

# Disparate effects of resonances in delay-based reservoir computers for chaotic attractor reconstruction

*Keywords: time-multiplexed networks, reservoir computing, network topology, resonance effects, attractor reconstruction*

## Extended Abstract

Recurrent neural networks are well known for their suitability as high dimensional dynamical systems in the framework of reservoir computing (RC)[1]. The coupling-topology of the network has a strong impact on the nonlinear mixing of injected information and, consequently, on the performance of the network as a reservoir computer. Using the analogy between spatially extended systems and systems with time delay, time-multiplexed RC has become a promising concept that allows the use of dynamical systems as processing units for RC[2, 3]. In the delay-based time-multiplexed RC approach, analogous to the coupling-topology of networks, different internal coupling topologies can be realized by tuning the timescales of the delay  $\tau$  or the data injection rate (inverse of the clock cycle  $T$ ) [4]. In this paper we investigate the influence of the coupling topology of delay-based RC on chaotic time-series prediction tasks. Particularly, we demonstrate the differences in the performance arising when operating the trained reservoir open- and closed-loop (see Fig.1a). In the open-loop setup the reservoir is always fed the true timeseries in order to make predictions, whereas, in the closed-loop operation the reservoir output is fed in as the next input, thereby acting as a surrogate of the dynamical system which evolves autonomously after training.

In the limit where the dynamical system (reservoir) reacts much faster than the sampling time, the topology of delay-based time-multiplexed RCs is determined by the ratio between the delay ( $\tau$ ) and the input clock cycle ( $T$ ). Equal values of  $\tau$  and  $T$  (1:1 resonance) lead to a network topology of uncoupled nodes with delayed self-coupling, i.e. without inter-node coupling. Whereas, asynchronous settings for  $\tau$  and  $T$  leads to delay-induced effective next-neighbor coupling, similar to a unidirectional ring (see Fig.1b). For open-loop chaotic timeseries prediction tasks, it is well known that  $\tau$ - $T$  resonances can have a detrimental effect on the performance [4, 5]. An example of the performance reduction at  $\tau$ - $T$  resonances can be seen in Fig.1c, where a 2D scan of a 2-step open-loop Mackey-Glass prediction task, performed with a semiconductor laser subjected to delayed self-feedback, is shown (prediction performance is plotted as color code with bright lines of bad performance appearing along resonances between  $T$  and  $\tau$ ).

For spatially multiplexed RC setups it was shown that an autonomous RC system operated in a closed loop configuration is more likely to produce a stable reconstruction of a chaotic timeseries if inter-node couplings are absent, i.e. only self-coupling [6]. Here we report that a similar effect can be observed in delay-based time-multiplexed RC. The impact of resonances, and therefore the impact of coupling topology, is disparate in delay-based RC as it depends on the operation mode after training. Figure 2 shows the performance of a reservoir build from a delayed-map both in open (a) and closed loop (b) operation as a function of the internal feedback delay time  $\tau$ . The scans indicate that delay-based time-multiplexed RC shows bad performance at  $\tau$ - $T$  resonances if operated open-loop, however, improved performance is evident at the resonances in autonomous closed-loop operation. While the impact of resonances

in the open-loop case can be traced back to the memory properties of the RC setup [4, 5], the open-loop case remains challenging and will be discussed in this contribution.

## References

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## Figures

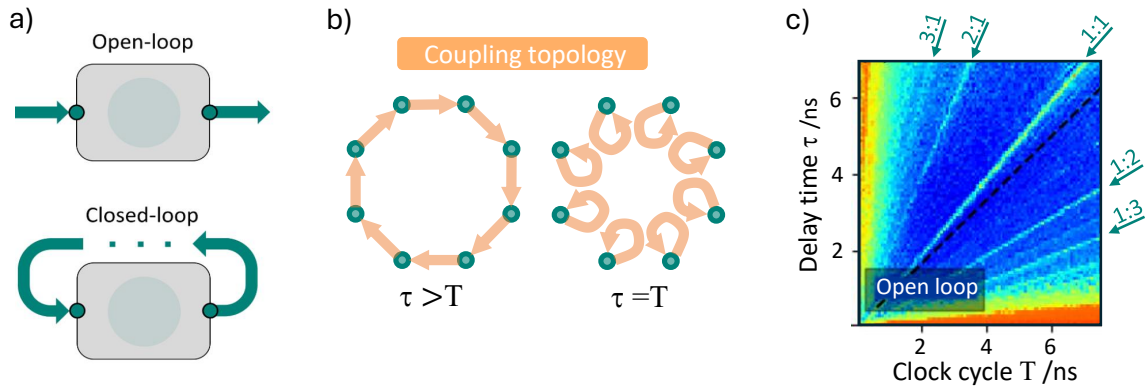


Figure 1: **Sketch** of a) open and closed loop RC configurations, b) Delay-induced coupling topology in time-multiplexed RC, c) Computing performance of a semiconductor laser with feedback for Mackey Glas open-loop 2-step prediction as a function of delay  $\tau$  and clock cycle  $T$  (arrows point at resonances and yellow/red (blue) colors indicate bad(good) performance).

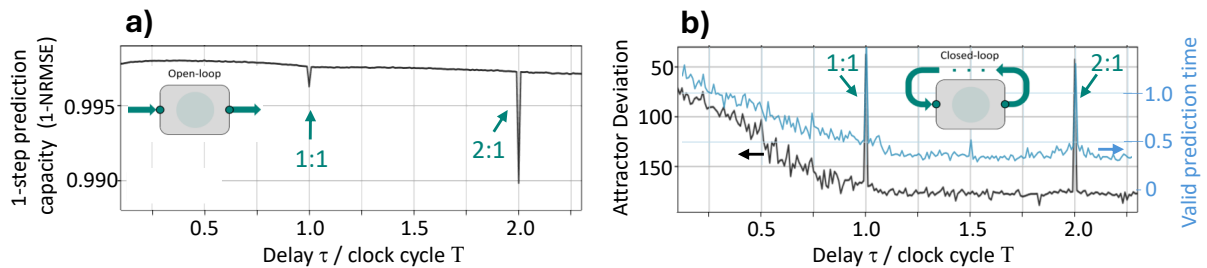


Figure 2: **Line scans** for increasing delay time  $\tau$ . (a) Prediction capacity of a delayed map for open loop 1-step ahead prediction (Lorenz xyz prediction is bad at resonances), (b) closed-loop operation of the same RC showing valid prediction time (blue) and attractor deviation (black), which are best at resonances.