

# What is a moment?

J. J. Hopfield and Carlos D. Brody

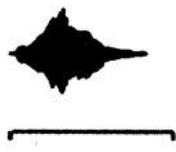
Ca. 2001

# Contents

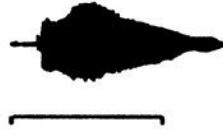
1. Brief discussion of problem + other solutions (context)
2. Behavior of current network
3. Context for this kind of behavior
4. Behavior of single-neurons in network
5. Conclusion

# How do neural networks “integrate” information across time?

"one"  
(speaker a)



"three"  
(speaker b)

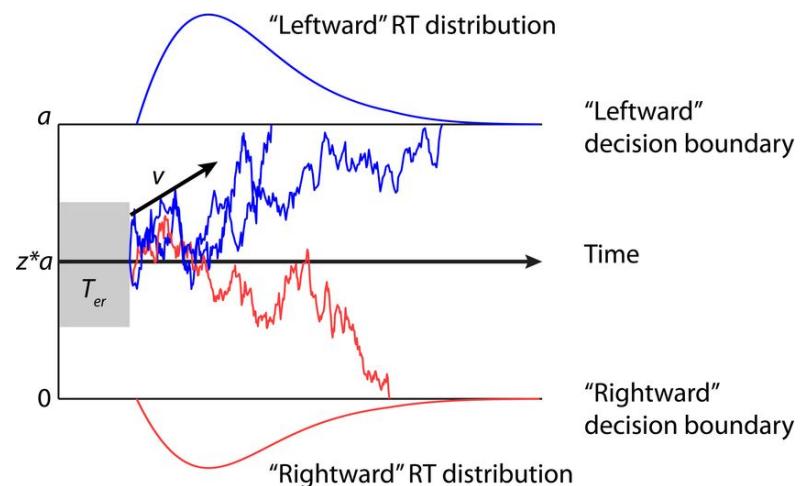


Transmission delays can only account for integrations on the order of 10s of milliseconds

# How do neural networks “integrate” information across time?

- Still ongoing topic of research
- Evolving dynamics converge onto a solution.
  - Illustration: Drift-diffusion
- What kind of networks are required for these non-linear solutions?
  - Drift diffusion network → Neural implementation?
  - Attractor networks
    - Point vs. ongoing stimuli
- Multi-scale / Scale-invariant behavior

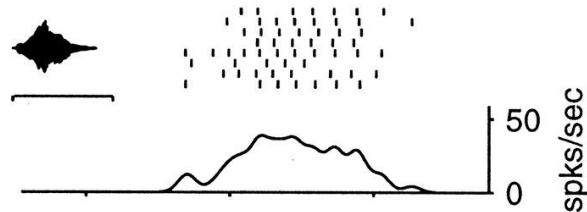
Drift-diffusion model of decision making



# Response of 1 neuron to different stimuli

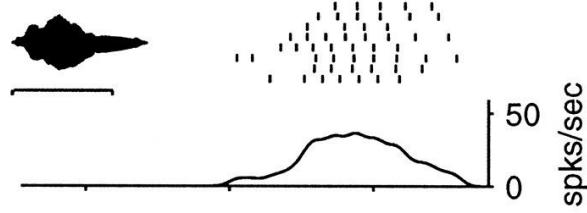
a

"one"  
(speaker a)



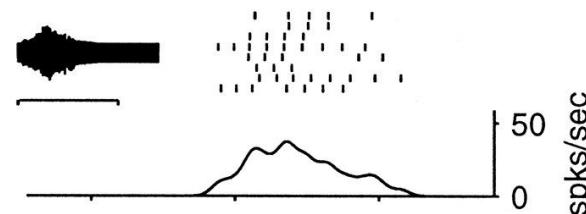
b

"one"  
(speaker b)



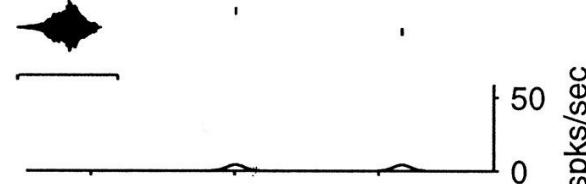
c

"one" + tone  
(speaker a)



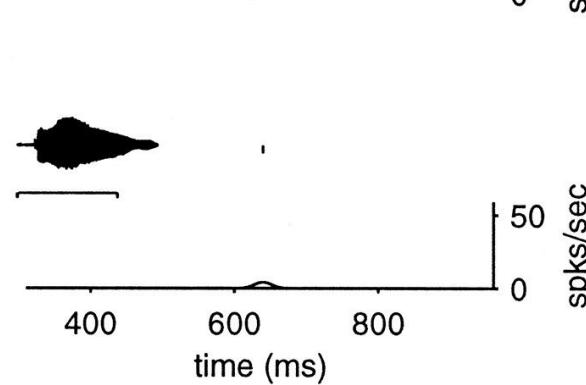
d

reversed "one"  
(speaker a)



