

Ye Hu

List of Publications by Year in descending order

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Version: 2024-02-01

60
papers

1,805
citations

257450

24
h-index

276875

41
g-index

61
all docs

61
docs citations

61
times ranked

1892
citing authors

#	ARTICLE	IF	CITATIONS
1	SAR Matrix Method for Large-Scale Analysis of Compound Structure-Activity Relationships and Exploration of Multitarget Activity Spaces. <i>Methods in Molecular Biology</i> , 2018, 1825, 339-352.	0.9	2
2	Structure-Promiscuity Relationship Puzzles-Extensively Assayed Analogs with Large Differences in Target Annotations. <i>AAPS Journal</i> , 2017, 19, 856-864.	4.4	14
3	How Frequently Are Pan-Assay Interference Compounds Active? Large-Scale Analysis of Screening Data Reveals Diverse Activity Profiles, Low Global Hit Frequency, and Many Consistently Inactive Compounds. <i>Journal of Medicinal Chemistry</i> , 2017, 60, 3879-3886.	6.4	97
4	Entering the "big data" era in medicinal chemistry: molecular promiscuity analysis revisited. <i>Future Science OA</i> , 2017, 3, FSO179.	1.9	53
5	Mapping of inhibitors and activity data to the human kinome and exploring promiscuity from a ligand and target perspective. <i>Chemical Biology and Drug Design</i> , 2017, 89, 834-845.	3.2	16
6	Recent Advances in Scaffold Hopping. <i>Journal of Medicinal Chemistry</i> , 2017, 60, 1238-1246.	6.4	213
7	Determining the Degree of Promiscuity of Extensively Assayed Compounds. <i>PLoS ONE</i> , 2016, 11, e0153873.	2.5	43
8	Network Variants for Analyzing Target-Ligand Interactions. <i>ACS Symposium Series</i> , 2016, , 35-51.	0.5	1
9	Exploring Molecular Promiscuity from a Ligand and Target Perspective. <i>ACS Symposium Series</i> , 2016, , 19-34.	0.5	1
10	Analog series-based scaffolds: computational design and exploration of a new type of molecular scaffolds for medicinal chemistry. <i>Future Science OA</i> , 2016, 2, FSO149.	1.9	28
11	Assessing the Growth of Bioactive Compounds and Scaffolds over Time: Implications for Lead Discovery and Scaffold Hopping. <i>Journal of Chemical Information and Modeling</i> , 2016, 56, 300-307.	5.4	16
12	Computational Exploration of Molecular Scaffolds in Medicinal Chemistry. <i>Journal of Medicinal Chemistry</i> , 2016, 59, 4062-4076.	6.4	100
13	Activity-relevant similarity values for fingerprints and implications for similarity searching. <i>F1000Research</i> , 2016, 5, 591.	1.6	41
14	Analyzing compound activity records and promiscuity degrees in light of publication statistics. <i>F1000Research</i> , 2016, 5, 1227.	1.6	7
15	Analyzing compound activity records and promiscuity degrees in light of publication statistics. <i>F1000Research</i> , 2016, 5, 1227.	1.6	4
16	Identification of Interaction Hot Spots in Structures of Drug Targets on the Basis of Three-Dimensional Activity Cliff Information. <i>Chemical Biology and Drug Design</i> , 2015, 86, 1458-1465.	3.2	8
17	Structural and Activity Profile Relationships Between Drug Scaffolds. <i>AAPS Journal</i> , 2015, 17, 609-619.	4.4	6
18	Extension of three-dimensional activity cliff information through systematic mapping of active analogs. <i>RSC Advances</i> , 2015, 5, 43006-43015.	3.6	9

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19	Identification and analysis of the currently available high-confidence three-dimensional activity cliffs. <i>RSC Advances</i> , 2015, 5, 43660-43668.	3.6	11
20	Exploring the Scaffold Universe of Kinase Inhibitors. <i>Journal of Medicinal Chemistry</i> , 2015, 58, 315-332.	6.4	21
21	Current Compound Coverage of the Kinome. <i>Journal of Medicinal Chemistry</i> , 2015, 58, 30-40.	6.4	56
22	Promiscuity progression of bioactive compounds over time. <i>F1000Research</i> , 2015, 4, 118.	1.6	8
23	AnalogExplorer2 “ Stereochemistry sensitive graphical analysis of large analog series. <i>F1000Research</i> , 2015, 4, 1031.	1.6	3
24	Quantifying the Tendency of Therapeutic Target Proteins to Bind Promiscuous or Selective Compounds. <i>PLoS ONE</i> , 2015, 10, e0126838.	2.5	15
25	Influence of Search Parameters and Criteria on Compound Selection, Promiscuity, and Pan Assay Interference Characteristics. <i>Journal of Chemical Information and Modeling</i> , 2014, 54, 3056-3066.	5.4	32
26	Many drugs contain unique scaffolds with varying structural relationships to scaffolds of currently available bioactive compounds. <i>European Journal of Medicinal Chemistry</i> , 2014, 76, 427-434.	5.5	12
27	Learning from “big data”: compounds and targets. <i>Drug Discovery Today</i> , 2014, 19, 357-360.	6.4	45
28	EXPLORING COMPOUND PROMISCUITY PATTERNS AND MULTI-TARGET ACTIVITY SPACES. <i>Computational and Structural Biotechnology Journal</i> , 2014, 9, e201401003.	4.1	54
29	Many Approved Drugs Have Bioactive Analogs With Different Target Annotations. <i>AAPS Journal</i> , 2014, 16, 847-859.	4.4	13
30	Matched molecular pair-based data sets for computer-aided medicinal chemistry. <i>F1000Research</i> , 2014, 3, 36.	1.6	5
31	Compound data sets and software tools for chemoinformatics and medicinal chemistry applications: update and data transfer. <i>F1000Research</i> , 2014, 3, 69.	1.6	4
32	Monitoring drug promiscuity over time. <i>F1000Research</i> , 2014, 3, 218.	1.6	17
33	Compound promiscuity: what can we learn from current data?. <i>Drug Discovery Today</i> , 2013, 18, 644-650.	6.4	135
34	What is the Likelihood of an Active Compound to Be Promiscuous? Systematic Assessment of Compound Promiscuity on the Basis of PubChem Confirmatory Bioassay Data. <i>AAPS Journal</i> , 2013, 15, 808-815.	4.4	36
35	Activity cliffs in PubChem confirmatory bioassays taking inactive compounds into account. <i>Journal of Computer-Aided Molecular Design</i> , 2013, 27, 115-124.	2.9	11
36	How Promiscuous Are Pharmaceutically Relevant Compounds? A Data-Driven Assessment. <i>AAPS Journal</i> , 2013, 15, 104-111.	4.4	32

