Markus Ringnér

List of Publications by Year in descending order

Source: https://exaly.com/author-pdf/224539/publications.pdf

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95 papers 16,943 citations

45 h-index 93 g-index

101 all docs

101 docs citations

times ranked

101

28703 citing authors

#	Article	IF	CITATIONS
1	Environmentally induced DNA methylation is inherited across generations in an aquatic keystone species. IScience, 2022, 25, 104303.	1.9	11
2	Mutational patterns and clonal evolution from diagnosis to relapse in pediatric acute lymphoblastic leukemia. Scientific Reports, 2021, 11, 15988.	1.6	6
3	Mitochondrial dysfunction in adult midbrain dopamine neurons triggers an early immune response. PLoS Genetics, 2021, 17, e1009822.	1.5	8
4	Tissue-specific transcriptional imprinting and heterogeneity in human innate lymphoid cells revealed by full-length single-cell RNA-sequencing. Cell Research, 2021, 31, 554-568.	5.7	97
5	Retrospective evaluation of whole exome and genome mutation calls in 746 cancer samples. Nature Communications, 2020, 11, 4748.	5.8	27
6	Sex differences in oncogenic mutational processes. Nature Communications, 2020, 11, 4330.	5.8	60
7	Analysis of DNA methylation patterns in the tumor immune microenvironment of metastatic melanoma. Molecular Oncology, 2020, 14, 933-950.	2.1	29
8	Pan-cancer analysis of whole genomes. Nature, 2020, 578, 82-93.	13.7	1,966
9	Partially methylated domains are hypervariable in breast cancer and fuel widespread CpG island hypermethylation. Nature Communications, 2019, 10, 1749.	5.8	46
10	A comprehensive map coupling histone modifications with gene regulation in adult dopaminergic and serotonergic neurons. Nature Communications, 2018, 9, 1226.	5.8	35
11			
	Spatially and functionally distinct subclasses of breast cancer-associated fibroblasts revealed by single cell RNA sequencing. Nature Communications, 2018, 9, 5150.	5.8	496
12	Spatially and functionally distinct subclasses of breast cancer-associated fibroblasts revealed by single cell RNA sequencing. Nature Communications, 2018, 9, 5150. Mutational and putative neoantigen load predict clinical benefit of adoptive T cell therapy in melanoma. Nature Communications, 2017, 8, 1738.	5.8 5.8	310
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12 13 14	Mutational and putative neoantigen load predict clinical benefit of adoptive T cell therapy in melanoma. Nature Communications, 2017, 8, 1738. Landscape of somatic mutations in 560 breast cancer whole-genome sequences. Nature, 2016, 534, 47-54. Breast cancer genome and transcriptome integration implicates specific mutational signatures with immune cell infiltration. Nature Communications, 2016, 7, 12910. An integrated genomics analysis of epigenetic subtypes in human breast tumors links DNA methylation	5.8 13.7 5.8	310 1,760 119
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19	Genome-Wide DNA Methylation Analysis in Melanoma Reveals the Importance of CpG Methylation in MITF Regulation. Journal of Investigative Dermatology, 2015, 135, 1820-1828.	0.3	46
20	Genome methylation patterns in male breast cancer $\hat{a} \in \text{``ldentification of an epitype with hypermethylation of polycomb target genes. Molecular Oncology, 2015, 9, 1565-1579.}$	2.1	14
21	DNA methylation and histone modifications regulate SOX11 expression in lymphoid and solid cancer cells. BMC Cancer, 2015, 15, 273.	1.1	14
22	Integrative epigenomic analysis of differential DNA methylation in urothelial carcinoma. Genome Medicine, 2015, 7, 23.	3.6	42
23	MicroRNA-200c-141 and â^†Np63 are required for breast epithelial differentiation and branching morphogenesis. Developmental Biology, 2015, 403, 150-161.	0.9	23
24	Molecular stratification of metastatic melanoma using gene expression profiling: Prediction of survival outcome and benefit from molecular targeted therapy. Oncotarget, 2015, 6, 12297-12309.	0.8	148
25	Making Breast Cancer Molecular Subtypes Robust?. Journal of the National Cancer Institute, 2014, 107, dju386-dju386.	3.0	10
26	Genome-wide DNA Methylation Analysis of Lung Carcinoma Reveals One Neuroendocrine and Four Adenocarcinoma Epitypes Associated with Patient Outcome. Clinical Cancer Research, 2014, 20, 6127-6140.	3.2	91
27	Genomic and Transcriptional Alterations in Lung Adenocarcinoma in Relation to Smoking History. Clinical Cancer Research, 2014, 20, 4912-4924.	3.2	24
28	Primary Melanoma Tumors from CDKN2A Mutation Carriers Do Not Belong to a Distinct Molecular Subclass. Journal of Investigative Dermatology, 2014, 134, 3000-3003.	0.3	8
29	Nonfamilial Breast Cancer Subtypes. Methods in Molecular Biology, 2013, 973, 279-295.	0.4	5
30	Exosomes reflect the hypoxic status of glioma cells and mediate hypoxia-dependent activation of vascular cells during tumor development. Proceedings of the National Academy of Sciences of the United States of America, 2013, 110, 7312-7317.	3.3	768
31	High expression of <scp><i>ZNF703</i></scp> independent of amplification indicates worse prognosis in patients with luminal B breast cancer. Cancer Medicine, 2013, 2, 437-446.	1.3	39
32	Monitoring of Technical Variation in Quantitative High-Throughput Datasets. Cancer Informatics, 2013, 12, CIN.S12862.	0.9	47
33	The Landscape of Candidate Driver Genes Differs between Male and Female Breast Cancer. PLoS ONE, 2013, 8, e78299.	1.1	46
34	A Molecular Taxonomy for Urothelial Carcinoma. Clinical Cancer Research, 2012, 18, 3377-3386.	3.2	729
35	The Retinoblastoma Gene Undergoes Rearrangements in <i>BRCA1</i> -Deficient Basal-like Breast Cancer. Cancer Research, 2012, 72, 4028-4036.	0.4	41
36	DNA methylation analyses of urothelial carcinoma reveal distinct epigenetic subtypes and an association between gene copy number and methylation status. Epigenetics, 2012, 7, 858-867.	1.3	44

