## Steffen Klamt

List of Publications by Year in descending order

Source: https://exaly.com/author-pdf/8408588/publications.pdf

Version: 2024-02-01

61984 53230 8,187 107 43 citations h-index papers

g-index 113 113 113 5964 docs citations times ranked citing authors all docs

85

#	Article	IF	CITATIONS
1	Cellâ€Free Multiâ€Enzyme Synthesis and Purification of Uridine Diphosphate Galactose. ChemBioChem, 2022, 23, .	2.6	7
2	Maximizing batch fermentation efficiency by constrained modelâ€based optimization and predictive control of adenosine triphosphate turnover. AICHE Journal, 2022, 68, .	3.6	5
3	CNApy: a CellNetAnalyzer GUI in Python for analyzing and designing metabolic networks. Bioinformatics, 2022, 38, 1467-1469.	4.1	8
4	Enabling anaerobic growth of Escherichia coli on glycerol in defined minimal medium using acetate as redox sink. Metabolic Engineering, 2022, 73, 50-57.	7.0	5
5	Analyzing and Resolving Infeasibility in Flux Balance Analysis of Metabolic Networks. Metabolites, 2022, 12, 585.	2.9	2
6	Increasing ATP turnover boosts productivity ofÂ2,3-butanediol synthesis inÂEscherichia coli. Microbial Cell Factories, 2021, 20, 63.	4.0	14
7	Blending industrial blast furnace gas with H2 enables Acetobacterium woodii to efficiently co-utilize CO, CO2 and H2. Bioresource Technology, 2021, 323, 124573.	9.6	23
8	Designing microbial communities to maximize the thermodynamic driving force for the production of chemicals. PLoS Computational Biology, 2021, 17, e1009093.	3.2	12
9	Systematizing the different notions of growthâ€coupled product synthesis and a single framework for computing corresponding strain designs. Biotechnology Journal, 2021, 16, e2100236.	3.5	10
10	Deciphering the physiological response of <i>Escherichia coli</i> under high ATP demand. Molecular Systems Biology, 2021, 17, e10504.	7.2	10
11	Speeding up the core algorithm for the dual calculation of minimal cut sets in large metabolic networks. BMC Bioinformatics, 2020, 21, 510.	2.6	8
12	ATPase-based implementation of enforced ATP wasting in Saccharomyces cerevisiae for improved ethanol production. Biotechnology for Biofuels, 2020, 13, 185.	6.2	17
13	An extended and generalized framework for the calculation of metabolic intervention strategies based on minimal cut sets. PLoS Computational Biology, 2020, 16, e1008110.	3.2	24
14	MEMOTE for standardized genome-scale metabolic model testing. Nature Biotechnology, 2020, 38, 272-276.	17.5	314
15	MEMO: A Method for Computing Metabolic Modules for Cell-Free Production Systems. ACS Synthetic Biology, 2020, 9, 556-566.	3.8	8
16	Automatic construction of metabolic models with enzyme constraints. BMC Bioinformatics, 2020, 21, 19.	2.6	69
17	<scp>SBML</scp> Level 3: an extensible format for the exchange and reuse of biological models. Molecular Systems Biology, 2020, 16, e9110.	7.2	178
18	Title is missing!. , 2020, 16, e1008110.		0

