

*A diverse range of factors affect the nature  
of neural representations underlying  
short-term memory*

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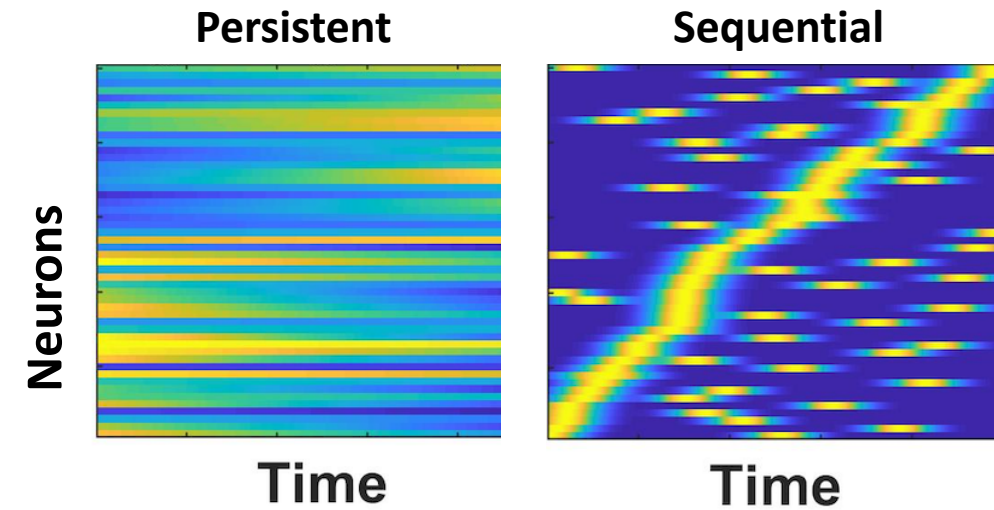
Rehan Chinoy

31 July 2020

# Research Goal and Experimental Setup

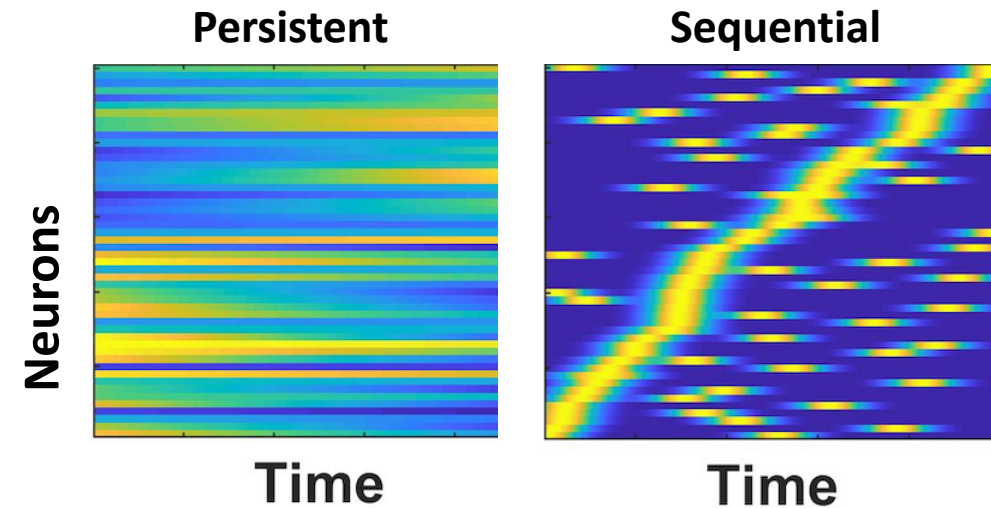
# Research Goal

- ❖ Persistent and sequential activity models are two prominent models of short-term memory in neural circuits.



