

Computational Neuroscience: Systems Neuroscience in the Context of the Hippocampus

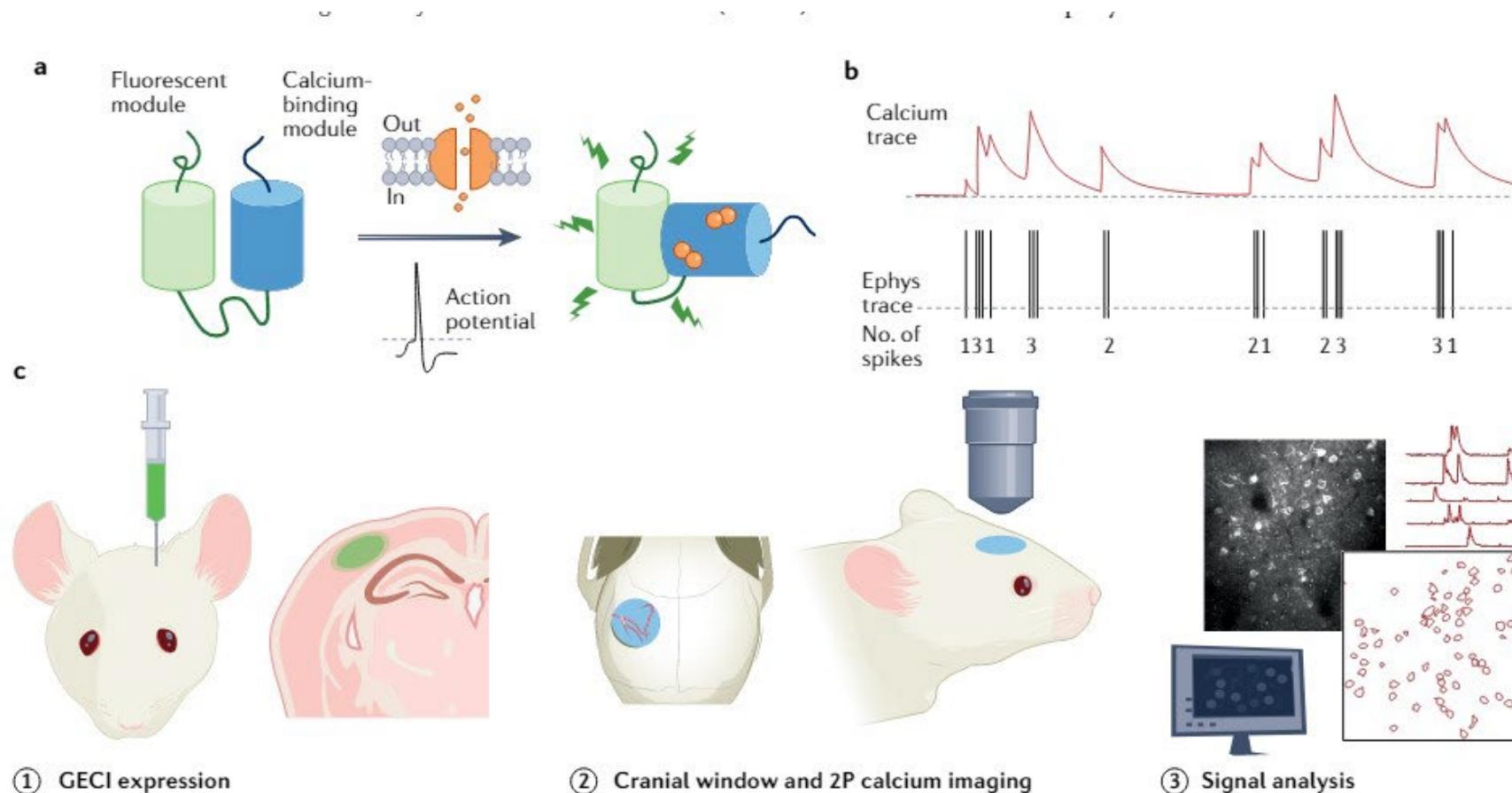
Heath Robinson, Ph.D

CWRU Undergraduate Neuroscience

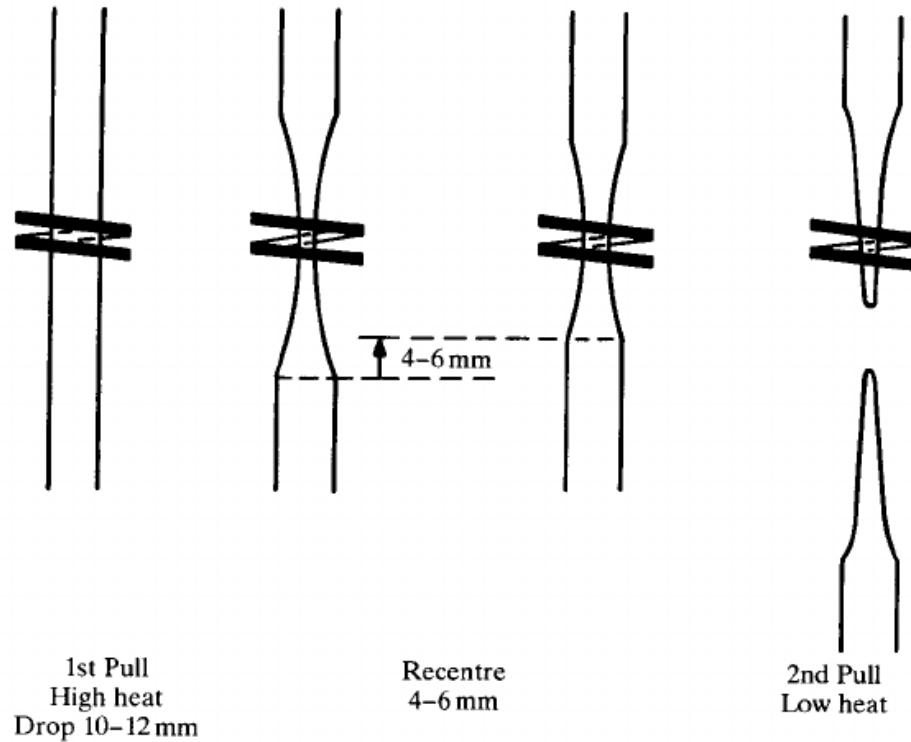
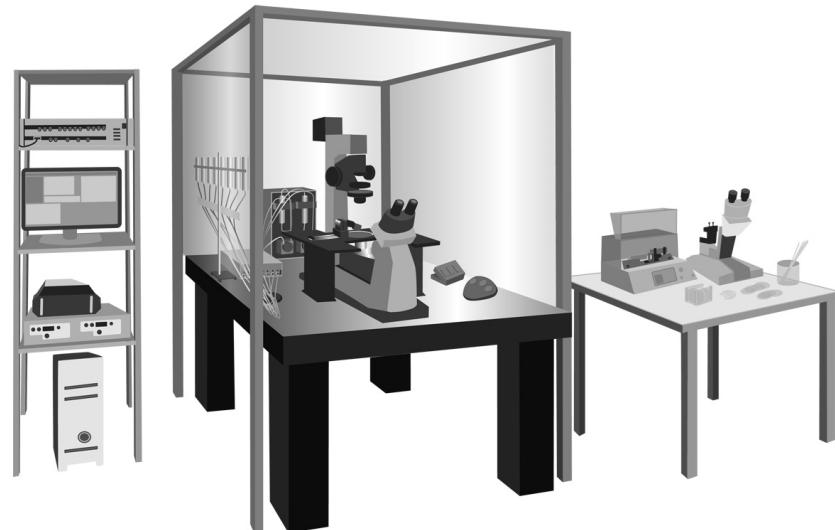
Objectives

