure 3.4.2.2). There's no reason to believe that the Lys234 rotation in Flx226b is due to the presence of the adduct since the other variants with a Ser70 adduct at pH 4.9 are observed with Lys234 in same conformation as wild type TEM-1. In all the models aside from Flx226b, Lys234 N $\zeta$  was observed within 2.8 Å



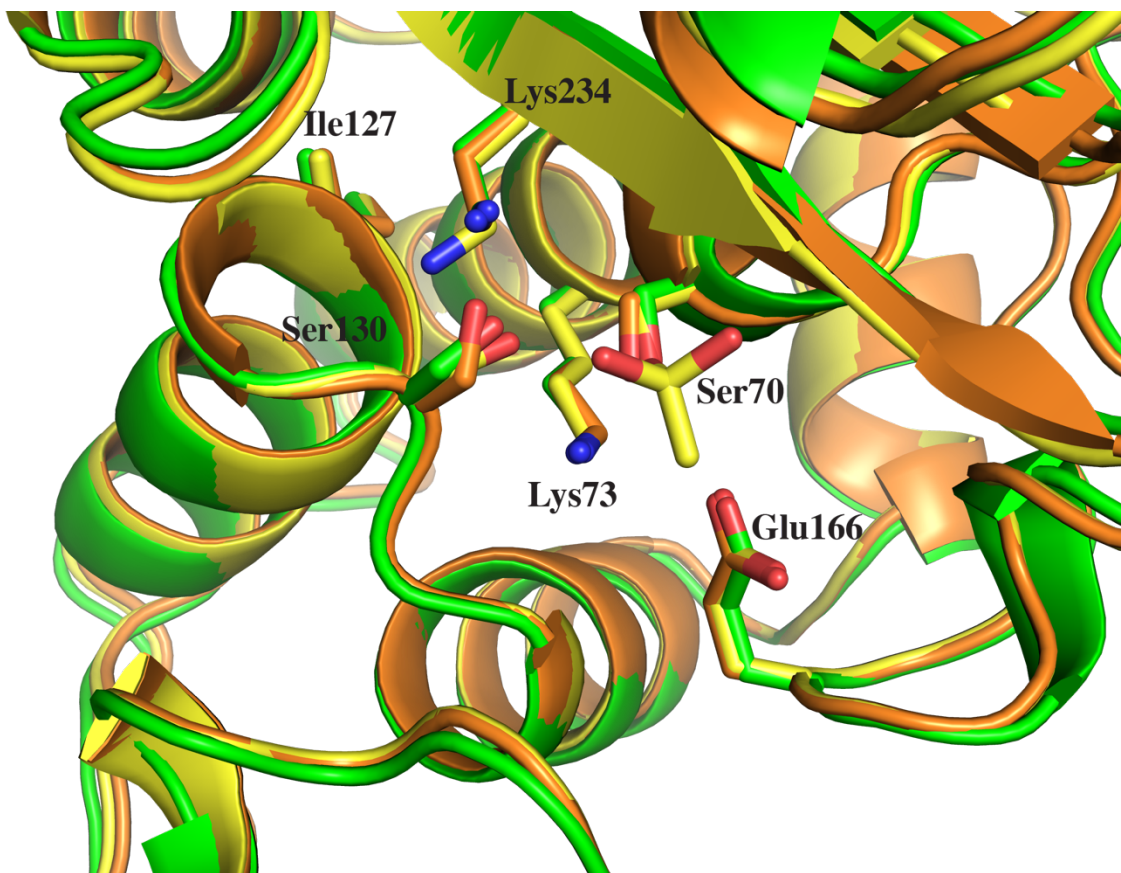
(hydrogen bonding distance) of Ser130 O $\gamma$ , the backbone carbonyl of Ser235, and a conserved water.



**Figure 3.4.2.1** Models of the active site of the crystallized TEM-1 variants at pH 4.9. (A) Models of the flexible designs, Flx226a (salmon), Flx226b (yellow), and Flx226c (blue) superimposed with wild type TEM-1 (PDB ID: 1btl) (green). The main catalytic residues are modeled as sticks. Aside from the torsion angle around Lys234 C $\epsilon$  rotation of  $\sim 75$  degrees toward the backbone of I127, (not shown) to form a sidechain-backbone hydrogen bond, no significant changes in side chain conformation with respect to wild type TEM-1 are observed in the models. (B) Models of the rigid designs, Rdg44c (teal) and Rdg262a (violet) superimposed with wild type TEM-1 (PDB ID: 1btl) (green). The main catalytic residues modeled as sticks. No significant changes in side chain conformation with respect to wild type TEM-1 are observed in the models.

**3.4.3 Analysis of Tyr105** Tyr105 is a Class A conserved residue that delineates one of the edges of the active site wall and has been identified as important to substrate recognition and stabilization (Doucet et al., 2004, 2007b). This stabilization is believed to be due to a stacking interaction between the aromatic ring of Tyr105 and the thiazolidine ring on penicillin-derived antibiotics (Doucet et al., 2004). The proximity of the aromatic ring of Tyr105 to the thiazolidine ring of penicillin-based substrates is apparent in an x-ray crystal structure model at 1.7 Å resolution of TEM-1 in complex with benzyl penicillin (PDB ID: 1fqg) with an E166N mutation preventing it from deacylating the



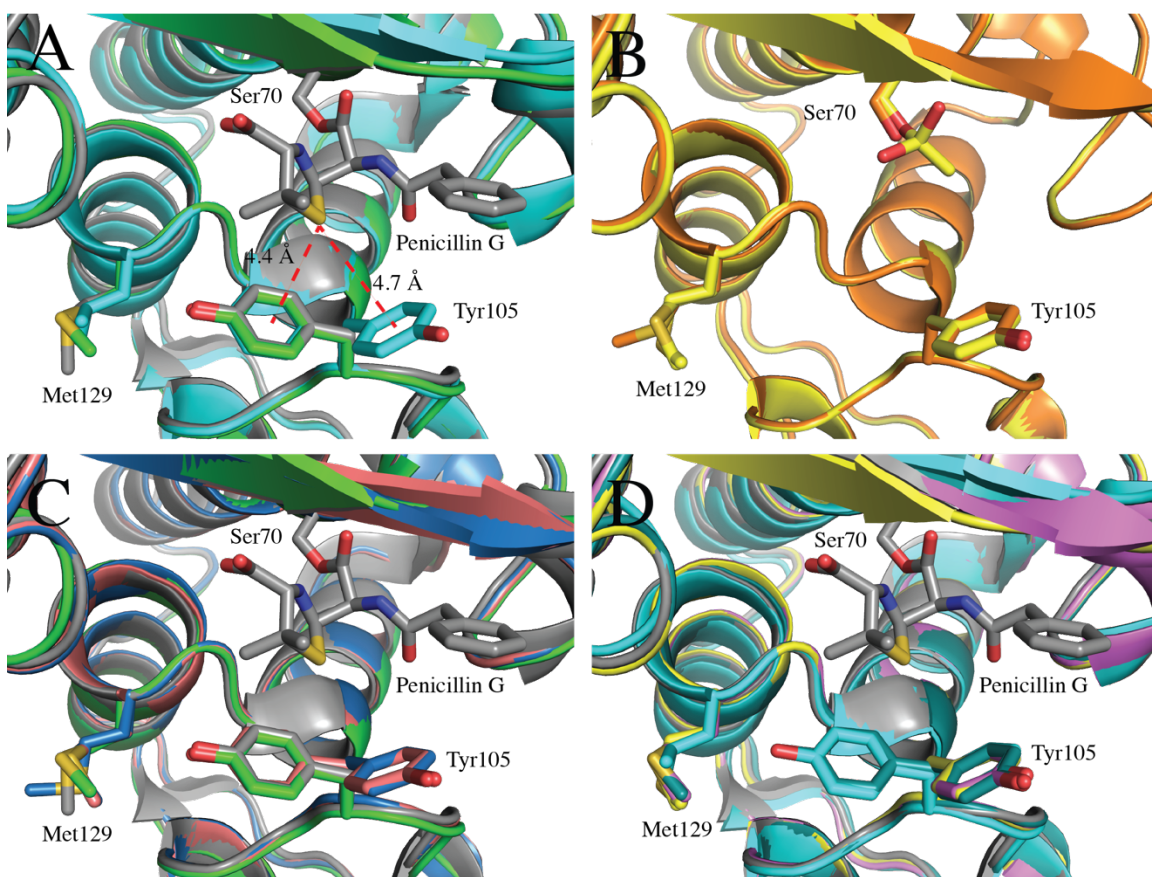


**Figure 3.4.2.2** Models of the Flx226b active site at pH 4.9 (yellow) and pH 7.5 (orange) superimposed with wild type TEM-1 (PDB ID: 1btl) (green). At pH 4.9, Ser70 is modeled with an adduct and the torsion angle around Lys234 C $\epsilon$  is observed rotated  $\sim$ 75 degrees toward the backbone of I127 to form a sidechain-backbone hydrogen bond. At pH 7.5, the torsion angle around Ser130 C $\beta$  is rotated 66 degrees toward Lys234.

acyl-enzyme intermediate in the reaction (Sielecki et al., 2003). In this model, Tyr105 has a torsion angle around the C $\beta$  atom of 63.7 degrees and the phenol is pointing toward Met129. In this “inward” conformation, Tyr105 O $\gamma$  is within hydrogen bonding distance (3.0 Å) of the backbone carbonyl of Met129. Additionally, the Tyr105 aromatic ring is in an edge-to-face stacking interaction with Pro107 as well as within the ideal sulfur- $\pi$  interaction distance of 4.4 Å of the sulfur atom on the penicillin thiazolidine ring (Ringer et al., 2007). Notably, the conformation of Tyr105 in the wild type model (PDB ID: 1btl)

is the same as that observed in the model of TEM-1 with an acyl-intermediate (PDB ID: 1fqg).

Previous studies have found that in a co-crystal structure of a TEM-1 homologue derived from *Staphylococcus aureus* named PC1, bound to an inhibitor, *p*-nitrophenyl[[*N*-(benzyloxycarbonyl)amino]methyl]phosphonate, (PDB ID: 1blh), Tyr105 will adopt an “outward” conformation that is not observed in the wild type TEM-1 structure. Nor is this “outward” conformation observed in the apo TEM-1 crystal structure as previously discussed. With the bound inhibitor, the Tyr105 torsion angle around C $\beta$  is 179.4 degrees so that its phenol points toward solvent (C. C. H. Chen et al., 1993). It has been hypothesized that the Tyr105 outward conformation is due to the presence of the inhibitor. It should be noted that the structure of the methylphosphonate inhibitor covalently bound to Ser70 does not preclude the “inward” Tyr105 conformation observed in the apo-TEM-1 structure and in the TEM-1 E166N structure co-crystallized with penicillin G. The methylphosphonate inhibitor also has a smaller chemical footprint than the acylated penicillin G structure. Additionally, when models of PC1 (PDB ID 1blh), and TEM-1 E166N bound with penicillin G are superimposed, the distance between the center of the Tyr105 aromatic ring in the outward-facing conformation to the sulfur atom on the thiazolidine ring on penicillin is 4.7 Å, also well within sulfur- $\pi$  interaction distance.



**Figure 3.4.3.1** Models of all the crystallized TEM-1 variants at pH 4.9 superimposed with wild type TEM-1 (PDB ID: 1btl) in green, GNCA (PDB ID: 4b88) in cyan and model of TEM-1 covalently bound to penicillin G (PDB ID: 1fqg) in grey. The conformation of Tyr105 varies in the models and no pattern between function against ampicillin and Tyr105 conformation is observed. (A) In wild type TEM-1, (green and grey) the phenol of Tyr105 points toward Met129 and makes a sidechain-backbone hydrogen bond with Met129. In GNCA, Tyr105 adopts the “outward” conformation with the phenol pointing toward solvent. The distance between the sulfur atom on the thiazolidine ring of penicillin G Ser70 adduct and the center of the Tyr105 aromatic ring is 4.4 Å when Tyr105 is in the “inward” conformation. In the “outward” conformation, the sulfur atom on the thiazolidine ring is 4.7 Å. (B) Model of Flx226b at pH 4.9 (yellow) with a Ser70 adduct overlaid with Flx226b at pH 7.5. Tyr105 is in the same “outward” conformation. (C) Models of apo TEM-1 PDB ID: 1btl (green), holo TEM-1 PDB ID: 1fqg (grey), Flx226a (salmon), Flx226c (blue) overlaid. Tyr105 is observed in an “outward” conformation in Flx226c and in two states in Flx226a. (D) Models of GNCA (cyan) and Rdg44c (teal) with Tyr105 in two states. Tyr105 is observed in the “outward facing conformation in models of Flx226b (yellow) and Rdg262b (violet).

Initially, I hypothesized that the “outward” conformation of Tyr105 was indicative of reduced function against penicillin derivatives. This hypothesis originated from the observation that Tyr105 was observed in the “inward” conformation in native TEM-1 (PDB ID: 1btl) while in the inhibited structure, (PDB ID: 1blh), Tyr105 is in the “outward” conformation. In addition, in the model of GNCA, the ancestral homologue of TEM-1 with reduced functionality against ampicillin and increased functionality against cephalosporins, Tyr105 is observed in the two conformations. The reduced activity of GNCA could be explained by the increased flexibility of Tyr105. However, the models of the TEM-1 variants from this study indicate that Tyr105 conformations are highly variable in the apo enzyme and are not linked to changes in function against penicillin-derived antibiotics. The same Tyr105 dual conformation observed in models of GNCA is also observed in the Rdg44c, and the Flx226a models. Rdg44c has a MIC<sub>amp</sub> value of 26 µg/mL, which is lower than GNCA with a MIC<sub>amp</sub> value of 43 µg/mL; while Flx226a has a MIC<sub>amp</sub> value on par with wild type TEM-1 (1500 µg/mL). In the Flx226b, Flx226c and Rdg262b models, Tyr105 adopts the outward-facing conformation only. Flx226b, Flx226c, and Rdg262b have MIC<sub>amp</sub> values of 376 µg/mL, 1500 µg/mL and 26 µg/mL, respectively (Table 1). In none of the models was Tyr105 observed solely in the inward-facing conformation. Therefore, the hypothesis that the Tyr105 conformation is indicative of native function is unsupported with these data.

### **3.5 Conclusions**

Crystallization studies of the TEM-1 variants with varying abilities to hydrolyze ampicillin were performed with the hopes that they would provide a structural

explanation for the observed changes in function and thermostability. Flx226a, Flx226b, Flx226c, Rdg44c, Rdg262b were crystallized, and models were analyzed for changes in structure with respect to wild type TEM-1. Initial analysis focused on the active site and on the large amount of electron density surrounding catalytic Ser70. It was concluded after data was collected on a second crystal of Flx226b at a higher pH that the electron density was due a pH dependent acetate adduct on Ser70. Aside from the presence of an acetate adduct, in all the variants, the catalytic residues in the active site, Lys73, Ser130, Glu166, and Lys234 are in the same conformations as wild type TEM-1.

In the models, Tyr105, a conserved residue that makes up one of the edges of the active site, was observed in two different conformations, an “inward” and “outward” conformation. It was initially hypothesized that the “outward” conformation could have been indicative of reduced functionality. However, this does not seem to be the case as there was no observable pattern between enzyme function and Tyr105 conformation.

### **3.6 Acknowledgements**

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## CHAPTER 4

### REMODELING THE LUCIFERASE ACTIVE SITE TO ACCOMMODATE NOVEL LUCIFERIN ANALOGUES

This chapter is adapted from the following publication: “Love A., Caldwell, D. R., Kolbaba-Kartchner, B., Townsend, K. M., Halbers, L. P., Yao, Z., Brennan, C. K., Ivanic, J., Hadjian, T., Mills, J. H., Schnermann, M. J., & Prescher, J. A. (2023). Red-Shifted Coumarin Luciferins for Improved Bioluminescence Imaging. *Journal of the American Chemical Society*, 145(6), 3335–3345.

This chapter is also adapted from a publication currently in review. “Expedient Synthesis and Characterization of  $\pi$ -Extended Luciferins” by Donald R. Caldwell, Katherine M. Townsend, Bethany Kolbaba-Kartchner, Tanya Hadjian, Joseph Ivanic, Jeremy Mills, Jennifer A. Prescher, Martin J. Schnermann.

In both publications, Bethany Kolbaba-Kartchner performed the Rosetta design methods and analysis.

#### 5.1 Abstract

*In vivo* multicomponent bioluminescence imaging requires a set of robust luciferase-luciferin pairs that can emit in the near-infrared range. However, novel luciferin analogues with red-shifted emission spectra often have a larger chemical footprint than the native D-luciferin that emits in the yellow-green range and the luciferase active site must be re-engineered to accommodate them. A common approach to mold an active site to accept a new ligand is to create a library of variants and to screen or select them for the desired function. The challenge lies in how to identify locations in the protein to target for these libraries. To this end, a high-throughput computational

platform was developed. This platform consists of two parts: a Rosetta docking and design combined with a data analysis portion that uses Jupyter notebooks. With this platform, five novel luciferin-luciferase systems were identified with a  $\lambda_{\max}$  of  $> 650$  nm making them good tools for multicomponent imaging techniques.

Analyzing a case study of one library campaign to identify luciferase variants that can accommodate a series of luciferin analogues inspired by the coumarin molecule indicates the successfulness of this platform. All five of the residues that were identified through library screening as being important to processing the coumarin-derived luciferin analogues were identified in the RosettaDesign platform. This work showcases a new platform for identifying locations to target for library design.

## 5.2 Introduction

Fireflies, potentially the most nostalgic insect, are often associated with fond memories of warm summer evenings as children delight in catching them in jars. Members of the Lampyridae family, fireflies attract curious children thanks to a bioluminescent 62 kDa enzyme aptly named luciferase (Fluc) that emits bursts of yellow-green light from their abdomen to communicate, attract mates, and deter predators (Kaskova et al., 2016). On an atomic level, in an aerobic environment, luciferase catalyzes the oxidation of a small molecule, D-luciferin, resulting in an adenylate intermediate. Oxidation of the adenylated intermediate leads to an electronically excited state oxyluciferin. As the excited state relaxes to its ground state, it releases energy in the form of a photon of light with a  $\lambda_{\max}$  of 554 nm (Branchini et al., 2005) (Figure 5.2.1).





is especially problematic when attempting to visualize events in deeper tissues (Rice et al., 2001).

Red-shifting the reaction emission maximum from 554 nm to > 650 nm would better isolate the luciferase signal from endogenous chromophores and increase the signal-to-noise ratio for better resolution. In addition, the development of new luciferase/luciferin pairs with varying emission maxima would confer the ability to visualize multiple cellular events simultaneously providing a more comprehensive view of complex biological phenomena. Thus, work has been undergoing to develop luciferase-luciferin pairs with varying, but distinct longer-wavelength emission spectra (Love et al., 2023; Yao et al., 2018). This has been successfully done using Directed Evolution techniques (F. H. Arnold et al., 2001; Renata et al., 2015) notably in the case of the development of the Akaluc/Alakumine luciferase/luciferin pair (Iwano et al., 2018). With 21 rounds of random mutagenesis, Iwano et al. identified a luciferase with 28 mutations relative to native Fluc with a  $\lambda_{\max}$  of 650 nm. Despite the success of this method, when one is initially agnostic with respect to locations to mutate in a library, the many rounds of random mutagenesis required is very time and resource intensive.

Rational design offers an alternative approach to identifying target mutations for library design. Typically, identifying locations to target is done by mining the literature for potential sites, performing alanine scans, or by structural studies of protein crystal structures. Although these methods have worked well in the past, they are low throughput. In addition, even when a suitable crystal structure of a protein bound with its endogenous ligand exists, it is still challenging to identify how a novel ligand will bind in the active site and to translate that to identify residues to target in a library.

To this end, I worked on developing a high-throughput computational platform to identify target locations in the active site to include in a semi-rational Combinatorial Codon Mutagenesis (CCM) library (Belsare et al., 2017) aimed at identifying novel luciferin/luciferase pairs with a  $\lambda_{\text{max}} > 650$  nm. The platform was composed of two parts: docking/design and analysis. The first part employed the RosettaMatch algorithm to dock novel luciferin analogues into a starting scaffold and the RosettaDesign algorithm to identify target locations for an experimental library (Richter et al., 2011). The second part employed Jupyter notebooks to analyze Rosetta output to 1) identify holo-enzymes where the ligand was in a conformation believed to be conducive to a productive chemical reaction and 2) calculate the number of times a particular location in the active site was mutated to accommodate the novel luciferin. Because the downstream application of this data was an CCM library, targeting “hot spot” residues identified by Rosetta, the data of interest was less the identity of the mutation, but the locations where Rosetta predicted mutations were necessary to accommodate the ligand in the active site. This workflow resulted in the development of five (CouLuc-3-NMe<sub>2</sub>, two FPLucs, CouLuc-1-NMe<sub>2</sub> and -OH) luciferase-luciferin pairs with emission maxima  $> 600$  nm (Figure 4.3.2.1).

### **4.3 Materials and Methods**

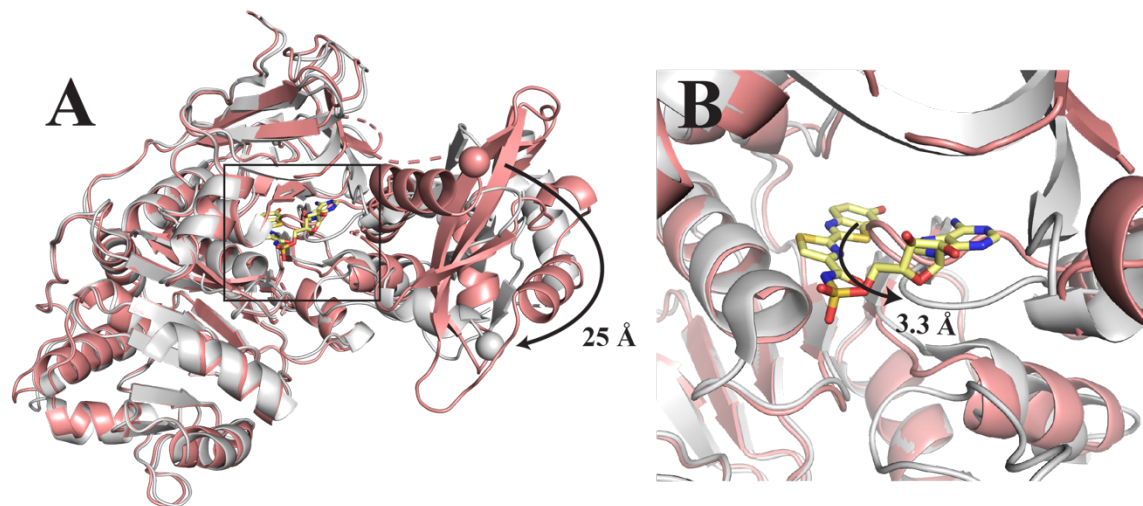
To develop more red-shifted luciferase-luciferin pairs, our collaborators in the Martin Schnermann lab at the NIH synthesized sets of molecules based on three small molecule fluorophores: coumarin, 1-naphthol, and the red fluorescent protein chromophore (Figure 4.3.2.1). They also developed more extended luciferin analogues based on the Akalumine structure (Iwano et al., 2018). Individual molecules in each set differed by the length of the pi wire and the addition of functional groups on the

heterocycles (Figure 4.3.2.1). These additions served to extend the conjugation of the molecule thereby increasing the  $\lambda_{\max}$  of the emission spectra. These novel luciferin analogues were not compatible with native Fluc as the emission max was typically very low at the outset. Therefore, our collaborators in the Jennifer Prescher lab at the University of California, Irvine set up CCM libraries using the NNK degenerate codon to screen for Fluc variants that could process the luciferin analogues with greater efficiency. To reduce the number of rounds of Directed Evolution they approached the Mills lab to employ Rosetta to identify target locations for library design.

Identifying target locations for a semi-rational library design began with building a solid model system which consisted of two separate, but equally important parts: the ligand and the scaffold. Unsurprisingly, both structures required preparative steps prior to modeling.

**4.3.1 Preparation of the Luciferase Scaffold.** In solution, luciferase folds into two domains: a large N-terminal domain and a small C-terminal domain (Conti et al., 1996). There is a dynamic cleft between the two domains that closes upon ligand binding undergoing a massive change in protein conformation including a 25 Å displacement of the C-terminus (Figure 4.3.1.1.A). Additionally, the active site itself undergoes drastic conformational changes upon ligand binding with the loop composed of residues 315-320 moving 3.3 Å to open up the space (Figure 4.3.1.1.B). With a dynamic enzyme like luciferase, it was imperative to choose a structurally accurate input for modeling. Ideally, a starting structure is one where the conformation of the protein is conducive to ligand binding. Fortunately, such a structure existed in Fluc co-crystallized with luciferyl-AMP (DLSA), a high-energy intermediate analogue (PDB ID: 4g36) (Sundlov et al.,

2012). Other crystal structures of Fluc or its homologues crystallized without a ligand, were in protein conformations inconducive to binding.



**Figure 4.3.1.1** Model of the dynamic Fluc. (PDB ID: 4g36) (grey) with bound high energy intermediate DLSA (yellow sticks) and Fluc (PDB ID: 1lci) (salmon) without a bound ligand. (A) C-terminal residues are shown as spheres to illustrate the  $\sim 25$  Å domain movement upon ligand binding. (B) This domain movement causes the loop composed of residues 315-320 to move 3.3 Å, opening up the active site for productive ligand binding.

Initial attempts to prepare the Fluc scaffold for modeling followed a well-established protocol to remove all non-proteinaceous molecules prior to equilibrating the scaffold with the Rosetta score function. However, close inspection of the equilibrated models overlaid with the input crystal structure showed that without a ligand in the active site, loop 315-320 adopted the lower energy conformation observed in the unliganded crystal structures. This “closed conformation” of the loop precluded productive ligand binding. Therefore, preparation of the structure included the co-crystallized ligand in the active

site which constrained loop 315-320 to the open conformation. This small change was integral to all downstream applications.

Our collaborators in the Jennifer Prescher lab have worked extensively with luciferase and over the years they have developed libraries of starting scaffolds for initial hit generation. Often, to get their foot-in-the-door they begin their selection efforts by screening for the best starting scaffold from this set. The majority of the starting scaffolds were variants of Fluc and contained from 2-28 mutations to wild type. Initially, to prepare models of the starting luciferase variants, I made the mutations in the model scaffold using an input “resfile” as an initial step prior to sampling the binding mode of the novel ligand. However, as I optimized the platform, I changed the order of events so that the mutations were made during the design step. The reasoning behind this change was two-fold. First, preparing the Fluc scaffold once for docking and design cut out a step and made the platform less computationally time-intensive. Second, I reasoned that making the required mutations in the context of a bound ligand would provide a better view of the interplay between the scaffold and the ligand.

**4.3.2 Preparation of the Luciferin Analogues.** The second preparatory step was to generate the ligand model. This step went through quite a few iterations before I settled on the best method because there were numerous parameters that went into this decision. As previously mentioned, the luciferase chemical reaction involves two half reactions: adenylation of the luciferin followed by oxidation (Figure 4.2.1). The first decision on how to properly model the novel luciferin analogues was whether the luciferin should be modeled as its starting structure or as the reaction intermediate, an adenyated luciferin. Because the product molecule is structurally very similar to the starting molecule, it was

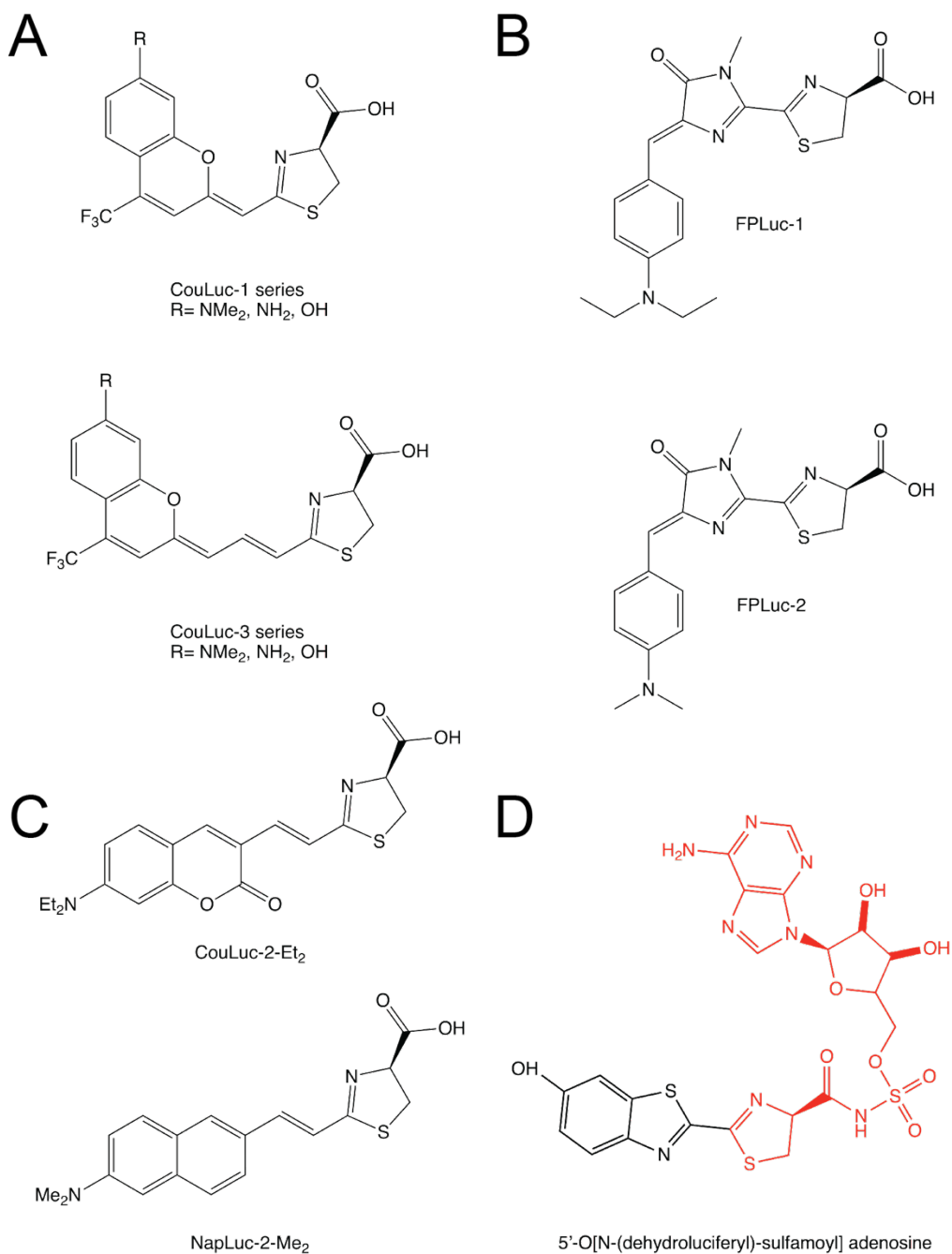
not a contender in this decision. Since current models of enzyme catalysis assert that the active site stabilizes the transition state intermediate. Therefore, modeling the luciferin analogue as a reaction intermediate was the highly favored option. Nevertheless, the starting structure of the luciferin analogues had a smaller chemical footprint and it was reasoned that it might be easier to dock the starting molecule as a first pass.

Computational experiments were performed where the luciferin analogue was docked as the starting structure. This approach was quickly abandoned because of the flexibility of the molecule. It was difficult to obtain docked models where the 4-carboxy-thiazoline ring of the analogue was in the native geometric orientation in the active site as identified by a structural study of the 4-carboxy-thiazoline ring on DLSA in the crystal structure. Therefore, the luciferin analogues were modeled as the intermediate adenylated luciferin form. To do this, the adenylated moiety of the DLSA structure from the co-crystallized Fluc model (Figure 4.3.2.1.D) was conjugated to the luciferin analogue to create a model of the luciferin intermediate. The adenylated moiety of the DLSA model thus served as a handle to anchor the luciferin analogues in the active site (Branchini et al., 2005).

The ligands were built in Avogadro an open-source molecular builder and visualization tool. Version 1.2.0. <http://avogadro.cc/> using the crystallized adenylated moiety of the DLSA model as the base (Hanwell et al., 2012). They were energy minimized using the UFF force field (Rappe et al., 1992) because it is a good general energy minimization algorithm. Another force field available in the Avogadro software, the Merck Molecular Force Field (MMFF94), was employed experimentally, but was found to be less than optimal. MMFF94 is a force field built to model a molecule in an aqueous environment (Halgren, 1996). Therefore, intramolecular interactions in the

molecule are upweighted. This caused the luciferin to take on a more compact structure during the energy minimization where interactions between the adenyly moiety and the heterocycles at the end of the pi wire are stronger. Post energy-minimization, the pi wire was more curved and no longer linear. Consequently, the luciferins subjected to the MMFF94 force field would not fit in the luciferase active site. I reasoned that because the luciferyl-AMP intermediate was only present in the protein active site environment, the ligand was less likely to be in the same geometric conformation as the MMFF94 predicted. The UFF force field minimization maintained the extended pi wire conformation that was more conducive to ligand binding. In all energy minimization events, the adenyly moiety, or the “handle” was always constrained from moving.

