

Supplementary Information for

Recent progress of protein tertiary structure prediction

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Table of Content

Supplementary Figures

Fig. S1. An illustration of co-evolutionary information contained in multiple sequence alignments and the corresponding relationship with residue-residue contact prediction.

Supplementary Tables

Table S1. Tools for template-based protein structure prediction.

Table S2. Tools for template free (free modeling) protein structure prediction.

Table S3. Tools for contact-based protein structure prediction.

Table S4. Tools for distance-based protein structure prediction.

Table S5. Tools for end-to-end protein structure prediction.

Table S6. Tools for protein language model-based protein structure prediction.

Table S7. Tools for multi-domain protein structure prediction.

Table S8. The monomer protein dataset from CASP14 used in our benchmark tests.

Supplementary Figures

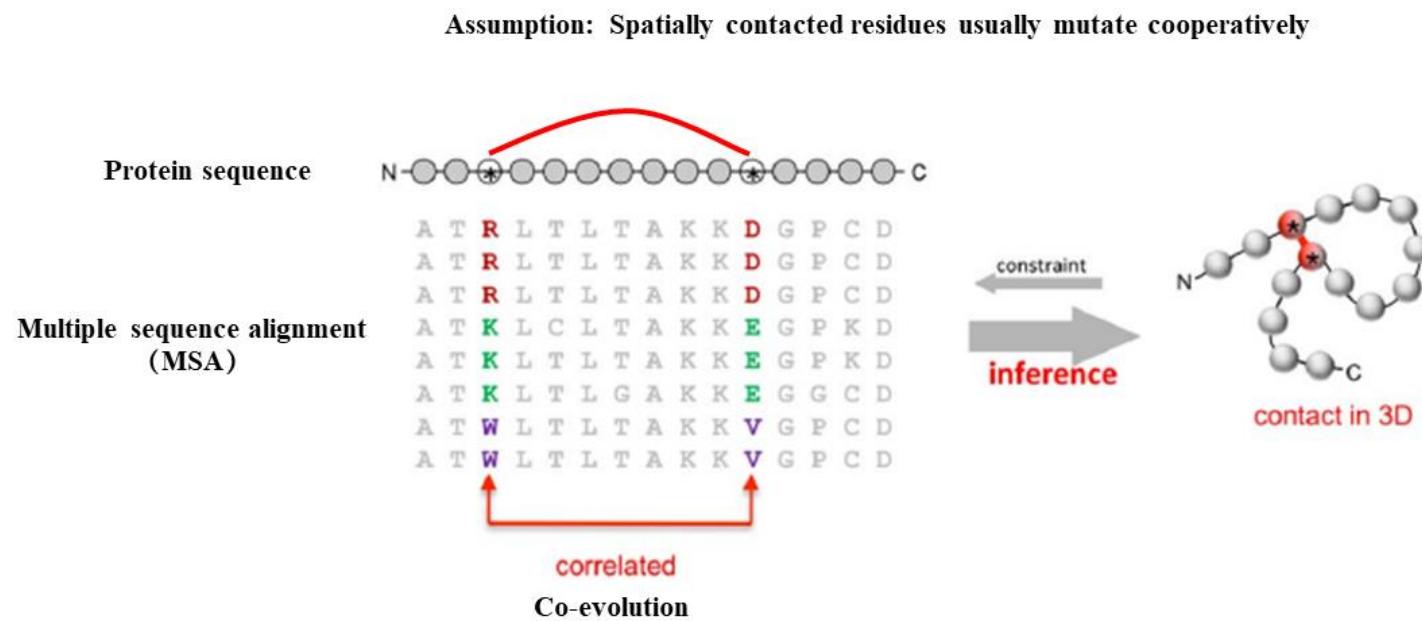


Figure S1. An illustration of co-evolutionary information contained in multiple sequence alignments and the corresponding relationship with residue-residue contact prediction.

Supplementary Tables

Table S1. Tools for template-based protein structure prediction.