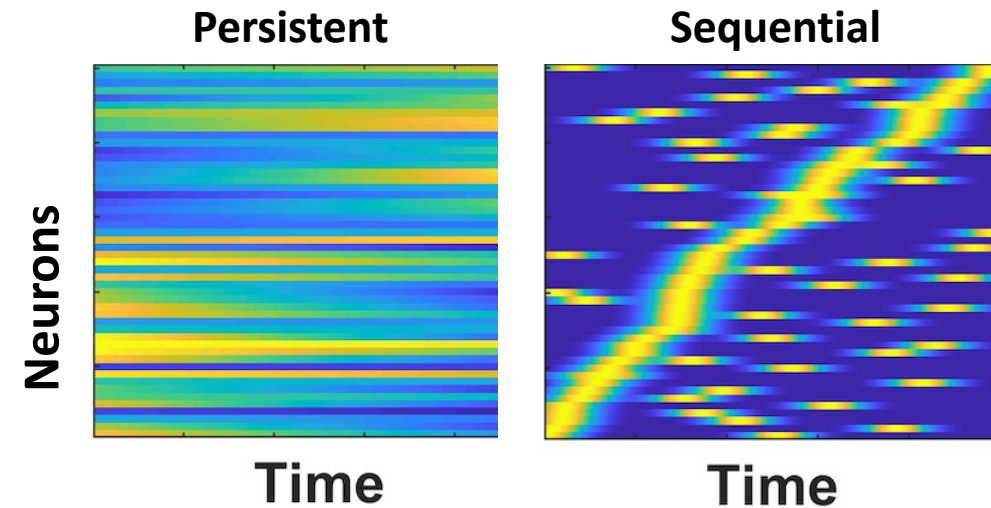
# Research Goal

- ❖ Persistent and sequential activity models are two prominent models of short-term memory in neural circuits.
- ❖ *What is the underlying circuit mechanism that determines whether a persistent or sequential solution will emerge in the network?*



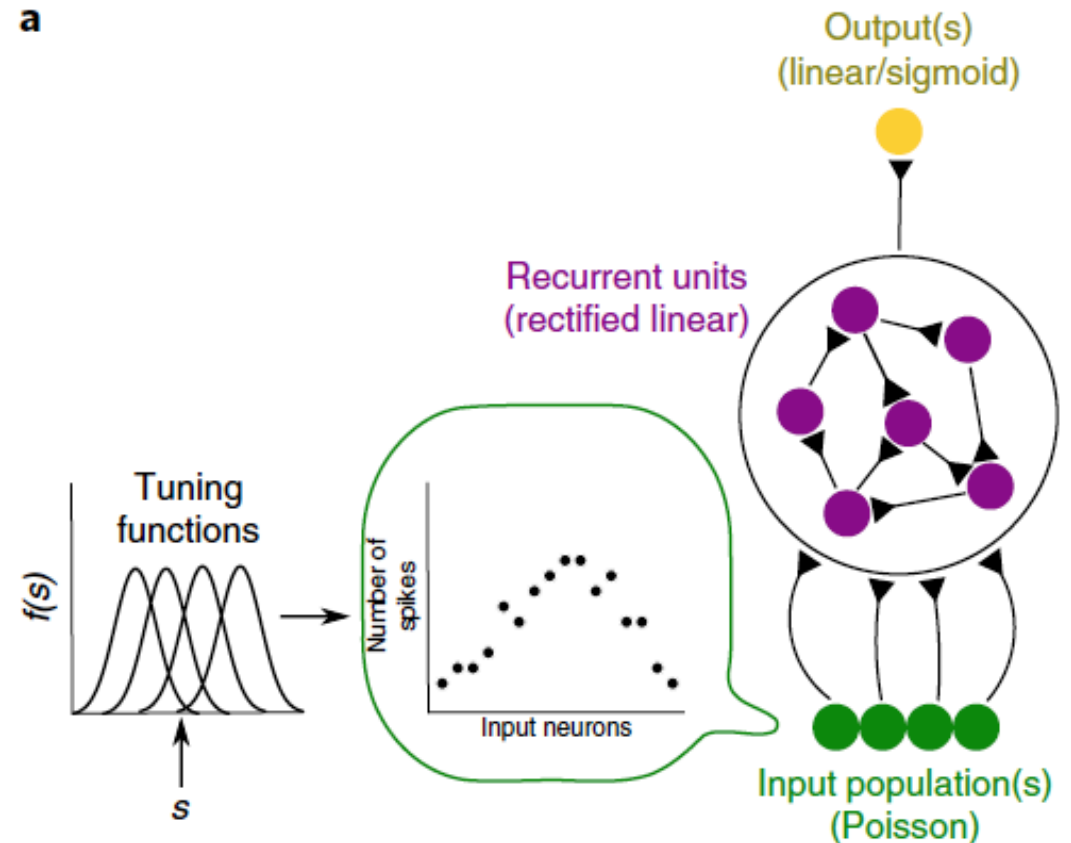
# Research Goal

- ❖ Persistent and sequential activity models are two prominent models of short-term memory in neural circuits.
- ❖ *What is the underlying circuit mechanism that determines whether a persistent or sequential solution will emerge in the network?*
- ❖ Authors address this question by training RNNs on several short-term memory tasks under a wide range of circuit and task manipulations.



