

# Rajarshi Guha

## List of Publications by Year in descending order

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102  
papers

7,077  
citations

66343

42  
h-index

60623

81  
g-index

112  
all docs

112  
docs citations

112  
times ranked

11658  
citing authors

#	ARTICLE	IF	CITATIONS
1	The synergy of BET inhibitors with aurora A kinase inhibitors in MYCN-amplified neuroblastoma is heightened with functional TP53. <i>Neoplasia</i> , 2021, 23, 624-633.	5.3	8
2	Moving targets in drug discovery. <i>Scientific Reports</i> , 2020, 10, 20213.	3.3	23
3	Matrix Drug Screen Identifies Synergistic Drug Combinations to Augment SMAC Mimetic Activity in Ovarian Cancer. <i>Cancers</i> , 2020, 12, 3784.	3.7	3
4	Drugs Targeting Tumor-Initiating Cells Prolong Survival in a Post-Surgery, Post-Chemotherapy Ovarian Cancer Relapse Model. <i>Cancers</i> , 2020, 12, 1645.	3.7	25
5	High-throughput screening identifies candidate drugs for the treatment of recurrent respiratory papillomatosis. <i>Papillomavirus Research (Amsterdam, Netherlands)</i> , 2019, 8, 100181.	4.5	18
6	Mutation Profiles in Glioblastoma 3D Oncospheres Modulate Drug Efficacy. <i>SLAS Technology</i> , 2019, 24, 28-40.	1.9	14
7	High-Throughput Screening for Drug Combinations. <i>Methods in Molecular Biology</i> , 2019, 1939, 11-35.	0.9	10
8	High-throughput Chemical Screening Identifies Focal Adhesion Kinase and Aurora Kinase B Inhibition as a Synergistic Treatment Combination in Ewing Sarcoma. <i>Clinical Cancer Research</i> , 2019, 25, 4552-4566.	7.0	30
9	Target Deconvolution of a Multikinase Inhibitor with Antimetastatic Properties Identifies TAOK3 as a Key Contributor to a Cancer Stem Cell-Like Phenotype. <i>Molecular Cancer Therapeutics</i> , 2019, 18, 2097-2110.	4.1	16
10	A High-Throughput Screen of a Library of Therapeutics Identifies Cytotoxic Substrates of P-glycoprotein. <i>Molecular Pharmacology</i> , 2019, 96, 629-640.	2.3	22
11	Therapeutic strategies for diffuse midline glioma from high-throughput combination drug screening. <i>Science Translational Medicine</i> , 2019, 11, .	12.4	129
12	A Combination CDK4/6 and IGF1R Inhibitor Strategy for Ewing Sarcoma. <i>Clinical Cancer Research</i> , 2019, 25, 1343-1357.	7.0	69
13	Glypican-3-Specific Antibody Drug Conjugates Targeting Hepatocellular Carcinoma. <i>Hepatology</i> , 2019, 70, 563-576.	7.3	65
14	Unexplored therapeutic opportunities in the human genome. <i>Nature Reviews Drug Discovery</i> , 2018, 17, 317-332.	46.4	263
15	Characterization of clinically used oral antiseptics as quadruplex-binding ligands. <i>Nucleic Acids Research</i> , 2018, 46, 2722-2732.	14.5	27
16	High-throughput screens identify HSP90 inhibitors as potent therapeutics that target inter-related growth and survival pathways in advanced prostate cancer. <i>Scientific Reports</i> , 2018, 8, 17239.	3.3	29
17	Canvass: A Crowd-Sourced, Natural-Product Screening Library for Exploring Biological Space. <i>ACS Central Science</i> , 2018, 4, 1727-1741.	11.3	32
18	Resistance to Epigenetic-Targeted Therapy Engenders Tumor Cell Vulnerabilities Associated with Enhancer Remodeling. <i>Cancer Cell</i> , 2018, 34, 922-938.e7.	16.8	63