| Methods | URLs | Reference |
|---------------|--|-----------|
| CEthreader | https://zhanglab.dcmb.med.umich.edu/CEthreader/ | [1] |
| DEDAL | https://github.com/google-research/google-research/tree/master/dedal | [2] |
| DisCovER | https://github.com/Bhattacharya-Lab/DisCovER | [3] |
| EBA | https://git.scicore.unibas.ch/schwede/EBA https://git.scicore.unibas.ch/schwede/eba_benchmark | [4] |
| EigenTHREADER | https://github.com/psipred/eigenthreader | [5] |
| FFAS-3D | https://ffas.godziklab.org/ffas-cgi/cgi/ffas.pl | [6] |
| HHpred | https://github.com/soedinglab/hh-suite/tree/master/scripts/hhpred | [7] |
| HHsearch | https://github.com/soedinglab/hh-suite | [8] |
| I-TASSER | https://zhanglab.dcmb.med.umich.edu/I-TASSER/ | [9-11] |
| LOMETS | https://zhanglab.dcmb.med.umich.edu/LOMETS/ | [12-14] |
| map_align | https://github.com/sokrypton/map_align https://github.com/gjoni/map_align | [15] |
| MODELLER | https://salilab.org/modeller/ | [16] |
| MRFalign | http://raptorx6.uchicago.edu | [17] |
| MULTICOM | https://multicom-toolbox.mu.hekademieia.org | [18, 19] |
| MUSTER | https://zhanglab.dcmb.med.umich.edu/MUSTER/ | [20] |
| Phyre2 | http://www.sbg.bio.ic.ac.uk/phyre2/ | [21] |
| pLM-BLAST | https://toolkit.tuebingen.mpg.de/tools/plmblast https://github.com/labstructbioinf/pLM-BLAST | [22] |
| RaptorX | https://github.com/j3xugit/RaptorX-3DModeling http://raptorx6.uchicago.edu | [23, 24] |

| | | |
|------------|---|------|
| RosettaCM | https://www.rosettacommons.org/software/license-and-download | [25] |
| SAdLSA | https://sites.gatech.edu/cssb/sadlsa/ | [26] |
| SparksX | https://sparks-lab.org/server/sparks-x/ | [27] |
| ThreaderAI | https://github.com/ShenLab/ThreaderAI | [28] |

Table S2. Tools for template free (free modeling) protein structure prediction.

| Methods | URLs | Reference |
|----------|--|-----------|
| FragFold | https://github.com/psipred/fragfold | [29] |
| QUARK | https://zhanglab.dcmb.med.umich.edu/QUARK/ | [30] |
| Rosetta | https://rosetta.bakerlab.org https://www.rosettacommons.org/software/license-and-download | [31, 32] |

Table S3. Tools for contact-based protein structure prediction.

| Methods | URLs | Reference |
|-----------------|--|-----------|
| CEthreader | https://zhanglab.dcmb.med.umich.edu/CEthreader/ | [1] |
| C-I-TASSER | https://zhanglab.dcmb.med.umich.edu/C-I-TASSER/ | [33] |
| CONFOLD2 | https://github.com/multicom-toolbox/CONFOLD2/ | [34] |
| C-QUARK | https://seq2fun.dcmb.med.umich.edu/C-QUARK/ | [35] |
| EigenTHREADER | https://github.com/psipred/eigenthreader | [5] |
| LOMETS2 | https://zhanglab.dcmb.med.umich.edu/LOMETS/ | [12-14] |
| map_align | https://github.com/sokrypton/map_align https://github.com/gjoni/map_align | [15] |
| RaptorX-Contact | https://github.com/j3xugit/RaptorX-Contact | [36] |

Table S4. Tools for distance-based protein structure prediction.

| Methods | URLs | Reference |
|---------------|---|-----------|
| DeepFold | https://zhanggroup.org/DeepFold/ | [37] |
| DeepPotential | https://zhanggroup.org/DeepPotential/ | [38] |
| DeepThreader | http://raptorx6.uchicago.edu | [39] |
| DisCovER | https://github.com/Bhattacharya-Lab/DisCovER | [3] |
| D-I-TASSER | https://zhanggroup.org/D-I-TASSER/ | [40, 41] |
| DMPfold | http://bioinf.cs.ucl.ac.uk/psipred/ https://github.com/psipred/DMPfold | [42] |
| D-QUARK | https://zhanggroup.org/D-QUARK/ | [41] |
| tFold | https://drug.ai.tencent.com/console/en/tfold | [43] |
| trRosetta | https://yanglab.qd.sdu.edu.cn/trRosetta/ https://github.com/gjoni/trRosetta https://github.com/RosettaCommons/trRosetta2 | [44, 45] |

Table S5. Tools for end-to-end protein structure prediction.

| Methods | URLs | Reference |
|-------------|---|-----------|
| AlphaFold2 | https://github.com/deepmind/ https://github.com/google-deepmind/alphafold | [46] |
| ColabFold | https://github.com/sokrypton/ColabFold | [47] |
| FastFold | https://github.com/hpcaitch/FastFold | [48] |
| HelixFold | https://github.com/PaddlePaddle/PaddleHelix/tree/dev/apps/protein_folding/helixfold | [49] |
| MEGA-Fold | https://gitee.com/mindspore/mindscience/tree/master/MindSPONGE/applications/MEGAProtein | [50] |
| OpenFold | https://github.com/aqlaboratory/openfold | [51] |
| RoseTTAFold | https://rosettta.bakerlab.org https://github.com/RosettaCommons/RoseTTAFold | [52] |
| Uni-Fold | https://github.com/dptech-corp/Uni-Fold | [53] |