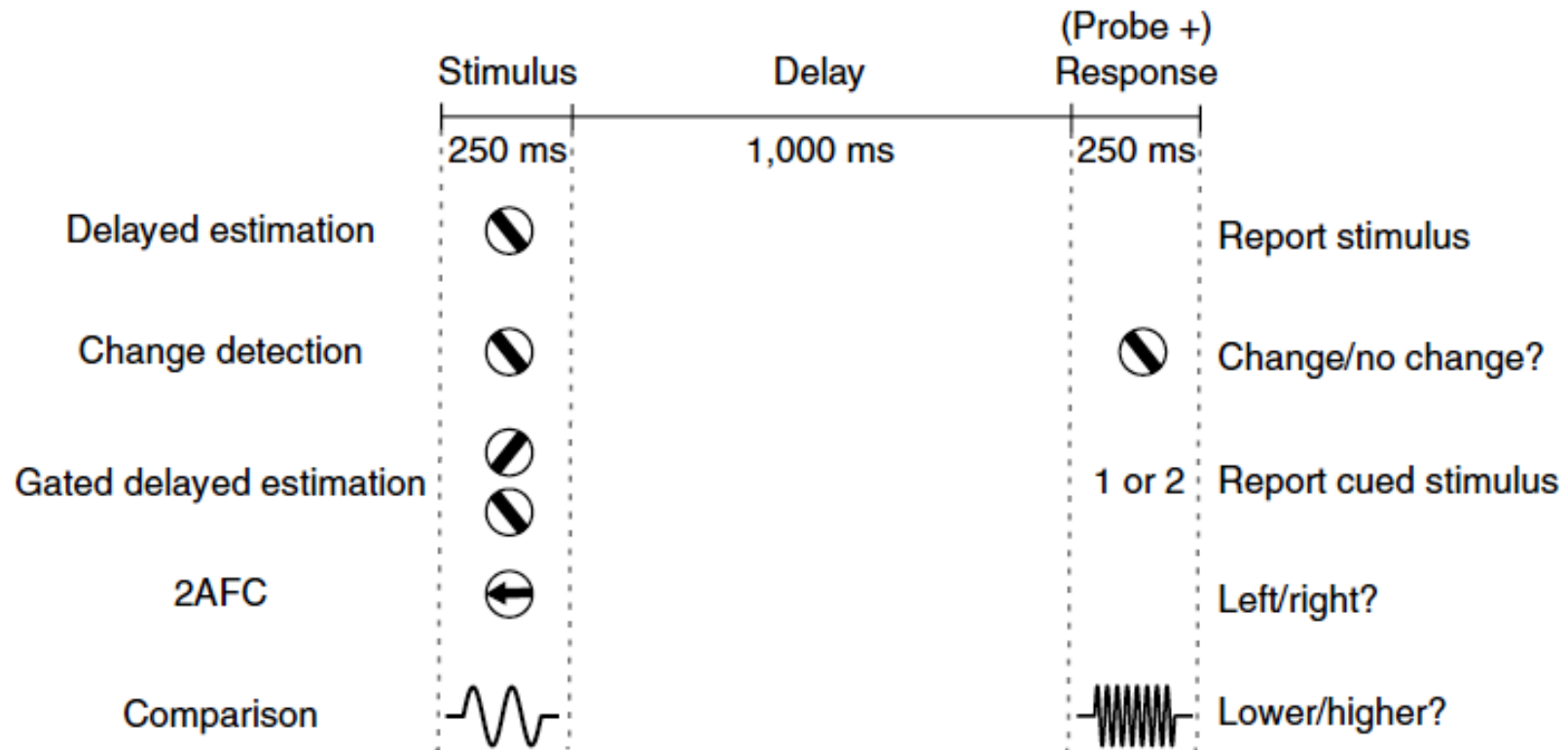
# Networks

- ❖ Each network is trained to learn a short-term memory task
- ❖ Tuning functions map a stimulus to the firing rate of input neurons
- ❖ Poisson input neurons fire independently at each time step of the task
- ❖ ReLU units clip activity such that the lower bound is zero