1. Understand the different methodologies used to record the brain *in vivo*
2. Understand the differences between EEG, ECoG, and LFP
3. Understand the physiology of place cells fire and reactivation during SWRs
4. Understand the importance of cell assemblies

Techniques in Systems Neuroscience – Single cell recording – Calcium transients from Calcium Imaging

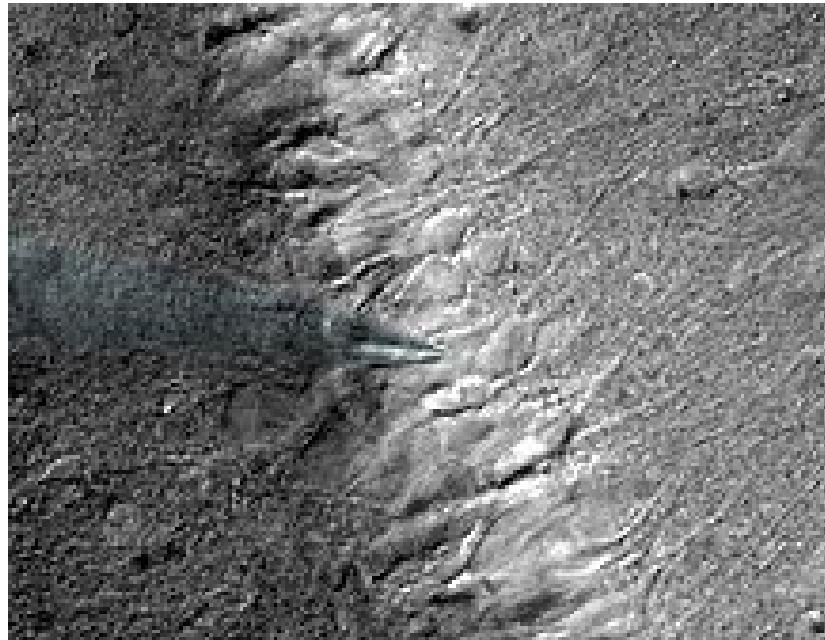


Techniques in Systems Neuroscience – Single cell recording – Patch Clamp

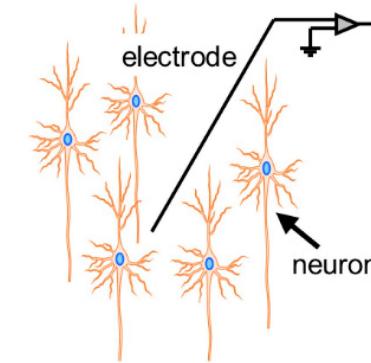


Ogden and Stanfield.(1994) *Patch clamp techniques for single channel and whole-cell recording.*

