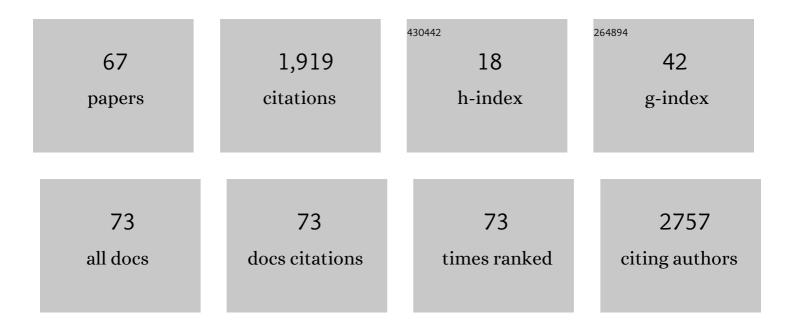
Cody Ashby

List of Publications by Year in descending order

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CODY ASHRY

#	Article	IF	CITATIONS
1	A high-risk, Double-Hit, group of newly diagnosed myeloma identified by genomic analysis. Leukemia, 2019, 33, 159-170.	3.3	313
2	Spatial genomic heterogeneity in multiple myeloma revealed by multi-region sequencing. Nature Communications, 2017, 8, 268.	5.8	277
3	Bridger: a new framework for de novo transcriptome assembly using RNA-seq data. Genome Biology, 2015, 16, 30.	3.8	258
4	Clonal selection and double-hit events involving tumor suppressor genes underlie relapse in myeloma. Blood, 2016, 128, 1735-1744.	0.6	170
5	Combination of flow cytometry and functional imaging for monitoring of residual disease in myeloma. Leukemia, 2019, 33, 1713-1722.	3.3	112
6	Revealing the Impact of Structural Variants in Multiple Myeloma. Blood Cancer Discovery, 2020, 1, 258-273.	2.6	81
7	Clonal evolution in myeloma: the impact of maintenance lenalidomide and depth of response on the genetics and sub-clonal structure of relapsed disease in uniformly treated newly diagnosed patients. Haematologica, 2019, 104, 1440-1450.	1.7	67
8	The level of deletion 17p and bi-allelic inactivation of <i>TP53</i> has a significant impact on clinical outcome in multiple myeloma. Haematologica, 2017, 102, e364-e367.	1.7	57
9	The molecular make up of smoldering myeloma highlights the evolutionary pathways leading to multiple myeloma. Nature Communications, 2021, 12, 293.	5.8	54
10	Bi-allelic inactivation is more prevalent at relapse in multiple myeloma, identifying RB1 as an independent prognostic marker. Blood Cancer Journal, 2017, 7, e535-e535.	2.8	48
11	Microhomology-mediated end joining drives complex rearrangements and overexpression of <i>MYC</i> and <i>PVT1</i> in multiple myeloma. Haematologica, 2020, 105, 1055-1066.	1.7	42
12	Accelerated single cell seeding in relapsed multiple myeloma. Nature Communications, 2020, 11, 3617.	5.8	41
13	Loss of heterozygosity as a marker of homologous repair deficiency in multiple myeloma: a role for PARP inhibition?. Leukemia, 2018, 32, 1561-1566.	3.3	39
14	<i>BRAF</i> and <i>DIS3</i> Mutations Associate with Adverse Outcome in a Long-term Follow-up of Patients with Multiple Myeloma. Clinical Cancer Research, 2020, 26, 2422-2432.	3.2	37
15	Bone marrow microenvironments that contribute to patient outcomes in newly diagnosed multiple myeloma: A cohort study of patients in the Total Therapy clinical trials. PLoS Medicine, 2020, 17, e1003323.	3.9	33
16	Genomic analysis of primary plasma cell leukemia reveals complex structural alterations and high-risk mutational patterns. Blood Cancer Journal, 2020, 10, 70.	2.8	27
17	Chromothripsis as a pathogenic driver of multiple myeloma. Seminars in Cell and Developmental Biology, 2022, 123, 115-123.	2.3	22
18	Monitoring treatment response and disease progression in myeloma with circulating cellâ€free DNA. European Journal of Haematology, 2021, 106, 230-240.	1.1	21