e

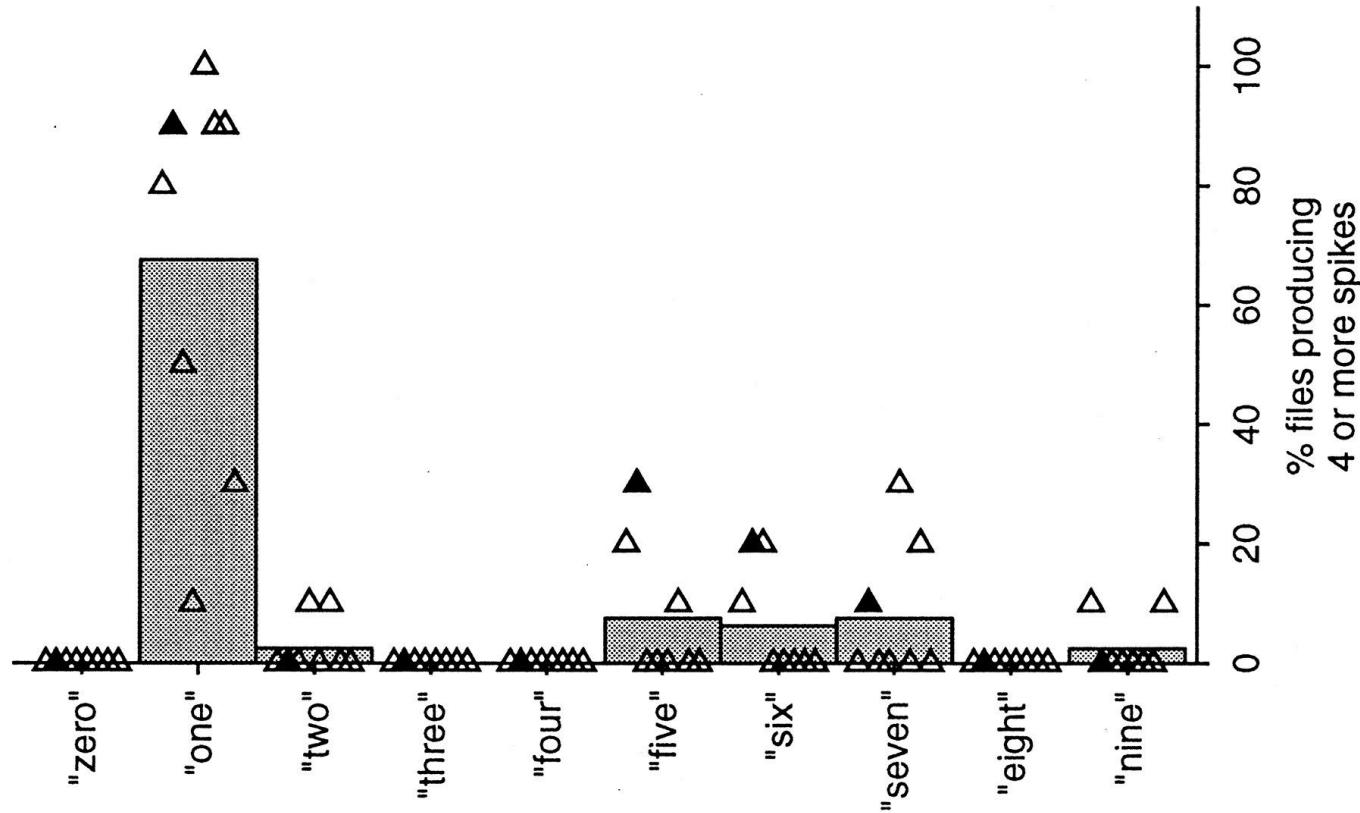
"three"  
(speaker b)



Notes:

- Resistant to warp + different speakers
- Resistant to background noise
- Specific to “directionality” of stimulus

