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llan Samish Editor

Computational Protein Design



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Contents

,	facetributors	ix
PAR	RT I COMPUTATIONAL PROTEIN DESIGN	
1	The Framework of Computational Protein Design	3
2	Achievements and Challenges in Computational Protein Design	21
3	Production of Computationally Designed Small Soluble- and Membrane-Proteins: Cloning, Expression, and Purification	95
4	Deterministic Search Methods for Computational Protein Design	107
5	Geometric Potentials for Computational Protein Sequence Design Jie Li and Patrice Koehl	125
6	Modeling Binding Affinity of Pathological Mutations for Computational Protein Design	139
7	Multistate Computational Protein Design with Backbone Ensembles James A. Davey and Roberto A. Chica	161
8	Integration of Molecular Dynamics Based Predictions into the Optimization of De Novo Protein Designs: Limitations and Benefits	181
9	Applications of Normal Mode Analysis Methods in Computational Protein Design	203
PAR	RT II SOFTWARE OF COMPUTATIONAL PROTEIN DESIGN APPLICATIONS	
10	Computational Protein Design Under a Given Backbone Structure with the ABACUS Statistical Energy Function Peng Xiong, Quan Chen, and Haiyan Liu	217
11	Computational Protein Design Through Grafting and Stabilization	227
12	An Evolution-Based Approach to De Novo Protein Design Jeffrey R. Brender, David Shultis, Naureen Aslam Khattak, and Yang Zhang	243

	_	
VIII	(`or	ntents
VIII	COL	пспо

13	Parallel Computational Protein Design	265
14	BindML/BindML+: Detecting Protein-Protein Interaction Interface Propensity from Amino Acid Substitution Patterns Qing Wei, David La, and Daisuke Kihara	279
15	OSPREY Predicts Resistance Mutations Using Positive and Negative Computational Protein Design	291
Par	T III COMPUTATIONAL PROTEIN DESIGN OF SPECIFIC TARGETS	
16	Evolution-Inspired Computational Design of Symmetric Proteins	309
17	A Protocol for the Design of Protein and Peptide Nanostructure Self-Assemblies Exploiting Synthetic Amino Acids	323
18	Probing Oligomerized Conformations of Defensin in the Membrane	353
19	Computational Design of Ligand Binding Proteins	363
20	EpiSweep: Computationally Driven Reengineering of Therapeutic Proteins to Reduce Immunogenicity While Maintaining Function	375
21	Computational Tools for Aiding Rational Antibody Design	399
22	Computational Design of Membrane Curvature-Sensing Peptides	417
23	Computational Tools for Allosteric Drug Discovery: Site Identification and Focus Library Design	439
Inde	200	447

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Chapter 11

Computational Protein Design Through Grafting and Stabilization

Cheng Zhu, David D. Mowrey, and Nikolay V. Dokholyan

Abstract

Computational grafting of target residues onto existing protein scaffolds is a powerful method for the design of proteins with novel function. In the grafting method side chain mutations are introduced into a preexisting protein scaffold to recreate a target functional motif. The success of this approach relies on two primary criteria: (1) the availability of compatible structural scaffolds, and (2) the introduction of mutations that do not affect the protein structure or stability. To identify compatible structural motifs we use the Erebus webserver, to search the protein data bank (PDB) for user-defined structural scaffolds. To identify potential design mutations we use the Eris webserver, which accurately predicts changes in protein stability resulting from mutations. Mutations that increase the protein stability are more likely to maintain the protein structure and therefore produce the desired function. Together these tools provide effective methods for identifying existing templates and guiding further design experiments. The software tools for scaffold searching and design are available at http://dokhlab.org.

Key words Scaffold search, Refinement, Stabilization, Mutation, Free energy, Protein design

1 Introduction

The goal of the protein design field is to engineer proteins with novel function with implications for developing new enzymes, biosensors, and therapeutics [1–4]. One approach to protein design is grafting, which has been successful in the developing of novel inhibitors [2, 5], biomarkers [6], enzymes [7], and antigens [8]. In the grafting approach one identifies an existing structural scaffold that can host a specific motif and replaces residues to match a desired active site or binding motif. This approach relies on the availability of potential scaffolds into which design mutations can be introduced. To this end, several methods have been developed to identify these scaffolds including Multigraft Match [9], GRAFTER [10], and MaDCaT [11]. These approaches rely on matching of backbone atoms, $C\alpha$ - $C\beta$ vectors, or $C\alpha$ distance maps to identify potential scaffolds. In our approach we use the

Erebus webserver, which allows greater user flexibility in matching user-specified atom types and residues, while maintaining high speed and accuracy [12].

