

# Maxim A Ziatdinov

## List of Publications by Year in Descending Order

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The third column is the impact factor (IF) of the journal, and the fourth column is the number of citations of the article.

104  
papers

1,232  
citations

18  
h-index

32  
g-index

116  
ext. papers

1,744  
ext. citations

8.2  
avg, IF

4.92  
L-index

| #   | Paper  | IF   | Citations |
|-----|--|------|-----------|
| 104 | Physics makes the difference: Bayesian optimization and active learning via augmented Gaussian process. <i>Machine Learning: Science and Technology</i> , <b>2022</b> , 3, 015022  | 5.1  | 2         |
| 103 | Machine learning in scanning transmission electron microscopy. <i>Nature Reviews Methods Primers</i> , <b>2022</b> , 2,  |      | 5         |
| 102 | Hypothesis learning in automated experiment: application to combinatorial materials libraries.. <i>Advanced Materials</i> , <b>2022</b> , e2201345   | 24   | 3         |
| 101 | Automated Experiment in 4D-STEM: Exploring Emergent Physics and Structural Behaviors.. <i>ACS Nano</i> , <b>2022</b> ,   | 16.7 | 3         |
| 100 | Tracking atomic structure evolution during directed electron beam induced Si-atom motion in graphene via deep machine learning. <i>Nanotechnology</i> , <b>2021</b> , 32, 035703   | 3.4  | 4         |
| 99  | Multi-objective Bayesian optimization of ferroelectric materials with interfacial control for memory and energy storage applications. <i>Journal of Applied Physics</i> , <b>2021</b> , 130, 204102  | 2.5  | 0         |
| 98  | High-Throughput Study of Antisolvents on the Stability of Multicomponent Metal Halide Perovskites through Robotics-Based Synthesis and Machine Learning Approaches. <i>Journal of the American Chemical Society</i> , <b>2021</b> , 143, 19945-19955 | 16.4 | 4         |
| 97  | Deep Bayesian local crystallography. <i>Npj Computational Materials</i> , <b>2021</b> , 7,   | 10.9 | 5         |
| 96  | Disentangling ferroelectric domain wall geometries and pathways in dynamic piezoresponse force microscopy via unsupervised machine learning. <i>Nanotechnology</i> , <b>2021</b> , 33,   | 3.4  | 5         |
| 95  | Machine learning for high-throughput experimental exploration of metal halide perovskites. <i>Joule</i> , <b>2021</b> ,  | 27.8 | 7         |
| 94  | Thermodynamics of order and randomness in dopant distributions inferred from atomically resolved imaging. <i>Npj Computational Materials</i> , <b>2021</b> , 7,  | 10.9 | 1         |
| 93  | Investigating phase transitions from local crystallographic analysis based on statistical learning of atomic environments in 2D MoS <sub>2</sub> -ReS <sub>2</sub> . <i>Applied Physics Reviews</i> , <b>2021</b> , 8, 011409                        | 17.3 | 1         |
| 92  | Exploring order parameters and dynamic processes in disordered systems via variational autoencoders. <i>Science Advances</i> , <b>2021</b> , 7,  | 14.3 | 11        |
| 91  | Disentangling Rotational Dynamics and Ordering Transitions in a System of Self-Organizing Protein Nanorods Rotationally Invariant Latent Representations. <i>ACS Nano</i> , <b>2021</b> , 15, 6471-6480  | 16.7 | 7         |
| 90  | Separating Physically Distinct Mechanisms in Complex Infrared Plasmonic Nanostructures via Machine Learning Enhanced Electron Energy Loss Spectroscopy. <i>Advanced Optical Materials</i> , <b>2021</b> , 9, 2001808                                 | 8.1  | 7         |
| 89  | Predictability of Localized Plasmonic Responses in Nanoparticle Assemblies. <i>Small</i> , <b>2021</b> , 17, e2100181  | 11   | 7         |
| 88  | Probing atomic-scale symmetry breaking by rotationally invariant machine learning of multidimensional electron scattering. <i>Npj Computational Materials</i> , <b>2021</b> , 7,   | 10.9 | 6         |