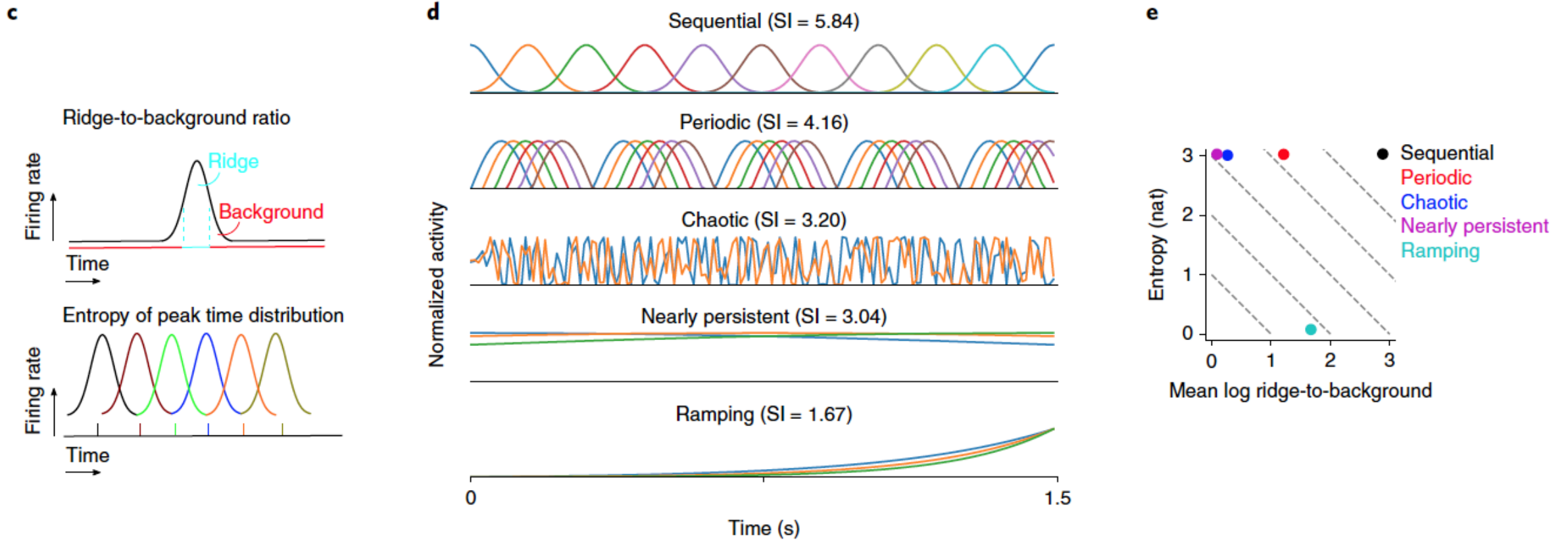
# Tasks

**b**



# Sequentiality Index

$SI = (\text{entropy of peak response time distribution of the recurrent neurons}) + (\text{mean log ridge-to-background ratio of the neurons})$

