

Luca Oneto

List of Publications by Year in descending order

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Version: 2024-02-01

141
papers

3,490
citations

236612

25
h-index

161609

54
g-index

149
all docs

149
docs citations

149
times ranked

3146
citing authors

| # | ARTICLE | IF | CITATIONS |
|----|---|-----|-----------|
| 1 | Human Activity Recognition on Smartphones Using a Multiclass Hardware-Friendly Support Vector Machine. Lecture Notes in Computer Science, 2012, , 216-223. | 1.0 | 540 |
| 2 | Transition-Aware Human Activity Recognition Using Smartphones. Neurocomputing, 2016, 171, 754-767. | 3.5 | 502 |
| 3 | Technical analysis and sentiment embeddings for market trend prediction. Expert Systems With Applications, 2019, 135, 60-70. | 4.4 | 175 |
| 4 | Big Data Analytics in the Cloud: Spark on Hadoop vs MPI/OpenMP on Beowulf. Procedia Computer Science, 2015, 53, 121-130. | 1.2 | 147 |
| 5 | Data-driven ship digital twin for estimating the speed loss caused by the marine fouling. Ocean Engineering, 2019, 186, 106063. | 1.9 | 131 |
| 6 | Vessels fuel consumption forecast and trim optimisation: A data analytics perspective. Ocean Engineering, 2017, 130, 351-370. | 1.9 | 127 |
| 7 | Condition Based Maintenance in Railway Transportation Systems Based on Big Data Streaming Analysis. Procedia Computer Science, 2015, 53, 437-446. | 1.2 | 98 |
| 8 | In-Sample and Out-of-Sample Model Selection and Error Estimation for Support Vector Machines. IEEE Transactions on Neural Networks and Learning Systems, 2012, 23, 1390-1406. | 7.2 | 95 |
| 9 | Statistical Learning Theory and ELM for Big Social Data Analysis. IEEE Computational Intelligence Magazine, 2016, 11, 45-55. | 3.4 | 88 |
| 10 | Train Delay Prediction Systems: A Big Data Analytics Perspective. Big Data Research, 2018, 11, 54-64. | 2.6 | 85 |
| 11 | Dynamic Delay Predictions for Large-Scale Railway Networks: Deep and Shallow Extreme Learning Machines Tuned via Thresholdout. IEEE Transactions on Systems, Man, and Cybernetics: Systems, 2017, 47, 2754-2767. | 5.9 | 72 |
| 12 | Model selection for support vector machines: Advantages and disadvantages of the Machine Learning Theory. , 2010, , . | | 62 |
| 13 | Condition-Based Maintenance of Naval Propulsion Systems with supervised Data Analysis. Ocean Engineering, 2018, 149, 268-278. | 1.9 | 57 |
| 14 | Machine learning approaches for improving condition-based maintenance of naval propulsion plants. Proceedings of the Institution of Mechanical Engineers Part M: Journal of Engineering for the Maritime Environment, 2016, 230, 136-153. | 0.3 | 53 |
| 15 | Condition-based maintenance of naval propulsion systems: Data analysis with minimal feedback. Reliability Engineering and System Safety, 2018, 177, 12-23. | 5.1 | 50 |
| 16 | Determining the most influential human factors in maritime accidents: A data-driven approach. Ocean Engineering, 2020, 211, 107588. | 1.9 | 50 |
| 17 | Human Activity Recognition on Smartphones with Awareness of Basic Activities and Postural Transitions. Lecture Notes in Computer Science, 2014, , 177-184. | 1.0 | 45 |
| 18 | Tikhonov, Ivanov and Morozov regularization for support vector machine learning. Machine Learning, 2016, 103, 103-136. | 3.4 | 44 |