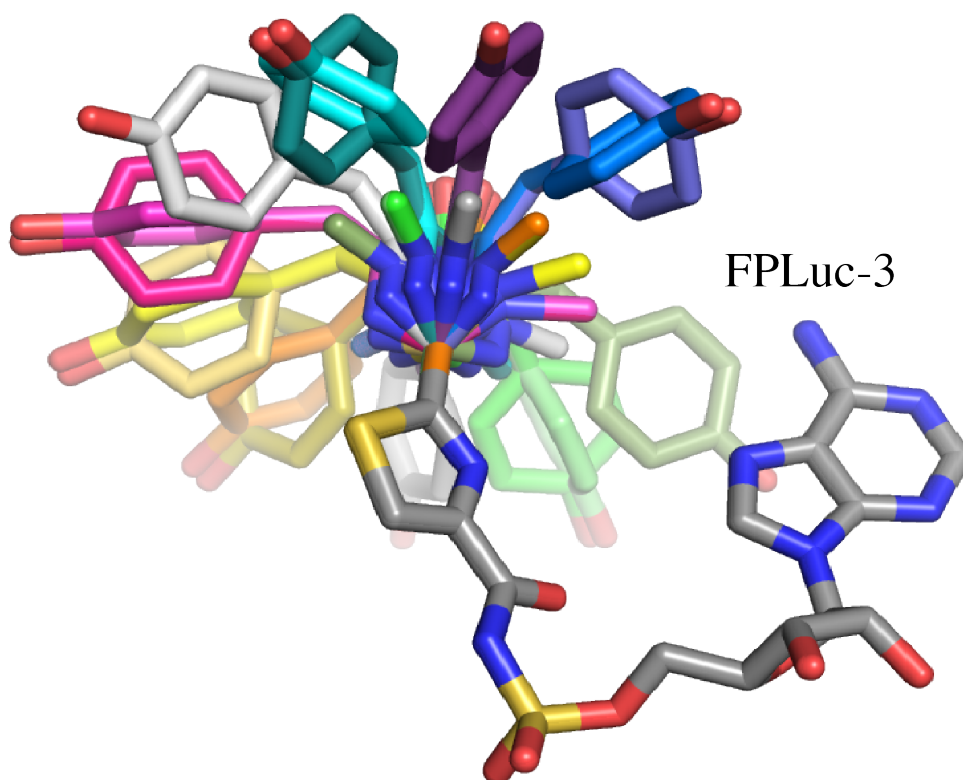




**Figure 4.3.2.1** Luciferin analogues that were incorporated into the rationally designed library protocol. (A) the CouLuc series built on the coumarin moiety. (B) the FPLuc series built on the red fluorescent protein chromophore (C) the NapLuc and CouLuc-2 series built around naphthalene and coumarin, respectively. (D) 5'-O[N-(dehydroLuciferyl)-sulfamoyl] adenosine (DLSA), an analogue of D-luciferin that was co-crystallized with Fluc (PDB ID: 4g36) and used as the base for building the luciferin

analogues. The red portion is the adenylyl moiety that was added to the novel luciferin analogues for computational modeling.

Originally, I performed a rigid docking of the luciferin analogues into the active site; relying on the Rosetta software to sample limited ligand torsion angles. However, I quickly discovered that this method was not conducive to identifying potential binding modes. When experimental data informed me that the one of the luciferins that I had previously believed impossible to dock into the Fluc scaffold generated light, I knew that my method needed to be improved. I increased the ligand sampling space by incorporating a file of enumerated ligand conformations generated using the OpenEye Omega software (Hawkins et al., 2010a). To limit the number of luciferin conformers in the library to only those that would be productive, I constrained the torsion angles in the handle of the ligand composed of the adenylyl moiety to the carbon past the thiazoline ring to the angles calculated from the crystal structure. Only the portion of the luciferin analogue that was composed of the starting molecule was allowed to sample torsional space (Figure 4.3.2.2). The inclusion of a ligand conformer library was integral to identifying potential binding modes of the analogues.



**Figure 4.3.2.2** Sample conformer library for FPLuc-3. For conformer generation, the torsion angles of the adenyl moiety (modeled as grey sticks) were constrained to the observed angles in the Fluc crystal structure (PDB ID: 4g36). The portion making up the starting molecule near the thiazoline ring were allowed to sample torsion space (shown as colorful sticks).

**4.3.3 Docking the Ligands and Sculpting the Luciferase Active Site.** To determine the native-like geometric alignment of D-luciferin in Fluc, a structural analysis of two crystal structures was performed. The first was one of a luciferase from the Japanese Genji-botaru (*Luciola cruciate*), a homologue of Fluc, co-crystallized with products AMP and oxyluciferin (PDB ID: 2d1r) (Nakatsu et al., 2006). The second was of Fluc co-crystallized with a luciferyl-AMP intermediate analogue, DLSA, (PDB ID: 4g36) (Sundlov et al., 2012). It was discovered that the carbonyl proximal to the thiazoline ring that reacts with ATP to form the luciferyl-AMP intermediate was within

3.4 Å of C $\alpha$  G315 on the 315-320 loop. Therefore, the C $\alpha$  of G315 was used as an anchor point for docking the luciferins in a native-like geometry. This measurement was incorporated into the constraints imposed on the RosettaMatch algorithm to properly dock the novel luciferins into the Fluc scaffold.

The RosettaMatch algorithm was employed to dock each novel luciferin in a native-like orientation in the prepared native Fluc structure (PDB ID: 4g36) (Tinberg & Khare 2017; Zanghellini et al., 2006). Docking was done in rounds with increasing weight placed on the constraints on ligand placement. The first round typically contained loose constraints to verify that the ligand could semi-productively bind in the active site. In subsequent rounds, the constraints were tightened until RosettaMatch could no longer find binding orientations for the luciferin. Docked models were chosen for design based on the closest adherence to the native binding mode as defined by 1) the distance of the carbonyl to loop 315-320 and 2) the angle between the C $\alpha$  of G315 and the carbonyl and proximal atoms on the luciferin analogue as calculated from the crystal structure.

Output files from the docking step were used as input to the RosettaDesign algorithm. RosettaDesign was used to randomly mutate residues near the docked luciferin analogue. The viability of each mutation was assessed using the Rosetta score function (Alford et al., 2017). Mutations that alleviated clashes between ligand atoms and protein atoms or that introduced new interactions (e.g., hydrogen bonds or hydrophobic packing interactions) between the ligand and protein typically improved the Rosetta energy. In this way, Rosetta was used to restructure the active site to accommodate novel ligands that initially were structural misfits.

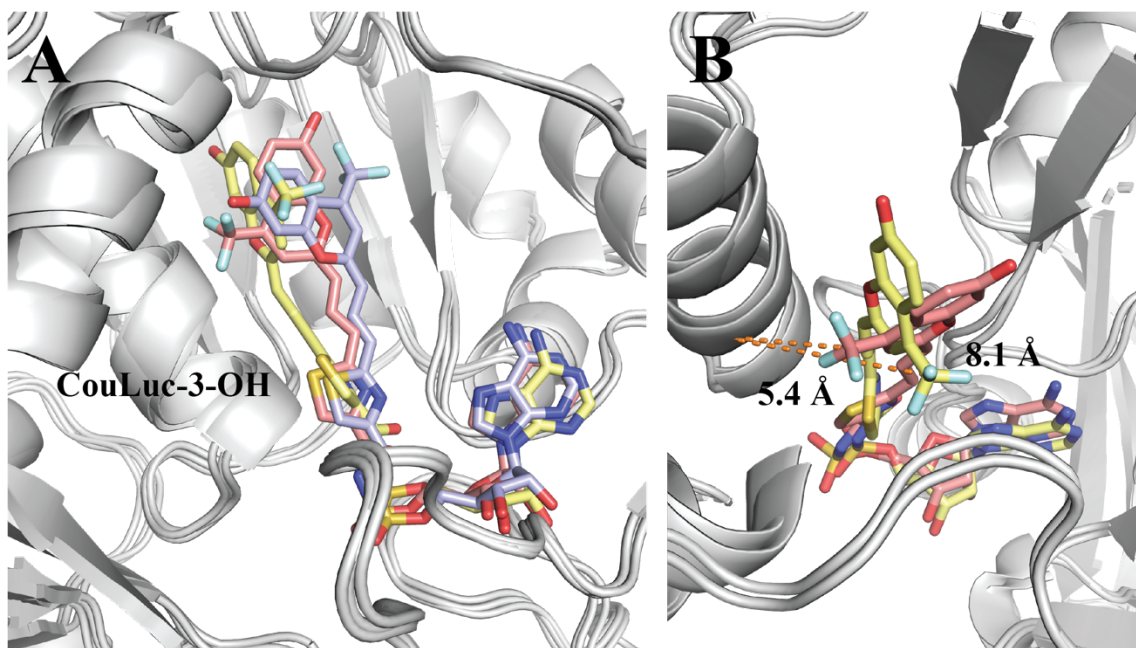
**4.3.4 Analyzing the Output Models.** Extracting data from the RosettaDesign outputs was where the most interesting platform development occurred. Rosetta proper is equipped with built-in methods to analyze potential output during the design process itself. These methods, called “filters,” are typically a robust way to eliminate models that do not meet specific criteria (Fleishman et al., 2011). The benefit of filters is that they eliminate all models that do not pass certain criteria on the fly, so they are never outputted. This is desirable when an appropriate filter exists, but in this case, one did not. Thus, the analysis of the Rosetta designs took place post-design.

The data that I wanted to extract from the RosettaDesign output was what locations did Rosetta mutate and how often that particular location was chosen in the context of ligand positioning in the active site. Since this analysis was integral to identifying initial target locations for a NNK library, the residue identity of the mutation was unimportant. The output pdb files from RosettaDesign were analyzed using code collected into Jupyter notebooks increasing both the reproducibility of the analysis and allowing it to be high throughput.

Each docked input structure, of which there were typically 10-15, were very similar with only slight variations in the torsion angles and geometric positioning of the ligand gave rise to ~10-25 output designs. Often the ligand placement in the output designs was similar to the input structure, but this was not a hard-and-fast rule. Rosetta typically performed rigid-body ligand translations during the design and side chain repacking steps. The variability of the Rosetta Monte Carlo algorithm made it a challenge to analyze the output designs in a high-throughput way. This challenge was overcome by binning the output files into sets based on observed ligand conformation in the active site.

This ligand-centric approach allowed for a clearer view of how the different ligand conformations affected the mutation locations chosen by Rosetta.

To bin the models, the distance was calculated between an atom on the heterocycle of the luciferin analogue and an  $C\alpha$  on a residue in the active site in proximity of the heterocycle atom. (Figure 4.3.4.1)



**Figure 4.3.4.1** Select Rosetta models of CouLuc-3-OH bound in the active site of Fluc. (A) CouLuc-3-OH adopts multiple conformations in the active site (B) To identify how the conformation of the luciferin analogue affects the locations mutated by Rosetta, the models were binned by the distance between a backbone atom in the active site and an atom on the ligand. There are two bins represented in this example: 5 Å and 8 Å.

The calculated distance was rounded up to the nearest integer and that integer was the bin. Typically, one RosettaDesign run contained from 4-8 bins with ~1-50 files in each bin. Within each bin, the locations where Rosetta made a mutation (relative to the input scaffold sequence) were identified and the frequency of each mutation location was

calculated. In addition, the frequency of each mutation location was also calculated for all the models regardless of bin status. This analysis provided a global view of the number of mutations as well as a very specific view by ligand binding mode.

## 4.4 Results and Discussion

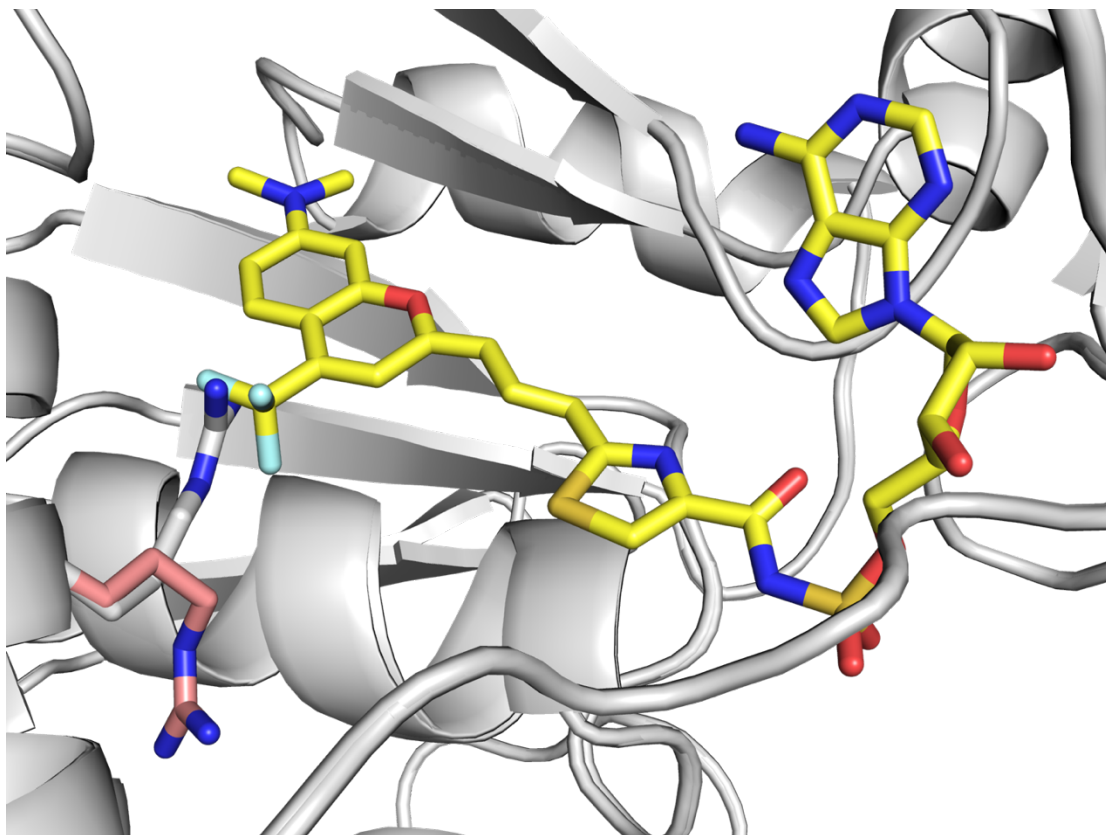
**4.4.1 The CouLuc-3 Series as a Case Study** Five novel luciferase-luciferin pairs with robust red-shifted photon output were identified using the data generated from the luciferase platform. To assess the efficacy of the platform, the CouLuc-3 series of luciferin analogues can be used as a case study. The campaign to identify a Fluc variant that would accommodate the coumarin derived luciferins (Figure 4.3.2.1.A) was particularly robust and involved iterative rounds of design-experimentation. The CouLuc-3 series includes three luciferins that differed in the electron-donating groups (-NMe<sub>2</sub>, -NH<sub>2</sub>, and -OH) installed on the coumarin heterocycle. These groups differ significantly in size from nine atoms in -NMe<sub>2</sub> to three atoms in NH<sub>2</sub> and two in -OH. When the light output of the three luciferins was measured with native Fluc, photon output of CouLuc-3-NMe<sub>2</sub> and CouLuc-3-NH<sub>2</sub> was too dim to be measured. However, CouLuc-3-OH was found to have the brightest emission maximum and a  $\lambda_{\text{max}}$  of 730 nm. I was reasoned that CouLuc-3-OH, with its smaller chemical footprint was better accommodated in the active site than the larger analogues in the series. However, the starting point of the emission maximum of CouLuc-3-OH was insufficient for imaging applications. Experimental efforts focused on developing Fluc variants for CouLuc-3-OH and CouLuc-3-NMe<sub>2</sub> because these two molecules were successfully synthesized in high yields while CouLuc-3-NH<sub>2</sub> was not. Additionally, it was reasoned that Fluc variants that could process the

smaller CouLuc-3-OH or the larger CouLuc-3-NMe<sub>2</sub> would likely be able to process CouLuc-3-NH<sub>2</sub> sufficiently well for imaging applications.

To engineer more robust luciferase/CouLuc-3-OH and -NMe<sub>2</sub> pairs, the computational luciferase platform was employed iteratively with input from experimental data. To identify target locations for the first generation RosettaDesign library, the CouLuc-3 luciferases were docked into wild type Fluc. Based on the docking output, it became immediately apparent that there was a significant steric clash between the sidechain of R218 and both luciferins because every design included a mutation to a smaller residue (Figure 4.4.1.1). When Rosetta was prohibited from mutating R218, it adopted a high-energy conformation. Therefore, this location was included in the first generation RosettaDesign library. Along with R218, there were 40 more locations that were identified as potential targets in the computational analysis. Most of these locations were in the active site, but there were also a few mutations in the AMP-binding domain. With respect to target locations in the AMP-binding domain, it was discovered after five rounds CCM libraries that none of the targets in the AMP binding region led to any improvements in CouLuc-3 processing. Therefore, future campaigns omitted this region in library design.

Our collaborators chose 20 locations to target for the RosettaDesign generation 1 library (Figure 4.4.1.A).

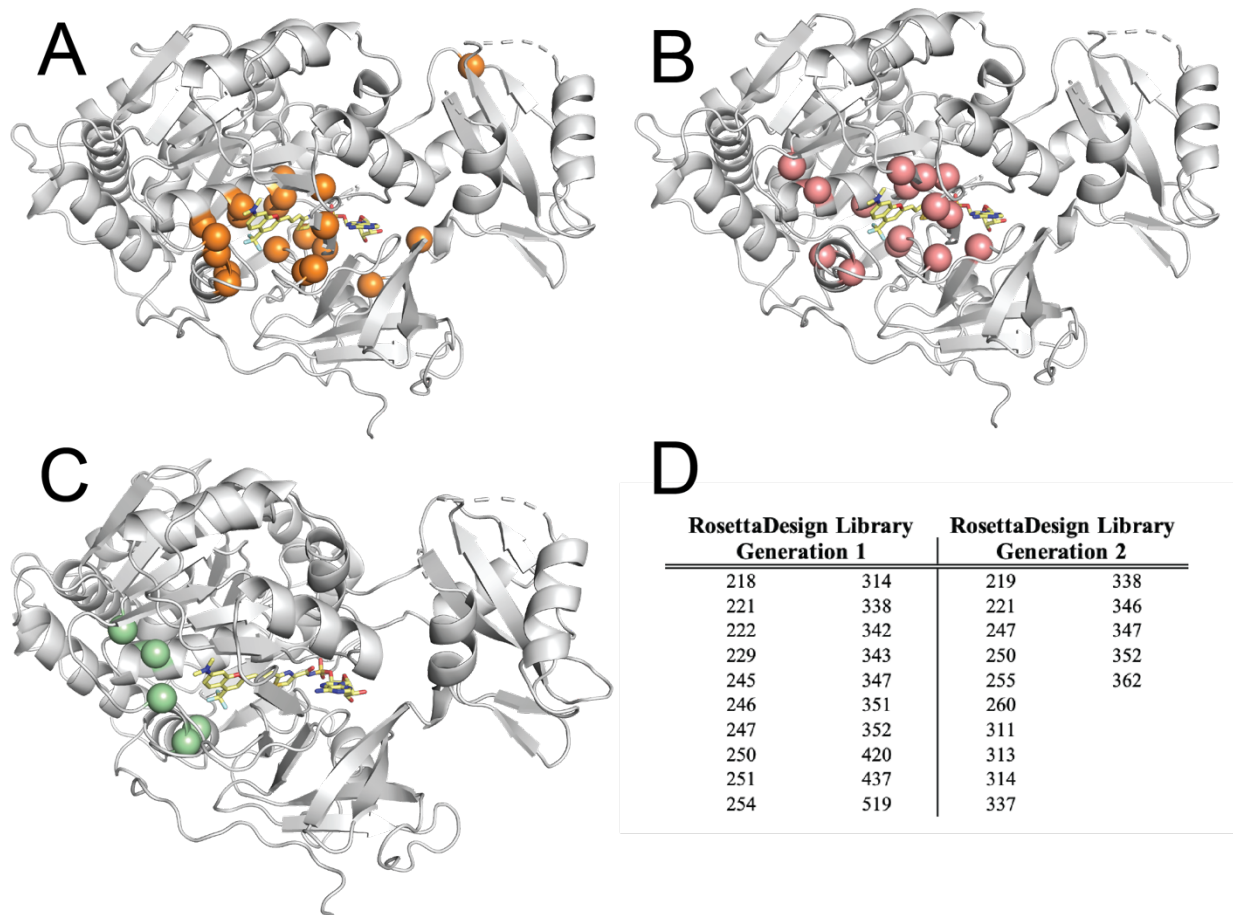




**Figure 4.4.1.1** A model of CouLuc-3-NMe<sub>2</sub> bound in the Fluc active site. R218 (grey) is in the conformation modeled in the crystal structure of PDB ID: 4g36. R218 (salmon) adopts a high energy conformation when CouLuc-3-NMe<sub>2</sub> is bound.

The brightest “hits” from the first generation RosettaDesign library contained the following mutations: R218S/V and N229F. However, the photon output was not as robust as desired, therefore a second round of modeling was undertaken using the hits from the first generation library as the starting scaffolds. Three scaffolds were prepared: R218S alone; R218S with N229F; and R218V with N229F. The second round of target hunting identified another ~ 30 potential target locations. Most of the locations were also found in the first set. However, a few new locations including two that would end up being important to CouLuc-3 processing appeared in low frequency. The first important location was F260 which was mutated 5% of the time when CouLuc3-NMe<sub>2</sub> was bound

and 2% of the time when CouLuc3-OH was bound. The second was Y255 which was mutated in 14% of the designs regardless of luciferin analogue bound. These locations along with 7 locations from the first-generation library and six new locations made up the second generation RosettaDesign library (Figure 4.4.1.2.B). Top hits from this library identified H221, Y255 and F260 as being important to CouLuc3 processing.



**Figure 4.4.1.2.** Locations chosen for the RosettaDesign libraries. (A) Locations chosen for the first generation (B) Locations chosen for the second generation (C) Locations that were identified as “hits” and (D) a table of the mutations chosen.

The first generation library included three hits that were focused on mutations at the following locations: R218, N229 and H221, all of which were Rosetta “hot spots.”

The second generation RosettaDesign library identified two more target locations: F260 and Y255. In an effort to increase emission output, two more error-prone PCR and random mutagenesis libraries were designed and screened. However, they failed to produce luciferase variants with emission output greater than those identified in the two RosettaDesign libraries.

The final output of the rounds of library screenings revealed two mutants with > 650 nm emission maxima for all three CouLuc-3 analogues and a total flux of 380 fold and 9 fold greater than wild type Fluc when processing CouLuc-3-NMe<sub>2</sub> and CouLuc-3-OH, respectively. The first mutant was named “Pistachio” after a Prescher lab tradition to name successful mutants after nuts. The second, was sadly neglected and kept its original library number of “709”. Pistachio contained the following mutations: R218S, H221I, N229F, and F260G. Unnamed mutant 709 contained the following mutations: R218V, N229F, Y255S, F260R. (Love et al., 2023).

Retrospectively analyzing the Rosetta-predicted hotspots with the mutations contained in the final mutants is an important step to improving the platform. Experimental data indicated that only five total mutations were necessary to process the CouLuc-3 analogues with sufficient photon output to be useful in biomedical applications. The mutation locations that were important to process all three luciferin analogues were R218, N229 and F260. One mutation was specific to the processing of CouLuc-3-NMe<sub>2</sub>: H221. The mutation specific to CouLuc-3-OH processing was Y255. All five locations appeared in the Rosetta-guided library. Three of them, R218, H221, and N229 were predicted as targets in 100%, 73% and 54% of the models, respectively. Y255 and F260 were only predicted to be targets in 14% and 2% of the models,

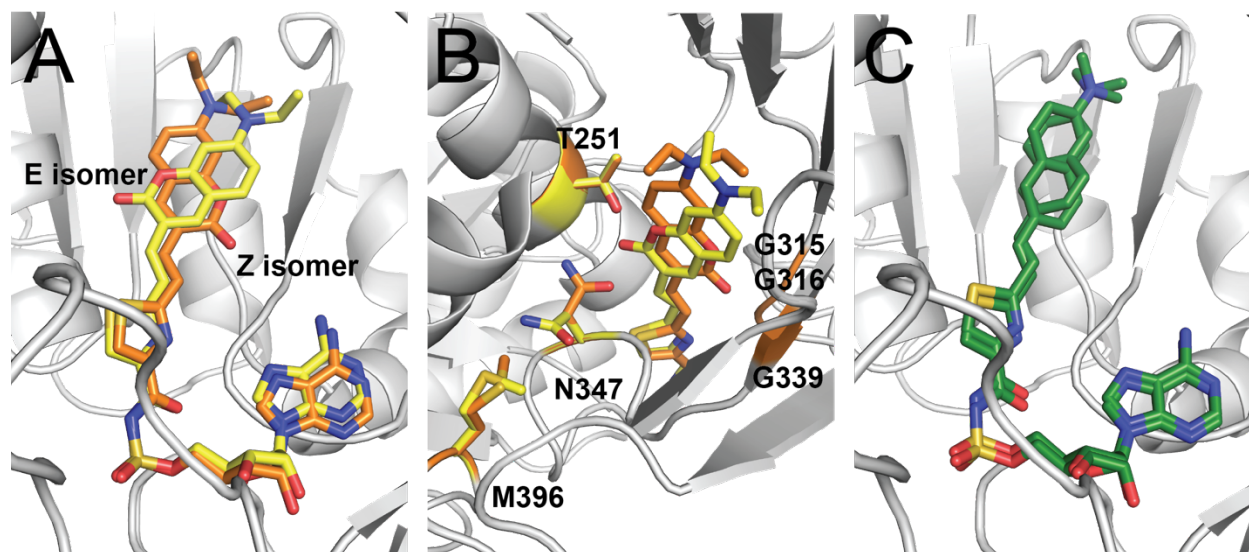
respectively. However, Rosetta predicted 36 more “hot spots”. This large signal-to-noise ratio provides ample room for platform improvement that would likely be made possible with tighter constraints during the RosettaDesign step and on increasing the number of outputs from RosettaDesign.

**4.4.2 Modeling Experimental Data with Rosetta** Frequently, my collaborators would be baffled by an experimental result and I would be asked to create a model of the protein-ligand interaction to better understand the data. Of course, solving the crystal structure of the holo protein, or undertaking more experiments is the best way to develop hypothesis to explain experimental data, but I found that using Rosetta to create models provided context and was a useful way to generate hypothesis. There are three examples of opportunities where I used Rosetta to generate hypothesis for experimental data. In this section, I will go into detail on one of them. The other two examples can be found in the following publications from the Leconte lab and the Prescher lab (Love et al., 2023; Williams et al., 2023).

One of the questions that our collaborators wanted to answer was why the light output was so much more robust when NapLuc-2-NMe<sub>2</sub> was bound over when CouLuc-2-NEt<sub>2</sub> was bound in Akaluc. Akaluc is an engineered luciferase from the Miyawaki lab with 28 mutations from wild type Fluc, most of which are outside the active site (Iwano et al., 2018). Akaluc is one of the scaffolds that the Prescher lab routinely uses as a starting point for their engineering efforts.

Two luciferins that were synthesized by the Martin Schnermann lab: CouLuc-2-NEt<sub>2</sub> and NapLuc-2-NMe<sub>2</sub> were found to have differences in light output when processed by Akaluc even though they are structurally very similar. (Figure 4.3.2.1.C). In Akaluc,

CouLuc-2-NEt<sub>2</sub> had 10<sup>7.5</sup> p/s total flux and NapLuc-2-NMe<sub>2</sub> had 10<sup>10</sup> p/s total flux. We hypothesized that the increased steric bulk of the diethylamino substituent in CouLuc-2-NEt<sub>2</sub> (relative to the dimethylamino substituent in NapLuc-2-NMe<sub>2</sub>), or the carbonyl on the pyrone ring of the coumarin moiety might have precluded productive binding of CouLuc-2-NEt<sub>2</sub> within the active site. To explore this, we first generated *in silico* models of NapLuc-2-NMe<sub>2</sub> and CouLuc-2-NEt<sub>2</sub>, using Avogadro and then generated libraries of low-energy conformers of each of these substrates using OMEGA 4.2.1.2: OpenEye Scientific Software, Santa Fe, NM (Hanwell et al., 2012; Hawkins et al., 2010b). Because no structure of Akaluc has been reported, the conformer libraries were docked into a structure of Fluc (PDB ID: 4g36) using the RosettaMatch algorithm; user-defined constraints were employed in this docking step to ensure a native-like binding geometry was achieved. The Rosetta CoupledMoves algorithm was used to 1) introduce the 28 mutations present in Akaluc, 2) sample low-energy ligand conformations, 3) sample low-energy side chain conformations, and 4) sample backbone torsions to alleviate any clashes between the ligand and the protein (Ollikainen et al., 2015). The resulting models were then analyzed to assess whether differences in the binding modes of the two ligands were present. In our models, the bound CouLuc-2-NEt<sub>2</sub> substrate was observed in both the E and Z isomeric forms (Figure 4.4.2.1.A). When the Z isomer is present, the ligand is well accommodated within the active site; namely, the carbonyl on the pyrone ring is observed in a glycine-rich microenvironment in which the C<sub>α</sub>s of G315, G316, and G339 fall within ~3.4-6 Å of the carbonyl oxygen atom (Figure 4.4.2.1.B).



**Figure 4.4.2.1** (A) The lowest scoring Rosetta models of E (yellow) and Z (orange) isomers of CouLuc-2-NEt<sub>2</sub> and bound in the Akaluc active site. (B) The Z isomer is well accommodated in the active site as the carbonyl on the pyrone ring is in a glycine-rich microenvironment with G315, G316 and G339 within 3.4-6 Å of the carbonyl oxygen atom. The CouLuc-2-NEt<sub>2</sub> E isomer is not well accommodated in the active site. In 70% of the models, the carbonyl on the pyrone ring is observed within 2.1 Å of O<sub>γ</sub> of T251. The side chain of N347 must also rotate ~150° about the C<sub>α</sub> and C<sub>β</sub> bond toward the side chain of M396 to accommodate the ligand, forcing it into a high energy conformer. In 30% of the models, the distance between the O<sub>δ</sub> of N347 and the C<sub>ε</sub> of M396 is < 3 Å. (C) Two low scoring Rosetta models of NapLuc-2-NMe<sub>2</sub> bound in the Akaluc active site.

However, when the E isomer is bound, the carbonyl on the pyrone ring is not as well accommodated in the pocket. In 70% of the models, the carbonyl on CouLuc-2-NEt<sub>2</sub> is observed within 2.1 Å of O<sub>γ</sub> of T251. In 30% of the models, the carbonyl is between 3-4.5 Å of the O<sub>γ</sub> of T251, but the side chain of N347 must rotate ~150° about the C<sub>α</sub> and C<sub>β</sub> bond toward the side chain of M396 to accommodate the ligand. In 30% of the models with the E isomer bound, the distance between the O<sub>δ</sub> of N347 and the C<sub>ε</sub> of M396 is < 3 Å. In addition, the Rosetta scores of N347 and M396 are 5.3 and 0.47 REU, respectively which indicates that these residues are higher in energy and could preclude proper

binding of the ligand. The proximity of the carbonyl group on CouLuc-2-NEt<sub>2</sub> to T251 combined with the proximity of N347 to M396, and the high energy conformation of N347 could explain the differences in the emissive output of Akaluc/CouLuc-2-NEt<sub>2</sub> and Akaluc/NapLuc-2-NMe<sub>2</sub>. However, further structural studies are warranted to verify this conclusion.

#### **4.5 Conclusions**

The use of Rosetta to identify target locations for semi-rational design is a novel technique that has the potential to reduce the number of rounds required for library screening. In the case outlined in this chapter, the two generations of RosettaDesign libraries identified all the necessary mutations to robustly process the luciferin analogues. However, the signal-to-noise ratio in the “hot spot” locations identified by Rosetta was high and leaves room for improvements to the platform. These improvement would likely be found in 1) increasing the constraints of the RosettaDesign step and 2) more exhaustively sampling sequence space through larger design runs.

Using Rosetta to create models to explain experimental results is another novel methodology that proved to be useful in multiple cases. The most notable was the case of identifying hypothesis for the differences in Akaluc’s processing of CouLuc-2-NEt<sub>2</sub> and NapLuc-2-NMe<sub>2</sub>. One of the keys to success in this arena is a very careful analysis of output models combined with qualitative assessments of said models. However, it must be noted that the hypothesis generated using Rosetta models can only be verified with experimental studies involving structure like crystallography or in some cases, NMR studies. Rosetta models should be used only as a way to generate hypothesis and not to form definite conclusions.

## 4.6 Acknowledgements

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## CHAPTER 5

### FINAL REMARKS

The world of protein design has undergone a tremendous revolution during the period this work was performed. Five years ago, machine learning was a nascent field that had not yet ventured into the protein design space, and the Rosetta software suite dominated. On July 15, 2021, with a massive bang, Google's Deep Mind introduced AlphaFold 2 to the world via *Nature* publishing. That same week, the Baker lab introduced RoseTTAFold via *Science*. To say that this week was a one-two-punch to the tools available to protein designers is a dramatic understatement; the week of July 15, 2021 was a paradigm-shift. Amino acid sequences without structural data were now able to be folded *in silico* with reasonable accuracy. The days of tedious homology modeling to predict protein structure were instantly over. Machine learning was now deeply embedded into the protein design space. Additionally, these tools were made open source, so anyone with a modicum of coding experience could not only use them, but adapt and hack them to address novel challenges.

Since 2021, Rosetta Commons developers, who were previously writing in C++ for the behemoth that is the Rosetta codebase, started shifting away from C++ and toward PyTorch and TensorFlow. In only two years, they developed and open-sourced game-changing machine learning packages like trRosetta, masif (molecular surface interaction fingerprints), ProteinMPNN, RFDiffusion, RFDiffusion-Ligand, and RoseTTAFold-All-Atom. Now, rational protein designers have the ability to create their own scaffold backbones, identify sequences that are likely to adopt the imagined backbone, and calculate the probability of obtaining soluble protein. Amusingly, the Rosetta design suite

of five years ago has been dubbed “Classic Rosetta” in order to differentiate it from the plethora of machine learning-based methodologies associated with the Rosetta Commons.

For the enormous leaps of progress made in the protein design space thanks to machine learning, there are still numerous outstanding challenges. One of these challenges is enzyme design. Recently, Yeh et al. used machine learning methods to design five *de novo* luciferases with function on par to native luciferases (Yeh et al., 2023) without the need for rounds of directed evolution to optimize them. This was an incredible achievement, but the reported success rate for the two rounds of design was 3/7648 and 2/46. This means that between the two rounds of method development, they went from a 0.04% success rate to a 4% success rate. The challenge of designing a well-folded functional enzyme has hardly been solved.

Thus far, machine learning tools have not yet cracked the code to decipher how allosteric interactions affect protein function, and they are unlikely to do so until there are more high-quality site saturation mutagenesis and pairwise mutation datasets available. Unfortunately, these types of data sets are incredibly time consuming and expensive to obtain. Another challenge is that once a large enough data set is obtained, there is no guaranteeing the rules that apply to one family of enzymes will be universal, as reaction mechanisms vary wildly from family to family.