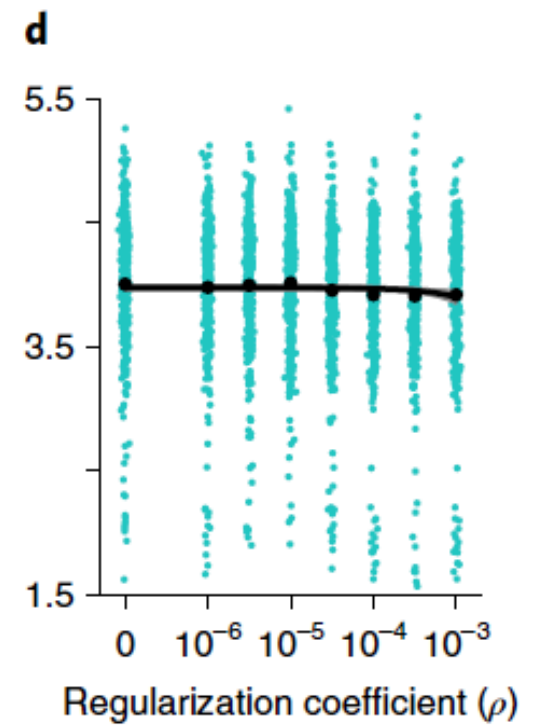
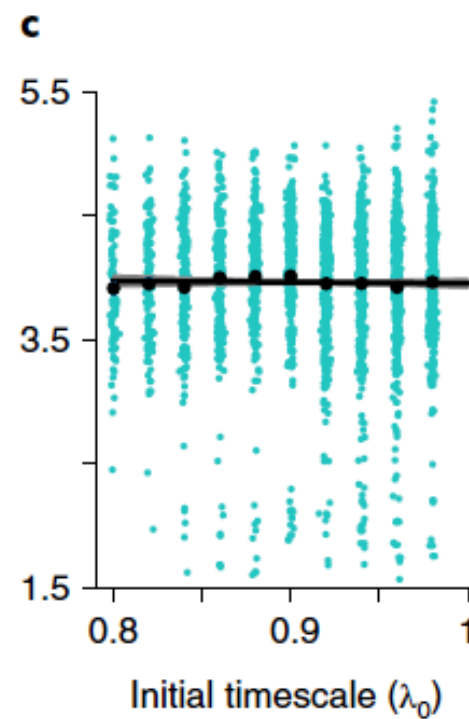
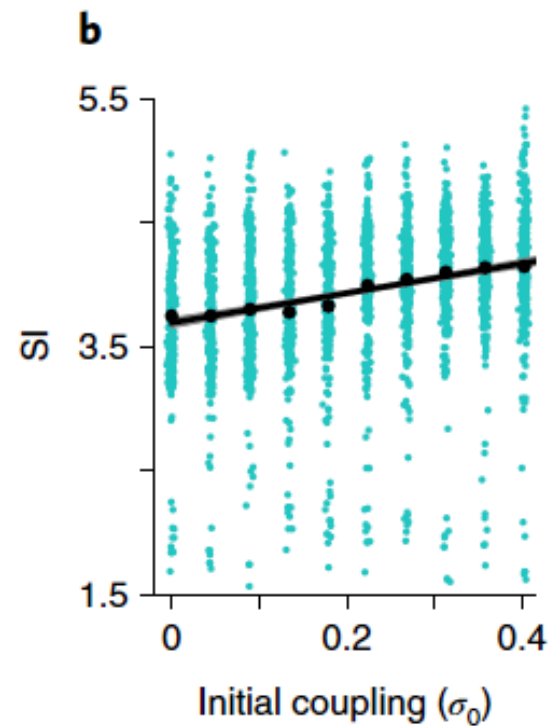
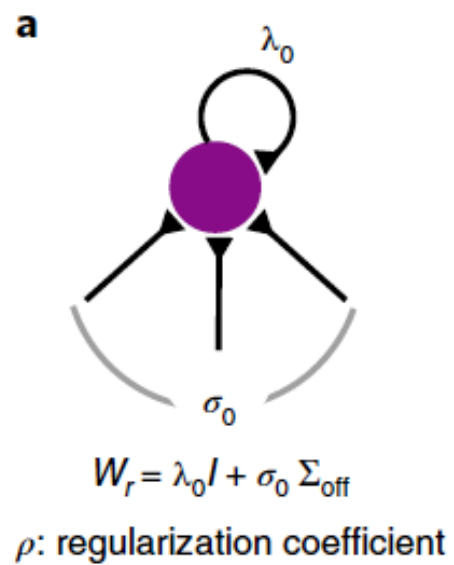




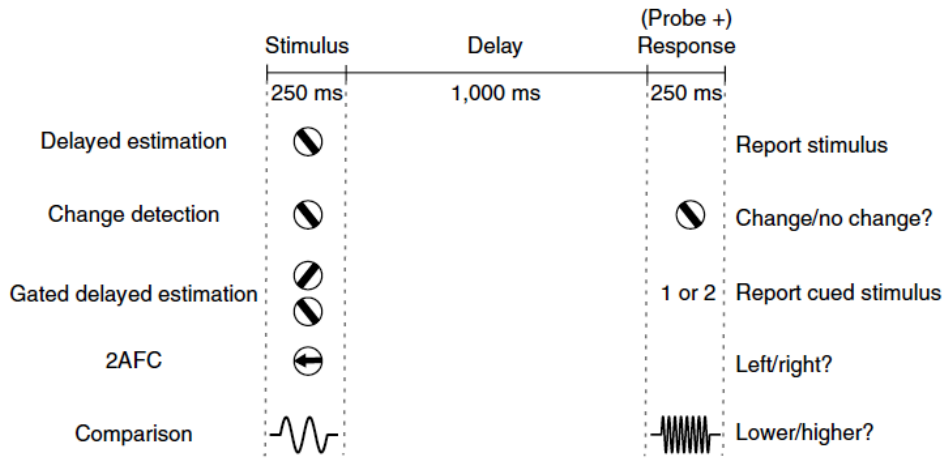
# Factors Affecting SI of Network Solutions

- ❖ Intrinsic circuit properties
- ❖ Temporal complexity of tasks
- ❖ Hebbian STSP
- ❖ Delay duration variability

