## Jason Ernst

## List of Publications by Year in descending order

Source: https://exaly.com/author-pdf/9108975/publications.pdf

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172457 214800 20,479 45 29 47 h-index citations g-index papers 61 61 61 34102 docs citations times ranked citing authors all docs

#	Article	IF	CITATIONS
1	Integrative analysis of 111 reference human epigenomes. Nature, 2015, 518, 317-330.	27.8	5,653
2	Mapping and analysis of chromatin state dynamics in nine human cell types. Nature, 2011, 473, 43-49.	27.8	2,630
3	ChromHMM: automating chromatin-state discovery and characterization. Nature Methods, 2012, 9, 215-216.	19.0	2,127
4	STEM: a tool for the analysis of short time series gene expression data. BMC Bioinformatics, 2006, 7, 191.	2.6	1,309
5	Identification of Functional Elements and Regulatory Circuits by <i>Drosophila</i> modENCODE. Science, 2010, 330, 1787-1797.	12.6	1,124
6	A high-resolution map of human evolutionary constraint using 29 mammals. Nature, 2011, 478, 476-482.	27.8	1,016
7	Discovery and characterization of chromatin states for systematic annotation of the human genome. Nature Biotechnology, 2010, 28, 817-825.	17.5	947
8	Alzheimer's disease: early alterations in brain DNA methylation at ANK1, BIN1, RHBDF2 and other loci. Nature Neuroscience, 2014, 17, 1156-1163.	14.8	800
9	Comprehensive analysis of the chromatin landscape in Drosophila melanogaster. Nature, 2011, 471, 480-485.	27.8	781
10	Chromatin-state discovery and genome annotation with ChromHMM. Nature Protocols, 2017, 12, 2478-2492.	12.0	613
11	Integrative annotation of chromatin elements from ENCODE data. Nucleic Acids Research, 2013, 41, 827-841.	14.5	490
12	Cooperative Binding of Transcription Factors Orchestrates Reprogramming. Cell, 2017, 168, 442-459.e20.	28.9	432
13	Large-scale imputation of epigenomic datasets for systematic annotation of diverse human tissues. Nature Biotechnology, 2015, 33, 364-376.	17.5	354
14	Systematic dissection of regulatory motifs in 2000 predicted human enhancers using a massively parallel reporter assay. Genome Research, 2013, 23, 800-811.	5.5	298
15	Long-Range Chromatin Contacts in Embryonic Stem Cells Reveal a Role for Pluripotency Factors and Polycomb Proteins in Genome Organization. Cell Stem Cell, 2013, 13, 602-616.	11.1	246
16	Reconstructing dynamic regulatory maps. Molecular Systems Biology, 2007, 3, 74.	7.2	183
17	Deep-coverage whole genome sequences and blood lipids among 16,324 individuals. Nature Communications, 2018, 9, 3391.	12.8	140
18	Genome-scale high-resolution mapping of activating and repressive nucleotides in regulatory regions. Nature Biotechnology, 2016, 34, 1180-1190.	17.5	132

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19	Transcriptional analysis of cystic fibrosis airways at single-cell resolution reveals altered epithelial cell states and composition. Nature Medicine, 2021, 27, 806-814.	30.7	101
20	Analysis of cardiomyocyte clonal expansion during mouse heart development and injury. Nature Communications, 2018, 9, 754.	12.8	94
21	A mammalian methylation array for profiling methylation levels at conserved sequences. Nature Communications, 2022, 13, 783.	12.8	93
22	A Semi-Supervised Method for Predicting Transcription Factor–Gene Interactions in Escherichia coli. PLoS Computational Biology, 2008, 4, e1000044.	3.2	85
23	Interplay between chromatin state, regulator binding, and regulatory motifs in six human cell types. Genome Research, 2013, 23, 1142-1154.	5.5	84
24	Deep coverage whole genome sequences and plasma lipoprotein(a) in individuals of European and African ancestries. Nature Communications, 2018, 9, 2606.	12.8	79
25	Systematic Epigenomic Analysis Reveals Chromatin States Associated with Melanoma Progression. Cell Reports, 2017, 19, 875-889.	6.4	78
26	Dual Roles of RNF2 in Melanoma Progression. Cancer Discovery, 2015, 5, 1314-1327.	9.4	57
27	Integrating and mining the chromatin landscape of cell-type specificity using self-organizing maps. Genome Research, 2013, 23, 2136-2148.	5.5	51
28	Mapping molecular landmarks of human skeletal ontogeny and pluripotent stem cell-derived articular chondrocytes. Nature Communications, 2018, 9, 3634.	12.8	49
29	Distinct and Predictive Histone Lysine Acetylation Patterns at Promoters, Enhancers, and Gene Bodies. G3: Genes, Genomes, Genetics, 2014, 4, 2051-2063.	1.8	39
30	Universal annotation of the human genome through integration of over a thousand epigenomic datasets. Genome Biology, 2022, 23, 9.	8.8	39
31	Evidence of reduced recombination rate in human regulatory domains. Genome Biology, 2017, 18, 193.	8.8	38
32	Circadian variation of the pancreatic islet transcriptome. Physiological Genomics, 2016, 48, 677-687.	2.3	28
33	Epigenetic clock and methylation studies in the rhesus macaque. GeroScience, 2021, 43, 2441-2453.	4.6	28
34	Epigenetic clock and methylation studies in vervet monkeys. GeroScience, 2022, 44, 699-717.	4.6	18
35	Systematic discovery of conservation states for single-nucleotide annotation of the human genome. Communications Biology, 2019, 2, 248.	4.4	15
36	Learning a genome-wide score of human–mouse conservation at the functional genomics level. Nature Communications, 2021, 12, 2495.	12.8	12

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37	Integration of evidence across human and model organism studies: A meeting report. Genes, Brain and Behavior, 2021, 20, e12738.	2.2	12
38	ChromTime: modeling spatio-temporal dynamics of chromatin marks. Genome Biology, 2018, 19, 109.	8.8	10
39	IL-3 and Oncogenic Abl Regulate the Myeloblast Transcriptome by Altering mRNA Stability. PLoS ONE, 2009, 4, e7469.	2.5	9
40	Mapping enhancer and promoter interactions. Cell Research, 2012, 22, 789-790.	12.0	7
41	ConsHMM Atlas: conservation state annotations for major genomes and human genetic variation. NAR Genomics and Bioinformatics, 2020, 2, Iqaa104.	3.2	4
42	Investigating enhancer evolution with massively parallel reporter assays. Genome Biology, 2018, 19, 114.	8.8	2
43	Single-nucleotide conservation state annotation of the SARS-CoV-2 genome. Communications Biology, 2021, 4, 698.	4.4	2
44	Identification and characterization of constrained non-exonic bases lacking predictive epigenomic and transcription factor binding annotations. Nature Communications, 2020, 11, 6168.	12.8	1
45	Marginal Contribution Feature Importance - an Axiomatic Approach for Explaining Data. Proceedings of Machine Learning Research, 2021, 139, 1324-1335.	0.3	1