

# Zachary W Ulissi

## List of Publications by Year in descending order

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42  
papers

3,766  
citations

304743

22  
h-index

254184

43  
g-index

44  
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44  
docs citations

44  
times ranked

4192  
citing authors

#	ARTICLE	IF	CITATIONS
1	Heterogeneous Catalysis in Grammar School. <i>Journal of Physical Chemistry C</i> , 2022, 126, 2931-2936.	3.1	4
2	Transfer learning using attentions across atomic systems with graph neural networks (TAAG). <i>Journal of Chemical Physics</i> , 2022, 156, 184702.	3.0	14
3	Open Challenges in Developing Generalizable Large-Scale Machine-Learning Models for Catalyst Discovery. <i>ACS Catalysis</i> , 2022, 12, 8572-8581.	11.2	18
4	Enabling robust offline active learning for machine learning potentials using simple physics-based priors. <i>Machine Learning: Science and Technology</i> , 2021, 2, 025007.	5.0	19
5	Elimination of Multidrug-Resistant Bacteria by Transition Metal Dichalcogenides Encapsulated by Synthetic Single-Stranded DNA. <i>ACS Applied Materials &amp; Interfaces</i> , 2021, 13, 8082-8094.	8.0	16
6	Efficient Discovery of Active, Selective, and Stable Catalysts for Electrochemical H <sub>2</sub> O <sub>2</sub> Synthesis through Active Motif Screening. <i>ACS Catalysis</i> , 2021, 11, 2483-2491.	11.2	44
7	Computational catalyst discovery: Active classification through myopic multiscale sampling. <i>Journal of Chemical Physics</i> , 2021, 154, 124118.	3.0	7
8	Open Catalyst 2020 (OC20) Dataset and Community Challenges. <i>ACS Catalysis</i> , 2021, 11, 6059-6072.	11.2	201
9	Deep reinforcement learning for predicting kinetic pathways to surface reconstruction in a ternary alloy. <i>Machine Learning: Science and Technology</i> , 2021, 2, 045018.	5.0	14
10	Capturing Structural Transitions in Surfactant Adsorption Isotherms at Solid/Solution Interfaces. <i>Langmuir</i> , 2020, 36, 819-826.	3.5	4
11	Differentiable Optimization for the Prediction of Ground State Structures (DOGSS). <i>Physical Review Letters</i> , 2020, 125, 173001.	7.8	8
12	Discovery of Acid-Stable Oxygen Evolution Catalysts: High-Throughput Computational Screening of Equimolar Bimetallic Oxides. <i>ACS Applied Materials &amp; Interfaces</i> , 2020, 12, 38256-38265.	8.0	47
13	In silico discovery of active, stable, CO-tolerant and cost-effective electrocatalysts for hydrogen evolution and oxidation. <i>Physical Chemistry Chemical Physics</i> , 2020, 22, 19454-19458.	2.8	12
14	Accelerated discovery of CO <sub>2</sub> electrocatalysts using active machine learning. <i>Nature</i> , 2020, 581, 178-183.	27.8	807
15	Parallelized Screening of Characterized and DFT-Modeled Bimetallic Colloidal Cocatalysts for Photocatalytic Hydrogen Evolution. <i>ACS Catalysis</i> , 2020, 10, 4244-4252.	11.2	41
16	Practical Deep-Learning Representation for Fast Heterogeneous Catalyst Screening. <i>Journal of Physical Chemistry Letters</i> , 2020, 11, 3185-3191.	4.6	63
17	Methods for comparing uncertainty quantifications for material property predictions. <i>Machine Learning: Science and Technology</i> , 2020, 1, 025006.	5.0	78
18	Toward a Design of Active Oxygen Evolution Catalysts: Insights from Automated Density Functional Theory Calculations and Machine Learning. <i>ACS Catalysis</i> , 2019, 9, 7651-7659.	11.2	118