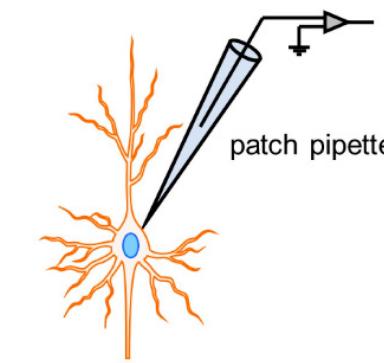
Slice In Vitro Electrophysiology



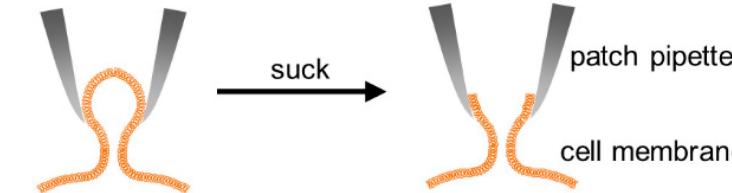
(a) Extracellular recording



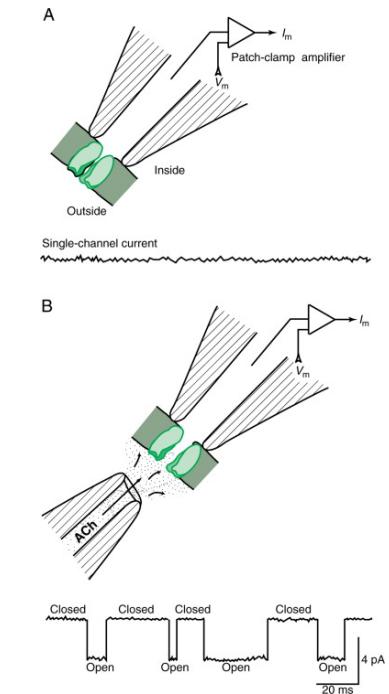
Intracellular recording



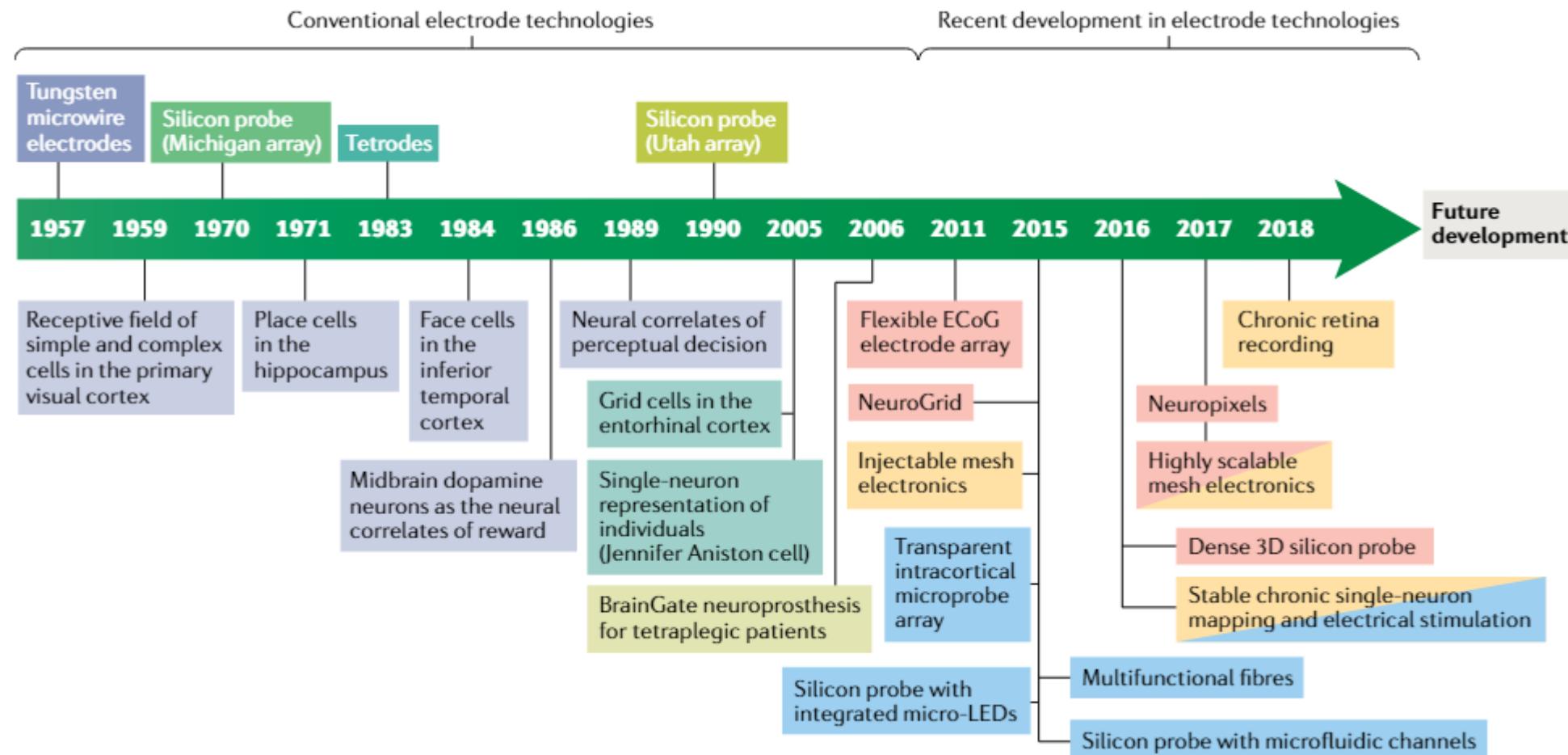
(b) Cell-attached (giga-ohm seal)



Wikipedia



Techniques in Systems Neuroscience – Single cell recording – Tetrodes and Silicon Probes



Techniques in Systems Neuroscience – Single cell recording – Tetrodes and Spike Sorting

