

# Heng Xiao

## List of Publications by Year in descending order

Source: <https://exaly.com/author-pdf/3986939/publications.pdf>

Version: 2024-02-01

63  
papers

3,864  
citations

201674

27  
h-index

149698

56  
g-index

64  
all docs

64  
docs citations

64  
times ranked

2113  
citing authors

#	ARTICLE	IF	CITATIONS
1	Frame-independent vector-cloud neural network for nonlocal constitutive modeling on arbitrary grids. <i>Computer Methods in Applied Mechanics and Engineering</i> , 2022, 388, 114211.	6.6	12
2	Acoustic Inversion for Uncertainty Reduction in Reynolds-Averaged Navier–Stokes-Based Jet Noise Prediction. <i>AIAA Journal</i> , 2022, 60, 2407-2422.	2.6	3
3	Neural network–based pore flow field prediction in porous media using super resolution. <i>Physical Review Fluids</i> , 2022, 7, .	2.5	13
4	Physics-informed machine learning: case studies for weather and climate modelling. <i>Philosophical Transactions Series A, Mathematical, Physical, and Engineering Sciences</i> , 2021, 379, 20200093.	3.4	167
5	End-to-end differentiable learning of turbulence models from indirect observations. <i>Theoretical and Applied Mechanics Letters</i> , 2021, 11, 100280.	2.8	13
6	Assimilation of disparate data for enhanced reconstruction of turbulent mean flows. <i>Computers and Fluids</i> , 2021, 224, 104962.	2.5	17
7	Learning nonlocal constitutive models with neural networks. <i>Computer Methods in Applied Mechanics and Engineering</i> , 2021, 384, 113927.	6.6	20
8	Recurrent neural network for end-to-end modeling of laminar-turbulent transition. <i>Data-Centric Engineering</i> , 2021, 2, .	2.3	13
9	Toward Transition Modeling in a Hypersonic Boundary Layer at Flight Conditions. , 2020, , .		1
10	Flows over periodic hills of parameterized geometries: A dataset for data-driven turbulence modeling from direct simulations. <i>Computers and Fluids</i> , 2020, 200, 104431.	2.5	67
11	Enforcing statistical constraints in generative adversarial networks for modeling chaotic dynamical systems. <i>Journal of Computational Physics</i> , 2020, 406, 109209.	3.8	77
12	Toward a Practical Method for Hypersonic Transition Prediction Based on Stability Correlations. <i>AIAA Journal</i> , 2020, 58, 4475-4484.	2.6	16
13	Enforcing boundary conditions on physical fields in Bayesian inversion. <i>Computer Methods in Applied Mechanics and Engineering</i> , 2020, 367, 113097.	6.6	4
14	Regularized ensemble Kalman methods for inverse problems. <i>Journal of Computational Physics</i> , 2020, 416, 109517.	3.8	29
15	Evaluation of ensemble methods for quantifying uncertainties in steady-state CFD applications with small ensemble sizes. <i>Computers and Fluids</i> , 2020, 203, 104530.	2.5	15
16	Convolutional neural network for transition modeling based on linear stability theory. <i>Physical Review Fluids</i> , 2020, 5, .	2.5	20
17	Physics-informed covariance kernel for model-form uncertainty quantification with application to turbulent flows. <i>Computers and Fluids</i> , 2019, 193, 104292.	2.5	12
18	Quantification of model uncertainty in RANS simulations: A review. <i>Progress in Aerospace Sciences</i> , 2019, 108, 1-31.	12.1	228

