

Jochen De Weerd

List of Publications by Year in descending order

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Version: 2024-02-01

71
papers

1,485
citations

430874

18
h-index

377865

34
g-index

77
all docs

77
docs citations

77
times ranked

780
citing authors

#	ARTICLE	IF	CITATIONS
1	Extending business failure prediction models with textual website content using deep learning. European Journal of Operational Research, 2023, 306, 348-357.	5.7	15
2	Inductive Graph Representation Learning for fraud detection. Expert Systems With Applications, 2022, 193, 116463.	7.6	15
3	A Bridging Model for Process Mining and IoT. Lecture Notes in Business Information Processing, 2022, , 98-110.	1.0	10
4	Learning uncertainty with artificial neural networks for predictive process monitoring. Applied Soft Computing Journal, 2022, 125, 109134.	7.2	11
5	tcc2vec: RFM-informed representation learning on call graphs for churn prediction. Information Sciences, 2021, 557, 270-285.	6.9	10
6	Special issue on business process intelligence. Computing (Vienna/New York), 2021, 103, 1-2.	4.8	12
7	A review of automated feedback systems for learners: Classification framework, challenges and opportunities. Computers and Education, 2021, 162, 104094.	8.3	63
8	Learning Uncertainty with Artificial Neural Networks for Improved Remaining Time Prediction of Business Processes. Lecture Notes in Computer Science, 2021, , 141-157.	1.3	2
9	Expert-driven trace clustering with instance-level constraints. Knowledge and Information Systems, 2021, 63, 1197-1220.	3.2	6
10	Supervised Conformance Checking Using Recurrent Neural Network Classifiers. Lecture Notes in Business Information Processing, 2021, , 175-187.	1.0	0
11	Process Model Forecasting Using Time Series Analysis of Event Sequence Data. Lecture Notes in Computer Science, 2021, , 47-61.	1.3	2
12	Mining Behavioral Sequence Constraints for Classification. IEEE Transactions on Knowledge and Data Engineering, 2020, 32, 1130-1142.	5.7	19
13	Churn modeling with probabilistic meta paths-based representation learning. Information Processing and Management, 2020, 57, 102052.	8.6	7
14	Recommendations for enhancing the usability and understandability of process mining in healthcare. Artificial Intelligence in Medicine, 2020, 109, 101962.	6.5	32
15	Representation Learning in Graphs for Credit Card Fraud Detection. Lecture Notes in Computer Science, 2020, , 32-46.	1.3	6
16	Conformance Checking Using Activity and Trace Embeddings. Lecture Notes in Business Information Processing, 2020, , 105-121.	1.0	2
17	Predicting student success in a blended learning environment. , 2020, , .		16
18	Mining Behavioural Patterns in Urban Mobility Sequences Using Foursquare Check-in Data from Tokyo. Studies in Computational Intelligence, 2020, , 931-943.	0.9	1

#	ARTICLE	IF	CITATIONS
19	A Comparative Study of Representation Learning Techniques for Dynamic Networks. Advances in Intelligent Systems and Computing, 2020, , 523-530.	0.6	0
20	A Comparative Study of Community Detection Techniques for Large Evolving Graphs. Communications in Computer and Information Science, 2020, , 368-384.	0.5	0
21	Scalable Mixed-Paradigm Trace Clustering using Super-Instances. , 2019, , .		6
22	Understanding Automated Feedback in Learning Processes by Mining Local Patterns. Lecture Notes in Business Information Processing, 2019, , 56-68.	1.0	5
23	A comparison of methods for link sign prediction with signed network embeddings. , 2019, , .		1
24	Predictive Process Monitoring in Operational Logistics: A Case Study in Aviation. Lecture Notes in Business Information Processing, 2019, , 250-262.	1.0	6
25	Trace Clustering. , 2019, , 1706-1711.		3
26	Discovering hidden dependencies in constraint-based declarative process models for improving understandability. Information Systems, 2018, 74, 40-52.	3.6	19
27	On the operational efficiency of different feature types for telco Churn prediction. European Journal of Operational Research, 2018, 267, 1141-1155.	5.7	29
28	Combining Temporal Aspects of Dynamic Networks with Node2Vec for a more Efficient Dynamic Link Prediction. , 2018, , .		25
29	Discovering the Impact of Studentsâ€™ Modeling Behavior on their Final Performance. Lecture Notes in Business Information Processing, 2018, , 335-350.	1.0	0
30	act2vec, trace2vec, log2vec, andÂmodel2vec: Representation Learning for Business Processes. Lecture Notes in Computer Science, 2018, , 305-321.	1.3	37
31	Dropout Prediction in MOOCs: A Comparison Between Process and Sequence Mining. Lecture Notes in Business Information Processing, 2018, , 243-255.	1.0	14
32	Trace Clustering. , 2018, , 1-6.		1
33	Fodina: A robust and flexible heuristic process discovery technique. Decision Support Systems, 2017, 100, 109-118.	5.9	84
34	Explaining clusterings of process instances. Data Mining and Knowledge Discovery, 2017, 31, 774-808.	3.7	15
35	Change visualisation: Analysing the resource and timing differences between two event logs. Information Systems, 2017, 65, 106-123.	3.6	19
36	Scalable RFM-enriched Representation Learning for Churn Prediction. , 2017, , .		4