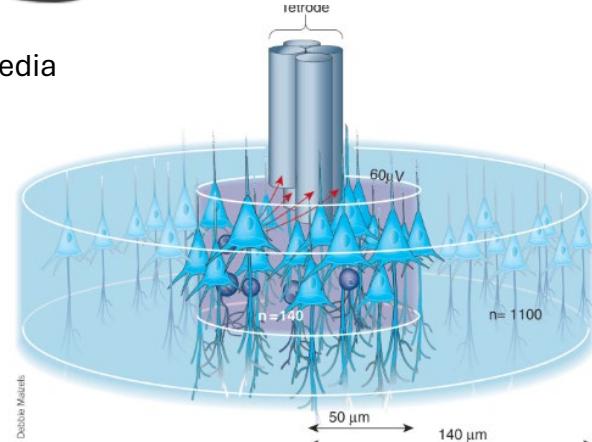
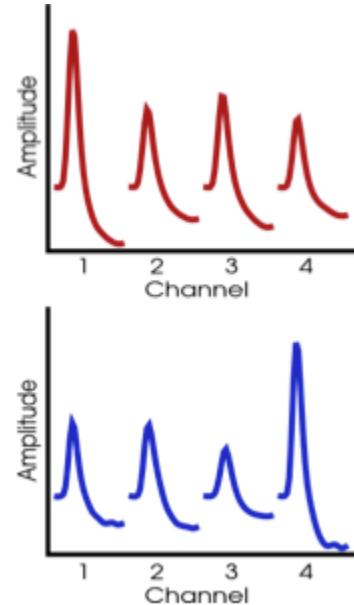
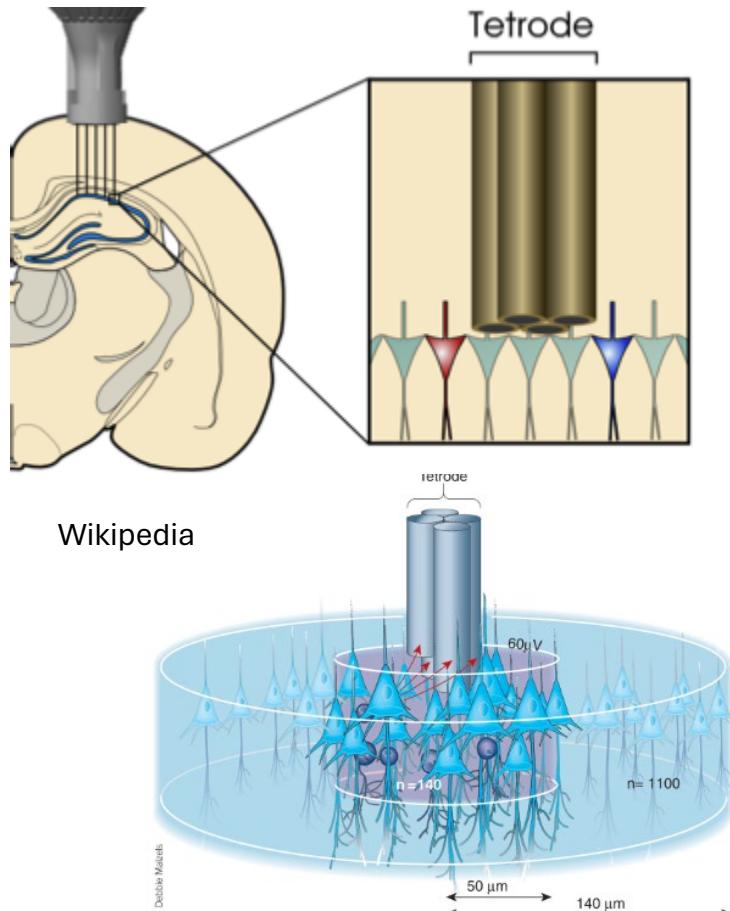
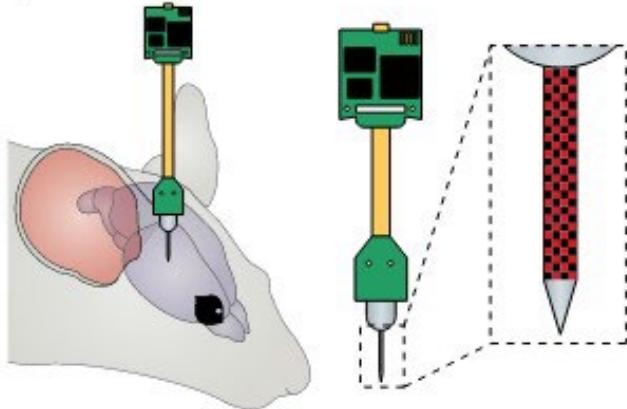


Figure 1 Unit isolation quality varies as a function of distance from the electrode. Multisite electrodes (a wire tetrode, for example) can estimate the position of the recorded neurons by triangulation. Distance of the visible electrode tips from a single pyramidal cell (triangles) is indicated by arrows. The spike amplitude of neurons ($>60 \mu\text{V}$) within the gray cylinder (50 μm radius), containing ~ 100 neurons, is large enough for separation by currently available clustering methods. Although the extracellularly recorded spike amplitude decreases rapidly with distance, neurons within a radius of 140 μm , containing $\sim 1,000$ neurons in the rat cortex^{19,21}, can be detected. Improved recording and clustering methods are therefore expected to record from larger number of neurons in the future. (Data are derived from simultaneous extracellular and intracellular recordings from the same pyramidal cells from ref. 19.)

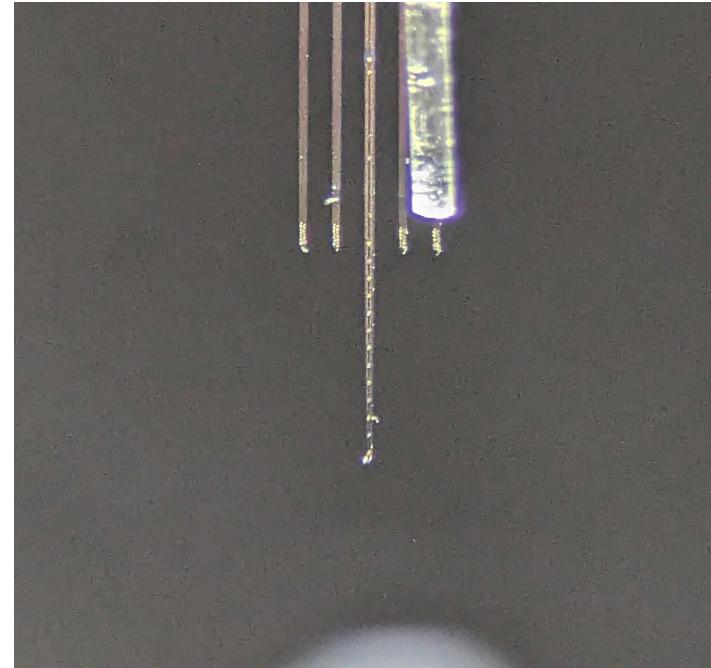
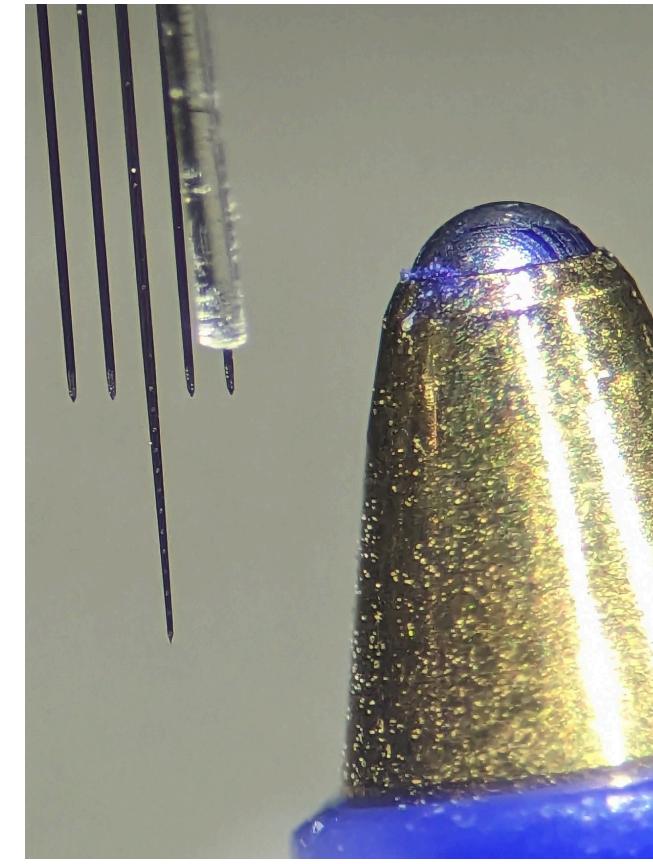
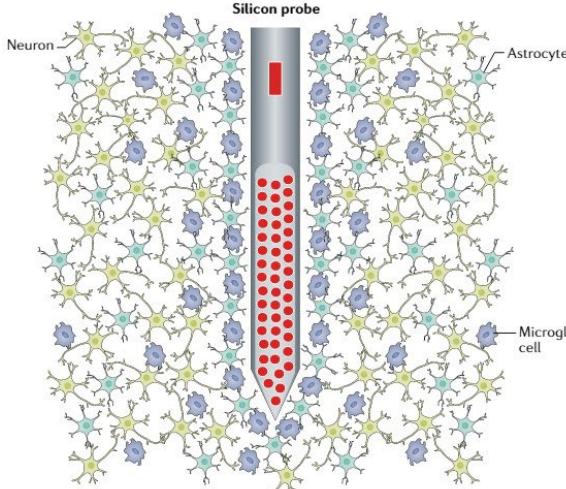
Techniques in Systems Neuroscience – Single cell recording – Tetrodes and Silicon Probes