#	ARTICLE	IF	CITATIONS
19	Reynolds-averaged Navier–Stokes equations with explicit data-driven Reynolds stress closure can be ill-conditioned. <i>Journal of Fluid Mechanics</i> , 2019, 869, 553-586.	3.4	109
20	Recent progress in augmenting turbulence models with physics-informed machine learning. <i>Journal of Hydrodynamics</i> , 2019, 31, 1153-1158.	3.2	23
21	Turbulence Modeling in the Age of Data. <i>Annual Review of Fluid Mechanics</i> , 2019, 51, 357-377.	25.0	755
22	Prediction of Reynolds stresses in high-Mach-number turbulent boundary layers using physics-informed machine learning. <i>Theoretical and Computational Fluid Dynamics</i> , 2019, 33, 1-19.	2.2	33
23	Representation of stress tensor perturbations with application in machine-learning-assisted turbulence modeling. <i>Computer Methods in Applied Mechanics and Engineering</i> , 2019, 346, 707-726.	6.6	31
24	Predictive large-eddy-simulation wall modeling via physics-informed neural networks. <i>Physical Review Fluids</i> , 2019, 4, .	2.5	149
25	Inferring tsunami flow depth and flow speed from sediment deposits based on Ensemble Kalman Filtering. <i>Geophysical Journal International</i> , 2018, 212, 646-658.	2.4	5
26	TSUFLIND-EnKF: Inversion of tsunami flow depth and flow speed from deposits with quantified uncertainties. <i>Marine Geology</i> , 2018, 396, 16-25.	2.1	13
27	Propagation of Input Uncertainty in Presence of Model-Form Uncertainty: A Multifidelity Approach for Computational Fluid Dynamics Applications. <i>ASCE-ASME Journal of Risk and Uncertainty in Engineering Systems, Part B: Mechanical Engineering</i> , 2018, 4, .	1.1	8
28	Investigating the settling dynamics of cohesive silt particles with particle-resolving simulations. <i>Advances in Water Resources</i> , 2018, 111, 406-422.	3.8	31
29	Seeing permeability from images: fast prediction with convolutional neural networks. <i>Science Bulletin</i> , 2018, 63, 1215-1222.	9.0	125
30	Physics-informed machine learning approach for augmenting turbulence models: A comprehensive framework. <i>Physical Review Fluids</i> , 2018, 3, .	2.5	309
31	Visualization of High Dimensional Turbulence Simulation Data using t-SNE. , 2017, , .		20
32	High-Mach-Number Turbulence Modeling using Machine Learning and Direct Numerical Simulation Database. , 2017, , .		3
33	A Physics-Informed Machine Learning Approach of Improving RANS Predicted Reynolds Stresses. , 2017, , .		6
34	Realistic representation of grain shapes in CFD–DEM simulations of sediment transport with a bonded-sphere approach. <i>Advances in Water Resources</i> , 2017, 107, 421-438.	3.8	24
35	An Implicitly Consistent Formulation of a Dual-Mesh Hybrid LES/RANS Method. <i>Communications in Computational Physics</i> , 2017, 21, 570-599.	1.7	11
36	A Priori Assessment of Prediction Confidence for Data-Driven Turbulence Modeling. <i>Flow, Turbulence and Combustion</i> , 2017, 99, 25-46.	2.6	51

