

# RenÃ©-Vinicio SÃ¡nchez

## List of Publications by Year in descending order

Source: <https://exaly.com/author-pdf/7370181/publications.pdf>

Version: 2024-02-01

75  
papers

3,505  
citations

236833

25  
h-index

189801

50  
g-index

76  
all docs

76  
docs citations

76  
times ranked

2821  
citing authors

| #  | ARTICLE  | IF  | CITATIONS |
|----|--|-----|-----------|
| 1  | A review on data-driven fault severity assessment in rolling bearings. <i>Mechanical Systems and Signal Processing</i> , 2018, 99, 169-196.  | 4.4 | 493       |
| 2  | Gearbox fault diagnosis based on deep random forest fusion of acoustic and vibratory signals. <i>Mechanical Systems and Signal Processing</i> , 2016, 76-77, 283-293.  | 4.4 | 339       |
| 3  | Fault diagnosis in spur gears based on genetic algorithm and random forest. <i>Mechanical Systems and Signal Processing</i> , 2016, 70-71, 87-103.   | 4.4 | 248       |
| 4  | Multimodal deep support vector classification with homologous features and its application to gearbox fault diagnosis. <i>Neurocomputing</i> , 2015, 168, 119-127.   | 3.5 | 245       |
| 5  | Gearbox Fault Identification and Classification with Convolutional Neural Networks. <i>Shock and Vibration</i> , 2015, 2015, 1-10.   | 0.3 | 208       |
| 6  | Fault Diagnosis for Rotating Machinery Using Vibration Measurement Deep Statistical Feature Learning. <i>Sensors</i> , 2016, 16, 895.  | 2.1 | 189       |
| 7  | Deep neural networks-based rolling bearing fault diagnosis. <i>Microelectronics Reliability</i> , 2017, 75, 327-333.   | 0.9 | 187       |
| 8  | Rolling element bearing defect detection using the generalized synchrosqueezing transform guided by time-frequency ridge enhancement. <i>ISA Transactions</i> , 2016, 60, 274-284.                             | 3.1 | 120       |
| 9  | Attribute clustering using rough set theory for feature selection in fault severity classification of rotating machinery. <i>Expert Systems With Applications</i> , 2017, 71, 69-86.                           | 4.4 | 92        |
| 10 | Bayesian approach and time series dimensionality reduction to LSTM-based model-building for fault diagnosis of a reciprocating compressor. <i>Neurocomputing</i> , 2020, 380, 51-66.                           | 3.5 | 90        |
| 11 | Extracting repetitive transients for rotating machinery diagnosis using multiscale clustered grey infogram. <i>Mechanical Systems and Signal Processing</i> , 2016, 76-77, 157-173.                            | 4.4 | 87        |
| 12 | A Systematic Review of Fuzzy Formalisms for Bearing Fault Diagnosis. <i>IEEE Transactions on Fuzzy Systems</i> , 2019, 27, 1362-1382.  | 6.5 | 86        |
| 13 | Multi-Stage Feature Selection by Using Genetic Algorithms for Fault Diagnosis in Gearboxes Based on Vibration Signal. <i>Sensors</i> , 2015, 15, 23903-23926.  | 2.1 | 76        |
| 14 | Feature ranking for multi-fault diagnosis of rotating machinery by using random forest and KNN. <i>Journal of Intelligent and Fuzzy Systems</i> , 2018, 34, 3463-3473.   | 0.8 | 75        |
| 15 | A comparison of fuzzy clustering algorithms for bearing fault diagnosis. <i>Journal of Intelligent and Fuzzy Systems</i> , 2018, 34, 3565-3580.  | 0.8 | 74        |
| 16 | Automatic feature extraction of time-series applied to fault severity assessment of helical gearbox in stationary and non-stationary speed operation. <i>Applied Soft Computing Journal</i> , 2017, 58, 53-64. | 4.1 | 59        |
| 17 | Fault Diagnosis of Wind Turbine Gearbox Based on the Optimized LSTM Neural Network with Cosine Loss. <i>Sensors</i> , 2020, 20, 2339.  | 2.1 | 59        |
| 18 | Hierarchical feature selection based on relative dependency for gear fault diagnosis. <i>Applied Intelligence</i> , 2016, 44, 687-703.   | 3.3 | 56        |

