

S Joshua Swamidass

List of Publications by Year in descending order

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Version: 2024-02-01

71
papers

5,196
citations

147801

31
h-index

91884

69
g-index

102
all docs

102
docs citations

102
times ranked

7417
citing authors

#	ARTICLE	IF	CITATIONS
1	“Black Box” to “Conversational” Machine Learning: Ondansetron Reduces Risk of Hospital-Acquired Venous Thromboembolism. <i>IEEE Journal of Biomedical and Health Informatics</i> , 2021, 25, 2204-2214.	6.3	24
2	Modeling the Bioactivation and Subsequent Reactivity of Drugs. <i>Chemical Research in Toxicology</i> , 2021, 34, 584-600.	3.3	11
3	Deep learning the structural determinants of protein biochemical properties by comparing structural ensembles with DiffNets. <i>Nature Communications</i> , 2021, 12, 3023.	12.8	42
4	Impacts of diphenylamine NSAID halogenation on bioactivation risks. <i>Toxicology</i> , 2021, 458, 152832.	4.2	5
5	Machine learning liver-injuring drug interactions with non-steroidal anti-inflammatory drugs (NSAIDs) from a retrospective electronic health record (EHR) cohort. <i>PLoS Computational Biology</i> , 2021, 17, e1009053.	3.2	33
6	Deep Learning Coordinate-Free Quantum Chemistry. <i>Journal of Physical Chemistry A</i> , 2021, 125, 8978-8986.	2.5	5
7	Deep learning quantification of percent steatosis in donor liver biopsy frozen sections. <i>EBioMedicine</i> , 2020, 60, 103029.	6.1	32
8	Site-Level Bioactivity of Small-Molecules from Deep-Learned Representations of Quantum Chemistry. <i>Journal of Physical Chemistry A</i> , 2020, 124, 9194-9202.	2.5	7
9	Metabolic Forest: Predicting the Diverse Structures of Drug Metabolites. <i>Journal of Chemical Information and Modeling</i> , 2020, 60, 4702-4716.	5.4	11
10	XenoNet: Inference and Likelihood of Intermediate Metabolite Formation. <i>Journal of Chemical Information and Modeling</i> , 2020, 60, 3431-3449.	5.4	14
11	The Metabolic Rainbow: Deep Learning Phase I Metabolism in Five Colors. <i>Journal of Chemical Information and Modeling</i> , 2020, 60, 1146-1164.	5.4	26
12	Dual mechanisms suppress meloxicam bioactivation relative to sudoxicam. <i>Toxicology</i> , 2020, 440, 152478.	4.2	16
13	A Time-Embedding Network Models the Ontogeny of 23 Hepatic Drug Metabolizing Enzymes. <i>Chemical Research in Toxicology</i> , 2019, 32, 1707-1721.	3.3	6
14	Comprehensive kinetic and modeling analyses revealed CYP2C9 and 3A4 determine terbinafine metabolic clearance and bioactivation. <i>Biochemical Pharmacology</i> , 2019, 170, 113661.	4.4	13
15	CYP2C19 and 3A4 Dominate Metabolic Clearance and Bioactivation of Terbinafine Based on Computational and Experimental Approaches. <i>Chemical Research in Toxicology</i> , 2019, 32, 1151-1164.	3.3	12
16	Precision Medicine in Pancreatic Disease—Knowledge Gaps and Research Opportunities. <i>Pancreas</i> , 2019, 48, 1250-1258.	1.1	9
17	Accounting for proximal variants improves neoantigen prediction. <i>Nature Genetics</i> , 2019, 51, 175-179.	21.4	43
18	Standard operating procedure for somatic variant refinement of sequencing data with paired tumor and normal samples. <i>Genetics in Medicine</i> , 2019, 21, 972-981.	2.4	67