| # | ARTICLE | IF | CITATIONS |
|----|---|-----|-----------|
| 19 | Fairness in Machine Learning. <i>Studies in Computational Intelligence</i> , 2020, , 155-196. | 0.7 | 42 |
| 20 | A Learning Analytics Approach to Correlate the Academic Achievements of Students with Interaction Data from an Educational Simulator. <i>Lecture Notes in Computer Science</i> , 2015, , 352-366. | 1.0 | 37 |
| 21 | Taking Advantage of Multitask Learning for Fair Classification. , 2019, , . | | 32 |
| 22 | Fully Empirical and Data-Dependent Stability-Based Bounds. <i>IEEE Transactions on Cybernetics</i> , 2015, 45, 1913-1926. | 6.2 | 29 |
| 23 | Model Selection and Error Estimation in a Nutshell. <i>Modeling and Optimization in Science and Technologies</i> , 2020, , . | 0.7 | 28 |
| 24 | Data-Driven Photovoltaic Power Production Nowcasting and Forecasting for Polygeneration Microgrids. <i>IEEE Systems Journal</i> , 2018, 12, 2842-2853. | 2.9 | 27 |
| 25 | Unsupervised Deep Learning for Induction Motor Bearings Monitoring. <i>Data-Enabled Discovery and Applications</i> , 2019, 3, 1. | 1.2 | 27 |
| 26 | Local Rademacher Complexity: Sharper risk bounds with and without unlabeled samples. <i>Neural Networks</i> , 2015, 65, 115-125. | 3.3 | 25 |
| 27 | Advanced Analytics for Train Delay Prediction Systems by Including Exogenous Weather Data. , 2016, , . | | 25 |
| 28 | A novelty detection approach to diagnosing hull and propeller fouling. <i>Ocean Engineering</i> , 2019, 176, 65-73. | 1.9 | 25 |
| 29 | Predicting the cavitating marine propeller noise at design stage: A deep learning based approach. <i>Ocean Engineering</i> , 2020, 209, 107481. | 1.9 | 24 |
| 30 | Marine dual fuel engines monitoring in the wild through weakly supervised data analytics. <i>Engineering Applications of Artificial Intelligence</i> , 2021, 100, 104179. | 4.3 | 23 |
| 31 | In-sample model selection for Support Vector Machines. , 2011, , . | | 20 |
| 32 | Selecting the hypothesis space for improving the generalization ability of Support Vector Machines. , 2011, , . | | 20 |
| 33 | A Deep Connection Between the Vapnikâ€“Chervonenkis Entropy and the Rademacher Complexity. <i>IEEE Transactions on Neural Networks and Learning Systems</i> , 2014, 25, 2202-2211. | 7.2 | 20 |
| 34 | Training Computationally Efficient Smartphoneâ€“Based Human Activity Recognition Models. <i>Lecture Notes in Computer Science</i> , 2013, , 426-433. | 1.0 | 19 |
| 35 | In-sample Model Selection for Trimmed Hinge Loss Support Vector Machine. <i>Neural Processing Letters</i> , 2012, 36, 275-283. | 2.0 | 17 |
| 36 | Global Rademacher Complexity Bounds: From Slow to Fast Convergence Rates. <i>Neural Processing Letters</i> , 2016, 43, 567-602. | 2.0 | 17 |