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37	Molecular Profiling Reveals Low- and High-Grade Forms of Primary Melanoma. Clinical Cancer Research, 2012, 18, 4026-4036.	3.2	96
38	Global H3K27 trimethylation and EZH2 abundance in breast tumor subtypes. Molecular Oncology, 2012, 6, 494-506.	2.1	136
39	The gene expression landscape of breast cancer is shaped by tumor protein p53 status and epithelial-mesenchymal transition. Breast Cancer Research, 2012, 14, R113.	2.2	49
40	Characterisation of amplification patterns and target genes at chromosome 11q13 in CCND1-amplified sporadic and familial breast tumours. Breast Cancer Research and Treatment, 2012, 133, 583-594.	1.1	44
41	Gene expression profiling of primary male breast cancers reveals two unique subgroups and identifies N-acetyltransferase-1 (NAT1) as a novel prognostic biomarker. Breast Cancer Research, 2012, 14, R31.	2.2	100
42	Landscape of somatic allelic imbalances and copy number alterations in HER2-amplified breast cancer. Breast Cancer Research, 2011, 13, R129.	2.2	25
43	GOBO: Gene Expression-Based Outcome for Breast Cancer Online. PLoS ONE, 2011, 6, e17911.	1.1	361
44	CD44 isoforms are heterogeneously expressed in breast cancer and correlate with tumor subtypes and cancer stem cell markers. BMC Cancer, 2011, 11, 418.	1.1	173
45	High-resolution genomic profiling of male breast cancer reveals differences hidden behind the similarities with female breast cancer. Breast Cancer Research and Treatment, 2011, 129, 747-760.	1.1	70
46	Endothelial Induced EMT in Breast Epithelial Cells with Stem Cell Properties. PLoS ONE, 2011, 6, e23833.	1.1	87
47	Genome-wide transcription factor binding site/promoter databases for the analysis of gene sets and co-occurrence of transcription factor binding motifs. BMC Genomics, 2010, 11, 145.	1.2	31
48	An integrative genomics screen uncovers ncRNA T-UCR functions in neuroblastoma tumours. Oncogene, 2010, 29, 3583-3592.	2.6	141
49	Gene Expression Profiling–Based Identification of Molecular Subtypes in Stage IV Melanomas with Different Clinical Outcome. Clinical Cancer Research, 2010, 16, 3356-3367.	3.2	235
50	Identification of Subtypes in Human Epidermal Growth Factor Receptor 2–Positive Breast Cancer Reveals a Gene Signature Prognostic of Outcome. Journal of Clinical Oncology, 2010, 28, 1813-1820.	0.8	145
51	Prediction of Stage, Grade, and Survival in Bladder Cancer Using Genome-wide Expression Data: A Validation Study. Clinical Cancer Research, 2010, 16, 4421-4433.	3.2	36
52	High-resolution genomic and expression analyses of copy number alterations in HER2-amplified breast cancer. Breast Cancer Research, 2010, 12, R25.	2.2	123
53	Molecular subtypes of breast cancer are associated with characteristic DNA methylation patterns. Breast Cancer Research, 2010, 12, R36.	2.2	251
54	Genomic subtypes of breast cancer identified by array-comparative genomic hybridization display distinct molecular and clinical characteristics. Breast Cancer Research, 2010, 12, R42.	2.2	167

