## Andrea Cerioli

List of Publications by Year in descending order

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Version: 2024-02-01

57	1,139	19	32
papers	citations	h-index	g-index
60	60	60	465
all docs	docs citations	times ranked	citing authors

#	Article	IF	CITATIONS
1	On Characterizations and Tests of Benford's Law. Journal of the American Statistical Association, 2022, 117, 1887-1903.	3.1	8
2	Covariance matrices of S robust regression estimators. Journal of Statistical Computation and Simulation, 2022, 92, 724-747.	1.2	0
3	Editorial for ADAC issue 2 of volume 15 (2021). Advances in Data Analysis and Classification, 2021, 15, 261-265.	1.4	O
4	Editorial for ADAC issue 3 of volume 15 (2021). Advances in Data Analysis and Classification, 2021, 15, 543-546.	1.4	0
5	Editorial for ADAC issue 4 of volume 15 (2021). Advances in Data Analysis and Classification, 2021, 15, 825.	1.4	O
6	Forum on Benford's law and statistical methods for the detection of frauds. Statistical Methods and Applications, 2021, 30, 767-778.	1.2	6
7	Editorial for ADAC issue 3 of volume 14 (2020). Advances in Data Analysis and Classification, 2020, 14, 513-515.	1.4	0
8	Wild adaptive trimming for robust estimation and cluster analysis. Scandinavian Journal of Statistics, 2019, 46, 235-256.	1.4	21
9	Assessing trimming methodologies for clustering linear regression data. Advances in Data Analysis and Classification, 2019, 13, 227-257.	1.4	9
10	Comments on: Data science, big data and statistics. Test, 2019, 28, 349-352.	1.1	3
10	Comments on: Data science, big data and statistics. Test, 2019, 28, 349-352.  Newcomb–Benford law and the detection of frauds in international trade. Proceedings of the National Academy of Sciences of the United States of America, 2019, 116, 106-115.	7.1	46
	Newcomb–Benford law and the detection of frauds in international trade. Proceedings of the		
11	Newcomb–Benford law and the detection of frauds in international trade. Proceedings of the National Academy of Sciences of the United States of America, 2019, 116, 106-115.  Efficient robust methods via monitoring for clustering and multivariate data analysis. Pattern	7.1	46
11 12	Newcomb–Benford law and the detection of frauds in international trade. Proceedings of the National Academy of Sciences of the United States of America, 2019, 116, 106-115.  Efficient robust methods via monitoring for clustering and multivariate data analysis. Pattern Recognition, 2019, 88, 246-260.  The power of monitoring: how to make the most of a contaminated multivariate sample. Statistical	7.1	10
11 12 13	Newcomb–Benford law and the detection of frauds in international trade. Proceedings of the National Academy of Sciences of the United States of America, 2019, 116, 106-115.  Efficient robust methods via monitoring for clustering and multivariate data analysis. Pattern Recognition, 2019, 88, 246-260.  The power of monitoring: how to make the most of a contaminated multivariate sample. Statistical Methods and Applications, 2018, 27, 559-587.  Goodness-of-Fit Testing for the Newcomb-Benford Law With Application to the Detection of Customs	7.1 8.1 1.2	46 10 29
11 12 13 14	Newcomb–Benford law and the detection of frauds in international trade. Proceedings of the National Academy of Sciences of the United States of America, 2019, 116, 106-115.  Efficient robust methods via monitoring for clustering and multivariate data analysis. Pattern Recognition, 2019, 88, 246-260.  The power of monitoring: how to make the most of a contaminated multivariate sample. Statistical Methods and Applications, 2018, 27, 559-587.  Goodness-of-Fit Testing for the Newcomb-Benford Law With Application to the Detection of Customs Fraud. Journal of Business and Economic Statistics, 2018, 36, 346-358.  Cluster detection and clustering with random start forward searches. Journal of Applied Statistics,	7.1 8.1 1.2 2.9	46 10 29 25
11 12 13 14	Newcomb–Benford law and the detection of frauds in international trade. Proceedings of the National Academy of Sciences of the United States of America, 2019, 116, 106-115.  Efficient robust methods via monitoring for clustering and multivariate data analysis. Pattern Recognition, 2019, 88, 246-260.  The power of monitoring: how to make the most of a contaminated multivariate sample. Statistical Methods and Applications, 2018, 27, 559-587.  Goodness-of-Fit Testing for the Newcomb-Benford Law With Application to the Detection of Customs Fraud. Journal of Business and Economic Statistics, 2018, 36, 346-358.  Cluster detection and clustering with random start forward searches. Journal of Applied Statistics, 2018, 45, 777-798.  Finding the Number of Normal Groups in Model-Based Clustering via Constrained Likelihoods. Journal	7.1 8.1 1.2 2.9	46 10 29 25