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| 87 | Exploring Responses of Contact Kelvin Probe Force Microscopy in Triple-Cation Double-Halide Perovskites. <i>Journal of Physical Chemistry C</i> , <b>2021</b> , 125, 12355-12365  | 3.8  | 0  |
| 86 | Revealing the Chemical Bonding in Adatom Arrays via Machine Learning of Hyperspectral Scanning Tunneling Spectroscopy Data. <i>ACS Nano</i> , <b>2021</b> ,   | 16.7 | 4  |
| 85 | Bayesian Learning of Adatom Interactions from Atomically Resolved Imaging Data. <i>ACS Nano</i> , <b>2021</b> , 15, 9649-9657   | 16.7 | 2  |
| 84 | Automated Experiment in SPM: Bayesian Optimization for efficient searching of parameter space to maximize functional response. <i>Microscopy and Microanalysis</i> , <b>2021</b> , 27, 470-471                              | 0.5  | 1  |
| 83 | Building an edge computing infrastructure for rapid multi-dimensional electron microscopy. <i>Microscopy and Microanalysis</i> , <b>2021</b> , 27, 56-57  | 0.5  | 0  |
| 82 | Towards Automating Structural Discovery in Scanning Transmission Electron Microscopy. <i>Microscopy and Microanalysis</i> , <b>2021</b> , 27, 2770-2772   | 0.5  |    |
| 81 | Probing atomic-scale symmetry breaking by rotationally invariant machine learning of 4D-STEM Data.. <i>Microscopy and Microanalysis</i> , <b>2021</b> , 27, 2200-2201   | 0.5  | 0  |
| 80 | Ensemble learning-iterative training machine learning for uncertainty quantification and automated experiment in atom-resolved microscopy. <i>Npj Computational Materials</i> , <b>2021</b> , 7,                            | 10.9 | 4  |
| 79 | Autonomous Experiments in Scanning Probe Microscopy and Spectroscopy: Choosing Where to Explore Polarization Dynamics in Ferroelectrics. <i>ACS Nano</i> , <b>2021</b> ,  | 16.7 | 8  |
| 78 | Bayesian Inference for Materials Physics from STEM Data: The Probability Distribution of Physical Parameters from Ferroelectric Domain Wall Observations. <i>Microscopy and Microanalysis</i> , <b>2021</b> , 27, 1212-1214 | 0.5  | 14 |
| 77 | AtomAI: Open-source software for applications of deep learning to microscopy data. <i>Microscopy and Microanalysis</i> , <b>2021</b> , 27, 3000-3002  | 0.5  | 1  |
| 76 | Automated and Autonomous Experiments in Electron and Scanning Probe Microscopy. <i>ACS Nano</i> , <b>2021</b> ,   | 16.7 | 11 |
| 75 | Probing potential energy landscapes via electron-beam-induced single atom dynamics. <i>Acta Materialia</i> , <b>2021</b> , 203, 116508  | 8.4  | 2  |
| 74 | Quantifying the Dynamics of Protein Self-Organization Using Deep Learning Analysis of Atomic Force Microscopy Data. <i>Nano Letters</i> , <b>2021</b> , 21, 158-165   | 11.5 | 7  |
| 73 | Alignment of Au nanorods along designed protein nanofibers studied with automated image analysis. <i>Soft Matter</i> , <b>2021</b> , 17, 6109-6115  | 3.6  | 3  |
| 72 | Toward Decoding the Relationship between Domain Structure and Functionality in Ferroelectrics via Hidden Latent Variables. <i>ACS Applied Materials &amp; Interfaces</i> , <b>2021</b> , 13, 1693-1703                      | 9.5  | 14 |
| 71 | Off-the-shelf deep learning is not enough, and requires parsimony, Bayesianity, and causality. <i>Npj Computational Materials</i> , <b>2021</b> , 7,  | 10.9 | 4  |
| 70 | Computational scanning tunneling microscope image database. <i>Scientific Data</i> , <b>2021</b> , 8, 57  | 8.2  | 4  |