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19	The functional epigenetic landscape of aberrant gene expression in molecular subgroups of newly diagnosed multiple myeloma. Journal of Hematology and Oncology, 2020, 13, 108.	6.9	20
20	Differential RNA splicing as a potentially important driver mechanism in multiple myeloma. Haematologica, 2021, 106, 736-745.	1.7	20
21	DNAp: A Pipeline for DNA-seq Data Analysis. Scientific Reports, 2018, 8, 6793.	1.6	14
22	Poor overall survival in hyperhaploid multiple myeloma is defined by double-hit bi-allelic inactivation of <i>TP53</i> . Oncotarget, 2019, 10, 732-737.	0.8	13
23	The genomic landscape of plasma cells in systemic light chain amyloidosis. Blood, 2018, 132, 2775-2777.	0.6	12
24	Mutations in CRBN and other cereblon pathway genes are infrequently associated with acquired resistance to immunomodulatory drugs. Leukemia, 2021, 35, 3017-3020.	3.3	11
25	TRIP13 modulates protein deubiquitination and accelerates tumor development and progression of B cell malignancies. Journal of Clinical Investigation, 2021, 131, .	3.9	10
26	Improving prognostic assignment in older adults with multiple myeloma using acquired genetic features, clonal hemopoiesis and telomere length. Leukemia, 2021, , .	3.3	8
27	Epigenomic translocation of H3K4me3 broad domains over oncogenes following hijacking of super-enhancers. Genome Research, 2022, 32, 1343-1354.	2.4	8
28	eMBI: Boosting Gene Expression-based Clustering for Cancer Subtypes. Cancer Informatics, 2014, 13s2, CIN.S13777.	0.9	7
29	FiNGS: high quality somatic mutations using filters for next generation sequencing. BMC Bioinformatics, 2021, 22, 77.	1.2	7
30	Structural variants shape the genomic landscape and clinical outcome of multiple myeloma. Blood Cancer Journal, 2022, 12, .	2.8	7
31	New enumeration algorithm for protein structure comparison and classification. BMC Genomics, 2013, 14, S1.	1.2	6
32	Enhancing cancer clonality analysis with integrative genomics. BMC Bioinformatics, 2015, 16, S7.	1.2	6
33	SPARCoC: A New Framework for Molecular Pattern Discovery and Cancer Gene Identification. PLoS ONE, 2015, 10, e0117135.	1.1	6
34	Plasma cells expression from smouldering myeloma to myeloma reveals the importance of the PRC2 complex, cell cycle progression, and the divergent evolutionary pathways within the different molecular subgroups. Leukemia, 2022, 36, 591-595.	3.3	6
35	Molecular characterization and clinical outcome of B-cell precursor acute lymphoblastic leukemia with IG-MYC rearrangement. Haematologica, 2023, 108, 717-731.	1.7	6
36	Myeloma Genome Project Panel is a Comprehensive Targeted Genomics Panel for Molecular Profiling of Patients with Multiple Myeloma. Clinical Cancer Research, 2022, 28, 2854-2864.	3.2	6

