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# Appendix: Multi-directional Recurrent Neural Networks: A Novel Method for Estimating Missing Data

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Jinsung Yoon<sup>1</sup> William R. Zame<sup>2</sup> Mihaela van der Schaar<sup>1,3,4</sup>

## 1. Configurations of the Experiments

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Method	Configurations
<b>M-RNN</b>	Model :Multi-Directional Recurrent Neural Networks Initialization: Xavier Initialization [13] Optimization: Adam* Optimization [14] (learning rate = 0.05) Batch size = 100, Iterations = 1000 Depth: 2 Constraint 1: The matrix parameters are diagonals for the interpolation part. Constraint 2:The diagonal part of the matrix is zero for the imputation part.

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<sup>1</sup>Department of Electrical Engineering, University of California, Los Angeles, California, USA. <sup>2</sup>Departments of Economics and Mathematics, University of California, Los Angeles, California, USA. <sup>3</sup>Man Institute, University of Oxford, Oxford, United Kingdom.

<sup>4</sup>Alan Turing Institute, London, United Kingdom. Correspondence to: Jinsung Yoon <jsyoon0823@g.ucla.edu>.

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