

# ANDRIJ VASYLENKO

Computational Materials Scientist | AI-Driven Discovery

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Computational materials scientist integrating DFT, machine learning, and high-throughput simulation workflows for autonomous materials discovery. Over a decade of experience bridging condensed-matter physics, solid-state chemistry, and AI. Developer of Bayesian-optimisation frameworks that accelerate first-principles screening; contributor to open-source tools (e.g., PyOD >16 M downloads). Proven record of mentoring researchers, managing HPC resources, and translating AI-driven models into experimentally validated discoveries (*Science*, *Nature Communications*, UK patent GB2627468).

## POSITIONS HELD & INDUSTRY COLLABORATIONS

2024–Present	<p>Research Associate (Permanent), UNIVERSITY OF LIVERPOOL (UoL)</p> <p>Lead computational materials modelling and machine-learning method development; responsible for project strategy, workflow design, and supervision of PhD students and postdoctoral researchers in AI-driven materials discovery across chemistry and computer science.</p>
2022–Present	<p>Research Consultant, CERES POWER VIA UoL</p> <p>Collaborate with Prof. Matthew Rosseinsky, OBE FRS, to design and optimise functional materials for energy applications. Developed DFT- and ML-driven workflows linking simulation with experimental validation; contributed to materials selection strategy and high-throughput screening pipelines for industrial deployment.</p>
2018–2024	<p>Research Associate, UNIVERSITY OF LIVERPOOL</p> <p>Developed supervised, semi-supervised, and unsupervised ML algorithms for materials representation, discovery, and functional optimisation. Led large-scale first-principles modelling of condensed-matter systems using VASP and ASE; integrated active-learning approaches for accelerated phase-field exploration.</p>
2015–2018	<p>Research Fellow, UNIVERSITY OF WARWICK</p> <p>Led high-throughput DFT-based crystal structure prediction of nanoconfined materials, including modification of VASP to model steric confinement effects. Developed predictive models validated experimentally, resulting in first-author publications in <i>ACS Nano</i>, <i>Physical Review B</i>, and <i>Physical Review Materials</i>.</p>

## EDUCATION

2013–2015	<p><i>PhD Physics</i> Adam Mickiewicz University, Poznań</p> <p>PhD in Physics — Adam Mickiewicz University, Poznań</p> <p><i>Marie Skłodowska-Curie Doctoral Fellowship (Eminence Programme)</i></p> <p>Thesis: <i>Nonequilibrium statistical theory and ab initio modelling of electron–ionic systems</i>.</p> <p>Advisors: Prof. Dr Sci. Mykhailo TOKARCHUK &amp; Prof. dr hab. Stefan JURGA</p>
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## AWARDS & GRANTS (SELECTED)

2025–2026	UK Alchemy Collaborative Travel Fund, <i>Principal Investigator</i> . Initiated collaboration with Empa Dübendorf on high-throughput computational–experimental integration.
2025	UK National Supercomputing Service (Isambard AI), <i>Principal Investigator</i> . Competitive allocation of 36,000 GPU-hours for AI-driven materials design.

2024–2025	UK EPSRC HPC National computational resource, <i>Principal Investigator</i> . Allocation of 1,382,400 CU hours for high-throughput materials simulation.
2018–2021	National Science Centre (NCN) (Poland), <i>Researcher Co-Investigator</i> , High-entropy alloys via high-power pulsed hybrid magnetron sputtering: modelling, synthesis, and investigation of functional properties.
2015–2018	NCN (Poland), <i>Researcher Co-Investigator</i> , Modelling and design of metal and transition metal oxides nanostructures on graphene for functional materials applications.
2013–2015	EU Marie Skłodowska-Curie Doctoral Fellowship , <i>PhD Fellow</i> . Awarded under the Eminence programme for doctoral research in theoretical condensed matter physics.

#### SCIENTIFIC COMPUTING

<i>Programming</i>	PYTHON, FORTRAN, C
<i>Parallel &amp; HPC</i>	MPI, OPENMP, SLURM, PYTORCH DISTRIBUTED (DDP)
<i>Machine Learning</i>	PYTORCH, SCIKIT-LEARN
<i>Automation</i>	ASE, PYMATGEN
<i>Data Handling</i>	NUMPY, PANDAS, MATPLOTLIB
<i>Atomistics</i>	VASP, QUANTUM ESPRESSO, LAMMPS, GULP
<i>Version Control</i>	GIT, DOCKER

#### SELECTED PUBLICATIONS & PATENTS

A. Vasylenko, et al. *Physics-informed diffusion models for extrapolating crystal structures beyond known motifs*, *Preprint*, arXiv:2501.23181 (2025). Introduces a physics-informed generative model and chemically grounded diversity metric that condition diffusion sampling beyond training distributions, producing novel, stable frameworks.

Z. Ursani, D. Antypov, K. Atkinson, M. Dyer, M. Rosseinsky, S. Schewe, and A. Vasylenko. *Protocol for Generating New Patterns Using Monte Carlo Ensemble of Hierarchical Models*, *AIPR* (2025), ID:AP0121. *Best Paper Award*. Introduces a probabilistic ensemble framework for pattern generation, linking hierarchical modelling with generative AI concepts.

G. Han, M. J. Rosseinsky, A. Vasylenko, et al. *UK Patent Application, University of Liverpool. GB2627468A - Lithium-Ion conductor materials*

G. Han, A. Vasylenko, L. M. Daniels, et al. *Superionic lithium transport via multiple coordination environments defined by two anion packing*, *Science*, 383 (6684), 739–744 (2024). Integrates AI-guided analysis with high-throughput DFT and AIMD simulations to identify and validate a novel superionic solid electrolyte.

A. Vasylenko, B. Asher, C. M. Collins, et al. *Inferring energy–composition relationships with Bayesian optimisation for enhanced exploration of inorganic materials*, *J. Chem. Phys.*, 160, 054110 (2024). Develops an active-learning framework coupling Bayesian optimisation with ab-initio data for accelerated phase-field exploration.

A. Vasylenko, J. Gamon, B. B. Duff, et al. *Element selection for crystalline inorganic solid discovery guided by unsupervised machine learning of experimentally explored chemistry*, *Nat. Commun.*, 12, 5561 (2021). Applies unsupervised representation learning to map chemical space and identify unexplored stable element combinations.