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19	Opportunities and obstacles for deep learning in biology and medicine. <i>Journal of the Royal Society Interface</i> , 2018, 15, 20170387.	3.4	1,282
20	Computationally Assessing the Bioactivation of Drugs by N-Dealkylation. <i>Chemical Research in Toxicology</i> , 2018, 31, 68-80.	3.3	30
21	Learning a Local-Variable Model of Aromatic and Conjugated Systems. <i>ACS Central Science</i> , 2018, 4, 52-62.	11.3	18
22	A deep learning approach to automate refinement of somatic variant calling from cancer sequencing data. <i>Nature Genetics</i> , 2018, 50, 1735-1743.	21.4	62
23	Lamisil (terbinafine) toxicity: Determining pathways to bioactivation through computational and experimental approaches. <i>Biochemical Pharmacology</i> , 2018, 156, 10-21.	4.4	17
24	Deep Learning Global Glomerulosclerosis in Transplant Kidney Frozen Sections. <i>IEEE Transactions on Medical Imaging</i> , 2018, 37, 2718-2728.	8.9	119
25	Modeling Small-Molecule Reactivity Identifies Promiscuous Bioactive Compounds. <i>Journal of Chemical Information and Modeling</i> , 2018, 58, 1483-1500.	5.4	28
26	Deep Learning to Predict the Formation of Quinone Species in Drug Metabolism. <i>Chemical Research in Toxicology</i> , 2017, 30, 642-656.	3.3	57
27	Computational Approach to Structural Alerts: Furans, Phenols, Nitroaromatics, and Thiophenes. <i>Chemical Research in Toxicology</i> , 2017, 30, 1046-1059.	3.3	32
28	Open Source Drug Discovery with the Malaria Box Compound Collection for Neglected Diseases and Beyond. <i>PLoS Pathogens</i> , 2016, 12, e1005763.	4.7	244
29	Inhibition of DNA Methyltransferases Blocks Mutant Huntingtin-Induced Neurotoxicity. <i>Scientific Reports</i> , 2016, 6, 31022.	3.3	28
30	Modeling Reactivity to Biological Macromolecules with a Deep Multitask Network. <i>ACS Central Science</i> , 2016, 2, 529-537.	11.3	76
31	Unsupervised detection of cancer driver mutations with parsimony-guided learning. <i>Nature Genetics</i> , 2016, 48, 1288-1294.	21.4	52
32	A simple model predicts UGT-mediated metabolism. <i>Bioinformatics</i> , 2016, 32, 3183-3189.	4.1	51
33	A survey of current trends in computational drug repositioning. <i>Briefings in Bioinformatics</i> , 2016, 17, 2-12.	6.5	459
34	Securely Measuring the Overlap between Private Datasets with Cryptosets. <i>PLoS ONE</i> , 2015, 10, e0117898.	2.5	6
35	Initiatives to bridge faith and science. <i>Nature</i> , 2015, 523, 531-531.	27.8	0
36	Statistically identifying tumor suppressors and oncogenes from pan-cancer genome-sequencing data. <i>Bioinformatics</i> , 2015, 31, 3561-3568.	4.1	55

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37	Modeling Epoxidation of Drug-like Molecules with a Deep Machine Learning Network. ACS Central Science, 2015, 1, 168-180.	11.3	130
38	Improved Prediction of CYP-Mediated Metabolism with Chemical Fingerprints. Journal of Chemical Information and Modeling, 2015, 55, 972-982.	5.4	13
39	Extending P450 site-of-metabolism models with region-resolution data. Bioinformatics, 2015, 31, 1966-1973.	4.1	15
40	Subcellular Localization and Ser-137 Phosphorylation Regulate Tumor-suppressive Activity of Profilin-1. Journal of Biological Chemistry, 2015, 290, 9075-9086.	3.4	23
41	Site of Reactivity Models Predict Molecular Reactivity of Diverse Chemicals with Glutathione. Chemical Research in Toxicology, 2015, 28, 797-809.	3.3	70
42	XenoSite server: a web-available site of metabolism prediction tool. Bioinformatics, 2015, 31, 1136-1137.	4.1	64
43	Sharing Chemical Relationships Does Not Reveal Structures. Journal of Chemical Information and Modeling, 2014, 54, 37-48.	5.4	7
44	Combined Analysis of Phenotypic and Target-Based Screening in Assay Networks. Journal of Biomolecular Screening, 2014, 19, 782-790.	2.6	23
45	Bigger data, collaborative tools and the future of predictive drug discovery. Journal of Computer-Aided Molecular Design, 2014, 28, 997-1008.	2.9	22
46	Managing missing measurements in small-molecule screens. Journal of Computer-Aided Molecular Design, 2013, 27, 469-478.	2.9	3
47	RS-WebPredictor: a server for predicting CYP-mediated sites of metabolism on drug-like molecules. Bioinformatics, 2013, 29, 497-498.	4.1	57
48	XenoSite: Accurately Predicting CYP-Mediated Sites of Metabolism with Neural Networks. Journal of Chemical Information and Modeling, 2013, 53, 3373-3383.	5.4	179
49	Accounting for noise when clustering biological data. Briefings in Bioinformatics, 2013, 14, 423-436.	6.5	24
50	Using economic optimization to design high-throughput screens. Future Medicinal Chemistry, 2013, 5, 9-11.	2.3	2
51	Scaffold network generator: a tool for mining molecular structures. Bioinformatics, 2013, 29, 2655-2656.	4.1	20
52	Automatically Detecting Workflows in PubChem. Journal of Biomolecular Screening, 2012, 17, 1071-1079.	2.6	9
53	Utility-Aware Screening with Clique-Oriented Prioritization. Journal of Chemical Information and Modeling, 2012, 52, 29-37.	5.4	7
54	Mining small-molecule screens to repurpose drugs. Briefings in Bioinformatics, 2011, 12, 327-335.	6.5	85