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|----|--|-----|-----------|
| 37 | Differential privacy and generalization: Sharper bounds with applications. Pattern Recognition Letters, 2017, 89, 31-38. | 2.6 | 17 |
| 38 | Prescriptive Maintenance of Railway Infrastructure: From Data Analytics to Decision Support. , 2019, , . | | 17 |
| 39 | Low-Resource Footprint, Data-Driven Malware Detection on Android. IEEE Transactions on Sustainable Computing, 2020, 5, 213-222. | 2.2 | 17 |
| 40 | PAC-bayesian analysis of distribution dependent priors: Tighter risk bounds and stability analysis. Pattern Recognition Letters, 2016, 80, 200-207. | 2.6 | 16 |
| 41 | Semi-supervised Learning for Affective Common-Sense Reasoning. Cognitive Computation, 2017, 9, 18-42. | 3.6 | 16 |
| 42 | An improved analysis of the Rademacher data-dependent bound using its self bounding property. Neural Networks, 2013, 44, 107-111. | 3.3 | 15 |
| 43 | Model selection and error estimation without the agonizing pain. Wiley Interdisciplinary Reviews: Data Mining and Knowledge Discovery, 2018, 8, e1252. | 4.6 | 15 |
| 44 | Mining Big Data with Random Forests. Cognitive Computation, 2019, 11, 294-316. | 3.6 | 15 |
| 45 | A dynamic, interpretable, and robust hybrid data analytics system for train movements in large-scale railway networks. International Journal of Data Science and Analytics, 2020, 9, 95-111. | 2.4 | 15 |
| 46 | Learning Resource-Aware Classifiers for Mobile Devices: From Regularization to Energy Efficiency. Neurocomputing, 2015, 169, 225-235. | 3.5 | 14 |
| 47 | Identifying the Determinants of Innovation Capability With Machine Learning and Patents. IEEE Transactions on Engineering Management, 2022, 69, 2144-2154. | 2.4 | 14 |
| 48 | A local Vapnik-Chervonenkis complexity. Neural Networks, 2016, 82, 62-75. | 3.3 | 13 |
| 49 | Can machine learning explain human learning?. Neurocomputing, 2016, 192, 14-28. | 3.5 | 13 |
| 50 | Data analytics and clinical feature ranking of medical records of patients with sepsis. BioData Mining, 2021, 14, 12. | 2.2 | 13 |
| 51 | Deep fair models for complex data: Graphs labeling and explainable face recognition. Neurocomputing, 2022, 470, 318-334. | 3.5 | 13 |
| 52 | Ship efficiency forecast based on sensors data collection: Improving numerical models through data analytics. , 2015, , . | | 12 |
| 53 | Unintrusive Monitoring of Induction Motors Bearings via Deep Learning on Stator Currents. Procedia Computer Science, 2018, 144, 42-51. | 1.2 | 12 |
| 54 | Understanding Violin Players' Skill Level Based on Motion Capture: a Data-Driven Perspective. Cognitive Computation, 2020, 12, 1356-1369. | 3.6 | 12 |

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| 55 | Unlabeled patterns to tighten Rademacher complexity error bounds for kernel classifiers. Pattern Recognition Letters, 2014, 37, 210-219. | 2.6 | 11 |
| 56 | Support vector machines and strictly positive definite kernel: The regularization hyperparameter is more important than the kernel hyperparameters. , 2015, , . | | 11 |
| 57 | SLT-Based ELM for Big Social Data Analysis. Cognitive Computation, 2017, 9, 259-274. | 3.6 | 11 |
| 58 | Ensemble of Technical Analysis and Machine Learning for Market Trend Prediction. , 2018, , . | | 11 |
| 59 | General Fair Empirical Risk Minimization. , 2020, , . | | 11 |
| 60 | Physical, data-driven and hybrid approaches to model engine exhaust gas temperatures in operational conditions. Ships and Offshore Structures, 2022, 17, 1360-1381. | 0.9 | 11 |
| 61 | Vessels Fuel Consumption: A Data Analytics Perspective to Sustainability. Studies in Fuzziness and Soft Computing, 2018, , 11-48. | 0.6 | 11 |
| 62 | Towards learning trustworthily, automatically, and with guarantees on graphs: An overview. Neurocomputing, 2022, 493, 217-243. | 3.5 | 11 |
| 63 | Large-Scale Railway Networks Train Movements: A Dynamic, Interpretable, and Robust Hybrid Data Analytics System. , 2018, , . | | 10 |
| 64 | Optimizing Fuel Consumption in Thrust Allocation for Marine Dynamic Positioning Systems. IEEE Transactions on Automation Science and Engineering, 2022, 19, 122-142. | 3.4 | 10 |
| 65 | Computationally aware estimation of ultimate strength reduction of stiffened panels caused by welding residual stress: From finite element to data-driven methods. Engineering Structures, 2022, 264, 114423. | 2.6 | 10 |
| 66 | Machine learning for wear forecasting of naval assets for condition-based maintenance applications. , 2015, , . | | 9 |
| 67 | Measuring the expressivity of graph kernels through Statistical Learning Theory. Neurocomputing, 2017, 268, 4-16. | 3.5 | 9 |
| 68 | Constraint-Aware Data Analysis on Mobile Devices. , 2017, , 127-149. | | 9 |
| 69 | Investigating Timing and Impact of News on the Stock Market. , 2018, , . | | 9 |
| 70 | Ensemble Application of Transfer Learning and Sample Weighting for Stock Market Prediction. , 2019, , . | | 9 |
| 71 | Computational intelligence identifies alkaline phosphatase (ALP), alpha-fetoprotein (AFP), and hemoglobin levels as most predictive survival factors for hepatocellular carcinoma. Health Informatics Journal, 2021, 27, 146045822098420. | 1.1 | 9 |
| 72 | An Enhanced Random Forests Approach to Predict Heart Failure From Small Imbalanced Gene Expression Data. IEEE/ACM Transactions on Computational Biology and Bioinformatics, 2021, 18, 2759-2765. | 1.9 | 9 |

