

Recurrent Neural Networks for Interval Duration Discrimination Task

Kyriacos Nikiforou^{1*}, Pedro A. M. Mediano^{2*}, Murray Shanahan
Department of Computing, Imperial College London, UK

* First co-authors

kn910@ic.ac.uk¹, pmediano@imperial.ac.uk²

Introduction and methods

- We analyse how a randomly connected network of firing rate neurons can perform computations on the temporal features of input stimuli.
- We extend previous work^{1,2} and conduct experiments whereby networks of a few hundred neurons were trained to discriminate whether the time between two input stimuli was larger or smaller than a set duration [150 ms].
- The discrimination was achieved by training the readout weights of the network with FORCE learning³ and an analysis was performed on the state of the network using principal component analysis and temporal mutual information⁴.

Simulations

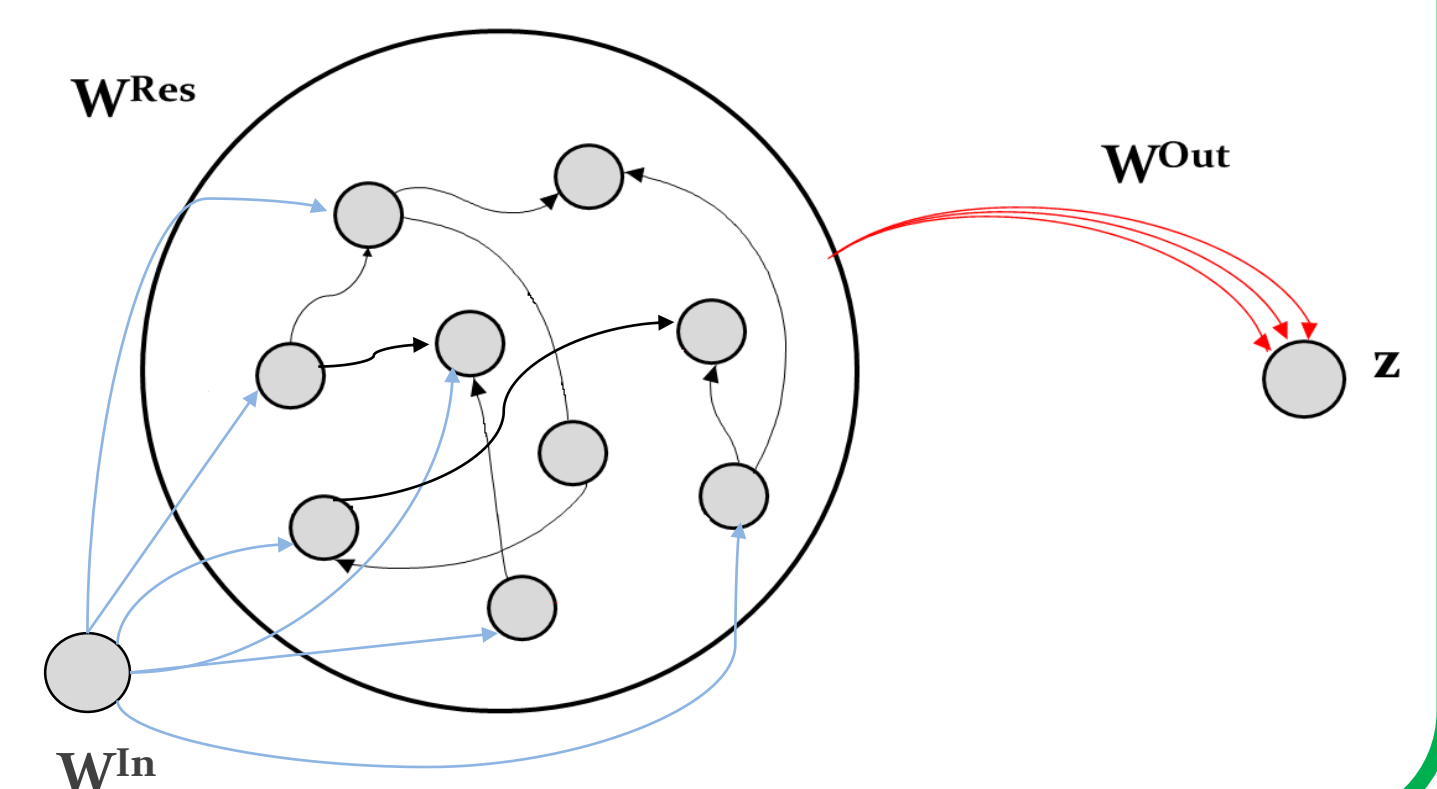
Neuron model

$$\tau \frac{dx_i}{dt} = -x_i + \sum_j W_{ij}^{Res} \tanh(x_j) + W_i^{In} Input$$

In the equation, x_i indicates neuron's i membrane potential and τ the time constant of the network.

Network configuration

W^{Res} (black), W^{Out} (red and trainable) and W^{In} (blue) are the connectivity matrices shown the figure.

