Stefano Curtarolo

List of Publications by Year in descending order

Source: https://exaly.com/author-pdf/7454092/publications.pdf

Version: 2024-02-01

169 papers 20,478 citations

19657 61 h-index 140 g-index

181 all docs

181 docs citations

times ranked

181

18032 citing authors

#	Article	IF	CITATIONS
1	Entropy-stabilized oxides. Nature Communications, 2015, 6, 8485.	12.8	1,624
2	The high-throughput highway to computational materials design. Nature Materials, 2013, 12, 191-201.	27.5	1,475
3	High-throughput electronic band structure calculations: Challenges and tools. Computational Materials Science, 2010, 49, 299-312.	3.0	1,140
4	High-entropy ceramics. Nature Reviews Materials, 2020, 5, 295-309.	48.7	902
5	AFLOW: An automatic framework for high-throughput materials discovery. Computational Materials Science, 2012, 58, 218-226.	3.0	898
6	AFLOWLIB.ORG: A distributed materials properties repository from high-throughput ab initio calculations. Computational Materials Science, 2012, 58, 227-235.	3.0	811
7	All The Catalytic Active Sites of MoS ₂ for Hydrogen Evolution. Journal of the American Chemical Society, 2016, 138, 16632-16638.	13.7	664
8	Charting the complete elastic properties of inorganic crystalline compounds. Scientific Data, 2015, 2, 150009.	5 . 3	642
9	High-entropy high-hardness metal carbides discovered by entropy descriptors. Nature Communications, 2018, 9, 4980.	12.8	604
10	Convergence of multi-valley bands as the electronic origin of high thermoelectric performance in CoSb3 skutterudites. Nature Materials, 2015, 14, 1223-1228.	27.5	587
11	Universal fragment descriptors for predicting properties of inorganic crystals. Nature Communications, 2017, 8, 15679.	12.8	435
12	QSAR without borders. Chemical Society Reviews, 2020, 49, 3525-3564.	38.1	427
13	Phase stability and mechanical properties of novel high entropy transition metal carbides. Acta Materialia, 2019, 166, 271-280.	7.9	422
14	Predicting Crystal Structures with Data Mining of Quantum Calculations. Physical Review Letters, 2003, 91, 135503.	7.8	354
15	SISSO: A compressed-sensing method for identifying the best low-dimensional descriptor in an immensity of offered candidates. Physical Review Materials, 2018, 2, .	2.4	349
16	Accuracy of ab initio methods in predicting the crystal structures of metals: A review of 80 binary alloys. Calphad: Computer Coupling of Phase Diagrams and Thermochemistry, 2005, 29, 163-211.	1.6	313
17	Machine learning modeling of superconducting critical temperature. Npj Computational Materials, 2018, 4, .	8.7	274
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The AFLOW standard for high throughput materials science calculations. Computational Materials Science, 2015, 108, 233-238. 20 Uptake of gases in bundles of carbon nanotubes. Physical Review 8, 2000, 62, 2173-2180. 21 The 2019 materials by design roadmap. Journal Physics D: Applied Physics, 2019, 52, 013001. 22 High throughput computational screening of thermal conductivity, Debye temperature, and GrAl/Anelsen parameter using a quasiharmonic Debye model. Physical Review 8, 2014, 90. 23 Low thermal conductivity and triaxial phononic anisotropy of SnSe. Applied Physics Letters, 2014, 105. 24 Dysprosium-doped cadmium oxide as a gateway material for mid-infrared plasmonics. Nature Materials. 27.5 216 28 High-Throughput Combinatorial Database of Electronic Band Structures for Inorganic Scintillator Materials. ACS Combinatorial Science, 2011, 13, 382-390. 29 Finding Unprecedentedly Low-Thermal-Conductivity Half-Heusler Semiconductors via High-Throughput Materials Modeling. Physical Review X, 2014, 4. 207 Materials Cartography: Reviewering and Mining Materials Space Using Structural and Electronic Fingerpoints. Chemistry of Materials, 2015, 27, 735-743. 208 Machine learning for alloys. Nature Reviews Materials, 2021, 6, 730-755. 48.7 202 48.7 202 48.7 202 48.7 203 Accelerated discovery of new magnets in the Heusler alloy family. Science Advances, 2017, 3, e1602241. 209 Theoretical prediction of high melting temperature for a ModC'RudC'TadC'W HCP multiprincipal element alloy, Npl Computational Materials, 2021, 7, 310 Theoretical prediction of high melting temperature for a ModC'RudC'TadC'W HCP multiprincipal element alloy, Npl Computational Materials, 2021, 7, 311 On the fly closed-loop materials discovery via Bayesian active learning. Nature Communications, 2020, 12.8 167 312 ARESTrul API for exchanging materials data in the AFLOWLIB.org consortium. Computational Materials Science, 2014, 93, 178-192. 313 The AFLOW Ubrary of Crystallegraphic Prototypes Part 1. Computational Materials, 2020, 6, . 31	#	Article	IF	CITATIONS