Table S6. Tools for protein language model-based protein structure prediction.

| Methods | URLs | Reference |
|-------------------|---|-----------|
| EMBER3D | https://github.com/kWeissenow/EMBER3D | [54] |
| ESM-Fold | https://github.com/facebookresearch/esm | [55] |
| HelixFold-Single | https://github.com/PaddlePaddle/PaddleHelix/tree/dev/apps/protein_folding/helixfold-single | [56] |
| IgFold | https://github.com/Graylab/IgFold | [57] |
| OmegaFold | https://github.com/HeliXonProtein/OmegaFold | [58] |
| RaptorX-Single | https://github.com/AndersJing/RaptorX-Single | [59] |
| trRosettaX-Single | https://yanglab.qd.sdu.edu.cn/trRosetta/benchmark_single | [60] |

Table S7. Tools for multi-domain protein structure prediction.

| Methods | URLs | Reference |
|--------------|---|-----------|
| AIDA | http://ffas.burnham.org/AIDA/ | [61] |
| DeepAssembly | http://zhanglab-bioinf.com/DeepAssembly/ | [62] |
| DEMO | https://zhanggroup.org/DEMO/ | [63, 64] |
| D-I-TASSER | https://zhanggroup.org/D-I-TASSER/ | [40, 41] |
| E2EDA | http://zhanglab-bioinf.com/E2EDA/ | [65] |
| I-TASSER-MTD | https://zhanggroup.org/I-TASSER-MTD/ | [66] |
| LOMETS3 | https://zhanglab.dcmb.med.umich.edu/LOMETS/ | [12-14] |
| SADA | http://zhanglab-bioinf.com/SADA/ | [67] |

Table S8. The CASP14 benchmark dataset, including 65 full-length proteins and the corresponding 91 domains. The 91 domains can be divided as 54 TBM domains and 37 FM domains. The original CASP ‘TBM-easy’ and ‘TBM-hard’ domains are categorized as ‘TBM’ domains, while ‘FM/TBM’ and ‘FM’ domains are categorized as ‘FM’ domains in this analysis.

| Target | Type | IDs |
|-------------|------|---|
| Domain | TBM | T1024-D1,T1024-D2,T1025-D1,T1026-D1,T1028-D1,T1030-D1,T1030-D2,T1032-D1, T1034-D1,T1036s1-D1,T1045s1-D1,T1045s2-D1,T1046s2-D1,T1047s2-D2,T1050-D1, T1050-D2,T1050-D3,T1052-D1,T1052-D2,T1054-D1,T1056-D1,T1057-D1,T1058-D2, T1060s2-D1,T1060s3-D1,T1061-D3,T1065s1-D1,T1067-D1,T1068-D1,T1070-D2, T1070-D3,T1070-D4,T1073-D1,T1076-D1,T1078-D1,T1079-D1,T1083-D1,T1084-D1, T1087-D1,T1089-D1,T1091-D1,T1091-D2,T1091-D3,T1091-D4,T1092-D1,T1092-D2, T1093-D2,T1094-D1,T1095-D1,T1099-D1,T1100-D1,T1100-D2,T1101-D1,T1101-D2 |
| | FM | T1027-D1,T1029-D1,T1031-D1,T1033-D1,T1035-D1,T1037-D1,T1038-D1,T1038-D2, T1039-D1,T1040-D1,T1041-D1,T1042-D1,T1043-D1,T1046s1-D1,T1047s1-D1, T1047s2-D1,T1047s2-D3,T1049-D1,T1052-D3,T1053-D1,T1053-D2,T1055-D1, T1058-D1,T1061-D1,T1061-D2,T1064-D1,T1065s2-D1,T1070-D1,T1074-D1, T1080-D1,T1082-D1,T1090-D1,T1093-D1,T1093-D3,T1094-D2,T1096-D1,T1096-D2 |
| Full-length | | T1024,T1025,T1026,T1027,T1028,T1029,T1030,T1031,T1032,T1033,T1034,T1035, T1036s1,T1037,T1038,T1039,T1040,T1041,T1042,T1043,T1045s1,T1045s2,T1046s1, T1046s2,T1047s1,T1047s2,T1049,T1050,T1052,T1053,T1054,T1055,T1056,T1057, T1058,T1060s2,T1060s3,T1061,T1064,T1065s1,T1065s2,T1067,T1068,T1070,T1073, T1074,T1076,T1078,T1079,T1080,T1082,T1083,T1084,T1087,T1089,T1090,T1091, T1092,T1093,T1094,T1095,T1096,T1099,T1100,T1101 |

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