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| 73 | A Machine Learning Analysis of Health Records of Patients With Chronic Kidney Disease at Risk of Cardiovascular Disease. <i>IEEE Access</i> , 2021, 9, 165132-165144. | 2.6 | 9 |
| 74 | Quantum computing and supervised machine learning. , 2017, , 33-83. | | 8 |
| 75 | Model scale cavitation noise spectra prediction: Combining physical knowledge with data science. <i>Ocean Engineering</i> , 2019, 178, 185-203. | 1.9 | 8 |
| 76 | Towards Online Discovery of Data-Aware Declarative Process Models from Event Streams. , 2020, , . | | 8 |
| 77 | Toward Learning Trustworthily from Data Combining Privacy, Fairness, and Explainability: An Application to Face Recognition. <i>Entropy</i> , 2021, 23, 1047. | 1.1 | 8 |
| 78 | Numerical methods for monitoring and evaluating the biofouling state and effects on vesselsâ€™ hull and propeller performance: A review. <i>Ocean Engineering</i> , 2022, 251, 110883. | 1.9 | 8 |
| 79 | Learning With Kernels: A Local Rademacher Complexity-Based Analysis With Application to Graph Kernels. <i>IEEE Transactions on Neural Networks and Learning Systems</i> , 2018, 29, 4660-4671. | 7.2 | 7 |
| 80 | A support vector machine classifier from a bit-constrained, sparse and localized hypothesis space. , 2013, , . | | 5 |
| 81 | Delay Prediction System for Large-Scale Railway Networks Based on Big Data Analytics. <i>Advances in Intelligent Systems and Computing</i> , 2017, , 139-150. | 0.5 | 5 |
| 82 | Randomized learning: Generalization performance of old and new theoretically grounded algorithms. <i>Neurocomputing</i> , 2018, 298, 21-33. | 3.5 | 5 |
| 83 | Digital twins of the mooring line tension for floating offshore wind turbines to improve monitoring, lifespan, and safety. <i>Journal of Ocean Engineering and Marine Energy</i> , 2022, 8, 1-16. | 0.9 | 5 |
| 84 | Vessel monitoring and design in industry 4.0: A data driven perspective. , 2016, , . | | 4 |
| 85 | Marine Safety and Data Analytics: Vessel Crash Stop Maneuvering Performance Prediction. <i>Lecture Notes in Computer Science</i> , 2017, , 385-393. | 1.0 | 4 |
| 86 | Multilayer Graph Node Kernels: Stacking While Maintaining Convexity. <i>Neural Processing Letters</i> , 2018, 48, 649-667. | 2.0 | 4 |
| 87 | Crash Stop Maneuvering Performance Prediction: a Data-Driven Solution for Safety and Collision Avoidance. <i>Data-Enabled Discovery and Applications</i> , 2018, 2, 1. | 1.2 | 4 |
| 88 | Simple continuous optimal regions of the space of data. <i>Neurocomputing</i> , 2019, 349, 91-104. | 3.5 | 4 |
| 89 | Local Rademacher Complexity Machine. <i>Neurocomputing</i> , 2019, 342, 24-32. | 3.5 | 4 |
| 90 | Advances in artificial neural networks, machine learning and computational intelligence. <i>Neurocomputing</i> , 2019, 342, 1-5. | 3.5 | 4 |