The 2019 materials by design roadmap. Journal Physics D: Applied Physics, 2019, 52, 013001. 2.8 236 High-throughput computational screening of thermal conductivity, Debye temperature, and GrĀ/Āneisen parameter using a quasiharmonic Debye model. Physical Review B, 2014, 90. Low thermal conductivity and triaxial phononic anisotropy of SnSe, Applied Physics Letters, 2014, 105. 3.3 226 Dysprosium-doped cadmium oxide as a gateway material for mid-infrared plasmonics. Nature Materials, 27,5 216 Physical Physics Letters, 2014, 105. High-Throughput Combinatorial Database of Electronic Band Structures for Inorganic Scintillator Air Materials, ACS Combinatorial Database of Electronic Band Structures for Inorganic Scintillator Materials, ACS Combinatorial Science, 2011, 13, 182-390. Finding Unprecedentedly Low-Thermal-Conductivity Half-Heusler Semiconductors via High-Throughput Materials Modeling, Physical Review X, 2014, 4, . Materials Cartography, Representing and Mining Materials Space Using Structural and Electronic Fingerprints. Chemistry of Materials, 2015, 27, 735-743. Machine learning for alloys. Nature Reviews Materials, 2021, 6, 730-755. 48.7 202 Accelerated discovery of new magnets in the Heusler alloy family, Science Advances, 2017, 3, e1602241. 10.3 197 Theoretical prediction of high melting temperature for a Model Rudel Physics Communications, 2020, 12.8 167 A RESTRIA Physics of Crystallographic Prototypes Part 1. Computational Materials Science, 2017, 136, 3.0 148 The AFLOW Ubrary of Crystallographic Prototypes Part 1. Computational Materials Science, 2017, 136, 3.0 142 The search for high entropy alloys: A high-throughput ab-initio approach. Acta Materials, 2018, 159, 364-383.	19	The AFLOW standard for high-throughput materials science calculations. Computational Materials Science, 2015, 108, 233-238.	3.0	244
High-throughput computational screening of thermal conductivity, Debye temperature, and GrÄ/Anelsen parameter using a quasiharmonic Debye model. Physical Review B, 2014, 90, . Low thermal conductivity and triaxial phononic anisotropy of SnSe. Applied Physics Letters, 2014, 105, . Dysprosium-doped cadmium oxide as a gateway material for mid-infrared plasmonics. Nature Materials, 2015, 14, 414420. Physh-Throughput Combinatorial Database of Electronic Band Structures for Inorganic Scintillator Materials, ACS Combinatorial Database of Electronic Band Structures for Inorganic Scintillator Materials, ACS Combinatorial Science, 2011, 13, 382-390. Finding Unprecedentedly Low-Thermal-Conductivity Half-Heusler Semiconductors via High-Throughput Materials Modeling, Physical Review X, 2014, 4, . Materials Cartography: Representing and Mining Materials Space Using Structural and Electronic fingerprints. Chemistry of Materials, 2015, 27, 735-743. Machine learning for alloys. Nature Reviews Materials, 2021, 6, 730-755. 48.7 202 Accelerated discovery of new magnets in the Heusler alloy family. Science Advances, 2017, 3, e1602241. Theoretical prediction of high melting temperature for a Mo8c*Ru8c*Ta8c*W HCP multiprincipal element alloy. Npj Computational Materials, 2021, 7, . ARESTGU API for exchanging materials data in the AFLOWUIB.org consortium. Computational Materials, 2020, 11, 5965. ARESTGU API for exchanging materials data in the AFLOWUIB.org consortium. Computational Materials Science, 2014, 93, 178-192. The Search for high entropy alloys: A high-throughput ab-initio approach. Acta Materialia, 2018, 159, 364-383.	20	Uptake of gases in bundles of carbon nanotubes. Physical Review B, 2000, 62, 2173-2180.	3.2	243
Low thermal conductivity and triaxial phononic anisotropy of SnSe. Applied Physics Letters, 2014, 105, 3.2 226 Low thermal conductivity and triaxial phononic anisotropy of SnSe. Applied Physics Letters, 2014, 105, 3.3 226 Dysprosium-doped cadmium oxide as a gateway material for mid-infrared plasmonics. Nature Materials, 2015, 14, 414-420. 216 High-Throughput Combinatorial Database of Electronic Band Structures for Inorganic Scintillator Materials, ACS Combinatorial Science, 2011, 13, 382-390. 210 Finding Unprecedentedly Low-Thermal-Conductivity Half-Heusler Semiconductors via High-Throughput Materials Modeling. Physical Review X, 2014, 4, . 8.9 210 Materials Cartography. Representing and Mining Materials Space Using Structural and Electronic Fingerprints. Chemistry of Materials, 2015, 27, 735-743. 48.7 202 Machine learning for alloys. Nature Reviews Materials, 2021, 6, 730-755. 48.7 202 Accelerated discovery of new magnets in the Heusler alloy family. Science Advances, 2017, 3, e1602241. 10.3 197 Theoretical prediction of high melting temperature for a Moše"Ruše"Taše"W HCP multiprincipal element alloy. Npj Computational Materials, 2021, 7. 186 A RESTful API for exchanging materials data in the AFLOWLIB.org consortium. Computational Materials Science, 2014, 93, 178-192. 187 The AFLOW Library of Crystallographic Prototypes: Part 1. Computational Materials Science, 2017, 136, 130 147 The search for high entropy alloys: A high-throughput ab-initio approach. Acta Materialia, 2018, 159, 79 142	21	The 2019 materials by design roadmap. Journal Physics D: Applied Physics, 2019, 52, 013001.	2.8	236