#	ARTICLE	IF	CITATIONS
37	Multi-objective Trace Clustering: Finding More Balanced Solutions. Lecture Notes in Business Information Processing, 2017, , 49-60.	1.0	3
38	An Approach for Incorporating Expert Knowledge in Trace Clustering. Lecture Notes in Computer Science, 2017, , 561-576.	1.3	10
39	Behavioral Constraint Template-Based Sequence Classification. Lecture Notes in Computer Science, 2017, , 20-36.	1.3	4
40	Churn Prediction Using Dynamic RFM-Augmented Node2vec. Lecture Notes in Computer Science, 2017, , 122-138.	1.3	3
41	Similarity-Based Approaches for Determining the Number of Trace Clusters in Process Discovery. Lecture Notes in Computer Science, 2017, , 19-42.	1.3	5
42	Gamification of Declarative Process Models for Learning and Model Verification. Lecture Notes in Business Information Processing, 2016, , 432-443.	1.0	7
43	A Stability Assessment Framework for Process Discovery Techniques. Lecture Notes in Computer Science, 2016, , 57-72.	1.3	3
44	Automating immediate and personalized feedback taking conceptual modelling education to a next level. , 2016, , .		12
45	Improving Understandability of Declarative Process Models by Revealing Hidden Dependencies. Lecture Notes in Computer Science, 2016, , 83-98.	1.3	16
46	Mixed-Paradigm Process Modeling with Intertwined State Spaces. Business and Information Systems Engineering, 2016, 58, 19-29.	6.1	27
47	Process-mining enabled feedback: “Tell me what I did wrong” vs. “tell me how to do it right”. Computers in Human Behavior, 2016, 57, 352-376.	8.5	44
48	Revising history for cost-informed process improvement. Computing (Vienna/New York), 2016, 98, 895-921.	4.8	5
49	A Full R/I-Net Construct Lexicon for Declare Constraints. SSRN Electronic Journal, 2015, , .	0.4	5
50	Fusion Miner: Process discovery for mixed-paradigm models. Decision Support Systems, 2015, 77, 123-136.	5.9	28
51	Multi-Paradigm Process Mining: Retrieving Better Models by Combining Rules and Sequences. SSRN Electronic Journal, 2014, , .	0.4	2
52	Process mining analysis of conceptual modeling behavior of novices “ empirical study using JMermaid modeling and experimental logging environment. Computers in Human Behavior, 2014, 41, 486-503.	8.5	33
53	Perturbing event logs to identify cost reduction opportunities: A genetic algorithm-based approach. , 2014, , .		7
54	Monitoring care processes in the gynecologic oncology department. Computers in Biology and Medicine, 2014, 44, 88-96.	7.0	44

#	ARTICLE	IF	CITATIONS
55	Determining Process Model Precision and Generalization with Weighted Artificial Negative Events. IEEE Transactions on Knowledge and Data Engineering, 2014, 26, 1877-1889.	5.7	60
56	Controlled automated discovery of collections of business process models. Information Systems, 2014, 46, 85-101.	3.6	24
57	SECPI: Searching for Explanations for Clustered Process Instances. Lecture Notes in Computer Science, 2014, , 408-415.	1.3	5
58	Multi-paradigm Process Mining: Retrieving Better Models by Combining Rules and Sequences. Lecture Notes in Computer Science, 2014, , 446-453.	1.3	6
59	Getting a Grasp on Clinical Pathway Data: An Approach Based on Process Mining. Lecture Notes in Computer Science, 2013, , 22-35.	1.3	12
60	Active Trace Clustering for Improved Process Discovery. IEEE Transactions on Knowledge and Data Engineering, 2013, 25, 2708-2720.	5.7	128
61	Process Mining for the multi-faceted analysis of business processes – A case study in a financial services organization. Computers in Industry, 2013, 64, 57-67.	9.9	102
62	A comprehensive benchmarking framework (CoBeFra) for conformance analysis between procedural process models and event logs in ProM. , 2013, , .		26
63	Leveraging process discovery with trace clustering and text mining for intelligent analysis of incident management processes. , 2012, , .		25
64	A multi-dimensional quality assessment of state-of-the-art process discovery algorithms using real-life event logs. Information Systems, 2012, 37, 654-676.	3.6	169
65	Improved Artificial Negative Event Generation to Enhance Process Event Logs. Notes on Numerical Fluid Mechanics and Multidisciplinary Design, 2012, , 254-269.	0.3	8
66	A robust F-measure for evaluating discovered process models. , 2011, , .		64
67	Process discovery in event logs: An application in the telecom industry. Applied Soft Computing Journal, 2011, 11, 1697-1710.	7.2	55
68	A Critical Evaluation Study of Model-Log Metrics in Process Discovery. Lecture Notes in Business Information Processing, 2011, , 158-169.	1.0	12
69	An Improved Process Event Log Artificial Negative Event Generator. SSRN Electronic Journal, 0, , .	0.4	10
70	Neural Machine Translation for Conditional Generation of Novel Procedures. , 0, , .		1
71	Leveraging Process Discovery with Trace Clustering and Text Mining for Intelligent Analysis of Incident Management Processes. SSRN Electronic Journal, 0, , .	0.4	1