The work presented in this dissertation focused on using “Classic Rosetta” to study the relationship between enzyme substrate specificity and global dynamics. This included a novel approach to targeting residues believed to be important to global dynamics in an effort to alter enzyme function. Moreover, platform development on how to mold an enzyme active site to accommodate novel ligands was undertaken. As

comprehensive as I tried to be, the work presented here is just the beginning, and although I successfully altered the specificity for Fluc by identifying mutations important to remodeling the active site to accommodate larger luciferins, I was unsuccessful in all attempts to alter substrate specificity by targeting allosteric locations distal to the active site in TEM-1. Understanding how allostery and enzyme function are related is an ongoing challenge. I look forward to discovering new ways of approaching the "dynamics mystery" as the field continues to generate data and develop new hypothesis.

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APPENDIX A  
ADDITIONAL PUBLICATIONS

In addition to the publications mentioned specifically in the chapters, I also contributed to the following projects. First, I did the molecular cloning, protein expression and purification of the streptavidin mutants in the following publication. Gleason, P.R., Kolbaba-Kartchner, B., Henderson, J. N., Stahl, E. P., Simmons, C. R., & Mills, J. H. (2021). Structural Origins of Altered Spectroscopic Properties upon Ligand Binding in Proteins Containing a Fluorescent Noncanonical Amino Acid. *Biochemistry (Easton)*, 60(34), 2577–2585.

Second, I used the RosettaDesign platform to model experimental data in the following publication. Williams, S., Gewing-Mullins, J. A., Lieberman, W. K., Kolbaba-Kartchner, B., Iqbal, R., Burgess, H. M., Colee, C. M., Ornelas, M. Y., Reid-McLaughlin, E. S., Mills, J. H., Prescher, J. A., & Leconte, A. M. (2023). Biochemical Analysis Leads to Improved Orthogonal Bioluminescent Tools. *Chembiochem : a European Journal of Chemical Biology*, 24(6), e202200726–n/a.

Third, I used the RosettaDesign platform to identify target locations for CCM luciferase library design in the following publication. Yao, Z., Caldwell, D. R., Love, A. C., Kolbaba-Kartchner, B., Mills, J. H., Schnermann, M. J., & Prescher, J. A. (2021). Coumarin luciferins and mutant luciferases for robust multi-component bioluminescence imaging. *Chemical Science (Cambridge)*, 12(35), 11684–11691.

Fourth, I used the RosettaDesign platform to model experimental data in the following publication which is under review at *The Journal of Organic Chemistry* at the time of this writing. Caldwell, Donald; Townsend, Katherine; Kolbaba-Kartchner, Bethany; Hadjian, Tanya; Ivanic, Joseph; Mills, Jeremy; Prescher, Jennifer; Schnermann, Martin. (2023). Expedient Synthesis and Characterization of  $\pi$ -Extended Luciferins. *The*



APPENDIX B

GENERAL ROSETTA SCRIPTS FROM CHAPTER 2



All calculations were carried out using Rosetta version:  
442bff4fb7bf2ccb44655e8d15276c9bccfbbd0.

The following command line was used to minimize the total energy of the 1btl crystal structure from the Protein Data Bank using the Rosetta relax protocol:

```
/Rosetta/main/source/bin/relax.default.linuxgccrelease -s pdb_file -
nstruct 1 -relax:default_repeats 5 -
relax:constrain_relax_to_start_coords -
relax:coord_constrain_sidechains -relax:ramp_constraints false -ex1
-ex2 -use_input_sc -flip_HNQ -ignore_unrecognized_res -
relax:coord_cst_stdev 0.5
```

The DesignAround protocol was initiated with the following command line:

```
/Rosetta/main/source/bin/rosetta_scripts.linuxgccrelease -
out:nstruct 25 -jd2:ntrials 50 -parser:protocol /design.xml -
packing:resfile [resfile] -database /Rosetta/main/database -
out::overwrite -s pdb_file -run:preserve_header -output_virtual true
-use_input_sc -no_his_his_pairE -score::hbond_params correct_params
-lj_hbond_hdis 1.75 -lj_hbond_OH_donor_dis 2.6 -linmem_ig 10 -
nblast_autoupdate true -in:ignore_unrecognized_res -out::overwrite
And the contents of design.xml was: "</> " resnums="" repack_shell=""
allow_design="1" resnums_allow_design="0"/>
```

The content of the resfile was:

```
ALLAA
EX 1
EX 2
USE_INPUT_SC
Start
2 A PIKAA P
19 A PIKAA V #rigid resi
20 A PIKAA G #rigid resi
37 A PIKAA P
42 A PIKAA P
45 A PIKAA S #Active site
48 A PIKAA K #Active site
51 A PIKAA L #rigid resi
82 A PIKAA P
97 A PIKAA L #rigid resi
105 A PIKAA S #Active site
107 A PIKAA N #Active site
120 A PIKAA P
141 A PIKAA E #Active site
142 A PIKAA P #This proline is important for folding stability
149 A PIKAA P
158 A PIKAA P
209 A PIKAA K #Active site
218 A PIKAA R #Active site
```

230 A PIKAA P  
235 A PIKAA V #rigid resi

Sequences of Designed Proteins in FASTA format

>Native  $\beta$ -lactamase signal peptide MSIQHFRVALIPFFAAFCLPVFA

>Rdg262a

HPETLVKVKDAEDQLGARVGFQLTDLNSGKILEYFRAEERFPMMSTFKVLLCGA  
VLSRIDAGQEQLGRRIHYSQNDLVEYSPVTEKHLTDGMTVRELCSAAITMSDNT  
AANLLTTIGGPKELTAFLHNMGDHSVTRLDRWEPELNEAIPNDERDTTQPKAMA  
QTLRKLTTGELLTLASRQQLIDWMEADKVAGPLLRSLPAGWFIACKSGAGERG  
SRGIIAALGPDGKPSRIVVIFTTGSQATMDERNRQIAEIGASLIKHW

>Rdg262b

HPETLVKVKDAEDQLGARVGFILLDLNSGKILESFRAEERFPMMSTFKVLLCGAV  
LSRIDAGQEQLGRRIHYSQNDLVEYSPVTEKHLTDGMTVRELCSAAITMSDNTA  
ANLLTTIGGPKELTAFLHNMGDHSVTRLDRWEPELNEAIPNDERDTTTTPRAMAT  
TLRKLTTGELLTLASRQQLIDWMEADKVAGPLLRSLPAGWFIADKSGAGERGS  
RGQIAALGPDGKPSRIVVIMTTGSQATMDERNRQIAEIGASLIKHW

>Rdg44a

HPETLVKVKDAVDQLGAPVGMIELDLNSGKILESYNPEERFPMMSTFKVLLCGV  
LSRIDAGQEQLGRRIHYSQNDLVEYSPVTEKHLTDGMTVRELCSAAITMSDNTA  
ANLLTTIGGPKELTAFLHNMGDHSVTRLDRWEPELNEAIPNDERDTTMPVAMAT  
TLRKLTTGELLTLASRQQLIDWMEADKVAGPLLRSLPAGWFIADKSGAGERGS  
RGIIAALGPDGKPSRIVVIMMTGSQATMDERNRAIAEIGASLIKHW

>Rdg44b

HPETLVKVKKAVDLGLAPVGFIELDLNSGKILESYPPEERFPMMSTFKVLLCGAV  
LSRIDAGQEQLGRRIHYSQNDLVEYSPVTEKHLTDGMTVRELCSAAITMSDNTA  
ANLLTTIGGPKELTAFLHNMGDHSVTRLDRWEPELNEAIPNDERDTTMPVAMAT  
TLRKLTTGELLTLASRQQLIDWMEADVAGPLLRSLPAGWFIADKSGAGERGSR  
GIIAALGPDGKPSRIVVTMTSGSQATMDERNRAIAEIGASLIKHW

>Rdg44c

HPETLVVVKQAEDKLGARVGYIELDLNSGKILESFRPEERFPMMSTFKVLLCGAV  
LSRIDAGQEQLGRRIHYSQNDLVEYSPVTEKHLTDGMTVRELCSAAITMSDNTA  
ANLLTTIGGPKELTAFLHNMGDHSVTRLDRWEPELNEAIPNDERDTTMPVAMAT  
TLRKLTTGELLTLASRQQLIDWMEADKVAGPLLRSLPAGWFIADKSGAGERGSI  
GIIAALGPDGKPSRIVVIYATGSQATMDELNRAIAEIGASLIKHW

>Flx226a

HPETLVKVKDAEDQLGARVGYIELDLNSGKILESFRPEERFPMMSTFKVLLCGAV  
LSRIDAGQEQLGRRIHYSQNDLVEYSPVTEKHLTDGMTVRELCSAAITMSDNTA  
ANLLTTIGGPKELTAFLHNMGDHSVTRLDRWEPELNEAIPNDERDTTMPVAMAT

TLRKLLTGELLTLASRQQLIDWMEADKVAGPLLRSAIPAGWFIADKSGAGERGS  
RGIIAALGPNGKPSRIVVIYTTGSQATMDERNRQIAEIGASLFKHW

>Flx226b

HPETLVKVKDAEDQLGARVGYIELDLNSGKILESFRPEERFPMMSTFKVLLCGAV  
LSRIDAGQEQLGRRIHYSQNDLVEYSPVTEKHLTDGMTVRELCSAAITMSDNTA  
ANLLLTIGGPKELTAFLHNMGDHSVTRLDRWEPENEAIPNDERDITMPVAMAT  
TLRKLLTGELLTLASRQQLIDWMEADVAGPLLRSAIPPGWFIADKSGAGERGSR  
GIIAALGPNVGPTRIVVIYTTGSQATMDERNRQIAEIGASLFKHY

>Flx226c

HPETLVKVKDAEDQLGARVGYIELDLNSGKILESFRPEERFPMMSTFKVLLCGAV  
LSRIDAGQEQLGRRIHYSQNDLVEYSPVTEKHLTDGMTVRELCSAAITMSDNTA  
ANLLLTIGGPKELTAFLHNMGDHSVTRLDRWEPENEAIPNDERDITMPVAMATT  
LRKLLTGELLTLASRQQLIDWMEADKVAGPLLRSAIPPGWFIADKSGAGERGSR  
GIIAALGPNVPSRIVVIYTTGSQATMDERNRQIAEIGASLFKHW

>Flx256

HPETLVKVKDAEDQLGARVGYIELDLNSGKILESFRPEERFPMMSTFKVLLCGAV  
LSRIDAGQEQLGRRIHYSQNDLVEYSPVTEKHLTDGMTVRELCSAAITMSDNTA  
ANLLLTIGGPKELTAFLHNMGDHSVTRLDRWEPENEAIPNDERDITMPVAMAT  
TLRKLLTGELLTLASRQQLIDWMAADKVAGPLLRSAIPPGWFIADKSGAGERGS  
GIIASLGPNGKPSRIVVIYTTGSQATMDERNRQIAEIGASLIKHW

>Flx55

HPETLVKVKDAEDQLGARVGYILLDADSGKILEAFRPEERFPMMSTFKVLLCGA  
VLSRIDAGQEQLGRRIHYSQNDLVEYSPVTEKHLTDGMTVRELCSAAITMSDNT  
AANLLLTIGGPKELTAFLHNMGDHSVTRLDRWEPENEAIPNDERDITMPRAMA  
ETLRKLLLGELLTLASRQQLIDWMEADKVAGPLLRSAIPAGWFIADKSGAGERG  
SRGIIAMLGPDGKPSRIVVIYTTGSQATMDERNRQIAEIGASLIKHW

APPENDIX B  
GENERAL ROSETTA SCRIPTS FROM CHAPTER 5

All calculations were carried out using the Rosetta software suite (version 2021.29+master.d8f55669792) with the ref15 score function (Alford et al., 2017a; Leaver-Fay, Tyka, et al., 2011).

### **Preparing the scaffolds**

A high-resolution (2.62 Å) crystal structure of *Photinus pyralis* luciferase (PDB ID: 4g36) was processed to remove waters and chain B (Sundlov et al., 2012). The structure was subjected to rounds of energy minimization using Rosetta FastRelax with the native co-crystallized ligand, 5'-O-[N-(dehydrolyciferyl)-sulfamoyl] adenosine (SLU) because it was noted that removal of the ligand prior to “relaxing” the protein caused motion in a flexible loop in the active site, precluding productive ligand binding in subsequent modeling efforts (Khatib et al., 2011; Maguire et al., 2021; Nivón et al., 2013). The params file used to input the geometric coordinates of each ligand atom, along with the bonding patterns of the atoms for SLU can be found in Appendix D.

The script used for the FastRelax application is as follows.

```
<Pathto>/Rosetta/main/source/bin/relax.default.linuxgccrelease -s  
<input_file> -nstruct 5 -relax:default_repeats 5 -  
relax:constrain_relax_to_start_coords -ex1 -ex2 -use_input_sc -flip_HNQ  
-ignore_unrecognized_res -relax:coord_cst_stdev 0.3 -extra_res_fa  
SLU.params
```

The lowest energy model as calculated by the Rosetta score function with a root-mean-square-deviation (RMSD) of < 2.0 Å from the input crystal structure was chosen and the SLU molecule was removed for all downstream applications.

### **Preparing the ligands**

To predict the binding modes of the luciferin analogues, two models of each of the CouLuc-3 and CouLuc-2 ligands (corresponding to the Z and E isomers), one model of the NapLuc analogues, and one model of the AkaLumine analogues were built in Avogadro: an open-source molecular builder and visualization tool, version 1.2.0. <http://avogadro.cc/> using the nonhydrolyzable adenylyl moiety of SLU as the base. (Hanwell et al., 2012) Each molecule was subjected to an energy minimization in Avogadro using the UFF force field (Rappe et al., 1992) Ligand conformers of both copies of each ligand were generated using OMEGA 4.1.2.0 (OpenEye Scientific Software, Santa Fe, NM) (Hawkins et al., 2010a). During conformer generation, torsion angle sampling was limited to the extended pi system of the novel luciferins by prohibiting movement of the torsions on the adenylyl moiety. An example command line for generating conformer libraries of each ligand is as follows. The SMILES string varied slightly with each ligand.

```
oeomega pose -in <input.mol2> -out output.mol2 -ewindow 12 -fixsmarts  
"C(=O)NS(=O)(=O)OCC1OC(C(C1O)O)[N+1]=2C=3C(=NC=2)[C+1](N=CN=3)N" -  
fromCT false -enumNitrogen false -enumRing false -searchff mmff_Trunc  
-strictatomtyping false
```

The Omega output was a .mol2 file containing the conformers, which was subsequently converted into a .params file with the identities and geometries of each atom specified; a .pdb file containing the library of conformers using an internal Rosetta script was also generated. The params files used to model the luciferin analogues can be found in Appendix D.

### **General RosettaDesign Methods**

The prepared Fluc (PDB ID: 4g36) structure was used as an input to the RosettaMatch protocol, which was used to dock the novel luciferins into the protein

structure based on user-defined constraints (Richter et al., 2011)The following command

line was used to call the RosettaMatch application.

```
<Path to>/Rosetta/main/source/bin/match.linuxgccrelease -s <input_file>
-extra_res_fa <params file> -match:geometric_constraint_file
<constraint file> -match:lig_name <ligand 3 letter code>
-match:scaffold_active_site_residues_for_geomcsts <pos_file> -ex1 -ex2
-ex2aro -ex1aro -extrachi_cutoff 0 -use_input_sc true -database
<path_to_Rosetta>/database/ -match:filter_colliding_upstream_residues -
match:filter_upstream_downstream_collisions -
match:upstream_residue_collision_tolerance 0.95 -
match:updown_collision_tolerance 0.3 -match::bump_tolerance 0.3 -
match_grouper SameSequenceAndDSPositionGrouper -
match:grouper_downstream_rmsd 0.5 -match:euclid_bin_size 0.5 -
match:euler_bin_size 5.0 -output_format PDB -exclude_patches
N_acetylated -consolidate_matches 1 -output_matches_per_group 5 -
output_matchres_only false -enumerate_ligand_rotamers -
only_enumerate_non_match_redundant_ligand_rotamers -
out::file::output_virtual
```

The contents of the .pos file was as follows.

```
N_CST 1
1: 308
```

The contents of each constraint file varied slightly with the name of the ligand. An example constraint file for NapLuc-2-NMe<sub>2</sub> is as follows.

```
CST::BEGIN
NATIVE
  TEMPLATE::  ATOM_MAP: 1 atom_name: O1 C21 C8
  TEMPLATE::  ATOM_MAP: 1 residue3: NLA

  TEMPLATE::  ATOM_MAP: 2 atom_name: N CA C ,
  TEMPLATE::  ATOM_MAP: 2 residue1: G
  TEMPLATE::  ATOM_MAP: 2 is_backbone

  CONSTRAINT:: distanceAB:  4.30  1.50  80.0  1  1
  CONSTRAINT::   angle_A:  135.3   5.0  10.0  360.  1
  CONSTRAINT::   angle_B:   43.6   5.0  10.0  360.  1
  CONSTRAINT::  torsion_A:   10.7   5.0  10.0  360.  1
  CONSTRAINT::  torsion_AB: -160.7   5.0  10.0  360.  1
  CONSTRAINT::  torsion_B: -134.1   5.0  10.0  360.  1

  ALGORITHM_INFO:: match
#    SECONDARY_MATCH: DOWNSTREAM
  CHI_STRATEGY:: CHI 1 EX_THREE_THIRD_STEP_STDDEVS
  CHI_STRATEGY:: CHI 2 EX_THREE_THIRD_STEP_STDDEVS
  ALGORITHM_INFO::END
CST::END
```

Matcher output .pdb files were used as input for the Rosetta CoupledMoves algorithm where mutagenesis, side chain conformational sampling and backbone minimization was undertaken to sculpt the binding pocket to accommodate the new ligand by alleviating any clashes between the ligand and the protein side chains for the CouLuc-2/NapLuc modeling. The Rosetta CoupledMoves application was called with the following script. (Ollikainen et al., 2015)

```
/path/to/Rosetta/coupled_moves.linuxgccrelease -jd2:ntrials 100 -
out:level 200 -s input.pdb -packing:resfile akaluc.resfile -database
~/Rosetta/main/database -mute protocols.backrub.BackrubMover -
extra_res_fa LCG_dimethyl.params -extra_res_fa NLA.params -ex1 -ex2 -
extrachi_cutoff 0 -nstruct 10 -coupled_moves::mc_kt 0.6 -
coupled_moves::ntrials 1000 -coupled_moves::initial_repack true -
coupled_moves::ligand_mode true -coupled_moves::fix_backbone false -
coupled_moves::bias_sampling true -coupled_moves::bump_check true -
coupled_moves::ligand_weight 2.0
```

For the CouLuc-3 modeling, the Matcher output files were used as input for the RosettaDesign algorithm where the active site pocket was also sculpted through side chain conformational sampling. The RosettaDesign application was called with the following script.

```
<path_to_Rosetta>/main/source/bin/rosetta_scripts.linuxgccrelease -
parser:script_vars ligand -out:nstruct 1 -extra_res_fa -
match:geometric_constraint_file -match:lig_name -jd2:ntrials 100 -
parser:protocol <xml file> -
match:scaffold_active_site_residues_for_geomcsts -in:file:native
4g36.pdb -database /main/database -s <input file from Matcher> -
run::preserve_header -unmute protocols.enzdes.EnzRepackMinimize -
enzdes::detect_design_interface -enzdes::cst_design -
enzdes::design_min_cycles 10 #The number of design-minimization
iterations -enzdes::minimize_ligand_torsions 5.0 -
enzdes::lig_packer_weight 1 -enzdes::cst_min -enzdes::bb_min -
enzdes:chi_min -packing::use_input_sc -packing::extrachi_cutoff 1 -
packing::ex1 -packing::ex2 -packing::soft_rep_design -linmem_ig 10 -
in:ignore_unrecognized_res -jd2:enzdes_out -nblast_autoupdate -
enzdes::bb_min_allowed_dev 0.05 -no_his_his_pairE
```

The contents of the RosettaDesign xml file were as follows.



```

<ROSETTASCRIPTS>

  <SCOREFXNS>
    <ScoreFunction name="ref15" weights="ref2015.wts"/>
    <ScoreFunction name="soft" weights="ref2015_soft.wts"/> #ref2015
recommended for protein repacking
    <ScoreFunction name="soft_cst" weights="ref2015_soft_cst.wts"/>
#soft-rep with constraints weights
    <ScoreFunction name="ref15_cst" weights="ref2015_cst.wts"/>
#hard-rep with constraints
  </SCOREFXNS>

  <RESIDUE_SELECTORS>
    <Index name="interface"
resnums="76,100,101,102,103,105,106,110,113,114,115,117,118,192,193,194
,195,196,197,198,199,210,213,214,219,220,221,229,233,234,236,237,238,23
9,243,246-
247,250,251,252,276,277,278,280,281,282,303,305,306,307,308,309,310,311
,312,313,316,317,320,326,327,328,239,330,332,332,333,334,335,336,338,33
9,340,343,344,345,348,349,350,351,352,353,354,355,414,423,426,428,429,4
31,433,434,435,437,438,439,440,441,442,445,446,449,469,470,471,498,499,
500,501,502"/>
    <ResidueName name="lig" residue_names="%%ligand%%"
residue_name3="%%ligand%%"/>
  </RESIDUE_SELECTORS>

  <FILTERS>
    <EnzScore name="allcst" score_type="cstE" whole_pose="1"
energy_cutoff="1000"/>
  </FILTERS>

  <SIMPLE_METRICS>
    <PerResidueEnergyMetric name="total_res_energies"
residue_selector="interface" scoretype="total_score" scorefxn="ref15"/>
    <PerResidueRMSDMetric name="res_RMSD"
residue_selector="interface" use_native="1" rmsd_type="rmsd_all"/>
  </SIMPLE_METRICS>

  <RESIDUE_LEVEL_TASK_OPERATIONS>
    <PreventRepackingRLT name="PreventRepacking"/>
  </RESIDUE_LEVEL_TASK_OPERATIONS>

  <TASKOPERATIONS>
    <DesignRestrictions name="notouch">
      <Action residue_selector="interface"
residue_level_operations="PreventRepacking"/>
      <Action residue_selector="lig"
residue_level_operations="PreventRepacking"/>
    </DesignRestrictions>

    <DesignRestrictions name="nomove">
      <Action residue_selector="lig"
residue_level_operations="PreventRepacking"/>
    </DesignRestrictions>

```

```

    <DetectProteinLigandInterface name="cuts_on" cut1="6" cut2="8"
cut3="10" cut4="12" design="1"/>
    <DetectProteinLigandInterface name="cuts_off" cut1="6" cut2="8"
cut3="10" cut4="12" design="0"/>

    <RestrictResiduesToRepacking name="pack_only"
residues="%%constrain%%"/>
    <ProteinLigandInterfaceUpweighter name="interface"
interface_weight="0.5"/>
    </TASKOPERATIONS>

<MOVERS>

    <FastRelax name="relax" scorefxn="ref15" disable_design="1"
task_operations="notouch"/>

    <RunSimpleMetrics name="total_initial_energy"
metrics="total_res_energies" prefix="initial_"/>
    <RunSimpleMetrics name="RMSD" metrics="res_RMSD"/>
    <RunSimpleMetrics name="total_final_energy"
metrics="total_res_energies" prefix="final_"/>

    #Favor Native Residues
    <FavorNativeResidue name="favor_natives" bonus="0.75"/>

    #Add constraints to file
    <AddOrRemoveMatchCsts name="addcst" cst_instruction="add_new"/>
    <AddOrRemoveMatchCsts name="rmvcst" cst_instruction="remove"/>

    #Optimize the pose per the cst file
    <EnzRepackMinimize name="cstopt" scorefxn_repack="soft_cst"
scorefxn_minimize="ref15_cst" cst_opt="1" design="0" repack_only="1"
fix_catalytic="1" minimize_rb="1" minimize_bb="1" minimize_sc="1"
minimize_lig="1" min_in_stages="1" cycles="1"
task_operations="cuts_off"/>

    #Design and repacking around the catalytic residues; keep
the catalytic residues fixed in this instance.
    <EnzRepackMinimize name="dsgn" scorefxn_repack="soft_cst"
scorefxn_minimize="ref15_cst" design="1" repack_only="0"
fix_catalytic="1" minimize_rb="1" minimize_bb="1" minimize_sc="1"
minimize_lig="1" min_in_stages="1" cycles="1"
task_operations="cuts_on,nomove"/> #Added nomove to prevent ligand from
moving too much 2/17/22

    #Minimize scaffold and ligand after each design
    <EnzRepackMinimize name="min_cst" scorefxn_repack="soft_cst"
scorefxn_minimize="ref15_cst" design="0" repack_only="1"
fix_catalytic="0" minimize_rb="1" minimize_bb="1" minimize_sc="1"
minimize_lig="1" min_in_stages="1" cycles="1"
task_operations="cuts_on"/>

```

```

        #Perform a final repacking step.
        <EnzRepackMinimize name="repack" scorefxn_repack="soft"
scorefxn_minimize="ref15" design="0" repack_only="1" fix_catalytic="0"
minimize_rb="1" minimize_bb="0" minimize_sc="1" minimize_lig="0"
min_in_stages="0" cycles="1" task_operations="cuts_off"/>

        #Perform a final minimization step.
        <EnzRepackMinimize name="min" scorefxn_repack="soft"
scorefxn_minimize="ref15" design="0" repack_only="0" fix_catalytic="0"
minimize_rb="1" minimize_bb="1" minimize_sc="1" minimize_lig="0"
min_in_stages="0" cycles="1" task_operations="cuts_off"/>
    </MOVERS>

    <PROTOCOLS>
        <Add mover_name="total_initial_energy"/>
        <Add mover_name="favor_natives"/>
        <Add mover_name="addcst"/>
        <Add mover_name="cstopt"/>
        <Add mover_name="dsgn"/>
        <Add mover_name="min_cst"/>
        <Add mover_name="repack"/>
        <Add mover_name="min"/>
        <Add mover_name="relax"/>
        <Add mover_name="total_final_energy"/>
        <Add mover_name="RMSD"/>
    </PROTOCOLS>

</ROSETTASCRIPTS>

```

Models of library variants were built by making the requisite mutations during the RosettaDesign algorithm and during the CoupledMoves algorithm using a “resfile.” A resfile is a Rosetta input file that indicates the locations and residue identities of the desired mutations. (Conway et al., 2014b; Nivón et al., 2013; Tyka et al., 2011). An example resfile to generate the mutations found in the Akaluc structure is as follows.

Akaluc resfile:

```

NATAA
start
36  A   PIKAA   A #T39A
45  A   PIKAA   Q #E48Q
48  A   PIKAA   V #I51V
65  A   PIKAA   R #K68R
83  A   PIKAA   S #L86S
131 A   PIKAA   R #Q134R
133 A   PIKAA   V #I136V
144 A   PIKAA   R #Q147R
172 A   PIKAA   S #G175S

```

221 A PIKAA Y #N229Y  
223 A PIKAA N #I231N  
286 A PIKAA C #F294C  
287 A PIKAA L #F295L  
300 A PIKAA S #N308S  
302 A PIKAA R #H310R  
324 A PIKAA R #H332R  
339 A PIKAA N #S347N  
341 A PIKAA V #I349V  
342 A PIKAA M #L350M  
349 A PIKAA R #D357R  
353 A PIKAA S #A361S  
369 A PIKAA V #D377V  
448 A PIKAA G #S456G  
#N463 missing density #N463Y  
505 A PIKAA R #K524R  
507 A PIKAA S #L526S  
521 A PIKAA T #I540T  
#G545D missing density