Dysprosium-doped cadmium oxide as a gateway material for mid-infrared plasmonics. Nature Materials, 27.5 216 2015, 14, 414-420. 2015, 14, 414-420. 3.8 211 High-Throughput Combinatorial Database of Electronic Band Structures for Inorganic Scintillator Materials. ACS Combinatorial Science, 2011, 13, 382-390. 3.8 211 Finding Unprecedentedly Low-Thermal-Conductivity Half-Heusler Semiconductors via High-Throughput Materials Modeling. Physical Review X, 2014, 4. 4. 8.9 210 Materials Cartography: Representing and Mining Materials Space Using Structural and Electronic Fingerprints. Chemistry of Materials, 2015, 27, 735-743. 48.7 202 Machine learning for alloys. Nature Reviews Materials, 2021, 6, 730-755. 48.7 202 Accelerated discovery of new magnets in the Heusler alloy family. Science Advances, 2017, 3, e1602241. 10.3 197 Theoretical prediction of high melting temperature for a Moá-C"Ruá-C"Taá-C"W HCP multiprincipal element alloy. Npj Computational Materials, 2021, 7, . 186 107 ARESTful API for exchanging materials data in the AFLOWUB org consortium. Computational Materials Science, 2014, 93, 178-192. 148 The search for high entropy alloys: A high-throughput ab-initio approach. Acta Materialia, 2018, 159, 7.9 142 364-383.	22	High-throughput computational screening of thermal conductivity, Debye temperature, and $Gr\tilde{A}\frac{1}{4}$ neisen parameter using a quasiharmonic Debye model. Physical Review B, 2014, 90, .	3.2	230
High-Throughput Combinatorial Database of Electronic Band Structures for Inorganic Scintillator Materials. ACS Combinatorial Science, 2011, 13, 382-390. 26 Finding Unprecedentedly Low-Thermal-Conductivity Half-Heusler Semiconductors via High-Throughput Materials Modeling. Physical Review X, 2014, 4. 27 Materials Cartography: Representing and Mining Materials Space Using Structural and Electronic Fingerprints. Chemistry of Materials, 2015, 27, 735-743. 28 Machine learning for alloys. Nature Reviews Materials, 2021, 6, 730-755. 29 Accelerated discovery of new magnets in the Heusler alloy family. Science Advances, 2017, 3, e1602241. 30 Theoretical prediction of high melting temperature for a Moâc "Ruâc "Taâc "W HCP multiprincipal element alloy. Npj Computational Materials, 2021, 7, . 31 On-the-fly closed-loop materials discovery via Bayesian active learning. Nature Communications, 2020, 11, 5966. 32 A RESTful API for exchanging materials data in the AFLOWLIB.org consortium. Computational Materials Science, 2014, 93, 178-192. 34 The AFLOW Library of Crystallographic Prototypes: Part 1. Computational Materials Science, 2017, 136, 3.0 147 35 The search for high entropy alloys: A high-throughput ab-initio approach. Acta Materialia, 2018, 159, 364-383.	23	Low thermal conductivity and triaxial phononic anisotropy of SnSe. Applied Physics Letters, 2014, 105, .	3.3	226
Materials. ACS Combinatorial Science, 2011, 13, 382-390. 26 Finding Unprecedentedly Low-Thermal-Conductivity Half-Heusler Semiconductors via High-Throughput Materials Modeling. Physical Review X, 2014, 4, . 27 Materials Cartography: Representing and Mining Materials Space Using Structural and Electronic Fingerprints. Chemistry of Materials, 2015, 27, 735-743. 28 Machine learning for alloys. Nature Reviews Materials, 2021, 6, 730-755. 48.7 202 29 Accelerated discovery of new magnets in the Heusler alloy family. Science Advances, 2017, 3, e1602241. 30 Theoretical prediction of high melting temperature for a Mo–Ru–Ta–W HCP multiprincipal element alloy. Npj Computational Materials, 2021, 7, . 31 On-the-fly closed-loop materials discovery via Bayesian active learning. Nature Communications, 2020, 11, 5966. 32 ARESTful API for exchanging materials data in the AFLOWLIB.org consortium. Computational Materials Science, 2014, 93, 178-192. 33 The AFLOW Library of Crystallographic Prototypes: Part 1. Computational Materials Science, 2017, 136, S1-5828. 34 The search for high entropy alloys: A high-throughput ab-initio approach. Acta Materialia, 2018, 159, 7.9 142	24	Dysprosium-doped cadmium oxide as a gateway material for mid-infrared plasmonics. Nature Materials, 2015, 14, 414-420.	27.5	216
High-Throughput Materials Modeling. Physical Review X, 2014, 4,	25		3.8	211
Fingerprints. Chemistry of Materials, 2015, 27, 735-743. Machine learning for alloys. Nature Reviews Materials, 2021, 6, 730-755. 48.7 202 Accelerated discovery of new magnets in the Heusler alloy family. Science Advances, 2017, 3, e1602241. Theoretical prediction of high melting temperature for a Mo–Ru–Ta–W HCP multiprincipal element alloy. Npj Computational Materials, 2021, 7, . 186 On-the-fly closed-loop materials discovery via Bayesian active learning. Nature Communications, 2020, 11, 5966. A RESTful API for exchanging materials data in the AFLOWLIB.org consortium. Computational Materials Science, 2014, 93, 178-192. A RESTful API for exchanging materials data in the AFLOWLIB.org consortium. Computational 3.0 148 The AFLOW Library of Crystallographic Prototypes: Part 1. Computational Materials Science, 2017, 136, S1-S828. The search for high entropy alloys: A high-throughput ab-initio approach. Acta Materialia, 2018, 159, 7.9 142	26		8.9	210
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Materials Science, 2014, 93, 178-192. The AFLOW Library of Crystallographic Prototypes: Part 1. Computational Materials Science, 2017, 136, S1-S828. The search for high entropy alloys: A high-throughput ab-initio approach. Acta Materialia, 2018, 159, 364-383.	31		12.8	167
The search for high entropy alloys: A high-throughput ab-initio approach. Acta Materialia, 2018, 159, 364-383.	32		3.0	148
34 364-383. 7.9 142	33		3.0	147
Discovery of high-entropy ceramics via machine learning. Npj Computational Materials, 2020, 6, . 8.7 133	34		7.9	142
	35	Discovery of high-entropy ceramics via machine learning. Npj Computational Materials, 2020, 6, .	8.7	133

