

Paris Perdikaris

List of Publications by Year in descending order

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Version: 2024-02-01

29
papers

5,743
citations

331538

21
h-index

477173

29
g-index

29
all docs

29
docs citations

29
times ranked

2450
citing authors

#	ARTICLE	IF	CITATIONS
1	When and why PINNs fail to train: A neural tangent kernel perspective. <i>Journal of Computational Physics</i> , 2022, 449, 110768.	1.9	262
2	Fast Characterization of Inducible Regions of Atrial Fibrillation Models With Multi-Fidelity Gaussian Process Classification. <i>Frontiers in Physiology</i> , 2022, 13, 757159.	1.3	3
3	Output-weighted sampling for multi-armed bandits with extreme payoffs. <i>Proceedings of the Royal Society A: Mathematical, Physical and Engineering Sciences</i> , 2022, 478, .	1.0	5
4	Improved Architectures and Training Algorithms for Deep Operator Networks. <i>Journal of Scientific Computing</i> , 2022, 92, .	1.1	17
5	Gaussian processes meet NeuralODEs: a Bayesian framework for learning the dynamics of partially observed systems from scarce and noisy data. <i>Philosophical Transactions Series A, Mathematical, Physical, and Engineering Sciences</i> , 2022, 380, .	1.6	10
6	Multiscale Modeling Meets Machine Learning: What Can We Learn?. <i>Archives of Computational Methods in Engineering</i> , 2021, 28, 1017-1037.	6.0	164
7	Deep learning of free boundary and Stefan problems. <i>Journal of Computational Physics</i> , 2021, 428, 109914.	1.9	41
8	Physics-Informed Neural Networks for Heat Transfer Problems. <i>Journal of Heat Transfer</i> , 2021, 143, .	1.2	304
9	Physics-informed machine learning. <i>Nature Reviews Physics</i> , 2021, 3, 422-440.	11.9	1,789
10	Accurate artificial boundary conditions for semi-discretized one-dimensional peridynamics. <i>Proceedings of the Royal Society A: Mathematical, Physical and Engineering Sciences</i> , 2021, 477, .	1.0	3
11	COVID-19 dynamics across the US: A deep learning study of human mobility and social behavior. <i>Computer Methods in Applied Mechanics and Engineering</i> , 2021, 382, 113891.	3.4	32
12	On the eigenvector bias of Fourier feature networks: From regression to solving multi-scale PDEs with physics-informed neural networks. <i>Computer Methods in Applied Mechanics and Engineering</i> , 2021, 384, 113938.	3.4	148
13	Learning the solution operator of parametric partial differential equations with physics-informed DeepONets. <i>Science Advances</i> , 2021, 7, eabi8605.	4.7	168
14	Understanding and Mitigating Gradient Flow Pathologies in Physics-Informed Neural Networks. <i>SIAM Journal of Scientific Computing</i> , 2021, 43, A3055-A3081.	1.3	385
15	Machine learning in cardiovascular flows modeling: Predicting arterial blood pressure from non-invasive 4D flow MRI data using physics-informed neural networks. <i>Computer Methods in Applied Mechanics and Engineering</i> , 2020, 358, 112623.	3.4	275
16	Physics-Informed Deep Neural Networks for Learning Parameters and Constitutive Relationships in Subsurface Flow Problems. <i>Water Resources Research</i> , 2020, 56, e2019WR026731.	1.7	194
17	Hydrodynamic and frictional modulation of deformations in switchable colloidal crystallites. <i>Proceedings of the National Academy of Sciences of the United States of America</i> , 2020, 117, 12700-12706.	3.3	4
18	Bayesian differential programming for robust systems identification under uncertainty. <i>Proceedings of the Royal Society A: Mathematical, Physical and Engineering Sciences</i> , 2020, 476, 20200290.	1.0	18

#	ARTICLE	IF	CITATIONS
19	Machine Learning of Space-Fractional Differential Equations. <i>SIAM Journal of Scientific Computing</i> , 2019, 41, A2485-A2509.	1.3	32
20	Multi-fidelity classification using Gaussian processes: Accelerating the prediction of large-scale computational models. <i>Computer Methods in Applied Mechanics and Engineering</i> , 2019, 357, 112602.	3.4	31
21	Adversarial uncertainty quantification in physics-informed neural networks. <i>Journal of Computational Physics</i> , 2019, 394, 136-152.	1.9	210
22	Conditional deep surrogate models for stochastic, high-dimensional, and multi-fidelity systems. <i>Computational Mechanics</i> , 2019, 64, 417-434.	2.2	29
23	Physics-constrained deep learning for high-dimensional surrogate modeling and uncertainty quantification without labeled data. <i>Journal of Computational Physics</i> , 2019, 394, 56-81.	1.9	510
24	Machine learning in drug development: Characterizing the effect of 30 drugs on the QT interval using Gaussian process regression, sensitivity analysis, and uncertainty quantification. <i>Computer Methods in Applied Mechanics and Engineering</i> , 2019, 348, 313-333.	3.4	76
25	Integrating machine learning and multiscale modeling—perspectives, challenges, and opportunities in the biological, biomedical, and behavioral sciences. <i>Npj Digital Medicine</i> , 2019, 2, 115.	5.7	319
26	Numerical Gaussian Processes for Time-Dependent and Nonlinear Partial Differential Equations. <i>SIAM Journal of Scientific Computing</i> , 2018, 40, A172-A198.	1.3	162
27	A probabilistic framework for multidisciplinary design: Application to the hydrostructural optimization of supercavitating hydrofoils. <i>International Journal for Numerical Methods in Engineering</i> , 2018, 116, 246-269.	1.5	7
28	Inferring solutions of differential equations using noisy multi-fidelity data. <i>Journal of Computational Physics</i> , 2017, 335, 736-746.	1.9	202
29	Machine learning of linear differential equations using Gaussian processes. <i>Journal of Computational Physics</i> , 2017, 348, 683-693.	1.9	343