

Benjamin J Raphael

List of Publications by Year in descending order

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Version: 2024-02-01

84
papers

17,852
citations

66234

42
h-index

66788

78
g-index

108
all docs

108
docs citations

108
times ranked

27998
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	International network of cancer genome projects. <i>Nature</i> , 2010, 464, 993-998.	13.7	2,114
3	Integrated Genomic Characterization of Pancreatic Ductal Adenocarcinoma. <i>Cancer Cell</i> , 2017, 32, 185-203.e13.	7.7	1,428
4	Multiplatform Analysis of 12 Cancer Types Reveals Molecular Classification within and across Tissues of Origin. <i>Cell</i> , 2014, 158, 929-944.	13.5	1,242
5	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
6	Eleven grand challenges in single-cell data science. <i>Genome Biology</i> , 2020, 21, 31.	3.8	742
7	Network propagation: a universal amplifier of genetic associations. <i>Nature Reviews Genetics</i> , 2017, 18, 551-562.	7.7	514
8	Comprehensive Pan-Genomic Characterization of Adrenocortical Carcinoma. <i>Cancer Cell</i> , 2016, 29, 723-736.	7.7	482
9	Algorithms for Detecting Significantly Mutated Pathways in Cancer. <i>Journal of Computational Biology</i> , 2011, 18, 507-522.	0.8	434
10	Analyses of non-coding somatic drivers in 2,658 cancer whole genomes. <i>Nature</i> , 2020, 578, 102-111.	13.7	424
11	De novo discovery of mutated driver pathways in cancer. <i>Genome Research</i> , 2012, 22, 375-385.	2.4	391
12	Pathway and network analysis of cancer genomes. <i>Nature Methods</i> , 2015, 12, 615-621.	9.0	297
13	The Integrated Genomic Landscape of Thymic Epithelial Tumors. <i>Cancer Cell</i> , 2018, 33, 244-258.e10.	7.7	270
14	Characterizing genetic intra-tumor heterogeneity across 2,658 human cancer genomes. <i>Cell</i> , 2021, 184, 2239-2254.e39.	13.5	260
15	Integrated analysis of germline and somatic variants in ovarian cancer. <i>Nature Communications</i> , 2014, 5, 3156.	5.8	253
16	Patterns and functional implications of rare germline variants across 12 cancer types. <i>Nature Communications</i> , 2015, 6, 10086.	5.8	243
17	Simultaneous Identification of Multiple Driver Pathways in Cancer. <i>PLoS Computational Biology</i> , 2013, 9, e1003054.	1.5	231
18	THetA: inferring intra-tumor heterogeneity from high-throughput DNA sequencing data. <i>Genome Biology</i> , 2013, 14, R80.	13.9	209

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19	Reconstruction of clonal trees and tumor composition from multi-sample sequencing data. <i>Bioinformatics</i> , 2015, 31, i62-i70.	1.8	194
20	Identifying driver mutations in sequenced cancer genomes: computational approaches to enable precision medicine. <i>Genome Medicine</i> , 2014, 6, 5.	3.6	186
21	CoMEt: a statistical approach to identify combinations of mutually exclusive alterations in cancer. <i>Genome Biology</i> , 2015, 16, 160.	3.8	182
22	Therapy-induced mutations drive the genomic landscape of relapsed acute lymphoblastic leukemia. <i>Blood</i> , 2020, 135, 41-55.	0.6	171
23	Expanding the computational toolbox for mining cancer genomes. <i>Nature Reviews Genetics</i> , 2014, 15, 556-570.	7.7	166
24	Single-cell sequencing data reveal widespread recurrence and loss of mutational hits in the life histories of tumors. <i>Genome Research</i> , 2017, 27, 1885-1894.	2.4	156
25	Inferring the Mutational History of a Tumor Using Multi-state Perfect Phylogeny Mixtures. <i>Cell Systems</i> , 2016, 3, 43-53.	2.9	140
26	A geometric approach for classification and comparison of structural variants. <i>Bioinformatics</i> , 2009, 25, i222-i230.	1.8	138
27	Identification of hierarchical chromatin domains. <i>Bioinformatics</i> , 2016, 32, 1601-1609.	1.8	134
28	An integrative probabilistic model for identification of structural variation in sequencing data. <i>Genome Biology</i> , 2012, 13, R22.	13.9	123
29	Quantifying tumor heterogeneity in whole-genome and whole-exome sequencing data. <i>Bioinformatics</i> , 2014, 30, 3532-3540.	1.8	115
30	Visible Machine Learning for Biomedicine. <i>Cell</i> , 2018, 173, 1562-1565.	13.5	115
31	Hierarchical HotNet: identifying hierarchies of altered subnetworks. <i>Bioinformatics</i> , 2018, 34, i972-i980.	1.8	102
32	A combinatorial approach for analyzing intra-tumor heterogeneity from high-throughput sequencing data. <i>Bioinformatics</i> , 2014, 30, i78-i86.	1.8	100
33	Characterizing allele- and haplotype-specific copy numbers in single cells with CHISEL. <i>Nature Biotechnology</i> , 2021, 39, 207-214.	9.4	97
34	Network analysis of GWAS data. <i>Current Opinion in Genetics and Development</i> , 2013, 23, 602-610.	1.5	95
35	Inferring parsimonious migration histories for metastatic cancers. <i>Nature Genetics</i> , 2018, 50, 718-726.	9.4	93
36	Pathway and network analysis of more than 2500 whole cancer genomes. <i>Nature Communications</i> , 2020, 11, 729.	5.8	73