a High-density electrodes

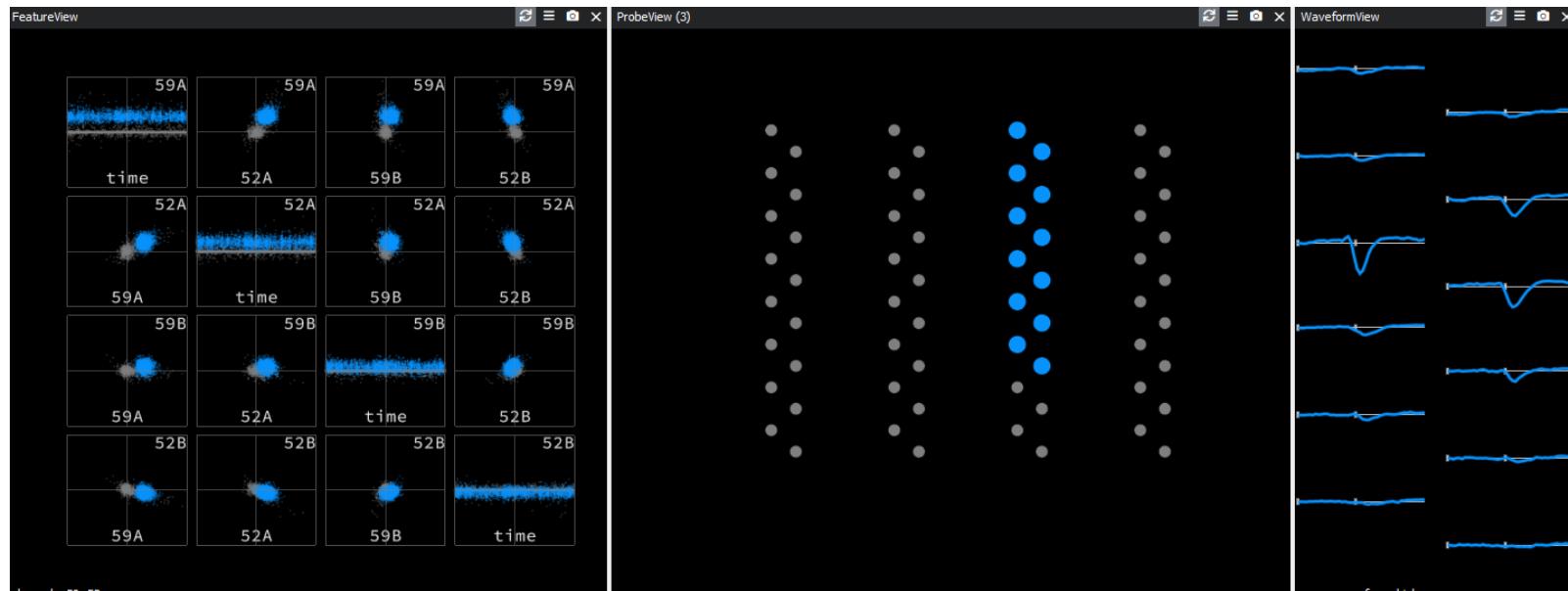
Neuropixels



c Neural probe-neural tissue interface



Techniques in Systems Neuroscience – Single cell recording – Tetrodes and Silicon Probes



Techniques in Systems Neuroscience – Single cell recording – EEG, ECoG, LFP and Laminar Recordings