Having obtained a viable scaffold onto which target residues can be introduced, a further challenge is to determine a priori whether desired mutations will distort the original protein scaffold or even completely destabilize the protein structure. To improve thermodynamic stabilities of designed proteins, we use the protein stability prediction software Eris [13]. Eris has been shown to effectively predict effects of mutations, even for the more challenging case of small to large mutations, without the need for computationally intensive molecular dynamics simulations.

The overall workflow for our protein design protocol is composed of three major steps. In the first step we identify protein scaffolds for redesign using the Erebus. In the second step we submit the structure to the Chiron webserver for pre-relaxation prior to introducing mutations. In the final step we use the Eris webserver to identify potential redesign sequences. The theory and protocol for each of these methods are outlined in the following sections.

1.1 Identify Protein Scaffolds for Redesign

The identification of promising candidates for redesign makes use of the Erebus substructure search [12]. Provided with a query structure Erebus scans the protein data bank (PDB) for matching structural scaffolds [14]. Atom pairs of target structures in the structural database matching the name, residues, and distances in the query structure are collected to create the candidate substructures. The best substructures are selected based on their final weights representing the agreement between query and target structures. The final weights are calculated according to the following equation:

$$W = \left(\prod_{i=1}^{N} W_i\right)^{1/N} \prod_{j=1}^{M} \left(1 - w_j\right)$$

where the final weight (W) is the geometric mean of the weights (W_i) for each of the N atom pairs. The product is multiplied by an additional penalty (w_j) for each of the M missing atoms. Individual weights (W_i) for each atom pair are calculated according to the formula:

$$W_i = e^{-\frac{\left(\Delta q_i - \Delta t_j\right)^2}{\sigma^2}}$$

Where the term $(\Delta q_i - \Delta t_j)$ represents the difference in distances between atom pairs i and j in the query (q) and target (t) structures, and σ^2 is a user-defined precision parameter.

1.2 Protein Preparation

As steric clashes are common structural artifacts observed in homology models and low-resolution crystal structures [15, 16], we first relax structures using Chiron before proceeding to Eris. In this method, a clash is defined as any atomic overlap resulting in energy greater than 0.3 kcal/mol (0.5 k_BT). Structural relaxation in Chiron is achieved by performing a series of short (~10 ps) discrete molecular dynamics simulations [17, 18], using a high heat exchange rate (5 fs) between the protein and the thermal bath. The exchange rate is used to effectively quench large atomic velocities resulting from large van der Waals clashes, which could result in broken bonds. To prevent large structural distortions, we also constrain backbone and C_β atoms with a harmonic potential. The algorithm alternates between high temperature (0.7 $\varepsilon/k_{\rm B}$, ~350 K) and low temperature (0.5 $\varepsilon/k_{\rm B}T$, ~250 K) simulations until the overall clash score is less than 0.02 kcal/mol/contact. Pre-relaxing the structures in this way significantly improves sidechain packing of the protein core, which improves the accuracy of $\Delta\Delta G$ evaluations.

1.3 Protein Redesign

Protein redesign of preexisting scaffolds is accomplished using Eris [13]. Eris introduces a mutation or set of mutations into a protein structure and calculates free energies for both mutant (ΔG_{MUT}) and native (ΔG_{NAT}) structures. For free energy calculations rapid side-chain repacking and backbone relaxation are performed around the mutation site(s) using a Monte Carlo algorithm. The free energies are the result of a weighted sum of van der Waals forces, solvation, hydrogen bonding, and backbone-dependent statistical energies derived from the Medusa force field [19]. The final prediction of protein stability induced by mutations is expressed as the $\Delta\Delta G$ ($\Delta G_{\text{MUT}} - \Delta G_{\text{NAT}}$). Weighting parameters for free energy calculations were independently trained on 34 highresolution X-ray protein structures and tested on a large dataset of 595 mutants where we found significant agreement between predicted and measured $\Delta\Delta G$ values ($R^2 = 0.75$) [13]. Furthermore, Eris can model the backbone flexibility, which is crucial for $\Delta\Delta G$ estimation of small-to-large mutations [20].