Соду Азнву

#	Article	IF	CITATIONS
37	Chromoplexy and Chromothripsis Are Important Prognostically in Myeloma and Deregulate Gene Function By a Range of Mechanisms. Blood, 2019, 134, 3767-3767.	0.6	5
38	Notch3 signaling between myeloma cells and osteocytes in the tumor niche promotes tumor growth and bone destruction. Neoplasia, 2022, 28, 100785.	2.3	5
39	Highâ€risk transcriptional profiles in multiple myeloma are an acquired feature that can occur in any subtype and more frequently with each subsequent relapse. British Journal of Haematology, 2021, 195, 283-286.	1.2	4
40	Baseline and on-Treatment Bone Marrow Microenvironments Predict Myeloma Patient Outcomes and Inform Potential Intervention Strategies. Blood, 2018, 132, 1882-1882.	0.6	3
41	Chromothripsis and Chromoplexy Are Associated with DNA Instability and Adverse Clinical Outcome in Multiple Myeloma. Blood, 2018, 132, 408-408.	0.6	3
42	TarPan: an easily adaptable targeted sequencing panel viewer for research and clinical use. BMC Bioinformatics, 2020, 21, 144.	1.2	2
43	The Mutational Landscape of Primary Plasma Cell Leukemia. Blood, 2018, 132, 114-114.	0.6	2
44	A High-Risk Multiple Myeloma Group Identified By Integrative Multi-Omics Segmentation of Newly Diagnosed Patients. Blood, 2018, 132, 3165-3165.	0.6	2
45	Analysis of the Sub-Clonal Structure of Smoldering Myeloma over Time Provides a New Means of Disease Monitoring and Highlights Evolutionary Trajectories Leading to Myeloma. Blood, 2019, 134, 4333-4333.	0.6	2
46	The Spectrum of Exomic Mutation in Elderly Myeloma Differs Substantially from Patients at Younger Ages Consistent with a Different Evolutionary Trajectory to Full Blown Disease Based on Age of Onset. Blood, 2019, 134, 4346-4346.	0.6	2
47	Extensive Regional Intra-Clonal Heterogeneity in Multiple Myeloma - Implications for Diagnostics, Risk Stratification and Targeted Treatment. Blood, 2016, 128, 3278-3278.	0.6	2
48	The Multiple Myeloma Genome Project: Development of a Molecular Segmentation Strategy for the Clinical Classification of Multiple Myeloma. Blood, 2016, 128, 196-196.	0.6	2
49	Multiple Myeloma with a Deletion of Chromosome 17p: TP53 Mutations Are Highly Prevalent and Negatively Affect Prognosis. Blood, 2016, 128, 3271-3271.	0.6	2
50	Influence of Aging Processes on the Biology and Outcome of Multiple Myeloma. Blood, 2020, 136, 8-9.	0.6	2
51	A detailed exploration of using RNA-Seq data in established multiple myeloma gene expression profile microarray based risk scores. Clinical Lymphoma, Myeloma and Leukemia, 2019, 19, e57-e58.	0.2	1
52	Long-Term Follow-up Identifies Double Hit and Key Mutations As Impacting Progression Free and Overall Survival in Multiple Myeloma. Blood, 2018, 132, 110-110.	0.6	1
53	High Risk Myeloma Is Characterized By the Bi-Allelic Inactivation of CDKN2C and RB1. Blood, 2016, 128, 4416-4416.	0.6	1
54	Extracting Prognostic Molecular Information from PET-CT Imaging of Multiple Myeloma Using Radiomic Approaches. Blood, 2018, 132, 1906-1906.	0.6	1

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55	Epigenetic Deregulation of Telomere-Related Genes in Newly Diagnosed Multiple Myeloma Patients. Cancers, 2021, 13, 6348.	1.7	1
56	Long-term Analysis Of Multiple Sequential Samples Reveals Patterns Of Progression In Smoldering Myeloma. Clinical Lymphoma, Myeloma and Leukemia, 2019, 19, e59-e60.	0.2	0
57	Global Expression Changes of Malignant Plasma Cells over Time Reveals the Evolutionary Development of Signatures of Aggressive Clinical Behavior. Blood, 2018, 132, 4457-4457.	0.6	0
58	Poor Overall Survival in Hyperhaploid Multiple Myeloma Is Defined By Double-Hit Bi-Allelic Inactivation of TP53. Blood, 2018, 132, 4441-4441.	0.6	0
59	Mutations and Copy Number Changes Predict Progression from Smoldering Myeloma to Symptomatic Myeloma in the Era of Novel IMWG Criteria. Blood, 2018, 132, 4456-4456.	0.6	Ο
60	Global 3D-Epigenetic Dysregulation of Cyclin D1 and D2 Actively Controls Their Expression Pattern in Multiple Myeloma. Blood, 2018, 132, 3904-3904.	0.6	0
61	Combination of Flow Cytometry and Functional Imaging for Monitoring of Residual Disease in Myeloma. Blood, 2018, 132, 3185-3185.	0.6	0
62	Hotspot Mutations in SF3B1 Result in Increased Alternative Splicing in Multiple Myeloma and Activation of Key Cellular Pathways. Blood, 2018, 132, 4454-4454.	0.6	0
63	High Levels of APOBEC3B Gene Expression Contribute to Poor Prognosis in Multiple Myeloma Patients. Blood, 2018, 132, 3897-3897.	0.6	0
64	Proliferation and Molecular Risk Score of Low Risk Myeloma Cells Are Increased in High Risk Microenvironment Via Augmented Bioavailability of Growth Factors. Blood, 2018, 132, 1929-1929.	0.6	0
65	Example of a Scalable and Adaptable Approach for NGS Analyses Leveraging High-Performance Computing. , 2019, , 247-269.		0
66	The mTOR Component, Rictor, Is Regulated By the Microenvironment to Control Dormancy and Proliferative States in Myeloma Cells. Blood, 2019, 134, 4412-4412.	0.6	0
67	Mutations in <i>CRBN</i> and Other Cereblon Pathway Genes Are Only Associated with Acquired Resistance to Immunomodulatory Drugs in a Subset of Patients and Cell Line Models. Blood, 2020, 136, 6-7.	0.6	0