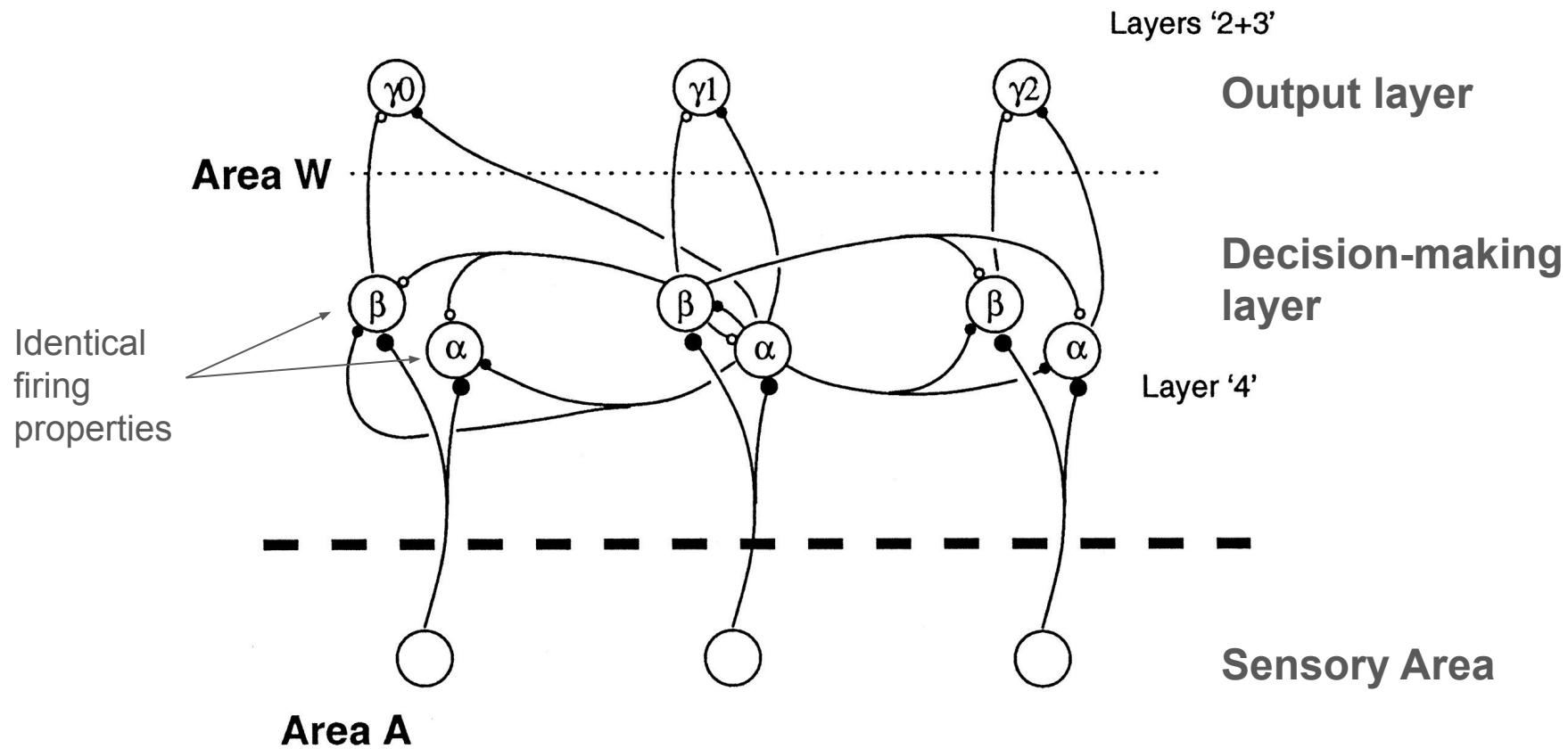
# Accuracy of “one” neuron



# How does it work?

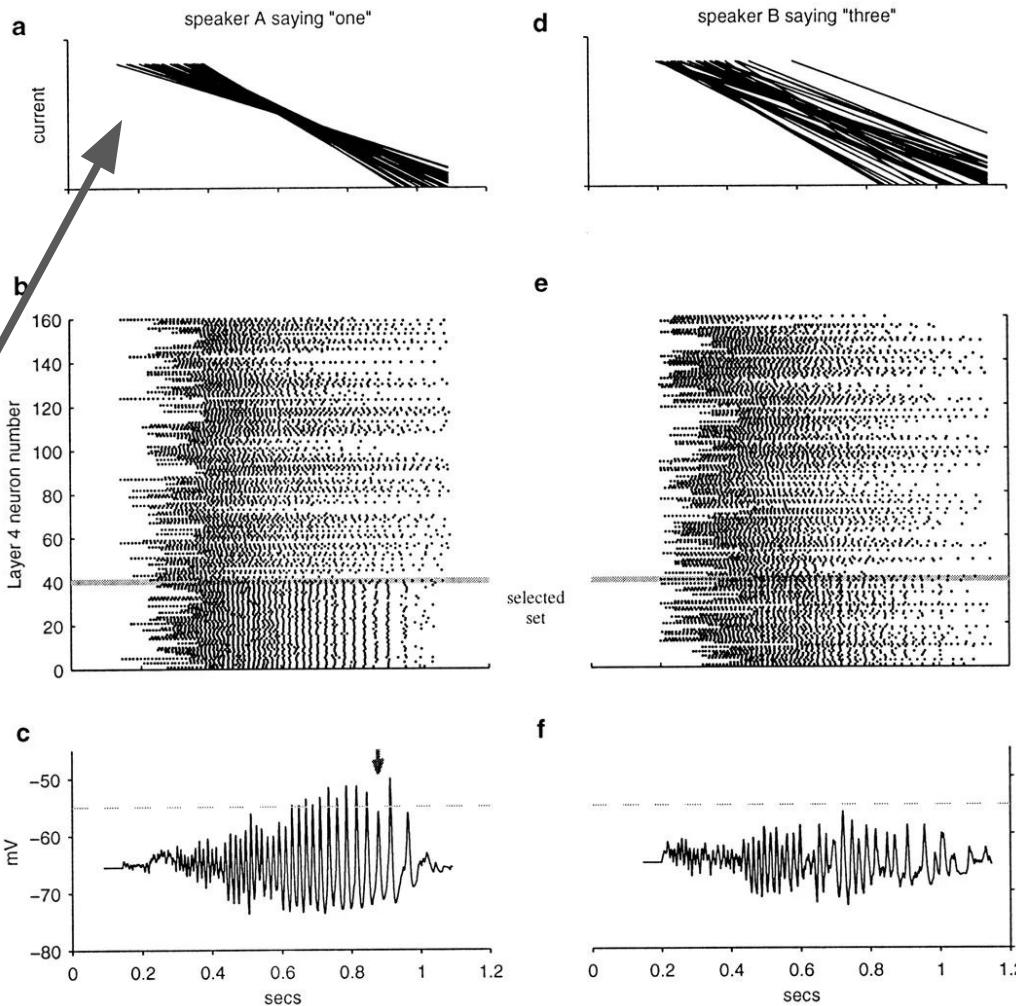
- Coincidence of variable timescales leading to neural synchronization
- Can be performed by simple leaky integrate-and-fire neurons
- No dependence on specific cell-types
- Varying synaptic strength randomly by “ $\pm 50\%$  has no appreciable effect on the response”.