2 Software Requirements

The webservers of Erebus, Chiron, and Eris (ddg module) are freely available on our group page (http://dokhlab.org). The current version of Eris (ddg, scan and design module) also supports Linux/Unix-like platforms with the C and C++ compilers installed. It has been tested on Linux, Microsoft Windows (with the Linux port Cygwin), and Mac OS X. Our methods usually require a molecular viewer for preparation of crystal structures and analysis of results. For these purposes PyMol (http://pymol.org), an Open Source molecular viewer available on Windows, Mac OS X, and Linux, is used [21].

3 Methods

3.1 Erebus: Structure Search and Grafting

Erebus is a protein substructure search server (http://Erebus.dokhlab.org). It searches the entire PDB database for a match to a user-defined substructure, which can be any atoms from the backbone (N, Cα, O) or functional sites (see the following example). Erebus reads ATOM and HETATM records for atom coordinates in the PDB format and will only match atoms in a target substructure with atom names exactly matching those in the query structure [3, 4]. This feature is useful for identifying proteins that have the same catalytic sites, bind to similar small molecules or have the same backbone structures.

As an example, we used the copper-binding site in Cu/Zn superoxide dismutase (SOD1) to prepare the query structure and to find similar metal binding sites in PDB.

1. Create a file containing the atoms for the substructure query in PDB format. In our example these atoms are the Nδ or Nε atoms of H46, H48, H63, and H120 forming the copper-binding site in SOD1 (an example of the file is below). We saved this file as Cu_His.pdb:

ATOM 2 ND1 HIS A 46 11.519 -11.568 8.749 1.00 9.92 N ATOM 4 NE2 HIS A 48 13.185 -15.110 8.862 1.00 11.95 N ATOM 6 NE2 HIS A 63 10.975 -13.745 10.609 1.00 8.89 N ATOM 8 NE2 HIS A 120 12.452 -13.024 6.807 1.00 2.00 N END

- 2. Upload the query PDB file to the to Erebus server (Fig. 1).
- 3. After uploading the query PDB structure the user will have the option to adjust parameters for each atom in the search. Adjustable parameters include:
 - 'Residue Name': Under the 'Residue Name' column the user can specify the particular residue to match via a dropdown menu or may specify ANY to return matches from any residue.
 - 'Matching precision' or σ : This is a user-defined precision parameter (*see* Subheading 1.1). Smaller values for σ result in smaller deviations between the query and the targets.
 - 'Minimum weight' or W: This parameter measures how well the query and target structure match (*see* Subheading 1.1). Values for W range from 0 to 1, where smaller W means a worse match (the RMSD is bigger) and W = 1 is an identical match. The user can define the minimum acceptable W.
- 4. For this example Erebus finds over 100 matching structures in the PDB. These structures are ranked based on their RMSD to the query structure (Fig. 2). For each match the user can

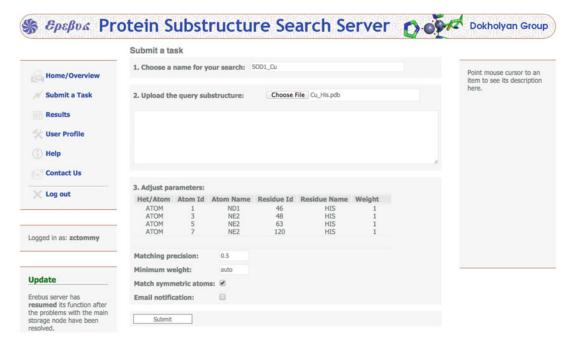


Fig. 1 The Erebus Web interface. Users may either upload a query PDB file or paste query coordinates into the field below in PDB format. Clicking a residue name under the 'Residue Name' column brings up a menu allowing the user to specify the particular residue to match. The 'ANY' selection allows for matches to any residue

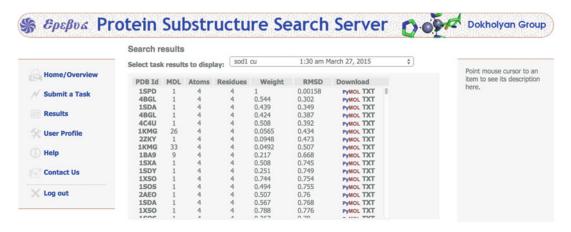


Fig. 2 Results from the Erebus scaffold search. The results are sorted by weight and root mean square deviation (RMSD) to the query structure

download a summary text file (TXT) and a structural model (PyMOL). The structural model file (.pml) is used to visualize the match between query and target structures (Fig. 3).

