

Alexander Sedykh

List of Publications by Year in descending order

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Version: 2024-02-01

40
papers

2,332
citations

218677

26
h-index

315739

38
g-index

40
all docs

40
docs citations

40
times ranked

2752
citing authors

#	ARTICLE	IF	CITATIONS
1	Quantitative Structure–Activity Relationship Modeling of Rat Acute Toxicity by Oral Exposure. <i>Chemical Research in Toxicology</i> , 2009, 22, 1913-1921.	3.3	210
2	Predicting Drug-Induced Hepatotoxicity Using QSAR and Toxicogenomics Approaches. <i>Chemical Research in Toxicology</i> , 2011, 24, 1251-1262.	3.3	190
3	Big Data in Chemical Toxicity Research: The Use of High-Throughput Screening Assays To Identify Potential Toxicants. <i>Chemical Research in Toxicology</i> , 2014, 27, 1643-1651.	3.3	119
4	Toxicokinetic Triage for Environmental Chemicals. <i>Toxicological Sciences</i> , 2015, 147, 55-67.	3.1	117
5	Integrative Chemical–Biological Read-Across Approach for Chemical Hazard Classification. <i>Chemical Research in Toxicology</i> , 2013, 26, 1199-1208.	3.3	107
6	Use of <i>in Vitro</i> HTS-Derived Concentration–Response Data as Biological Descriptors Improves the Accuracy of QSAR Models of <i>in Vivo</i> Toxicity. <i>Environmental Health Perspectives</i> , 2011, 119, 364-370.	6.0	103
7	Alarms about structural alerts. <i>Green Chemistry</i> , 2016, 18, 4348-4360.	9.0	103
8	An updated review on drug-induced cholestasis: Mechanisms and investigation of physicochemical properties and pharmacokinetic parameters. <i>Journal of Pharmaceutical Sciences</i> , 2013, 102, 3037-3057.	3.3	95
9	Predicting Nano–Bio Interactions by Integrating Nanoparticle Libraries and Quantitative Nanostructure Activity Relationship Modeling. <i>ACS Nano</i> , 2017, 11, 12641-12649.	14.6	80
10	Identification of putative estrogen receptor-mediated endocrine disrupting chemicals using QSAR- and structure-based virtual screening approaches. <i>Toxicology and Applied Pharmacology</i> , 2013, 272, 67-76.	2.8	78
11	Discovery of Novel Antimalarial Compounds Enabled by QSAR-Based Virtual Screening. <i>Journal of Chemical Information and Modeling</i> , 2013, 53, 475-492.	5.4	77
12	Construction of a web-based nanomaterial database by big data curation and modeling friendly nanostructure annotations. <i>Nature Communications</i> , 2020, 11, 2519.	12.8	77
13	Human Intestinal Transporter Database: QSAR Modeling and Virtual Profiling of Drug Uptake, Efflux and Interactions. <i>Pharmaceutical Research</i> , 2013, 30, 996-1007.	3.5	76
14	Critical Evaluation of Human Oral Bioavailability for Pharmaceutical Drugs by Using Various Cheminformatics Approaches. <i>Pharmaceutical Research</i> , 2014, 31, 1002-1014.	3.5	76
15	A Data Analysis Pipeline Accounting for Artifacts in Tox21 Quantitative High-Throughput Screening Assays. <i>Journal of Biomolecular Screening</i> , 2015, 20, 887-897.	2.6	75
16	Developing Enhanced Blood–Brain Barrier Permeability Models: Integrating External Bio-Assay Data in QSAR Modeling. <i>Pharmaceutical Research</i> , 2015, 32, 3055-3065.	3.5	70
17	Predictive Modeling of Chemical Hazard by Integrating Numerical Descriptors of Chemical Structures and Short-term Toxicity Assay Data. <i>Toxicological Sciences</i> , 2012, 127, 1-9.	3.1	64
18	<i>In silico</i> profiling nanoparticles: predictive nanomodeling using universal nanodescriptors and various machine learning approaches. <i>Nanoscale</i> , 2019, 11, 8352-8362.	5.6	64

