

# Linfeng Zhang

## List of Publications by Year in descending order

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48  
papers

4,297  
citations

218677

26  
h-index

223800

46  
g-index

48  
all docs

48  
docs citations

48  
times ranked

2334  
citing authors

#	ARTICLE	IF	CITATIONS
1	Deep Potential Molecular Dynamics: A Scalable Model with the Accuracy of Quantum Mechanics. <i>Physical Review Letters</i> , 2018, 120, 143001.	7.8	1,006
2	DeePMD-kit: A deep learning package for many-body potential energy representation and molecular dynamics. <i>Computer Physics Communications</i> , 2018, 228, 178-184.	7.5	727
3	Active learning of uniformly accurate interatomic potentials for materials simulation. <i>Physical Review Materials</i> , 2019, 3, .	2.4	299
4	DP-GEN: A concurrent learning platform for the generation of reliable deep learning based potential energy models. <i>Computer Physics Communications</i> , 2020, 253, 107206.	7.5	271
5	Deep Potential: A General Representation of a Many-Body Potential Energy Surface. <i>Communications in Computational Physics</i> , 2018, 23, .	1.7	164
6	DeePCG: Constructing coarse-grained models via deep neural networks. <i>Journal of Chemical Physics</i> , 2018, 149, 034101.	3.0	141
7	Phase Diagram of a Deep Potential Water Model. <i>Physical Review Letters</i> , 2021, 126, 236001.	7.8	140
8	Free energy of proton transfer at the water-TiO <sub>2</sub> interface from <i>ab initio</i> deep potential molecular dynamics. <i>Chemical Science</i> , 2020, 11, 2335-2341.	7.4	134
9	Signatures of a liquid-liquid transition in an <i>ab initio</i> deep neural network model for water. <i>Proceedings of the National Academy of Sciences of the United States of America</i> , 2020, 117, 26040-26046.	7.1	112
10	Solving many-electron Schrödinger equation using deep neural networks. <i>Journal of Computational Physics</i> , 2019, 399, 108929.	3.8	108
11	86 PFLOPS Deep Potential Molecular Dynamics simulation of 100 million atoms with <i>ab initio</i> accuracy. <i>Computer Physics Communications</i> , 2021, 259, 107624.	7.5	100
12	Raman spectrum and polarizability of liquid water from deep neural networks. <i>Physical Chemistry Chemical Physics</i> , 2020, 22, 10592-10602.	2.8	80
13	Pushing the Limit of Molecular Dynamics with <i>Ab Initio</i> Accuracy to 100 Million Atoms with Machine Learning. , 2020, , .		69
14	When do short-range atomistic machine-learning models fall short?. <i>Journal of Chemical Physics</i> , 2021, 154, 034111.	3.0	61
15	Deep potentials for materials science. <i>Materials Futures</i> , 2022, 1, 022601.	8.4	61
16	Deep neural network for the dielectric response of insulators. <i>Physical Review B</i> , 2020, 102, .	3.2	60
17	A deep potential model with long-range electrostatic interactions. <i>Journal of Chemical Physics</i> , 2022, 156, 124107.	3.0	57
18	Isotope effects in liquid water via deep potential molecular dynamics. <i>Molecular Physics</i> , 2019, 117, 3269-3281.	1.7	52