## APPENDIX D

### ADDITIONAL LIGAND PARAMS FILES FOR LUCIFERIN MODELS

The SLU.params file was as follows.

```
NAME SLU
IO_STRING SLU Z
TYPE LIGAND
AA UNK
ATOM S3 S X 1.62
ATOM N8 Ntrp X -0.79
ATOM C16 CNH2 X 0.81
ATOM C14 CH1 X 0.21
ATOM C8 aroC X 0.08
ATOM S2 S X -0.28
ATOM C11 aroC X 0.68
ATOM C7 aroC X 0.29
ATOM N1 Nhis X -0.57
ATOM C4 aroC X 0.23
ATOM C1 aroC X -0.15
ATOM C2 aroC X -0.15
ATOM C5 aroC X 0.08
ATOM C3 aroC X -0.15
ATOM C6 aroC X 0.04
ATOM S1 S X -0.08
ATOM H3 Haro X 0.15
ATOM O5 OH X -0.53
ATOM H16 Hpol X 0.45
ATOM H2 Haro X 0.15
ATOM H1 Haro X 0.15
ATOM N5 Nhis X -0.66
ATOM H4 Haro X 0.06
ATOM O1 ONH2 X -0.57
ATOM H15 Hpol X 0.42
ATOM O2 OOC X -0.65
ATOM O3 OOC X -0.65
ATOM O8 OH X -0.46
ATOM C21 CH2 X 0.28
ATOM C19 CH1 X 0.28
ATOM C17 CH1 X 0.28
ATOM C18 CH1 X 0.28
ATOM C20 CH1 X 0.63
ATOM N6 Npro X -0.08
ATOM C10 aroC X 0.55
ATOM N4 Nhis X -0.66
ATOM C12 aroC X 0.46
ATOM C13 aroC X 0.78
ATOM N3 Nhis X -0.66
ATOM C9 aroC X 0.60
ATOM N2 Nhis X -0.65
ATOM C15 CH1 X 0.51
ATOM N7 NH2O X -0.12
ATOM H13 Hpol X 0.40
ATOM H14 Hpol X 0.40
ATOM H5 Haro X 0.06
ATOM H6 Haro X 0.06
ATOM O4 OH X -0.56
ATOM H10 Hapo X 0.00
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ATOM	O7	OH	X	-0.68
ATOM	H18	Hpol	X	0.40
ATOM	H8	Hapo	X	0.00
ATOM	O6	OH	X	-0.68
ATOM	H17	Hpol	X	0.40
ATOM	H7	Hapo	X	0.00
ATOM	H9	Hapo	X	0.00
ATOM	H11	Hapo	X	0.00
ATOM	H12	Hapo	X	0.00
BOND_TYPE	C1	C2	4	
BOND_TYPE	C1	C4	4	
BOND_TYPE	C2	C5	4	
BOND_TYPE	C3	C5	4	
BOND_TYPE	C3	C6	4	
BOND_TYPE	C4	C6	4	
BOND_TYPE	C4	N1	1	
BOND_TYPE	C5	O5	1	
BOND_TYPE	C6	S1	1	
BOND_TYPE	C7	C11	1	
BOND_TYPE	C7	N1	2	
BOND_TYPE	C7	S1	1	
BOND_TYPE	C8	C14	1	
BOND_TYPE	C8	S2	2	
BOND_TYPE	C9	N2	2	
BOND_TYPE	C9	N3	1	
BOND_TYPE	C10	N4	1	
BOND_TYPE	C10	N6	2	
BOND_TYPE	C11	N5	2	
BOND_TYPE	C11	S2	1	
BOND_TYPE	C12	C13	1	
BOND_TYPE	C12	C15	1	
BOND_TYPE	C12	N4	2	
BOND_TYPE	C13	N3	2	
BOND_TYPE	C13	N6	1	
BOND_TYPE	C14	C16	1	
BOND_TYPE	C14	N5	1	
BOND_TYPE	C15	N2	1	
BOND_TYPE	C15	N7	1	
BOND_TYPE	C16	N8	4	
BOND_TYPE	C16	O1	2	
BOND_TYPE	C17	C18	1	
BOND_TYPE	C17	C19	1	
BOND_TYPE	C17	O6	1	
BOND_TYPE	C18	C20	1	
BOND_TYPE	C18	O7	1	
BOND_TYPE	C19	C21	1	
BOND_TYPE	C19	O4	1	
BOND_TYPE	C20	N6	1	
BOND_TYPE	C20	O4	1	
BOND_TYPE	C21	O8	1	
BOND_TYPE	N8	S3	1	
BOND_TYPE	O2	S3	2	
BOND_TYPE	O3	S3	2	
BOND_TYPE	O8	S3	1	
BOND_TYPE	C1	H1	1	

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BOND_TYPE  C2  H2  1
BOND_TYPE  C3  H3  1
BOND_TYPE  C8  H4  1
BOND_TYPE  C9  H5  1
BOND_TYPE  C10 H6  1
BOND_TYPE  C17 H7  1
BOND_TYPE  C18 H8  1
BOND_TYPE  C19 H9  1
BOND_TYPE  C20 H10 1
BOND_TYPE  C21 H11 1
BOND_TYPE  C21 H12 1
BOND_TYPE  N7  H13 1
BOND_TYPE  N7  H14 1
BOND_TYPE  N8  H15 1
BOND_TYPE  O5  H16 1
BOND_TYPE  O6  H17 1
BOND_TYPE  O7  H18 1
CHI 1  C2  C5  O5  H16
PROTON_CHI 1 SAMPLES 2 0 180 EXTRA 0
CHI 2  C19 C17 O6  H17
PROTON_CHI 2 SAMPLES 3 60 -60 180 EXTRA 0
CHI 3  C17 C18 O7  H18
PROTON_CHI 3 SAMPLES 3 60 -60 180 EXTRA 0
CHI 4  S2  C11 C7  N1
CHI 5  N8  C16 C14 C8
CHI 6  O8  C21 C19 C17
CHI 7  C18 C20 N6  C10
CHI 8  S3  O8  C21 C19
CHI 9  O2  S3  N8  C16
CHI 10 N8  S3  O8  C21
NBR_ATOM  S3
NBR_RADIUS 15.331690
ICOOR_INTERNAL  S3  0.000000  0.000000  0.000000  S3  N8
C16
ICOOR_INTERNAL  N8  0.000000  180.000000  1.649468  S3  N8
C16
ICOOR_INTERNAL  C16 0.000000  58.301303  1.468440  N8  S3
C16
ICOOR_INTERNAL  C14 -179.026765  66.190937  1.467099  C16  N8
S3
ICOOR_INTERNAL  C8  150.001919  53.612295  1.382663  C14  C16
N8
ICOOR_INTERNAL  S2 -179.276669  69.544208  1.714009  C8  C14
C16
ICOOR_INTERNAL  C11  0.364755  89.734176  1.730152  S2  C8
C14
ICOOR_INTERNAL  C7  179.943935  57.759050  1.444385  C11  S2
C8
ICOOR_INTERNAL  N1  -0.837493  56.799721  1.318117  C7  C11
S2
ICOOR_INTERNAL  C4  179.869950  69.803179  1.384206  N1  C7
C11
ICOOR_INTERNAL  C1  179.911901  54.184816  1.409523  C4  N1
C7

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ICOOR_INTERNAL N1	C2	179.936163	60.300737	1.398062	C1	C4
ICOOR_INTERNAL C4	C5	0.003526	59.887010	1.387361	C2	C1
ICOOR_INTERNAL C1	C3	0.007226	58.207192	1.394926	C5	C2
ICOOR_INTERNAL C2	C6	-0.009962	62.553954	1.406198	C3	C5
ICOOR_INTERNAL C5	S1	-179.977454	51.050316	1.720909	C6	C3
ICOOR_INTERNAL C6	H3	-179.989158	57.933132	1.083926	C3	C5
ICOOR_INTERNAL C3	O5	179.997951	62.706448	1.365157	C5	C2
ICOOR_INTERNAL C2	H16	179.998826	70.905029	0.971545	O5	C5
ICOOR_INTERNAL C5	H2	179.984819	60.095414	1.087046	C2	C1
ICOOR_INTERNAL C2	H1	179.992009	60.262532	1.085520	C1	C4
ICOOR_INTERNAL C7	N5	179.242791	66.371076	1.317856	C11	S2
ICOOR_INTERNAL S2	H4	179.991860	55.230173	1.031940	C8	C14
ICOOR_INTERNAL C14	O1	177.613201	56.617971	1.227139	C16	N8
ICOOR_INTERNAL C16	H15	-179.995484	60.845852	0.984556	N8	S3
ICOOR_INTERNAL C16	O2	-83.829697	72.117552	1.437687	S3	N8
ICOOR_INTERNAL O2	O3	-110.978610	68.096527	1.445149	S3	N8
ICOOR_INTERNAL O3	O8	-129.796645	67.882755	1.508902	S3	N8
ICOOR_INTERNAL N8	C21	47.972090	59.846860	1.411589	O8	S3
ICOOR_INTERNAL S3	C19	157.664803	70.737827	1.511020	C21	O8
ICOOR_INTERNAL O8	C17	-162.089472	70.934687	1.469290	C19	C21
ICOOR_INTERNAL C21	C18	-144.250635	77.197183	1.465518	C17	C19
ICOOR_INTERNAL C19	C20	31.359911	77.525336	1.475945	C18	C17
ICOOR_INTERNAL C17	N6	91.982589	65.434427	1.445330	C20	C18
ICOOR_INTERNAL C18	C10	-93.612286	52.726401	1.352130	N6	C20
ICOOR_INTERNAL C20	N4	-179.125529	69.963424	1.324697	C10	N6
ICOOR_INTERNAL N6	C12	0.434724	70.201937	1.328569	N4	C10
ICOOR_INTERNAL C10	C13	-0.108966	73.543980	1.420687	C12	N4

ICOOR_INTERNAL N4	N3	179.727076	60.631876	1.342184	C13	C12
ICOOR_INTERNAL C12	C9	0.605797	59.617650	1.325985	N3	C13
ICOOR_INTERNAL C13	N2	-0.843694	57.615869	1.331578	C9	N3
ICOOR_INTERNAL N3	C15	0.543846	58.116997	1.342047	N2	C9
ICOOR_INTERNAL C9	N7	-179.885925	60.505121	1.449333	C15	N2
ICOOR_INTERNAL N2	H13	179.857903	59.996676	0.984436	N7	C15
ICOOR_INTERNAL H13	H14	-179.996362	60.003205	0.984528	N7	C15
ICOOR_INTERNAL N2	H5	179.991347	61.186872	1.031972	C9	N3
ICOOR_INTERNAL N4	H6	179.994533	55.014941	1.032043	C10	N6
ICOOR_INTERNAL N6	O4	-122.220593	71.733603	1.413878	C20	C18
ICOOR_INTERNAL O4	H10	-121.131406	73.232551	1.069922	C20	C18
ICOOR_INTERNAL C20	O7	-116.507140	66.309831	1.374491	C18	C17
ICOOR_INTERNAL C17	H18	179.997609	70.527483	0.969954	O7	C18
ICOOR_INTERNAL O7	H8	-120.201693	66.638819	1.112365	C18	C17
ICOOR_INTERNAL C18	O6	-122.102731	69.950997	1.380152	C17	C19
ICOOR_INTERNAL C19	H17	-64.962631	70.528340	0.970025	O6	C17
ICOOR_INTERNAL O6	H7	-117.021009	64.837781	1.070175	C17	C19
ICOOR_INTERNAL C17	H9	119.990690	70.467539	1.070000	C19	C21
ICOOR_INTERNAL C19	H11	-119.973622	70.474860	1.069935	C21	O8
ICOOR_INTERNAL H11	H12	-119.929165	70.409662	1.070081	C21	O8

The CouLuc-3-NMe<sub>2</sub> .params file for the Z isomer is below.

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NAME LCA
IO_STRING LCA Z
TYPE LIGAND
AA UNK
ATOM C21 CNH2 X 0.89
ATOM C8 aroC X 0.14
ATOM C4 aroC X -0.11
ATOM S1 S X -0.08
ATOM C9 aroC X 0.33
ATOM C18 aroC X -0.10
ATOM C20 aroC X -0.15
ATOM C19 aroC X -0.15

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ATOM	C15	aroC	X	0.08
ATOM	C10	aroC	X	-0.15
ATOM	C13	aroC	X	-0.17
ATOM	C5	aroC	X	0.03
ATOM	C1	aroC	X	-0.15
ATOM	C2	aroC	X	-0.15
ATOM	C6	aroC	X	0.10
ATOM	C3	aroC	X	-0.15
ATOM	C7	aroC	X	0.08
ATOM	O4	Oaro	X	-0.16
ATOM	H3	Haro	X	0.15
ATOM	N8	Nhis	X	-0.84
ATOM	C26	CH3	X	0.37
ATOM	H15	Hapo	X	0.00
ATOM	H16	Hapo	X	0.00
ATOM	H17	Hapo	X	0.00
ATOM	C27	CH3	X	0.37
ATOM	H18	Hapo	X	0.00
ATOM	H19	Hapo	X	0.00
ATOM	H20	Hapo	X	0.00
ATOM	H2	Haro	X	0.15
ATOM	H1	Haro	X	0.15
ATOM	C29	CH1	X	1.16
ATOM	F1	F	X	-0.34
ATOM	F2	F	X	-0.34
ATOM	F3	F	X	-0.34
ATOM	H5	Haro	X	0.15
ATOM	H9	Haro	X	0.15
ATOM	H10	Haro	X	0.15
ATOM	H8	Haro	X	0.15
ATOM	N1	Nhis	X	-0.57
ATOM	H4	Haro	X	0.15
ATOM	N7	Ntrp	X	-0.79
ATOM	S2	S	X	1.62
ATOM	O2	OOC	X	-0.65
ATOM	O3	OOC	X	-0.65
ATOM	O8	OH	X	-0.46
ATOM	C28	CH2	X	0.28
ATOM	C24	CH1	X	0.28
ATOM	C22	CH1	X	0.28
ATOM	C23	CH1	X	0.28
ATOM	C25	CH1	X	0.63
ATOM	N5	Npro	X	-0.08
ATOM	C12	aroC	X	0.55
ATOM	N4	Nhis	X	-0.66
ATOM	C14	aroC	X	0.46
ATOM	C16	aroC	X	0.78
ATOM	N3	Nhis	X	-0.66
ATOM	C11	aroC	X	0.60
ATOM	N2	Nhis	X	-0.65
ATOM	C17	CH1	X	0.51
ATOM	N6	NH2O	X	-0.12
ATOM	H23	Hpol	X	0.40
ATOM	H24	Hpol	X	0.40
ATOM	H6	Haro	X	0.06

ATOM	H7	Haro	X	0.06
ATOM	O5	OH	X	-0.56
ATOM	H14	Hapo	X	0.00
ATOM	O7	OH	X	-0.68
ATOM	H27	Hpo1	X	0.40
ATOM	H12	Hapo	X	0.00
ATOM	O6	OH	X	-0.68
ATOM	H26	Hpo1	X	0.40
ATOM	H11	Hapo	X	0.00
ATOM	H13	Hapo	X	0.00
ATOM	H21	Hapo	X	0.00
ATOM	H22	Hapo	X	0.00
ATOM	H25	Hpo1	X	0.42
ATOM	O1	ONH2	X	-0.57
BOND_TYPE	C1	C2	4	
BOND_TYPE	C1	C5	4	
BOND_TYPE	C2	C6	4	
BOND_TYPE	C3	C6	4	
BOND_TYPE	C3	C7	4	
BOND_TYPE	C4	C8	2	
BOND_TYPE	C4	S1	1	
BOND_TYPE	C5	C7	4	
BOND_TYPE	C5	C13	1	
BOND_TYPE	C6	N8	1	
BOND_TYPE	C7	O4	1	
BOND_TYPE	C8	C21	1	
BOND_TYPE	C8	N1	1	
BOND_TYPE	C9	C18	1	
BOND_TYPE	C9	N1	2	
BOND_TYPE	C9	S1	1	
BOND_TYPE	C10	C13	2	
BOND_TYPE	C10	C15	1	
BOND_TYPE	C11	N2	2	
BOND_TYPE	C11	N3	1	
BOND_TYPE	C12	N4	1	
BOND_TYPE	C12	N5	2	
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BOND_TYPE	C14	C16	1	
BOND_TYPE	C14	C17	1	
BOND_TYPE	C14	N4	2	
BOND_TYPE	C15	C19	2	
BOND_TYPE	C15	O4	1	
BOND_TYPE	C16	N3	2	
BOND_TYPE	C16	N5	1	
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BOND_TYPE	C22	O6	1	
BOND_TYPE	C23	C25	1	
BOND_TYPE	C23	O7	1	

BOND_TYPE	C24	C28	1						
BOND_TYPE	C24	O5	1						
BOND_TYPE	C25	N5	1						
BOND_TYPE	C25	O5	1						
BOND_TYPE	C26	N8	1						
BOND_TYPE	C27	N8	1						
BOND_TYPE	C28	O8	1						
BOND_TYPE	C29	F1	1						
BOND_TYPE	C29	F2	1						
BOND_TYPE	C29	F3	1						
BOND_TYPE	N7	S2	1						
BOND_TYPE	O2	S2	2						
BOND_TYPE	O3	S2	2						
BOND_TYPE	O8	S2	1						
BOND_TYPE	C1	H1	1						
BOND_TYPE	C2	H2	1						
BOND_TYPE	C3	H3	1						
BOND_TYPE	C4	H4	1						
BOND_TYPE	C10	H5	1						
BOND_TYPE	C11	H6	1						
BOND_TYPE	C12	H7	1						
BOND_TYPE	C18	H8	1						
BOND_TYPE	C19	H9	1						
BOND_TYPE	C20	H10	1						
BOND_TYPE	C22	H11	1						
BOND_TYPE	C23	H12	1						
BOND_TYPE	C24	H13	1						
BOND_TYPE	C25	H14	1						
BOND_TYPE	C26	H15	1						
BOND_TYPE	C26	H16	1						
BOND_TYPE	C26	H17	1						
BOND_TYPE	C27	H18	1						
BOND_TYPE	C27	H19	1						
BOND_TYPE	C27	H20	1						
BOND_TYPE	C28	H21	1						
BOND_TYPE	C28	H22	1						
BOND_TYPE	N6	H23	1						
BOND_TYPE	N6	H24	1						
BOND_TYPE	N7	H25	1						
BOND_TYPE	O6	H26	1						
BOND_TYPE	O7	H27	1						
CHI 1	C24	C22	O6	H26					
PROTON_CHI	1	SAMPLES	3	60	-60	180	EXTRA	0	
CHI 2	C22	C23	O7	H27					
PROTON_CHI	2	SAMPLES	3	60	-60	180	EXTRA	0	
CHI 3	C2	C6	N8	C26					
CHI 4	N7	C21	C8	C4					
CHI 5	S1	C9	C18	C20					
CHI 6	C10	C13	C29	F1					
CHI 7	C18	C20	C19	C15					
CHI 8	O8	C28	C24	C22					
CHI 9	C23	C25	N5	C12					
CHI 10	S2	O8	C28	C24					
CHI 11	C21	N7	S2	O2					
CHI 12	N7	S2	O8	C28					

NBR\_ATOM C21

NBR\_RADIUS 17.590121

ICOOR_INTERNAL	C21	0.000000	0.000000	0.000000	C21	C8	C4
ICOOR_INTERNAL	C8	0.000000	179.999999	1.467129	C21	C8	C4
ICOOR_INTERNAL	C4	0.000001	53.617032	1.382669	C8	C21	C4
ICOOR_INTERNAL	S1	-179.281387	69.551912	1.714111	C4	C8	C21
ICOOR_INTERNAL	C9	0.368538	89.731913	1.730053	S1	C4	C8
ICOOR_INTERNAL	C18	-179.983428	55.061880	1.476720	C9	S1	C4
ICOOR_INTERNAL	C20	179.107511	57.154412	1.342614	C18	C9	S1
ICOOR_INTERNAL	C19	179.963866	59.541585	1.485000	C20	C18	C9
ICOOR_INTERNAL	C15	179.999695	54.513145	1.349071	C19	C20	C18
ICOOR_INTERNAL	C10	-0.053775	56.133742	1.491760	C15	C19	C20
ICOOR_INTERNAL	C13	-179.979235	59.844397	1.351898	C10	C15	C19
ICOOR_INTERNAL	C5	-0.016465	61.367697	1.498679	C13	C10	C15
ICOOR_INTERNAL	C1	-179.978742	57.380727	1.404750	C5	C13	C10
ICOOR_INTERNAL	C2	179.978603	59.280050	1.396556	C1	C5	C13
ICOOR_INTERNAL	C6	-0.011294	58.773291	1.412211	C2	C1	C5
ICOOR_INTERNAL	C3	0.011612	62.377933	1.412764	C6	C2	C1
ICOOR_INTERNAL	C7	-0.002535	58.740853	1.398770	C3	C6	C2
ICOOR_INTERNAL	O4	179.994190	61.003563	1.354369	C7	C3	C6
ICOOR_INTERNAL	H3	179.996756	58.715470	1.080834	C3	C6	C7
ICOOR_INTERNAL	N8	-179.978952	58.833041	1.444598	C6	C2	C3
ICOOR_INTERNAL	C26	-0.005874	59.369540	1.456963	N8	C6	C2
ICOOR_INTERNAL	H15	-179.999668	70.119058	1.110614	C26	N8	C6
ICOOR_INTERNAL	H16	-120.878814	70.148362	1.110679	C26	N8	H15
ICOOR_INTERNAL	H17	-119.538468	67.587037	1.110219	C26	N8	H16
ICOOR_INTERNAL	C27	179.991082	59.353693	1.457042	N8	C6	C26
ICOOR_INTERNAL	H18	179.997383	70.129063	1.110567	C27	N8	C6
ICOOR_INTERNAL	H19	-119.544038	67.600229	1.110291	C27	N8	H18
ICOOR_INTERNAL	H20	-119.542316	70.127139	1.110547	C27	N8	H19
ICOOR_INTERNAL	H2	-179.980340	62.733103	1.080662	C2	C1	C6
ICOOR_INTERNAL	H1	-179.996344	58.605071	1.080428	C1	C5	C2
ICOOR_INTERNAL	C29	-179.995527	59.589585	1.520781	C13	C10	C5
ICOOR_INTERNAL	F1	179.998863	67.610783	1.385073	C29	C13	C10
ICOOR_INTERNAL	F2	120.037494	70.475588	1.383039	C29	C13	F1
ICOOR_INTERNAL	F3	119.916915	70.471215	1.382932	C29	C13	F2
ICOOR_INTERNAL	H5	179.977690	59.689688	1.080359	C10	C15	C13
ICOOR_INTERNAL	H9	179.992072	62.715893	1.087842	C19	C20	C15
ICOOR_INTERNAL	H10	-179.998350	61.078692	1.080735	C20	C18	C19
ICOOR_INTERNAL	H8	-179.998975	62.397857	1.087144	C18	C9	C20
ICOOR_INTERNAL	N1	179.169058	66.367720	1.317939	C9	S1	C18
ICOOR_INTERNAL	H4	178.050628	50.571171	1.080506	C4	C8	S1
ICOOR_INTERNAL	N7	179.999349	66.189768	1.468382	C21	C8	C4
ICOOR_INTERNAL	S2	-179.028528	58.301695	1.649480	N7	C21	C8
ICOOR_INTERNAL	O2	-83.828030	72.115214	1.437728	S2	N7	C21
ICOOR_INTERNAL	O3	-110.979907	68.095120	1.445125	S2	N7	O2
ICOOR_INTERNAL	O8	-129.796485	67.884564	1.508898	S2	N7	O3
ICOOR_INTERNAL	C28	47.972804	59.847207	1.411561	O8	S2	N7
ICOOR_INTERNAL	C24	157.660576	70.738957	1.511080	C28	O8	S2
ICOOR_INTERNAL	C22	-162.092351	70.936507	1.469262	C24	C28	O8
ICOOR_INTERNAL	C23	-144.249833	77.197932	1.465575	C22	C24	C28
ICOOR_INTERNAL	C25	31.366047	77.527376	1.475916	C23	C22	C24
ICOOR_INTERNAL	N5	91.982349	65.429978	1.445332	C25	C23	C22
ICOOR_INTERNAL	C12	-93.611780	52.728629	1.352156	N5	C25	C23
ICOOR_INTERNAL	N4	-179.129471	69.965230	1.324715	C12	N5	C25

ICOOR_INTERNAL	C14	0.433861	70.203038	1.328601	N4	C12	N5
ICOOR_INTERNAL	C16	-0.110422	73.543260	1.420679	C14	N4	C12
ICOOR_INTERNAL	N3	179.726204	60.629497	1.342216	C16	C14	N4
ICOOR_INTERNAL	C11	0.608176	59.624923	1.326020	N3	C16	C14
ICOOR_INTERNAL	N2	-0.848435	57.610071	1.331554	C11	N3	C16
ICOOR_INTERNAL	C17	0.552053	58.120141	1.342096	N2	C11	N3
ICOOR_INTERNAL	N6	-179.890730	60.506917	1.449335	C17	N2	C11
ICOOR_INTERNAL	H23	179.859865	60.001164	0.984524	N6	C17	N2
ICOOR_INTERNAL	H24	179.991250	60.000412	0.984519	N6	C17	H23
ICOOR_INTERNAL	H6	-179.999269	61.192473	1.031974	C11	N3	N2
ICOOR_INTERNAL	H7	179.997573	55.017697	1.031995	C12	N5	N4
ICOOR_INTERNAL	O5	-122.223815	71.733653	1.413900	C25	C23	N5
ICOOR_INTERNAL	H14	-121.130496	73.235732	1.069989	C25	C23	O5
ICOOR_INTERNAL	O7	-116.511459	66.309694	1.374530	C23	C22	C25
ICOOR_INTERNAL	H27	-179.997776	70.530952	0.969950	O7	C23	C22
ICOOR_INTERNAL	H12	-118.778660	69.018066	1.070015	C23	C22	O7
ICOOR_INTERNAL	O6	-122.097254	69.957154	1.380198	C22	C24	C23
ICOOR_INTERNAL	H26	-64.966530	70.526952	0.970044	O6	C22	C24
ICOOR_INTERNAL	H11	-117.021961	64.834299	1.070064	C22	C24	O6
ICOOR_INTERNAL	H13	120.000334	70.466931	1.069977	C24	C28	C22
ICOOR_INTERNAL	H21	-119.967044	70.476305	1.069971	C28	O8	C24
ICOOR_INTERNAL	H22	-119.934519	70.408073	1.070003	C28	O8	H21
ICOOR_INTERNAL	H25	-179.999258	60.849549	0.984516	N7	C21	S2
ICOOR_INTERNAL	O1	-177.630746	57.240813	1.227206	C21	C8	N7

PDB\_ROTAMERS LCAa\_conformers.pdb

The CouLuc-3-NMe<sub>2</sub> .params file for the E isomer is below.

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NAME LCA
IO_STRING LCA Z
TYPE LIGAND
AA UNK
ATOM C21 CNH2 X 0.89
ATOM C8 aroC X 0.14
ATOM C4 aroC X -0.11
ATOM S1 S X -0.08
ATOM C9 aroC X 0.33
ATOM C18 aroC X -0.10
ATOM C20 aroC X -0.15
ATOM C19 aroC X -0.15
ATOM C15 aroC X 0.08
ATOM C10 aroC X -0.15
ATOM C13 aroC X -0.17
ATOM C5 aroC X 0.03
ATOM C1 aroC X -0.15
ATOM C2 aroC X -0.15
ATOM C6 aroC X 0.10
ATOM C3 aroC X -0.15
ATOM C7 aroC X 0.08
ATOM O4 Oaro X -0.16
ATOM H3 Haro X 0.15
ATOM N8 Nhis X -0.84
ATOM C26 CH3 X 0.37
ATOM H15 Hapo X 0.00

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ATOM	H16	Hapo	X	0.00
ATOM	H17	Hapo	X	0.00
ATOM	C27	CH3	X	0.37
ATOM	H18	Hapo	X	0.00
ATOM	H19	Hapo	X	0.00
ATOM	H20	Hapo	X	0.00
ATOM	H2	Haro	X	0.15
ATOM	H1	Haro	X	0.15
ATOM	C29	CH1	X	1.16
ATOM	F1	F	X	-0.34
ATOM	F2	F	X	-0.34
ATOM	F3	F	X	-0.34
ATOM	H5	Haro	X	0.15
ATOM	H9	Haro	X	0.15
ATOM	H10	Haro	X	0.15
ATOM	H8	Haro	X	0.15
ATOM	N1	Nhis	X	-0.57
ATOM	H4	Haro	X	0.15
ATOM	N7	Ntrp	X	-0.79
ATOM	S2	S	X	1.62
ATOM	O2	OOC	X	-0.65
ATOM	O3	OOC	X	-0.65
ATOM	O8	OH	X	-0.46
ATOM	C28	CH2	X	0.28
ATOM	C24	CH1	X	0.28
ATOM	C22	CH1	X	0.28
ATOM	C23	CH1	X	0.28
ATOM	C25	CH1	X	0.63
ATOM	N5	Npro	X	-0.08
ATOM	C12	aroC	X	0.55
ATOM	N4	Nhis	X	-0.66
ATOM	C14	aroC	X	0.46
ATOM	C16	aroC	X	0.78
ATOM	N3	Nhis	X	-0.66
ATOM	C11	aroC	X	0.60
ATOM	N2	Nhis	X	-0.65
ATOM	C17	CH1	X	0.51
ATOM	N6	NH2O	X	-0.12
ATOM	H23	Hpol	X	0.40
ATOM	H24	Hpol	X	0.40
ATOM	H6	Haro	X	0.06
ATOM	H7	Haro	X	0.06
ATOM	O5	OH	X	-0.56
ATOM	H14	Hapo	X	0.00
ATOM	O7	OH	X	-0.68
ATOM	H27	Hpol	X	0.40
ATOM	H12	Hapo	X	0.00
ATOM	O6	OH	X	-0.68
ATOM	H26	Hpol	X	0.40
ATOM	H11	Hapo	X	0.00
ATOM	H13	Hapo	X	0.00
ATOM	H21	Hapo	X	0.00
ATOM	H22	Hapo	X	0.00
ATOM	H25	Hpol	X	0.42
ATOM	O1	ONH2	X	-0.57



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BOND_TYPE	C6	N8	1
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BOND_TYPE	C8	C21	1
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BOND_TYPE	C16	N3	2
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BOND_TYPE	C17	N2	1
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BOND_TYPE	C21	N7	4
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BOND_TYPE	C22	C23	1
BOND_TYPE	C22	C24	1
BOND_TYPE	C22	O6	1
BOND_TYPE	C23	C25	1
BOND_TYPE	C23	O7	1
BOND_TYPE	C24	C28	1
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BOND_TYPE	C25	N5	1
BOND_TYPE	C25	O5	1
BOND_TYPE	C26	N8	1
BOND_TYPE	C27	N8	1
BOND_TYPE	C28	O8	1
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BOND_TYPE	C29	F3	1
BOND_TYPE	N7	S2	1
BOND_TYPE	O2	S2	2
BOND_TYPE	O3	S2	2
BOND_TYPE	O8	S2	1

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BOND_TYPE C2 H2 1
BOND_TYPE C3 H3 1
BOND_TYPE C4 H4 1
BOND_TYPE C10 H5 1
BOND_TYPE C11 H6 1
BOND_TYPE C12 H7 1
BOND_TYPE C18 H8 1
BOND_TYPE C19 H9 1
BOND_TYPE C20 H10 1
BOND_TYPE C22 H11 1
BOND_TYPE C23 H12 1
BOND_TYPE C24 H13 1
BOND_TYPE C25 H14 1
BOND_TYPE C26 H15 1
BOND_TYPE C26 H16 1
BOND_TYPE C26 H17 1
BOND_TYPE C27 H18 1
BOND_TYPE C27 H19 1
BOND_TYPE C27 H20 1
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BOND_TYPE C28 H22 1
BOND_TYPE N6 H23 1
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BOND_TYPE N7 H25 1
BOND_TYPE O6 H26 1
BOND_TYPE O7 H27 1
CHI 1 C24 C22 O6 H26
PROTON_CHI 1 SAMPLES 3 60 -60 180 EXTRA 0
CHI 2 C22 C23 O7 H27
PROTON_CHI 2 SAMPLES 3 60 -60 180 EXTRA 0
CHI 3 C2 C6 N8 C26
CHI 4 N7 C21 C8 C4
CHI 5 S1 C9 C18 C20
CHI 6 C10 C13 C29 F1
CHI 7 C18 C20 C19 C15
CHI 8 O8 C28 C24 C22
CHI 9 C23 C25 N5 C12
CHI 10 S2 O8 C28 C24
CHI 11 C21 N7 S2 O2
CHI 12 N7 S2 O8 C28
NBR_ATOM C21
NBR_RADIUS 17.643938
ICOOR_INTERNAL C21 0.000000 0.000000 0.000000 C21 C8 C4
ICOOR_INTERNAL C8 0.000000 180.000000 1.467099 C21 C8 C4
ICOOR_INTERNAL C4 0.000001 53.613584 1.382650 C8 C21 C4
ICOOR_INTERNAL S1 -179.277244 69.545521 1.714033 C4 C8 C21
ICOOR_INTERNAL C9 0.366000 89.733106 1.730169 S1 C4 C8
ICOOR_INTERNAL C18 178.480672 53.956919 1.472190 C9 S1 C4
ICOOR_INTERNAL C20 -179.253693 55.590211 1.341657 C18 C9 S1
ICOOR_INTERNAL C19 -178.860703 61.081220 1.480117 C20 C18 C9
ICOOR_INTERNAL C15 179.997551 56.075757 1.342682 C19 C20 C18
ICOOR_INTERNAL C10 -179.356063 60.754970 1.477869 C15 C19 C20
ICOOR_INTERNAL C13 179.033457 58.995367 1.347258 C10 C15 C19
ICOOR_INTERNAL C5 0.096313 62.200717 1.513467 C13 C10 C15