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| 69 | Predictability as a probe of manifest and latent physics: The case of atomic scale structural, chemical, and polarization behaviors in multiferroic Sm-doped BiFeO <sub>3</sub> . <i>Applied Physics Reviews</i> , <b>2021</b> , 8, 011403 | 17.3 | 2  |
| 68 | Deep LearningBased Workflow for Analyzing Helium Bubbles in Transmission Electron Microscopy Images. <i>Microscopy and Microanalysis</i> , <b>2021</b> , 27, 2132-2133   | 0.5  |    |
| 67 | Causal Analysis of Parameterized Atomic HAADF-STEM Across a Doped Ferroelectric Phase Boundary. <i>Microscopy and Microanalysis</i> , <b>2021</b> , 27, 2762-2764  | 0.5  |    |
| 66 | Direct mapping of polarization fields from STEM images: A Deep Learning based exploration of ferroelectrics. <i>Microscopy and Microanalysis</i> , <b>2021</b> , 27, 2990-2992   | 0.5  |    |
| 65 | Denosing STEM Electron Energy Loss Spectra using Convolutional Autoencoders. <i>Microscopy and Microanalysis</i> , <b>2021</b> , 27, 1180-1182   | 0.5  | 0  |
| 64 | Predicting local plasmon resonances and geometries using autoencoder networks in complex nanoparticle assemblies. <i>Microscopy and Microanalysis</i> , <b>2021</b> , 27, 2766-2768  | 0.5  |    |
| 63 | Beyond NMF: Advanced Signal Processing and Machine Learning Methodologies for Hyperspectral Analysis in EELS. <i>Microscopy and Microanalysis</i> , <b>2021</b> , 27, 322-324  | 0.5  | 0  |
| 62 | Deep learning ferroelectric polarization distributions from STEM data via with and without atom finding. <i>Npj Computational Materials</i> , <b>2021</b> , 7,   | 10.9 | 1  |
| 61 | Disentangling Ferroelectric Wall Dynamics and Identification of Pinning Mechanisms via Deep Learning. <i>Advanced Materials</i> , <b>2021</b> , 33, e2103680   | 24   | 7  |
| 60 | Gaussian process analysis of electron energy loss spectroscopy data: multivariate reconstruction and kernel control. <i>Npj Computational Materials</i> , <b>2021</b> , 7,   | 10.9 | 2  |
| 59 | Probing polarization dynamics at specific domain configurations: Computer-vision based automated experiment in piezoresponse force microscopy. <i>Applied Physics Letters</i> , <b>2021</b> , 119, 132902                                  | 3.4  | 3  |
| 58 | Identification and correction of temporal and spatial distortions in scanning transmission electron microscopy. <i>Ultramicroscopy</i> , <b>2021</b> , 229, 113337   | 3.1  | 2  |
| 57 | Mapping the Distortion Function via Multivariate Analysis of Atomically Resolved Images. <i>Microscopy and Microanalysis</i> , <b>2020</b> , 26, 2134-2135   | 0.5  |    |
| 56 | Reconstruction of the interatomic forces from dynamic scanning transmission electron microscopy data. <i>Journal of Applied Physics</i> , <b>2020</b> , 127, 224301  | 2.5  | 1  |
| 55 | Guided search for desired functional responses via Bayesian optimization of generative model: Hysteresis loop shape engineering in ferroelectrics. <i>Journal of Applied Physics</i> , <b>2020</b> , 128, 024102                           | 2.5  | 4  |
| 54 | Reconstruction of effective potential from statistical analysis of dynamic trajectories. <i>AIP Advances</i> , <b>2020</b> , 10, 065034  | 1.5  | 2  |
| 53 | Exploration of Electrochemical Reactions at OrganicInorganic Halide Perovskite Interfaces via Machine Learning in In Situ Time-of-Flight Secondary Ion Mass Spectrometry. <i>Advanced Functional Materials</i> , <b>2020</b> , 30, 2001995 | 15.6 | 15 |
| 52 | Imaging mechanism for hyperspectral scanning probe microscopy via Gaussian process modelling. <i>Npj Computational Materials</i> , <b>2020</b> , 6,  | 10.9 | 9  |