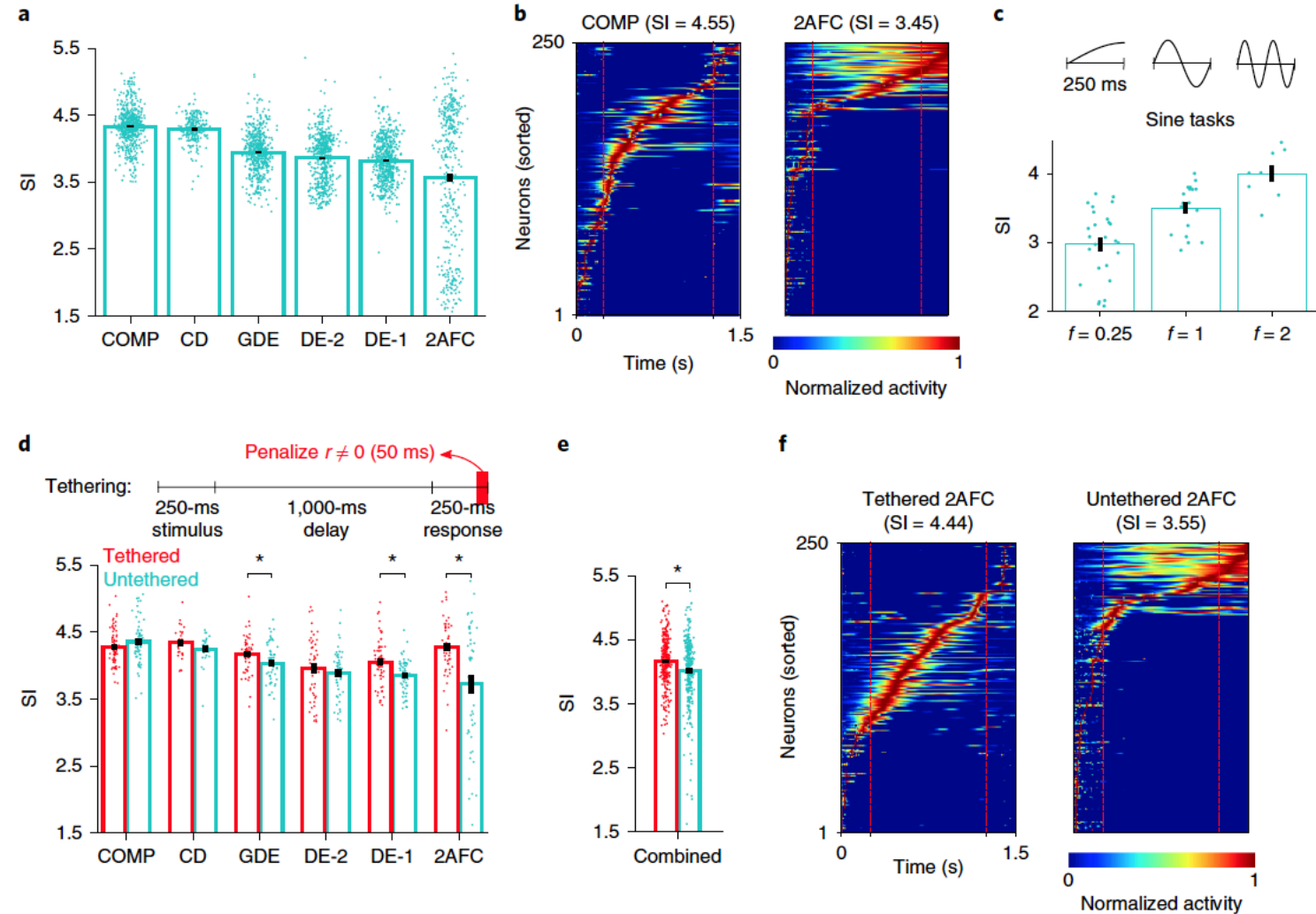
# Intrinsic circuit properties



# Temporal complexity of tasks increases SI



Hypothesis: increasing target function complexity (i.e. frequency) increases sequentiality of the network solution