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ICOOR_INTERNAL	C1	-179.627334	56.000846	1.402666	C5	C13	C10
ICOOR_INTERNAL	C2	-179.917507	57.810018	1.397502	C1	C5	C13
ICOOR_INTERNAL	C6	0.093158	59.285238	1.414135	C2	C1	C5
ICOOR_INTERNAL	C3	-0.127674	62.800534	1.416784	C6	C2	C1
ICOOR_INTERNAL	C7	-0.012295	58.232499	1.401153	C3	C6	C2
ICOOR_INTERNAL	O4	-179.599495	61.355171	1.358145	C7	C3	C6
ICOOR_INTERNAL	H3	-178.593087	58.090090	1.078016	C3	C6	C7
ICOOR_INTERNAL	N8	-179.971338	58.616652	1.449221	C6	C2	C3
ICOOR_INTERNAL	C26	-0.001984	57.458357	1.457852	N8	C6	C2
ICOOR_INTERNAL	H15	179.999884	65.780358	1.102888	C26	N8	C6
ICOOR_INTERNAL	H16	-122.635202	70.083946	1.110682	C26	N8	H15
ICOOR_INTERNAL	H17	-119.309917	70.640070	1.110827	C26	N8	H16
ICOOR_INTERNAL	C27	179.959626	57.499291	1.457905	N8	C6	C26
ICOOR_INTERNAL	H18	179.995176	70.637767	1.110918	C27	N8	C6
ICOOR_INTERNAL	H19	-118.086256	65.797498	1.102983	C27	N8	H18
ICOOR_INTERNAL	H20	-122.600548	70.079213	1.110695	C27	N8	H19
ICOOR_INTERNAL	H2	178.693040	63.235478	1.078324	C2	C1	C6
ICOOR_INTERNAL	H1	179.717317	57.022015	1.070466	C1	C5	C2
ICOOR_INTERNAL	C29	-179.946573	64.677245	1.528239	C13	C10	C5
ICOOR_INTERNAL	F1	179.945748	64.850453	1.374843	C29	C13	C10
ICOOR_INTERNAL	F2	120.797426	70.974799	1.384186	C29	C13	F1
ICOOR_INTERNAL	F3	118.349656	70.966161	1.384147	C29	C13	F2
ICOOR_INTERNAL	H5	-179.768550	61.078044	1.088585	C10	C15	C13
ICOOR_INTERNAL	H9	-179.310715	62.326928	1.088433	C19	C20	C15
ICOOR_INTERNAL	H10	179.101411	59.204346	1.085987	C20	C18	C19
ICOOR_INTERNAL	H8	-178.976872	63.534159	1.087683	C18	C9	C20
ICOOR_INTERNAL	N1	-179.293981	66.372571	1.317944	C9	S1	C18
ICOOR_INTERNAL	H4	178.052410	50.569171	1.080507	C4	C8	S1
ICOOR_INTERNAL	N7	179.999853	66.190937	1.468440	C21	C8	C4
ICOOR_INTERNAL	S2	-179.026765	58.301303	1.649468	N7	C21	C8
ICOOR_INTERNAL	O2	-83.829697	72.117552	1.437687	S2	N7	C21
ICOOR_INTERNAL	O3	-110.978610	68.096527	1.445149	S2	N7	O2
ICOOR_INTERNAL	O8	-129.796645	67.882755	1.508902	S2	N7	O3
ICOOR_INTERNAL	C28	47.972090	59.846860	1.411589	O8	S2	N7
ICOOR_INTERNAL	C24	157.664803	70.737827	1.511020	C28	O8	S2
ICOOR_INTERNAL	C22	-162.089472	70.934687	1.469290	C24	C28	O8
ICOOR_INTERNAL	C23	-144.250635	77.197183	1.465518	C22	C24	C28
ICOOR_INTERNAL	C25	31.359911	77.525336	1.475945	C23	C22	C24
ICOOR_INTERNAL	N5	91.982589	65.434427	1.445330	C25	C23	C22
ICOOR_INTERNAL	C12	-93.612286	52.726401	1.352130	N5	C25	C23
ICOOR_INTERNAL	N4	-179.125529	69.963424	1.324697	C12	N5	C25
ICOOR_INTERNAL	C14	0.434724	70.201937	1.328569	N4	C12	N5
ICOOR_INTERNAL	C16	-0.108966	73.543980	1.420687	C14	N4	C12
ICOOR_INTERNAL	N3	179.727076	60.631876	1.342184	C16	C14	N4
ICOOR_INTERNAL	C11	0.605797	59.617650	1.325985	N3	C16	C14
ICOOR_INTERNAL	N2	-0.843694	57.615869	1.331578	C11	N3	C16
ICOOR_INTERNAL	C17	0.543846	58.116997	1.342047	N2	C11	N3
ICOOR_INTERNAL	N6	-179.885925	60.505121	1.449333	C17	N2	C11
ICOOR_INTERNAL	H23	179.857903	59.996676	0.984436	N6	C17	N2
ICOOR_INTERNAL	H24	-179.996362	60.003205	0.984528	N6	C17	H23
ICOOR_INTERNAL	H6	179.991347	61.186872	1.031972	C11	N3	N2
ICOOR_INTERNAL	H7	179.994533	55.014941	1.032043	C12	N5	N4
ICOOR_INTERNAL	O5	-122.220593	71.733603	1.413878	C25	C23	N5
ICOOR_INTERNAL	H14	-121.131406	73.232551	1.069922	C25	C23	O5
ICOOR_INTERNAL	O7	-116.507140	66.309831	1.374491	C23	C22	C25

ICOOOR_INTERNAL	H27	179.997609	70.527483	0.969954	O7	C23	C22
ICOOOR_INTERNAL	H12	-120.201693	66.638819	1.112365	C23	C22	O7
ICOOOR_INTERNAL	O6	-122.102731	69.950997	1.380152	C22	C24	C23
ICOOOR_INTERNAL	H26	-64.962631	70.528340	0.970025	O6	C22	C24
ICOOOR_INTERNAL	H11	-117.021009	64.837781	1.070175	C22	C24	O6
ICOOOR_INTERNAL	H13	119.990690	70.467539	1.070000	C24	C28	C22
ICOOOR_INTERNAL	H21	-119.973622	70.474860	1.069935	C28	O8	C24
ICOOOR_INTERNAL	H22	-119.929165	70.409662	1.070081	C28	O8	H21
ICOOOR_INTERNAL	H25	179.995484	60.852845	0.984556	N7	C21	S2
ICOOOR_INTERNAL	O1	-177.629925	57.236261	1.227139	C21	C8	N7
PDB_ROTAMERS LCAb_conformers.pdb							

The CouLuc-3-OH .params file for the Z isomer is below.

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NAME LCB
IO_STRING LCB Z
TYPE LIGAND
AA UNK
ATOM C21 CNH2 X 0.89
ATOM C8 aroC X 0.14
ATOM C4 aroC X -0.11
ATOM S1 S X -0.08
ATOM C9 aroC X 0.33
ATOM C18 aroC X -0.10
ATOM C20 aroC X -0.15
ATOM C19 aroC X -0.15
ATOM C15 aroC X 0.08
ATOM C10 aroC X -0.15
ATOM C13 aroC X -0.17
ATOM C5 aroC X 0.03
ATOM C1 aroC X -0.15
ATOM C2 aroC X -0.15
ATOM C7 aroC X 0.08
ATOM C3 aroC X -0.15
ATOM C6 aroC X 0.08
ATOM O4 Oaro X -0.16
ATOM H3 Haro X 0.15
ATOM O6 OH X -0.53
ATOM H20 Hpol X 0.45
ATOM H2 Haro X 0.15
ATOM H1 Haro X 0.15
ATOM C27 CH1 X 1.16
ATOM F1 F X -0.34
ATOM F2 F X -0.34
ATOM F3 F X -0.34
ATOM H5 Haro X 0.15
ATOM H9 Haro X 0.15
ATOM H10 Haro X 0.15
ATOM H8 Haro X 0.15
ATOM N1 Nhis X -0.57
ATOM H4 Haro X 0.15
ATOM N7 Ntrp X -0.79
ATOM S2 S X 1.62
ATOM O2 OOC X -0.65

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ATOM	O3	OOC	X	-0.65
ATOM	O9	OH	X	-0.46
ATOM	C26	CH2	X	0.28
ATOM	C24	CH1	X	0.28
ATOM	C22	CH1	X	0.28
ATOM	C23	CH1	X	0.28
ATOM	C25	CH1	X	0.63
ATOM	N5	Npro	X	-0.08
ATOM	C12	aroC	X	0.55
ATOM	N4	Nhis	X	-0.66
ATOM	C14	aroC	X	0.46
ATOM	C16	aroC	X	0.78
ATOM	N3	Nhis	X	-0.66
ATOM	C11	aroC	X	0.60
ATOM	N2	Nhis	X	-0.65
ATOM	C17	CH1	X	0.51
ATOM	N6	NH2O	X	-0.12
ATOM	H17	Hpol	X	0.40
ATOM	H18	Hpol	X	0.40
ATOM	H6	Haro	X	0.06
ATOM	H7	Haro	X	0.06
ATOM	O5	OH	X	-0.56
ATOM	H14	Hapo	X	0.00
ATOM	O8	OH	X	-0.68
ATOM	H22	Hpol	X	0.40
ATOM	H12	Hapo	X	0.00
ATOM	O7	OH	X	-0.68
ATOM	H21	Hpol	X	0.40
ATOM	H11	Hapo	X	0.00
ATOM	H13	Hapo	X	0.00
ATOM	H15	Hapo	X	0.00
ATOM	H16	Hapo	X	0.00
ATOM	H19	Hpol	X	0.42
ATOM	O1	ONH2	X	-0.57
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BOND_TYPE	C2	C7	4	
BOND_TYPE	C3	C6	4	
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BOND_TYPE	C4	C8	2	
BOND_TYPE	C4	S1	1	
BOND_TYPE	C5	C6	4	
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BOND_TYPE	C6	O4	1	
BOND_TYPE	C7	O6	1	
BOND_TYPE	C8	C21	1	
BOND_TYPE	C8	N1	1	
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BOND_TYPE	C11	N2	2	
BOND_TYPE	C11	N3	1	
BOND_TYPE	C12	N4	1	

BOND_TYPE	C12	N5	2	
BOND_TYPE	C13	C27	1	
BOND_TYPE	C14	C16	1	
BOND_TYPE	C14	C17	1	
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BOND_TYPE	C15	C19	2	
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BOND_TYPE	C22	C23	1	
BOND_TYPE	C22	C24	1	
BOND_TYPE	C22	O7	1	
BOND_TYPE	C23	C25	1	
BOND_TYPE	C23	O8	1	
BOND_TYPE	C24	C26	1	
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BOND_TYPE	C27	F3	1	
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BOND_TYPE	O3	S2	2	
BOND_TYPE	O9	S2	1	
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BOND_TYPE	C2	H2	1	
BOND_TYPE	C3	H3	1	
BOND_TYPE	C4	H4	1	
BOND_TYPE	C10	H5	1	
BOND_TYPE	C11	H6	1	
BOND_TYPE	C12	H7	1	
BOND_TYPE	C18	H8	1	
BOND_TYPE	C19	H9	1	
BOND_TYPE	C20	H10	1	
BOND_TYPE	C22	H11	1	
BOND_TYPE	C23	H12	1	
BOND_TYPE	C24	H13	1	
BOND_TYPE	C25	H14	1	
BOND_TYPE	C26	H15	1	
BOND_TYPE	C26	H16	1	
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BOND_TYPE	N6	H18	1	
BOND_TYPE	N7	H19	1	
BOND_TYPE	O6	H20	1	
BOND_TYPE	O7	H21	1	
BOND_TYPE	O8	H22	1	
CHI 1	C2	C7	O6	H20

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PROTON_CHI 1 SAMPLES 2 0 180 EXTRA 0
CHI 2 C24 C22 O7 H21
PROTON_CHI 2 SAMPLES 3 60 -60 180 EXTRA 0
CHI 3 C22 C23 O8 H22
PROTON_CHI 3 SAMPLES 3 60 -60 180 EXTRA 0
CHI 4 N7 C21 C8 C4
CHI 5 S1 C9 C18 C20
CHI 6 C10 C13 C27 F1
CHI 7 C18 C20 C19 C15
CHI 8 O9 C26 C24 C22
CHI 9 C23 C25 N5 C12
CHI 10 S2 O9 C26 C24
CHI 11 C21 N7 S2 O2
CHI 12 N7 S2 O9 C26
NBR_ATOM C21
NBR_RADIUS 16.869989
ICOOR_INTERNAL C21 0.000000 0.000000 0.000000 C21 C8 C4
ICOOR_INTERNAL C8 0.000000 180.000000 1.467142 C21 C8 C4
ICOOR_INTERNAL C4 0.000000 53.603197 1.382530 C8 C21 C4
ICOOR_INTERNAL S1 -179.253533 69.546811 1.714309 C4 C8 C21
ICOOR_INTERNAL C9 0.329386 89.738119 1.729975 S1 C4 C8
ICOOR_INTERNAL C18 -179.992720 54.987450 1.476047 C9 S1 C4
ICOOR_INTERNAL C20 179.236740 57.142540 1.341649 C18 C9 S1
ICOOR_INTERNAL C19 179.982036 59.264291 1.481055 C20 C18 C9
ICOOR_INTERNAL C15 -179.996448 57.789060 1.343942 C19 C20 C18
ICOOR_INTERNAL C10 179.999050 59.670344 1.483658 C15 C19 C20
ICOOR_INTERNAL C13 -179.991608 59.833308 1.351148 C10 C15 C19
ICOOR_INTERNAL C5 -0.004444 61.513320 1.499877 C13 C10 C15
ICOOR_INTERNAL C1 -179.996621 57.435974 1.409032 C5 C13 C10
ICOOR_INTERNAL C2 179.998465 59.522518 1.397663 C1 C5 C13
ICOOR_INTERNAL C7 -0.004785 59.667532 1.400900 C2 C1 C5
ICOOR_INTERNAL C3 0.002290 60.326517 1.397182 C7 C2 C1
ICOOR_INTERNAL C6 0.000334 59.593488 1.398917 C3 C7 C2
ICOOR_INTERNAL O4 -179.995549 60.901929 1.354343 C6 C3 C7
ICOOR_INTERNAL H3 -179.998838 60.183084 1.083274 C3 C7 C6
ICOOR_INTERNAL O6 179.996981 59.186210 1.348361 C7 C2 C3
ICOOR_INTERNAL H20 0.003741 57.860590 0.967914 O6 C7 C2
ICOOR_INTERNAL H2 -179.998714 60.540712 1.082888 C2 C1 C7
ICOOR_INTERNAL H1 179.997594 58.496322 1.080365 C1 C5 C2
ICOOR_INTERNAL C27 179.996201 59.614613 1.519435 C13 C10 C5
ICOOR_INTERNAL F1 179.999197 67.754658 1.385061 C27 C13 C10
ICOOR_INTERNAL F2 -120.033162 70.462510 1.382985 C27 C13 F1
ICOOR_INTERNAL F3 -119.934302 70.463260 1.383016 C27 C13 F2
ICOOR_INTERNAL H5 179.994447 61.221673 1.086269 C10 C15 C13
ICOOR_INTERNAL H9 179.988904 61.465856 1.087703 C19 C20 C15
ICOOR_INTERNAL H10 -179.987437 59.955724 1.085672 C20 C18 C19
ICOOR_INTERNAL H8 179.984007 62.342041 1.087001 C18 C9 C20
ICOOR_INTERNAL N1 179.286483 66.355734 1.317806 C9 S1 C18
ICOOR_INTERNAL H4 179.166138 54.100665 1.081209 C4 C8 S1
ICOOR_INTERNAL N7 179.999622 66.137129 1.468346 C21 C8 C4
ICOOR_INTERNAL S2 -179.172745 58.297716 1.649429 N7 C21 C8
ICOOR_INTERNAL O2 -83.826517 72.116620 1.437749 S2 N7 C21
ICOOR_INTERNAL O3 -110.983663 68.091418 1.445053 S2 N7 O2
ICOOR_INTERNAL O9 -129.798996 67.886550 1.508919 S2 N7 O3
ICOOR_INTERNAL C26 47.974515 59.844760 1.411599 O9 S2 N7

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ICOOR_INTERNAL	C24	157.659727	70.735303	1.511000	C26	O9	S2
ICOOR_INTERNAL	C22	-162.087031	70.934522	1.469223	C24	C26	O9
ICOOR_INTERNAL	C23	-144.252608	77.196882	1.465612	C22	C24	C26
ICOOR_INTERNAL	C25	31.365562	77.532435	1.475932	C23	C22	C24
ICOOR_INTERNAL	N5	91.986334	65.428356	1.445367	C25	C23	C22
ICOOR_INTERNAL	C12	-93.616101	52.734633	1.352203	N5	C25	C23
ICOOR_INTERNAL	N4	-179.129993	69.969290	1.324719	C12	N5	C25
ICOOR_INTERNAL	C14	0.432095	70.198470	1.328585	N4	C12	N5
ICOOR_INTERNAL	C16	-0.108979	73.546376	1.420752	C14	N4	C12
ICOOR_INTERNAL	N3	179.722517	60.629252	1.342146	C16	C14	N4
ICOOR_INTERNAL	C11	0.610914	59.624551	1.326052	N3	C16	C14
ICOOR_INTERNAL	N2	-0.850532	57.608301	1.331525	C11	N3	C16
ICOOR_INTERNAL	C17	0.549648	58.122539	1.342123	N2	C11	N3
ICOOR_INTERNAL	N6	-179.889809	60.506863	1.449372	C17	N2	C11
ICOOR_INTERNAL	H17	179.858300	60.000088	0.984533	N6	C17	N2
ICOOR_INTERNAL	H18	-179.994767	60.001461	0.984518	N6	C17	H17
ICOOR_INTERNAL	H6	-179.998716	61.190168	1.032048	C11	N3	N2
ICOOR_INTERNAL	H7	179.996347	55.015832	1.031901	C12	N5	N4
ICOOR_INTERNAL	O5	-122.232955	71.729727	1.413896	C25	C23	N5
ICOOR_INTERNAL	H14	-121.126627	73.236014	1.070090	C25	C23	O5
ICOOR_INTERNAL	O8	-116.509427	66.306380	1.374434	C23	C22	C25
ICOOR_INTERNAL	H22	-179.997499	70.525025	0.969944	O8	C23	C22
ICOOR_INTERNAL	H12	-118.788948	69.018463	1.070044	C23	C22	O8
ICOOR_INTERNAL	O7	-122.096059	69.952681	1.380155	C22	C24	C23
ICOOR_INTERNAL	H21	-64.965720	70.523320	0.970083	O7	C22	C24
ICOOR_INTERNAL	H11	-117.021116	64.832231	1.070051	C22	C24	O7
ICOOR_INTERNAL	H13	119.997304	70.465137	1.070087	C24	C26	C22
ICOOR_INTERNAL	H15	-119.965789	70.478627	1.070027	C26	O9	C24
ICOOR_INTERNAL	H16	-119.933086	70.407815	1.069992	C26	O9	H15
ICOOR_INTERNAL	H19	179.999381	60.851423	0.984579	N7	C21	S2
ICOOR_INTERNAL	O1	-177.781693	57.290531	1.227179	C21	C8	N7

PDB\_ROTAMERS LCBb\_conformers.pdb

The CouLuc-3-OH .params file for the E isomer is below.

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NAME LCB
IO_STRING LCB Z
TYPE LIGAND
AA UNK
ATOM C21 CNH2 X 0.89
ATOM C8 aroC X 0.14
ATOM C4 aroC X -0.11
ATOM S1 S X -0.08
ATOM C9 aroC X 0.33
ATOM C18 aroC X -0.10
ATOM C20 aroC X -0.15
ATOM C19 aroC X -0.15
ATOM C15 aroC X 0.08
ATOM C10 aroC X -0.15
ATOM C13 aroC X -0.17
ATOM C5 aroC X 0.03
ATOM C1 aroC X -0.15
ATOM C2 aroC X -0.15
ATOM C7 aroC X 0.08
ATOM C3 aroC X -0.15

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ATOM	C6	aroC	X	0.08
ATOM	O4	Oaro	X	-0.16
ATOM	H3	Haro	X	0.15
ATOM	O6	OH	X	-0.53
ATOM	H20	Hpol	X	0.45
ATOM	H2	Haro	X	0.15
ATOM	H1	Haro	X	0.15
ATOM	C27	CH1	X	1.16
ATOM	F1	F	X	-0.34
ATOM	F2	F	X	-0.34
ATOM	F3	F	X	-0.34
ATOM	H5	Haro	X	0.15
ATOM	H9	Haro	X	0.15
ATOM	H10	Haro	X	0.15
ATOM	H8	Haro	X	0.15
ATOM	N1	Nhis	X	-0.57
ATOM	H4	Haro	X	0.15
ATOM	N7	Ntrp	X	-0.79
ATOM	S2	S	X	1.62
ATOM	O2	OOC	X	-0.65
ATOM	O3	OOC	X	-0.65
ATOM	O9	OH	X	-0.46
ATOM	C26	CH2	X	0.28
ATOM	C24	CH1	X	0.28
ATOM	C22	CH1	X	0.28
ATOM	C23	CH1	X	0.28
ATOM	C25	CH1	X	0.63
ATOM	N5	Npro	X	-0.08
ATOM	C12	aroC	X	0.55
ATOM	N4	Nhis	X	-0.66
ATOM	C14	aroC	X	0.46
ATOM	C16	aroC	X	0.78
ATOM	N3	Nhis	X	-0.66
ATOM	C11	aroC	X	0.60
ATOM	N2	Nhis	X	-0.65
ATOM	C17	CH1	X	0.51
ATOM	N6	NH2O	X	-0.12
ATOM	H17	Hpol	X	0.40
ATOM	H18	Hpol	X	0.40
ATOM	H6	Haro	X	0.06
ATOM	H7	Haro	X	0.06
ATOM	O5	OH	X	-0.56
ATOM	H14	Hapo	X	0.00
ATOM	O8	OH	X	-0.68
ATOM	H22	Hpol	X	0.40
ATOM	H12	Hapo	X	0.00
ATOM	O7	OH	X	-0.68
ATOM	H21	Hpol	X	0.40
ATOM	H11	Hapo	X	0.00
ATOM	H13	Hapo	X	0.00
ATOM	H15	Hapo	X	0.00
ATOM	H16	Hapo	X	0.00
ATOM	H19	Hpol	X	0.42
ATOM	O1	ONH2	X	-0.57
BOND_TYPE	C1	C2	4	

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BOND_TYPE	C6	O4	1
BOND_TYPE	C7	O6	1
BOND_TYPE	C8	C21	1
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BOND_TYPE	C14	C17	1
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BOND_TYPE	C15	C19	2
BOND_TYPE	C15	O4	1
BOND_TYPE	C16	N3	2
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BOND_TYPE	C17	N2	1
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BOND_TYPE	C18	C20	2
BOND_TYPE	C19	C20	1
BOND_TYPE	C21	N7	4
BOND_TYPE	C21	O1	2
BOND_TYPE	C22	C23	1
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BOND_TYPE	C22	O7	1
BOND_TYPE	C23	C25	1
BOND_TYPE	C23	O8	1
BOND_TYPE	C24	C26	1
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BOND_TYPE	C27	F3	1
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BOND_TYPE	O2	S2	2
BOND_TYPE	O3	S2	2
BOND_TYPE	O9	S2	1
BOND_TYPE	C1	H1	1
BOND_TYPE	C2	H2	1
BOND_TYPE	C3	H3	1

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BOND_TYPE  C11 H6  1
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BOND_TYPE  C19 H9  1
BOND_TYPE  C20 H10 1
BOND_TYPE  C22 H11 1
BOND_TYPE  C23 H12 1
BOND_TYPE  C24 H13 1
BOND_TYPE  C25 H14 1
BOND_TYPE  C26 H15 1
BOND_TYPE  C26 H16 1
BOND_TYPE  N6  H17 1
BOND_TYPE  N6  H18 1
BOND_TYPE  N7  H19 1
BOND_TYPE  O6  H20 1
BOND_TYPE  O7  H21 1
BOND_TYPE  O8  H22 1
CHI 1  C2  C7  O6  H20
PROTON_CHI 1 SAMPLES 2 0 180 EXTRA 0
CHI 2  C24 C22 O7  H21
PROTON_CHI 2 SAMPLES 3 60 -60 180 EXTRA 0
CHI 3  C22 C23 O8  H22
PROTON_CHI 3 SAMPLES 3 60 -60 180 EXTRA 0
CHI 4  N7  C21 C8  C4
CHI 5  S1  C9  C18 C20
CHI 6  C10 C13 C27 F1
CHI 7  C18 C20 C19 C15
CHI 8  O9  C26 C24 C22
CHI 9  C23 C25 N5  C12
CHI 10 S2  O9  C26 C24
CHI 11 C21 N7  S2  O2
CHI 12 N7  S2  O9  C26
NBR_ATOM  C21
NBR_RADIUS 16.869989
ICOOR_INTERNAL  C21  0.000000  0.000000  0.000000  C21  C8  C4
ICOOR_INTERNAL  C8  0.000000  180.000000  1.467142  C21  C8  C4
ICOOR_INTERNAL  C4  0.000000  53.603197  1.382530  C8  C21  C4
ICOOR_INTERNAL  S1 -179.253533  69.546811  1.714309  C4  C8  C21
ICOOR_INTERNAL  C9  0.329386  89.738119  1.729975  S1  C4  C8
ICOOR_INTERNAL  C18 174.789967  60.958622  1.472921  C9  S1  C4
ICOOR_INTERNAL  C20 -174.817607  52.374999  1.341456  C18 C9  S1
ICOOR_INTERNAL  C19 176.669795  63.190030  1.489780  C20 C18 C9
ICOOR_INTERNAL  C15 179.999613  52.099225  1.343116  C19 C20 C18
ICOOR_INTERNAL  C10 -3.940119  55.615188  1.487039  C15 C19 C20
ICOOR_INTERNAL  C13 -178.767537  60.324351  1.356134  C10 C15 C19
ICOOR_INTERNAL  C5  -1.908797  61.119747  1.501774  C13 C10 C15
ICOOR_INTERNAL  C1  -177.946289  57.079642  1.409867  C5  C13 C10
ICOOR_INTERNAL  C2  179.974256  59.489131  1.399093  C1  C5  C13
ICOOR_INTERNAL  C7  0.123786  59.683863  1.400785  C2  C1  C5
ICOOR_INTERNAL  C3  -0.006688  60.396556  1.396000  C7  C2  C1
ICOOR_INTERNAL  C6  -0.112866  59.618090  1.399076  C3  C7  C2
ICOOR_INTERNAL  O4  -179.840928  60.986484  1.350060  C6  C3  C7
ICOOR_INTERNAL  H3  -179.857894  60.167913  1.083461  C3  C7  C6

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ICOOR_INTERNAL	O6	-179.963095	59.088533	1.348698	C7	C2	C3
ICOOR_INTERNAL	H20	0.008392	57.812798	0.967746	O6	C7	C2
ICOOR_INTERNAL	H2	-179.987008	60.467903	1.082729	C2	C1	C7
ICOOR_INTERNAL	H1	-179.811248	58.444172	1.080036	C1	C5	C2
ICOOR_INTERNAL	C27	179.219739	59.314271	1.518084	C13	C10	C5
ICOOR_INTERNAL	F1	-179.212895	67.765697	1.382974	C27	C13	C10
ICOOR_INTERNAL	F2	-119.216427	70.652560	1.383140	C27	C13	F1
ICOOR_INTERNAL	F3	-119.858809	70.297200	1.382780	C27	C13	F2
ICOOR_INTERNAL	H5	178.116472	60.466234	1.083816	C10	C15	C13
ICOOR_INTERNAL	H9	179.518166	63.514338	1.088961	C19	C20	C15
ICOOR_INTERNAL	H10	-179.436665	59.109031	1.084693	C20	C18	C19
ICOOR_INTERNAL	H8	179.998924	65.772904	1.087972	C18	C9	C20
ICOOR_INTERNAL	N1	-175.496204	66.355734	1.317806	C9	S1	C18
ICOOR_INTERNAL	H4	178.032084	50.570774	1.080736	C4	C8	S1
ICOOR_INTERNAL	N7	179.999622	66.137129	1.468346	C21	C8	C4
ICOOR_INTERNAL	S2	-179.172745	58.297716	1.649429	N7	C21	C8
ICOOR_INTERNAL	O2	-83.826517	72.116620	1.437749	S2	N7	C21
ICOOR_INTERNAL	O3	-110.983663	68.091418	1.445053	S2	N7	O2
ICOOR_INTERNAL	O9	-129.798996	67.886550	1.508919	S2	N7	O3
ICOOR_INTERNAL	C26	47.974515	59.844760	1.411599	O9	S2	N7
ICOOR_INTERNAL	C24	157.659727	70.735303	1.511000	C26	O9	S2
ICOOR_INTERNAL	C22	-162.087031	70.934522	1.469223	C24	C26	O9
ICOOR_INTERNAL	C23	-144.252608	77.196882	1.465612	C22	C24	C26
ICOOR_INTERNAL	C25	31.365562	77.532435	1.475932	C23	C22	C24
ICOOR_INTERNAL	N5	91.986334	65.428356	1.445367	C25	C23	C22
ICOOR_INTERNAL	C12	-93.616101	52.734633	1.352203	N5	C25	C23
ICOOR_INTERNAL	N4	-179.129993	69.969290	1.324719	C12	N5	C25
ICOOR_INTERNAL	C14	0.432095	70.198470	1.328585	N4	C12	N5
ICOOR_INTERNAL	C16	-0.108979	73.546376	1.420752	C14	N4	C12
ICOOR_INTERNAL	N3	179.722517	60.629252	1.342146	C16	C14	N4
ICOOR_INTERNAL	C11	0.610914	59.624551	1.326052	N3	C16	C14
ICOOR_INTERNAL	N2	-0.850532	57.608301	1.331525	C11	N3	C16
ICOOR_INTERNAL	C17	0.549648	58.122539	1.342123	N2	C11	N3
ICOOR_INTERNAL	N6	-179.889809	60.506863	1.449372	C17	N2	C11
ICOOR_INTERNAL	H17	179.858300	60.000088	0.984533	N6	C17	N2
ICOOR_INTERNAL	H18	-179.994767	60.001461	0.984518	N6	C17	H17
ICOOR_INTERNAL	H6	-179.998716	61.190168	1.032048	C11	N3	N2
ICOOR_INTERNAL	H7	179.996347	55.015832	1.031901	C12	N5	N4
ICOOR_INTERNAL	O5	-122.232955	71.729727	1.413896	C25	C23	N5
ICOOR_INTERNAL	H14	-121.126627	73.236014	1.070090	C25	C23	O5
ICOOR_INTERNAL	O8	-116.509427	66.306380	1.374434	C23	C22	C25
ICOOR_INTERNAL	H22	-179.997499	70.525025	0.969944	O8	C23	C22
ICOOR_INTERNAL	H12	-118.788948	69.018463	1.070044	C23	C22	O8
ICOOR_INTERNAL	O7	-122.096059	69.952681	1.380155	C22	C24	C23
ICOOR_INTERNAL	H21	-64.965720	70.523320	0.970083	O7	C22	C24
ICOOR_INTERNAL	H11	-117.021116	64.832231	1.070051	C22	C24	O7
ICOOR_INTERNAL	H13	119.997304	70.465137	1.070087	C24	C26	C22
ICOOR_INTERNAL	H15	-119.965789	70.478627	1.070027	C26	O9	C24
ICOOR_INTERNAL	H16	-119.933086	70.407815	1.069992	C26	O9	H15
ICOOR_INTERNAL	H19	179.999381	60.851423	0.984579	N7	C21	S2
ICOOR_INTERNAL	O1	-177.781693	57.290531	1.227179	C21	C8	N7

PDB\_ROTAMERS LCBa\_conformers.pdb

The contents of the CouLuc-3-NH<sub>2</sub>.params file for the E isomer is below.