#	Article	IF	CITATIONS
19	A new family of tempered distributions. Electronic Journal of Statistics, 2016, 10, .	0.7	11
20	Discussion of †Asymptotic Theory of Outlier Detection Algorithms for Linear Time Series Regression Models' by Johansen and Nielsen. Scandinavian Journal of Statistics, 2016, 43, 349-352.	1.4	1
21	Reliable Robust Regression Diagnostics. International Statistical Review, 2016, 84, 99-127.	1.9	9
22	Modeling international trade data with the Tweedie distribution for anti-fraud and policy support. European Journal of Operational Research, 2016, 248, 1031-1043.	<b>5.</b> 7	15
23	How to Marry Robustness and Applied Statistics. , 2016, , 51-64.		1
24	Simulating mixtures of multivariate data with fixed cluster overlap in FSDA library. Advances in Data Analysis and Classification, 2015, 9, 461-481.	1.4	19
25	Finding the Number of Disparate Clusters with Background Contamination. Studies in Classification, Data Analysis, and Knowledge Organization, 2015, , 29-42.	0.2	2
26	The Forward Search for Very Large Datasets. Journal of Statistical Software, 2015, 67, .	3.7	7
27	Monitoring robust regression. Electronic Journal of Statistics, 2014, 8, .	0.7	35
28	Robust clustering around regression lines with high density regions. Advances in Data Analysis and Classification, 2014, 8, 5-26.	1.4	25
29	On consistency factors and efficiency of robust S-estimators. Test, 2014, 23, 356-387.	1.1	29
30	Strong consistency and robustness of the Forward Search estimator of multivariate location and scatter. Journal of Multivariate Analysis, 2014, 126, 167-183.	1.0	29
31	Robust distances for outlier-free goodness-of-fit testing. Computational Statistics and Data Analysis, 2013, 65, 29-45.	1.2	26
32	Size and Power of Multivariate Outlier Detection Rules. Studies in Classification, Data Analysis, and Knowledge Organization, 2013, , 3-17.	0.2	6
33	Robustness Issues in Text Mining. Advances in Intelligent Systems and Computing, 2013, , 263-272.	0.6	4
34	Problems and Challenges in the Analysis of Complex Data: Static and Dynamic Approaches. , 2012, , 145-157.		4
35	Error rates for multivariate outlier detection. Computational Statistics and Data Analysis, 2011, 55, 544-553.	1.2	52
36	Some Perspectives on Multivariate Outlier Detection. Studies in Classification, Data Analysis, and Knowledge Organization, 2011, , 231-238.	0.2	2

#	Article	IF	CITATIONS
37	Special Issue on Robust Methods for Classification and Data Analysis. Advances in Data Analysis and Classification, 2010, 4, 85-87.	1.4	O
38	The forward search: Theory and data analysis. Journal of the Korean Statistical Society, 2010, 39, 117-134.	0.4	66
39	Reply to discussion of "The Forward Search: Theory and data analysis― Journal of the Korean Statistical Society, 2010, 39, 161-163.	0.4	3
40	Multivariate Outlier Detection With High-Breakdown Estimators. Journal of the American Statistical Association, 2010, 105, 147-156.	3.1	105
41	Robust Clustering for Performance Evaluation. Studies in Classification, Data Analysis, and Knowledge Organization, 2010, , 381-390.	0.2	1
42	Controlling the size of multivariate outlier tests with the MCD estimator of scatter. Statistics and Computing, 2009, 19, 341-353.	1.5	30
43	Finding an Unknown Number of Multivariate Outliers. Journal of the Royal Statistical Society Series B: Statistical Methodology, 2009, 71, 447-466.	2.2	134
44	Monitoring Random Start Forward Searches for Multivariate Data. , 2008, , 447-458.		0
45	Special issue on robust multivariate analysis and classification. Statistical Methods and Applications, 2007, 15, 267-269.	1.2	0
46	Automatic Classification of Functional Data with Extremal Information. Studies in Classification, Data Analysis, and Knowledge Organization, 2007, , 99-106.	0.2	0
47	Random Start Forward Searches with Envelopes for Detecting Clusters in Multivariate Data. , 2006, , 163-171.		11
48	Robust Transformations and Outlier Detection with Autocorrelated Data., 2006,, 262-269.		0
49	Functional Cluster Analysis of Financial Time Series. , 2005, , 333-341.		2
50	Exploring Multivariate Data with the Forward Search. Springer Series in Statistics, 2004, , .	0.9	141
51	Robust methods for the analysis of spatially autocorrelated data. Statistical Methods and Applications, 2002, 11, 335-358.	1.2	12
52	Tests of homogeneity for spatial populations. Statistics and Probability Letters, 2002, 58, 123-130.	0.7	4
53	Testing Mutual Independence Between Two Discrete-Valued Spatial Processes: A Correction to Pearson Chi-Squared. Biometrics, 2002, 58, 888-897.	1.4	14
54	The Ordering of Spatial Data and the Detection of Multiple Outliers. Journal of Computational and Graphical Statistics, 1999, 8, 239-258.	1.7	22

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55	The Ordering of Spatial Data and the Detection of Multiple Outliers. Journal of Computational and Graphical Statistics, 1999, 8, 239.	1.7	22
56	Modified Tests of Independence in $2 \times 2$ Tables with Spatial Data. Biometrics, 1997, 53, 619.	1.4	55
57	Comparing three partitions: an inferential approach based on multi-way contingency tables. Communications in Statistics - Theory and Methods, 1997, 26, 2457-2471.	1.0	1