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|----|---|-----|-----------|
| 19 | A statistical comparison of neuroclassifiers and feature selection methods for gearbox fault diagnosis under realistic conditions. <i>Neurocomputing</i> , 2016, 194, 192-206.                | 3.5 | 54        |
| 20 | Generative Adversarial Networks Selection Approach for Extremely Imbalanced Fault Diagnosis of Reciprocating Machinery. <i>IEEE Access</i> , 2019, 7, 70643-70653.                            | 2.6 | 48        |
| 21 | Observer-biased bearing condition monitoring: From fault detection to multi-fault classification. <i>Engineering Applications of Artificial Intelligence</i> , 2016, 50, 287-301.             | 4.3 | 47        |
| 22 | Fault diagnosis of spur gearbox based on random forest and wavelet packet decomposition. <i>Frontiers of Mechanical Engineering</i> , 2015, 10, 277-286.                                      | 2.5 | 46        |
| 23 | A Bayesian approach to consequent parameter estimation in probabilistic fuzzy systems and its application to bearing fault classification. <i>Knowledge-Based Systems</i> , 2017, 129, 39-60. | 4.0 | 39        |
| 24 | Fusing convolutional generative adversarial encoders for 3D printer fault detection with only normal condition signals. <i>Mechanical Systems and Signal Processing</i> , 2021, 147, 107108.  | 4.4 | 33        |
| 25 | Finite-time and fixed-time impulsive synchronization of chaotic systems. <i>Journal of the Franklin Institute</i> , 2020, 357, 11545-11557.   | 1.9 | 27        |
| 26 | A fuzzy transition based approach for fault severity prediction in helical gearboxes. <i>Fuzzy Sets and Systems</i> , 2018, 337, 52-73.   | 1.6 | 26        |
| 27 | Fuzzy determination of informative frequency band for bearing fault detection. <i>Journal of Intelligent and Fuzzy Systems</i> , 2016, 30, 3513-3525.   | 0.8 | 24        |
| 28 | Spur Gear Fault Diagnosis Using a Multilayer Gated Recurrent Unit Approach With Vibration Signal. <i>IEEE Access</i> , 2019, 7, 56880-56889.  | 2.6 | 24        |
| 29 | Vibration signal analysis using symbolic dynamics for gearbox fault diagnosis. <i>International Journal of Advanced Manufacturing Technology</i> , 2019, 104, 2195-2214.                      | 1.5 | 23        |
| 30 | Gear and bearing fault classification under different load and speed by using Poincaré plot features and SVM. <i>Journal of Intelligent Manufacturing</i> , 2022, 33, 1031-1055.              | 4.4 | 22        |
| 31 | A comparative feature analysis for gear pitting level classification by using acoustic emission, vibration and current signals. <i>IFAC-PapersOnLine</i> , 2018, 51, 346-352.                 | 0.5 | 19        |
| 32 | Using a Support Vector Machine Based Decision Stage to Improve the Fault Diagnosis on Gearboxes. <i>Computational Intelligence and Neuroscience</i> , 2019, 2019, 1-13.                       | 1.1 | 19        |
| 33 | Vibration-based gearbox fault diagnosis using deep neural networks. <i>Journal of Vibroengineering</i> , 2017, 19, 2475-2496.   | 0.5 | 18        |
| 34 | Reciprocating Compressor Multi-Fault Classification Using Symbolic Dynamics and Complex Correlation Measure. <i>Applied Sciences (Switzerland)</i> , 2020, 10, 2512.                          | 1.3 | 16        |
| 35 | Feature engineering based on ANOVA, cluster validity assessment and KNN for fault diagnosis in bearings. <i>Journal of Intelligent and Fuzzy Systems</i> , 2018, 34, 3451-3462.               | 0.8 | 15        |
| 36 | From fault detection to one-class severity discrimination of 3D printers with one-class support vector machine. <i>ISA Transactions</i> , 2021, 110, 357-367.                                 | 3.1 | 15        |

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|----|--|-----|-----------|
| 37 | One-Shot Fault Diagnosis of Three-Dimensional Printers Through Improved Feature Space Learning. IEEE Transactions on Industrial Electronics, 2021, 68, 8768-8776.                      | 5.2 | 15        |
| 38 | Multi-fault Diagnosis of Rotating Machinery by Using Feature Ranking Methods and SVM-based Classifiers. , 2017, , .  |     | 14        |
| 39 | Exploiting Generative Adversarial Networks as an Oversampling Method for Fault Diagnosis of an Industrial Robotic Manipulator. Applied Sciences (Switzerland), 2020, 10, 7712.         | 1.3 | 13        |
| 40 | AutoML for Feature Selection and Model Tuning Applied to Fault Severity Diagnosis in Spur Gearboxes. Mathematical and Computational Applications, 2022, 27, 6.                         | 0.7 | 13        |
| 41 | Rolling bearing fault diagnosis based on Deep Boltzmann machines. , 2016, , .  |     | 12        |
| 42 | Echo state network and variational autoencoder for efficient one-class learning on dynamical systems. Journal of Intelligent and Fuzzy Systems, 2018, 34, 3799-3809.                   | 0.8 | 12        |
| 43 | Finite-time leaderless consensus control of a group of Euler-Lagrangian systems with backlash nonlinearities. Journal of the Franklin Institute, 2019, 356, 9286-9301.                 | 1.9 | 11        |
| 44 | A hybrid prototype selection-based deep learning approach for anomaly detection in industrial machines. Expert Systems With Applications, 2022, 204, 117528.                           | 4.4 | 10        |
| 45 | Gearbox fault classification using dictionary sparse based representations of vibration signals. Journal of Intelligent and Fuzzy Systems, 2018, 34, 3605-3618.                        | 0.8 | 9         |
| 46 | Deep Learning-Based Gear Pitting Severity Assessment Using Acoustic Emission, Vibration and Currents Signals. , 2019, , .  |     | 9         |
| 47 | Knowledge extraction from deep convolutional neural networks applied to cyclo-stationary time-series classification. Information Sciences, 2020, 524, 1-14.                            | 4.0 | 9         |
| 48 | Gearbox Fault Diagnosis Based on a Novel Hybrid Feature Reduction Method. IEEE Access, 2018, 6, 75813-75823.   | 2.6 | 8         |
| 49 | A semi-supervised approach based on evolving clusters for discovering unknown abnormal condition patterns in gearboxes. Journal of Intelligent and Fuzzy Systems, 2018, 34, 3581-3593. | 0.8 | 8         |
| 50 | Evaluation of Time and Frequency Condition Indicators from Vibration Signals for Crack Detection in Railway Axles. Applied Sciences (Switzerland), 2020, 10, 4367.                     | 1.3 | 7         |
| 51 | Some Preliminary Results on the Comparison of FCM, GK, FCMFP and FN-DBSCAN for Bearing Fault Diagnosis. , 2017, , .  |     | 6         |
| 52 | ANOVA and Cluster Distance Based Contributions for Feature Empirical Analysis to Fault Diagnosis in Rotating Machinery. , 2017, , .  |     | 5         |
| 53 | Convolutional Neural Networks Using Fourier Transform Spectrogram to Classify the Severity of Gear Tooth Breakage. , 2018, , .   |     | 5         |
| 54 | Gear Crack Level Classification by Using KNN and Time-Domain Features from Acoustic Emission Signals Under Different Motor Speeds and Loads. , 2018, , .                               |     | 5         |