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55	Multiple metastases from cutaneous malignant melanoma patients may display heterogeneous genomic and epigenomic patterns. Melanoma Research, 2010, 20, 381-391.	0.6	30
56	Tiling array-CGH for the assessment of genomic similarities among synchronous unilateral and bilateral invasive breast cancer tumor pairs. BMC Clinical Pathology, 2008, 8, 6.	1.8	28
57	What is principal component analysis?. Nature Biotechnology, 2008, 26, 303-304.	9.4	1,634
58	Normalization of Illumina Infinium whole-genome SNP data improves copy number estimates and allelic intensity ratios. BMC Bioinformatics, 2008, 9, 409.	1.2	114
59	The CD44+/CD24-phenotype is enriched in basal-like breast tumors. Breast Cancer Research, 2008, 10, R53.	2.2	464
60	Segmentation-based detection of allelic imbalance and loss-of-heterozygosity in cancer cells using whole genome SNP arrays. Genome Biology, 2008, 9, R136.	13.9	127
61	High Myc pathway activity and low stage of neuronal differentiation associate with poor outcome in neuroblastoma. Proceedings of the National Academy of Sciences of the United States of America, 2008, 105, 14094-14099.	3.3	149
62	Poor prognosis in carcinoma is associated with a gene expression signature of aberrant PTEN tumor suppressor pathway activity. Proceedings of the National Academy of Sciences of the United States of America, 2007, 104, 7564-7569.	3.3	445
63	Revealing signaling pathway deregulation by using gene expression signatures and regulatory motif analysis. Genome Biology, 2007, 8, R77.	13.9	18
64	Classification of Genomic and Proteomic Data Using Support Vector Machines., 2007, , 187-202.		4
65	High-resolution genomic profiles of breast cancer cell lines assessed by tiling BAC array comparative genomic hybridization. Genes Chromosomes and Cancer, 2007, 46, 543-558.	1.5	176
66	Genomic profiling of malignant melanoma using tiling-resolution arrayCGH. Oncogene, 2007, 26, 4738-4748.	2.6	118
67	Normalization of array-CGH data: influence of copy number imbalances. BMC Genomics, 2007, 8, 382.	1.2	57
68	Detection and Identification of Protein Isoforms Using Cluster Analysis of MALDIâ [^] 'MS Mass Spectra. Journal of Proteome Research, 2006, 5, 785-792.	1.8	68
69	Osteopontin is a downstream effector of the PI3-kinase pathway in melanomas that is inversely correlated with functional PTEN. Carcinogenesis, 2006, 27, 1778-1786.	1.3	55
70	Distinct Genomic Profiles in Hereditary Breast Tumors Identified by Array-Based Comparative Genomic Hybridization. Cancer Research, 2005, 65, 7612-7621.	0.4	147
71	Gene expression profiling demonstrates that TGF- \hat{l}^21 signals exclusively through receptor complexes involving Alk5 and identifies targets of TGF- \hat{l}^2 signaling. Physiological Genomics, 2005, 21, 396-403.	1.0	33
72	Folding Free Energies of 5′-UTRs Impact Post-Transcriptional Regulation on a Genomic Scale in Yeast. PLoS Computational Biology, 2005, 1, e72.	1.5	116