#	ARTICLE	IF	CITATIONS
37	A random matrix approach for quantifying model-form uncertainties in turbulence modeling. Computer Methods in Applied Mechanics and Engineering, 2017, 313, 941-965.	6.6	29
38	Physics-informed machine learning approach for reconstructing Reynolds stress modeling discrepancies based on DNS data. Physical Review Fluids, 2017, 2, .	2.5	403
39	Data-driven CFD modeling of turbulent flows through complex structures. International Journal of Heat and Fluid Flow, 2016, 62, 138-149.	2.4	25
40	CFD-DEM simulations of current-induced dune formation and morphological evolution. Advances in Water Resources, 2016, 92, 228-239.	3.8	47
41	Quantification of uncertainties in turbulence modeling: A comparison of physics-based and random matrix theoretic approaches. International Journal of Heat and Fluid Flow, 2016, 62, 577-592.	2.4	10
42	Sediment micromechanics in sheet flows induced by asymmetric waves: A CFD-DEM study. Computers and Geosciences, 2016, 96, 35-46.	4.2	5
43	Quantifying and reducing model-form uncertainties in Reynolds-averaged Navier-Stokes simulations: A data-driven, physics-informed Bayesian approach. Journal of Computational Physics, 2016, 324, 115-136.	3.8	209
44	A Bayesian Calibration-Prediction Method for Reducing Model-Form Uncertainties with Application in RANS Simulations. Flow, Turbulence and Combustion, 2016, 97, 761-786.	2.6	42
45	SediFoam: A general-purpose, open-source CFD-DEM solver for particle-laden flow with emphasis on sediment transport. Computers and Geosciences, 2016, 89, 207-219.	4.2	112
46	INCORPORATING PRIOR KNOWLEDGE FOR QUANTIFYING AND REDUCING MODEL-FORM UNCERTAINTY IN RANS SIMULATIONS. , 2016, 6, 109-126.		9
47	Diffusion-based coarse graining in hybrid continuum-discrete solvers: Applications in CFD-DEM. International Journal of Multiphase Flow, 2015, 72, 233-247.	3.4	69
48	Diffusion-based coarse graining in hybrid continuum-discrete solvers: Theoretical formulation and a priori tests. International Journal of Multiphase Flow, 2015, 77, 142-157.	3.4	93
49	Analysis of four-dimensional Mie imaging using fiber-based endoscopes. Applied Optics, 2014, 53, 6389.	1.8	19
50	Dynamic Evaluation of Mesh Resolution and Its Application in Hybrid LES/RANS Methods. Flow, Turbulence and Combustion, 2014, 93, 141-170.	2.6	9
51	Coupling of solvers with non-conforming computational domains in a dual-mesh hybrid LES/RANS framework. Computers and Fluids, 2013, 88, 653-662.	2.5	5
52	Preliminary Evaluation and Applications of a Consistent Hybrid LES/RANS Method. Notes on Numerical Fluid Mechanics and Multidisciplinary Design, 2012, , 91-100.	0.3	2
53	A consistent dual-mesh framework for hybrid LES/RANS modeling. Journal of Computational Physics, 2012, 231, 1848-1865.	3.8	50
54	Algorithms in a Robust Hybrid CFD-DEM Solver for Particle-Laden Flows. Communications in Computational Physics, 2011, 9, 297-323.	1.7	81

#	ARTICLE	IF	CITATIONS
55	Hydro- and morpho-dynamic modeling of breaking solitary waves over a fine sand beach. Part II: Numerical simulation. <i>Marine Geology</i> , 2010, 269, 119-131.	2.1	46
56	Scaling of transient wave-soil interaction problems. <i>International Journal for Numerical and Analytical Methods in Geomechanics</i> , 2010, 34, 839-858.	3.3	1
57	Time-scale analysis in unsaturated porous media under external wave loads. <i>International Journal for Numerical and Analytical Methods in Geomechanics</i> , 2010, 34, 1935-1959.	3.3	3
58	Hydro- and morpho-dynamic modeling of breaking solitary waves over a fine sand beach. Part I: Experimental study. <i>Marine Geology</i> , 2010, 269, 107-118.	2.1	50
59	Parametric study of breaking solitary wave induced liquefaction of coastal sand slopes. <i>Ocean Engineering</i> , 2010, 37, 1546-1553.	4.3	18
60	Liquefaction potential of coastal slopes induced by solitary waves. <i>Acta Geotechnica</i> , 2009, 4, 17-34.	5.7	49
61	Numerical study of segregation using multiscale models. <i>International Journal of Computational Fluid Dynamics</i> , 2009, 23, 81-92.	1.2	20
62	Dynamic Interactions Between the Vadose and Phreatic Zones During Breaking Solitary Wave Runup and Drawdown Over a Fine Sand Beach. , 2009, , .		0
63	Assessment of Regularized Ensemble Kalman Method for Inversion of Turbulence Quantity Fields. <i>AIAA Journal</i> , 0, , 1-11.	2.6	1