

Mark D M Leiserson

List of Publications by Year in descending order

Source: <https://exaly.com/author-pdf/2890340/publications.pdf>

Version: 2024-02-01

24
papers

5,737
citations

623574

14
h-index

610775

24
g-index

26
all docs

26
docs citations

26
times ranked

12908
citing authors

#	ARTICLE	IF	CITATIONS
1	Mutational landscape and significance across 12 major cancer types. <i>Nature</i> , 2013, 502, 333-339.	13.7	3,695
2	Pan-cancer network analysis identifies combinations of rare somatic mutations across pathways and protein complexes. <i>Nature Genetics</i> , 2015, 47, 106-114.	9.4	830
3	Integrated analysis of germline and somatic variants in ovarian cancer. <i>Nature Communications</i> , 2014, 5, 3156.	5.8	253
4	Patterns and functional implications of rare germline variants across 12 cancer types. <i>Nature Communications</i> , 2015, 6, 10086.	5.8	243
5	CoMEt: a statistical approach to identify combinations of mutually exclusive alterations in cancer. <i>Genome Biology</i> , 2015, 16, 160.	3.8	182
6	Precision Oncology: The Road Ahead. <i>Trends in Molecular Medicine</i> , 2017, 23, 874-898.	3.5	131
7	Hierarchical HotNet: identifying hierarchies of altered subnetworks. <i>Bioinformatics</i> , 2018, 34, i972-i980.	1.8	102
8	Network analysis of GWAS data. <i>Current Opinion in Genetics and Development</i> , 2013, 23, 602-610.	1.5	95
9	MAGI: visualization and collaborative annotation of genomic aberrations. <i>Nature Methods</i> , 2015, 12, 483-484.	9.0	25
10	A systematic genome-wide mapping of oncogenic mutation selection during CRISPR-Cas9 genome editing. <i>Nature Communications</i> , 2021, 12, 6512.	5.8	24
11	Criticality in tumor evolution and clinical outcome. <i>Proceedings of the National Academy of Sciences of the United States of America</i> , 2018, 115, E11101-E11110.	3.3	23
12	Functional protein representations from biological networks enable diverse cross-species inference. <i>Nucleic Acids Research</i> , 2019, 47, e51-e51.	6.5	23
13	Hidden Markov models lead to higher resolution maps of mutation signature activity in cancer. <i>Genome Medicine</i> , 2019, 11, 49.	3.6	22
14	Network-based approaches elucidate differences within APOBEC and clock-like signatures in breast cancer. <i>Genome Medicine</i> , 2020, 12, 52.	3.6	20
15	Mutational Signatures: From Methods to Mechanisms. <i>Annual Review of Biomedical Data Science</i> , 2021, 4, 189-206.	2.8	19
16	Modeling clinical and molecular covariates of mutational process activity in cancer. <i>Bioinformatics</i> , 2019, 35, i492-i500.	1.8	15
17	A multifactorial model of T cell expansion and durable clinical benefit in response to a PD-L1 inhibitor. <i>PLoS ONE</i> , 2018, 13, e0208422.	1.1	14
18	A mixture model for signature discovery from sparse mutation data. <i>Genome Medicine</i> , 2021, 13, 173.	3.6	8

#	ARTICLE	IF	CITATIONS
19	A Sticky Multinomial Mixture Model of Strand-Coordinated Mutational Processes in Cancer. <i>IScience</i> , 2020, 23, 100900.	1.9	5
20	Reply: Co-occurrence of MYC amplification and TP53 mutations in human cancer. <i>Nature Genetics</i> , 2016, 48, 106-108.	9.4	2
21	SuperDendrix algorithm integrates genetic dependencies and genomic alterations across pathways and cancer types. <i>Cell Genomics</i> , 2022, 2, 100099.	3.0	2
22	A data-driven approach for constructing mutation categories for mutational signature analysis. <i>PLoS Computational Biology</i> , 2021, 17, e1009542.	1.5	1
23	Matrix (factorization) reloaded: flexible methods for imputing genetic interactions with cross-species and side information. <i>Bioinformatics</i> , 2020, 36, i866-i874.	1.8	1
24	<scp>ScapelSig</scp> Designs Targeted Genomic Panels from Data to Detect Activity of Mutational Signatures. <i>Journal of Computational Biology</i> , 2022, 29, 56-73.	0.8	1