**firestar**—advances in the prediction of functionally important residues

Gonzalo Lopez¹ ², Paolo Maietta¹, Jose Manuel Rodriguez¹, Alfonso Valencia¹ and Michael L. Tress*

¹Structural Computational Biology Group, Spanish National Cancer Research Centre (CNIO), c. Melchor Fernandez Almagro, 3, 28029 Madrid, Spain and ²Center for Computational Biology and Bioinformatics, Columbia University, New York, NY 10032, USA

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**ABSTRACT**

**firestar** is a server for predicting catalytic and ligand-binding residues in protein sequences. Here, we present the important developments since the first release of **firestar**. Previous versions of the server required human interpretation of the results; the server is now fully automatized. **firestar** has been implemented as a web service and can now be run in high-throughput mode. Prediction coverage has been greatly improved with the extension of the FireDB database and the addition of alignments generated by HHsearch. Ligands in FireDB are now classified for biological relevance. Many of the changes have been motivated by the critical assessment of techniques for protein structure prediction (CASP) ligand-binding prediction experiment, which provided us with a framework to test the performance of **firestar**. URL: http://firedb.bioinfo.cnio.es/Php/FireStar.php.

**INTRODUCTION**

The ultimate goal for researchers working with experimental protein sequences is to determine function. Computational methods form the basis of initial approaches to function determination because most approaches for the characterization of molecular function are difficult, expensive and time consuming. Many methods have been developed to predict protein function in recent years and the power of homology-based function prediction methods has increased thanks to the prodigious growth in the sequence and structural databases that are due to genome sequencing projects (1) and structural genomics initiatives (2) have increased the power of homology-based function prediction methods.

As the structural databases expand and populate structural space, a great deal of interesting biological information is being generated. Much of this, such as the amino acid residues implicated in molecular interactions or catalysis, can be found at the residue level. FireDB (3) and the **firestar** web server (4) were developed specifically to make use of this data in order to predict biologically important residues in protein sequences. FireDB is a database of annotated catalytic residues and ligand-binding residues culled from the protein structures deposited in the Protein Data Bank (PDB, 5). **firestar** uses the functional information in FireDB to make predictions of ligand-binding residues and catalytic residues.

The identification of potential ligand-binding or catalytic residues can provide important clues for the design of targeted biochemical experiments, and can be a vital part of drug design and virtual screening. Ligand-binding site predictions can also be helpful in predicting general protein function, while predicted binding sites may also act as anchoring regions in the generation of structural models. Baker et al. (6) used predicted zinc-binding residues as an important constraint to limit the structural space of possible decoys in their ROSETTA algorithm.

A number of ligand-binding prediction methods have been published since 2007 (7,8), mostly motivated by the critical assessment of techniques for protein structure prediction (CASP) ligand-binding prediction experiments (9,10), which provided a blind test framework for the evaluation of ligand-binding methods.

Here, we present the new developments of **firestar**. Several new features have been incorporated into the server to improve the quality of the predictions and the usability of the web interface. CASP blind tests show that **firestar** predictions are state of the art.

**DESCRIPTION OF THE TOOL**

We developed **firestar** with the aim of predicting functional residues from the information extracted from remotely related structures. The server makes predictions...
based on local sequence conservation matches to the biologically relevant small molecule ligand binding residues in FireDB and annotated catalytic residues from the Catalytic Site Atlas (CSA, 11).

**Protocol**

The *firestar* web server works as follows:

1. Most users will input a single protein sequence, but there is also an option to search with a protein structure, either directly from the PDB or user uploaded. The sequence is extracted from the 3D structure.

2. PSI-BLAST (12) profiles are generated for the sequences from a locally generated 70% redundant database. The profiles are used to search against the FireDB template database.

3. Users may specify the BLAST e-value cut-off for the final search of the FireDB templates; note that the default e-value is intentionally high as functional information will be present in even very distantly related proteins.

4. At the same time HHsearch (13) uses hidden Markov models generated from PSI-BLAST sequences to search against a profile database created *ad hoc* from all the FireDB template sequences. *firestar* uses all the templates detected by HHsearch to predict binding residues.