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73	Folding free energies of 5'-UTRs impact post-transcriptional regulation on a genomic scale in yeast. PLoS Computational Biology, 2005, preprint, e72.	1.5	0
74	ACID: a database for microarray clone information. Bioinformatics, 2004, 20, 2305-2306.	1.8	18
75	Microarray expression profiling in melanoma reveals a BRAF mutation signature. Oncogene, 2004, 23, 4060-4067.	2.6	169
76	Multiclass discovery in array data. BMC Bioinformatics, 2004, 5, 70.	1.2	11
77	A strategy for identifying putative causes of gene expression variation in human cancers. Journal of the Franklin Institute, 2004, 341, 77-88.	1.9	21
78	Analyzing tumor gene expression profiles. Artificial Intelligence in Medicine, 2003, 28, 59-74.	3.8	43
79	Gene expression profile in multiple sclerosis patients and healthy controls: identifying pathways relevant to disease. Human Molecular Genetics, 2003, 12, 2191-2199.	1.4	191
80	Matching protein structures with fuzzy alignments. Proceedings of the National Academy of Sciences of the United States of America, 2003, 100, 11936-11940.	3.3	29
81	Molecular classification of familial non-BRCA1/BRCA2 breast cancer. Proceedings of the National Academy of Sciences of the United States of America, 2003, 100, 2532-2537.	3.3	182
82	Classification of Expression Patterns Using Artificial Neural Networks., 2003,, 201-215.		3
83	Microarray-Based Cancer Diagnosis with Artificial Neural Networks. BioTechniques, 2003, 34, S30-S35.	0.8	37
83	Microarray-Based Cancer Diagnosis with Artificial Neural Networks. BioTechniques, 2003, 34, S30-S35. Microarray-based cancer diagnosis with artificial neural networks. BioTechniques, 2003, Suppl, 30-5.	0.8	10
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84	Microarray-based cancer diagnosis with artificial neural networks. BioTechniques, 2003, Suppl, 30-5. Gene expression in inherited breast cancer. Advances in Cancer Research, 2002, 84, 1-34.	0.8	10
84 85 86	Microarray-based cancer diagnosis with artificial neural networks. BioTechniques, 2003, Suppl, 30-5. Gene expression in inherited breast cancer. Advances in Cancer Research, 2002, 84, 1-34. Analyzing array data using supervised methods. Pharmacogenomics, 2002, 3, 403-415. Expression profiling to predict outcome in breast cancer: the influence of sample selection. Breast	0.8 1.9 0.6	10 44 67
84 85 86	Microarray-based cancer diagnosis with artificial neural networks. BioTechniques, 2003, Suppl, 30-5. Gene expression in inherited breast cancer. Advances in Cancer Research, 2002, 84, 1-34. Analyzing array data using supervised methods. Pharmacogenomics, 2002, 3, 403-415. Expression profiling to predict outcome in breast cancer: the influence of sample selection. Breast Cancer Research, 2002, 5, 23-6. Impact of DNA amplification on gene expression patterns in breast cancer. Cancer Research, 2002, 62,	0.8 1.9 0.6	10 44 67 35

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91	The Feynman–Wilson gas and the Lund model. European Physical Journal C, 1999, 7, 251.	1.4	o
92	Transverse and longitudinal Bose-Einstein correlations. Physics Letters, Section B: Nuclear, Elementary Particle and High-Energy Physics, 1998, 421, 283-288.	1.5	19
93	Bose-Einstein correlations in the Lund model. Nuclear Physics B, 1998, 513, 627-644.	0.9	27
94	Is there screwiness at the end of the QCD cascades?. Journal of High Energy Physics, 1998, 1998, 014-014.	1.6	10
95	Bose-Einstein and colour interference in W-pair decays. European Physical Journal C, 1998, 5, 275.	1.4	3