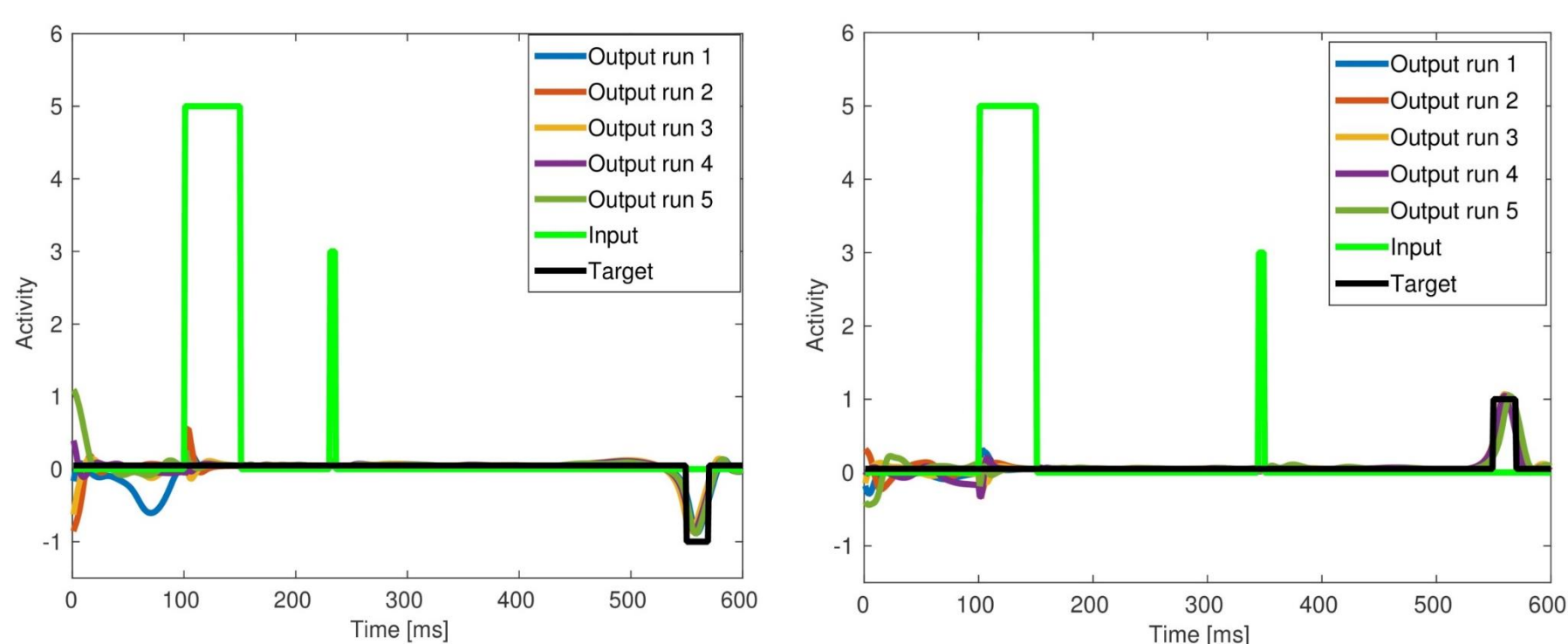


Results

I. Interval duration discrimination experiments

Output

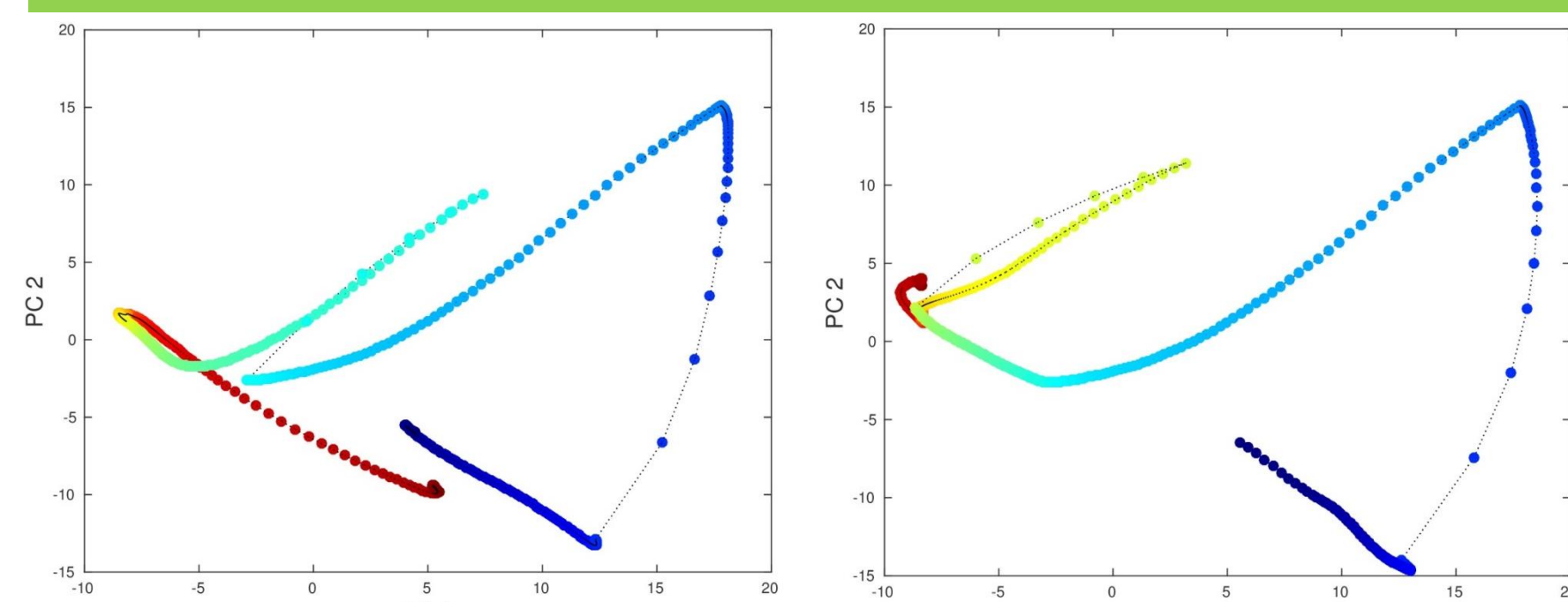
Plots of the network's output activity overlaid on the input (green) and correct output (black).



95.75% Interval Classification Accuracy

- FORCE learning on output weights only can successfully learn to discriminate intervals.
- The temporal computation is encoded in the state of the network.

