InterEvDock2: an expanded server for protein docking using evolutionary and biological information from homology models and multimeric inputs

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ABSTRACT

Computational protein docking is a powerful strategy to predict structures of protein-protein interactions and provides crucial insights for the functional characterization of macromolecular cross-talks. We previously developed InterEvDock, a server for ab initio protein docking based on rigid-body sampling followed by consensus scoring using physics-based and statistical potentials, including the InterEvScore function specifically developed to incorporate co-evolutionary information in docking. InterEvDock2 is a major evolution of InterEvDock which allows users to submit input sequences – not only structures – and multimeric inputs and to specify constraints for the pairwise docking process based on previous knowledge about the interaction. For this purpose, we added modules in InterEvDock2 for automatic template search and comparative modeling of the input proteins. The InterEvDock2 pipeline was benchmarked on 812 complexes for which unbound homology models of the two partners and co-evolutionary information are available in the PPI4DOCK database. InterEvDock2 identified a correct model among the top 10 consensus in 29% of these cases (compared to 15–24% for individual scoring functions) and at least one correct interface residue among 10 predicted in 91% of these cases. InterEvDock2 is thus a unique protein docking server, designed to be useful for the experimental biology community. The InterEvDock2 web interface is available at http://bioserv.rpbs.univ-paris-diderot.fr/services/InterEvDock2/.

INTRODUCTION

Computational modeling of protein assemblies provides crucial insights for the functional characterization of macromolecular interactions occurring in the crowded cellular environment. Predictions of protein–protein interfaces can be used to design experiments to investigate the role of important interactions and possibly interfere with them, typically using mutagenesis. Models of macromolecular complexes are also useful to complement integrative structural biology (1) and to deepen our understanding of disease-associated mutations (2) and protein interaction networks (3).

A number of servers have been developed for protein–protein docking, which can be separated into template-based modeling servers, which aim to identify suitable structural templates for the protein complex, and template-free (ab initio) docking servers, which implement sampling of interface structural models followed by scoring protocols. Recent resources to find templates for interface modeling starting from the sequences of two protein partners include KBDock (4), focused on domain–domain interactions and PPI3D (5). Recently released servers taking protein sequences as input for homology-based interface modeling include SnapDock (6), HOMCOS (7) and SWISS-MODEL Oligo (8). Many template-free docking servers implement a rigid-body docking approach, sometimes followed by rescoring: PatchDock (9), FireDock (10), HexServer (11), ZDOCK (12), FRODOCK 2.0 (13), pyDockWEB (14), ClusPro (15), GRAMM-X (16), InterEvDock (17). A hybrid approach combining template-based and template-free docking was recently proposed in the HDOCK server (18). Some free docking servers include specific features such as symmetric docking (SymmDock (9), ZDOCK); local docking around an initial guess (RosettaDock (19), re-
cently moved to the ROSIE server (20); docking with more
than two proteins (ClusPro, GRAMM-X); and docking in-
cluding degrees of flexibility (SwarmDock (21), ATTRACT
(22), HADDOCK (23)).

Attempts to address the limitations of computational
docking have led to placing increasing focus on data-driven
docking (24) and many servers now allow the user to spe-
cify interface residues and/or distance restraints, including
ZDOCK, FRODOCK 2.0 (for refinement), pyDockWEB,
ClusPro, GRAMM-X, SwarmDock, HADDOCK and AT-
TRACT. Some servers such as ClusPro and pyDockSAXS
(25) can specifically use experimental small angle X-ray
scattering (SAXS) data.

One of the main features differentiating existing docking
servers is the nature of the scoring function used to discrim-
inate correct from incorrect docking models. Most scoring
strategies use either physics-based or statistical potentials.
Understanding how binding partners co-evolved can pro-
vide essential clues to improve the structural prediction of
protein interfaces. Following pioneering studies (26–28),
the field of covariation-based contact prediction has been very
active in recent years, with several servers enabling the pre-
diction of inter-molecular contacts such as EVcomplex (29),
GREMLIN (30) and I-COMS (31); however, such meth-
ods still have limited applicability due to the difficulty in
building large enough joint multiple sequence alignments
(MSAs) for the two protein partners. We developed the In-
terEvScore scoring function incorporating co-evolutionary
information into the docking process, which improves pre-
dictions for as few as 10 sequences in the joint MSAs (32).
We integrated this scoring function into the InterEvDock
pipeline (17). InterEvDock is based on rigid-body sampling
by FRODOCK (33) followed by re-scoring using the SOAP-
PP atomic statistical potential (34) and InterEvScore (32)
and consensus model selection. To date, the InterEvDock
web server is the only free docking server allowing to di-
rectly predict the structure of protein–protein interactions
using co-evolutionary information. We successfully used
the InterEvDock strategy to guide our predictions in re-
cent Critical Assessment of Predicted Interactions (CAPRI)
rounds: for CAPRI rounds 28–35, our group ranked first by
making correct predictions for 10 out of 18 targets (35).