# Schematic of Network Structure



Synchronous  
input activates  
downstream  
decision  
neurons

Different onset  
times + different  
decay slopes



Decision-making layer

Output neuron

# Other literature on utility of synchronization

- Is this commonly seen?
  - Expected from Hebbian plasticity

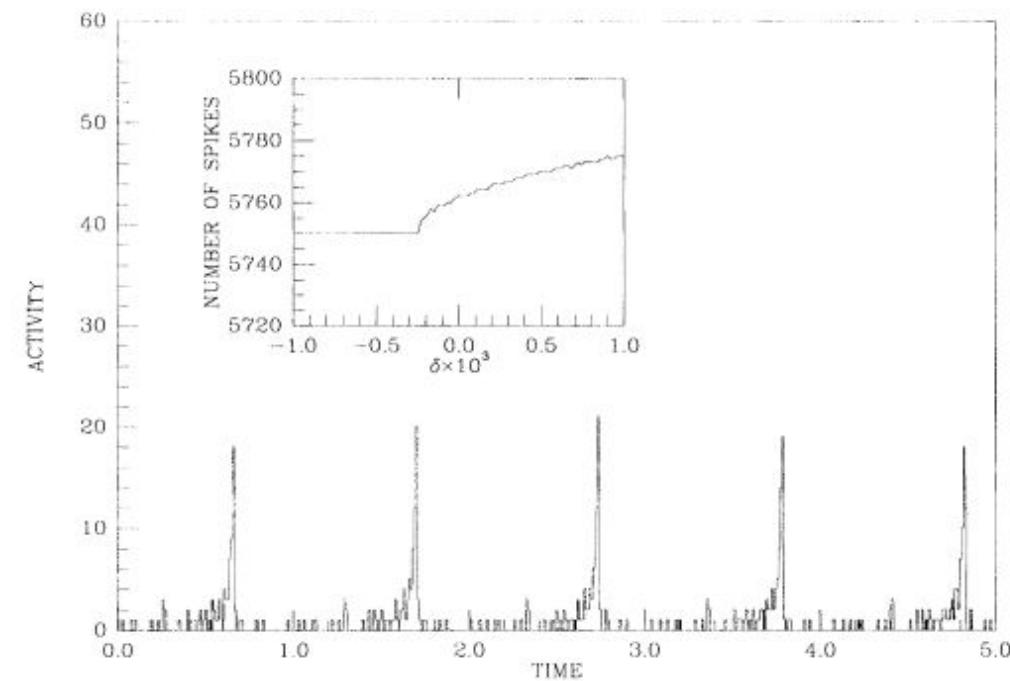


FIG. 2. Simulated activity of an inhomogeneous network of 100 neurons, with  $\Delta = 10^{-3}$ . Other parameters are the same as in Fig. 1. The number of spikes emitted by the network in the time interval  $dt = 0.01$  is plotted vs time. The transient up to  $t = 5000$  was discarded. Inset: the number of spikes emitted by a neuron in a time window of length 6000, against the deviation of its local external current  $I^0$  from the mean  $I$ .

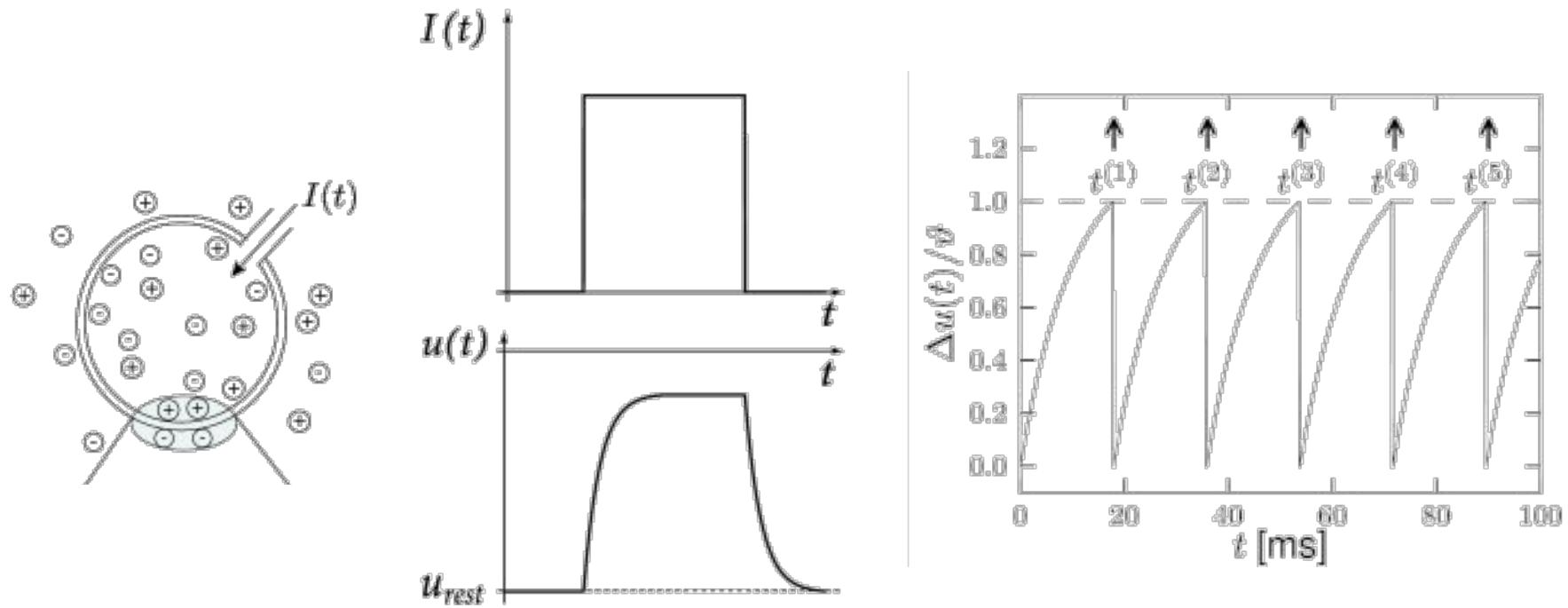
# Other literature on utility of synchronization

- Plenty of interest in measuring synchrony:
  - Kreuz, Thomas, Daniel Chicharro, et al. “Monitoring Spike Train Synchrony.” *Journal of Neurophysiology*, vol. 109, no. 5, Mar. 2013, pp. 1457–72. DOI.org (Crossref), <https://doi.org/10.1152/jn.00873.2012>.
  - Kreuz, Thomas, Mario Mulansky, et al. “SPIKY: A Graphical User Interface for Monitoring Spike Train Synchrony.” *Journal of Neurophysiology*, vol. 113, no. 9, May 2015, pp. 3432–45. PubMed Central, <https://doi.org/10.1152/jn.00848.2014>.
  - Lama, Nikesh, et al. “Spike Train Synchrony Analysis of Neuronal Cultures.” 2018 International Joint Conference on Neural Networks (IJCNN), 2018, pp. 1–8. IEEE Xplore, <https://doi.org/10.1109/IJCNN.2018.8489728>.
  - Mulansky, Mario, and Thomas Kreuz. “PySpike—A Python Library for Analyzing Spike Train Synchrony.” *SoftwareX*, vol. 5, Jan. 2016, pp. 183–89. ScienceDirect, <https://doi.org/10.1016/j.softx.2016.07.006>.
  - Satuvuori, Eero, et al. “Measures of Spike Train Synchrony for Data with Multiple Time Scales.” *Journal of Neuroscience Methods*, vol. 287, Aug. 2017, pp. 25–38. PubMed Central, <https://doi.org/10.1016/j.jneumeth.2017.05.028>.

# Other literature on utility of synchronization

- Relation to actual brain function?
  - Signal propagation...improve signal-to-noise ratio
  - Seen in coordination of brain regions (putting cart before horse?)