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| 91 | Innovation Capability of Firms: A Big Data Approach with Patents. Proceedings of the International Neural Networks Society, 2020, , 169-179. | 0.6 | 4 |
| 92 | A Learning Analytics Methodology to Profile Students Behavior and Explore Interactions with a Digital Electronics Simulator. Lecture Notes in Computer Science, 2014, , 596-597. | 1.0 | 4 |
| 93 | Introduction to Digital Systems Design. , 2019, , . | | 4 |
| 94 | Keep it Simple: Handcrafting Feature and Tuning Random Forests and XGBoost to face the Affective Movement Recognition Challenge 2021. , 2021, , . | | 4 |
| 95 | Nested Sequential Minimal Optimization for Support Vector Machines. Lecture Notes in Computer Science, 2012, , 156-163. | 1.0 | 3 |
| 96 | Performance assessment and uncertainty quantification of predictive models for smart manufacturing systems. , 2015, , . | | 3 |
| 97 | Learning Hardware-Friendly Classifiers Through Algorithmic Stability. Transactions on Embedded Computing Systems, 2016, 15, 1-29. | 2.1 | 3 |
| 98 | ReForeSt: Random Forests in Apache Spark. Lecture Notes in Computer Science, 2017, , 331-339. | 1.0 | 3 |
| 99 | Randomized learning and generalization of fair and private classifiers: From PAC-Bayes to stability and differential privacy. Neurocomputing, 2020, 416, 231-243. | 3.5 | 3 |
| 100 | Learning fair models and representations. Intelligenza Artificiale, 2020, 14, 151-178. | 1.0 | 3 |
| 101 | Accuracy and Intrusiveness in Data-Driven Violin Players Skill Levels Prediction: MOCAP Against MYO Against KINECT. Lecture Notes in Computer Science, 2021, , 367-379. | 1.0 | 3 |
| 102 | Communication platform concept for virtual testing of novel applications for railway traffic management systems. Transportation Research Procedia, 2022, 62, 832-839. | 0.8 | 3 |
| 103 | The benefits of adversarial defense in generalization. Neurocomputing, 2022, 505, 125-141. | 3.5 | 3 |
| 104 | Advances in artificial neural networks, machine learning and computational intelligence. Neurocomputing, 2020, 416, 172-176. | 3.5 | 2 |
| 105 | Big Data Analytics for Train Delay Prediction. Advances in Civil and Industrial Engineering Book Series, 2018, , 320-348. | 0.2 | 2 |
| 106 | Digital Twin of the Mooring Line Tension for Floating Offshore Wind Turbines. , 2021, , . | | 2 |
| 107 | Physical and Data-Driven Models Hybridisation for Modelling the Dynamic State of Four-Stroke Marine Diesel Engine. Energy, Environment, and Sustainability, 2022, , 145-193. | 0.6 | 2 |
| 108 | Rademacher Complexity and Structural Risk Minimization: An Application to Human Gene Expression Datasets. Lecture Notes in Computer Science, 2012, , 491-498. | 1.0 | 1 |