Very often, the individual structures of the exact pro-
teins involved in a complex of biological interest are not
known. On the other hand, structural models can be ob-
tained for a large fraction of proteins in interaction net-
works thanks to homology modeling (36), making them
amenable to protein–protein docking. To date, most free
docking servers, except the HDOCK server, allow users to
provide only input structures but no input sequences for the
protein partners.

Based on the user-oriented considerations mentioned
above, here we introduce the InterEvDock2 server which
represents a major evolution over the original InterEvDock.
Protein sequences can now be provided as input, and not
only 3D structures. To handle sequence inputs, we have
added a module that performs comparative modeling prior
to docking based on an automatic template search protocol.
In case the user has biological input such as a position that is
known to be involved in the interface between the two pro-
tein partners or a pair of residues known to be in contact,
Figure 1. Workflow of the InterEvDock2 pipeline. Eight steps can be performed in InterEvDock2 depending on the user input, out of which three mandatory steps (iv), (vii) and (viii) are always performed as they were in the original InterEvDock pipeline (17), except that the FRODOCK algorithm was updated to version 2.1 (13) and the consensus calculation was slightly modified to save time and improve results. New features are available allowing the user to provide only an input sequence for one or both partners: (i) if the user does not provide a template, search for a suitable template using HHsearch (38,39); (ii) if the user provides a template but no query-template alignment, alignment of query sequence with template sequence using MAFFT (40); (iii) once a template and a query-template alignment are available for each partner with no user-provided structure, comparative modeling using a Rosetta Scripts (41) protocol based on RosettaCM (42) to build a 3D model for (at least part of) the input sequence. Once a 3D structure or a structural model is available for each partner: (iv) exhaustive sampling using the rigid-body method FRODOCK 2.1; (v) (new feature) if the user provides information on residues (or pairs of residues) involved in the interface, applying constraints to filter sampled solutions; (vi) if the user does not provide a joint MSA for the two protein partners, co-MSA generation; (vii) clustering and scoring by three scores, FRODOCK 2.1, SOAP-PP (34) and InterEvScore (32); (viii) clustering and selection of the InterEvDock2 consensus top ten decoys. Green text indicates user input. Red text indicates possible breakpoints and hot restart. Italics indicate the optional steps in the pipeline, depending on the input provided by the user. Details for each step are provided in the Supplementary Methods and Figure S1.

used for each constraint. An InterEvDock2 session identifier can also be provided in order to re-use docking results from a previous run and test different constraints. As in InterEvDock, users may input the joint MSAs used for co-evolution-based scoring; otherwise the joint MSAs will be built by the server through an automated procedure. In case an oligomeric structure is submitted as one of the two docking partners, the joint MSAs will also be automatically calculated and processed by the server for every chain of the oligomer. A demonstration case using sequences as input to the docking (and optionally a constraint) is available from the InterEvDock2 submission page.

The web page resulting from an InterEvDock2 submission contains information about the best ranked decoys, which can be explored interactively thanks to the PV WebGL applet (M. Biasini, https://dx.doi.org/10.5281/zenodo.12620). Detailed results are available in a downloadable archive, also containing a script for easy loading and offline visualization of the best docking solutions with PyMOL (The PyMOL Molecular Graphics System, Schrödinger, LLC, New York, NY).

The InterEvDock2 server benefits from parallelized implementation in the dedicated infrastructure built at RPBS and from data privacy ensured in the Mobyle framework.