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55	Probabilistic Substructure Mining From Small-Molecule Screens. <i>Molecular Informatics</i> , 2011, 30, 809-815.	2.5	12
56	Enhancing the rate of scaffold discovery with diversity-oriented prioritization. <i>Bioinformatics</i> , 2011, 27, 2271-2278.	4.1	9
57	A CROC stronger than ROC: measuring, visualizing and optimizing early retrieval. <i>Bioinformatics</i> , 2010, 26, 1348-1356.	4.1	88
58	An Economic Framework to Prioritize Confirmatory Tests after a High-Throughput Screen. <i>Journal of Biomolecular Screening</i> , 2010, 15, 680-686.	2.6	14
59	Large scale study of multiple-molecule queries. <i>Journal of Cheminformatics</i> , 2009, 1, 7.	6.1	24
60	Influence Relevance Voting: An Accurate And Interpretable Virtual High Throughput Screening Method. <i>Journal of Chemical Information and Modeling</i> , 2009, 49, 756-766.	5.4	49
61	Discovery of Power-Laws in Chemical Space. <i>Journal of Chemical Information and Modeling</i> , 2008, 48, 1138-1151.	5.4	58
62	ChemDB update full-text search and virtual chemical space. <i>Bioinformatics</i> , 2007, 23, 2348-2351.	4.1	117
63	Lossless Compression of Chemical Fingerprints Using Integer Entropy Codes Improves Storage and Retrieval. <i>Journal of Chemical Information and Modeling</i> , 2007, 47, 2098-2109.	5.4	45
64	One- to Four-Dimensional Kernels for Virtual Screening and the Prediction of Physical, Chemical, and Biological Properties. <i>Journal of Chemical Information and Modeling</i> , 2007, 47, 965-974.	5.4	56
65	Mathematical Correction for Fingerprint Similarity Measures to Improve Chemical Retrieval. <i>Journal of Chemical Information and Modeling</i> , 2007, 47, 952-964.	5.4	61
66	Bounds and Algorithms for Fast Exact Searches of Chemical Fingerprints in Linear and Sublinear Time. <i>Journal of Chemical Information and Modeling</i> , 2007, 47, 302-317.	5.4	87
67	Functional Census of Mutation Sequence Spaces: The Example of p53 Cancer Rescue Mutants. <i>IEEE/ACM Transactions on Computational Biology and Bioinformatics</i> , 2006, 3, 114-125.	3.0	28
68	Structure-based inhibitor design of AccD5, an essential acyl-CoA carboxylase carboxyltransferase domain of <i>Mycobacterium tuberculosis</i> . <i>Proceedings of the National Academy of Sciences of the United States of America</i> , 2006, 103, 3072-3077.	7.1	112
69	Graph kernels for chemical informatics. <i>Neural Networks</i> , 2005, 18, 1093-1110.	5.9	329
70	Kernels for small molecules and the prediction of mutagenicity, toxicity and anti-cancer activity. <i>Bioinformatics</i> , 2005, 21, i359-i368.	4.1	145
71	ChemDB: a public database of small molecules and related cheminformatics resources. <i>Bioinformatics</i> , 2005, 21, 4133-4139.	4.1	155