# Detour: Leaky Integrate-and-fire neurons

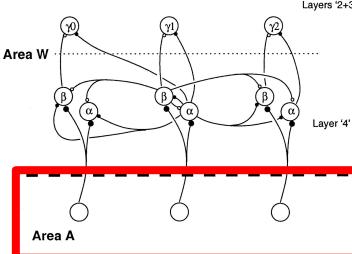


# Properties of the network

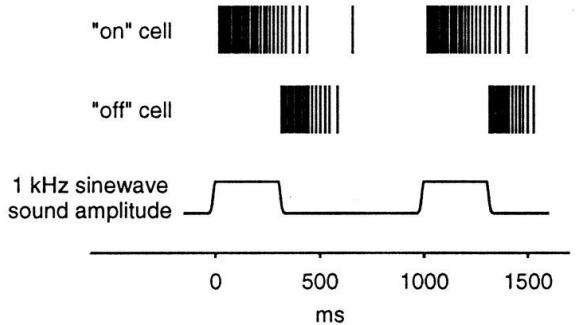
- “Good” specificity to input
- Invariance to temporal warp of input

Note: Goal of paper was not to improve on the state-of-the-art of neural networks, but to demonstrate a mechanism of network function.

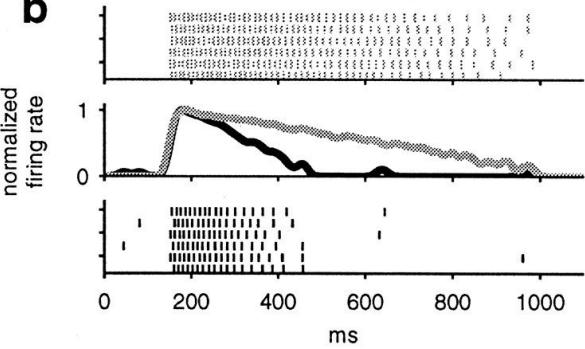
# Response of “sensory” neurons (Area A)



