

# Peter DÃ¼ben

## List of Publications by Year in descending order

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Version: 2024-02-01

60  
papers

1,974  
citations

331538

21  
h-index

276775

41  
g-index

84  
all docs

84  
docs citations

84  
times ranked

1850  
citing authors

#	ARTICLE	IF	CITATIONS
1	DYAMOND: the DYNAMICS of the Atmospheric general circulation Modeled On Non-hydrostatic Domains. <i>Progress in Earth and Planetary Science</i> , 2019, 6, .	1.1	239
2	Challenges and design choices for global weather and climate models based on machine learning. <i>Geoscientific Model Development</i> , 2018, 11, 3999-4009.	1.3	179
3	Global Cloud-Resolving Models. <i>Current Climate Change Reports</i> , 2019, 5, 172-184.	2.8	164
4	WeatherBench: A Benchmark Data Set for Data-Driven Weather Forecasting. <i>Journal of Advances in Modeling Earth Systems</i> , 2020, 12, e2020MS002203.	1.3	150
5	The digital revolution of Earth-system science. <i>Nature Computational Science</i> , 2021, 1, 104-113.	3.8	98
6	Single Precision in Weather Forecasting Models: An Evaluation with the IFS. <i>Monthly Weather Review</i> , 2017, 145, 495-502.	0.5	75
7	Deep learning for post-processing ensemble weather forecasts. <i>Philosophical Transactions Series A, Mathematical, Physical, and Engineering Sciences</i> , 2021, 379, 20200092.	1.6	60
8	Opportunities and challenges for machine learning in weather and climate modelling: hard, medium and soft AI. <i>Philosophical Transactions Series A, Mathematical, Physical, and Engineering Sciences</i> , 2021, 379, 20200083.	1.6	55
9	A Baseline for Global Weather and Climate Simulations at 1 km Resolution. <i>Journal of Advances in Modeling Earth Systems</i> , 2020, 12, e2020MS002192.	1.3	54
10	Benchmark Tests for Numerical Weather Forecasts on Inexact Hardware. <i>Monthly Weather Review</i> , 2014, 142, 3809-3829.	0.5	50
11	The use of imprecise processing to improve accuracy in weather & climate prediction. <i>Journal of Computational Physics</i> , 2014, 271, 2-18.	1.9	48
12	Assessing the scales in numerical weather and climate predictions: will exascale be the rescue?. <i>Philosophical Transactions Series A, Mathematical, Physical, and Engineering Sciences</i> , 2019, 377, 20180148.	1.6	48
13	Bridging observations, theory and numerical simulation of the ocean using machine learning. <i>Environmental Research Letters</i> , 2021, 16, 073008.	2.2	40
14	rpe v5: an emulator for reduced floating-point precision in large numerical simulations. <i>Geoscientific Model Development</i> , 2017, 10, 2221-2230.	1.3	39
15	TRU-NET: a deep learning approach to high resolution prediction of rainfall. <i>Machine Learning</i> , 2021, 110, 2035-2062.	3.4	33
16	Machine Learning Emulation of Gravity Wave Drag in Numerical Weather Forecasting. <i>Journal of Advances in Modeling Earth Systems</i> , 2021, 13, e2021MS002477.	1.3	33
17	On the use of programmable hardware and reduced numerical precision in earth-system modeling. <i>Journal of Advances in Modeling Earth Systems</i> , 2015, 7, 1393-1408.	1.3	29
18	On the use of scale-dependent precision in Earth System modelling. <i>Quarterly Journal of the Royal Meteorological Society</i> , 2017, 143, 897-908.	1.0	28

#	ARTICLE	IF	CITATIONS
19	Tropical Cyclones in Global Storm-Resolving Models. Journal of the Meteorological Society of Japan, 2021, 99, 579-602.	0.7	28
20	Global Simulations of the Atmosphere at 1.45 km Grid-Spacing with the Integrated Forecasting System. Journal of the Meteorological Society of Japan, 2020, 98, 551-572.	0.7	27
21	On the use of inexact, pruned hardware in atmospheric modelling. Philosophical Transactions Series A, Mathematical, Physical, and Engineering Sciences, 2014, 372, 20130276.	1.6	25
22	Improving Weather Forecast Skill through Reduced-Precision Data Assimilation. Monthly Weather Review, 2018, 146, 49-62.	0.5	24
23	Number Formats, Error Mitigation, and Scope for 16â€Bit Arithmetics in Weather and Climate Modeling Analyzed With a Shallow Water Model. Journal of Advances in Modeling Earth Systems, 2020, 12, e2020MS002246.	1.3	24
24	Systematic detection of local CH&lt;sub&gt;4&lt;/sub& anomalies by combining satellite measurements with high-resolution forecasts. Atmospheric Chemistry and Physics, 2021, 21, 5117-5136.	1.9	24
25	Posits as an alternative to floats for weather and climate models. , 2019, , .		23
26	Opportunities for Energy Efficient Computing: A Study of Inexact General Purpose Processors for High-Performance and Big-Data Applications. , 2015, , .		22
27	Scale-Selective Precision for Weather and Climate Forecasting. Monthly Weather Review, 2019, 147, 645-655.	0.5	22
28	Building Tangentâ€Linear and Adjoint Models for Data Assimilation With Neural Networks. Journal of Advances in Modeling Earth Systems, 2021, 13, e2021MS002521.	1.3	22
29	Accelerating High-Resolution Weather Models with Deep-Learning Hardware. , 2019, , .		20
30	Reliable low precision simulations in land surface models. Climate Dynamics, 2018, 51, 2657-2666.	1.7	18
31	Choosing the Optimal Numerical Precision for Data Assimilation in the Presence of Model Error. Journal of Advances in Modeling Earth Systems, 2018, 10, 2177-2191.	1.3	18
32	Machine Learning for Earth System Observation and Prediction. Bulletin of the American Meteorological Society, 2021, 102, E710-E716.	1.7	18
33	A study of reduced numerical precision to make superparameterization more competitive using a hardware emulator in the OpenIFS model. Journal of Advances in Modeling Earth Systems, 2017, 9, 566-584.	1.3	16
34	Model intercomparison of COSMO 5.0 and IFS 45r1 at kilometer-scale grid spacing. Geoscientific Model Development, 2021, 14, 4617-4639.	1.3	15
35	More accuracy with less precision. Quarterly Journal of the Royal Meteorological Society, 2021, 147, 4358-4370.	1.0	13
36	Machine Learning Emulation of 3D Cloud Radiative Effects. Journal of Advances in Modeling Earth Systems, 2022, 14, .	1.3	12