Molecular docking procedure

Figure 1 presents the InterEvDock2 pipeline which consists of eight steps; more details about each step are provided in Supplementary Figures S1 and S2 and in the Supplementary Methods. The three core docking steps – sampling
Figure 2. Illustration of new features available in InterEvDock2. Among advanced options, interactive template selection prior to modeling and docking is now available as an advanced workflow option (breakpoint). The list of templates returned by the server can be interactively explored by the user to select a template. The provided query-template alignment can then be submitted as input to the server for modeling and docking.

with FRODOCK2.1 (iv), clustering with FRODOCK2.1 and scoring with InterEvScore and SOAP-PP (vii) and consensus calculation (viii) – are always performed. Step (vi) consists in automatically generating the joint MSAs used by InterEvScore to account for co-evolution in the scoring process, unless the joint MSAs are provided by the user. Steps (iv), (vi) and (vii) are unchanged compared to the original InterEvDock pipeline (17), except that the FRODOCK algorithm was updated to version 2.1 (13). In the final step (viii) a consensus list of 10 most likely models is calculated. Since decoys well ranked by at least two different scoring methods (out of the three methods used in InterEvDock2) have higher chances of being correct, the 3*top 10 models for each score are re-ranked according to the number of similar decoys (defined as ligand RMSD ≤ 10 Å) within the top 50 models of the other two scores (down to a minimum of two similar decoys). In case of a tie, priority is given to InterEvScore top 10 models, then SOAP-PP, then FRODOCK. If necessary, the consensus list is then filled up to 10 models by selecting the best models from each score (four from InterEvScore, three from SOAP-PP and three from FRODOCK). When building the consensus, models that are structurally redundant (i.e. ligand RMSD ≤ 10 Å) with previously selected models are excluded, so that the final list contains 10 structurally non-redundant models.

Docking from input sequences

If the user provides only an input sequence for one or both partners, steps (i) to (iii) can be applied. (i) If the user does not provide a template, the profile–profile comparison tool HHsearch is used to search for templates (38,39); only templates with HHsearch probability higher than 95% are selected. The web server returns a list of up to 20 templates selected according to HHsearch probability, query-template sequence identity and structural resolution (see details in the Supplementary Methods). By setting the breakpoint after template search, the user can choose to start modeling from any of these templates by copy-pasting the query-template alignment to the server submission form; otherwise the best template found by the automatic procedure is used. If no suitable template is identified, no modeling is performed. (ii) If the user provides a template but no query-template alignment, the query sequence is aligned with the template sequence using MAFFT (40). (iii) Once a template and a query-template alignment are available for each protein with no user-provided structure, comparative modeling using a RosettaScripts (41) protocol based on RosettaCM (42) is performed to build a 3D model for (at least part of) the input sequence. Due to runtime considerations, compared to the procedure used to build the PPI4DOCK database (37), the comparative modeling protocol implemented in the InterEvDock2 web server involves fewer optimization cycles (see protocol details in the Supplementary Methods). This protocol is quite robust for templates with relatively high homology but it can lead to loss of precision for more remote templates (typically when both templates have <50% sequence identity with the query proteins). By default, to avoid spending time reconstructing regions that are not present in the template, only loops (insertions) shorter than 14 residues are rebuilt during the modeling and N-terminal and C-terminal extensions are not mod-
eled, but maximal lengths for modeling of loops, N-terminal and C-terminal extensions can be tuned by the user.

User-defined constraints

Step (v) applies if the user provides information on residues (or pairs of residues) involved in the interface: restraints are applied to filter sampled solutions. The distance used to enforce restraints can be modulated which offers the possibility to integrate data from various sources. The default distance was set to 8 Å for constraints on single positions and 11 Å for pair constraints (see Supplementary Results for a detailed justification of these thresholds). When constraints are provided by the user, the output returned by the server will provide information about whether or not each constraint was used during docking (e.g. constraints on residues not exposed on the surface of the protein are excluded).

Runtime

The core docking steps (iv), (vii) and (viii) take altogether around 30 min for proteins of size 200 residues and 1 h for proteins of size 400–500 residues. Template search and query-template alignment steps (i) and (ii) take only a few minutes, whatever the size of the proteins. The comparative modeling step (iii) was optimized for speed as reported above and typically takes 5–20 min depending on the size of the proteins and the query-template sequence identities. Compared to InterEvDock, InterEvDock2 benefits from a large speed-up in step (vi) for the generation of joint MSAs for two protein partners which was a key bottleneck. This step now typically lasts ~3 min for proteins of 200 residues and ~15 min for proteins of 400–500 residues.