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19	Mechanism Profiling of Hepatotoxicity Caused by Oxidative Stress Using Antioxidant Response Element Reporter Gene Assay Models and Big Data. <i>Environmental Health Perspectives</i> , 2016, 124, 634-641.	6.0	56
20	Conditional Toxicity Value (CTV) Predictor: An <i>In Silico</i> Approach for Generating Quantitative Risk Estimates for Chemicals. <i>Environmental Health Perspectives</i> , 2018, 126, 057008.	6.0	52
21	Quantitative High-Throughput Screening for Chemical Toxicity in a Population-Based In Vitro Model. <i>Toxicological Sciences</i> , 2012, 126, 578-588.	3.1	47
22	Experimental Errors in QSAR Modeling Sets: What We Can Do and What We Cannot Do. <i>ACS Omega</i> , 2017, 2, 2805-2812.	3.5	47
23	The Use of Pseudo-Equilibrium Constant Affords Improved QSAR Models of Human Plasma Protein Binding. <i>Pharmaceutical Research</i> , 2013, 30, 1790-1798.	3.5	43
24	Predicting Chemical Ocular Toxicity Using a Combinatorial QSAR Approach. <i>Chemical Research in Toxicology</i> , 2012, 25, 2763-2769.	3.3	42
25	Hybrid <i>in silico</i> models for drug-induced liver injury using chemical descriptors and <i>in vitro</i> cell-imaging information. <i>Journal of Applied Toxicology</i> , 2014, 34, 281-288.	2.8	41
26	Real-time cell toxicity profiling of Tox21 10K compounds reveals cytotoxicity dependent toxicity pathway linkage. <i>PLoS ONE</i> , 2017, 12, e0177902.	2.5	40
27	Identifying Compounds with Genotoxicity Potential Using Tox21 High-Throughput Screening Assays. <i>Chemical Research in Toxicology</i> , 2019, 32, 1384-1401.	3.3	27
28	Design, synthesis and experimental validation of novel potential chemopreventive agents using random forest and support vector machine binary classifiers. <i>Journal of Computer-Aided Molecular Design</i> , 2014, 28, 631-646.	2.9	25
29	Analysis of model PM2.5-induced inflammation and cytotoxicity by the combination of a virtual carbon nanoparticle library and computational modeling. <i>Ecotoxicology and Environmental Safety</i> , 2020, 191, 110216.	6.0	20
30	Application of Benchmark Concentration (BMC) Analysis on Zebrafish Data: A New Perspective for Quantifying Toxicity in Alternative Animal Models. <i>Toxicological Sciences</i> , 2019, 167, 92-104.	3.1	19
31	Using Tox21 High-Throughput Screening Assays for the Evaluation of Botanical and Dietary Supplements. <i>Applied in Vitro Toxicology</i> , 2019, 5, 10-25.	1.1	15
32	Short Communication: Cheminformatics Analysis to Identify Predictors of Antiviral Drug Penetration into the Female Genital Tract. <i>AIDS Research and Human Retroviruses</i> , 2014, 30, 1058-1064.	1.1	14
33	Universal nanohydrophobicity predictions using virtual nanoparticle library. <i>Journal of Cheminformatics</i> , 2019, 11, 6.	6.1	14
34	Integrative Approaches for Predicting In Vivo Effects of Chemicals from their Structural Descriptors and the Results of Short-Term Biological Assays. <i>Current Topics in Medicinal Chemistry</i> , 2014, 14, 1356-1364.	2.1	14
35	Modelability Criteria: Statistical Characteristics Estimating Feasibility to Build Predictive QSAR Models for a Dataset. , 2014, , 187-230.		11
36	Harnessing In Silico, In Vitro, and In Vivo Data to Understand the Toxicity Landscape of Polycyclic Aromatic Compounds (PACs). <i>Chemical Research in Toxicology</i> , 2021, 34, 268-285.	3.3	9

#	ARTICLE	IF	CITATIONS
37	Chemistry-Wide Association Studies (CWAS): A Novel Framework for Identifying and Interpreting Structure-Activity Relationships. <i>Journal of Chemical Information and Modeling</i> , 2018, 58, 2203-2213.	5.4	7
38	CurveP Method for Rendering High-Throughput Screening Dose-Response Data into Digital Fingerprints. <i>Methods in Molecular Biology</i> , 2016, 1473, 135-141.	0.9	5
39	Predicting Adverse Drug Effects from Literature- and Database-Mined Assertions. <i>Drug Safety</i> , 2018, 41, 1059-1072.	3.2	3
40	CurveP Method for Rendering High-Throughput Screening Dose-Response Data into Digital Fingerprints. <i>Methods in Molecular Biology</i> , 2022, 2474, 147-154.	0.9	0