Erebus identified several protein families containing copperbinding site, including superoxide dismutase, laccase, and multicopper oxidase. Several iron and zinc binding sites were also found.

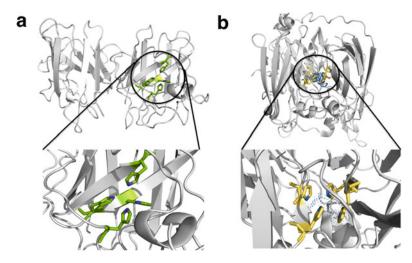


Fig. 3 Comparison of the copper binding site in the query structure (a) and one target structure (b). The query is based on the crystal structure of SOD1 (1SPD) and the target is a laccase (2XUW). The copper-interacting atoms (N δ or N ϵ of histidine) are shown in *blue color*

3.2 Chiron: Preparation of Input Structures: Minimizes Steric Clashes Before Redesign

Chiron estimates the quality of a structure with respect to clashes and minimizes clashes using a series of short discrete molecular dynamics simulations [17, 18]. For homology models and low-resolution crystal structures in which structural artifacts often exist, Chiron is a useful refinement tool.

The Chiron webserver is freely available at http://chiron.dokhlab.org. After logging into the server, the user will be directed to the task submission page by clicking 'Submit Task' in the left panel. The user provides a structural model in PDB format or a PDB ID. During relaxation all backbone and Cβ atoms are constrained. To constrain all side-chain atoms not involved in steric clashes check the 'Constrain Sidechain' box (Fig. 4).

Results of the calculation can be accessed from the 'Home/ Overview' page. Chiron outputs a refined structural model ({JOBID}.pdb) and the record of clashes ({JOBID}.py).

As an example, we submitted the crystal structure of Cu/Zn superoxide dismutase (SOD1, PDB ID: 1SPD) to the Chiron server. The resulting files are 12489.pdb and 12489.py. The following steps can be applied to visualize the clashes before and after the refinement:

- 1. Open 12489.pdb with PyMOL.
- 2. In the PyMOL command prompt, type 'run {Path}/ {JOBID}. py' and enter, omitting the quotation marks. The path to the . py should be indicated. In the provided example the command is 'run ~/user/12489.py'

The script generates two structural models: i-12489 is the structure of SOD1 before refinement and f-12489 is the structure

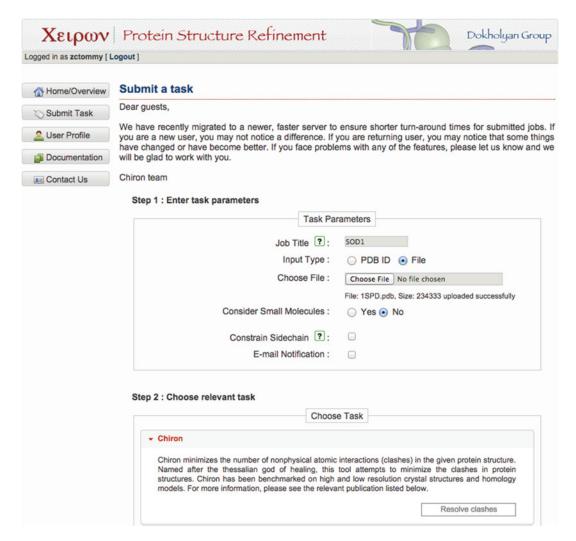


Fig. 4 Image of the Chiron Web interface

after refinement. The steric clashes are represented as color-coded (rainbow spectrum) cylinders of different radii. The clashes with higher repulsion energy are denoted as cylinders of larger radii. For SOD1, Chiron refinement successfully reduced the number of clashes and eliminated all major clashes (Fig. 5).