RESULTS

Benchmarking on PPI4DOCK

To assess the predictive power of the InterEvDock2 server on 3D models, we have set up the most extensive benchmark to date, using unbound models as input of the docking simulations. The PPI4DOCK database (37) was designed to ensure unbiased evaluation of free docking performance and contains 1417 non-redundant heterodimeric docking targets based on unbound homology models. The InterEvDock2 pipeline was tested on the subset of 812 protein complexes from PPI4DOCK for which pairs of joint MSAs with more than 10 sequences could be obtained (excluding any antibody complex) and FRODOCK 2.1 (13) was able to generate at least one acceptable or better decoy among the top 50 000 decoys. The list of the 812 complexes used for benchmarking and detailed results are provided in http://bioserv.rpbs.univ-paris-diderot.fr/services/InterEvDock2/table.html. This benchmark dataset is roughly an order of magnitude larger than other typical docking benchmarks, among which the widely used Weng benchmark (43). For each of the 812 targets, PPI4DOCK provides unbound homology models of the two protein partners as well as the joint MSAs used for docking and scoring in the InterEvDock2 pipeline. As on the web server, the predictions for each case consist in the top 10 consensus interface models and the top 10 interface residues, which are used to assess the InterEvDock2 performance. A solution is defined as acceptable or better according to the criteria defined by the CAPRI consortium (44).

The prediction performance of InterEvDock2 is reported in Table 1. Among the 812 targets, 29% (239) have at least one model of acceptable or better quality in the top 10 consensus obtained from the InterEvDock2 pipeline, which represents a significant improvement over the top 10 success rates of the three individual scores used to build the consensus (see also Supplementary Figure S3). The 812 complexes belong to four difficulty levels (PPI4DOCK categories) based on the quality of the superimposed interface model (two unbound models superimposed on the bound structure): ‘very easy’ (174 complexes), ‘easy’ (498 complexes), ‘hard’ (118 complexes) and ‘very hard’ (22 complexes). Other ‘very hard’ and all ‘super hard’ PPI4DOCK targets do not satisfy the condition that FRODOCK 2.1 was able to generate at least one acceptable or better decoy among the top 50 000 decoys, since they may require flexibility in the docking process (37), and are therefore not included in the present benchmark. As expected, the InterEvDock2 top 10 consensus success rate decreases with increasing difficulty of the test cases, from 43% for the ‘very easy’ PPI4DOCK category to 30% for the ‘easy’ category, 11% for the ‘hard’ category and 5% for the ‘very hard’ category. Analysis of InterEvDock2 performance depending on the minimum sequence identity between the target and template shows a moderate drop in success rate for models built with remote templates (<30% sequence identity) and an increased success rate for models built with very close templates (≥95% sequence identity), compared to the overall InterEvDock2 success rate (see Supplementary Table S1).

Direct comparisons with previous benchmarks are difficult because the benchmark dataset used here is much larger than other datasets typically used to assess docking and scoring performance. Comparison with previously reported success rates on the Weng benchmark (17,43) are provided in Supplementary Tables S2 and S3. An interesting feature of the Weng benchmark compared to PPI4DOCK is that it contains targets where one partner is multimeric. Out of the 85 cases from the Weng benchmark that can be used for InterEvDock2 benchmarking, 16 contain a multimeric partner. The InterEvDock2 top 10 consensus contains an acceptable or better solution for 7 out of these 16 cases (44%). This success rate is comparable to the overall success rate of InterEvDock2 on the much larger PPI4DOCK benchmark (29%) and on the 85 cases of the Weng benchmark (32%). Additionally, docking using multimeric partners has the advantage that potentially ‘sticky’ interface regions involved in multimeric interactions of one partner are buried in the multimeric interface and therefore masked for the docking process.

In Table 1, the InterEvDock2 performance is also compared to the performance of the widely used rigid-body docking program Zdock3.0.2 (45), assessed on the same 812 complexes from the PPI4DOCK benchmark. For each case, 54 000 decoys are generated and ranked by Zdock3.0.2. In 126 out of 812 cases (15%), an acceptable or better solution is found among the top 10 decoys. Altogether, these benchmarking results highlight the added value of the Inter-
Table 1. Prediction performance of the InterEvDock2 server on 812 complexes of the PPI4DOCK benchmark, split into four levels of difficulty: very easy, easy, hard and very hard.