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19	Using Machine Learning to Predict Synergistic Antimalarial Compound Combinations With Novel Structures. <i>Frontiers in Pharmacology</i> , 2018, 9, 1096.	3.5	27
20	A single nucleotide polymorphism in the <i>Plasmodium falciparum</i> atg18 gene associates with artemisinin resistance and confers enhanced parasite survival under nutrient deprivation. <i>Malaria Journal</i> , 2018, 17, 391.	2.3	30
21	Identification of Combinations of Approved Drugs With Synergistic Activity Against Ebola Virus in Cell Cultures. <i>Journal of Infectious Diseases</i> , 2018, 218, S672-S678.	4.0	49
22	A systematic and prospectively validated approach for identifying synergistic drug combinations against malaria. <i>Malaria Journal</i> , 2018, 17, 160.	2.3	19
23	Pharmacological and genomic profiling of neurofibromatosis type 1 plexiform neurofibroma-derived schwann cells. <i>Scientific Data</i> , 2018, 5, 180106.	5.3	20
24	Pharos: Collating protein information to shed light on the druggable genome. <i>Nucleic Acids Research</i> , 2017, 45, D995-D1002.	14.5	271
25	A High-Throughput Screening Model of the Tumor Microenvironment for Ovarian Cancer Cell Growth. <i>SLAS Discovery</i> , 2017, 22, 494-506.	2.7	26
26	Chemotaxis of Molecular Dyes in Polymer Gradients in Solution. <i>Journal of the American Chemical Society</i> , 2017, 139, 15588-15591.	13.7	28
27	Matrix Screen Identifies Synergistic Combination of PARP Inhibitors and Nicotinamide Phosphoribosyltransferase (NAMPT) Inhibitors in Ewing Sarcoma. <i>Clinical Cancer Research</i> , 2017, 23, 7301-7311.	7.0	44
28	The Chemistry Development Kit (CDK) v2.0: atom typing, depiction, molecular formulas, and substructure searching. <i>Journal of Cheminformatics</i> , 2017, 9, 33.	6.1	275
29	The phosphatidylinositol-3-phosphate 5-kinase inhibitor apilimod blocks filoviral entry and infection. <i>PLoS Neglected Tropical Diseases</i> , 2017, 11, e0005540.	3.0	97
30	Drug target ontology to classify and integrate drug discovery data. <i>Journal of Biomedical Semantics</i> , 2017, 8, 50.	1.6	63
31	Evaluation of the Activity of Lamivudine and Zidovudine against Ebola Virus. <i>PLoS ONE</i> , 2016, 11, e0166318.	2.5	28
32	Large-scale pharmacological profiling of 3D tumor models of cancer cells. <i>Cell Death and Disease</i> , 2016, 7, e2492-e2492.	6.3	26
33	Modelling of compound combination effects and applications to efficacy and toxicity: state-of-the-art, challenges and perspectives. <i>Drug Discovery Today</i> , 2016, 21, 225-238.	6.4	162
34	Ranking Differential Drug Activities from Dose-Response Synthetic Lethality Screens. <i>Journal of Biomolecular Screening</i> , 2016, 21, 942-955.	2.6	4
35	A Druggable TCF4- and BRD4-Dependent Transcriptional Network Sustains Malignancy in Blastic Plasmacytoid Dendritic Cell Neoplasm. <i>Cancer Cell</i> , 2016, 30, 764-778.	16.8	116
36	mQC: A Heuristic Quality-Control Metric for High-Throughput Drug Combination Screening. <i>Scientific Reports</i> , 2016, 6, 37741.	3.3	8