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37	Activity profile relationships between structurally similar promiscuous compounds. <i>European Journal of Medicinal Chemistry</i> , 2013, 69, 393-398.	5.5	10
38	Introduction of Target Cliffs as a Concept To Identify and Describe Complex Molecular Selectivity Patterns. <i>Journal of Chemical Information and Modeling</i> , 2013, 53, 545-552.	5.4	8
39	Systematic Identification of Scaffolds Representing Compounds Active against Individual Targets and Single or Multiple Target Families. <i>Journal of Chemical Information and Modeling</i> , 2013, 53, 312-326.	5.4	23
40	Visualization of Activity Landscapes and Chemogenomics Data. <i>Molecular Informatics</i> , 2013, 32, 954-963.	2.5	2
41	Advancing the activity cliff concept. <i>F1000Research</i> , 2013, 2, 199.	1.6	65
42	High-resolution view of compound promiscuity. <i>F1000Research</i> , 2013, 2, 144.	1.6	39
43	Many structurally related drugs bind different targets whereas distinct drugs display significant target overlap. <i>RSC Advances</i> , 2012, 2, 3481.	3.6	14
44	Systematic Identification and Classification of Three-Dimensional Activity Cliffs. <i>Journal of Chemical Information and Modeling</i> , 2012, 52, 1490-1498.	5.4	26
45	Extending the Activity Cliff Concept: Structural Categorization of Activity Cliffs and Systematic Identification of Different Types of Cliffs in the ChEMBL Database. <i>Journal of Chemical Information and Modeling</i> , 2012, 52, 1806-1811.	5.4	54
46	Exploration of 3D Activity Cliffs on the Basis of Compound Binding Modes and Comparison of 2D and 3D Cliffs. <i>Journal of Chemical Information and Modeling</i> , 2012, 52, 670-677.	5.4	23
47	Growth of Ligand-Target Interaction Data in ChEMBL Is Associated with Increasing and Activity Measurement-Dependent Compound Promiscuity. <i>Journal of Chemical Information and Modeling</i> , 2012, 52, 2550-2558.	5.4	37
48	Rationalizing Structure and Target Relationships between Current Drugs. <i>AAPS Journal</i> , 2012, 14, 764-771.	4.4	3
49	Chemical Transformations That Yield Compounds with Distinct Activity Profiles. <i>ACS Medicinal Chemistry Letters</i> , 2011, 2, 523-527.	2.8	18
50	Target Family-Directed Exploration of Scaffolds with Different SAR Profiles. <i>Journal of Chemical Information and Modeling</i> , 2011, 51, 3138-3148.	5.4	9
51	Lessons Learned from Molecular Scaffold Analysis. <i>Journal of Chemical Information and Modeling</i> , 2011, 51, 1742-1753.	5.4	82
52	Activity Profile Sequences: a Concept to Account for the Progression of Compound Activity in Target Space and to Extract SAR Information from Analogue Series with Multiple Target Annotations. <i>ChemMedChem</i> , 2011, 6, 2150-2154.	3.2	4
53	Exploring Target-Selectivity Patterns of Molecular Scaffolds. <i>ACS Medicinal Chemistry Letters</i> , 2010, 1, 54-58.	2.8	14
54	Scaffold Distributions in Bioactive Molecules, Clinical Trials Compounds, and Drugs. <i>ChemMedChem</i> , 2010, 5, 187-190.	3.2	22

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55	Molecular Scaffolds with High Propensity to Form Multi-Target Activity Cliffs. <i>Journal of Chemical Information and Modeling</i> , 2010, 50, 500-510.	5.4	45
56	Polypharmacology Directed Compound Data Mining: Identification of Promiscuous Chemotypes with Different Activity Profiles and Comparison to Approved Drugs. <i>Journal of Chemical Information and Modeling</i> , 2010, 50, 2112-2118.	5.4	56
57	Systematic Analysis of Public Domain Compound Potency Data Identifies Selective Molecular Scaffolds across Druggable Target Families. <i>Journal of Medicinal Chemistry</i> , 2010, 53, 752-758.	6.4	43
58	Global assessment of scaffold hopping potential for current pharmaceutical targets. <i>MedChemComm</i> , 2010, 1, 339-344.	3.4	25
59	Inside Cover: From Structure-Activity to Structure-Selectivity Relationships: Quantitative Assessment, Selectivity Cliffs, and Key Compounds (ChemMedChem 11/2009). <i>ChemMedChem</i> , 2009, 4, 1766-1766.	3.2	0
60	RelACCSâ€œFP: A Structural Minimalist Approach to Fingerprint Design. <i>Chemical Biology and Drug Design</i> , 2008, 72, 341-349.	3.2	13