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37	Nanograined Halfâ€Heusler Semiconductors as Advanced Thermoelectrics: An Ab Initio Highâ€Throughput Statistical Study. Advanced Functional Materials, 2014, 24, 7427-7432.	14.9	117
38	Spectral descriptors for bulk metallic glasses based on the thermodynamics of competing crystalline phases. Nature Communications, 2016, 7, 12315.	12.8	104
39	How Chemical Composition Alone Can Predict Vibrational Free Energies and Entropies of Solids. Chemistry of Materials, 2017, 29, 6220-6227.	6.7	103
40	Hafnium binary alloys from experiments and first principles. Acta Materialia, 2010, 58, 2887-2897.	7.9	102
41	High-throughput and data mining with ab initio methods. Measurement Science and Technology, 2005, 16, 296-301.	2.6	100
42	Modeling the melting of supported clusters. Applied Physics Letters, 2006, 88, 133110.	3.3	95
43	Assessing the Thermoelectric Properties of Sintered Compounds via High-Throughput < i > Ab-Initio < /i > Calculations. Physical Review X, 2011, 1, .	8.9	92
44	Thermopower of Molecular Junctions: An ab Initio Study. Nano Letters, 2009, 9, 1011-1014.	9.1	91
45	Decoupling Local Disorder and Optical Effects in Infrared Spectra: Differentiating Between Calcites with Different Origins. Advanced Materials, 2011, 23, 550-554.	21.0	91
46	Theoretical study of metal borides stability. Physical Review B, 2006, 74, .	3.2	79
47	Modeling Off-Stoichiometry Materials with a High-Throughput Ab-Initio Approach. Chemistry of Materials, 2016, 28, 6484-6492.	6.7	78
48	Prediction of different crystal structure phases in metal borides: A lithium monoboride analog toMgB2. Physical Review B, 2006, 73, .	3.2	77
49	PAOFLOW: A utility to construct and operate on ab initio Hamiltonians from the projections of electronic wavefunctions on atomic orbital bases, including characterization of topological materials. Computational Materials Science, 2018, 143, 462-472.	3.0	74
50	Predicting superhard materials via a machine learning informed evolutionary structure search. Npj Computational Materials, 2019, 5, .	8.7	74
51	Theoretical study of the thermal behavior of free and alumina-supported Fe-C nanoparticles. Physical Review B, 2007, 75, .	3.2	73
52	Reduced Carbon Solubility in Fe Nanoclusters and Implications for the Growth of Single-Walled Carbon Nanotubes. Physical Review Letters, 2008, 100, 195502.	7.8	73
53	AFLOW-ML: A RESTful API for machine-learning predictions of materials properties. Computational Materials Science, 2018, 152, 134-145.	3.0	72
54	A computational high-throughput search for new ternary superalloys. Acta Materialia, 2017, 122, 438-447.	7.9	70