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19	Title is missing!. , 2020, 16, e1008110.		O
20	Title is missing!. , 2020, 16, e1008110.		0
21	Title is missing!. , 2020, 16, e1008110.		0
22	Augmenting Biogas Process Modeling by Resolving Intracellular Metabolic Activity. Frontiers in Microbiology, 2019, 10, 1095.	3.5	14
23	Broadening the Scope of Enforced ATP Wasting as a Tool for Metabolic Engineering in <i>Escherichia coli</i> li>. Biotechnology Journal, 2019, 14, e1800438.	3.5	32
24	RedCom: A strategy for reduced metabolic modeling of complex microbial communities and its application for analyzing experimental datasets from anaerobic digestion. PLoS Computational Biology, 2019, 15, e1006759.	3.2	29
25	Characterizing and ranking computed metabolic engineering strategies. Bioinformatics, 2019, 35, 3063-3072.	4.1	17
26	Designing Optimal Experiments to Discriminate Interaction Graph Models. IEEE/ACM Transactions on Computational Biology and Bioinformatics, 2019, 16, 925-935.	3.0	1
27	A mathematical framework for yield (vs. rate) optimization in constraint-based modeling and applications in metabolic engineering. Metabolic Engineering, 2018, 47, 153-169.	7.0	37
28	Temperatureâ€dependent dynamic control of the TCA cycle increases volumetric productivity of itaconic acid production by <i>Escherichia coli</i> . Biotechnology and Bioengineering, 2018, 115, 156-164.	3.3	97
29	When Do Twoâ€Stage Processes Outperform Oneâ€Stage Processes?. Biotechnology Journal, 2018, 13, 1700539.	3.5	52
30	MoVE identifies metabolic valves to switch between phenotypic states. Nature Communications, 2018, 9, 5332.	12.8	37
31	OptMDFpathway: Identification of metabolic pathways with maximal thermodynamic driving force and its application for analyzing the endogenous CO2 fixation potential of Escherichia coli. PLoS Computational Biology, 2018, 14, e1006492.	3.2	45
32	One-Stage vs. Two-Stage Bioprocesses: Comparative Computational and Experimental Studies and Consequences for Strain Design. IFAC-PapersOnLine, 2018, 51, 7.	0.9	4
33	Generalizing Diffuse Interface Methods on Graphs: Nonsmooth Potentials and Hypergraphs. SIAM Journal on Applied Mathematics, 2018, 78, 1350-1377.	1.8	12
34	Establishment of a five-enzyme cell-free cascade for the synthesis of uridine diphosphate N-acetylglucosamine. Journal of Biotechnology, 2018, 283, 120-129.	3.8	26
35	Use of CellNetAnalyzer in biotechnology and metabolic engineering. Journal of Biotechnology, 2017, 261, 221-228.	3.8	83
36	EColiCore2: a reference network model of the central metabolism of Escherichia coli and relationships to its genome-scale parent model. Scientific Reports, 2017, 7, 39647.	3.3	41

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37	Growth-coupled overproduction is feasible for almost all metabolites in five major production organisms. Nature Communications, 2017, 8, 15956.	12.8	105
38	From elementary flux modes to elementary flux vectors: Metabolic pathway analysis with arbitrary linear flux constraints. PLoS Computational Biology, 2017, 13, e1005409.	3.2	60
39	TRaCE+: Ensemble inference of gene regulatory networks from transcriptional expression profiles of gene knock-out experiments. BMC Bioinformatics, 2016, 17, 252.	2.6	13
40	MUFINS: multi-formalism interaction network simulator. Npj Systems Biology and Applications, 2016, 2, 16032.	3.0	18
41	Model-based metabolic engineering enables high yield itaconic acid production by Escherichia coli. Metabolic Engineering, 2016, 38, 29-37.	7.0	144
42	Predicting compositions of microbial communities from stoichiometric models with applications for the biogas process. Biotechnology for Biofuels, 2016, 9, 17.	6.2	20
43	Exact quantification of cellular robustness in genome-scale metabolic networks. Bioinformatics, 2016, 32, 730-737.	4.1	15
44	Manipulation of the ATP pool as a tool for metabolic engineering. Biochemical Society Transactions, 2015, 43, 1140-1145.	3.4	19
45	Extended notions of sign consistency to relate experimental data to signaling and regulatory network topologies. BMC Bioinformatics, 2015, 16, 345.	2.6	12
46	Enforced ATP futile cycling increases specific productivity and yield of anaerobic lactate production in <i>Escherichia coli</i> in Siotechnology and Bioengineering, 2015, 112, 2195-2199.	3.3	50
47	Disentangling the Complexity of HGF Signaling by Combining Qualitative and Quantitative Modeling. PLoS Computational Biology, 2015, 11, e1004192.	3.2	15
48	On the feasibility of growth-coupled product synthesis in microbial strains. Metabolic Engineering, 2015, 30, 166-178.	7.0	83
49	Genome-scale strain designs based on regulatory minimal cut sets. Bioinformatics, 2015, 31, 2844-2851.	4.1	26
50	Silence on the relevant literature and errors in implementation. Nature Biotechnology, 2015, 33, 336-339.	17.5	14
51	T160â€phosphorylated <scp>CDK</scp> 2 defines threshold for <scp>HGF</scp> â€dependent proliferation in primary hepatocytes. Molecular Systems Biology, 2015, 11, 795.	7.2	19
52	Network-Based Analysis of Nutraceuticals in Human Hepatocellular Carcinomas Reveals Mechanisms of Chemopreventive Action. CPT: Pharmacometrics and Systems Pharmacology, 2015, 4, 350-361.	2.5	4
53	An algorithm for the reduction of genome-scale metabolic network models to meaningful core models. BMC Systems Biology, 2015, 9, 48.	3.0	61
54	Enumeration of Smallest Intervention Strategies in Genome-Scale Metabolic Networks. PLoS Computational Biology, 2014, 10, e1003378.	3.2	97