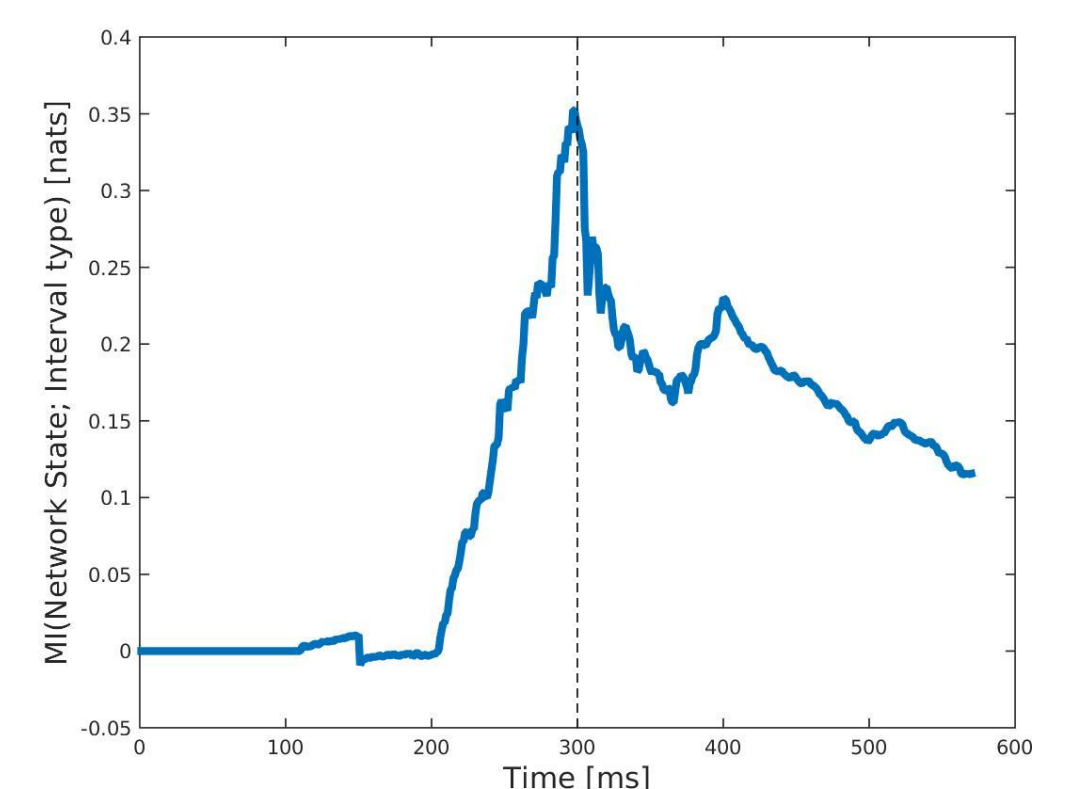
Principal component analysis



Projection of the network activity on its first two principal components.

II. Temporal mutual information (MI)

- Average mutual information between individual neuron state and interval type (short or long) at every timestep.
- Peak at 300 ms indicates the timestep where highest amount of information about interval type can be extracted from a single neuron in the network.



Conclusions

- The dynamical state of the network has an intrinsic temporal memory that can be associated through simple linear readout units with temporal computations such as the interval duration discrimination task.
- Small cortical circuits can perform such temporal computations in a distributed manner.

References

- Maniadakis, M., & Trahanias, P. (2015). Artificial Agents Perceiving and Processing Time. *IEEE IJCNN*.
- Laje, R., & Buonomano, D. V. (2013). Robust timing and motor patterns by taming chaos in recurrent neural networks. *Nature Neuroscience*, 16(7), 925-933.
- Sussillo, D., & Abbott, L. F. (2009). Generating coherent patterns of activity from chaotic neural networks. *Neuron*, 63(4), 544-57.
- Beer, R. D., & Williams, P. L. (2014). Information Processing and Dynamics in Minimally Cognitive Agents. *Cognitive Science*, 1-38.

This work was supported by EPSRC Doctoral training fund (Award reference 1506349) and Timestorm EU project.