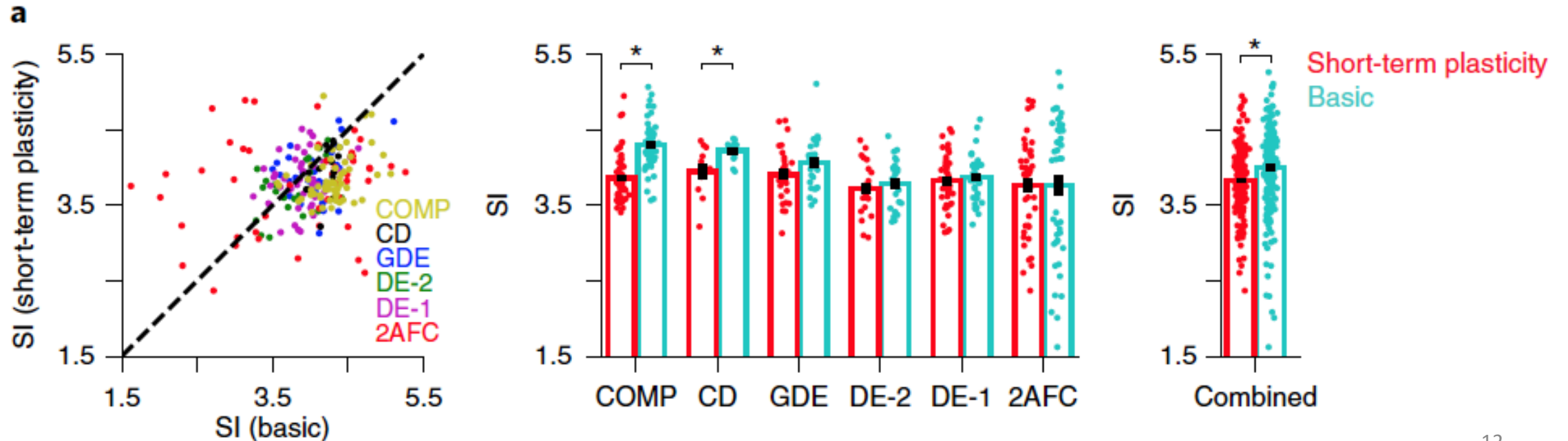
# Symmetric Hebbian STSP decreases SI

Basic discrete-time formulation:  $\mathbf{r}_t = f(W_r \mathbf{r}_{t-1} + W_h \mathbf{h}_t + \mathbf{b})$

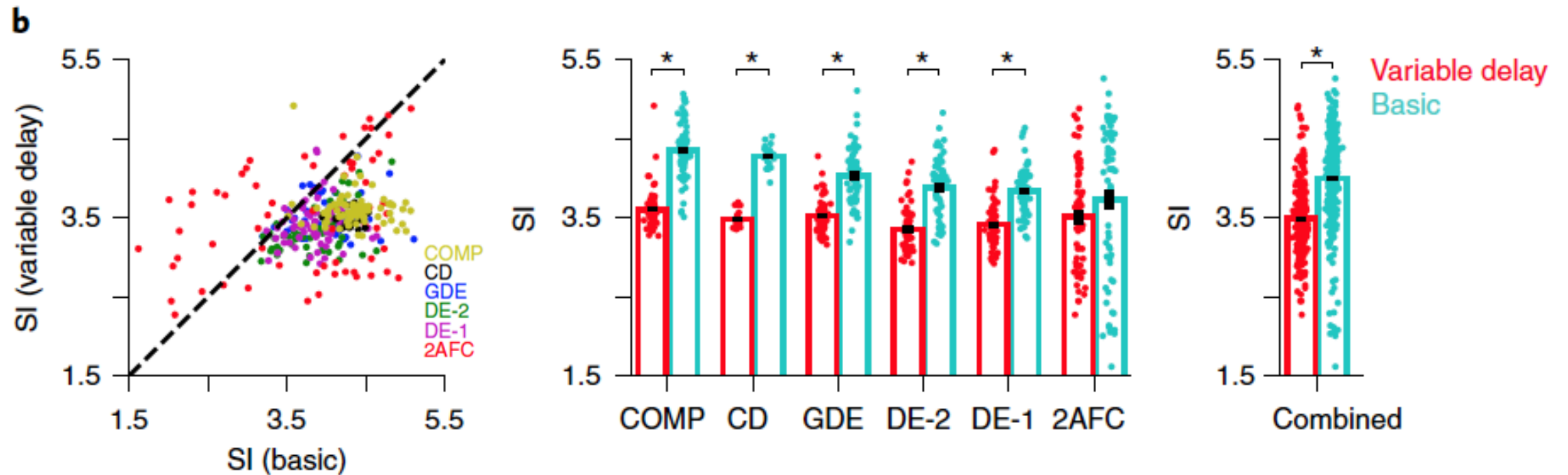
With Hebbian STSP,  $\mathbf{r}_t = f\left(\left[W_r + \sum_{\tau=1}^T \gamma^\tau \mathbf{r}_{t-\tau-1} \mathbf{r}_{t-\tau-1}^\top\right] \mathbf{r}_{t-1} + W_h \mathbf{h}_t + \mathbf{b}\right)$

\* Where  $f$  is ReLU,  $\gamma=0.0007$  is Hebbian contribution, and  $\kappa()$  is a clipping function for stability