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19	Convolutional Neural Network of Atomic Surface Structures To Predict Binding Energies for High-Throughput Screening of Catalysts. <i>Journal of Physical Chemistry Letters</i> , 2019, 10, 4401-4408.	4.6	151
20	Toward Predicting Intermetallics Surface Properties with High-Throughput DFT and Convolutional Neural Networks. <i>Journal of Chemical Information and Modeling</i> , 2019, 59, 4742-4749.	5.4	45
21	Optimization-Based Design of Active and Stable Nanostructured Surfaces. <i>Journal of Physical Chemistry C</i> , 2019, 123, 29209-29218.	3.1	8
22	Dynamic Workflows for Routine Materials Discovery in Surface Science. <i>Journal of Chemical Information and Modeling</i> , 2018, 58, 2392-2400.	5.4	39
23	Active learning across intermetallics to guide discovery of electrocatalysts for CO <sub>2</sub> reduction and H <sub>2</sub> evolution. <i>Nature Catalysis</i> , 2018, 1, 696-703.	34.4	497
24	Theoretical Investigations of Transition Metal Surface Energies under Lattice Strain and CO Environment. <i>Journal of Physical Chemistry C</i> , 2018, 122, 14481-14487.	3.1	14
25	Copper Silver Thin Films with Metastable Miscibility for Oxygen Reduction Electrocatalysis in Alkaline Electrolytes. <i>ACS Applied Energy Materials</i> , 2018, 1, 1990-1999.	5.1	40
26	To address surface reaction network complexity using scaling relations machine learning and DFT calculations. <i>Nature Communications</i> , 2017, 8, 14621.	12.8	399
27	Machine-Learning Methods Enable Exhaustive Searches for Active Bimetallic Facets and Reveal Active Site Motifs for CO <sub>2</sub> Reduction. <i>ACS Catalysis</i> , 2017, 7, 6600-6608.	11.2	300
28	Automated Discovery and Construction of Surface Phase Diagrams Using Machine Learning. <i>Journal of Physical Chemistry Letters</i> , 2016, 7, 3931-3935.	4.6	74
29	Persistently Auxetic Materials: Engineering the Poisson Ratio of 2D Self-Avoiding Membranes under Conditions of Non-Zero Anisotropic Strain. <i>ACS Nano</i> , 2016, 10, 7542-7549.	14.6	13
30	2D Equation-of-State Model for Corona Phase Molecular Recognition on Single-Walled Carbon Nanotube and Graphene Surfaces. <i>Langmuir</i> , 2015, 31, 628-636.	3.5	22
31	A Mathematical Formulation and Solution of the CoPhMoRe Inverse Problem for Helically Wrapping Polymer Corona Phases on Cylindrical Substrates. <i>Journal of Physical Chemistry C</i> , 2015, 119, 13876-13886.	3.1	40
32	Quantitative Theory of Adsorptive Separation for the Electronic Sorting of Single-Walled Carbon Nanotubes. <i>ACS Nano</i> , 2014, 8, 3367-3379.	14.6	31
33	Low Dimensional Carbon Materials for Applications in Mass and Energy Transport. <i>Chemistry of Materials</i> , 2014, 26, 172-183.	6.7	42
34	Spatiotemporal Intracellular Nitric Oxide Signaling Captured Using Internalized, Near-Infrared Fluorescent Carbon Nanotube Nanosensors. <i>Nano Letters</i> , 2014, 14, 4887-4894.	9.1	91
35	Control of nano and microchemical systems. <i>Computers and Chemical Engineering</i> , 2013, 51, 149-156.	3.8	20
36	Molecular recognition using corona phase complexes made of synthetic polymers adsorbed on carbon nanotubes. <i>Nature Nanotechnology</i> , 2013, 8, 959-968.	31.5	282

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37	Charge Transfer at Junctions of a Single Layer of Graphene and a Metallic Single Walled Carbon Nanotube. <i>Small</i> , 2013, 9, 1954-1963.	10.0	24
38	Stochastic Pore Blocking and Gating in PDMS-Glass Nanopores from Vapor-Liquid Phase Transitions. <i>Journal of Physical Chemistry C</i> , 2013, 117, 9641-9651.	3.1	15
39	A Quantitative and Predictive Model of Electromigration-Induced Breakdown of Metal Nanowires. <i>Journal of Physical Chemistry C</i> , 2013, 117, 12373-12378.	3.1	9
40	Observation of Oscillatory Surface Reactions of Riboflavin, Trolox, and Singlet Oxygen Using Single Carbon Nanotube Fluorescence Spectroscopy. <i>ACS Nano</i> , 2012, 6, 10632-10645.	14.6	58
41	Applicability of Birth-Death Markov Modeling for Single-Molecule Counting Using Single-Walled Carbon Nanotube Fluorescent Sensor Arrays. <i>Journal of Physical Chemistry Letters</i> , 2011, 2, 1690-1694.	4.6	11
42	Carbon Nanotubes as Molecular Conduits: Advances and Challenges for Transport through Isolated Sub-2 nm Pores. <i>Journal of Physical Chemistry Letters</i> , 2011, 2, 2892-2896.	4.6	19