```

NAME LCC
IO_STRING LCC Z
TYPE LIGAND
AA UNK
ATOM C21 CNH2 X 0.89
ATOM C8 aroC X 0.14
ATOM C4 aroC X -0.11
ATOM S1 S X -0.08
ATOM C9 aroC X 0.33
ATOM C18 aroC X -0.10
ATOM C20 aroC X -0.15
ATOM C19 aroC X -0.15
ATOM C15 aroC X 0.08
ATOM C10 aroC X -0.15
ATOM C13 aroC X -0.17
ATOM C5 aroC X 0.03
ATOM C1 aroC X -0.15
ATOM C2 aroC X -0.15
ATOM C6 aroC X 0.10
ATOM C3 aroC X -0.15
ATOM C7 aroC X 0.08
ATOM O4 Oaro X -0.16
ATOM H3 Haro X 0.15
ATOM N6 NH2O X -0.90
ATOM H17 Hpol X 0.40
ATOM H18 Hpol X 0.40
ATOM H2 Haro X 0.15
ATOM H1 Haro X 0.15
ATOM C27 CH1 X 1.16
ATOM F1 F X -0.34
ATOM F2 F X -0.34
ATOM F3 F X -0.34
ATOM H5 Haro X 0.15
ATOM H9 Haro X 0.15
ATOM H10 Haro X 0.15
ATOM H8 Haro X 0.15
ATOM N1 Nhis X -0.57
ATOM H4 Haro X 0.15
ATOM N8 Ntrp X -0.79
ATOM S2 S X 1.62
ATOM O2 OOC X -0.65
ATOM O3 OOC X -0.65
ATOM O8 OH X -0.46
ATOM C26 CH2 X 0.28
ATOM C24 CH1 X 0.28
ATOM C22 CH1 X 0.28
ATOM C23 CH1 X 0.28
ATOM C25 CH1 X 0.63
ATOM N5 Npro X -0.08
ATOM C12 aroC X 0.55
ATOM N4 Nhis X -0.66
ATOM C14 aroC X 0.46
ATOM C16 aroC X 0.78
ATOM N3 Nhis X -0.66

```

ATOM	C11	aroC	X	0.60
ATOM	N2	Nhis	X	-0.65
ATOM	C17	CH1	X	0.51
ATOM	N7	NH2O	X	-0.12
ATOM	H19	Hpol	X	0.40
ATOM	H20	Hpol	X	0.40
ATOM	H6	Haro	X	0.06
ATOM	H7	Haro	X	0.06
ATOM	O5	OH	X	-0.56
ATOM	H14	Hapo	X	0.00
ATOM	O7	OH	X	-0.68
ATOM	H23	Hpol	X	0.40
ATOM	H12	Hapo	X	0.00
ATOM	O6	OH	X	-0.68
ATOM	H22	Hpol	X	0.40
ATOM	H11	Hapo	X	0.00
ATOM	H13	Hapo	X	0.00
ATOM	H15	Hapo	X	0.00
ATOM	H16	Hapo	X	0.00
ATOM	H21	Hpol	X	0.42
ATOM	O1	ONH2	X	-0.57
BOND_TYPE	C1	C2	4	
BOND_TYPE	C1	C5	4	
BOND_TYPE	C2	C6	4	
BOND_TYPE	C3	C6	4	
BOND_TYPE	C3	C7	4	
BOND_TYPE	C4	C8	2	
BOND_TYPE	C4	S1	1	
BOND_TYPE	C5	C7	4	
BOND_TYPE	C5	C13	1	
BOND_TYPE	C6	N6	1	
BOND_TYPE	C7	O4	1	
BOND_TYPE	C8	C21	1	
BOND_TYPE	C8	N1	1	
BOND_TYPE	C9	C18	1	
BOND_TYPE	C9	N1	2	
BOND_TYPE	C9	S1	1	
BOND_TYPE	C10	C13	2	
BOND_TYPE	C10	C15	1	
BOND_TYPE	C11	N2	2	
BOND_TYPE	C11	N3	1	
BOND_TYPE	C12	N4	1	
BOND_TYPE	C12	N5	2	
BOND_TYPE	C13	C27	1	
BOND_TYPE	C14	C16	1	
BOND_TYPE	C14	C17	1	
BOND_TYPE	C14	N4	2	
BOND_TYPE	C15	C19	2	
BOND_TYPE	C15	O4	1	
BOND_TYPE	C16	N3	2	
BOND_TYPE	C16	N5	1	
BOND_TYPE	C17	N2	1	
BOND_TYPE	C17	N7	1	
BOND_TYPE	C18	C20	2	
BOND_TYPE	C19	C20	1	

BOND_TYPE	C21	N8	4						
BOND_TYPE	C21	O1	2						
BOND_TYPE	C22	C23	1						
BOND_TYPE	C22	C24	1						
BOND_TYPE	C22	O6	1						
BOND_TYPE	C23	C25	1						
BOND_TYPE	C23	O7	1						
BOND_TYPE	C24	C26	1						
BOND_TYPE	C24	O5	1						
BOND_TYPE	C25	N5	1						
BOND_TYPE	C25	O5	1						
BOND_TYPE	C26	O8	1						
BOND_TYPE	C27	F1	1						
BOND_TYPE	C27	F2	1						
BOND_TYPE	C27	F3	1						
BOND_TYPE	N8	S2	1						
BOND_TYPE	O2	S2	2						
BOND_TYPE	O3	S2	2						
BOND_TYPE	O8	S2	1						
BOND_TYPE	C1	H1	1						
BOND_TYPE	C2	H2	1						
BOND_TYPE	C3	H3	1						
BOND_TYPE	C4	H4	1						
BOND_TYPE	C10	H5	1						
BOND_TYPE	C11	H6	1						
BOND_TYPE	C12	H7	1						
BOND_TYPE	C18	H8	1						
BOND_TYPE	C19	H9	1						
BOND_TYPE	C20	H10	1						
BOND_TYPE	C22	H11	1						
BOND_TYPE	C23	H12	1						
BOND_TYPE	C24	H13	1						
BOND_TYPE	C25	H14	1						
BOND_TYPE	C26	H15	1						
BOND_TYPE	C26	H16	1						
BOND_TYPE	N6	H17	1						
BOND_TYPE	N6	H18	1						
BOND_TYPE	N7	H19	1						
BOND_TYPE	N7	H20	1						
BOND_TYPE	N8	H21	1						
BOND_TYPE	O6	H22	1						
BOND_TYPE	O7	H23	1						
CHI 1	C24	C22	O6	H22					
PROTON_CHI	1	SAMPLES	3	60	-60	180	EXTRA	0	
CHI 2	C22	C23	O7	H23					
PROTON_CHI	2	SAMPLES	3	60	-60	180	EXTRA	0	
CHI 3	N8	C21	C8	C4					
CHI 4	S1	C9	C18	C20					
CHI 5	C10	C13	C27	F1					
CHI 6	C18	C20	C19	C15					
CHI 7	O8	C26	C24	C22					
CHI 8	C23	C25	N5	C12					
CHI 9	S2	O8	C26	C24					
CHI 10	C21	N8	S2	O2					
CHI 11	N8	S2	O8	C26					

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NBR_ATOM C21
NBR_RADIUS 16.870172
ICOOR_INTERNAL C21 0.000000 0.000000 0.000000 C21 C8
C4
ICOOR_INTERNAL C8 0.000000 180.000000 1.467099 C21 C8
C4
ICOOR_INTERNAL C4 0.000001 53.613584 1.382650 C8 C21
C4
ICOOR_INTERNAL S1 -
179.277244 69.545521 1.714033 C4 C8 C21
ICOOR_INTERNAL C9 0.366000 89.733106 1.730169 S1 C4
C8
ICOOR_INTERNAL C18 178.553405 53.909816 1.470946 C9 S1
C4
ICOOR_INTERNAL C20 -
179.324030 55.671334 1.340708 C18 C9 S1
ICOOR_INTERNAL C19 -
178.869499 61.124213 1.479088 C20 C18 C9
ICOOR_INTERNAL C15 179.996690 56.167418 1.341760 C19 C20
C18
ICOOR_INTERNAL C10 -
179.328007 60.815784 1.477883 C15 C19 C20
ICOOR_INTERNAL C13 179.018646 58.951715 1.346924 C10 C15
C19
ICOOR_INTERNAL C5 0.083660 62.211270 1.512953 C13 C10
C15
ICOOR_INTERNAL C1 -
179.720574 56.106976 1.406783 C5 C13 C10
ICOOR_INTERNAL C2 -
179.792112 57.923251 1.397442 C1 C5 C13
ICOOR_INTERNAL C6 -
0.121670 60.422992 1.400044 C2 C1 C5
ICOOR_INTERNAL C3 -
0.014294 60.447619 1.402716 C6 C2 C1
ICOOR_INTERNAL C7 0.064356 59.281962 1.401587 C3 C6
C2
ICOOR_INTERNAL O4 -
179.893959 61.173171 1.357107 C7 C3 C6
ICOOR_INTERNAL H3 -
179.996790 60.096543 1.083290 C3 C6 C7
ICOOR_INTERNAL N6 -
179.957136 59.786317 1.416636 C6 C2 C3
ICOOR_INTERNAL H17 -
0.010267 59.286668 1.030553 N6 C6 C2
ICOOR_INTERNAL H18 -
179.981802 59.285011 1.030602 N6 C6 H17
ICOOR_INTERNAL H2 -
179.960013 60.099881 1.083143 C2 C1 C6
ICOOR_INTERNAL H1 179.883915 56.831332 1.070336 C1 C5
C2
ICOOR_INTERNAL C27 -
179.974949 64.668326 1.528344 C13 C10 C5
ICOOR_INTERNAL F1 179.970880 64.860846 1.374701 C27 C13
C10

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ICOOR_INTERNAL F1	F2	120.803880	70.973223	1.384120	C27	C13
ICOOR_INTERNAL F2	F3	118.347981	70.969959	1.384140	C27	C13
ICOOR_INTERNAL 179.791393	H5 - 61.100806	1.088681	C10 C15	C13		
ICOOR_INTERNAL 179.288766	H9 - 62.277338	1.088530	C19 C20	C15		
ICOOR_INTERNAL C19	H10	179.084778	59.182765	1.086090	C20	C18
ICOOR_INTERNAL 178.987338	H8 - 63.499105	1.087867	C18 C9	C20		
ICOOR_INTERNAL 179.366714	N1 - 66.372571	1.317944	C9 S1	C18		
ICOOR_INTERNAL S1	H4	178.052410	50.569171	1.080507	C4	C8
ICOOR_INTERNAL C4	N8	179.999853	66.190937	1.468440	C21	C8
ICOOR_INTERNAL 179.026765	S2 - 58.301303	1.649468	N8 C21	C8		
ICOOR_INTERNAL 83.829697	O2 - 72.117552	1.437687	S2 N8	C21		
ICOOR_INTERNAL 110.978610	O3 - 68.096527	1.445149	S2 N8	O2		
ICOOR_INTERNAL 129.796645	O8 - 67.882755	1.508902	S2 N8	O3		
ICOOR_INTERNAL N8	C26	47.972090	59.846860	1.411589	O8	S2
ICOOR_INTERNAL S2	C24	157.664803	70.737827	1.511020	C26	O8
ICOOR_INTERNAL 162.089472	C22 - 70.934687	1.469290	C24 C26	O8		
ICOOR_INTERNAL 144.250635	C23 - 77.197183	1.465518	C22 C24	C26		
ICOOR_INTERNAL C24	C25	31.359911	77.525336	1.475945	C23	C22
ICOOR_INTERNAL C22	N5	91.982589	65.434427	1.445330	C25	C23
ICOOR_INTERNAL 93.612286	C12 - 52.726401	1.352130	N5 C25	C23		
ICOOR_INTERNAL 179.125529	N4 - 69.963424	1.324697	C12 N5	C25		
ICOOR_INTERNAL N5	C14	0.434724	70.201937	1.328569	N4	C12
ICOOR_INTERNAL 0.108966	C16 - 73.543980	1.420687	C14 N4	C12		
ICOOR_INTERNAL N4	N3	179.727076	60.631876	1.342184	C16	C14
ICOOR_INTERNAL C14	C11	0.605797	59.617650	1.325985	N3	C16
ICOOR_INTERNAL 0.843694	N2 - 57.615869	1.331578	C11 N3	C16		
ICOOR_INTERNAL N3	C17	0.543846	58.116997	1.342047	N2	C11

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ICOOOR_INTERNAL      N7  -
179.885925  60.505121  1.449333  C17  N2  C11
ICOOOR_INTERNAL      H19 179.857903  59.996676  0.984436  N7  C17
N2
ICOOOR_INTERNAL      H20 -
179.996362  60.003205  0.984528  N7  C17  H19
ICOOOR_INTERNAL      H6  179.991347  61.186872  1.031972  C11  N3
N2
ICOOOR_INTERNAL      H7  179.994533  55.014941  1.032043  C12  N5
N4
ICOOOR_INTERNAL      O5  -
122.220593  71.733603  1.413878  C25  C23  N5
ICOOOR_INTERNAL      H14 -
121.131406  73.232551  1.069922  C25  C23  O5
ICOOOR_INTERNAL      O7  -
116.507140  66.309831  1.374491  C23  C22  C25
ICOOOR_INTERNAL      H23 179.997609  70.527483  0.969954  O7  C23
C22
ICOOOR_INTERNAL      H12 -
120.201693  66.638819  1.112365  C23  C22  O7
ICOOOR_INTERNAL      O6  -
122.102731  69.950997  1.380152  C22  C24  C23
ICOOOR_INTERNAL      H22 -
64.962631  70.528340  0.970025  O6  C22  C24
ICOOOR_INTERNAL      H11 -
117.021009  64.837781  1.070175  C22  C24  O6
ICOOOR_INTERNAL      H13 119.990690  70.467539  1.070000  C24  C26
C22
ICOOOR_INTERNAL      H15 -
119.973622  70.474860  1.069935  C26  O8  C24
ICOOOR_INTERNAL      H16 -
119.929165  70.409662  1.070081  C26  O8  H15
ICOOOR_INTERNAL      H21 179.995484  60.852845  0.984556  N8  C21
S2
ICOOOR_INTERNAL      O1  -
177.629925  57.236261  1.227139  C21  C8  N8
PDB_ROTAMERS LCCb_conformers.pdb

