

Quantum Loop Topography for Machine Learning

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Citation Report

#	ARTICLE	IF	CITATIONS
1	Self-learning Monte Carlo method: Continuous-time algorithm. Physical Review B, 2017, 96, .	1.1	55
2	Machine learning quantum phases of matter beyond the fermion sign problem. Scientific Reports, 2017, 7, 8823.	1.6	252
3	Probing many-body localization with neural networks. Physical Review B, 2017, 95, .	1.1	117
4	Solving the Bose-Hubbard Model with Machine Learning. Journal of the Physical Society of Japan, 2017, 86, 093001.	0.7	97
5	Machine learning Z^2 quantum spin liquids with quasiparticle statistics. Physical Review B, 2017, 96, .	1.1	99
6	Kernel methods for interpretable machine learning of order parameters. Physical Review B, 2017, 96, .	1.1	99
7	Restricted Boltzmann machine learning for solving strongly correlated quantum systems. Physical Review B, 2017, 96, .	1.1	198
8	Phase Diagrams of Three-Dimensional Anderson and Quantum Percolation Models Using Deep Three-Dimensional Convolutional Neural Network. Journal of the Physical Society of Japan, 2017, 86, 113704.	0.7	22
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20	Approximating quantum many-body wave functions using artificial neural networks. Physical Review B, 2018, 97, .	1.1	146
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