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55	The AFLOW Library of Crystallographic Prototypes: Part 2. Computational Materials Science, 2019, 161, S1-S1011.	3.0	70
56	AFLOWÏ€: A minimalist approach to high-throughput ab initio calculations including the generation of tight-binding hamiltonians. Computational Materials Science, 2017, 136, 76-84.	3.0	70
57	AFLOW-CHULL: Cloud-Oriented Platform for Autonomous Phase Stability Analysis. Journal of Chemical Information and Modeling, 2018, 58, 2477-2490.	5.4	69
58	Uncovering Compounds by Synergy of Cluster Expansion and High-Throughput Methods. Journal of the American Chemical Society, 2010, 132, 4830-4833.	13.7	68
59	An efficient and accurate framework for calculating lattice thermal conductivity of solids: AFLOW—AAPL Automatic Anharmonic Phonon Library. Npj Computational Materials, 2017, 3, .	8.7	65
60	The New Face of Rhodium Alloys: Revealing Ordered Structures from First Principles. Journal of the American Chemical Society, 2010, 132, 833-837.	13.7	64
61	Effective and accurate representation of extended Bloch states on finite Hilbert spaces. Physical Review B, 2013, 88, .	3.2	64
62	Crystallization behavior upon heating and cooling in Cu50Zr50 metallic glass thin films. Acta Materialia, 2016, 121, 68-77.	7.9	64
63	Dynamics of an Inhomogeneously Coarse Grained Multiscale System. Physical Review Letters, 2002, 88, 255504.	7.8	62
64	Ordered Structures in Rhenium Binary Alloys from First-Principles Calculations. Journal of the American Chemical Society, 2011, 133, 158-163.	13.7	62
65	High-Throughput Computation of Thermal Conductivity of High-Temperature Solid Phases: The Case of Oxide and Fluoride Perovskites. Physical Review X, 2016, 6, .	8.9	61
66	Unavoidable disorder and entropy in multi-component systems. Npj Computational Materials, 2019, 5, .	8.7	61
67	Revealing the Impact of Catalyst Phase Transition on Carbon Nanotube Growth by <i>in Situ</i> Raman Spectroscopy. ACS Nano, 2013, 7, 1100-1107.	14. 6	60
68	Hidden features of the catalyst nanoparticles favorable for single-walled carbon nanotube growth. Applied Physics Letters, 2007, 90, 163120.	3.3	59
69	Comparative Study of Nonproportionality and Electronic Band Structures Features in Scintillator Materials. IEEE Transactions on Nuclear Science, 2009, 56, 2989-2996.	2.0	59
70	Challenges in Ceramic Science: A Report from the Workshop on Emerging Research Areas in Ceramic Science. Journal of the American Ceramic Society, 2012, 95, 3699-3712.	3.8	59
71	Improved predictions of the physical properties of Zn- and Cd-based wide band-gap semiconductors: A validation of the ACBNO functional. Physical Review B, 2015, 91, .	3.2	56
72	AFLUX: The LUX materials search API for the AFLOW data repositories. Computational Materials Science, 2017, 137, 362-370.	3.0	56