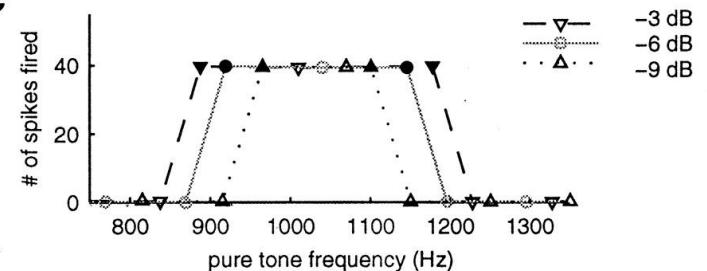
a



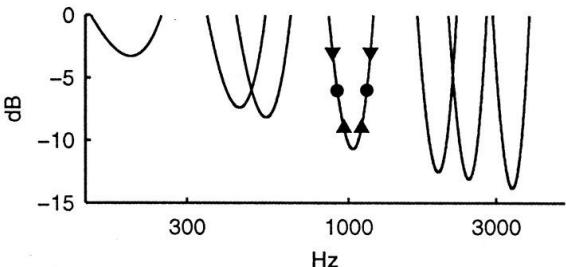
b



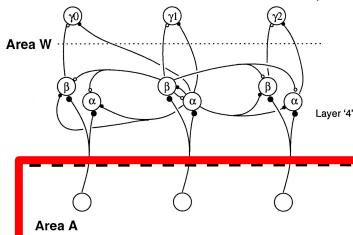
c



d

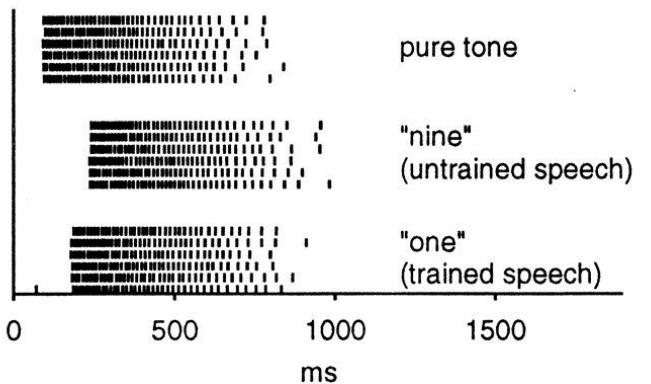


- All-or-nothing responses

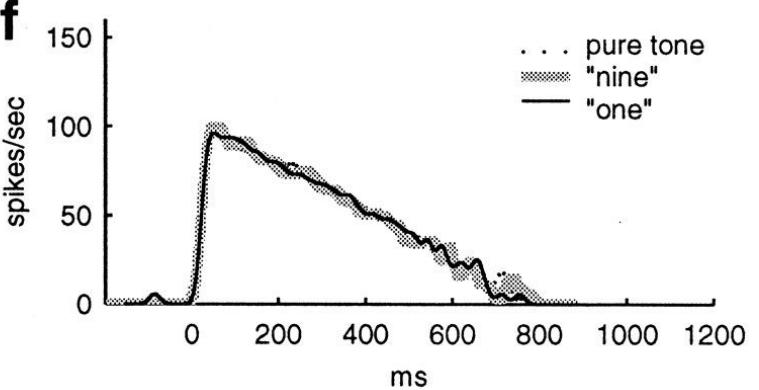


# Response of “sensory” neurons

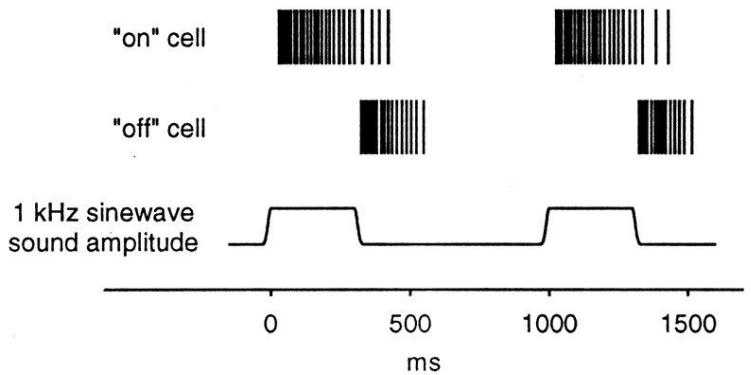
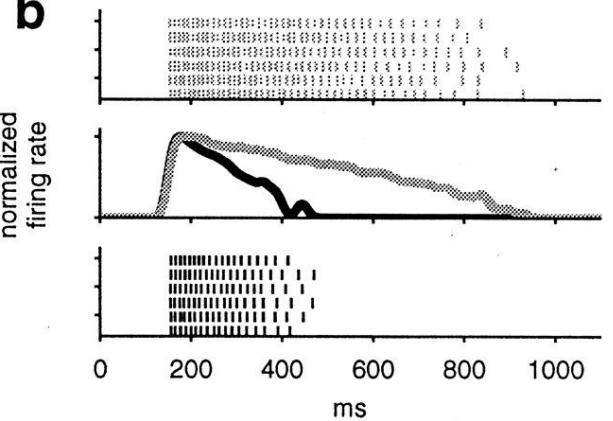
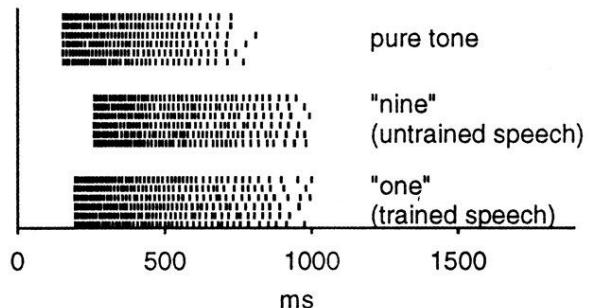
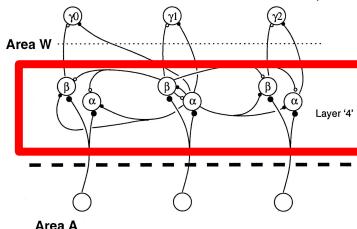
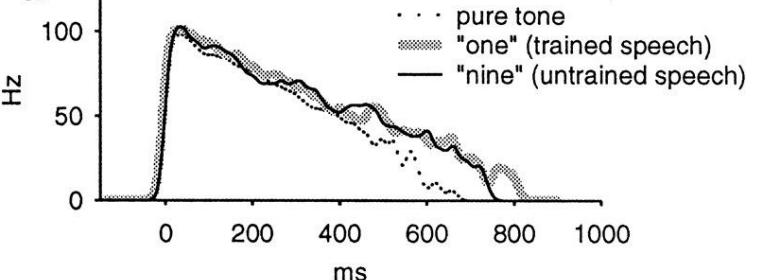
e



