INTREPID: a web server for prediction of functionally important residues by evolutionary analysis

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Received March 3, 2009; Revised April 13, 2009; Accepted April 21, 2009

ABSTRACT

We present the INTREPID web server for predicting functionally important residues in proteins. INTREPID has been shown to boost the recall and precision of catalytic residue prediction over other sequence-based methods and can be used to identify other types of functional residues. The web server takes an input protein sequence, gathers homologs, constructs a multiple sequence alignment and phylogenetic tree and finally runs the INTREPID method to assign a score to each position. Residues predicted to be functionally important are displayed on homologous 3D structures (where available), highlighting spatial patterns of conservation at various significance thresholds. The INTREPID web server is available at http://phylogenomics.berkeley.edu/intrepid.

INTRODUCTION

Predicting the functional residues in a protein is an important and challenging problem in bioinformatics. Prediction methods provide a useful starting point to understand the functions of proteins and serve to prioritize site-directed mutagenesis experiments. A variety of approaches have been developed, including those that primarily or exclusively exploit sequence information [e.g. (1–6)] and those that make use of both sequence and structure information [e.g. (7–9)]. INTREPID (Information-theoretic Tree Traversal for Protein Functional Site Identification) falls into the first class of sequence-only methods, and uses phylogenetic analysis of a family of homologous sequences to identify positions that are conserved at different levels of an evolutionary tree. It has been shown to outperform other sequence-only methods at detecting catalytic sites in proteins (10).

Structural information clearly provides a significant boost in prediction accuracy, but is available for only a small fraction of proteins. For this reason, sequence-based methods play a key role in bioinformatics prediction of functional residues. Sequence-only methods for functional residue prediction are based on the assumption that mutations disrupting function will not be tolerated by evolution; i.e. we can exploit nature’s mutagenesis experiments to reveal positions playing critical roles. Such positions are identified using multiple sequence alignment (MSA) analysis: positions that display a high degree of conservation against a backdrop of divergence across related sequences can be posited as functionally (or perhaps structurally) important.

Notably, while sequence-based methods depend on the evolutionary context to predict functional residues, many methods include only moderately divergent homologs in an alignment. This restriction may be designed to limit the changes in function and structure that accumulate with evolutionary distance (11), to reduce the likelihood of alignment errors (12,13) or for reasons of computational efficiency, but effectively reduces the total possible information available to a method.

INTREPID is designed to make full use of the information in a protein family containing many distantly related sequences through the use of tree traversal: INTREPID computes an information-theoretic score for each position in the sequence at each subtree encountered on a path from the root to the leaf corresponding to the sequence of interest. The score for each position is set to the maximal score obtained on the path. Positions that are conserved across the entire family obtain their maximum score on the path. Positions that are conserved across the entire family obtain their maximum score at the root of the tree, while other positions will achieve maximum scores at nodes corresponding to one of the nested subtrees. This tree traversal gives INTREPID the ability to detect subtle evolutionary patterns that other methods might miss. For instance, positions that are critical for function for one subfamily but variable across the family as a whole may remain undetected by prediction methods that do not use tree
traversal to extract the evolutionary signal. In contrast, since INTREPID computes the importance of each position within all subtrees containing the sequence of interest, it suffices for a position to become conserved within some subtree for it to be detectable. In analysis of enzymes from the manually curated section of the Catalytic Site Atlas (14), INTREPID has been shown to achieve significantly higher levels of recall and precision at catalytic residue prediction than other sequence-based methods, including those that use evolutionary tree analysis, primarily due to the use of tree traversal to mine the information contained in highly divergent datasets (10).

THE INTREPID WEB SERVER
The INTREPID web server provides a pipeline for homolog selection and alignment, identification of homologous 3D structures and calculation of INTREPID scores. It is available at http://phylogenomics.berkeley.edu/intrepid/.

INPUT
The input to the web server is a protein sequence (seed) in FASTA format. The user can supply an email address to which results will be sent, or bookmark the page displayed after clicking ‘Submit’.

PROCESSING STEPS
Initial processing steps
The INTREPID functional residue prediction server is a multi-stage automated analysis pipeline. We start by gathering homologs from the UniProt protein sequence database using PSI-BLAST. Since INTREPID performance improves as sequence divergence in the dataset increases (10), we use four iterations of PSI-BLAST with an E-value inclusion cutoff of $1 \times 10^{-04}$. We retrieve the full-length proteins for all hits found by PSI-BLAST and align these sequences using MUSCLE (15). The MSA is edited to remove columns corresponding to a gap in the seed, followed by removing sequences that are composed entirely of gaps. A Neighbor-Joining phylogenetic tree is estimated using the QuickTree software (16) and rooted at the midpoint of the longest span. The tree and MSA are then given as input to the INTREPID algorithm. We run additional analyses such as identification of PFAM domains and prediction of transmembrane helices, retrieve GO annotations and biological literature, identify homologous 3D structures and gather other data from various external resources. These are not used by INTREPID, but are provided to the user to assist in interpreting INTREPID program output.