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37	Augmented efficacy of brentuximab vedotin combined with ruxolitinib and/or Navitoclax in a murine model of human Hodgkin's lymphoma. Proceedings of the National Academy of Sciences of the United States of America, 2016, 113, 1624-1629.	7.1	38
38	Chemical Screens Identify Drugs that Enhance or Mitigate Cellular Responses to Antibody-Toxin Fusion Proteins. PLoS ONE, 2016, 11, e0161415.	2.5	8
39	High-throughput matrix screening identifies synergistic and antagonistic antimalarial drug combinations. Scientific Reports, 2015, 5, 13891.	3.3	92
40	Selective targeting of JAK/STAT signaling is potentiated by Bcl-xL blockade in IL-2-dependent adult T-cell leukemia. Retrovirology, 2015, 12, .	2.0	3
41	Novel Phenotypic Outcomes Identified for a Public Collection of Approved Drugs from a Publicly Accessible Panel of Assays. PLoS ONE, 2015, 10, e0130796.	2.5	18
42	Synergy Maps: exploring compound combinations using network-based visualization. Journal of Cheminformatics, 2015, 7, 36.	6.1	32
43	Chemogenomic Profiling of Endogenous <i>PARK2</i> Expression Using a Genome-Edited Coincidence Reporter. ACS Chemical Biology, 2015, 10, 1188-1197.	3.4	52
44	Whole-genome RNAi screen highlights components of the endoplasmic reticulum/Golgi as a source of resistance to immunotoxin-mediated cytotoxicity. Proceedings of the National Academy of Sciences of the United States of America, 2015, 112, E1135-42.	7.1	22
45	Selective targeting of JAK/STAT signaling is potentiated by Bcl-xL blockade in IL-2-dependent adult T-cell leukemia. Proceedings of the National Academy of Sciences of the United States of America, 2015, 112, 12480-12485.	7.1	81
46	Actinoramide A Identified as a Potent Antimalarial from Titration-Based Screening of Marine Natural Product Extracts. Journal of Natural Products, 2015, 78, 2411-2422.	3.0	30
47	Aurora B kinase is a potent and selective target in MYCN-driven neuroblastoma. Oncotarget, 2015, 6, 35247-35262.	1.8	52
48	Genome Editing-Enabled HTS Assays Expand Drug Target Pathways for Charcot-Marie-Tooth Disease. ACS Chemical Biology, 2014, 9, 2594-2602.	3.4	31
49	High-throughput combinatorial screening identifies drugs that cooperate with ibrutinib to kill activated B-cell-like diffuse large B-cell lymphoma cells. Proceedings of the National Academy of Sciences of the United States of America, 2014, 111, 2349-2354.	7.1	355
50	Blockade of oncogenic Î²B kinase activity in diffuse large B-cell lymphoma by bromodomain and extraterminal domain protein inhibitors. Proceedings of the National Academy of Sciences of the United States of America, 2014, 111, 11365-11370.	7.1	166
51	On the validity versus utility of activity landscapes: are all activity cliffs statistically significant?. Journal of Cheminformatics, 2014, 6, 11.	6.1	9
52	A High-Throughput Assay for Small Molecule Destabilizers of the KRAS Oncoprotein. PLoS ONE, 2014, 9, e103836.	2.5	18
53	High-content genome-wide RNAi screens identify regulators of parkin upstream of mitophagy. Nature, 2013, 504, 291-295.	27.8	301
54	Large-scale screening identifies a novel microRNA, miR-15a-3p, which induces apoptosis in human cancer cell lines. RNA Biology, 2013, 10, 287-300.	3.1	41