```

The contents of the CouLuc-3-NH<sub>2</sub> .params file for the Z isomer is below.

```

NAME LCC
IO_STRING LCC Z
TYPE LIGAND
AA UNK
ATOM C21 CNH2 X 0.89
ATOM C8 aroC X 0.14
ATOM C4 aroC X -0.11
ATOM S1 S X -0.08
ATOM C9 aroC X 0.33
ATOM C18 aroC X -0.10
ATOM C20 aroC X -0.15
ATOM C19 aroC X -0.15
ATOM C15 aroC X 0.08
ATOM C10 aroC X -0.15
ATOM C13 aroC X -0.17

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ATOM	C5	aroC	X	0.03
ATOM	C1	aroC	X	-0.15
ATOM	C2	aroC	X	-0.15
ATOM	C6	aroC	X	0.10
ATOM	C3	aroC	X	-0.15
ATOM	C7	aroC	X	0.08
ATOM	O4	Oaro	X	-0.16
ATOM	H3	Haro	X	0.15
ATOM	N6	NH2O	X	-0.90
ATOM	H17	Hpol	X	0.40
ATOM	H18	Hpol	X	0.40
ATOM	H2	Haro	X	0.15
ATOM	H1	Haro	X	0.15
ATOM	C27	CH1	X	1.16
ATOM	F1	F	X	-0.34
ATOM	F2	F	X	-0.34
ATOM	F3	F	X	-0.34
ATOM	H5	Haro	X	0.15
ATOM	H9	Haro	X	0.15
ATOM	H10	Haro	X	0.15
ATOM	H8	Haro	X	0.15
ATOM	N1	Nhis	X	-0.57
ATOM	H4	Haro	X	0.15
ATOM	N8	Ntrp	X	-0.79
ATOM	S2	S	X	1.62
ATOM	O2	OOC	X	-0.65
ATOM	O3	OOC	X	-0.65
ATOM	O8	OH	X	-0.46
ATOM	C26	CH2	X	0.28
ATOM	C24	CH1	X	0.28
ATOM	C22	CH1	X	0.28
ATOM	C23	CH1	X	0.28
ATOM	C25	CH1	X	0.63
ATOM	N5	Npro	X	-0.08
ATOM	C12	aroC	X	0.55
ATOM	N4	Nhis	X	-0.66
ATOM	C14	aroC	X	0.46
ATOM	C16	aroC	X	0.78
ATOM	N3	Nhis	X	-0.66
ATOM	C11	aroC	X	0.60
ATOM	N2	Nhis	X	-0.65
ATOM	C17	CH1	X	0.51
ATOM	N7	NH2O	X	-0.12
ATOM	H19	Hpol	X	0.40
ATOM	H20	Hpol	X	0.40
ATOM	H6	Haro	X	0.06
ATOM	H7	Haro	X	0.06
ATOM	O5	OH	X	-0.56
ATOM	H14	Hapo	X	0.00
ATOM	O7	OH	X	-0.68
ATOM	H23	Hpol	X	0.40
ATOM	H12	Hapo	X	0.00
ATOM	O6	OH	X	-0.68
ATOM	H22	Hpol	X	0.40
ATOM	H11	Hapo	X	0.00

ATOM	H13	Hapo	X	0.00
ATOM	H15	Hapo	X	0.00
ATOM	H16	Hapo	X	0.00
ATOM	H21	Hpo1	X	0.42
ATOM	O1	ONH2	X	-0.57
BOND_TYPE	C1	C2	4	
BOND_TYPE	C1	C5	4	
BOND_TYPE	C2	C6	4	
BOND_TYPE	C3	C6	4	
BOND_TYPE	C3	C7	4	
BOND_TYPE	C4	C8	2	
BOND_TYPE	C4	S1	1	
BOND_TYPE	C5	C7	4	
BOND_TYPE	C5	C13	1	
BOND_TYPE	C6	N6	1	
BOND_TYPE	C7	O4	1	
BOND_TYPE	C8	C21	1	
BOND_TYPE	C8	N1	1	
BOND_TYPE	C9	C18	1	
BOND_TYPE	C9	N1	2	
BOND_TYPE	C9	S1	1	
BOND_TYPE	C10	C13	2	
BOND_TYPE	C10	C15	1	
BOND_TYPE	C11	N2	2	
BOND_TYPE	C11	N3	1	
BOND_TYPE	C12	N4	1	
BOND_TYPE	C12	N5	2	
BOND_TYPE	C13	C27	1	
BOND_TYPE	C14	C16	1	
BOND_TYPE	C14	C17	1	
BOND_TYPE	C14	N4	2	
BOND_TYPE	C15	C19	2	
BOND_TYPE	C15	O4	1	
BOND_TYPE	C16	N3	2	
BOND_TYPE	C16	N5	1	
BOND_TYPE	C17	N2	1	
BOND_TYPE	C17	N7	1	
BOND_TYPE	C18	C20	2	
BOND_TYPE	C19	C20	1	
BOND_TYPE	C21	N8	4	
BOND_TYPE	C21	O1	2	
BOND_TYPE	C22	C23	1	
BOND_TYPE	C22	C24	1	
BOND_TYPE	C22	O6	1	
BOND_TYPE	C23	C25	1	
BOND_TYPE	C23	O7	1	
BOND_TYPE	C24	C26	1	
BOND_TYPE	C24	O5	1	
BOND_TYPE	C25	N5	1	
BOND_TYPE	C25	O5	1	
BOND_TYPE	C26	O8	1	
BOND_TYPE	C27	F1	1	
BOND_TYPE	C27	F2	1	
BOND_TYPE	C27	F3	1	
BOND_TYPE	N8	S2	1	

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BOND_TYPE O2 S2 2
BOND_TYPE O3 S2 2
BOND_TYPE O8 S2 1
BOND_TYPE C1 H1 1
BOND_TYPE C2 H2 1
BOND_TYPE C3 H3 1
BOND_TYPE C4 H4 1
BOND_TYPE C10 H5 1
BOND_TYPE C11 H6 1
BOND_TYPE C12 H7 1
BOND_TYPE C18 H8 1
BOND_TYPE C19 H9 1
BOND_TYPE C20 H10 1
BOND_TYPE C22 H11 1
BOND_TYPE C23 H12 1
BOND_TYPE C24 H13 1
BOND_TYPE C25 H14 1
BOND_TYPE C26 H15 1
BOND_TYPE C26 H16 1
BOND_TYPE N6 H17 1
BOND_TYPE N6 H18 1
BOND_TYPE N7 H19 1
BOND_TYPE N7 H20 1
BOND_TYPE N8 H21 1
BOND_TYPE O6 H22 1
BOND_TYPE O7 H23 1
CHI 1 C24 C22 O6 H22
PROTON_CHI 1 SAMPLES 3 60 -60 180 EXTRA 0
CHI 2 C22 C23 O7 H23
PROTON_CHI 2 SAMPLES 3 60 -60 180 EXTRA 0
CHI 3 N8 C21 C8 C4
CHI 4 S1 C9 C18 C20
CHI 5 C10 C13 C27 F1
CHI 6 C18 C20 C19 C15
CHI 7 O8 C26 C24 C22
CHI 8 C23 C25 N5 C12
CHI 9 S2 O8 C26 C24
CHI 10 C21 N8 S2 O2
CHI 11 N8 S2 O8 C26
NBR_ATOM C21
NBR_RADIUS 16.870131
ICOOR_INTERNAL C21 0.000000 0.000000 0.000000 C21 C8 C4
ICOOR_INTERNAL C8 0.000000 179.999999 1.467129 C21 C8 C4
ICOOR_INTERNAL C4 0.000001 53.617032 1.382669 C8 C21 C4
ICOOR_INTERNAL S1 -179.281387 69.551912 1.714111 C4 C8 C21
ICOOR_INTERNAL C9 0.368538 89.731913 1.730053 S1 C4 C8
ICOOR_INTERNAL C18 -179.983428 55.061880 1.476720 C9 S1 C4
ICOOR_INTERNAL C20 179.108433 57.299028 1.341098 C18 C9 S1
ICOOR_INTERNAL C19 -179.997376 59.599319 1.483349 C20 C18 C9
ICOOR_INTERNAL C15 179.998708 54.652274 1.347872 C19 C20 C18
ICOOR_INTERNAL C10 -0.025247 56.152013 1.492261 C15 C19 C20
ICOOR_INTERNAL C13 -179.987967 59.792909 1.351295 C10 C15 C19
ICOOR_INTERNAL C5 0.003565 61.420008 1.497790 C13 C10 C15
ICOOR_INTERNAL C1 179.993562 57.471192 1.407653 C5 C13 C10
ICOOR_INTERNAL C2 179.998015 59.547970 1.396492 C1 C5 C13

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ICOOR_INTERNAL	C6	-0.001108	59.464747	1.399978	C2	C1	C5
ICOOR_INTERNAL	C3	0.005501	60.566274	1.399914	C6	C2	C1
ICOOR_INTERNAL	C7	0.002166	59.609795	1.398888	C3	C6	C2
ICOOR_INTERNAL	O4	179.998709	60.952031	1.353058	C7	C3	C6
ICOOR_INTERNAL	H3	179.989981	59.941927	1.083245	C3	C6	C7
ICOOR_INTERNAL	N6	179.998939	59.722662	1.416621	C6	C2	C3
ICOOR_INTERNAL	H17	0.000439	59.291948	1.030503	N6	C6	C2
ICOOR_INTERNAL	H18	179.996618	59.286299	1.030608	N6	C6	H17
ICOOR_INTERNAL	H2	-179.991617	60.851468	1.080264	C2	C1	C6
ICOOR_INTERNAL	H1	-179.993980	58.418820	1.080282	C1	C5	C2
ICOOR_INTERNAL	C27	179.982858	59.552969	1.520825	C13	C10	C5
ICOOR_INTERNAL	F1	-179.987880	67.606252	1.384916	C27	C13	C10
ICOOR_INTERNAL	F2	120.054389	70.477005	1.383014	C27	C13	F1
ICOOR_INTERNAL	F3	119.904788	70.480419	1.383074	C27	C13	F2
ICOOR_INTERNAL	H5	-179.992947	59.638075	1.080279	C10	C15	C13
ICOOR_INTERNAL	H9	-179.990293	62.633293	1.087843	C19	C20	C15
ICOOR_INTERNAL	H10	179.985237	61.090752	1.080489	C20	C18	C19
ICOOR_INTERNAL	H8	-179.976657	62.342554	1.087064	C18	C9	C20
ICOOR_INTERNAL	N1	179.169058	66.367720	1.317939	C9	S1	C18
ICOOR_INTERNAL	H4	178.050628	50.571171	1.080506	C4	C8	S1
ICOOR_INTERNAL	N8	179.999349	66.189768	1.468382	C21	C8	C4
ICOOR_INTERNAL	S2	-179.028528	58.301695	1.649480	N8	C21	C8
ICOOR_INTERNAL	O2	-83.828030	72.115214	1.437728	S2	N8	C21
ICOOR_INTERNAL	O3	-110.979907	68.095120	1.445125	S2	N8	O2
ICOOR_INTERNAL	O8	-129.796485	67.884564	1.508898	S2	N8	O3
ICOOR_INTERNAL	C26	47.972804	59.847207	1.411561	O8	S2	N8
ICOOR_INTERNAL	C24	157.660576	70.738957	1.511080	C26	O8	S2
ICOOR_INTERNAL	C22	-162.092351	70.936507	1.469262	C24	C26	O8
ICOOR_INTERNAL	C23	-144.249833	77.197932	1.465575	C22	C24	C26
ICOOR_INTERNAL	C25	31.366047	77.527376	1.475916	C23	C22	C24
ICOOR_INTERNAL	N5	91.982349	65.429978	1.445332	C25	C23	C22
ICOOR_INTERNAL	C12	-93.611780	52.728629	1.352156	N5	C25	C23
ICOOR_INTERNAL	N4	-179.129471	69.965230	1.324715	C12	N5	C25
ICOOR_INTERNAL	C14	0.433861	70.203038	1.328601	N4	C12	N5
ICOOR_INTERNAL	C16	-0.110422	73.543260	1.420679	C14	N4	C12
ICOOR_INTERNAL	N3	179.726204	60.629497	1.342216	C16	C14	N4
ICOOR_INTERNAL	C11	0.608176	59.624923	1.326020	N3	C16	C14
ICOOR_INTERNAL	N2	-0.848435	57.610071	1.331554	C11	N3	C16
ICOOR_INTERNAL	C17	0.552053	58.120141	1.342096	N2	C11	N3
ICOOR_INTERNAL	N7	-179.890730	60.506917	1.449335	C17	N2	C11
ICOOR_INTERNAL	H19	179.859865	60.001164	0.984524	N7	C17	N2
ICOOR_INTERNAL	H20	179.991250	60.000412	0.984519	N7	C17	H19
ICOOR_INTERNAL	H6	-179.999269	61.192473	1.031974	C11	N3	N2
ICOOR_INTERNAL	H7	179.997573	55.017697	1.031995	C12	N5	N4
ICOOR_INTERNAL	O5	-122.223815	71.733653	1.413900	C25	C23	N5
ICOOR_INTERNAL	H14	-121.130496	73.235732	1.069989	C25	C23	O5
ICOOR_INTERNAL	O7	-116.511459	66.309694	1.374530	C23	C22	C25
ICOOR_INTERNAL	H23	-179.997776	70.530952	0.969950	O7	C23	C22
ICOOR_INTERNAL	H12	-118.778660	69.018066	1.070015	C23	C22	O7
ICOOR_INTERNAL	O6	-122.097254	69.957154	1.380198	C22	C24	C23
ICOOR_INTERNAL	H22	-64.966530	70.526952	0.970044	O6	C22	C24
ICOOR_INTERNAL	H11	-117.021961	64.834299	1.070064	C22	C24	O6
ICOOR_INTERNAL	H13	120.000334	70.466931	1.069977	C24	C26	C22
ICOOR_INTERNAL	H15	-119.967044	70.476305	1.069971	C26	O8	C24
ICOOR_INTERNAL	H16	-119.934519	70.408073	1.070003	C26	O8	H15

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ICOOOR_INTERNAL    H21 -179.999258   60.849549   0.984516   N8    C21   S2
ICOOOR_INTERNAL    O1  -177.630746   57.240813   1.227206   C21   C8    N8
PDB_ROTAMERS      LCCa_conformers.pdb

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CouLuc-2-NEt<sub>2</sub> params file was as follows.

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NAME LCG
IO_STRING LCG Z
TYPE LIGAND
AA UNK
ATOM  C20 CNH2  X   0.81
ATOM  C16 CH1   X   0.21
ATOM  C8  aroC  X   0.08
ATOM  S1  S     X  -0.28
ATOM  C15 aroC  X   0.58
ATOM  C19 aroC  X  -0.14
ATOM  C18 aroC  X  -0.15
ATOM  C11 aroC  X   0.01
ATOM  C7  aroC  X  -0.18
ATOM  C4  aroC  X   0.03
ATOM  C1  aroC  X  -0.15
ATOM  C2  aroC  X  -0.15
ATOM  C5  aroC  X   0.10
ATOM  C3  aroC  X  -0.15
ATOM  C6  aroC  X   0.08
ATOM  O5  Oaro  X  -0.23
ATOM  C14 COO   X   0.71
ATOM  O1  OOC   X  -0.57
ATOM  H3  Haro  X   0.15
ATOM  N8  Nhis  X  -0.84
ATOM  C28 CH2   X   0.37
ATOM  C25 CH3   X   0.00
ATOM  H14 Hapo  X   0.00
ATOM  H15 Hapo  X   0.00
ATOM  H16 Hapo  X   0.00
ATOM  H22 Hapo  X   0.00
ATOM  H23 Hapo  X   0.00
ATOM  C29 CH2   X   0.37
ATOM  C26 CH3   X   0.00
ATOM  H17 Hapo  X   0.00
ATOM  H18 Hapo  X   0.00
ATOM  H19 Hapo  X   0.00
ATOM  H24 Hapo  X   0.00
ATOM  H25 Hapo  X   0.00
ATOM  H2  Haro  X   0.15
ATOM  H1  Haro  X   0.15
ATOM  H4  Haro  X   0.15
ATOM  H8  Haro  X   0.15
ATOM  H9  Haro  X   0.15
ATOM  N4  Nhis  X  -0.66
ATOM  H5  Haro  X   0.06
ATOM  N7  Ntrp  X  -0.79
ATOM  S2  S     X   1.62
ATOM  O3  OOC   X  -0.65
ATOM  O4  OOC   X  -0.65

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ATOM	O9	OH	X	-0.46
ATOM	C27	CH2	X	0.28
ATOM	C23	CH1	X	0.28
ATOM	C21	CH1	X	0.28
ATOM	C22	CH1	X	0.28
ATOM	C24	CH1	X	0.63
ATOM	N5	Npro	X	-0.08
ATOM	C10	aroC	X	0.55
ATOM	N3	Nhis	X	-0.66
ATOM	C12	aroC	X	0.46
ATOM	C13	aroC	X	0.78
ATOM	N2	Nhis	X	-0.66
ATOM	C9	aroC	X	0.60
ATOM	N1	Nhis	X	-0.65
ATOM	C17	CH1	X	0.51
ATOM	N6	NH2O	X	-0.12
ATOM	H26	Hpol	X	0.40
ATOM	H27	Hpol	X	0.40
ATOM	H6	Haro	X	0.06
ATOM	H7	Haro	X	0.06
ATOM	O6	OH	X	-0.56
ATOM	H13	Hapo	X	0.00
ATOM	O8	OH	X	-0.68
ATOM	H30	Hpol	X	0.40
ATOM	H11	Hapo	X	0.00
ATOM	O7	OH	X	-0.68
ATOM	H29	Hpol	X	0.40
ATOM	H10	Hapo	X	0.00
ATOM	H12	Hapo	X	0.00
ATOM	H20	Hapo	X	0.00
ATOM	H21	Hapo	X	0.00
ATOM	H28	Hpol	X	0.42
ATOM	O2	ONH2	X	-0.57
BOND_TYPE	C1	C2	4	
BOND_TYPE	C1	C4	4	
BOND_TYPE	C2	C5	4	
BOND_TYPE	C3	C5	4	
BOND_TYPE	C3	C6	4	
BOND_TYPE	C4	C6	4	
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BOND_TYPE	C5	N8	1	
BOND_TYPE	C6	O5	1	
BOND_TYPE	C7	C11	2	
BOND_TYPE	C8	C16	1	
BOND_TYPE	C8	S1	2	
BOND_TYPE	C9	N1	2	
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BOND_TYPE	C11	C14	1	
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BOND_TYPE	C13	N2	2	



BOND_TYPE	C13	N5	1
BOND_TYPE	C14	O1	2
BOND_TYPE	C14	O5	1
BOND_TYPE	C15	C19	1
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BOND_TYPE	C16	C20	1
BOND_TYPE	C16	N4	1
BOND_TYPE	C17	N1	1
BOND_TYPE	C17	N6	1
BOND_TYPE	C18	C19	2
BOND_TYPE	C20	N7	4
BOND_TYPE	C20	O2	2
BOND_TYPE	C21	C22	1
BOND_TYPE	C21	C23	1
BOND_TYPE	C21	O7	1
BOND_TYPE	C22	C24	1
BOND_TYPE	C22	O8	1
BOND_TYPE	C23	C27	1
BOND_TYPE	C23	O6	1
BOND_TYPE	C24	N5	1
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BOND_TYPE	C28	N8	1
BOND_TYPE	C29	N8	1
BOND_TYPE	N7	S2	1
BOND_TYPE	O3	S2	2
BOND_TYPE	O4	S2	2
BOND_TYPE	O9	S2	1
BOND_TYPE	C1	H1	1
BOND_TYPE	C2	H2	1
BOND_TYPE	C3	H3	1
BOND_TYPE	C7	H4	1
BOND_TYPE	C8	H5	1
BOND_TYPE	C9	H6	1
BOND_TYPE	C10	H7	1
BOND_TYPE	C18	H8	1
BOND_TYPE	C19	H9	1
BOND_TYPE	C21	H10	1
BOND_TYPE	C22	H11	1
BOND_TYPE	C23	H12	1
BOND_TYPE	C24	H13	1
BOND_TYPE	C25	H14	1
BOND_TYPE	C25	H15	1
BOND_TYPE	C25	H16	1
BOND_TYPE	C26	H17	1
BOND_TYPE	C26	H18	1
BOND_TYPE	C26	H19	1
BOND_TYPE	C27	H20	1
BOND_TYPE	C27	H21	1
BOND_TYPE	C28	H22	1
BOND_TYPE	C28	H23	1
BOND_TYPE	C29	H24	1

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BOND_TYPE  C29  H25  1
BOND_TYPE  N6   H26  1
BOND_TYPE  N6   H27  1
BOND_TYPE  N7   H28  1
BOND_TYPE  O7   H29  1
BOND_TYPE  O8   H30  1
CHI 1  C23  C21  O7   H29
#PROTON_CHI 1 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 2  C21  C22  O8   H30
#PROTON_CHI 1 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 3  C2   C5   N8   C28
#PROTON_CHI 3 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 4  C19  C18  C11  C7
#PROTON_CHI 4 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 5  S1   C15  C19  C18
#PROTON_CHI 5 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 6  N7   C20  C16  C8
CHI 7  O9   C27  C23  C21
CHI 8  C22  C24  N5   C10
CHI 9  S2   O9   C27  C23
CHI 10 C5    N8   C28  C25
#PROTON_CHI 10 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 11 C5    N8   C29  C26
#PROTON_CHI 11 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 12 C20  N7   S2   O3
CHI 13 N7   S2   O9   C27
NBR_ATOM  C20
NBR_RADIUS 18.224748
ICOOR_INTERNAL  C20  0.000000  0.000000  0.000000  C20  C16
C8
ICOOR_INTERNAL  C16  0.000000  180.000000  1.496021  C20  C16
C8
ICOOR_INTERNAL  C8   0.000000  59.872347  1.372318  C16  C20
C8
ICOOR_INTERNAL  S1  -179.270493  77.666388  1.751132  C8   C16
C20
ICOOR_INTERNAL  C15  0.492363  84.355877  1.747855  S1   C8
C16
ICOOR_INTERNAL  C19 -178.965358  51.788039  1.472562  C15  S1
C8
ICOOR_INTERNAL  C18  -1.305325  59.030868  1.347370  C19  C15
S1
ICOOR_INTERNAL  C11 -179.274860  53.166278  1.497084  C18  C19
C15
ICOOR_INTERNAL  C7   179.451266  63.097829  1.404762  C11  C18
C19
ICOOR_INTERNAL  C4  -179.793688  59.532570  1.395584  C7   C11
C18
ICOOR_INTERNAL  C1   179.796054  58.392558  1.400612  C4   C7
C11
ICOOR_INTERNAL  C2  -179.921545  60.040291  1.397529  C1   C4
C7
ICOOR_INTERNAL  C5  -0.116957  59.289576  1.406615  C2   C1
C4

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ICOOR_INTERNAL C1	C3	0.075829	61.084161	1.406503	C5	C2
ICOOR_INTERNAL C2	C6	0.052323	59.387740	1.398536	C3	C5
ICOOR_INTERNAL C5	O5	179.736043	60.042539	1.327524	C6	C3
ICOOR_INTERNAL C3	C14	-179.964796	56.803196	1.330734	O5	C6
ICOOR_INTERNAL C6	O1	-179.927603	61.884600	1.215817	C14	O5
ICOOR_INTERNAL C6	H3	178.842523	59.779401	1.082402	C3	C5
ICOOR_INTERNAL C3	N8	-179.993686	59.454771	1.434979	C5	C2
ICOOR_INTERNAL C2	C28	-0.002457	57.690884	1.466778	N8	C5
ICOOR_INTERNAL C5	C25	-91.854520	64.095211	1.534414	C28	N8
ICOOR_INTERNAL N8	H14	60.685126	69.691289	1.110806	C25	C28
ICOOR_INTERNAL H14	H15	-118.802735	71.019176	1.110603	C25	C28
ICOOR_INTERNAL H15	H16	-118.056018	67.538937	1.105051	C25	C28
ICOOR_INTERNAL C25	H22	-118.146161	72.185330	1.112371	C28	N8
ICOOR_INTERNAL H22	H23	-117.451347	70.323011	1.111477	C28	N8
ICOOR_INTERNAL C28	C29	-179.995647	57.678678	1.466716	N8	C5
ICOOR_INTERNAL C5	C26	-91.850411	64.105071	1.534379	C29	N8
ICOOR_INTERNAL N8	H17	60.685925	67.549597	1.105211	C26	C29
ICOOR_INTERNAL H17	H18	-123.147627	69.685529	1.110768	C26	C29
ICOOR_INTERNAL H18	H19	-118.811804	71.020849	1.110601	C26	C29
ICOOR_INTERNAL C26	H24	-118.150108	72.182516	1.112383	C29	N8
ICOOR_INTERNAL H24	H25	-117.454404	70.317350	1.111475	C29	N8
ICOOR_INTERNAL C5	H2	-178.850362	61.102830	1.081904	C2	C1
ICOOR_INTERNAL C2	H1	-179.822022	59.308872	1.083577	C1	C4
ICOOR_INTERNAL C4	H4	179.822139	59.822077	1.084739	C7	C11
ICOOR_INTERNAL C11	H8	179.434073	62.837482	1.087641	C18	C19
ICOOR_INTERNAL C18	H9	-179.278443	63.724142	1.078562	C19	C15
ICOOR_INTERNAL C19	N4	178.910262	74.692820	1.329868	C15	S1

ICOOR_INTERNAL S1	H5	179.269279	46.636101	1.077222	C8	C16
ICOOR_INTERNAL C8	N7	-9.923225	63.813358	1.468378	C20	C16
ICOOR_INTERNAL C16	S2	178.603099	58.300039	1.649518	N7	C20
ICOOR_INTERNAL C20	O3	-83.829480	72.117504	1.437768	S2	N7
ICOOR_INTERNAL O3	O4	-110.978218	68.095528	1.445070	S2	N7
ICOOR_INTERNAL O4	O9	-129.803411	67.884839	1.508885	S2	N7
ICOOR_INTERNAL N7	C27	47.977324	59.848471	1.411546	O9	S2
ICOOR_INTERNAL S2	C23	157.655249	70.737770	1.511083	C27	O9
ICOOR_INTERNAL O9	C21	-162.093482	70.933727	1.469270	C23	C27
ICOOR_INTERNAL C27	C22	-144.247783	77.196739	1.465552	C21	C23
ICOOR_INTERNAL C23	C24	31.367605	77.529154	1.475957	C22	C21
ICOOR_INTERNAL C21	N5	91.983558	65.428300	1.445328	C24	C22
ICOOR_INTERNAL C22	C10	-93.613244	52.728471	1.352163	N5	C24
ICOOR_INTERNAL C24	N3	-179.130300	69.967367	1.324677	C10	N5
ICOOR_INTERNAL N5	C12	0.433704	70.197497	1.328561	N3	C10
ICOOR_INTERNAL C10	C13	-0.110640	73.547282	1.420761	C12	N3
ICOOR_INTERNAL N3	N2	179.726171	60.632752	1.342181	C13	C12
ICOOR_INTERNAL C12	C9	0.608583	59.621629	1.326036	N2	C13
ICOOR_INTERNAL C13	N1	-0.848335	57.613744	1.331577	C9	N2
ICOOR_INTERNAL N2	C17	0.553656	58.116069	1.342041	N1	C9
ICOOR_INTERNAL C9	N6	-179.889752	60.503702	1.449271	C17	N1
ICOOR_INTERNAL N1	H26	-152.990791	73.020370	1.019094	N6	C17
ICOOR_INTERNAL H26	H27	125.651839	73.003358	1.011416	N6	C17
ICOOR_INTERNAL N1	H6	-179.346779	61.198218	1.081267	C9	N2
ICOOR_INTERNAL N3	H7	176.292279	47.119880	1.018607	C10	N5
ICOOR_INTERNAL N5	O6	-122.227965	71.734996	1.413847	C24	C22
ICOOR_INTERNAL O6	H13	-118.377782	70.274077	1.114966	C24	C22

ICOOR_INTERNAL C24	O8	-116.514168	66.311108	1.374557	C22	C21
ICOOR_INTERNAL C21	H30	-179.998770	69.489626	0.994258	O8	C22
ICOOR_INTERNAL O8	H11	-120.121966	66.508026	1.112708	C22	C21
ICOOR_INTERNAL C22	O7	-122.099284	69.960468	1.380236	C21	C23
ICOOR_INTERNAL C23	H29	-64.969184	73.081754	0.992629	O7	C21
ICOOR_INTERNAL O7	H10	-126.339895	68.634703	1.120794	C21	C23
ICOOR_INTERNAL C21	H12	120.013706	68.521390	1.115529	C23	C27
ICOOR_INTERNAL C23	H20	-117.558614	72.406306	1.115829	C27	O9
ICOOR_INTERNAL H20	H21	-120.541374	68.798568	1.089626	C27	O9
ICOOR_INTERNAL S2	H28	179.382356	56.641763	1.023812	N7	C20
ICOOR_INTERNAL N7	O2	-179.988605	59.568862	1.227216	C20	C16

The NapLuc-2-NMe<sub>2</sub> params file was as follows.

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NAME NLA
IO_STRING NLA Z
TYPE LIGAND
AA UNK
ATOM C21 CNH2 X 0.89
ATOM C12 aroC X 0.14
ATOM C7 aroC X -0.11
ATOM S1 S X -0.08
ATOM C13 aroC X 0.33
ATOM C20 aroC X -0.10
ATOM C19 aroC X -0.18
ATOM C10 aroC X 0.03
ATOM C3 aroC X -0.15
ATOM C1 aroC X -0.15
ATOM C9 aroC X 0.00
ATOM C6 aroC X -0.15
ATOM C11 aroC X 0.10
ATOM C4 aroC X -0.15
ATOM C2 aroC X -0.15
ATOM C8 aroC X 0.00
ATOM C5 aroC X -0.15
ATOM H5 Haro X 0.15
ATOM H2 Haro X 0.15
ATOM H4 Haro X 0.15
ATOM N8 Nhis X -0.84
ATOM C26 CH3 X 0.37
ATOM H16 Hapo X 0.00
ATOM H17 Hapo X 0.00
ATOM H18 Hapo X 0.00
ATOM C27 CH3 X 0.37

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ATOM	H19	Hapo	X	0.00
ATOM	H20	Hapo	X	0.00
ATOM	H21	Hapo	X	0.00
ATOM	H6	Haro	X	0.15
ATOM	H1	Haro	X	0.15
ATOM	H3	Haro	X	0.15
ATOM	H10	Haro	X	0.15
ATOM	H11	Haro	X	0.15
ATOM	N1	Nhis	X	-0.57
ATOM	H7	Haro	X	0.15
ATOM	N7	Ntrp	X	-0.79
ATOM	S2	S	X	1.62
ATOM	O2	OOC	X	-0.65
ATOM	O3	OOC	X	-0.65
ATOM	O7	OH	X	-0.46
ATOM	C28	CH2	X	0.28
ATOM	C24	CH1	X	0.28
ATOM	C22	CH1	X	0.28
ATOM	C23	CH1	X	0.28
ATOM	C25	CH1	X	0.63
ATOM	N5	Npro	X	-0.08
ATOM	C15	aroC	X	0.55
ATOM	N4	Nhis	X	-0.66
ATOM	C16	aroC	X	0.46
ATOM	C17	aroC	X	0.78
ATOM	N3	Nhis	X	-0.66
ATOM	C14	aroC	X	0.60
ATOM	N2	Nhis	X	-0.65
ATOM	C18	CH1	X	0.51
ATOM	N6	NH2O	X	-0.12
ATOM	H24	Hpol	X	0.40
ATOM	H25	Hpol	X	0.40
ATOM	H8	Haro	X	0.06
ATOM	H9	Haro	X	0.06
ATOM	O4	OH	X	-0.56
ATOM	H15	Hapo	X	0.00
ATOM	O6	OH	X	-0.68
ATOM	H28	Hpol	X	0.40
ATOM	H13	Hapo	X	0.00
ATOM	O5	OH	X	-0.68
ATOM	H27	Hpol	X	0.40
ATOM	H12	Hapo	X	0.00
ATOM	H14	Hapo	X	0.00
ATOM	H22	Hapo	X	0.00
ATOM	H23	Hapo	X	0.00
ATOM	H26	Hpol	X	0.42
ATOM	O1	ONH2	X	-0.57
BOND_TYPE	C1	C3	4	
BOND_TYPE	C1	C9	4	
BOND_TYPE	C2	C4	4	
BOND_TYPE	C2	C8	4	
BOND_TYPE	C3	C10	4	
BOND_TYPE	C4	C11	4	
BOND_TYPE	C5	C8	4	
BOND_TYPE	C5	C10	4	

BOND_TYPE	C6	C9	4
BOND_TYPE	C6	C11	4
BOND_TYPE	C7	C12	2
BOND_TYPE	C7	S1	1
BOND_TYPE	C8	C9	4
BOND_TYPE	C10	C19	1
BOND_TYPE	C11	N8	1
BOND_TYPE	C12	C21	1
BOND_TYPE	C12	N1	1
BOND_TYPE	C13	C20	1
BOND_TYPE	C13	N1	2
BOND_TYPE	C13	S1	1
BOND_TYPE	C14	N2	2
BOND_TYPE	C14	N3	1
BOND_TYPE	C15	N4	1
BOND_TYPE	C15	N5	2
BOND_TYPE	C16	C17	1
BOND_TYPE	C16	C18	1
BOND_TYPE	C16	N4	2
BOND_TYPE	C17	N3	2
BOND_TYPE	C17	N5	1
BOND_TYPE	C18	N2	1
BOND_TYPE	C18	N6	1
BOND_TYPE	C19	C20	2
BOND_TYPE	C21	N7	4
BOND_TYPE	C21	O1	2
BOND_TYPE	C22	C23	1
BOND_TYPE	C22	C24	1
BOND_TYPE	C22	O5	1
BOND_TYPE	C23	C25	1
BOND_TYPE	C23	O6	1
BOND_TYPE	C24	C28	1
BOND_TYPE	C24	O4	1
BOND_TYPE	C25	N5	1
BOND_TYPE	C25	O4	1
BOND_TYPE	C26	N8	1
BOND_TYPE	C27	N8	1
BOND_TYPE	C28	O7	1
BOND_TYPE	N7	S2	1
BOND_TYPE	O2	S2	2
BOND_TYPE	O3	S2	2
BOND_TYPE	O7	S2	1
BOND_TYPE	C1	H1	1
BOND_TYPE	C2	H2	1
BOND_TYPE	C3	H3	1
BOND_TYPE	C4	H4	1
BOND_TYPE	C5	H5	1
BOND_TYPE	C6	H6	1
BOND_TYPE	C7	H7	1
BOND_TYPE	C14	H8	1
BOND_TYPE	C15	H9	1
BOND_TYPE	C19	H10	1
BOND_TYPE	C20	H11	1
BOND_TYPE	C22	H12	1
BOND_TYPE	C23	H13	1

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BOND_TYPE  C24  H14  1
BOND_TYPE  C25  H15  1
BOND_TYPE  C26  H16  1
BOND_TYPE  C26  H17  1
BOND_TYPE  C26  H18  1
BOND_TYPE  C27  H19  1
BOND_TYPE  C27  H20  1
BOND_TYPE  C27  H21  1
BOND_TYPE  C28  H22  1
BOND_TYPE  C28  H23  1
BOND_TYPE  N6   H24  1
BOND_TYPE  N6   H25  1
BOND_TYPE  N7   H26  1
BOND_TYPE  O5   H27  1
BOND_TYPE  O6   H28  1
CHI 1  C24  C22  O5   H27
#PROTON_CHI 1 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 2  C22  C23  O6   H28
#PROTON_CHI 2 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 3  C20  C19  C10  C3
#PROTON_CHI 3 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 4  C6   C11  N8   C26
#PROTON_CHI 4 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 5  N7   C21  C12  C7
CHI 6  S1   C13  C20  C19
#PROTON_CHI 6 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 7  O7   C28  C24  C22
CHI 8  C23  C25  N5   C15
CHI 9  S2   O7   C28  C24
CHI 10 C21  N7   S2   O2
CHI 11 N7   S2   O7   C28
NBR_ATOM  C21
NBR_RADIUS 16.998837
ICOOR_INTERNAL  C21  0.000000  0.000000  0.000000  C21  C12
C7
ICOOR_INTERNAL  C12  0.000000  180.000000  1.496021  C21  C12
C7
ICOOR_INTERNAL  C7   0.000000  59.872347  1.372318  C12  C21
C7
ICOOR_INTERNAL  S1  -179.270493  77.666388  1.751132  C7   C12
C21
ICOOR_INTERNAL  C13  0.492363  84.355877  1.747855  S1   C7
C12
ICOOR_INTERNAL  C20 -173.771619  61.683298  1.726380  C13  S1
C7
ICOOR_INTERNAL  C19  171.268212  57.214464  1.342253  C20  C13
S1
ICOOR_INTERNAL  C10  179.597277  55.226679  1.489884  C19  C20
C13
ICOOR_INTERNAL  C3  -160.002537  61.928409  1.404869  C10  C19
C20
ICOOR_INTERNAL  C1  -179.968581  59.087473  1.396545  C3   C10
C19
ICOOR_INTERNAL  C9   -0.006186  59.677313  1.401283  C1   C3
C10

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ICOOR_INTERNAL C3	C6	179.987822	59.415405	1.403816	C9	C1
ICOOR_INTERNAL C1	C11	-179.987152	58.846339	1.414986	C6	C9
ICOOR_INTERNAL C9	C4	-0.006973	62.221683	1.413997	C11	C6
ICOOR_INTERNAL C6	C2	0.010676	58.765899	1.396493	C4	C11
ICOOR_INTERNAL C11	C8	-0.008063	59.529665	1.399099	C2	C4
ICOOR_INTERNAL C4	C5	179.985804	59.390822	1.404579	C8	C2
ICOOR_INTERNAL C2	H5	0.004226	61.554529	1.080184	C5	C8
ICOOR_INTERNAL C8	H2	-179.996492	61.005780	1.083803	C2	C4
ICOOR_INTERNAL C2	H4	179.988820	58.440586	1.080434	C4	C11
ICOOR_INTERNAL C4	N8	179.998366	58.840031	1.445016	C11	C6
ICOOR_INTERNAL C6	C26	-179.996136	59.372524	1.457096	N8	C11
ICOOR_INTERNAL C11	H16	179.999819	67.581418	1.110235	C26	N8
ICOOR_INTERNAL H16	H17	-119.548441	70.125007	1.110555	C26	N8
ICOOR_INTERNAL H17	H18	-120.914948	70.131366	1.110560	C26	N8
ICOOR_INTERNAL C26	C27	179.987400	59.307240	1.456978	N8	C11
ICOOR_INTERNAL C11	H19	179.999960	70.119296	1.110451	C27	N8
ICOOR_INTERNAL H19	H20	-119.532614	67.589575	1.110170	C27	N8
ICOOR_INTERNAL H20	H21	-119.522494	70.131372	1.110635	C27	N8
ICOOR_INTERNAL C11	H6	179.997205	61.890803	1.081476	C6	C9
ICOOR_INTERNAL C9	H1	-179.993733	60.976718	1.083611	C1	C3
ICOOR_INTERNAL C1	H3	179.982726	59.580568	1.083325	C3	C10
ICOOR_INTERNAL C10	H10	179.898474	62.179240	1.084956	C19	C20
ICOOR_INTERNAL C19	H11	179.205681	66.627727	1.084284	C20	C13
ICOOR_INTERNAL C20	N1	173.716522	74.692820	1.329868	C13	S1
ICOOR_INTERNAL S1	H7	179.269279	46.636101	1.077222	C7	C12
ICOOR_INTERNAL C7	N7	-9.923225	63.813358	1.468378	C21	C12
ICOOR_INTERNAL C12	S2	178.603099	58.300039	1.649518	N7	C21

ICOOR_INTERNAL C21	O2	-83.829480	72.117504	1.437768	S2	N7
ICOOR_INTERNAL O2	O3	-110.978218	68.095528	1.445070	S2	N7
ICOOR_INTERNAL O3	O7	-129.803411	67.884839	1.508885	S2	N7
ICOOR_INTERNAL N7	C28	47.977324	59.848471	1.411546	O7	S2
ICOOR_INTERNAL S2	C24	157.655249	70.737770	1.511083	C28	O7
ICOOR_INTERNAL O7	C22	-162.093482	70.933727	1.469270	C24	C28
ICOOR_INTERNAL C28	C23	-144.247783	77.196739	1.465552	C22	C24
ICOOR_INTERNAL C24	C25	31.367605	77.529154	1.475957	C23	C22
ICOOR_INTERNAL C22	N5	91.983558	65.428300	1.445328	C25	C23
ICOOR_INTERNAL C23	C15	-93.613244	52.728471	1.352163	N5	C25
ICOOR_INTERNAL C25	N4	-179.130300	69.967367	1.324677	C15	N5
ICOOR_INTERNAL N5	C16	0.433704	70.197497	1.328561	N4	C15
ICOOR_INTERNAL C15	C17	-0.110640	73.547282	1.420761	C16	N4
ICOOR_INTERNAL N4	N3	179.726171	60.632752	1.342181	C17	C16
ICOOR_INTERNAL C16	C14	0.608583	59.621629	1.326036	N3	C17
ICOOR_INTERNAL C17	N2	-0.848335	57.613744	1.331577	C14	N3
ICOOR_INTERNAL N3	C18	0.553656	58.116069	1.342041	N2	C14
ICOOR_INTERNAL C14	N6	-179.889752	60.503702	1.449271	C18	N2
ICOOR_INTERNAL N2	H24	-152.990791	73.020370	1.019094	N6	C18
ICOOR_INTERNAL H24	H25	125.651839	73.003358	1.011416	N6	C18
ICOOR_INTERNAL N2	H8	-179.346779	61.198218	1.081267	C14	N3
ICOOR_INTERNAL N4	H9	176.292279	47.119880	1.018607	C15	N5
ICOOR_INTERNAL N5	O4	-122.227965	71.734996	1.413847	C25	C23
ICOOR_INTERNAL O4	H15	-118.377782	70.274077	1.114966	C25	C23
ICOOR_INTERNAL C25	O6	-116.514168	66.311108	1.374557	C23	C22
ICOOR_INTERNAL C22	H28	-179.998770	69.489626	0.994258	O6	C23
ICOOR_INTERNAL O6	H13	-120.121966	66.508026	1.112708	C23	C22

ICOOR_INTERNAL C23	O5	-122.099284	69.960468	1.380236	C22	C24
ICOOR_INTERNAL C24	H27	-64.969184	73.081754	0.992629	O5	C22
ICOOR_INTERNAL O5	H12	-126.339895	68.634703	1.120794	C22	C24
ICOOR_INTERNAL C22	H14	120.013706	68.521390	1.115529	C24	C28
ICOOR_INTERNAL C24	H22	-117.558614	72.406306	1.115829	C28	O7
ICOOR_INTERNAL H22	H23	-120.541374	68.798568	1.089626	C28	O7
ICOOR_INTERNAL S2	H26	179.382356	56.641763	1.023812	N7	C21
ICOOR_INTERNAL N7	O1	-179.988605	59.568862	1.227216	C21	C12

## FPLuc 1b

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NAME RSB
IO_STRING RSB Z
TYPE LIGAND
AA UNK
ATOM C19 CNH2 X 0.81
ATOM C16 CH1 X 0.21
ATOM C7 aroC X 0.08
ATOM S1 S X -0.28
ATOM C14 aroC X 0.59
ATOM C12 aroC X 0.51
ATOM N4 Nhis X -0.62
ATOM C11 aroC X 0.19
ATOM C15 CNH2 X 0.62
ATOM N6 Npro X -0.42
ATOM C24 CH3 X 0.30
ATOM H13 Hapo X 0.00
ATOM H14 Hapo X 0.00
ATOM H15 Hapo X 0.00
ATOM O1 ONH2 X -0.57
ATOM C18 aroC X -0.18
ATOM C5 aroC X 0.03
ATOM C1 aroC X -0.15
ATOM C3 aroC X -0.15
ATOM C6 aroC X 0.10
ATOM C4 aroC X -0.15
ATOM C2 aroC X -0.15
ATOM H2 Haro X 0.15
ATOM H4 Haro X 0.15
ATOM N10 Nhis X -0.84
ATOM C25 CH3 X 0.37
ATOM H16 Hapo X 0.00
ATOM H17 Hapo X 0.00
ATOM H18 Hapo X 0.00

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ATOM	C26	CH3	X	0.37
ATOM	H19	Hapo	X	0.00
ATOM	H20	Hapo	X	0.00
ATOM	H21	Hapo	X	0.00
ATOM	H3	Haro	X	0.15
ATOM	H1	Haro	X	0.15
ATOM	H8	Haro	X	0.15
ATOM	N5	Nhis	X	-0.66
ATOM	H5	Haro	X	0.06
ATOM	N9	Ntrp	X	-0.79
ATOM	S2	S	X	1.62
ATOM	O3	OOC	X	-0.65
ATOM	O4	OOC	X	-0.65
ATOM	O8	OH	X	-0.46
ATOM	C27	CH2	X	0.28
ATOM	C22	CH1	X	0.28
ATOM	C20	CH1	X	0.28
ATOM	C21	CH1	X	0.28
ATOM	C23	CH1	X	0.63
ATOM	N7	Npro	X	-0.08
ATOM	C9	aroC	X	0.55
ATOM	N3	Nhis	X	-0.66
ATOM	C10	aroC	X	0.46
ATOM	C13	aroC	X	0.78
ATOM	N2	Nhis	X	-0.66
ATOM	C8	aroC	X	0.60
ATOM	N1	Nhis	X	-0.65
ATOM	C17	CH1	X	0.51
ATOM	N8	NH2O	X	-0.12
ATOM	H24	Hpol	X	0.40
ATOM	H25	Hpol	X	0.40
ATOM	H6	Haro	X	0.06
ATOM	H7	Haro	X	0.06
ATOM	O5	OH	X	-0.56
ATOM	H12	Hapo	X	0.00
ATOM	O7	OH	X	-0.68
ATOM	H28	Hpol	X	0.40
ATOM	H10	Hapo	X	0.00
ATOM	O6	OH	X	-0.68
ATOM	H27	Hpol	X	0.40
ATOM	H9	Hapo	X	0.00
ATOM	H11	Hapo	X	0.00
ATOM	H22	Hapo	X	0.00
ATOM	H23	Hapo	X	0.00
ATOM	H26	Hpol	X	0.42
ATOM	O2	ONH2	X	-0.57
BOND_TYPE	C1	C3	4	
BOND_TYPE	C1	C5	4	
BOND_TYPE	C2	C4	4	
BOND_TYPE	C2	C5	4	
BOND_TYPE	C3	C6	4	
BOND_TYPE	C4	C6	4	
BOND_TYPE	C5	C18	1	
BOND_TYPE	C6	N10	1	
BOND_TYPE	C7	C16	1	

BOND_TYPE	C7	S1	2
BOND_TYPE	C8	N1	2
BOND_TYPE	C8	N2	1
BOND_TYPE	C9	N3	1
BOND_TYPE	C9	N7	2
BOND_TYPE	C10	C13	1
BOND_TYPE	C10	C17	1
BOND_TYPE	C10	N3	2
BOND_TYPE	C11	C15	1
BOND_TYPE	C11	C18	2
BOND_TYPE	C11	N4	1
BOND_TYPE	C12	C14	1
BOND_TYPE	C12	N4	2
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BOND_TYPE	C13	N2	2
BOND_TYPE	C13	N7	1
BOND_TYPE	C14	N5	2
BOND_TYPE	C14	S1	1
BOND_TYPE	C15	N6	4
BOND_TYPE	C15	O1	2
BOND_TYPE	C16	C19	1
BOND_TYPE	C16	N5	1
BOND_TYPE	C17	N1	1
BOND_TYPE	C17	N8	1
BOND_TYPE	C19	N9	4
BOND_TYPE	C19	O2	2
BOND_TYPE	C20	C21	1
BOND_TYPE	C20	C22	1
BOND_TYPE	C20	O6	1
BOND_TYPE	C21	C23	1
BOND_TYPE	C21	O7	1
BOND_TYPE	C22	C27	1
BOND_TYPE	C22	O5	1
BOND_TYPE	C23	N7	1
BOND_TYPE	C23	O5	1
BOND_TYPE	C24	N6	1
BOND_TYPE	C25	N10	1
BOND_TYPE	C26	N10	1
BOND_TYPE	C27	O8	1
BOND_TYPE	N9	S2	1
BOND_TYPE	O3	S2	2
BOND_TYPE	O4	S2	2
BOND_TYPE	O8	S2	1
BOND_TYPE	C1	H1	1
BOND_TYPE	C2	H2	1
BOND_TYPE	C3	H3	1
BOND_TYPE	C4	H4	1
BOND_TYPE	C7	H5	1
BOND_TYPE	C8	H6	1
BOND_TYPE	C9	H7	1
BOND_TYPE	C18	H8	1
BOND_TYPE	C20	H9	1
BOND_TYPE	C21	H10	1
BOND_TYPE	C22	H11	1
BOND_TYPE	C23	H12	1

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BOND_TYPE  C24  H13  1
BOND_TYPE  C24  H14  1
BOND_TYPE  C24  H15  1
BOND_TYPE  C25  H16  1
BOND_TYPE  C25  H17  1
BOND_TYPE  C25  H18  1
BOND_TYPE  C26  H19  1
BOND_TYPE  C26  H20  1
BOND_TYPE  C26  H21  1
BOND_TYPE  C27  H22  1
BOND_TYPE  C27  H23  1
BOND_TYPE  N8   H24  1
BOND_TYPE  N8   H25  1
BOND_TYPE  N9   H26  1
BOND_TYPE  O6   H27  1
BOND_TYPE  O7   H28  1
CHI 1  C22  C20  O6   H27
PROTON_CHI 1 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 2  C20  C21  O7   H28
PROTON_CHI 2 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 3  C11  C18  C5   C1
CHI 4  C3   C6   N10  C25
CHI 5  S1   C14  C12  N4
CHI 6  N9   C19  C16  C7
CHI 7  O8   C27  C22  C20
CHI 8  C21  C23  N7   C9
CHI 9  S2   O8   C27  C22
CHI 10 C19  N9   S2   O3
CHI 11 N9   S2   O8   C27
NBR_ATOM  C19
NBR_RADIUS 16.914558
ICOOR_INTERNAL  C19  0.000000  0.000000  0.000000  C19  C16
C7
ICOOR_INTERNAL  C16  0.000000  180.000000  1.496095  C19  C16
C7
