Jean Paul Barddal

List of Publications by Year in descending order

Source: https://exaly.com/author-pdf/2360389/publications.pdf

Version: 2024-02-01

42 papers 1,338 citations

759233 12 h-index 24 g-index

43 all docs

43 docs citations 43 times ranked 1022 citing authors

| # | Article | IF | CITATIONS |
|----|--|------|-----------|
| 1 | Hierarchical classification of data streams: a systematic literature review. Artificial Intelligence Review, 2022, 55, 3243-3282. | 15.7 | 4 |
| 2 | A systematic review on computer vision-based parking lot management applied on public datasets. Expert Systems With Applications, 2022, 198, 116731. | 7.6 | 18 |
| 3 | Dynamically Selected Ensemble for Data Stream Classification. , 2021, , . | | 1 |
| 4 | A case study of batch and incremental recommender systems in supermarket data under concept drifts and cold start. Expert Systems With Applications, 2021, 176, 114890. | 7.6 | 12 |
| 5 | Interactive Process Drift Detection Framework. Lecture Notes in Computer Science, 2021, , 192-204. | 1.3 | 1 |
| 6 | Classifying Potentially Unbounded Hierarchical Data Streams with Incremental Gaussian Naive Bayes. Lecture Notes in Computer Science, 2021, , 421-436. | 1.3 | 1 |
| 7 | Adaptive Global k-Nearest Neighbors for Hierarchical Classification of Data Streams [*] ., 2021,,. | | O |
| 8 | Lessons learned from data stream classification applied to credit scoring. Expert Systems With Applications, 2020, 162, 113899. | 7.6 | 28 |
| 9 | Regularized and incremental decision trees for data streams. Annales Des Telecommunications/Annals of Telecommunications, 2020, 75, 493-503. | 2.5 | 1 |
| 10 | Cost-sensitive learning for imbalanced data streams. , 2020, , . | | 14 |
| 11 | Combining Slow and Fast Learning for Improved Credit Scoring. , 2020, , . | | O |
| 12 | Na $	ilde{A}^{\scriptscriptstyle-}$ ve Approaches to Deal With Concept Drifts. , 2020, , . | | 2 |
| 13 | Decision tree-based feature ranking in concept drifting data streams. , 2019, , . | | 4 |
| 14 | Addressing feature drift in data streams using iterative subset selection. ACM SIGAPP Applied Computing Review: A Publication of the Special Interest Group on Applied Computing, 2019, 19, 20-33. | 0.9 | 4 |
| 15 | Boosting decision stumps for dynamic feature selection on data streams. Information Systems, 2019, 83, 13-29. | 3.6 | 24 |
| 16 | Machine learning for streaming data. SIGKDD Explorations: Newsletter of the Special Interest Group (SIG) on Knowledge Discovery & Data Mining, 2019, 21, 6-22. | 4.0 | 137 |
| 17 | Vertical and Horizontal Partitioning in Data Stream Regression Ensembles. , 2019, , . | | 4 |
| 18 | Merit-guided dynamic feature selection filter for data streams. Expert Systems With Applications, 2019, 116, 227-242. | 7.6 | 28 |

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|----|--|------|-----------|
| 19 | Learning regularized hoeffding trees from data streams. , 2019, , . | | 7 |
| 20 | On Social Network-Based Algorithms for Data Stream Clustering. Studies in Big Data, 2019, , 297-317. | 1.1 | 1 |
| 21 | A Survey on Ensemble Learning for Data Stream Classification. ACM Computing Surveys, 2018, 50, 1-36. | 23.0 | 342 |
| 22 | An Experimental Perspective on Sampling Methods for Imbalanced Learning From Financial Databases. , 2018, , . | | 0 |
| 23 | Are fintechs really a hype? A machine learning-based polarity analysis of Brazilian posts on social media. , 2018, , . | | 3 |
| 24 | Iterative subset selection for feature drifting data streams. , 2018, , . | | 11 |
| 25 | A survey on feature drift adaptation: Definition, benchmark, challenges and future directions. Journal of Systems and Software, 2017, 127, 278-294. | 4.5 | 76 |
| 26 | Adaptive random forests for evolving data stream classification. Machine Learning, 2017, 106, 1469-1495. | 5.4 | 415 |
| 27 | Improving Credit Risk Prediction in Online Peer-to-Peer (P2P) Lending Using Imbalanced Learning Techniques. , 2017, , . | | 9 |
| 28 | A benchmark of classifiers on feature drifting data streams. , 2016, , . | | 3 |
| 29 | Overcoming feature drifts via dynamic feature weighted k-nearest neighbor learning. , 2016, , . | | 2 |
| 30 | Towards emotion-based reputation guessing learning agents. , 2016, , . | | 2 |
| 31 | On Dynamic Feature Weighting for Feature Drifting Data Streams. Lecture Notes in Computer Science, 2016, , 129-144. | 1.3 | 21 |
| 32 | SNCStream+: Extending a high quality true anytime data stream clustering algorithm. Information Systems, 2016, 62, 60-73. | 3.6 | 15 |
| 33 | Advances on Concept Drift Detection in Regression Tasks Using Social Networks Theory. International Journal of Natural Computing Research, 2015, 5, 26-41. | 0.5 | 7 |
| 34 | Analyzing the Impact of Feature Drifts in Streaming Learning. Lecture Notes in Computer Science, 2015, , 21-28. | 1.3 | 8 |
| 35 | On the Discovery of Time Distance Constrained Temporal Association Rules. Lecture Notes in Computer Science, 2015, , 510-519. | 1.3 | 1 |
| 36 | A Complex Network-Based Anytime Data Stream Clustering Algorithm. Lecture Notes in Computer Science, 2015, , 615-622. | 1.3 | 2 |

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|----|--|------|-----------|
| 37 | A Survey on Feature Drift Adaptation. , 2015, , . | | 12 |
| 38 | Trust and Reputation Models for Multiagent Systems. ACM Computing Surveys, 2015, 48, 1-42. | 23.0 | 53 |
| 39 | SNCStream., 2015, , . | | 20 |
| 40 | Pairwise combination of classifiers for ensemble learning on data streams. , 2015, , . | | 4 |
| 41 | SFNClassifier., 2014, , . | | 13 |
| 42 | LEARNING DRIFTING NEGOTIATIONS. Applied Artificial Intelligence, 2007, 21, 861-881. | 3.2 | 17 |