3.3 Eris: Identify Mutations That Stabilize a Protein Scaffold Eris has three modules: ddg, scan, and design. The 'ddg' module exhaustively calculates the $\Delta\Delta G$ for individual mutations. The 'scan' module is used to rapidly search for stabilizing single mutations at a specified site. The 'design' function identifies the lowest-energy sequence for a given backbone, which can be a complete protein structure or a user-defined region in the whole structure. In the following section we elaborate the methods and procedures for each module.

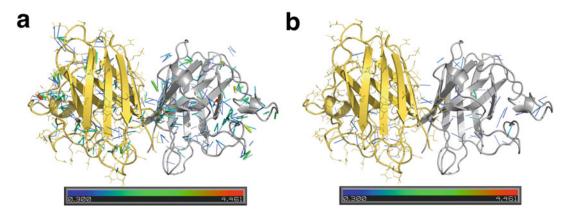


Fig. 5 Resolution of clashes using the Chiron webserver. Structures of SOD1 (PDB ID: 1SPD) are shown (a) before and (b) after refinement by Chiron. Colored cylinders connect atoms involved in clashes. The color and thickness of the cylinder denote the energy associated with the clash. Comparing panel **a** to panel **b** demonstrates that Chiron greatly reduces the number of clashes, and eliminates the large steric clashes

Table 1 Eris usage flags

Flag	Usage
-W	Specify the working directory
-j	Specify the 'JOBID'
-f	Flexible backbone (Not valid in 'scan' mode)
-r	Provide random seeds in Monte Carlo simulation
-m	Specify the mutation(s) for $\Delta\Delta G$ evaluations in 'ddg' mode
-s	Specify the site(s) in 'scan' mode
-d	Specify the path to a 'DesignTable' in 'design' mode
-h	Print the help, then exit
-V	Verbose output

3.3.1 Installation and Preparation for Input Files

The Eris package is available at http://moleculesinaction.com. Installation instructions are provided with the package (*see* **Note 1**).

The input file for all three modules should be in the PDB format. Currently, the Eris server will only read the first chain of a PDB file (*see* **Note 2**). Eris also renumbers the index of the first residue as 1. The common flags in Eris for command line usage are listed in Table 1.

3.3.2 ddg: An Estimation of $\Delta \Delta G$ for Given Mutations

For a single mutation or multiple substitutions, Eris-ddg repacks the sidechain 20 times using simulated annealing and computes stabilities by averaging all the conformations for both wild type and mutant. The stability change, $\Delta\Delta G$, is computed as the difference between the average stabilities of mutant and native structure $(\Delta G_{\text{MUT}} - \Delta G_{\text{NAT}})$.

As an example, we calculate the stability change induced by Y36E mutation in protein kinase B (PDB ID: 1UNP).

1. Submit the job:

 $\ eris\ ddg\ -m\ \{MUTATION\}\ -w\ \{DIRECTORY\}\ -j\ \{JOBID\}\ INPUT.pdb$

MUTATION is a comma-delimited string of all mutations to be performed in a particular eris run in the format {native residue}{residue number}{target residue}. For example to change an alanine dipeptide to a serine dipeptide the input would be A1S, A2S.

In our case we used the following command line and the results were kept in a folder called 'Eris_ddg':

\$ ~/Eris/eris ddg -m Y36E -w ~/Eris_ddg -j 1UNP_Y36E ~/ Eris_ddg /1UNP.pdb

2. Analyzing the results:

A summary of the results can be found in ~/Eris_ddg/eris_ddg.txt, which looks like:

1UNP_Y36E ddg ~/Eris_ddg/1UNP.pdb Y36E 9.95 9.95 7.11 0.11 2.84 0 0.81 0.05 0.40 0.46 -4.62 3.37

The job (JOBID = 1UNP_Y36E) is the 'ddg' calculation for input structural model of 1UNP.pdb. The mutation (Y36E) is the substitution of Tyrosine at position 36 to Glutamic acid. The total $\Delta\Delta G$ equals to 9.95, i.e., destabilizing. In the second line the total stability change and its decomposition are illustrated (*see* Note 3).