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37	Compressing atmospheric data into its real information content. <i>Nature Computational Science</i> , 2021, 1, 713-724.	3.8	12
38	A discontinuous/continuous low order finite element shallow water model on the sphere. <i>Journal of Computational Physics</i> , 2012, 231, 2396-2413.	1.9	11
39	Monte Carlo simulations of the randomly forced Burgers equation. <i>Europhysics Letters</i> , 2008, 84, 40002.	0.7	10
40	Exploiting the chaotic behaviour of atmospheric models with reconfigurable architectures. <i>Computer Physics Communications</i> , 2017, 221, 160-173.	3.0	10
41	Stochastic modelling and energy-efficient computing for weather and climate prediction. <i>Philosophical Transactions Series A, Mathematical, Physical, and Engineering Sciences</i> , 2014, 372, 20140118.	1.6	9
42	Bitwise efficiency in chaotic models. <i>Proceedings of the Royal Society A: Mathematical, Physical and Engineering Sciences</i> , 2017, 473, 20170144.	1.0	9
43	A Comparison of Data-Driven Approaches to Build Low-Dimensional Ocean Models. <i>Journal of Advances in Modeling Earth Systems</i> , 2021, 13, e2021MS002537.	1.3	9
44	Rounding errors may be beneficial for simulations of atmospheric flow: results from the forced 1D Burgers equation. <i>Theoretical and Computational Fluid Dynamics</i> , 2015, 29, 311-328.	0.9	8
45	Fluid Simulations Accelerated With 16 Bits: Approaching 4x Speedup on A64FX by Squeezing ShallowWaters.jl Into Float16. <i>Journal of Advances in Modeling Earth Systems</i> , 2022, 14, .	1.3	8
46	Atmosphere and Ocean Modeling on Grids of Variable Resolution- A 2D Case Study. <i>Monthly Weather Review</i> , 2014, 142, 1997-2017.	0.5	7
47	A power law for reduced precision at small spatial scales: Experiments with an SQG model. <i>Quarterly Journal of the Royal Meteorological Society</i> , 2018, 144, 1179-1188.	1.0	7
48	Resilience and fault tolerance in high-performance computing for numerical weather and climate prediction. <i>International Journal of High Performance Computing Applications</i> , 2021, 35, 285-311.	2.4	7
49	Architectures and Precision Analysis for Modelling Atmospheric Variables with Chaotic Behaviour. , 2015, , .		6
50	Reduced-precision parametrization: lessons from an intermediate-complexity atmospheric model. <i>Quarterly Journal of the Royal Meteorological Society</i> , 2020, 146, 1590-1607.	1.0	6
51	Deep Learning to Estimate Model Biases in an Operational NWP Assimilation System. <i>Journal of Advances in Modeling Earth Systems</i> , 2022, 14, .	1.3	6
52	Lower precision for higher accuracy: Precision and resolution exploration for shallow water equations. , 2015, , .		5
53	An approach to secure weather and climate models against hardware faults. <i>Journal of Advances in Modeling Earth Systems</i> , 2017, 9, 501-513.	1.3	4
54	A Stochastic Representation of Subgrid Uncertainty for Dynamical Core Development. <i>Bulletin of the American Meteorological Society</i> , 2019, 100, 1091-1101.	1.7	4

#	ARTICLE	IF	CITATIONS
55	Single-Precision in the Tangent-Linear and Adjoint Models of Incremental 4D-Var. Monthly Weather Review, 2020, 148, 1541-1552.	0.5	4
56	Machine Learning Emulation of Urban Land Surface Processes. Journal of Advances in Modeling Earth Systems, 2022, 14, .	1.3	4
57	A New Number Format for Ensemble Simulations. Journal of Advances in Modeling Earth Systems, 2018, 10, 2983-2991.	1.3	3
58	The Relationship between Numerical Precision and Forecast Lead Time in the Lorenzâ€™95 System. Monthly Weather Review, 2020, 148, 849-855.	0.5	2
59	Validating optimisations for chaotic simulations. , 2017, , .		1
60	New Methods for Data Storage of Model Output from Ensemble Simulations. Monthly Weather Review, 2019, 147, 677-689.	0.5	1