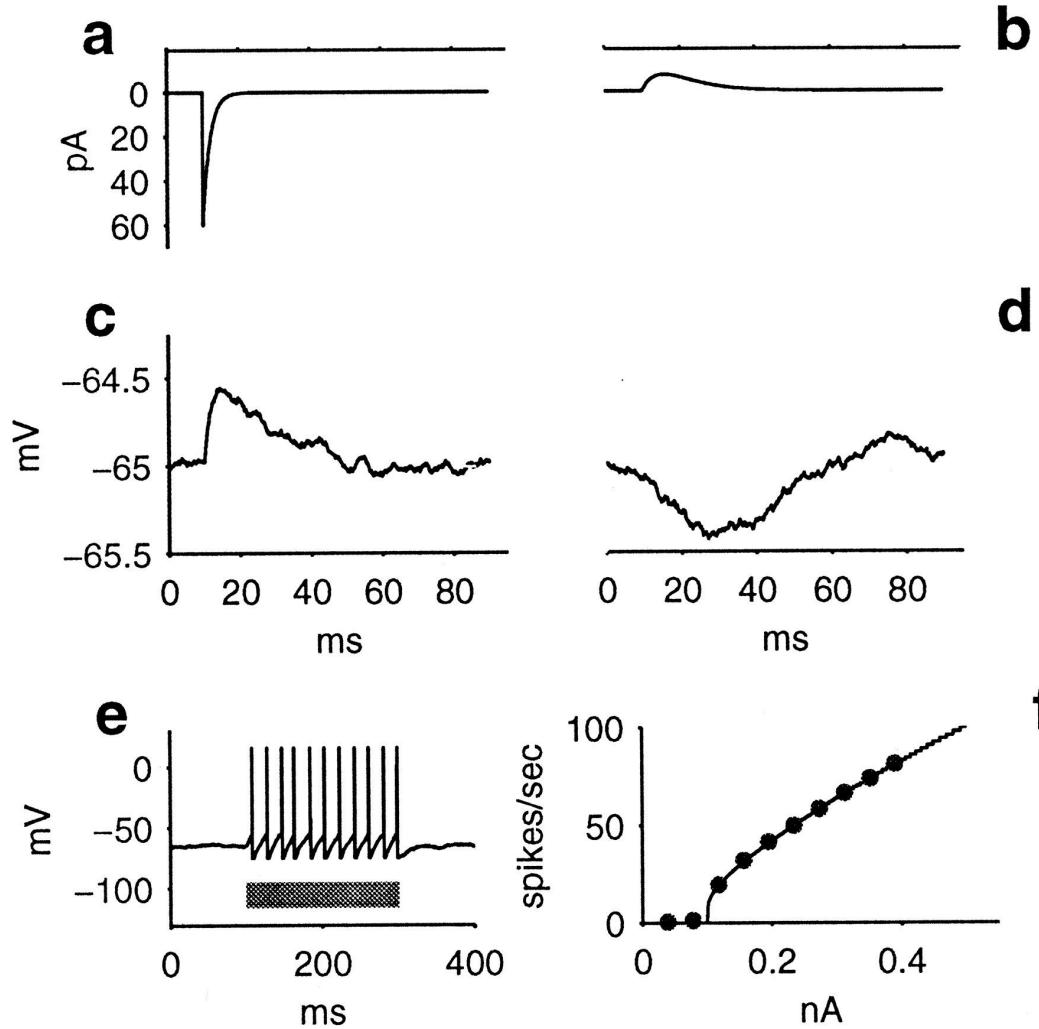
f



# Response of Decision Neurons (Area W, Layer 4)

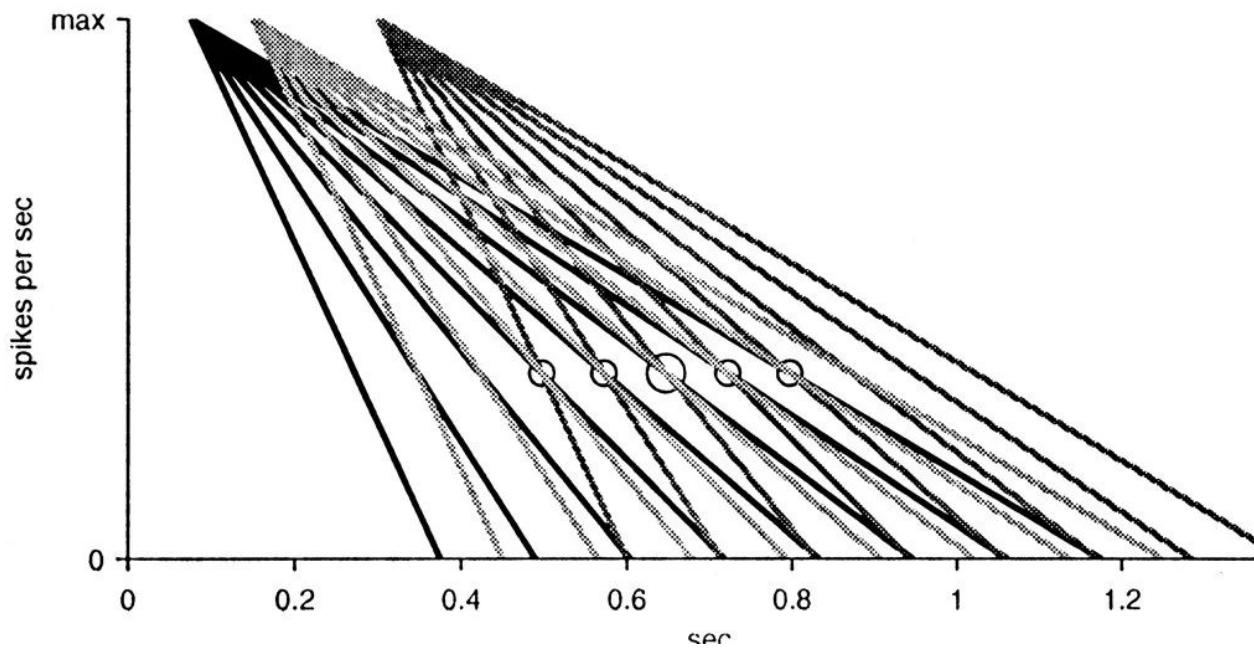
**a****b****c****d**

# Synaptic Properties of Circuit

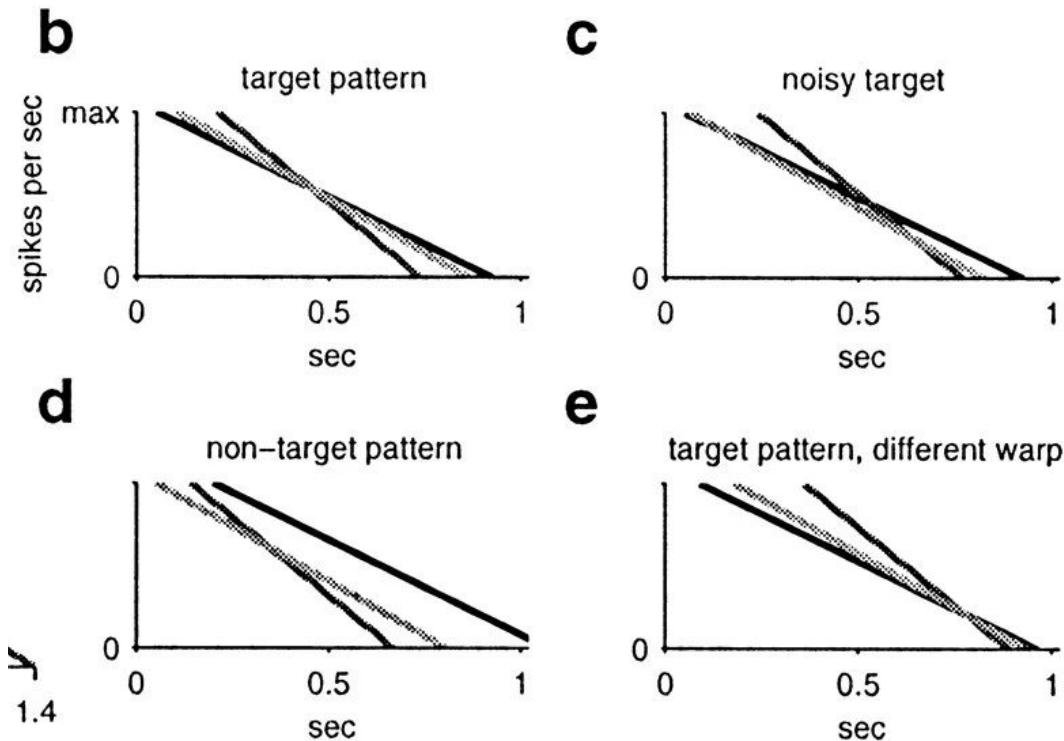


# Propagation of activity based on coincidence of multiple timescales

a

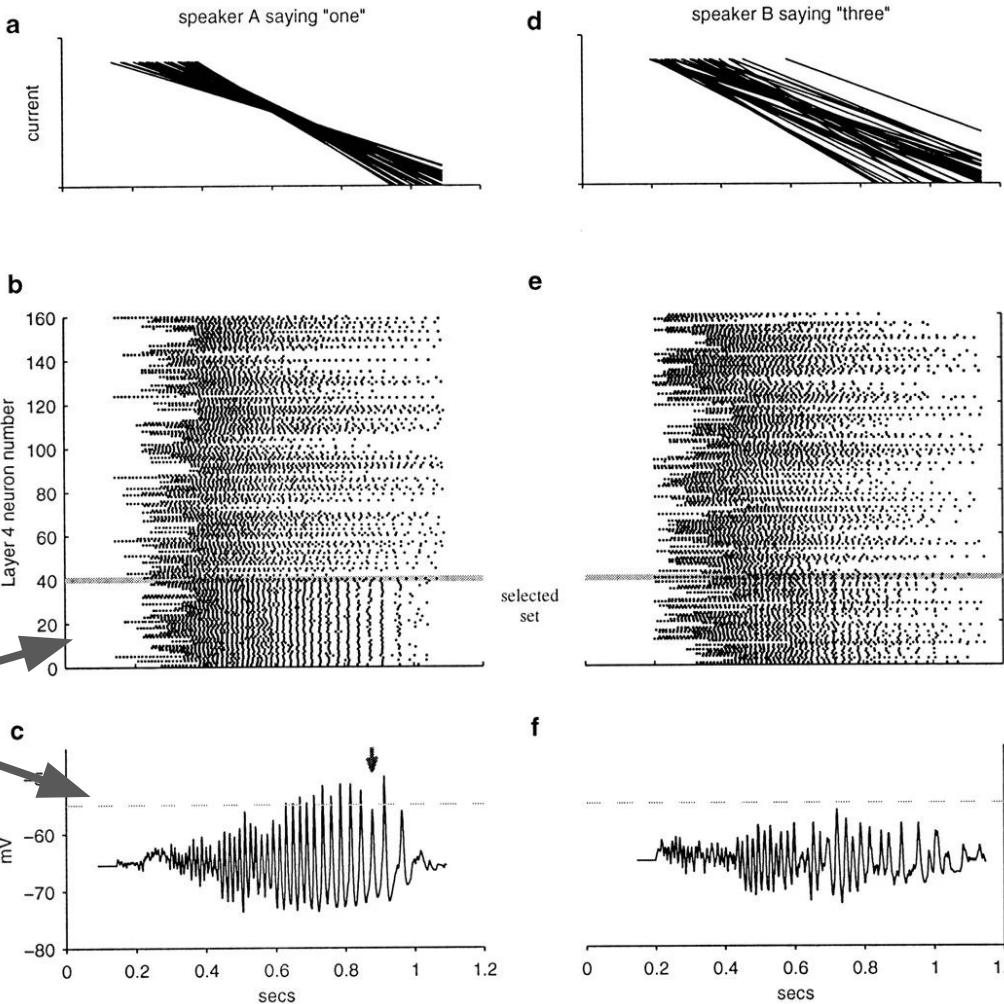


# Propagation of activity based on coincidence of multiple timescales



# Synchronous input activates downstream decision neurons

Given biological parameters (not discussed), synchronization results in gamma-frequency activity