INTREPID scoring
Given an MSA, a corresponding phylogenetic tree and a specified seed sequence, INTREPID examines each subtree containing the seed (i.e. corresponding to the subtrees encountered on the path from the root to the seed sequence leaf). For each position in the seed and for each subtree independently, INTREPID computes the Jensen–Shannon divergence between the amino acid distribution at the position and the background distribution in the subtree. This produces a set of scores for each position in the sequence, one for each subtree encountered in the tree traversal. Upon reaching the leaf, the maximal score for each position is identified. The scores are then normalized and sorted to produce a rank-ordered list. Details on the algorithm are available in (10).

Identification of homologous 3D structures
We search for homologous 3D structures using an HMM constructed from the MSA using the UCSC SAM w0.5 software (17). Note that INTREPID does not make use of structural information; the PDB structures are simply provided to enhance user interpretation of results.

Time to complete
The bulk of the INTREPID computational complexity is due to phylogenetic tree construction (in contrast, computing the INTREPID scores is very fast once an MSA and tree are available, and typically takes $<5$ min). The time required from input to final results can range from under 15 min to many hours, depending primarily on the number of sequences retrieved.

OUTPUT
Two links are provided, either sent by email or on the web page bookmarked by the user. The first link gives the INTREPID score results (Figure 1). The second link is to a web page displaying the alignment, phylogenetic tree and additional data for the protein family and includes results from a second functional site prediction protocol (Figure 2).

INTREPID results
Results include a normalized score for each residue in the input protein sequence. Scores can be viewed on the output page or downloaded in CSV format. If a homologous PDB structure can be detected, these scores will be used to highlight top-ranked residues on the structure; if not, we report, ‘This sequence has no evident homology to any PDB structure.’ Homologous PDB structures are displayed using the Jmol Java-based structure viewer (18). The top 10% of residues are colored by default. A scale on the right of the Jmol display shows the range of INTREPID scores and allows the user to select an appropriate threshold. A link is also provided to allow users to view scores for all residues and to sort the results based on residue position, residue type and INTREPID score (Figure 1).

Protein family page
This page includes an MSA, phylogenetic tree, predicted subfamilies, hidden Markov models for the family and subfamilies, GO annotations, matching PFAM domains and homologous PDB structures,
These data are displayed on the web page using the same formatting as PhyloFacts family ‘books’ (20). The structure icon is linked to a page presenting results from a different method for predicting functional sites (using family and subfamily conservation patterns) and is provided as a complementary analysis to the INTREPID scores. Interactive viewers obtained using a slightly modified version of the PhyloBuilder pipeline (19). These data are displayed on the page using the same formatting as PhyloFacts family ‘books’ (20). The structure icon is linked to a
are provided for the phylogenetic tree, structure and MSA. We provide two phylogenetic tree viewers: the Java-based ATV software (21) and a new Javascript phylogenetic tree viewer, PhyloScope. The alignment, tree and HMMs can be downloaded from the web site (Figure 2).
identified using SCI-PHY (22). Subtrees can be collapsed or expanded individually by clicking on internal nodes (or collapsed terminal nodes). The availability of experimental data, literature and external resources with additional data is indicated by icons that are hyperlinked to these resources (Figure 2b).

INTREPID program dependencies

INTREPID depends on the availability of homologs to predict functional positions. If fewer than four sequences are retrieved using PSI-BLAST, the program terminates with the message ‘Functional Residue Prediction – terminated. Too few homologs were found to perform phylogenomic analysis.’

DISCUSSION

The INTREPID server predicts functional residues based on sequence information alone. The tree traversal enables INTREPID to identify sites that are important for the query sequence even if they are variable across the family as a whole. As a result, INTREPID is robust to functional divergence in specific lineages of the family and to alignment and phylogeny errors. The computational efficiency of the tree traversal allows INTREPID to scale to very large and divergent protein families, enabling it to exploit the information in these families to boost its predictive power. In fact, we have shown in (10) that the accuracy of INTREPID improves as we increase the evolutionary divergence of the input data.

The INTREPID web server takes as input a protein sequence, retrieves and aligns homologs, constructs a phylogenetic tree and searches for homologous 3D structures. The time required to complete these analyses can vary from a few minutes to several hours, dependent primarily on the number of homologs retrieved. Results are plotted on homologous 3D structures, when available, and can be downloaded or viewed online. Additional outputs of the INTREPID web server include a web page containing an MSA, phylogenetic tree, a spreadsheet of data, predicted subfamilies, homologous 3D structures, Gene Ontology annotations with evidence codes, biological literature and hidden Markov models for the protein family as a whole.

FUTURE PLANS

Our future plans for this server include expanding the functionality to utilize information from 3D structure, where available, and providing for prediction of specificity-determining residues (e.g. substrate recognition) and positions involved in protein–protein interaction. We also plan to construct comparative (homology) models for user-submitted sequences to assist in interpretation of scores.

ACKNOWLEDGEMENTS

We thank Dr Tom Alber and anonymous referees for helpful comments on the manuscript and recommendations for future developments of the INTREPID web server.

FUNDING

Presidential Early Career Award for Scientists and Engineers (from the National Science Foundation, grant number 0238311 to K.S.); Microbial Genome Sequencing Program of the National Science Foundation (grant number 0732065 to K.S.); National Institutes of Health (grant number HG002769 to K.S.). Funding for open access charge: National Institutes of Health grant HG002769 and National Science Foundation grant 0732065 to KS.

Conflict of interest statement. None declared.

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