| Number of cases | All | Very easy | Easy | Hard | Very hard |
|-----------------|-----|-----------|------|------|-----------|
| Top 10 success rate | InterEvScore | 171 (21%) | 44 (25%) | 115 (23%) | 11 (9%) | 1 (5%) |
|                  | SOAP_PP | 194 (24%) | 55 (32%) | 126 (25%) | 12 (10%) | 1 (5%) |
|                  | FRODOCK 2.1 | 164 (20%) | 55 (32%) | 102 (20%) | 5 (4%) | 2 (9%) |
|                  | InterEvDock2 consensus | 239 (29%) | 75 (43%) | 150 (30%) | 13 (11%) | 1 (5%) |
| Residue interface prediction (≥1 correct in top 5 receptor or top 5 ligand) | Zdock 3.0.2 | 126 (15%) | 33 (19%) | 83 (17%) | 9 (8%) | 1 (5%) |
|                  | InterEvDock2 | 735 (91%) | 160 (92%) | 450 (90%) | 103 (87%) | 22 (100%) |
| Residue interface prediction (≥1 correct in top 5 receptor and top 5 ligand) | Zdock 3.0.2 | 680 (84%) | 145 (83%) | 427 (86%) | 91 (77%) | 17 (79%) |
|                  | InterEvDock2 | 414 (51%) | 103 (59%) | 263 (53%) | 39 (33%) | 9 (41%) |
| Residue interface prediction (≥1 correct in top 1 receptor or top 1 ligand) | Zdock 3.0.2 | 345 (43%) | 76 (44%) | 228 (46%) | 33 (28%) | 8 (34%) |
|                  | InterEvDock2 | 613 (75%) | 140 (80%) | 385 (77%) | 71 (60%) | 17 (77%) |
| Residue interface prediction (≥1 correct in top 1 receptor and top 1 ligand) | Zdock 3.0.2 | 532 (66%) | 111 (64%) | 344 (69%) | 64 (54%) | 13 (58%) |
|                  | InterEvDock2 | 278 (34%) | 75 (43%) | 184 (37%) | 17 (14%) | 2 (9%) |

The benchmark is made of the 812 targets of the PPI4DOCK benchmark (1417 cases) (37) for which pairs of co-evolved MSAs with more than 10 sequences could be obtained and FRODOCK 2.1 (13) was able to generate at least one acceptable or better decoy (44) among the top 50 000 decoys. In the lower part of the table, top 10 success rates are reported as the number of cases (and percentage between brackets) for which at least one model out of 10 is an acceptable or better solution. Assessed methods are InterEvScore (32), SOAP-PP (34), FRODOCK 2.1 (13), InterEvDock2 consensus (this work and 17) and Zdock3.0.2 (45). In the lower part of the table, the number (and percentage) of cases for which at least one residue out of the top 10 or top 2 residues was correctly predicted as present in the complex interface is assessed for InterEvDock2 and Zdock3.0.2 (details are provided in the Supplementary Data).

The best results for each category are highlighted in bold.

terEvDock2 processing pipeline, in particular the clustering and consensus scoring steps.

Of key interest for experimental biologists, the InterEvDock2 output offers a list of 10 residues most likely involved in the complex interface (5 predicted residues on each partner) that can be targeted for mutagenesis. For these residue predictions, we reach 91% success rate, with 735 of the 812 benchmark cases having at least one of the 10 predicted residues involved in the actual interface (Table 1). As was found for the 85 cases from the Weng benchmark used to assess the original InterEvDock performance (17), this success rate is remarkably stable with increasing difficulty: from 92% for very easy cases to 90% for easy cases to 87% for hard cases. Predictions of the InterEvDock2 server can also be used as a prior to constrain more thorough docking simulations including flexibility. In that perspective, in 51% of the cases, at least one correct residue is predicted on both sides of the interface (59% for very easy targets, 53% for easy targets and 33% for hard targets). Results are also presented in Table 1 and Supplementary Figure S4 for only the top 2 predicted residues (one on each partner): at least one of the two predicted residues is correct in 75% of the cases and both are correct in 34% of the cases, highlighting the practical value of InterEvDock2 residue prediction. All those results are significantly higher than a reference interval given by random selection of residues on the surface of the protein (see Supplementary Methods and Figure S4).