ICOOR_INTERNAL  C7   0.000000  59.870629  1.372257  C16  C19
C7
ICOOR_INTERNAL  S1  -179.266651  77.664513  1.751119  C7   C16
C19
ICOOR_INTERNAL  C14  0.491524  84.357409  1.747864  S1   C7
C16
ICOOR_INTERNAL  C12  177.286844  49.940206  1.622431  C14  S1
C7
ICOOR_INTERNAL  N4   3.119648  68.967221  1.289347  C12  C14
S1
ICOOR_INTERNAL  C11  179.585573  69.443259  1.404266  N4   C12
C14
ICOOR_INTERNAL  C15  0.217288  73.173984  1.447123  C11  N4
C12
ICOOR_INTERNAL  N6   0.110403  69.777641  1.351244  C15  C11
N4
ICOOR_INTERNAL  C24  178.762000  62.670855  1.536108  N6   C15
C11
ICOOR_INTERNAL  H13  -88.646762  91.959140  1.070045  C24  N6
C15

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ICOOR_INTERNAL H13	H14	-116.010915	68.624428	1.110449	C24	N6
ICOOR_INTERNAL H14	H15	-128.473712	68.605163	1.110298	C24	N6
ICOOR_INTERNAL N6	O1	179.476428	54.497599	1.223134	C15	C11
ICOOR_INTERNAL C15	C18	179.948502	53.009608	1.348754	C11	N4
ICOOR_INTERNAL N4	C5	-0.005324	52.277498	1.495927	C18	C11
ICOOR_INTERNAL C11	C1	119.973951	55.415443	1.406595	C5	C18
ICOOR_INTERNAL C18	C3	179.939001	59.106594	1.398201	C1	C5
ICOOR_INTERNAL C5	C6	0.170504	58.470265	1.414210	C3	C1
ICOOR_INTERNAL C1	C4	-0.075253	63.008161	1.413693	C6	C3
ICOOR_INTERNAL C3	C2	-0.087128	58.688472	1.396733	C4	C6
ICOOR_INTERNAL C6	H2	-179.642280	61.602795	1.083640	C2	C4
ICOOR_INTERNAL C2	H4	-178.616178	57.596882	1.077836	C4	C6
ICOOR_INTERNAL C4	N10	179.989872	58.488243	1.449368	C6	C3
ICOOR_INTERNAL C3	C25	-0.005764	57.461630	1.457935	N10	C6
ICOOR_INTERNAL C6	H16	-179.999783	65.758567	1.102772	C25	N10
ICOOR_INTERNAL H16	H17	-122.593261	70.093874	1.110752	C25	N10
ICOOR_INTERNAL H17	H18	-119.299184	70.638298	1.110932	C25	N10
ICOOR_INTERNAL C25	C26	-179.999514	57.472830	1.457900	N10	C6
ICOOR_INTERNAL C6	H19	179.998676	65.765352	1.102846	C26	N10
ICOOR_INTERNAL H19	H20	-122.588868	70.088248	1.110753	C26	N10
ICOOR_INTERNAL H20	H21	-119.298877	70.638601	1.110919	C26	N10
ICOOR_INTERNAL C6	H3	178.680828	63.778744	1.077983	C3	C1
ICOOR_INTERNAL C3	H1	179.779170	57.524698	1.074940	C1	C5
ICOOR_INTERNAL C5	H8	179.968044	63.933978	1.087264	C18	C11
ICOOR_INTERNAL C12	N5	-177.338451	74.691871	1.329846	C14	S1
ICOOR_INTERNAL S1	H5	179.272402	46.637967	1.077202	C7	C16
ICOOR_INTERNAL C7	N9	-9.928572	63.816343	1.468382	C19	C16

ICOOR_INTERNAL C16	S2	178.603415	58.301695	1.649480	N9	C19
ICOOR_INTERNAL C19	O3	-83.828030	72.115214	1.437728	S2	N9
ICOOR_INTERNAL O3	O4	-110.979907	68.095120	1.445125	S2	N9
ICOOR_INTERNAL O4	O8	-129.796485	67.884564	1.508898	S2	N9
ICOOR_INTERNAL N9	C27	47.972804	59.847207	1.411561	O8	S2
ICOOR_INTERNAL S2	C22	157.660576	70.738957	1.511080	C27	O8
ICOOR_INTERNAL O8	C20	-162.092351	70.936507	1.469262	C22	C27
ICOOR_INTERNAL C27	C21	-144.249833	77.197932	1.465575	C20	C22
ICOOR_INTERNAL C22	C23	31.366047	77.527376	1.475916	C21	C20
ICOOR_INTERNAL C20	N7	91.982349	65.429978	1.445332	C23	C21
ICOOR_INTERNAL C21	C9	-93.611780	52.728629	1.352156	N7	C23
ICOOR_INTERNAL C23	N3	-179.129471	69.965230	1.324715	C9	N7
ICOOR_INTERNAL N7	C10	0.433861	70.203038	1.328601	N3	C9
ICOOR_INTERNAL C9	C13	-0.110422	73.543260	1.420679	C10	N3
ICOOR_INTERNAL N3	N2	179.726204	60.629497	1.342216	C13	C10
ICOOR_INTERNAL C10	C8	0.608176	59.624923	1.326020	N2	C13
ICOOR_INTERNAL C13	N1	-0.848435	57.610071	1.331554	C8	N2
ICOOR_INTERNAL N2	C17	0.552053	58.120141	1.342096	N1	C8
ICOOR_INTERNAL C8	N8	-179.890730	60.506917	1.449335	C17	N1
ICOOR_INTERNAL N1	H24	179.876742	59.668192	1.030159	N8	C17
ICOOR_INTERNAL H24	H25	179.955977	59.596310	1.030220	N8	C17
ICOOR_INTERNAL N1	H6	-179.344486	61.198071	1.081258	C8	N2
ICOOR_INTERNAL N3	H7	177.596100	51.693780	1.047707	C9	N7
ICOOR_INTERNAL N7	O5	-122.223815	71.733653	1.413900	C23	C21
ICOOR_INTERNAL O5	H12	-118.383618	70.268459	1.114938	C23	C21
ICOOR_INTERNAL C23	O7	-116.511459	66.309694	1.374530	C21	C20
ICOOR_INTERNAL C20	H28	-179.999072	69.490695	0.994264	O7	C21



ICOOR_INTERNAL O7	H10	-120.171312	66.625332	1.112440	C21	C20
ICOOR_INTERNAL C21	O6	-122.097254	69.957154	1.380198	C20	C22
ICOOR_INTERNAL C22	H27	-64.967598	73.550547	0.992744	O6	C20
ICOOR_INTERNAL O6	H9	-124.983998	69.865294	1.119984	C20	C22
ICOOR_INTERNAL C20	H11	120.706989	68.439637	1.115607	C22	C27
ICOOR_INTERNAL C22	H22	-111.245097	74.994281	1.117155	C27	O8
ICOOR_INTERNAL H22	H23	-110.732374	65.739063	1.081951	C27	O8
ICOOR_INTERNAL S2	H26	179.382979	56.649450	1.023796	N9	C19
ICOOR_INTERNAL N9	O2	-179.982443	59.569098	1.227206	C19	C16

PDB\_ROTAMERS RSB\_conformers.pdb

### FPLuc 1a:

NAME	RSC			
IO_STRING	RSC	Z		
TYPE	LIGAND			
AA	UNK			
ATOM	C19	CNH2	X	0.81
ATOM	C16	CH1	X	0.21
ATOM	C7	aroC	X	0.08
ATOM	S1	S	X	-0.28
ATOM	C14	aroC	X	0.59
ATOM	C12	aroC	X	0.51
ATOM	N4	Nhis	X	-0.62
ATOM	C11	aroC	X	0.19
ATOM	C15	CNH2	X	0.62
ATOM	N6	Npro	X	-0.42
ATOM	C26	CH3	X	0.30
ATOM	H19	Hapo	X	0.00
ATOM	H20	Hapo	X	0.00
ATOM	H21	Hapo	X	0.00
ATOM	O1	ONH2	X	-0.57
ATOM	C18	aroC	X	-0.18
ATOM	C5	aroC	X	0.03
ATOM	C1	aroC	X	-0.15
ATOM	C3	aroC	X	-0.15
ATOM	C6	aroC	X	0.10
ATOM	C4	aroC	X	-0.15
ATOM	C2	aroC	X	-0.15
ATOM	H2	Haro	X	0.15
ATOM	H4	Haro	X	0.15
ATOM	N10	Nhis	X	-0.84
ATOM	C28	CH2	X	0.37
ATOM	C24	CH3	X	0.00
ATOM	H13	Hapo	X	0.00
ATOM	H14	Hapo	X	0.00
ATOM	H15	Hapo	X	0.00

ATOM	H24	Hapo	X	0.00
ATOM	H25	Hapo	X	0.00
ATOM	C29	CH2	X	0.37
ATOM	C25	CH3	X	0.00
ATOM	H16	Hapo	X	0.00
ATOM	H17	Hapo	X	0.00
ATOM	H18	Hapo	X	0.00
ATOM	H26	Hapo	X	0.00
ATOM	H27	Hapo	X	0.00
ATOM	H3	Haro	X	0.15
ATOM	H1	Haro	X	0.15
ATOM	H8	Haro	X	0.15
ATOM	N5	Nhis	X	-0.66
ATOM	H5	Haro	X	0.06
ATOM	N9	Ntrp	X	-0.79
ATOM	S2	S	X	1.62
ATOM	O3	OOC	X	-0.65
ATOM	O4	OOC	X	-0.65
ATOM	O8	OH	X	-0.46
ATOM	C27	CH2	X	0.28
ATOM	C22	CH1	X	0.28
ATOM	C20	CH1	X	0.28
ATOM	C21	CH1	X	0.28
ATOM	C23	CH1	X	0.63
ATOM	N7	Npro	X	-0.08
ATOM	C9	aroC	X	0.55
ATOM	N3	Nhis	X	-0.66
ATOM	C10	aroC	X	0.46
ATOM	C13	aroC	X	0.78
ATOM	N2	Nhis	X	-0.66
ATOM	C8	aroC	X	0.60
ATOM	N1	Nhis	X	-0.65
ATOM	C17	CH1	X	0.51
ATOM	N8	NH2O	X	-0.12
ATOM	H28	Hpol	X	0.40
ATOM	H29	Hpol	X	0.40
ATOM	H6	Haro	X	0.06
ATOM	H7	Haro	X	0.06
ATOM	O5	OH	X	-0.56
ATOM	H12	Hapo	X	0.00
ATOM	O7	OH	X	-0.68
ATOM	H32	Hpol	X	0.40
ATOM	H10	Hapo	X	0.00
ATOM	O6	OH	X	-0.68
ATOM	H31	Hpol	X	0.40
ATOM	H9	Hapo	X	0.00
ATOM	H11	Hapo	X	0.00
ATOM	H22	Hapo	X	0.00
ATOM	H23	Hapo	X	0.00
ATOM	H30	Hpol	X	0.42
ATOM	O2	ONH2	X	-0.57
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BOND_TYPE	C1	C5	4	
BOND_TYPE	C2	C4	4	
BOND_TYPE	C2	C5	4	

BOND_TYPE	C3	C6	4
BOND_TYPE	C4	C6	4
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BOND_TYPE	C6	N10	1
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BOND_TYPE	C8	N1	2
BOND_TYPE	C8	N2	1
BOND_TYPE	C9	N3	1
BOND_TYPE	C9	N7	2
BOND_TYPE	C10	C13	1
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BOND_TYPE	C13	N2	2
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BOND_TYPE	C14	S1	1
BOND_TYPE	C15	N6	4
BOND_TYPE	C15	O1	2
BOND_TYPE	C16	C19	1
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BOND_TYPE	C17	N1	1
BOND_TYPE	C17	N8	1
BOND_TYPE	C19	N9	4
BOND_TYPE	C19	O2	2
BOND_TYPE	C20	C21	1
BOND_TYPE	C20	C22	1
BOND_TYPE	C20	O6	1
BOND_TYPE	C21	C23	1
BOND_TYPE	C21	O7	1
BOND_TYPE	C22	C27	1
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BOND_TYPE	C23	O5	1
BOND_TYPE	C24	C28	1
BOND_TYPE	C25	C29	1
BOND_TYPE	C26	N6	1
BOND_TYPE	C27	O8	1
BOND_TYPE	C28	N10	1
BOND_TYPE	C29	N10	1
BOND_TYPE	N9	S2	1
BOND_TYPE	O3	S2	2
BOND_TYPE	O4	S2	2
BOND_TYPE	O8	S2	1
BOND_TYPE	C1	H1	1
BOND_TYPE	C2	H2	1
BOND_TYPE	C3	H3	1
BOND_TYPE	C4	H4	1
BOND_TYPE	C7	H5	1

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BOND_TYPE C8 H6 1
BOND_TYPE C9 H7 1
BOND_TYPE C18 H8 1
BOND_TYPE C20 H9 1
BOND_TYPE C21 H10 1
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BOND_TYPE N9 H30 1
BOND_TYPE O6 H31 1
BOND_TYPE O7 H32 1
CHI 1 C22 C20 O6 H31
PROTON_CHI 1 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 2 C20 C21 O7 H32
PROTON_CHI 2 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 3 C11 C18 C5 C1
CHI 4 C3 C6 N10 C28
CHI 5 S1 C14 C12 N4
CHI 6 N9 C19 C16 C7
CHI 7 O8 C27 C22 C20
CHI 8 C21 C23 N7 C9
CHI 9 S2 O8 C27 C22
CHI 10 C6 N10 C28 C24
CHI 11 C6 N10 C29 C25
CHI 12 C19 N9 S2 O3
CHI 13 N9 S2 O8 C27
NBR_ATOM C19
NBR_RADIUS 18.357181
ICOOR_INTERNAL C19 0.000000 0.000000 0.000000 C19 C16
C7
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C7
ICOOR_INTERNAL C7 0.000000 59.870629 1.372257 C16 C19
C7
ICOOR_INTERNAL S1 -179.266651 77.664513 1.751119 C7 C16
C19
ICOOR_INTERNAL C14 0.491524 84.357409 1.747864 S1 C7
C16

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ICOOR_INTERNAL C7	C12	177.131963	49.951342	1.622461	C14	S1
ICOOR_INTERNAL S1	N4	3.301075	68.968546	1.289316	C12	C14
ICOOR_INTERNAL C14	C11	179.673582	69.437629	1.404149	N4	C12
ICOOR_INTERNAL C12	C15	0.161299	73.173999	1.447125	C11	N4
ICOOR_INTERNAL N4	N6	0.181994	69.783309	1.351228	C15	C11
ICOOR_INTERNAL C11	C26	178.677200	62.669354	1.536056	N6	C15
ICOOR_INTERNAL C15	H19	-88.595944	91.952084	1.070100	C26	N6
ICOOR_INTERNAL H19	H20	-116.093967	68.619238	1.110427	C26	N6
ICOOR_INTERNAL H20	H21	-128.469706	68.609586	1.110443	C26	N6
ICOOR_INTERNAL N6	O1	179.401549	54.495891	1.223158	C15	C11
ICOOR_INTERNAL C15	C18	179.860624	53.010262	1.348848	C11	N4
ICOOR_INTERNAL N4	C5	0.003284	52.305008	1.495972	C18	C11
ICOOR_INTERNAL C11	C1	120.070998	55.430092	1.406398	C5	C18
ICOOR_INTERNAL C18	C3	179.967042	59.160770	1.397509	C1	C5
ICOOR_INTERNAL C5	C6	0.212443	58.346730	1.413992	C3	C1
ICOOR_INTERNAL C1	C4	-0.403339	63.071742	1.414433	C6	C3
ICOOR_INTERNAL C3	C2	0.304625	58.736712	1.396961	C4	C6
ICOOR_INTERNAL C6	H2	-179.781615	61.616139	1.083742	C2	C4
ICOOR_INTERNAL C2	H4	-178.714813	57.320247	1.077185	C4	C6
ICOOR_INTERNAL C4	N10	-179.563712	58.921400	1.451968	C6	C3
ICOOR_INTERNAL C3	C28	-179.998648	58.921642	1.465370	N10	C6
ICOOR_INTERNAL C6	C24	-85.908602	68.563472	1.530605	C28	N10
ICOOR_INTERNAL N10	H13	56.324982	67.991405	1.107744	C24	C28
ICOOR_INTERNAL H13	H14	-120.027716	70.551019	1.110234	C24	C28
ICOOR_INTERNAL H14	H15	-119.264434	70.029409	1.110271	C24	C28
ICOOR_INTERNAL C24	H24	-124.088393	69.562449	1.109579	C28	N10
ICOOR_INTERNAL H24	H25	-116.943216	69.153819	1.112113	C28	N10

ICOOR_INTERNAL C28	C29	-177.241545	57.897030	1.465675	N10	C6
ICOOR_INTERNAL C6	C25	-81.719351	68.653456	1.528170	C29	N10
ICOOR_INTERNAL N10	H16	-55.348410	70.204033	1.110611	C25	C29
ICOOR_INTERNAL H16	H17	-120.954177	68.560352	1.109065	C25	C29
ICOOR_INTERNAL H17	H18	-119.661097	70.470188	1.110361	C25	C29
ICOOR_INTERNAL C25	H26	121.715185	66.938713	1.102697	C29	N10
ICOOR_INTERNAL H26	H27	116.510841	70.745118	1.111779	C29	N10
ICOOR_INTERNAL C6	H3	179.417917	63.355002	1.079223	C3	C1
ICOOR_INTERNAL C3	H1	179.940498	57.463167	1.074843	C1	C5
ICOOR_INTERNAL C5	H8	-179.924366	63.925149	1.087298	C18	C11
ICOOR_INTERNAL C12	N5	-177.183569	74.691871	1.329846	C14	S1
ICOOR_INTERNAL S1	H5	179.272402	46.637967	1.077202	C7	C16
ICOOR_INTERNAL C7	N9	-9.928572	63.816343	1.468382	C19	C16
ICOOR_INTERNAL C16	S2	178.603415	58.301695	1.649480	N9	C19
ICOOR_INTERNAL C19	O3	-83.828030	72.115214	1.437728	S2	N9
ICOOR_INTERNAL O3	O4	-110.979907	68.095120	1.445125	S2	N9
ICOOR_INTERNAL O4	O8	-129.796485	67.884564	1.508898	S2	N9
ICOOR_INTERNAL N9	C27	47.972804	59.847207	1.411561	O8	S2
ICOOR_INTERNAL S2	C22	157.660576	70.738957	1.511080	C27	O8
ICOOR_INTERNAL O8	C20	-162.092351	70.936507	1.469262	C22	C27
ICOOR_INTERNAL C27	C21	-144.249833	77.197932	1.465575	C20	C22
ICOOR_INTERNAL C22	C23	31.366047	77.527376	1.475916	C21	C20
ICOOR_INTERNAL C20	N7	91.982349	65.429978	1.445332	C23	C21
ICOOR_INTERNAL C21	C9	-93.611780	52.728629	1.352156	N7	C23
ICOOR_INTERNAL C23	N3	-179.129471	69.965230	1.324715	C9	N7
ICOOR_INTERNAL N7	C10	0.433861	70.203038	1.328601	N3	C9
ICOOR_INTERNAL C9	C13	-0.110422	73.543260	1.420679	C10	N3

ICOOR_INTERNAL N3	N2	179.726204	60.629497	1.342216	C13	C10
ICOOR_INTERNAL C10	C8	0.608176	59.624923	1.326020	N2	C13
ICOOR_INTERNAL C13	N1	-0.848435	57.610071	1.331554	C8	N2
ICOOR_INTERNAL N2	C17	0.552053	58.120141	1.342096	N1	C8
ICOOR_INTERNAL C8	N8	-179.890730	60.506917	1.449335	C17	N1
ICOOR_INTERNAL N1	H28	179.876742	59.668192	1.030159	N8	C17
ICOOR_INTERNAL H28	H29	179.955977	59.596310	1.030220	N8	C17
ICOOR_INTERNAL N1	H6	-179.344486	61.198071	1.081258	C8	N2
ICOOR_INTERNAL N3	H7	177.596100	51.693780	1.047707	C9	N7
ICOOR_INTERNAL N7	O5	-122.223815	71.733653	1.413900	C23	C21
ICOOR_INTERNAL O5	H12	-118.383618	70.268459	1.114938	C23	C21
ICOOR_INTERNAL C23	O7	-116.511459	66.309694	1.374530	C21	C20
ICOOR_INTERNAL C20	H32	-179.999072	69.490695	0.994264	O7	C21
ICOOR_INTERNAL O7	H10	-120.171312	66.625332	1.112440	C21	C20
ICOOR_INTERNAL C21	O6	-122.097254	69.957154	1.380198	C20	C22
ICOOR_INTERNAL C22	H31	-64.972041	73.546813	0.992763	O6	C20
ICOOR_INTERNAL O6	H9	-124.983998	69.865294	1.119984	C20	C22
ICOOR_INTERNAL C20	H11	120.706989	68.439637	1.115607	C22	C27
ICOOR_INTERNAL C22	H22	-111.245097	74.994281	1.117155	C27	O8
ICOOR_INTERNAL H22	H23	-110.732374	65.739063	1.081951	C27	O8
ICOOR_INTERNAL S2	H30	179.382979	56.649450	1.023796	N9	C19
ICOOR_INTERNAL N9	O2	-179.982443	59.569098	1.227206	C19	C16

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### CouLuc-2-NEt<sub>2</sub> params file:

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NAME LCG
IO_STRING LCG Z
TYPE LIGAND
AA UNK
ATOM C20 CNH2 X 0.81
ATOM C16 CH1 X 0.21
ATOM C8 aroC X 0.08

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ATOM	S1	S	X	-0.28
ATOM	C15	aroC	X	0.58
ATOM	C19	aroC	X	-0.14
ATOM	C18	aroC	X	-0.15
ATOM	C11	aroC	X	0.01
ATOM	C7	aroC	X	-0.18
ATOM	C4	aroC	X	0.03
ATOM	C1	aroC	X	-0.15
ATOM	C2	aroC	X	-0.15
ATOM	C5	aroC	X	0.10
ATOM	C3	aroC	X	-0.15
ATOM	C6	aroC	X	0.08
ATOM	O5	Oaro	X	-0.23
ATOM	C14	COO	X	0.71
ATOM	O1	OOC	X	-0.57
ATOM	H3	Haro	X	0.15
ATOM	N8	Nhis	X	-0.84
ATOM	C28	CH2	X	0.37
ATOM	C25	CH3	X	0.00
ATOM	H14	Hapo	X	0.00
ATOM	H15	Hapo	X	0.00
ATOM	H16	Hapo	X	0.00
ATOM	H22	Hapo	X	0.00
ATOM	H23	Hapo	X	0.00
ATOM	C29	CH2	X	0.37
ATOM	C26	CH3	X	0.00
ATOM	H17	Hapo	X	0.00
ATOM	H18	Hapo	X	0.00
ATOM	H19	Hapo	X	0.00
ATOM	H24	Hapo	X	0.00
ATOM	H25	Hapo	X	0.00
ATOM	H2	Haro	X	0.15
ATOM	H1	Haro	X	0.15
ATOM	H4	Haro	X	0.15
ATOM	H8	Haro	X	0.15
ATOM	H9	Haro	X	0.15
ATOM	N4	Nhis	X	-0.66
ATOM	H5	Haro	X	0.06
ATOM	N7	Ntrp	X	-0.79
ATOM	S2	S	X	1.62
ATOM	O3	OOC	X	-0.65
ATOM	O4	OOC	X	-0.65
ATOM	O9	OH	X	-0.46
ATOM	C27	CH2	X	0.28
ATOM	C23	CH1	X	0.28
ATOM	C21	CH1	X	0.28
ATOM	C22	CH1	X	0.28
ATOM	C24	CH1	X	0.63
ATOM	N5	Npro	X	-0.08
ATOM	C10	aroC	X	0.55
ATOM	N3	Nhis	X	-0.66
ATOM	C12	aroC	X	0.46
ATOM	C13	aroC	X	0.78
ATOM	N2	Nhis	X	-0.66
ATOM	C9	aroC	X	0.60



ATOM	N1	Nhis	X	-0.65
ATOM	C17	CH1	X	0.51
ATOM	N6	NH2O	X	-0.12
ATOM	H26	Hpol	X	0.40
ATOM	H27	Hpol	X	0.40
ATOM	H6	Haro	X	0.06
ATOM	H7	Haro	X	0.06
ATOM	O6	OH	X	-0.56
ATOM	H13	Hapo	X	0.00
ATOM	O8	OH	X	-0.68
ATOM	H30	Hpol	X	0.40
ATOM	H11	Hapo	X	0.00
ATOM	O7	OH	X	-0.68
ATOM	H29	Hpol	X	0.40
ATOM	H10	Hapo	X	0.00
ATOM	H12	Hapo	X	0.00
ATOM	H20	Hapo	X	0.00
ATOM	H21	Hapo	X	0.00
ATOM	H28	Hpol	X	0.42
ATOM	O2	ONH2	X	-0.57
BOND_TYPE	C1	C2	4	
BOND_TYPE	C1	C4	4	
BOND_TYPE	C2	C5	4	
BOND_TYPE	C3	C5	4	
BOND_TYPE	C3	C6	4	
BOND_TYPE	C4	C6	4	
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BOND_TYPE	C5	N8	1	
BOND_TYPE	C6	O5	1	
BOND_TYPE	C7	C11	2	
BOND_TYPE	C8	C16	1	
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BOND_TYPE	C9	N1	2	
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BOND_TYPE	C16	N4	1	
BOND_TYPE	C17	N1	1	
BOND_TYPE	C17	N6	1	
BOND_TYPE	C18	C19	2	
BOND_TYPE	C20	N7	4	
BOND_TYPE	C20	O2	2	

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BOND_TYPE C21 C22 1
BOND_TYPE C21 C23 1
BOND_TYPE C21 O7 1
BOND_TYPE C22 C24 1
BOND_TYPE C22 O8 1
BOND_TYPE C23 C27 1
BOND_TYPE C23 O6 1
BOND_TYPE C24 N5 1
BOND_TYPE C24 O6 1
BOND_TYPE C25 C28 1
BOND_TYPE C26 C29 1
BOND_TYPE C27 O9 1
BOND_TYPE C28 N8 1
BOND_TYPE C29 N8 1
BOND_TYPE N7 S2 1
BOND_TYPE O3 S2 2
BOND_TYPE O4 S2 2
BOND_TYPE O9 S2 1
BOND_TYPE C1 H1 1
BOND_TYPE C2 H2 1
BOND_TYPE C3 H3 1
BOND_TYPE C7 H4 1
BOND_TYPE C8 H5 1
BOND_TYPE C9 H6 1
BOND_TYPE C10 H7 1
BOND_TYPE C18 H8 1
BOND_TYPE C19 H9 1
BOND_TYPE C21 H10 1
BOND_TYPE C22 H11 1
BOND_TYPE C23 H12 1
BOND_TYPE C24 H13 1
BOND_TYPE C25 H14 1
BOND_TYPE C25 H15 1
BOND_TYPE C25 H16 1
BOND_TYPE C26 H17 1
BOND_TYPE C26 H18 1
BOND_TYPE C26 H19 1
BOND_TYPE C27 H20 1
BOND_TYPE C27 H21 1
BOND_TYPE C28 H22 1
BOND_TYPE C28 H23 1
BOND_TYPE C29 H24 1
BOND_TYPE C29 H25 1
BOND_TYPE N6 H26 1
BOND_TYPE N6 H27 1
BOND_TYPE N7 H28 1
BOND_TYPE O7 H29 1
BOND_TYPE O8 H30 1
CHI 1 C23 C21 O7 H29
#PROTON_CHI 1 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 2 C21 C22 O8 H30
#PROTON_CHI 1 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 3 C2 C5 N8 C28
#PROTON_CHI 3 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 4 C19 C18 C11 C7

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#PROTON_CHI 4 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 5 S1 C15 C19 C18
#PROTON_CHI 5 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 6 N7 C20 C16 C8
CHI 7 O9 C27 C23 C21
CHI 8 C22 C24 N5 C10
CHI 9 S2 O9 C27 C23
CHI 10 C5 N8 C28 C25
#PROTON_CHI 10 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 11 C5 N8 C29 C26
#PROTON_CHI 11 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 12 C20 N7 S2 O3
CHI 13 N7 S2 O9 C27
NBR_ATOM C20
NBR_RADIUS 18.224748
ICOOR_INTERNAL C20 0.000000 0.000000 0.000000 C20 C16
C8
ICOOR_INTERNAL C16 0.000000 180.000000 1.496021 C20 C16
C8
ICOOR_INTERNAL C8 0.000000 59.872347 1.372318 C16 C20
C8
ICOOR_INTERNAL S1 -179.270493 77.666388 1.751132 C8 C16
C20
ICOOR_INTERNAL C15 0.492363 84.355877 1.747855 S1 C8
C16
ICOOR_INTERNAL C19 -178.965358 51.788039 1.472562 C15 S1
C8
ICOOR_INTERNAL C18 -1.305325 59.030868 1.347370 C19 C15
S1
ICOOR_INTERNAL C11 -179.274860 53.166278 1.497084 C18 C19
C15
ICOOR_INTERNAL C7 179.451266 63.097829 1.404762 C11 C18
C19
ICOOR_INTERNAL C4 -179.793688 59.532570 1.395584 C7 C11
C18
ICOOR_INTERNAL C1 179.796054 58.392558 1.400612 C4 C7
C11
ICOOR_INTERNAL C2 -179.921545 60.040291 1.397529 C1 C4
C7
ICOOR_INTERNAL C5 -0.116957 59.289576 1.406615 C2 C1
C4
ICOOR_INTERNAL C3 0.075829 61.084161 1.406503 C5 C2
C1
ICOOR_INTERNAL C6 0.052323 59.387740 1.398536 C3 C5
C2
ICOOR_INTERNAL O5 179.736043 60.042539 1.327524 C6 C3
C5
ICOOR_INTERNAL C14 -179.964796 56.803196 1.330734 O5 C6
C3
ICOOR_INTERNAL O1 -179.927603 61.884600 1.215817 C14 O5
C6
ICOOR_INTERNAL H3 178.842523 59.779401 1.082402 C3 C5
C6
ICOOR_INTERNAL N8 -179.993686 59.454771 1.434979 C5 C2
C3

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ICOOR_INTERNAL C2	C28	-0.002457	57.690884	1.466778	N8	C5
ICOOR_INTERNAL C5	C25	-91.854520	64.095211	1.534414	C28	N8
ICOOR_INTERNAL N8	H14	60.685126	69.691289	1.110806	C25	C28
ICOOR_INTERNAL H14	H15	-118.802735	71.019176	1.110603	C25	C28
ICOOR_INTERNAL H15	H16	-118.056018	67.538937	1.105051	C25	C28
ICOOR_INTERNAL C25	H22	-118.146161	72.185330	1.112371	C28	N8
ICOOR_INTERNAL H22	H23	-117.451347	70.323011	1.111477	C28	N8
ICOOR_INTERNAL C28	C29	-179.995647	57.678678	1.466716	N8	C5
ICOOR_INTERNAL C5	C26	-91.850411	64.105071	1.534379	C29	N8
ICOOR_INTERNAL N8	H17	60.685925	67.549597	1.105211	C26	C29
ICOOR_INTERNAL H17	H18	-123.147627	69.685529	1.110768	C26	C29
ICOOR_INTERNAL H18	H19	-118.811804	71.020849	1.110601	C26	C29
ICOOR_INTERNAL C26	H24	-118.150108	72.182516	1.112383	C29	N8
ICOOR_INTERNAL H24	H25	-117.454404	70.317350	1.111475	C29	N8
ICOOR_INTERNAL C5	H2	-178.850362	61.102830	1.081904	C2	C1
ICOOR_INTERNAL C2	H1	-179.822022	59.308872	1.083577	C1	C4
ICOOR_INTERNAL C4	H4	179.822139	59.822077	1.084739	C7	C11
ICOOR_INTERNAL C11	H8	179.434073	62.837482	1.087641	C18	C19
ICOOR_INTERNAL C18	H9	-179.278443	63.724142	1.078562	C19	C15
ICOOR_INTERNAL C19	N4	178.910262	74.692820	1.329868	C15	S1
ICOOR_INTERNAL S1	H5	179.269279	46.636101	1.077222	C8	C16
ICOOR_INTERNAL C8	N7	-9.923225	63.813358	1.468378	C20	C16
ICOOR_INTERNAL C16	S2	178.603099	58.300039	1.649518	N7	C20
ICOOR_INTERNAL C20	O3	-83.829480	72.117504	1.437768	S2	N7
ICOOR_INTERNAL O3	O4	-110.978218	68.095528	1.445070	S2	N7
ICOOR_INTERNAL O4	O9	-129.803411	67.884839	1.508885	S2	N7
ICOOR_INTERNAL N7	C27	47.977324	59.848471	1.411546	O9	S2

ICOOR_INTERNAL S2	C23	157.655249	70.737770	1.511083	C27	O9
ICOOR_INTERNAL O9	C21	-162.093482	70.933727	1.469270	C23	C27
ICOOR_INTERNAL C27	C22	-144.247783	77.196739	1.465552	C21	C23
ICOOR_INTERNAL C23	C24	31.367605	77.529154	1.475957	C22	C21
ICOOR_INTERNAL C21	N5	91.983558	65.428300	1.445328	C24	C22
ICOOR_INTERNAL C22	C10	-93.613244	52.728471	1.352163	N5	C24
ICOOR_INTERNAL C24	N3	-179.130300	69.967367	1.324677	C10	N5
ICOOR_INTERNAL N5	C12	0.433704	70.197497	1.328561	N3	C10
ICOOR_INTERNAL C10	C13	-0.110640	73.547282	1.420761	C12	N3
ICOOR_INTERNAL N3	N2	179.726171	60.632752	1.342181	C13	C12
ICOOR_INTERNAL C12	C9	0.608583	59.621629	1.326036	N2	C13
ICOOR_INTERNAL C13	N1	-0.848335	57.613744	1.331577	C9	N2
ICOOR_INTERNAL N2	C17	0.553656	58.116069	1.342041	N1	C9
ICOOR_INTERNAL C9	N6	-179.889752	60.503702	1.449271	C17	N1
ICOOR_INTERNAL N1	H26	-152.990791	73.020370	1.019094	N6	C17
ICOOR_INTERNAL H26	H27	125.651839	73.003358	1.011416	N6	C17
ICOOR_INTERNAL N1	H6	-179.346779	61.198218	1.081267	C9	N2
ICOOR_INTERNAL N3	H7	176.292279	47.119880	1.018607	C10	N5
ICOOR_INTERNAL N5	O6	-122.227965	71.734996	1.413847	C24	C22
ICOOR_INTERNAL O6	H13	-118.377782	70.274077	1.114966	C24	C22
ICOOR_INTERNAL C24	O8	-116.514168	66.311108	1.374557	C22	C21
ICOOR_INTERNAL C21	H30	-179.998770	69.489626	0.994258	O8	C22
ICOOR_INTERNAL O8	H11	-120.121966	66.508026	1.112708	C22	C21
ICOOR_INTERNAL C22	O7	-122.099284	69.960468	1.380236	C21	C23
ICOOR_INTERNAL C23	H29	-64.969184	73.081754	0.992629	O7	C21
ICOOR_INTERNAL O7	H10	-126.339895	68.634703	1.120794	C21	C23
ICOOR_INTERNAL C21	H12	120.013706	68.521390	1.115529	C23	C27

ICOOR_INTERNAL C23	H20	-117.558614	72.406306	1.115829	C27	O9
ICOOR_INTERNAL H20	H21	-120.541374	68.798568	1.089626	C27	O9
ICOOR_INTERNAL S2	H28	179.382356	56.641763	1.023812	N7	C20
ICOOR_INTERNAL N7	O2	-179.988605	59.568862	1.227216	C20	C16

The NapLuc-2-NMe<sub>2</sub> params file:

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NAME NLA
IO_STRING NLA Z
TYPE LIGAND
AA UNK
ATOM C21 CNH2 X 0.89
ATOM C12 aroC X 0.14
ATOM C7 aroC X -0.11
ATOM S1 S X -0.08
ATOM C13 aroC X 0.33
ATOM C20 aroC X -0.10
ATOM C19 aroC X -0.18
ATOM C10 aroC X 0.03
ATOM C3 aroC X -0.15
ATOM C1 aroC X -0.15
ATOM C9 aroC X 0.00
ATOM C6 aroC X -0.15
ATOM C11 aroC X 0.10
ATOM C4 aroC X -0.15
ATOM C2 aroC X -0.15
ATOM C8 aroC X 0.00
ATOM C5 aroC X -0.15
ATOM H5 Haro X 0.15
ATOM H2 Haro X 0.15
ATOM H4 Haro X 0.15
ATOM N8 Nhis X -0.84
ATOM C26 CH3 X 0.37
ATOM H16 Hapo X 0.00
ATOM H17 Hapo X 0.00
ATOM H18 Hapo X 0.00
ATOM C27 CH3 X 0.37
ATOM H19 Hapo X 0.00
ATOM H20 Hapo X 0.00
ATOM H21 Hapo X 0.00
ATOM H6 Haro X 0.15
ATOM H1 Haro X 0.15
ATOM H3 Haro X 0.15
ATOM H10 Haro X 0.15
ATOM H11 Haro X 0.15
ATOM N1 Nhis X -0.57
ATOM H7 Haro X 0.15
ATOM N7 Ntrp X -0.79
ATOM S2 S X 1.62
ATOM O2 OOC X -0.65
ATOM O3 OOC X -0.65

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ATOM	O7	OH	X	-0.46
ATOM	C28	CH2	X	0.28
ATOM	C24	CH1	X	0.28
ATOM	C22	CH1	X	0.28
ATOM	C23	CH1	X	0.28
ATOM	C25	CH1	X	0.63
ATOM	N5	Npro	X	-0.08
ATOM	C15	aroC	X	0.55
ATOM	N4	Nhis	X	-0.66
ATOM	C16	aroC	X	0.46
ATOM	C17	aroC	X	0.78
ATOM	N3	Nhis	X	-0.66
ATOM	C14	aroC	X	0.60
ATOM	N2	Nhis	X	-0.65
ATOM	C18	CH1	X	0.51
ATOM	N6	NH2O	X	-0.12
ATOM	H24	Hpol	X	0.40
ATOM	H25	Hpol	X	0.40
ATOM	H8	Haro	X	0.06
ATOM	H9	Haro	X	0.06
ATOM	O4	OH	X	-0.56
ATOM	H15	Hapo	X	0.00
ATOM	O6	OH	X	-0.68
ATOM	H28	Hpol	X	0.40
ATOM	H13	Hapo	X	0.00
ATOM	O5	OH	X	-0.68
ATOM	H27	Hpol	X	0.40
ATOM	H12	Hapo	X	0.00
ATOM	H14	Hapo	X	0.00
ATOM	H22	Hapo	X	0.00
ATOM	H23	Hapo	X	0.00
ATOM	H26	Hpol	X	0.42
ATOM	O1	ONH2	X	-0.57
BOND_TYPE	C1	C3	4	
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BOND_TYPE	C3	C10	4	
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BOND_TYPE	C5	C8	4	
BOND_TYPE	C5	C10	4	
BOND_TYPE	C6	C9	4	
BOND_TYPE	C6	C11	4	
BOND_TYPE	C7	C12	2	
BOND_TYPE	C7	S1	1	
BOND_TYPE	C8	C9	4	
BOND_TYPE	C10	C19	1	
BOND_TYPE	C11	N8	1	
BOND_TYPE	C12	C21	1	
BOND_TYPE	C12	N1	1	
BOND_TYPE	C13	C20	1	
BOND_TYPE	C13	N1	2	
BOND_TYPE	C13	S1	1	
BOND_TYPE	C14	N2	2	
BOND_TYPE	C14	N3	1	

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BOND_TYPE	C19	C20	2
BOND_TYPE	C21	N7	4
BOND_TYPE	C21	O1	2
BOND_TYPE	C22	C23	1
BOND_TYPE	C22	C24	1
BOND_TYPE	C22	O5	1
BOND_TYPE	C23	C25	1
BOND_TYPE	C23	O6	1
BOND_TYPE	C24	C28	1
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BOND_TYPE	C28	O7	1
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BOND_TYPE	C28	H22	1
BOND_TYPE	C28	H23	1
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BOND_TYPE	N6	H25	1
BOND_TYPE	N7	H26	1
BOND_TYPE	O5	H27	1



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BOND_TYPE  O6  H28  1
CHI 1  C24  C22  O5  H27
#PROTON_CHI 1 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 2  C22  C23  O6  H28
#PROTON_CHI 2 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 3  C20  C19  C10  C3
#PROTON_CHI 3 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 4  C6  C11  N8  C26
#PROTON_CHI 4 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 5  N7  C21  C12  C7
CHI 6  S1  C13  C20  C19
#PROTON_CHI 6 SAMPLES 3 60 -60 180 EXTRA 1 20
CHI 7  O7  C28  C24  C22
CHI 8  C23  C25  N5  C15
CHI 9  S2  O7  C28  C24
CHI 10  C21  N7  S2  O2
CHI 11  N7  S2  O7  C28
NBR_ATOM  C21
NBR_RADIUS 16.998837
ICOOR_INTERNAL  C21  0.000000  0.000000  0.000000  C21  C12
C7
ICOOR_INTERNAL  C12  0.000000  180.000000  1.496021  C21  C12
C7
ICOOR_INTERNAL  C7  0.000000  59.872347  1.372318  C12  C21
C7
ICOOR_INTERNAL  S1  -179.270493  77.666388  1.751132  C7  C12
C21
ICOOR_INTERNAL  C13  0.492363  84.355877  1.747855  S1  C7
C12
ICOOR_INTERNAL  C20  -173.771619  61.683298  1.726380  C13  S1
C7
ICOOR_INTERNAL  C19  171.268212  57.214464  1.342253  C20  C13
S1
ICOOR_INTERNAL  C10  179.597277  55.226679  1.489884  C19  C20
C13
ICOOR_INTERNAL  C3  -160.002537  61.928409  1.404869  C10  C19
C20
ICOOR_INTERNAL  C1  -179.968581  59.087473  1.396545  C3  C10
C19
ICOOR_INTERNAL  C9  -0.006186  59.677313  1.401283  C1  C3
C10
ICOOR_INTERNAL  C6  179.987822  59.415405  1.403816  C9  C1
C3
ICOOR_INTERNAL  C11  -179.987152  58.846339  1.414986  C6  C9
C1
ICOOR_INTERNAL  C4  -0.006973  62.221683  1.413997  C11  C6
C9
ICOOR_INTERNAL  C2  0.010676  58.765899  1.396493  C4  C11
C6
ICOOR_INTERNAL  C8  -0.008063  59.529665  1.399099  C2  C4
C11
ICOOR_INTERNAL  C5  179.985804  59.390822  1.404579  C8  C2
C4
ICOOR_INTERNAL  H5  0.004226  61.554529  1.080184  C5  C8
C2

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ICOOR_INTERNAL C8	H2	-179.996492	61.005780	1.083803	C2	C4
ICOOR_INTERNAL C2	H4	179.988820	58.440586	1.080434	C4	C11
ICOOR_INTERNAL C4	N8	179.998366	58.840031	1.445016	C11	C6
ICOOR_INTERNAL C6	C26	-179.996136	59.372524	1.457096	N8	C11
ICOOR_INTERNAL C11	H16	179.999819	67.581418	1.110235	C26	N8
ICOOR_INTERNAL H16	H17	-119.548441	70.125007	1.110555	C26	N8
ICOOR_INTERNAL H17	H18	-120.914948	70.131366	1.110560	C26	N8
ICOOR_INTERNAL C26	C27	179.987400	59.307240	1.456978	N8	C11
ICOOR_INTERNAL C11	H19	179.999960	70.119296	1.110451	C27	N8
ICOOR_INTERNAL H19	H20	-119.532614	67.589575	1.110170	C27	N8
ICOOR_INTERNAL H20	H21	-119.522494	70.131372	1.110635	C27	N8
ICOOR_INTERNAL C11	H6	179.997205	61.890803	1.081476	C6	C9
ICOOR_INTERNAL C9	H1	-179.993733	60.976718	1.083611	C1	C3
ICOOR_INTERNAL C1	H3	179.982726	59.580568	1.083325	C3	C10
ICOOR_INTERNAL C10	H10	179.898474	62.179240	1.084956	C19	C20
ICOOR_INTERNAL C19	H11	179.205681	66.627727	1.084284	C20	C13
ICOOR_INTERNAL C20	N1	173.716522	74.692820	1.329868	C13	S1
ICOOR_INTERNAL S1	H7	179.269279	46.636101	1.077222	C7	C12
ICOOR_INTERNAL C7	N7	-9.923225	63.813358	1.468378	C21	C12
ICOOR_INTERNAL C12	S2	178.603099	58.300039	1.649518	N7	C21
ICOOR_INTERNAL C21	O2	-83.829480	72.117504	1.437768	S2	N7
ICOOR_INTERNAL O2	O3	-110.978218	68.095528	1.445070	S2	N7
ICOOR_INTERNAL O3	O7	-129.803411	67.884839	1.508885	S2	N7
ICOOR_INTERNAL N7	C28	47.977324	59.848471	1.411546	O7	S2
ICOOR_INTERNAL S2	C24	157.655249	70.737770	1.511083	C28	O7
ICOOR_INTERNAL O7	C22	-162.093482	70.933727	1.469270	C24	C28
ICOOR_INTERNAL C28	C23	-144.247783	77.196739	1.465552	C22	C24

ICOOR_INTERNAL C24	C25	31.367605	77.529154	1.475957	C23	C22
ICOOR_INTERNAL C22	N5	91.983558	65.428300	1.445328	C25	C23
ICOOR_INTERNAL C23	C15	-93.613244	52.728471	1.352163	N5	C25
ICOOR_INTERNAL C25	N4	-179.130300	69.967367	1.324677	C15	N5
ICOOR_INTERNAL N5	C16	0.433704	70.197497	1.328561	N4	C15
ICOOR_INTERNAL C15	C17	-0.110640	73.547282	1.420761	C16	N4
ICOOR_INTERNAL N4	N3	179.726171	60.632752	1.342181	C17	C16
ICOOR_INTERNAL C16	C14	0.608583	59.621629	1.326036	N3	C17
ICOOR_INTERNAL C17	N2	-0.848335	57.613744	1.331577	C14	N3
ICOOR_INTERNAL N3	C18	0.553656	58.116069	1.342041	N2	C14
ICOOR_INTERNAL C14	N6	-179.889752	60.503702	1.449271	C18	N2
ICOOR_INTERNAL N2	H24	-152.990791	73.020370	1.019094	N6	C18
ICOOR_INTERNAL H24	H25	125.651839	73.003358	1.011416	N6	C18
ICOOR_INTERNAL N2	H8	-179.346779	61.198218	1.081267	C14	N3
ICOOR_INTERNAL N4	H9	176.292279	47.119880	1.018607	C15	N5
ICOOR_INTERNAL N5	O4	-122.227965	71.734996	1.413847	C25	C23
ICOOR_INTERNAL O4	H15	-118.377782	70.274077	1.114966	C25	C23
ICOOR_INTERNAL C25	O6	-116.514168	66.311108	1.374557	C23	C22
ICOOR_INTERNAL C22	H28	-179.998770	69.489626	0.994258	O6	C23
ICOOR_INTERNAL O6	H13	-120.121966	66.508026	1.112708	C23	C22
ICOOR_INTERNAL C23	O5	-122.099284	69.960468	1.380236	C22	C24
ICOOR_INTERNAL C24	H27	-64.969184	73.081754	0.992629	O5	C22
ICOOR_INTERNAL O5	H12	-126.339895	68.634703	1.120794	C22	C24
ICOOR_INTERNAL C22	H14	120.013706	68.521390	1.115529	C24	C28
ICOOR_INTERNAL C24	H22	-117.558614	72.406306	1.115829	C28	O7
ICOOR_INTERNAL H22	H23	-120.541374	68.798568	1.089626	C28	O7
ICOOR_INTERNAL S2	H26	179.382356	56.641763	1.023812	N7	C21

ICOOR\_INTERNAL      O1   -179.988605    59.568862      1.227216      C21    C12  
N7