#	ARTICLE	IF	CITATIONS
19	Deep potential generation scheme and simulation protocol for the Li <sub>10</sub> GeP <sub>2</sub> S <sub>12</sub> -type superionic conductors. <i>Journal of Chemical Physics</i> , 2021, 154, 094703.	3.0	49
20	Reinforced dynamics for enhanced sampling in large atomic and molecular systems. <i>Journal of Chemical Physics</i> , 2018, 148, 124113.	3.0	48
21	Ground State Energy Functional with Hartree-Fock Efficiency and Chemical Accuracy. <i>Journal of Physical Chemistry A</i> , 2020, 124, 7155-7165.	2.5	42
22	DeePKS: A Comprehensive Data-Driven Approach toward Chemically Accurate Density Functional Theory. <i>Journal of Chemical Theory and Computation</i> , 2021, 17, 170-181.	5.3	40
23	Modeling Liquid Water by Climbing up Jacob's Ladder in Density Functional Theory Facilitated by Using Deep Neural Network Potentials. <i>Journal of Physical Chemistry B</i> , 2021, 125, 11444-11456.	2.6	40
24	Deep learning of accurate force field of ferroelectric $\text{HfO}_2$ . <i>Physical Review B</i> , 2021, 103, .	3.2	39
25	Deep learning inter-atomic potential model for accurate irradiation damage simulations. <i>Applied Physics Letters</i> , 2019, 114, .	3.3	31
26	Heat transport in liquid water from first-principles and deep neural network simulations. <i>Physical Review B</i> , 2021, 104, .	3.2	29
27	Specialising neural network potentials for accurate properties and application to the mechanical response of titanium. <i>Npj Computational Materials</i> , 2021, 7, .	8.7	26
28	Accurate Deep Potential model for the Al-Cu-Mg alloy in the full concentration space*. <i>Chinese Physics B</i> , 2021, 30, 050706.	1.4	25
29	Structural phase transitions in $\text{SrTiO}_3$ from deep potential molecular dynamics. <i>Physical Review B</i> , 2022, 105, .	3.2	25
30	Viscosity in water from first-principles and deep-neural-network simulations. <i>Npj Computational Materials</i> , 2022, 8, .	8.7	23
31	Exploring the Chemical Space of Linear Alkane Pyrolysis via Deep Potential GENERator. <i>Energy &amp; Fuels</i> , 2021, 35, 762-769.	5.1	22
32	Isotope effects in molecular structures and electronic properties of liquid water via deep potential molecular dynamics based on the SCAN functional. <i>Physical Review B</i> , 2020, 102, .	3.2	22
33	The chemical origin of temperature-dependent lithium-ion concerted diffusion in sulfide solid electrolyte Li <sub>10</sub> GeP <sub>2</sub> S <sub>12</sub> . <i>Journal of Energy Chemistry</i> , 2022, 70, 59-66.	12.9	22
34	Warm dense matter simulation via electron temperature dependent deep potential molecular dynamics. <i>Physics of Plasmas</i> , 2020, 27, .	1.9	19
35	Combined Deep Learning and Classical Potential Approach for Modeling Diffusion in UiO-66. <i>Journal of Chemical Theory and Computation</i> , 2022, 18, 3593-3606.	5.3	19
36	Efficiently Trained Deep Learning Potential for Graphane. <i>Journal of Physical Chemistry C</i> , 2021, 125, 14874-14882.	3.1	18

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37	Efficient sampling of high-dimensional free energy landscapes using adaptive reinforced dynamics. <i>Nature Computational Science</i> , 2022, 2, 20-29.	8.0	18
38	Neural Canonical Transformation with Symplectic Flows. <i>Physical Review X</i> , 2020, 10, .	8.9	13
39	Adaptive coupling of a deep neural network potential to a classical force field. <i>Journal of Chemical Physics</i> , 2018, 149, 154107.	3.0	11
40	Crystal Structure Prediction of Binary Alloys via Deep Potential. <i>Frontiers in Chemistry</i> , 2020, 8, 589795.	3.6	11
41	Machine-learning-assisted modeling. <i>Physics Today</i> , 2021, 74, 36-41.	0.3	11
42	Deep Density: Circumventing the Kohn-Sham equations via symmetry preserving neural networks. <i>Journal of Computational Physics</i> , 2021, 443, 110523.	3.8	11
43	Machine learning phase space quantum dynamics approaches. <i>Journal of Chemical Physics</i> , 2021, 154, 184104.	3.0	10
44	Extending the limit of molecular dynamics with <i>ab initio</i> accuracy to 10 billion atoms. , 2022, , .		8
45	A generalizable machine learning potential of Ag-Au nanoalloys and its application to surface reconstruction, segregation and diffusion. <i>Modelling and Simulation in Materials Science and Engineering</i> , 2022, 30, 025003.	2.0	8
46	Isotope effects in x-ray absorption spectra of liquid water. <i>Physical Review B</i> , 2020, 102, .	3.2	6
47	Neural network representation of electronic structure from <i>ab initio</i> molecular dynamics. <i>Science Bulletin</i> , 2022, 67, 29-37.	9.0	5
48	Universal approximation of symmetric and anti-symmetric functions. <i>Communications in Mathematical Sciences</i> , 2022, 20, 1397-1408.	1.0	4