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| 51 | Big and Deep Data Methods in Scanning Probe Microscopy and Spectroscopy <b>2020</b> , 301-362  |      |    |
| 50 | Deep Machine Learning in Electron and Scanning Probe Microscopy <b>2020</b> , 363-395  |      |    |
| 49 | Exploration of lattice Hamiltonians for functional and structural discovery via Gaussian process-based exploration. <i>Journal of Applied Physics</i> , <b>2020</b> , 128, 164304                        | 2.5  | 4  |
| 48 | Reconstruction and uncertainty quantification of lattice Hamiltonian model parameters from observations of microscopic degrees of freedom. <i>Journal of Applied Physics</i> , <b>2020</b> , 128, 214103 | 2.5  | 1  |
| 47 | Deep learning of interface structures from simulated 4D STEM data: cation intermixing vs. roughening. <i>Machine Learning: Science and Technology</i> , <b>2020</b> , 1, 04LT01                          | 5.1  | 3  |
| 46 | Chemical Robotics Enabled Exploration of Stability in Multicomponent Lead Halide Perovskites via Machine Learning. <i>ACS Energy Letters</i> , <b>2020</b> , 5, 3426-3436                                | 20.1 | 24 |
| 45 | Causal Learning from Structural and Spectral Electron Microscopy Data. <i>Microscopy and Microanalysis</i> , <b>2020</b> , 26, 6-6   | 0.5  |    |
| 44 | Fast Scanning Probe Microscopy via Machine Learning: Non-Rectangular Scans with Compressed Sensing and Gaussian Process Optimization. <i>Small</i> , <b>2020</b> , 16, e2002878                          | 11   | 19 |
| 43 | Super-resolution and signal separation in contact Kelvin probe force microscopy of electrochemically active ferroelectric materials. <i>Journal of Applied Physics</i> , <b>2020</b> , 128, 055101       | 2.5  | 3  |
| 42 | Causal analysis of competing atomistic mechanisms in ferroelectric materials from high-resolution scanning transmission electron microscopy data. <i>Npj Computational Materials</i> , <b>2020</b> , 6,  | 10.9 | 10 |
| 41 | Exploring physics of ferroelectric domain walls via Bayesian analysis of atomically resolved STEM data. <i>Nature Communications</i> , <b>2020</b> , 11, 6361  | 17.4 | 7  |
| 40 | FerroNet: Machine Learning Flow for Analysis of Ferroelectric and Ferroelastic Materials. <i>Microscopy and Microanalysis</i> , <b>2019</b> , 25, 170-171  | 0.5  |    |
| 39 | Building and exploring libraries of atomic defects in graphene: Scanning transmission electron and scanning tunneling microscopy study. <i>Science Advances</i> , <b>2019</b> , 5, eaaw8989              | 14.3 | 41 |
| 38 | Towards Atomic Scale Quantum Structure Fabrication in 2D Materials. <i>Microscopy and Microanalysis</i> , <b>2019</b> , 25, 940-941  | 0.5  |    |
| 37 | Deep learning analysis of defect and phase evolution during electron beam-induced transformations in WS <sub>2</sub> . <i>Npj Computational Materials</i> , <b>2019</b> , 5,                             | 10.9 | 74 |
| 36 | Atom-by-atom fabrication with electron beams. <i>Nature Reviews Materials</i> , <b>2019</b> , 4, 497-507   | 73.3 | 42 |
| 35 | Deep data analytics for genetic engineering of diatoms linking genotype to phenotype via machine learning. <i>Npj Computational Materials</i> , <b>2019</b> , 5,   | 10.9 | 9  |
| 34 | Toward Electrochemical Studies on the Nanometer and Atomic Scales: Progress, Challenges, and Opportunities. <i>ACS Nano</i> , <b>2019</b> , 13, 9735-9780  | 16.7 | 18 |

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| 33 | Building ferroelectric from the bottom up: The machine learning analysis of the atomic-scale ferroelectric distortions. <i>Applied Physics Letters</i> , <b>2019</b> , 115, 052902                       | 3.4  | 13 |
| 32 | Statistical Physics-based Framework and Bayesian Inference for Model Selection and Uncertainty Quantification. <i>Microscopy and Microanalysis</i> , <b>2019</b> , 25, 130-131                           | 0.5  | 2  |
| 31 | Lab on a beam Big data and artificial intelligence in scanning transmission electron microscopy. <i>MRS Bulletin</i> , <b>2019</b> , 44, 565-575   | 3.2  | 15 |
| 30 | Materials Science in the AI age: high-throughput library generation, machine learning and a pathway from correlations to the underpinning physics. <i>MRS Communications</i> , <b>2019</b> , 9, 821      | 2.7  | 56 |
| 29 | Nanoscale interlayer defects in iron arsenides. <i>Journal of Solid State Chemistry</i> , <b>2019</b> , 277, 422-426   | 3.3  | 1  |
| 28 | Atomic Mechanisms for the Si Atom Dynamics in Graphene: Chemical Transformations at the Edge and in the Bulk. <i>Advanced Functional Materials</i> , <b>2019</b> , 29, 1904480                           | 15.6 | 17 |
| 27 | Learning from Imperfections: Predicting Structure and Thermodynamics from Atomic Imaging of Fluctuations. <i>ACS Nano</i> , <b>2019</b> , 13, 718-727  | 16.7 | 19 |
| 26 | Data mining for better material synthesis: The case of pulsed laser deposition of complex oxides. <i>Journal of Applied Physics</i> , <b>2018</b> , 123, 115303  | 2.5  | 18 |
| 25 | Deep data analysis via physically constrained linear unmixing: universal framework, domain examples, and a community-wide platform. <i>Advanced Structural and Chemical Imaging</i> , <b>2018</b> , 4, 6 | 3.9  | 27 |
| 24 | Deep Learning for Atomically Resolved Imaging. <i>Microscopy and Microanalysis</i> , <b>2018</b> , 24, 60-61   | 0.5  | 2  |
| 23 | Leveraging Single Atom Dynamics to Measure the Electron-Beam-Induced Force and Atomic Potentials. <i>Microscopy and Microanalysis</i> , <b>2018</b> , 24, 96-97  | 0.5  |    |
| 22 | Deep Convolutional Neural Networks for Symmetry Detection. <i>Microscopy and Microanalysis</i> , <b>2018</b> , 24, 112-113   | 0.5  | 2  |
| 21 | Machine Learning for the Dynamic Scanning Transmission Electron Microscopy Experiment on Solid State Transformations. <i>Microscopy and Microanalysis</i> , <b>2018</b> , 24, 1600-1601                  | 0.5  |    |
| 20 | 167-PFlops Deep Learning for Electron Microscopy: From Learning Physics to Atomic Manipulation <b>2018</b> ,   |      | 15 |
| 19 | Deep Data Analytics in Structural and Functional Imaging of Nanoscale Materials. <i>Springer Series in Materials Science</i> , <b>2018</b> , 103-128   | 0.9  | 3  |
| 18 | Mapping mesoscopic phase evolution during E-beam induced transformations via deep learning of atomically resolved images. <i>Npj Computational Materials</i> , <b>2018</b> , 4,                          | 10.9 | 24 |
| 17 | Chemically induced topological zero mode at graphene armchair edges. <i>Physical Chemistry Chemical Physics</i> , <b>2017</b> , 19, 5145-5154  | 3.6  | 11 |
| 16 | Identifying local structural states in atomic imaging by computer vision. <i>Advanced Structural and Chemical Imaging</i> , <b>2017</b> , 2, 14  | 3.9  | 9  |