### Predictions of CAPRI targets

The InterEvDock2 pipeline was challenged through our participation in all CAPRI rounds since 2013. Focusing on heteromeric targets evaluated at the sixth CAPRI evaluation meeting (rounds 28–35), our group ranked first with 10 correctly predicted targets out of 18. Among those 10 targets, our best prediction among 10 submitted models was of high quality in one case, medium quality in seven cases and acceptable quality in two cases (35). In 15 of the total 18 targets, evolutionary information was available in the form of either co-evolution or conservation, providing key constraints to guide docking toward the correct solution. Although the InterEvDock2 pipeline was not specifically designed to handle homo-oligomeric docking, we were also among the highest ranking groups in the two joint CASP-CAPRI experiments involving mostly predictions of homodimers (46,47). Of note, for most CAPRI targets since 2013, only sequence information was provided to the participants. Figure 3 illustrates an InterEvDock2 run for CAPRI target T95 (round 31) involving docking between the nucleosome (a decameric structure) and the PRC1 ubiquitin ligase (a trimeric structure).

### Description of docking case studies from input sequences and using constraints

To illustrate the biological relevance of InterEvDock2 predictions, we consider two docking case studies derived from the PPI4DOCK benchmark (Figure 4). The first case is a complex between the Rho-family GTPase Cdc42 and the conserved, catalytic domains of exchange factor intersectin. Details of this interaction (PDB identifier: 1ki1) and structure-based mutagenesis revealed key features of the activation of Cdc42 by intersectin (48). This case was tested on the InterEvDock2 server by providing input sequences of the interacting regions in the two partners. Unbound templates were imposed for both proteins as in the PPI4DOCK benchmark; otherwise the automatic template search might
have found the bound partners belonging to PDB identifier: 1ki1 or other bound templates. The unbound templates (4f38A and 3odoA) have sequence identities of 54 and 25% with the modeled regions of Cdc42 and intersectin, respectively. Among the top 10 consensus models returned by InterEvDock2, one acceptable solution is found as top 2 (Figure 4A).

The second case illustrates the interest of docking with user-defined restraints. We consider a complex between the RING domain of E3 ubiquitin protein ligase IDOL and ubiquitin-conjugating enzyme E2 UBE2D (PDB identifier: 2yho) (49). This interaction is involved in the regulation of cholesterol uptake. Nuclear magnetic resonance (NMR) chemical shift mapping was used to confirm the interacting region prior to crystallographic studies. This NMR analysis showed four residues (M388, V389, C390 and C391) in the RING domain of IDOL to have particularly high chemical shift variation upon binding of UBE2D. The PPI4DOCK models of the interacting regions of IDOL and UBE2D (built by homology modeling using unbound templates for the two proteins, respectively, 2yhnA and 3bzhA with sequence identities of 100 and 61%) were submitted to InterEvDock2. Two runs were performed, one without constraints and one using the four residues identified by NMR as interface constraints. Among the top 10 consensus models returned by InterEvDock2, the highest-ranked acceptable solution (medium quality according to the CAPRI criteria) was ranked number six in the run without constraints. When using the constraints derived from experimental NMR data, there were two acceptable or better solutions in the InterEvDock2 top 10 consensus: one was ranked first (Figure 4B) and the second ranked sixth.

CONCLUSION

InterEvDock2 represents a major, user-oriented evolution of InterEvDock. InterEvDock2 is still the only free docking server taking into account co-evolutionary information, relying on a combination of complementary scoring functions to identify the most likely interface models. The previous InterEvDock version was limited by its requirement of only dealing with monomeric inputs. InterEvDock2 greatly expands the range of applications to homo- and hetero-oligomers by handling multimeric chains in the two input proteins used for pairwise docking and the automated processing of their joint MSAs. Benchmarking results on PPI4DOCK emphasize the usefulness of InterEvDock2 in generating interface models of good quality in the scope of integrative structural biology. The InterEvDock2 server returns docking results within typical runtimes of 30 min (for proteins of around 100 residues) to 2 h (for proteins of around 500 residues) even when starting from input sequences, while performing well on our benchmark of 812 cases docked from unbound homology models. The server also benefits from a user-friendly submission and visualization interface, including breakpoints after template search and homology modeling, and options for offline in-depth analysis with PyMOL. InterEvDock2 is thus designed as a useful tool for biologists who can very easily submit docking runs starting from simple input sequences and specify constraints to make use of any previously acquired experimental knowledge. InterEvDock2 results can assist biologists in designing hypotheses about molecular interaction mechanisms and interface mutations to investigate the functional role of an interaction.

SUPPLEMENTARY DATA

Supplementary Data are available at NAR Online.
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