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| 109 | Some results about the Vapnik-Chervonenkis entropy and the rademacher complexity. , 2013, , . | | 1 |
| 110 | Smartphone battery saving by bit-based hypothesis spaces and local Rademacher Complexities. , 2014, , . | | 1 |
| 111 | Shrinkage learning to improve SVM with hints. , 2015, , . | | 1 |
| 112 | Crack random forest for arbitrary large datasets. , 2017, , . | | 1 |
| 113 | Deep graph node kernels: A convex approach. , 2017, , . | | 1 |
| 114 | Hybrid Model for Cavitation Noise Spectra Prediction. , 2019, , . | | 1 |
| 115 | Improving Railway Maintenance Actions with Big Data and Distributed Ledger Technologies. Proceedings of the International Neural Networks Society, 2020, , 120-125. | 0.6 | 1 |
| 116 | Train Overtaking Prediction in Railway Networks: A Big Data Perspective. Proceedings of the International Neural Networks Society, 2020, , 142-151. | 0.6 | 1 |
| 117 | Deep Learning for Cavitating Marine Propeller Noise Prediction at Design Stage. , 2020, , . | | 1 |
| 118 | A Novel Procedure for Training L1-L2 Support Vector Machine Classifiers. Lecture Notes in Computer Science, 2013, , 434-441. | 1.0 | 1 |
| 119 | Natural language processing for aviation safety: extracting knowledge from publicly-available loss of separation reports. Open Research Europe, 0, 1, 110. | 2.0 | 1 |
| 120 | Out-of-Sample Error Estimation: The Blessing of High Dimensionality. , 2014, , . | | 0 |
| 121 | Fast convergence of extended Rademacher Complexity bounds. , 2015, , . | | 0 |
| 122 | Sentic Computing for Social Network Analysis. , 2017, , 71-90. | | 0 |
| 123 | Introduzione al Progetto di Sistemi a Microprocessore. , 2021, , . | | 0 |
| 124 | An Efficient Hybrid Planning Framework for In-Station Train Dispatching. Lecture Notes in Computer Science, 2021, , 168-182. | 1.0 | 0 |
| 125 | Distribution-Dependent Weighted Union Bound. Entropy, 2021, 23, 101. | 1.1 | 0 |
| 126 | Learn and Visually Explain Deep Fair Models: an Application to Face Recognition. , 2021, , . | | 0 |

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| 127 | Natural language processing for aviation safety: extracting knowledge from publicly-available loss of separation reports. Open Research Europe, 0, 1, 110. | 2.0 | 0 |
| 128 | Learning Hardware Friendly Classifiers Through Algorithmic Risk Minimization. Smart Innovation, Systems and Technologies, 2016, , 403-413. | 0.5 | 0 |
| 129 | Introduzione al Progetto di Sistemi Digitali. , 2018, , . | | 0 |
| 130 | Visual Analytics for Supporting Conflict Resolution in Large Railway Networks. Proceedings of the International Neural Networks Society, 2020, , 206-215. | 0.6 | 0 |
| 131 | Restoration Time Prediction in Large Scale Railway Networks: Big Data and Interpretability. Proceedings of the International Neural Networks Society, 2020, , 136-141. | 0.6 | 0 |
| 132 | Algorithmic Stability Theory. Modeling and Optimization in Science and Technologies, 2020, , 65-74. | 0.7 | 0 |
| 133 | Resampling Methods. Modeling and Optimization in Science and Technologies, 2020, , 25-31. | 0.7 | 0 |
| 134 | The "Five W" of MS and EE. Modeling and Optimization in Science and Technologies, 2020, , 5-11. | 0.7 | 0 |
| 135 | Complexity-Based Methods. Modeling and Optimization in Science and Technologies, 2020, , 33-57. | 0.7 | 0 |
| 136 | PAC-Bayes Theory. Modeling and Optimization in Science and Technologies, 2020, , 75-86. | 0.7 | 0 |
| 137 | Compression Bound. Modeling and Optimization in Science and Technologies, 2020, , 59-63. | 0.7 | 0 |
| 138 | Conclusions and Further Readings. Modeling and Optimization in Science and Technologies, 2020, , 99-100. | 0.7 | 0 |
| 139 | L'interfacciamento di dispositivi. , 2021, , 323-440. | | 0 |
| 140 | Sistemi a Microprocessore Su FPGA. , 2021, , 441-555. | | 0 |
| 141 | Computational prediction of underwater radiated noise of cavitating marine propellers: On the accuracy of semi-empirical models. Ocean Engineering, 2022, 259, 111477. | 1.9 | 0 |