5. Both sets of alignments between query sequences and FireDB templates, with their accompanying functional information, are used to predict functional sites and likely bound ligands.

6. The predicted sites are evaluated by SQUARE (14).

Over the 26 targets it predicted (see Supplementary Table S1), the sensitivity of the *firestar* predictions in CASP8 was 0.9 (90% of known functional residues were in the predictions), while the precision 0.67 (67% of predicted residues were known functional residues) suggesting a certain overprediction. Indeed, *firestar* was tuned to make predictions at a distance of 1.5 Å in CASP8, while the official distance used to define a ligand-binding residue was 0.5 Å. Most false positive predictions were residues that were next to the official binding site. Of 87 *firestar* false positives in CASP8, 63 were within 2.5 Å of the bound ligand. At a distance of 1.5 Å, *firestar* precision was 0.8 (80% of predicted residues were known functional residues), although the sensitivity dropped to 0.84. *firestar* had a better mean MCC score than all officially participating groups in CASP8, the human predictors as well as the server groups (Figures 1 and 2).

The *firestar* server participated in CASP9 and we can report the preliminary official results for the ligand-binding site prediction category. Over the 25 targets predicted by both servers, the I-TASSER server (16) had a similar performance to *firestar* (I-TASSER MCC of 0.7 and *firestar* MCC of 0.71). The human *firestar* and I-TASSER predictors were marginally better than their servers in head-to-head comparisons and two other human groups (neither of which have publicly available servers) had slightly lower MCC than *firestar*. The other 12 server groups that participated in CASP had substantially lower MCC scores than *firestar*.

Unfortunately, for technical reasons the results from CASP9 were not directly comparable with those of CASP8. CASP9 assessors had to include targets that bound to non-biological ligands such as solvents and buffers in the assessment. Over the small subset of 10 targets that did bind biological ligands, *firestar* had higher MCC scores than all server groups (a mean MCC performance of 0.5 for all predictions against the known ligand-binding residues). At a distance of 1.5 Å, *firestar* was tuned to make predictions at a distance of 1.5 Å in CASP8, while the official distance used to define a ligand-binding residue was 0.5 Å. Most false positive predictions were residues that were next to the official binding site. Of 87 *firestar* false positives in CASP8, 63 were within 2.5 Å of the bound ligand. At a distance of 1.5 Å, *firestar* precision was 0.8 (80% of predicted residues were known functional residues), although the sensitivity dropped to 0.84. *firestar* had a better mean MCC score than all officially participating groups in CASP8, the human predictors as well as the server groups (Figures 1 and 2).

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score of 0.72 against 0.65 for I-TASSER, see Supplementary Table S2).

In total, the automatic firestar server made predictions for 82 assessed targets over three CASP editions. The server failed to make a prediction for five targets (most because of a technical problem that has now been fixed). firestar correctly predicted the binding site for all targets, though not for all the binding residues. These predictions included the three free modelling targets (those without detectable structural templates) that bound ligands in the CASP7 and CASP8 editions. firestar was able to predict the binding sites for these targets because firestar does not need to build accurate 3D models to make reliable predictions.

**NEW ADDITIONS AND IMPROVEMENTS in firestar**

The three main new developments in firestar have all contributed to huge improvements in the server. From a technical point of view, the server is now much easier to use and the fact that firestar now allows high-throughput predictions adds another dimension to the prediction of functional residues. The definition of the biological

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**Figure 1.** Outstanding firestar prediction for CASP8. The prediction for target T0407, 1 of 12 targets for which firestar would have recorded the best MCC score. (A) The prediction from firestar. The residues highlighted in yellow were the prediction. (B) T0407 was a predicted metal-dependent phosphoesterase and was crystallized with three calcium atoms (shown in light green). firestar predicted all the calcium-binding residues (shown in red) without any over prediction.
relevance of bound ligands improves the accuracy of the predictions by removing false positives generated from ligands of dubious functional importance. The addition of HHsearch alignments increases the coverage of the firestar alignments.