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55	Cyanobacterial biofuels: new insights and strain design strategies revealed by computational modeling. Microbial Cell Factories, 2014, 13, 128.	4.0	76
56	The avian cell line AGE1.CR.pIX characterized by metabolic flux analysis. BMC Biotechnology, 2014, 14, 72.	3.3	18
57	A model integration approach linking signalling and gene-regulatory logic with kinetic metabolic models. BioSystems, 2014, 124, 26-38.	2.0	18
58	Stoichiometric and Constraint-Based Analysis of Biochemical Reaction Networks. Modeling and Simulation in Science, Engineering and Technology, 2014, , 263-316.	0.6	12
59	Modeling approaches for qualitative and semi-quantitative analysis of cellular signaling networks. Cell Communication and Signaling, 2013, 11, 43.	6.5	106
60	Evaluating differences of metabolic performances: Statistical methods and their application to animal cell cultivations. Biotechnology and Bioengineering, 2013, 110, 2633-2642.	3.3	5
61	In silico profiling of Escherichia coli and Saccharomyces cerevisiae as terpenoid factories. Microbial Cell Factories, 2013, 12, 84.	4.0	78
62	SBML qualitative models: a model representation format and infrastructure to foster interactions between qualitative modelling formalisms and tools. BMC Systems Biology, 2013, 7, 135.	3.0	145
63	Bridging the layers: towards integration of signal transduction, regulation and metabolism into mathematical models. Molecular BioSystems, 2013, 9, 1576.	2.9	83
	mathematical models. Molecular 5105/5001116, 2513, 3, 137 61		
64	Host–Pathogen Interactions. , 2013, , 107-126.		1
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	Host–Pathogen Interactions., 2013, , 107-126.  Detecting and Removing Inconsistencies between Experimental Data and Signaling Network Topologies Using Integer Linear Programming on Interaction Graphs. PLoS Computational Biology, 2013, 9,	3.2 2.6	
65	Host–Pathogen Interactions., 2013,, 107-126.  Detecting and Removing Inconsistencies between Experimental Data and Signaling Network Topologies Using Integer Linear Programming on Interaction Graphs. PLoS Computational Biology, 2013, 9, e1003204.  Comparison and improvement of algorithms for computing minimal cut sets. BMC Bioinformatics,		40
65	Host–Pathogen Interactions., 2013,, 107-126.  Detecting and Removing Inconsistencies between Experimental Data and Signaling Network Topologies Using Integer Linear Programming on Interaction Graphs. PLoS Computational Biology, 2013, 9, e1003204.  Comparison and improvement of algorithms for computing minimal cut sets. BMC Bioinformatics, 2013, 14, 318.  Reconstruction of large-scale regulatory networks based on perturbation graphs and transitive	2.6	22
65 66 67	Host–Pathogen Interactions., 2013, , 107-126.  Detecting and Removing Inconsistencies between Experimental Data and Signaling Network Topologies Using Integer Linear Programming on Interaction Graphs. PLoS Computational Biology, 2013, 9, e1003204.  Comparison and improvement of algorithms for computing minimal cut sets. BMC Bioinformatics, 2013, 14, 318.  Reconstruction of large-scale regulatory networks based on perturbation graphs and transitive reduction: improved methods and their evaluation. BMC Systems Biology, 2013, 7, 73.  Minimal cut sets in a metabolic network are elementary modes in a dual network. Bioinformatics, 2012,	2.6 3.0	40 22 9
65 66 67	Host–Pathogen Interactions., 2013, , 107-126.  Detecting and Removing Inconsistencies between Experimental Data and Signaling Network Topologies Using Integer Linear Programming on Interaction Graphs. PLoS Computational Biology, 2013, 9, e1003204.  Comparison and improvement of algorithms for computing minimal cut sets. BMC Bioinformatics, 2013, 14, 318.  Reconstruction of large-scale regulatory networks based on perturbation graphs and transitive reduction: improved methods and their evaluation. BMC Systems Biology, 2013, 7, 73.  Minimal cut sets in a metabolic network are elementary modes in a dual network. Bioinformatics, 2012, 28, 381-387.  An integrative model links multiple inputs and signaling pathways to the onset of DNA synthesis in	2.6 3.0 4.1	40 22 9 54
65 66 67 68	Host–Pathogen Interactions., 2013, , 107-126.  Detecting and Removing Inconsistencies between Experimental Data and Signaling Network Topologies Using Integer Linear Programming on Interaction Graphs. PLoS Computational Biology, 2013, 9, e1003204.  Comparison and improvement of algorithms for computing minimal cut sets. BMC Bioinformatics, 2013, 14, 318.  Reconstruction of large-scale regulatory networks based on perturbation graphs and transitive reduction: improved methods and their evaluation. BMC Systems Biology, 2013, 7, 73.  Minimal cut sets in a metabolic network are elementary modes in a dual network. Bioinformatics, 2012, 28, 381-387.  An integrative model links multiple inputs and signaling pathways to the onset of DNA synthesis in hepatocytes. FEBS Journal, 2012, 279, 3290-3313.  Large-scale network models of IL-1 and IL-6 signalling and their hepatocellular specification.	2.6 3.0 4.1 4.7	40 22 9 54 30