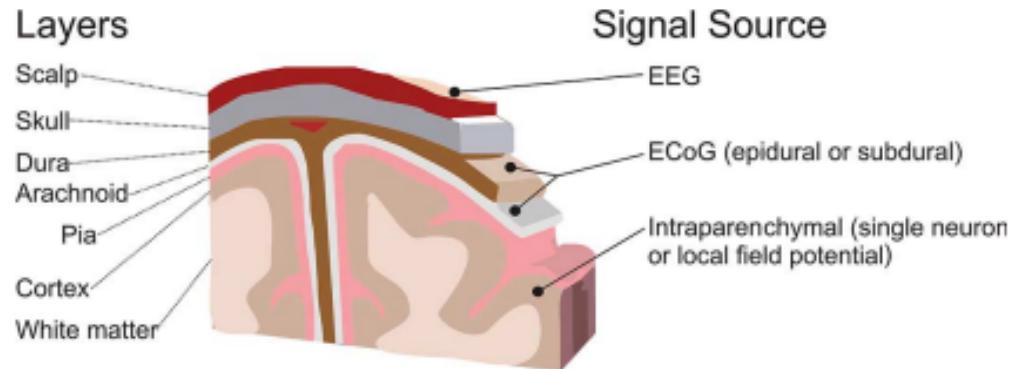
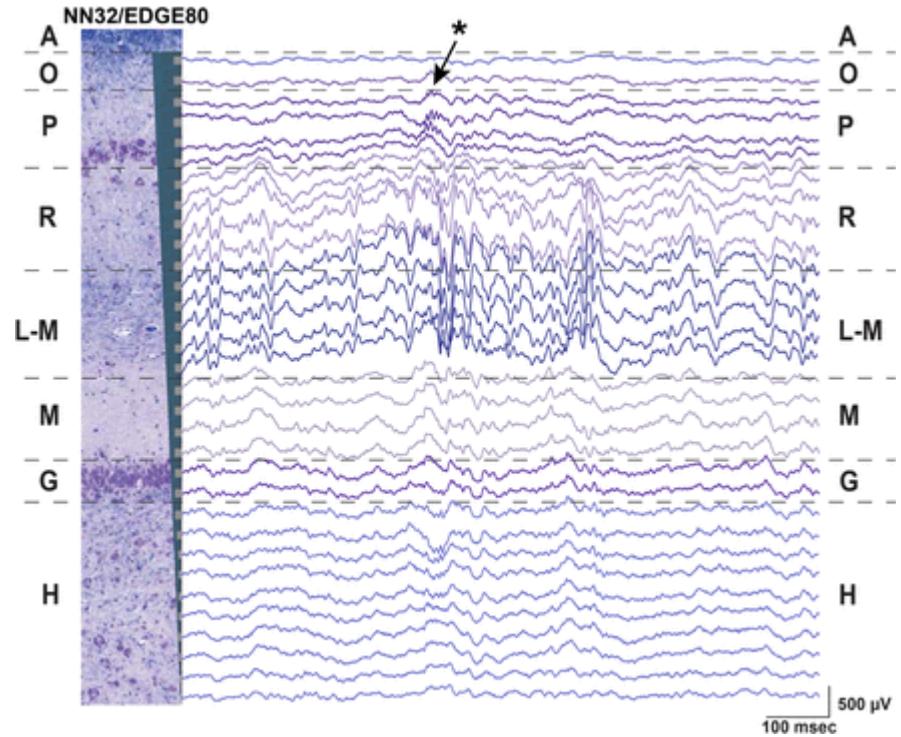
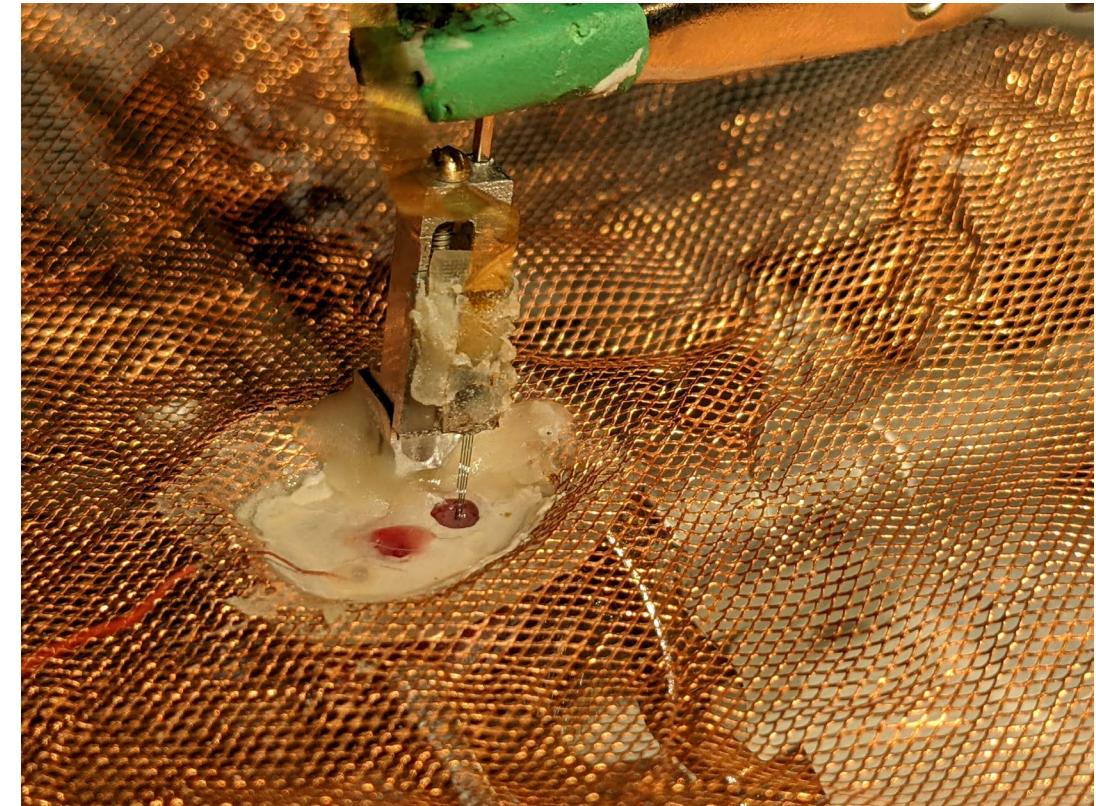
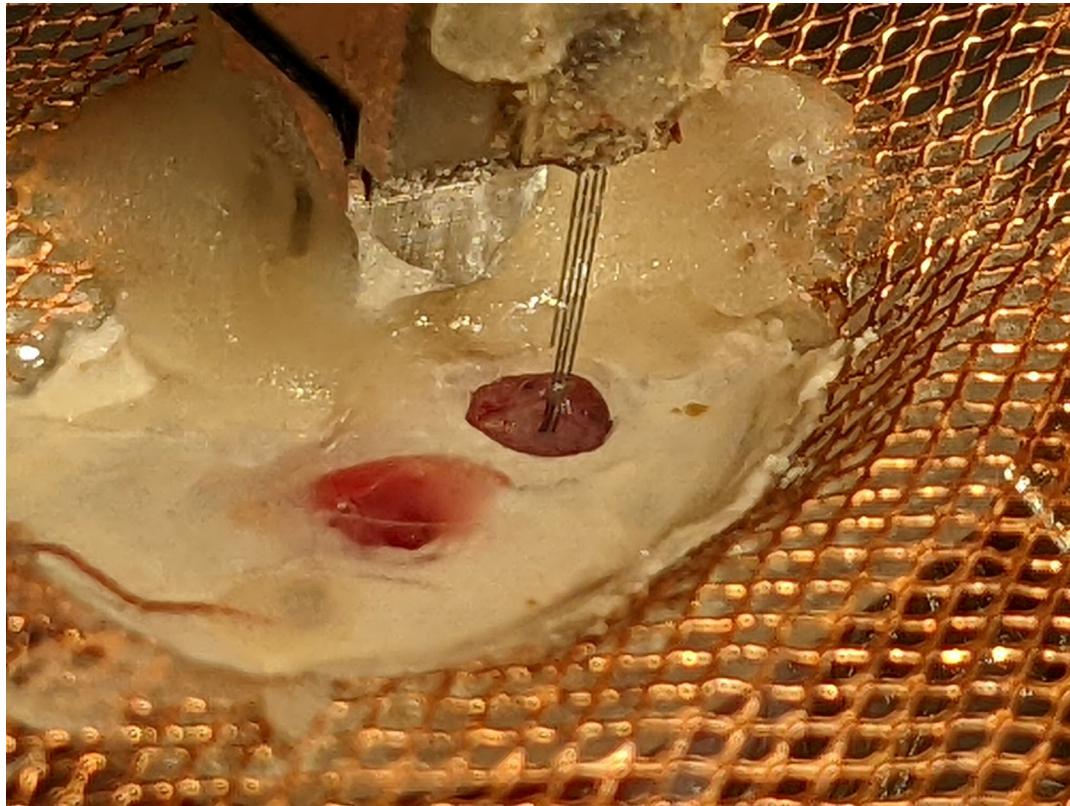