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37	Alignment and integration of spatial transcriptomics data. <i>Nature Methods</i> , 2022, 19, 567-575.	9.0	73
38	Expression Profiling of Primary and Metastatic Ovarian Tumors Reveals Differences Indicative of Aggressive Disease. <i>PLoS ONE</i> , 2014, 9, e94476.	1.1	66
39	Accurate quantification of copy-number aberrations and whole-genome duplications in multi-sample tumor sequencing data. <i>Nature Communications</i> , 2020, 11, 4301.	5.8	62
40	netNMF-sc: leveraging gene-gene interactions for imputation and dimensionality reduction in single-cell expression analysis. <i>Genome Research</i> , 2020, 30, 195-204.	2.4	61
41	SCARLET: Single-Cell Tumor Phylogeny Inference with Copy-Number Constrained Mutation Losses. <i>Cell Systems</i> , 2020, 10, 323-332.e8.	2.9	61
42	Identifying structural variants using linked-read sequencing data. <i>Bioinformatics</i> , 2018, 34, 353-360.	1.8	60
43	DISCOVERY OF MUTATED SUBNETWORKS ASSOCIATED WITH CLINICAL DATA IN CANCER. , 2011, , .		59
44	Comprehensive characterization of 536 patient-derived xenograft models prioritizes candidates for targeted treatment. <i>Nature Communications</i> , 2021, 12, 5086.	5.8	58
45	Characterization of structural variants with single molecule and hybrid sequencing approaches. <i>Bioinformatics</i> , 2014, 30, 3458-3466.	1.8	56
46	Gene and Network Analysis of Common Variants Reveals Novel Associations in Multiple Complex Diseases. <i>Genetics</i> , 2016, 204, 783-798.	1.2	56
47	A weighted exact test for mutually exclusive mutations in cancer. <i>Bioinformatics</i> , 2016, 32, i736-i745.	1.8	46
48	Tumor phylogeny inference using tree-constrained importance sampling. <i>Bioinformatics</i> , 2017, 33, i152-i160.	1.8	46
49	CALDER: Inferring Phylogenetic Trees from Longitudinal Tumor Samples. <i>Cell Systems</i> , 2019, 8, 514-522.e5.	2.9	46
50	Detecting non-allelic homologous recombination from high-throughput sequencing data. <i>Genome Biology</i> , 2015, 16, 72.	3.8	40
51	The Clonal Evolution of Metastatic Osteosarcoma as Shaped by Cisplatin Treatment. <i>Molecular Cancer Research</i> , 2019, 17, 895-906.	1.5	40
52	Reconstructing cancer genomes from paired-end sequencing data. <i>BMC Bioinformatics</i> , 2012, 13, S10.	1.2	36
53	STARCH: copy number and clone inference from spatial transcriptomics data. <i>Physical Biology</i> , 2021, 18, 035001.	0.8	35
54	Structural variation analysis with strobe reads. <i>Bioinformatics</i> , 2010, 26, 1291-1298.	1.8	33