Structure files of both native and mutant proteins for each of the 20 rounds of calculation are stored in PDB format in the folder ~/Eris_ddg/1UNP_Y36E.

3.3.3 Scan: Search for Stabilizing Mutations

In the scan module a native amino acid at a specified site is substituted to all other 19 types of amino acids and only the stabilizing substitution ($\Delta\Delta G < 0$) are kept. If positions are not explicitly specified using '-s', then all the residues are scanned.

As an example, we used Eris-scan to find stabilizing mutations at position 37 in protein kinase B (PDB ID: 1UNP).

1. Submit the job:

\$ eris scan -s {SITE} -w {DIRECTORY} -j {JOBID} INPUT. pdb

SITE is a comma-delimited string of integers specifying the residue positions to be scanned. Residue positions are determined from their order in the PDB file and Eris-scan

renumbers the index of the first residue as 0 ('-sN' means the scan is performed on the N+1 site).

In our case we used the following command line and the results were kept in a folder called 'Eris_scan':

\$ ~/Eris/eris scan -s 36 -w ~/Eris_scan -j 1UNP ~/Eris_scan /1UNP.pdb

2. Analyzing the results:

The stabilizing mutations were listed in 'Eris_scan/1UNP/output/ddgStabilizing.dat', which looks like:

```
K37L -3.04264 -0.0944399 -0.0642849 -1.14021 0 -0.132146...

K37V -2.11104 0.0127475 -0.064736 -1.33434 0 0.0623585...

K37Q -0.111982 -0.878894 0.05282570.354524 0 0.0364916...

K37N -0.817968 0.106138 -0.10984 -0.245153 0 0.0981676...
```

Each line in the ddgStabilizing.dat specifies the stabilizing single mutations (*see* **Note 4**). In the same line, the numbers starting from the second column correspond to the total stability change and its decomposition.

The atomic structures of repacked conformations are stored in the same folder. The calculation results for all 19 substitutions were stored in ddgAll.dat.

3.3.4 Design: Find the Optimal Amino Acid Sequence for the Given Protein Backbone

In the 'design' module, users specify the protein segments to optimize and which subset of amino acids can be used to substitute the original one (polar, hydrophobic or user-defined subsets). The search can be performed using either a fixed backbone protocol (C, O, CA, and N positions fixed during design), or a flexible backbone protocol (allowing small adjustments of the backbone atoms to minimize energy). Eris-design then searches the lowest-energy sequence that satisfies the constraints listed in the design table.

Before submitting the 'design' job, the user should prepare a design table in .txt format. The design table consists of two columns: the first column (Index) specifies the mutation sites and the second column (Keyword) defines the subset of amino acids for substitution.

Values in the index column can be a single integer (m), a set of integers separated by commas (m,n,..), a range defined as m-n (m < n), or a mixture such as (a-b,c,d,e-f). The "DEFAULT" keyword can be used to represent all residues that have not been explicitly specified.

The keyword column takes as an argument one of a list of predefined flags. The flags and definitions are listed in Table 2.

For clarity an example design table for protein kinase B (PDB ID: 1UNP) is shown below:

Table 2					
Keywords and	definitions	for	Eris	design	table

Flag	
ALLAA	All available amino acids and the corresponding rotamers
NATAA	Fixed with native amino acid, but with all its available rotamers
NAROT	Fix the amino acid in its native rotamer but with sub-rotamer allowed
FIXNR	Fix the amino acid in its native rotamer without sub-rotamer motion
POLAR ^a	Polar amino acids and their rotamers
HYDPH ^b	Hydrophobic amino acids
AROMA ^c	Aromatic amino acids
PIKAA	User selected amino acids represented by single letter

^aPOLAR includes: SER, THR, GLN, GLU, ASN, ASP, LYS, ARG, HIS

#Index Keyword DEFAULT NATAA 4-10 ALLAA 102, 112 PIKAA STYWL 50,69-76 HYDPH QEND

This design table will perform all-amino substitutions for residues 4 through 10, substitute residues 102 and 112 with Ser, Thr, Tyr, Trp, and Leu. And substitute residues 50 and 69 through 76 with hydrophobic amino acids. The QEND flag serves to denote the end of the design table