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73	Computing complex metabolic intervention strategies using constrained minimal cut sets. Metabolic Engineering, 2011, 13, 204-213.	7.0	100
74	CASOP: A Computational Approach for Strain Optimization aiming at high Productivity. Journal of Biotechnology, 2010, 147, 88-101.	3.8	78
75	From Binary to Multivalued to Continuous Models: The lac Operon as a Case Study. Journal of Integrative Bioinformatics, 2010, 7, .	1.5	8
76	TRANSWESD: inferring cellular networks with transitive reduction. Bioinformatics, 2010, 26, 2160-2168.	4.1	27
77	Computing Combinatorial Intervention Strategies and Failure Modes in Signaling Networks. Journal of Computational Biology, 2010, 17, 39-53.	1.6	44
78	Discrete logic modelling as a means to link protein signalling networks with functional analysis of mammalian signal transduction. Molecular Systems Biology, 2009, 5, 331.	7.2	308
79	Hypergraphs and Cellular Networks. PLoS Computational Biology, 2009, 5, e1000385.	3.2	316
80	The Logic of EGFR/ErbB Signaling: Theoretical Properties and Analysis of High-Throughput Data. PLoS Computational Biology, 2009, 5, e1000438.	3.2	164
81	Transforming Boolean models to continuous models: methodology and application to T-cell receptor signaling. BMC Systems Biology, 2009, 3, 98.	3.0	212
82	Computing paths and cycles in biological interaction graphs. BMC Bioinformatics, 2009, 10, 181.	2.6	29
83	Logical network of genotoxic stress-induced NF-kappaB signal transduction predicts putative target structures for therapeutic intervention strategies. Advances and Applications in Bioinformatics and Chemistry, 2009, 2, 125.	2.6	3
84	Host-pathogen systems biology: logical modelling of hepatocyte growth factor and Helicobacter pylori induced c-Met signal transduction. BMC Systems Biology, 2008, 2, 4.	3.0	59
85	Systems-level analysis of redox signaling in Rhodospirillum rubrum. Journal of Biotechnology, 2008, 136, S108.	3 <b>.</b> 8	0
86	Computing Knock-Out Strategies in Metabolic Networks. Journal of Computational Biology, 2008, 15, 259-268.	1.6	60
87	Modeling the electron transport chain of purple nonâ€sulfur bacteria. Molecular Systems Biology, 2008, 4, 156.	7.2	73
88	A Logical Model Provides Insights into T Cell Receptor Signaling. PLoS Computational Biology, 2007, 3, e163.	3.2	311
89	GSMN-TB: a web-based genome-scale network model of Mycobacterium tuberculosis metabolism. Genome Biology, 2007, 8, R89.	9.6	197
90	Structural and functional analysis of cellular networks with CellNetAnalyzer. BMC Systems Biology, 2007, 1, 2.	3.0	454