Fig. 1. Recording domains: from single-unit recordings within the brain to EEG recordings on the surface of the scalp. From [1].

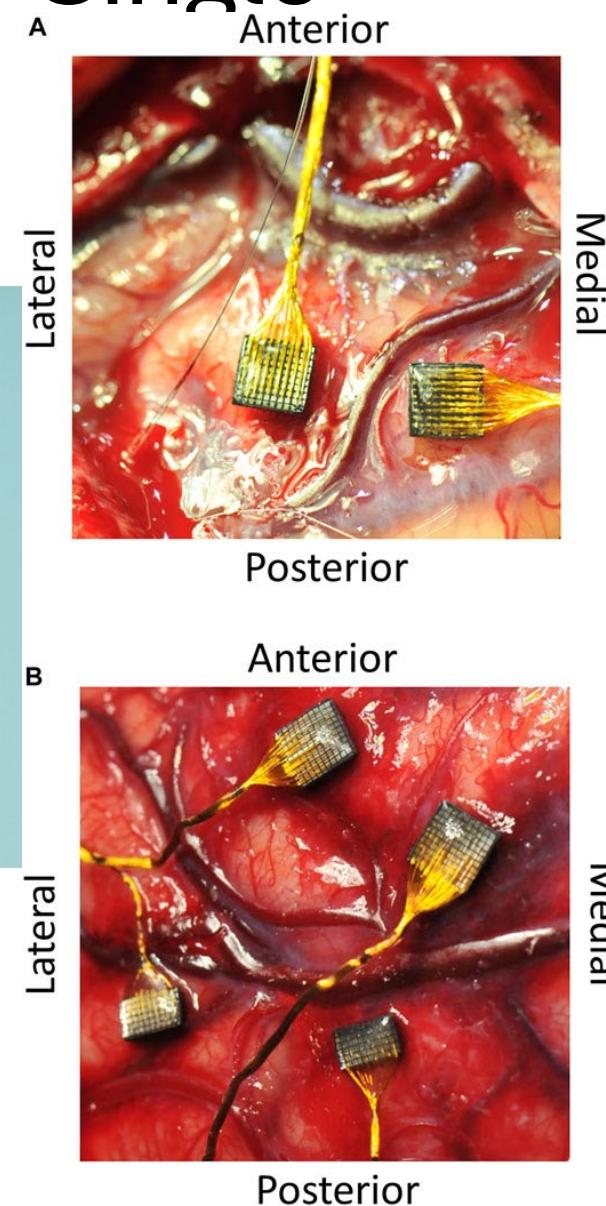
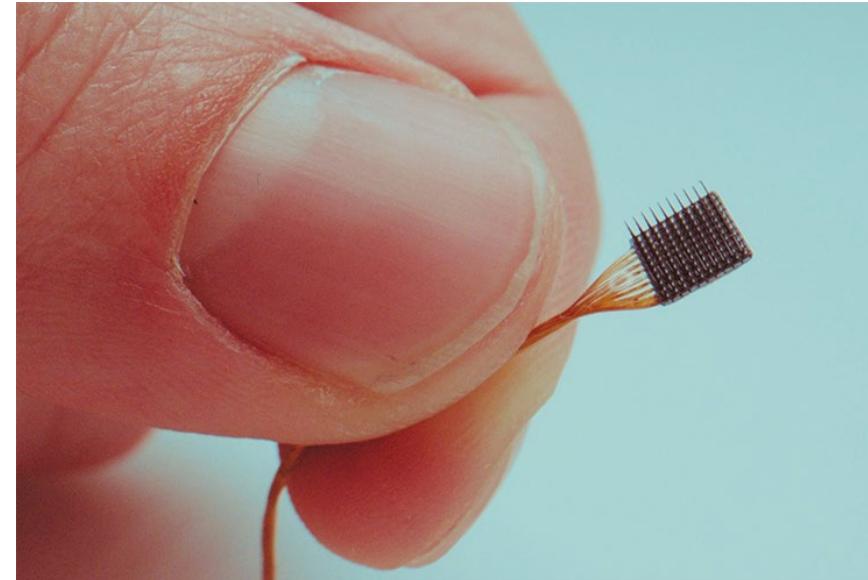
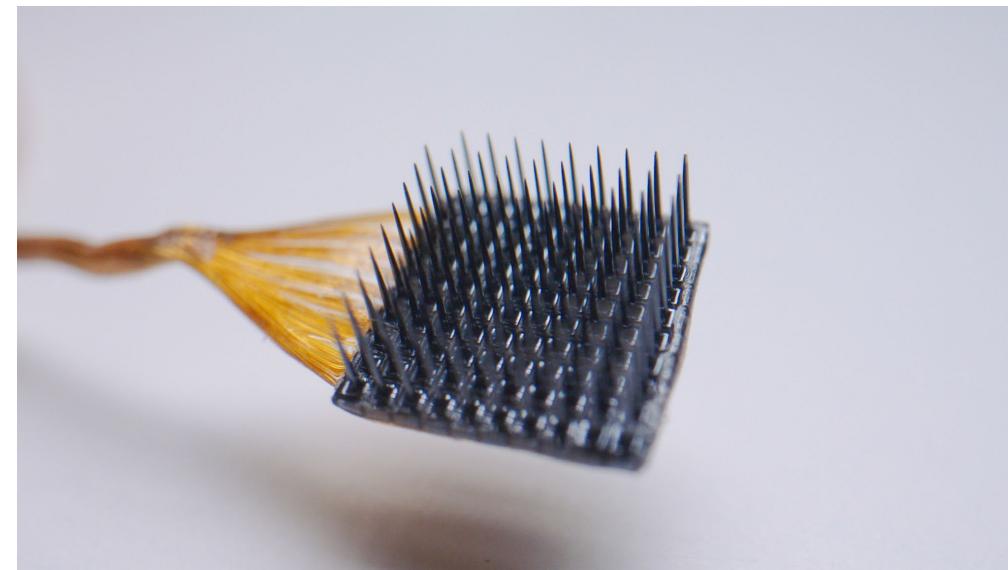