**Automatic interpretation**

Predictions are made from the alignments generated with the templates in FireDB. Previous versions of firestar required human interpretation of the results—probable functional residues could be gleaned from a by-eye inspection of the pairwise alignments between the query sequence and the templates with bound ligands. The detailed results pages with the PSI-BLAST and HHsearch alignments are still available and are linked from the main output page. These extended results show all pairwise alignments between the query sequence and those FireDB templates that have functional annotations. These pages are important in those cases where the firestar summary pages do not return a result, because the alignments evaluated by SQUARE that are found in these pages can often give clues to possible binding sites.

Whereas the old version of firestar required expert input, the new process of predicting functional residues is completely automated. As previously the predictions from each PSI-BLAST and HHsearch alignment are evaluated separately, but now the predictions from each alignment are collated to generate an overall functional site prediction that is incorporated into a single results page. The graphical output (Figure 3A) shows the query amino acid chain coloured by relative local conservation scores and highlights predicted catalytic (green) or ligand-binding (yellow) residues. Each pocket shown in this section is the result of merging predicted functional sites where at least 40% of residues overlap.

A text summary provides information for each individual predicted binding and catalytic site, including a list of predicted residues, the mean SQUARE score for the site and which ligands are found in the homologues. In the text summary, predicted binding sites with at least 60% residue overlap are merged. Sites that bind metals are differentiated from non-metal sites regardless of the overlapping percentage.

In addition to the summary page, there are two other levels of output available to the user, the detailed HHsearch and PSI-BLAST alignment evaluation pages (Figure 3B) and the raw PSI-BLAST/HHsearch output. The detailed alignment evaluation pages show the SQUARE evaluations of each template–target alignment and how these scores relate to ligand-binding and catalytic residues. The raw output contains all the target–template alignments, including those FireDB templates with no site information.

**Alignments with HHSearch**

The new firestar release includes HHsearch as an additional means of generating alignments between the query sequence and FireDB templates. HHsearch will find different homologues and (just as important) will create different alignments from PSI-BLAST. Both PSI-BLAST and HHsearch provide a pool of input alignments that are used to generate the initial prediction. Although the
alignments generated from HHsearch are in theory more powerful than PSI-BLAST (HHpred, based on HHsearch, was rated the best performing server in the official CASP9 evaluation at the meeting in December) the alignments from PSI-BLAST and HHsearch are complementary and equally valid for the prediction of ligand-binding residues in firestar.

Both methods are set up with lax cut-offs. The reason for this is that many of the short low-scoring local alignments generated by HHsearch and PSI-BLAST include functional information. With these cut-offs, the two methods will detect remote homologues, some false positive hits and many short alignments. However, this is not important because the alignments are only used as the initial input to firestar. The evaluation is carried out by SQUARE. SQUARE locates highly conserved local regions of residues (14) within the PSI-BLAST and HHsearch alignments through profile–profile comparison. Only those template ligand-binding residues that are in aligned regions with high local conservation (according to SQUARE) can be considered as binding residues in the target. SQUARE has been shown to be particularly effective at predicting ligand-binding residues from alignments (17).

Once these potential binding residues are localized from all the alignments from HHsearch and PSI-BLAST, firestar determines whether the ligand-binding residues do form part of a functionally relevant binding site according to the numbers of residues detected (one limitation is that each binding site needs to be composed of a minimum number of highly conserved residues) and based on the biological relevance of the ligand.

The evaluation process weeds out the vast majority of the initial predictions. For example, HHsearch and PSI-BLAST generate 276 different alignments for the recent CASP target, T0614, but despite all the alignments no site was predicted for target T0614.

Above all, the effect of combining PSI-BLAST and HHsearch alignments is to extend the coverage of firestar predictions. The extended coverage will come
from two different sources: from those extra FireDB homologues that only HHsearch detects and from those alignments where HHsearch aligns correctly and PSI-BLAST does not.

For example, we have run firestar with HHsearch and PSI-BLAST alignments for part of the human genome. Adding HHsearch alignments increases coverage by 34%. We ran firestar for all 798 genes in chromosomes 21 and 22 annotated by Gencode in their 3C release (18). For the transcripts from these genes, PSI-BLAST–firestar predicts 12,657 ligand-binding residues. HHsearch–firestar predicts 15,078 residues and combining alignments from the two methods helps firestar to predict 17,027 ligand-binding residues.