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55	A sequence-based survey of the complex structural organization of tumor genomes. <i>Genome Biology</i> , 2008, 9, R59.	13.9	31
56	Simultaneous Inference of Cancer Pathways and Tumor Progression from Cross-Sectional Mutation Data. <i>Journal of Computational Biology</i> , 2015, 22, 510-527.	0.8	28
57	Phylogenetic Copy-Number Factorization of Multiple Tumor Samples. <i>Journal of Computational Biology</i> , 2018, 25, 689-708.	0.8	28
58	Complexity and algorithms for copy-number evolution problems. <i>Algorithms for Molecular Biology</i> , 2017, 12, 13.	0.3	27
59	Chapter 6: Structural Variation and Medical Genomics. <i>PLoS Computational Biology</i> , 2012, 8, e1002821.	1.5	26
60	MAGI: visualization and collaborative annotation of genomic aberrations. <i>Nature Methods</i> , 2015, 12, 483-484.	9.0	25
61	Accurate Computation of Survival Statistics in Genome-Wide Studies. <i>PLoS Computational Biology</i> , 2015, 11, e1004071.	1.5	24
62	Reconstructing tumor amplicomes. <i>Bioinformatics</i> , 2004, 20, i265-i273.	1.8	22
63	Detecting independent and recurrent copy number aberrations using interval graphs. <i>Bioinformatics</i> , 2014, 30, i195-i203.	1.8	22
64	Copy-Number Evolution Problems: Complexity and Algorithms. <i>Lecture Notes in Computer Science</i> , 2016, , 137-149.	1.0	17
65	GenomeVIP: a cloud platform for genomic variant discovery and interpretation. <i>Genome Research</i> , 2017, 27, 1450-1459.	2.4	15
66	Identifying tumor clones in sparse single-cell mutation data. <i>Bioinformatics</i> , 2020, 36, i186-i193.	1.8	15
67	Haplotype phasing in single-cell DNA-sequencing data. <i>Bioinformatics</i> , 2018, 34, i211-i217.	1.8	13
68	Reconstruction of clone- and haplotype-specific cancer genome karyotypes from bulk tumor samples. <i>Genome Research</i> , 2020, 30, 1274-1290.	2.4	12
69	DeCiFering the elusive cancer cell fraction in tumor heterogeneity and evolution. <i>Cell Systems</i> , 2021, 12, 1004-1018.e10.	2.9	12
70	Copy number evolution with weighted aberrations in cancer. <i>Bioinformatics</i> , 2020, 36, i344-i352.	1.8	11
71	NetMix: A Network-Structured Mixture Model for Reduced-Bias Estimation of Altered Subnetworks. <i>Journal of Computational Biology</i> , 2021, 28, 469-484.	0.8	8
72	Novel Gene and Network Associations Found for Acute Lymphoblastic Leukemia Using Case-€“Control and Family-Based Studies in Multiethnic Populations. <i>Cancer Epidemiology Biomarkers and Prevention</i> , 2017, 26, 1531-1539.	1.1	7

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73	On the Sample Complexity of Cancer Pathways Identification. Journal of Computational Biology, 2016, 23, 30-41.	0.8	6
74	Open adjacencies and k-breaks: detecting simultaneous rearrangements in cancer genomes. BMC Genomics, 2014, 15, S4.	1.2	5
75	Identifying simultaneous rearrangements in cancer genomes. Bioinformatics, 2018, 34, 346-352.	1.8	4
76	Reply: Co-occurrence of MYC amplification and TP53 mutations in human cancer. Nature Genetics, 2016, 48, 106-108.	9.4	2
77	SuperDendrix algorithm integrates genetic dependencies and genomic alterations across pathways and cancer types. Cell Genomics, 2022, 2, 100099.	3.0	2
78	A PARSIMONY APPROACH TO ANALYSIS OF HUMAN SEGMENTAL DUPLICATIONS. , 2008, , .		1
79	MODELING CELL HETEROGENEITY: FROM SINGLE-CELL VARIATIONS TO MIXED CELLS. , 2012, , .		1
80	Workshop: Algorithms for discovery of mutated pathways in cancer. , 2012, , .		0
81	POST-NGS: INTERPRETATION AND ANALYSIS OF NEXT GENERATION SEQUENCING DATA FOR BASIC AND TRANSLATIONAL SCIENCE. , 2012, , .		0
82	Identifying significant mutations in large cohorts of cancer genomes. , 2013, , .		0
83	Workshop: Reconstructing the organization of cancer genomes. , 2013, , .		0
84	Using controls to limit false discovery in the era of big data. BMC Bioinformatics, 2018, 19, 323.	1.2	0