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|----|--|-----|-----------|
| 55 | Accelerometer Placement Comparison for Crack Detection in Railway Axles Using Vibration Signals and Machine Learning. , 2019, , .                                  |     | 5         |
| 56 | A hybrid heuristic algorithm for evolving models in simultaneous scenarios of classification and clustering. Knowledge and Information Systems, 2019, 61, 755-798. | 2.1 | 5         |
| 57 | Deep Ensemble-Based Classifier for Transfer Learning in Rotating Machinery Fault Diagnosis. IEEE Access, 2022, 10, 29778-29787.                                    | 2.6 | 5         |
| 58 | SOA Based Integrated Software to Develop Fault Diagnosis Models Using Machine Learning in Rotating Machinery. , 2017, , .  |     | 3         |
| 59 | Poincaré plot features from vibration signal for gearbox fault diagnosis. , 2017, , .  |     | 3         |
| 60 | A LSTM Neural Network Approach using Vibration Signals for Classifying Faults in a Gearbox. , 2019, , .  |     | 3         |
| 61 | Fast feature selection based on cluster validity index applied on data-driven bearing fault detection. , 2020, , .   |     | 3         |
| 62 | An adaptive genomic difference based genetic algorithm and its application to memetic continuous optimization. Intelligent Data Analysis, 2018, 22, 363-382.       | 0.4 | 2         |
| 63 | SOA based smartphone system for the fault detection in rotating machines. , 2020, , .  |     | 2         |
| 64 | Fault Diagnosis for Controlled Continuous Systems from a Hybrid Approach: A Case Study. , 2015, , .  |     | 1         |
| 65 | A methodological framework using statistical tests for comparing machine learning based models applied to fault diagnosis in rotating machinery. , 2016, , .       |     | 1         |
| 66 | Clustering algorithm using rough set theory for unsupervised feature selection. , 2016, , .  |     | 1         |
| 67 | GKFP: A New Fuzzy Clustering Method Applied to Bearings Diagnosis. , 2018, , .   |     | 1         |
| 68 | Multilayer Gated Recurrent Unit for Spur Gear Fault Diagnosis. , 2019, , .   |     | 1         |
| 69 | Introduction to the special issue on the VIII Latin-American Congress on Mechanical Engineering. Frontiers of Mechanical Engineering, 2015, 10, 219-220.           | 2.5 | 0         |
| 70 | Framework for Discovering Unknown Abnormal Condition Patterns in Gearboxes Using a Semi-supervised Approach. , 2017, , .   |     | 0         |
| 71 | A Dictionary Sparse Based Representation of Vibration Signals for Gearbox Fault Detection. , 2017, , .   |     | 0         |
| 72 | Influence of Accelerometer Position on Gearbox Fault Severity Classification through Evaluation of Deep Learning Models. , 2019, , .                               |     | 0         |

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|----|--|-----|-----------|
| 73 | Using the Kullback-Leibler Divergence and Kolmogorov-Smirnov Test to Select Input Sizes to the Fault Diagnosis Problem Based on a CNN Model. Learning and Nonlinear Models, 2021, 18, 16-26. | 0.2 | 0         |
| 74 | A Method for the Estimation of the Constant Load Torque by Using the Steady-State Electrical Torque Signal. , 2021, , .  |     | 0         |
| 75 | Data-Driven Gearbox Fault Severity Diagnosis Based on Concept Drift. , 2021, , .   |     | 0         |