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91	Generalized concept of minimal cut sets in biochemical networks. BioSystems, 2006, 83, 233-247.	2.0	107
92	Visual setup of logical models of signaling and regulatory networks with ProMoT. BMC Bioinformatics, 2006, 7, 506.	2.6	16
93	A methodology for the structural and functional analysis of signaling and regulatory networks. BMC Bioinformatics, 2006, 7, 56.	2.6	330
94	Stoichiometric and Constraint-based Modeling. , 2006, , 73-96.		18
95	Algorithmic approaches for computing elementary modes in large biochemical reaction networks. IET Systems Biology, 2005, 152, 249.	2.0	62
96	A Logical Model Provides Insights into T-cell Receptor Signaling. PLoS Computational Biology, 2005, preprint, e163.	3.2	0
97	Minimal cut sets in biochemical reaction networks. Bioinformatics, 2004, 20, 226-234.	4.1	239
98	Comparison of network-based pathway analysis methods. Trends in Biotechnology, 2004, 22, 400-405.	9.3	347
99	Computation of elementary modes: a unifying framework and the new binary approach. BMC Bioinformatics, 2004, 5, 175.	2.6	207
100	Workbench zur Modellbildung, Simulation und Analyse zellulÄ <b>r</b> er Systeme (Workbench for Model Set) Tj ETQq0	0 0 rgBT /	Overlock 10 <sup>-</sup>
101	Two approaches for metabolic pathway analysis?. Trends in Biotechnology, 2003, 21, 64-69.	9.3	257
102	FluxAnalyzer: exploring structure, pathways, and flux distributions in metabolic networks on interactive flux maps. Bioinformatics, 2003, 19, 261-269.	4.1	187
103	Applying metabolic pathway analysis to make good use of methanol. Trends in Biotechnology, 2002, 20, 322.	9.3	1
104	Calculability analysis in underdetermined metabolic networks illustrated by a model of the central metabolism in purple nonsulfur bacteria. Biotechnology and Bioengineering, 2002, 77, 734-751.	3.3	112
105	Metabolic network structure determines key aspects of functionality and regulation. Nature, 2002, 420, 190-193.	27.8	712
106	Combinatorial complexity of pathway analysis in metabolic networks. Molecular Biology Reports, 2002, 29, 233-236.	2.3	196
107	Calculating as many fluxes as possible in underdetermined metabolic networks. Molecular Biology Reports, 2002, 29, 243-248.	2.3	15