**Biological relevance**

The PDB contains a diverse range of functional information that can be automatically collected. Unfortunately, much of it is redundant and many PDB files contain artefacts and molecular data without strict biological meaning. Solvent molecules and crystal packing effects produce interatomic contacts between amino acids and heteroatoms that may not have biological relevance. Many PDB structures are crystallized with inhibitors.

FireDB collects and organizes all protein–small ligand interactions in the PDB. FireDB is built around templates generated from a 97% redundant version of the PDB and all protein–small ligand interactions are mapped onto these templates. All functional residues in the FireDB repository are now classified in terms of their biological relevance using evolutionary information, structural data and lists of known cognate ligands. All protein–ligand interactions in FireDB are classified as biologically relevant, putative or non-relevant.

Cognate ligands are those found in PROCOGNATE (19). However, we filter out those that are commonly added as a part of the crystallization process. This excludes most ions, water and solvent molecules such as glycerol. Inhibitors are accepted as ligands as long as they can act as analogues of the cognate ligands.

Evolutionary information for biological relevance analysis is obtained by running firestar for all FireDB templates against the FireDB template database. This allows us to cluster together all binding sites that are evolutionarily related. This information is accessible through the FireDB web services and a detailed description is provided in the online help. FireDB also computes the average number of residues that bind each ligand. It has been previously reported that high connectivity is a good descriptor of biological relevance (20).

Biologically relevant protein–ligand interactions are those that involve cognate ligands with at least one evolutionarily related site in FireDB. In the absence of evolutionary information, protein–ligand interactions are considered putative if the ligand is in the cognate list and the number of residues implicated in binding is over two-thirds of the average number for the ligand. Predictions by firestar are only made from biologically relevant or putative binding sites.

Further information on the decision-making process involved in determining biological relevance can be found on the web pages.

FireDB is regularly updated with new structures. The greater the amount of functional information in FireDB, the more sequence space can be covered by firestar. There has been an increase of 8,608 templates with functional sites, since firestar was first presented in 2007. The most recent version of FireDB contains 18,048 templates, of which 14,770 contain putative or biologically relevant sites.

The number of binding sites in the database has more than doubled from 38,865 to 86,379, and half of these (41,063) are classified as putative or biologically relevant sites. Only biologically relevant and putative sites are considered by firestar for the predictions on the summary pages. The remaining sites are still available through PDB code queries in the FireDB web pages.

**High-throughput mode**

The firestar server has now been enabled to work in high-throughput mode and can be easily integrated into servers either through the server or as a web service. At present it plays an important role as a part of the APPRIS pipeline to annotate splice variants (21) as a part of the ENCODE project. Predictions for the human genome are accessible through APPRIS (appris.bioinfo.cnio.es). The web service differs from the web server in that it predicts only ligand-binding residues and a confidence score for each residue.

**FUTURE IMPROVEMENTS**

During the CASP9 prediction edition, our group participated with two predictors, the fully automatic server and a version of firestar that used 3D models to extend firestar predictions. The preliminary results that suggested a slight improvement is obtained by using models.

Given that structural information frequently gives insights about binding mechanisms and ligand-binding specificities, we are working to implement 3D model prediction in firestar. Future versions of firestar will allow users to retrieve models with the predicted ligand bound to the structure. This is an important feature in which potential users of firestar will be interested, even if the improvement in the accuracy of ligand-binding prediction is not always substantial.

We would like to add more sources of annotated functional residues beyond those that are in the PDB and CSA, such as the annotated functionally important residues that are available in a number of sequence databases. Adding further search and alignment methods ought to generate incremental improvements in coverage, although this would affect the performance of firestar.

We are working to refine our definition of biological ligands by highlighting those non-cognate ligands that are of pharmacological or chemical importance.
SUPPLEMENTARY DATA
Supplementary Data are available at NAR Online.

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Conflict of interest statement. None declared.

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