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| 15 | Learning surface molecular structures via machine vision. <i>Npj Computational Materials</i> , <b>2017</b> , 3,  | 10.9 | 60  |
| 14 | Deep Learning of Atomically Resolved Scanning Transmission Electron Microscopy Images: Chemical Identification and Tracking Local Transformations. <i>ACS Nano</i> , <b>2017</b> , 11, 12742-12752                     | 16.7 | 183 |
| 13 | Bowl Inversion and Electronic Switching of Buckybowls on Gold. <i>Journal of the American Chemical Society</i> , <b>2016</b> , 138, 12142-9  | 16.4 | 38  |
| 12 | Data mining graphene: correlative analysis of structure and electronic degrees of freedom in graphenic monolayers with defects. <i>Nanotechnology</i> , <b>2016</b> , 27, 495703                                       | 3.4  | 13  |
| 11 | Deep Data Mining in a Real Space: Application to Scanning Probe Microscopy Studies on a Parent State of a High Temperature Superconductor. <i>Microscopy and Microanalysis</i> , <b>2016</b> , 22, 1418-1419           | 0.5  |     |
| 10 | Phase determination from atomically resolved images: physics-constrained deep data analysis through an unmixing approach. <i>Microscopy and Microanalysis</i> , <b>2016</b> , 22, 1452-1453                            | 0.5  |     |
| 9  | Atomic Level Structure-Property Relationship in a Spin-Orbit Mott insulator: Scanning Transmission Electron and Scanning Tunneling Microscopy Studies. <i>Microscopy and Microanalysis</i> , <b>2016</b> , 22, 908-909 | 0.5  |     |
| 8  | Deep data mining in a real space: separation of intertwined electronic responses in a lightly doped BaFeAs. <i>Nanotechnology</i> , <b>2016</b> , 27, 475706   | 3.4  | 20  |
| 7  | Analysis of citation networks as a new tool for scientific research. <i>MRS Bulletin</i> , <b>2016</b> , 41, 1009-1016   | 3.2  | 6   |
| 6  | Atomic-scale observation of structural and electronic orders in the layered compound $\text{RuCl}_2$ . <i>Nature Communications</i> , <b>2016</b> , 7, 13774   | 17.4 | 50  |
| 5  | Phases and Interfaces from Real Space Atomically Resolved Data: Physics-Based Deep Data Image Analysis. <i>Nano Letters</i> , <b>2016</b> , 16, 5574-81  | 11.5 | 26  |
| 4  | Role of edge geometry and chemistry in the electronic properties of graphene nanostructures. <i>Faraday Discussions</i> , <b>2014</b> , 173, 173-99  | 3.6  | 54  |
| 3  | Direct imaging of monovacancy-hydrogen complexes in a single graphitic layer. <i>Physical Review B</i> , <b>2014</b> , 89,   | 3.3  | 43  |
| 2  | Visualization of electronic states on atomically smooth graphitic edges with different types of hydrogen termination. <i>Physical Review B</i> , <b>2013</b> , 87,   | 3.3  | 41  |
| 1  | Latent Mechanisms of Polarization Switching from In Situ Electron Microscopy Observations. <i>Advanced Functional Materials</i> , 2100271  | 15.6 | 1   |