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73	Finding the stable structures of <mml:math xmlns:mml="http://www.w3.org/1998/Math/MathML"> <mml:mrow> <mml:msub> <mml:mtext> N </mml:mtext> < n an <i>ab initio </i> high-throughput approach. Physical Review B, 2015, 91, .</mml:msub></mml:mrow></mml:math>	า เลไร mrow ว	> 6raml:mn>
74	High-throughput prediction of finite-temperature properties using the quasi-harmonic approximation. Computational Materials Science, 2016, 125, 82-91.	3.0	51
7 5	Ordered phases in ruthenium binary alloys from high-throughput first-principles calculations. Physical Review B, 2011, 84, .	3.2	50
76	Surface segregation in nanoparticles from first principles: The case of FePt. Scripta Materialia, 2010, 62, 179-182.	5.2	49
77	OPTIMADE, an API for exchanging materials data. Scientific Data, 2021, 8, 217.	5.3	49
78	Data-Mining-Driven Quantum Mechanics for the Prediction of Structure. MRS Bulletin, 2006, 31, 981-985.	3.5	48
79	Guiding the experimental discovery of magnesium alloys. Physical Review B, 2011, 84, .	3.2	47
80	Comprehensive Search for New Phases and Compounds in Binary Alloy Systems Based on Platinum-Group Metals, Using a Computational First-Principles Approach. Physical Review X, 2013, 3, .	8.9	47
81	Combining the AFLOW GIBBS and elastic libraries to efficiently and robustly screen thermomechanical properties of solids. Physical Review Materials, 2017, 1, .	2.4	47
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