Decision-making layer

Output neuron

# “Training” the network

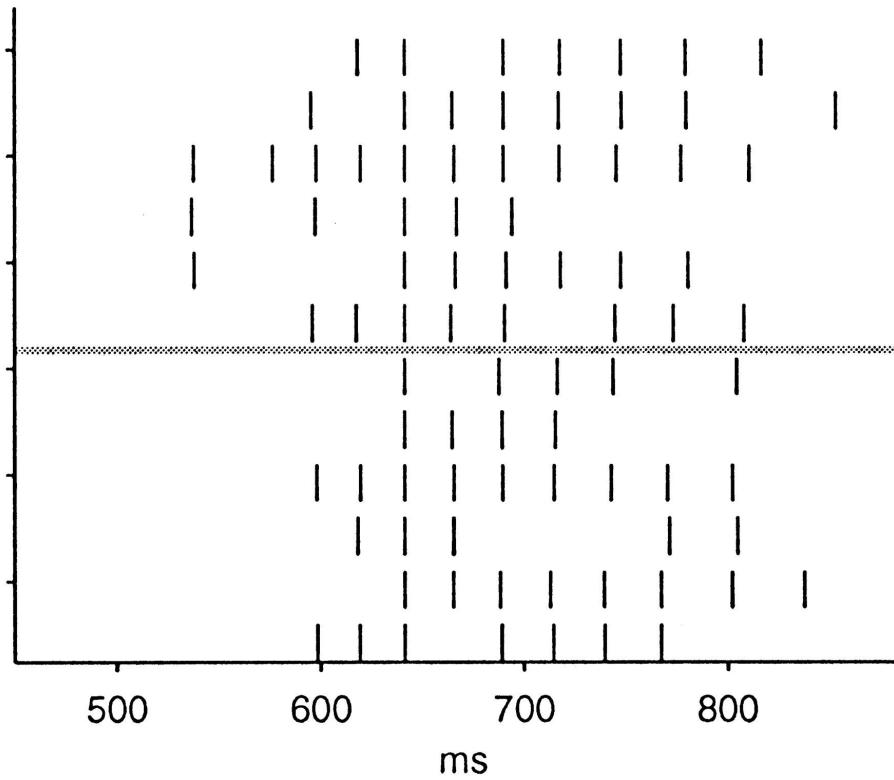
1. Find neurons with converging firing rates in response to a stimulus
2. Creating mutual all-to-all connections within this set

Kinda-sorta like a reservoir network.

# How does this happen?

- Proper synchronization (spike-by-spike) can ONLY happen at similar firing rates
- Anything special about 60Hz?

time-aligned rasters



# Questions?

- Does synchronization happen in GC?
  - Maybe as an output signal?
  - Or to trigger a state transition?