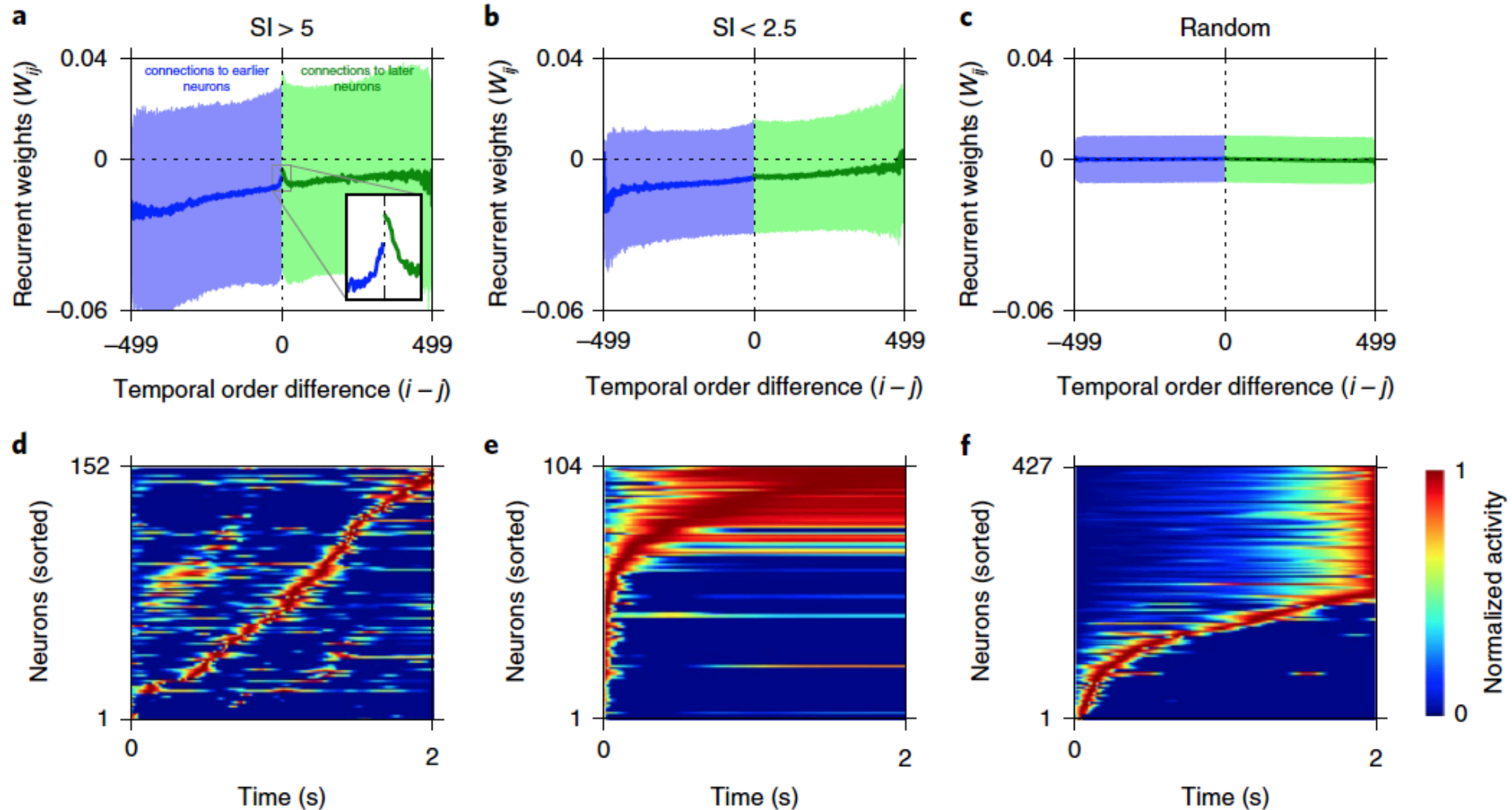
$T > 1$  is unstable, so set  $T=1$ :  $\mathbf{r}_t = f(W_r \mathbf{r}_{t-1} + \gamma \kappa(\mathbf{r}_{t-2}^\top \mathbf{r}_{t-1}) \mathbf{r}_{t-2} + W_h \mathbf{h}_t + \mathbf{b})$



# Delay duration variability decreases SI



# Circuit mechanism that generates sequential versus persistent activity



# Concluding Remarks

- ❖ This paper establishes a mechanism for the maintenance of STM and shows that persistent and sequential solutions are ends of a spectrum that emerges from training
- ❖ Why does the network's mechanism for maintaining short-term memory rely on non-normal dynamics?
- ❖ Is the model biologically realistic?