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H. Meidani</td></tr>
<tr><th>title</th><td>Adaptive Physics-Informed Neural Networks for Markov-Chain Monte Carlo</td></tr>
<tr><th>publication_date</th><td>2020-08-03 00:00:00</td></tr>
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