1. Submit the job:

\$ eris design –d {DesignTable} –w {DIRECTORY} –j {JOBID} STUCTURE.pdb

In our case we used the following command line and the results were kept in a folder called 'Eris_design' (*see* **Note 5**):

\$ ~/Eris/eris design -d ~/DesignTable.txt -w ~/Eris_design -j 1UNP ~/Eris_design/1UNP.pdb

2. Analyzing the results

The output of Eris-design is a PDB file of redesigned structural model and its free energy. In this module 20 rounds of Monte

bHYDPH includes: GLY, ALA, MET, VAL, LEU, ILE, PHE, TYR, TRP, PRO, CYS

^cAROMA includes: PHE, TYR, TRP, HIS

Carlo simulations are performed. The results are given as design.run [00-19] and kept in Eris_design/1UNP/design. In each ".run" file, the first 20 lines record the temperature, Monte Carlo acceptance rate, total energy and its decompositions. The following lines are in PDB format so that the user can open it with PyMOL.

For example, design.run00 looks like:

0 10 0.763302 222.253 -409.689 371.847 319.997... 1 6.35799 0.70095 120.507 -402.49 196.314 296.97...

. . .

19 0.101625 0.0279097 -184.007 -464.69 8.64136 316.316... ATOM 1 N ASP A 1 31.522 1.268 -6.333 1.00 0.00 N ATOM 2 CA ASP A 1 30.972 2.648 -6.220 1.00 0.00 C

. . .

ATOM 1245 2HH2 ARG A 119 22.086 -5.596 -26.587 1.00 0.00 H TER

3.3.5 An Online Server for 'ddg' Calculations

A Web-based Eris server for $\Delta\Delta G$ estimation is freely accessible online (http://eris.dokhlab.org). The users can follow the simple procedures listed below to submit their own task after registration:

- 1. Use the 'Submit a Task' bar on the left to submit the protein structure file. It can be a PDB ID or your own .pdb file. Eris only recognizes the first chain by reading the 'TER' line in a . pdb file.
- 2. Click on any residue site and choose the amino acids you want to substitute (Fig. 6).
- 3. Choose 'Fixed Backbone' or 'Flexible Backbone' and choose whether you want to include a pre-relaxation of backbone structure. Pre-relaxation remarkably improves the prediction accuracy when a high-resolution protein structure is not available. Alternatively the users can use Chiron to minimize steric clashes in the input structure (*see* Note 6).

4 Notes

- 1. After installation, typing "eris" without any command line arguments will display the brief help information. Typing "eris –h" will bring more detailed instructions.
- 2. The Eris webserver only reads the first chain of the provided PDB file. To modify the protein chains, the user can apply the 'alter chain' and 'alter resi' command in PyMOL for this modification (http://www.pymolwiki.org).



Fig. 6 Image of the Eris Web interface. Clicking on a residue results in a pop-up window allowing the user to select the target residue

- 3. The energy decomposition of total $\Delta\Delta G$ is composed of ten values. From left to right they are van der Waals attraction, van der Waals repulsion, solvation, backbone hydrogen bonds, backbone–side chain hydrogen bonds, side chain–side chain hydrogen bonds, backbone dependent statistical energy for amino acid, backbone-dependent statistical energy of the rotamer, reference energy of the unfolded states, and the correction for reference energy, respectively [19].
- 4. Eris-scan does not support flexible-backbone protocol and the Monte Carlo simulation is performed only once. Due to the limitations of the fixed-backbone method and sampling inefficiency, atomic clashes may not be resolved during structure minimization. These clashes can be identified by checking the van der Waals repulsion energy terms in the results. We find that the predictions are relatively more accurate for buried residues than exposed residues.
- 5. Eris uses a Monte Carlo algorithm for identifying changes in ΔΔG. As such there is a certain amount of stochasticity in the results. We suggest that the user run the calculation multiple times to check for convergence in the distribution of energies. Performing multiple rounds of Eris calculation can be performed using the following bash script.

~/eris ddg -m Y36E -w \$CUR_DIR -j 1UNP_Y36E_\${i} -r \$RANDOM \$CUR_DIR/1UNP.pdb

done

6. If there is difficulty with formatting or running the calculations it is suggested that the user submits the structural model to Chiron first. The server will both reformat the file to be compatible with Eris and fix clashes that could produce problems with the Eris calculations.

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