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Automated machine learning for dynamic energy management using time-series data

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- Can Wang, Thomas Bäck, Holger H. Hoos, Mitra Baratchi, Steffen Limmer, and Markus Olhofer. Automated Machine Learning for Short-term Electric Load Forecasting. In *Proceedings of the IEEE Symposium Series on Computational Intelligence, Xiamen, China, December 6-9, 2019*, pages 314–321.
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