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55	Profile of the GSK Published Protein Kinase Inhibitor Set Across ATP-Dependent and-Independent Luciferases: Implications for Reporter-Gene Assays. <i>PLoS ONE</i> , 2013, 8, e57888.	2.5	65
56	On Exploring Structure–Activity Relationships. <i>Methods in Molecular Biology</i> , 2013, 993, 81-94.	0.9	102
57	A Survey of Quantitative Descriptions of Molecular Structure. <i>Current Topics in Medicinal Chemistry</i> , 2013, 12, 1946-1956.	2.1	6
58	A 1536-Well Quantitative High-Throughput Screen to Identify Compounds Targeting Cancer Stem Cells. <i>Journal of Biomolecular Screening</i> , 2012, 17, 1231-1242.	2.6	35
59	A Survey of Quantitative Descriptions of Molecular Structure. <i>Current Topics in Medicinal Chemistry</i> , 2012, 12, 1946-1956.	2.1	48
60	Exploring Uncharted Territories: Predicting Activity Cliffs in Structure–Activity Landscapes. <i>Journal of Chemical Information and Modeling</i> , 2012, 52, 2181-2191.	5.4	33
61	High-throughput screening for genes that prevent excess DNA replication in human cells and for molecules that inhibit them. <i>Methods</i> , 2012, 57, 234-248.	3.8	11
62	A furoxan–amodiaquine hybrid as a potential therapeutic for three parasitic diseases. <i>MedChemComm</i> , 2012, 3, 1505.	3.4	21
63	Exploring structure–activity data using the landscape paradigm. <i>Wiley Interdisciplinary Reviews: Computational Molecular Science</i> , 2012, 2, 829-841.	14.6	34
64	Exploiting Synthetic Lethality for the Therapy of ABC Diffuse Large B Cell Lymphoma. <i>Cancer Cell</i> , 2012, 21, 723-737.	16.8	460
65	Chemical Genomic Profiling for Antimalarial Therapies, Response Signatures, and Molecular Targets. <i>Science</i> , 2011, 333, 724-729.	12.6	130
66	Using a neural network for mining interpretable relationships of West Nile risk factors. <i>Social Science and Medicine</i> , 2011, 72, 418-429.	3.8	13
67	Exploratory analysis of kinetic solubility measurements of a small molecule library. <i>Bioorganic and Medicinal Chemistry</i> , 2011, 19, 4127-4134.	3.0	27
68	Open Data, Open Source and Open Standards in chemistry: The Blue Obelisk five years on. <i>Journal of Cheminformatics</i> , 2011, 3, 37.	6.1	63
69	KNIME Workflow to Assess PAINS Filters in SMARTS Format. Comparison of RDKit and Indigo Cheminformatics Libraries. <i>Molecular Informatics</i> , 2011, 30, 847-850.	2.5	118
70	Discovery of new antimalarial chemotypes through chemical methodology and library development. <i>Proceedings of the National Academy of Sciences of the United States of America</i> , 2011, 108, 6775-6780.	7.1	42
71	The Ups and Downs of Structure–Activity Landscapes. <i>Methods in Molecular Biology</i> , 2010, 672, 101-117.	0.9	28
72	Use of genetic algorithm and neural network approaches for risk factor selection: A case study of West Nile virus dynamics in an urban environment. <i>Computers, Environment and Urban Systems</i> , 2010, 34, 189-203.	7.1	14

