

# Ivan Anishchenko

## List of Publications by Year in Descending Order

**Source:** <https://exaly.com/author-pdf/8926204/ivan-anishchenko-publications-by-year.pdf>  
**Version:** 2024-04-09

This document has been generated based on the publications and citations recorded by exaly.com. For the latest version of this publication list, visit the link given above.  
The third column is the impact factor (IF) of the journal, and the fourth column is the number of citations of the article.

38 papers	1,871 citations	17 h-index	41 g-index
41 ext. papers	4,268 ext. citations	10.9 avg, IF	5.48 L-index

#	Paper	IF	Citations
38	Computed structures of core eukaryotic protein complexes. <i>Science</i> , <b>2021</b> , 374, eabm4805	33.3	51
37	The trRosetta server for fast and accurate protein structure prediction. <i>Nature Protocols</i> , <b>2021</b> , 16, 5634-5651	18.1	36
36	De novo protein design by deep network hallucination. <i>Nature</i> , <b>2021</b> ,	50.4	33
35	Protein sequence design by conformational landscape optimization. <i>Proceedings of the National Academy of Sciences of the United States of America</i> , <b>2021</b> , 118,	11.5	29
34	Improved protein structure refinement guided by deep learning based accuracy estimation. <i>Nature Communications</i> , <b>2021</b> , 12, 1340	17.4	50
33	In silico detection of SARS-CoV-2 specific B-cell epitopes and validation in ELISA for serological diagnosis of COVID-19. <i>Scientific Reports</i> , <b>2021</b> , 11, 4290	4.9	9
32	Accurate prediction of protein structures and interactions using a three-track neural network. <i>Science</i> , <b>2021</b> , 373, 871-876	33.3	522
31	Protein tertiary structure prediction and refinement using deep learning and Rosetta in CASP14. <i>Proteins: Structure, Function and Bioinformatics</i> , <b>2021</b> , 89, 1722-1733	4.2	9
30	Protein oligomer modeling guided by predicted interchain contacts in CASP14. <i>Proteins: Structure, Function and Bioinformatics</i> , <b>2021</b> , 89, 1824-1833	4.2	4
29	Deep learning enables the atomic structure determination of the Fanconi Anemia core complex from cryoEM. <i>IUCrJ</i> , <b>2020</b> , 7, 881-892	4.7	5
28	Improved protein structure prediction using predicted interresidue orientations. <i>Proceedings of the National Academy of Sciences of the United States of America</i> , <b>2020</b> , 117, 1496-1503	11.5	496
27	Structural basis for autophagy inhibition by the human Rubicon-Rab7 complex. <i>Proceedings of the National Academy of Sciences of the United States of America</i> , <b>2020</b> , 117, 17003-17010	11.5	9
26	Protein contact prediction using metagenome sequence data and residual neural networks. <i>Bioinformatics</i> , <b>2020</b> , 36, 41-48	7.2	43
25	High-accuracy refinement using Rosetta in CASP13. <i>Proteins: Structure, Function and Bioinformatics</i> , <b>2019</b> , 87, 1276-1282	4.2	26
24	Template-based modeling by ClusPro in CASP13 and the potential for using co-evolutionary information in docking. <i>Proteins: Structure, Function and Bioinformatics</i> , <b>2019</b> , 87, 1241-1248	4.2	7
23	Protein interaction networks revealed by proteome coevolution. <i>Science</i> , <b>2019</b> , 365, 185-189	33.3	112
22	Gene ontology improves template selection in comparative protein docking. <i>Proteins: Structure, Function and Bioinformatics</i> , <b>2019</b> , 87, 245-253	4.2	2

21	Modeling CAPRI targets 110-120 by template-based and free docking using contact potential and combined scoring function. <i>Proteins: Structure, Function and Bioinformatics</i> , <b>2018</b> , 86 Suppl 1, 302-310	4.2	13
20	Dockground: A comprehensive data resource for modeling of protein complexes. <i>Protein Science</i> , <b>2018</b> , 27, 172-181	6.3	32
19	Contact Potential for Structure Prediction of Proteins and Protein Complexes from Potts Model. <i>Biophysical Journal</i> , <b>2018</b> , 115, 809-821	2.9	12
18	Origins of coevolution between residues distant in protein 3D structures. <i>Proceedings of the National Academy of Sciences of the United States of America</i> , <b>2017</b> , 114, 9122-9127	11.5	92
17	Modeling complexes of modeled proteins. <i>Proteins: Structure, Function and Bioinformatics</i> , <b>2017</b> , 85, 470-478	4.2	18
16	Structural quality of unrefined models in protein docking. <i>Proteins: Structure, Function and Bioinformatics</i> , <b>2017</b> , 85, 39-45	4.2	4
15	Prediction of homoprotein and heteroprotein complexes by protein docking and template-based modeling: A CASP-CAPRI experiment. <i>Proteins: Structure, Function and Bioinformatics</i> , <b>2016</b> , 84 Suppl 1, 323-48	4.2	111
14	Protein models docking benchmark 2. <i>Proteins: Structure, Function and Bioinformatics</i> , <b>2015</b> , 83, 891-7	4.2	17
13	Structural templates for comparative protein docking. <i>Proteins: Structure, Function and Bioinformatics</i> , <b>2015</b> , 83, 1563-70	4.2	19
12	Protein models: the Grand Challenge of protein docking. <i>Proteins: Structure, Function and Bioinformatics</i> , <b>2014</b> , 82, 278-87	4.2	18
11	Structural basis for autophagy inhibition by the human Rubicon-Rab7 complex		1
10	Deep learning methods for designing proteins scaffolding functional sites		4
9	Structures of core eukaryotic protein complexes		7
8	Deep learning enables the atomic structure determination of the Fanconi Anemia core complex from cryoEM		1
7	In silico detection of SARS-CoV-2 specific B-cell epitopes and validation in ELISA for serological diagnosis of COVID-19		4
6	Improved protein structure refinement guided by deep learning based accuracy estimation		10
5	De novo protein design by deep network hallucination		26
4	Protein sequence design by explicit energy landscape optimization		4

3	Design of proteins presenting discontinuous functional sites using deep learning	8
2	Improved protein structure prediction using predicted inter-residue orientations	17
1	Accurate prediction of protein structures and interactions using a 3-track network	9