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73	Towards interoperable and reproducible QSAR analyses: Exchange of datasets. <i>Journal of Cheminformatics</i> , 2010, 2, 5.	6.1	39
74	Advances in Cheminformatics Methodologies and Infrastructure to Support the Data Mining of Large, Heterogeneous Chemical Datasets. <i>Current Computer-Aided Drug Design</i> , 2010, 6, 50-67.	1.2	28
75	A Risk Factor Analysis of West Nile Virus: Extraction of Relationships from a Neural-Network Model. <i>Lecture Notes in Computer Science</i> , 2010, , 208-217.	1.3	1
76	Navigating structure-activity landscapes. <i>Drug Discovery Today</i> , 2009, 14, 698-705.	6.4	161
77	Chemoinformatic Analysis of Combinatorial Libraries, Drugs, Natural Products, and Molecular Libraries Small Molecule Repository. <i>Journal of Chemical Information and Modeling</i> , 2009, 49, 1010-1024.	5.4	138
78	Utilizing high throughput screening data for predictive toxicology models: protocols and application to MLSCN assays. <i>Journal of Computer-Aided Molecular Design</i> , 2008, 22, 367-384.	2.9	43
79	On the interpretation and interpretability of quantitative structure-activity relationship models. <i>Journal of Computer-Aided Molecular Design</i> , 2008, 22, 857-871.	2.9	70
80	Structure-Activity Landscape Index: Identifying and Quantifying Activity Cliffs. <i>Journal of Chemical Information and Modeling</i> , 2008, 48, 646-658.	5.4	281
81	Flexible Web Service Infrastructure for the Development and Deployment of Predictive Models. <i>Journal of Chemical Information and Modeling</i> , 2008, 48, 456-464.	5.4	22
82	Assessing How Well a Modeling Protocol Captures a Structure-Activity Landscape. <i>Journal of Chemical Information and Modeling</i> , 2008, 48, 1716-1728.	5.4	84
83	Chemical Data Mining of the NCI Human Tumor Cell Line Database. <i>Journal of Chemical Information and Modeling</i> , 2007, 47, 2063-2076.	5.4	30
84	Ensemble Feature Selection: Consistent Descriptor Subsets for Multiple QSAR Models. <i>Journal of Chemical Information and Modeling</i> , 2007, 47, 989-997.	5.4	51
85	Counting Clusters Using R-NN Curves. <i>Journal of Chemical Information and Modeling</i> , 2007, 47, 1308-1318.	5.4	3
86	Web Service Infrastructure for Cheminformatics. <i>Journal of Chemical Information and Modeling</i> , 2007, 47, 1303-1307.	5.4	41
87	Userscripts for the Life Sciences. <i>BMC Bioinformatics</i> , 2007, 8, 487.	2.6	14
88	Local Lazy Regression: Making Use of the Neighborhood to Improve QSAR Predictions. <i>Journal of Chemical Information and Modeling</i> , 2006, 46, 1836-1847.	5.4	65
89	The Blue Obelisk: Interoperability in Chemical Informatics. <i>Journal of Chemical Information and Modeling</i> , 2006, 46, 991-998.	5.4	366
90	Scalable Partitioning and Exploration of Chemical Spaces Using Geometric Hashing. <i>Journal of Chemical Information and Modeling</i> , 2006, 46, 321-333.	5.4	22

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91	R-NN Curves: An Intuitive Approach to Outlier Detection Using a Distance Based Method. Journal of Chemical Information and Modeling, 2006, 46, 1713-1722.	5.4	20
92	Recent Developments of the Chemistry Development Kit (CDK) - An Open-Source Java Library for Chemo- and Bioinformatics. Current Pharmaceutical Design, 2006, 12, 2111-2120.	1.9	418
93	Development of Linear, Ensemble, and Nonlinear Models for the Prediction and Interpretation of the Biological Activity of a Set of PDGFR Inhibitors.. ChemInform, 2005, 36, no.	0.0	0
94	Determining the Validity of a QSAR Model ? A Classification Approach.. ChemInform, 2005, 36, no.	0.0	0
95	Interpreting Computational Neural Network Quantitative Structure-Activity Relationship Models: A Detailed Interpretation of the Weights and Biases.. ChemInform, 2005, 36, no.	0.0	0
96	Interpreting Computational Neural Network Quantitative Structure-Activity Relationship Models: A Detailed Interpretation of the Weights and Biases. Journal of Chemical Information and Modeling, 2005, 45, 1109-1121.	5.4	63
97	Determining the Validity of a QSAR Model - A Classification Approach. Journal of Chemical Information and Modeling, 2005, 45, 65-73.	5.4	88
98	Interpreting Computational Neural Network QSAR Models: A Measure of Descriptor Importance. Journal of Chemical Information and Modeling, 2005, 45, 800-806.	5.4	94
99	Development of QSAR Models to Predict and Interpret the Biological Activity of Artemisinin Analogues.. ChemInform, 2004, 35, no.	0.0	0
100	Generation of QSAR sets with a self-organizing map. Journal of Molecular Graphics and Modelling, 2004, 23, 1-14.	2.4	52
101	Development of Linear, Ensemble, and Nonlinear Models for the Prediction and Interpretation of the Biological Activity of a Set of PDGFR Inhibitors. Journal of Chemical Information and Computer Sciences, 2004, 44, 2179-2189.	2.8	78
102	Development of QSAR Models To Predict and Interpret the Biological Activity of Artemisinin Analogues. Journal of Chemical Information and Computer Sciences, 2004, 44, 1440-1449.	2.8	60