Ulyanova et al. *Frontiers in Neuroscience* 2019

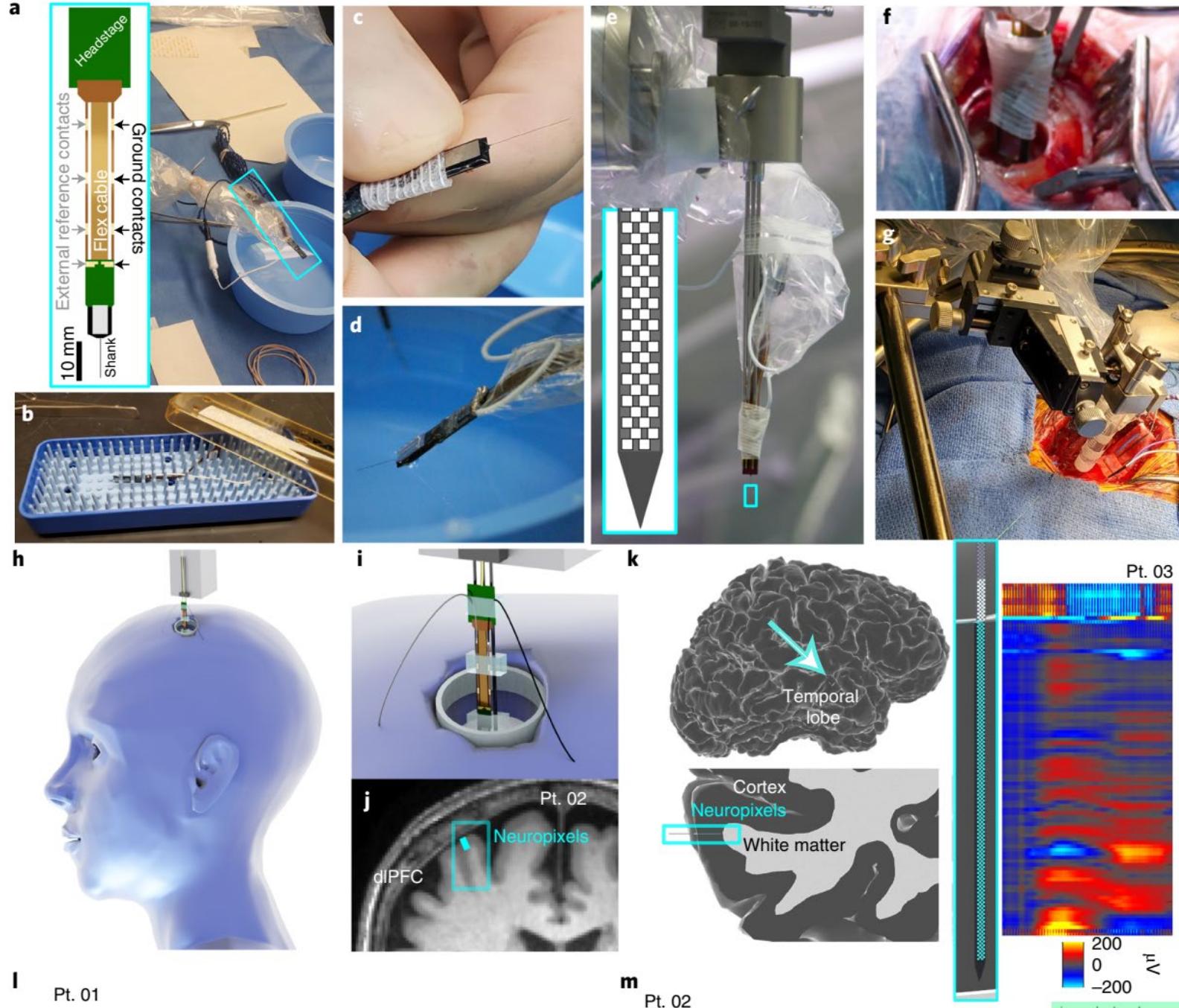
Techniques in Systems Neuroscience – Single cell recording – Silicon Probes



Techniques in Systems Neuroscience – Single cell recording – Utah Array

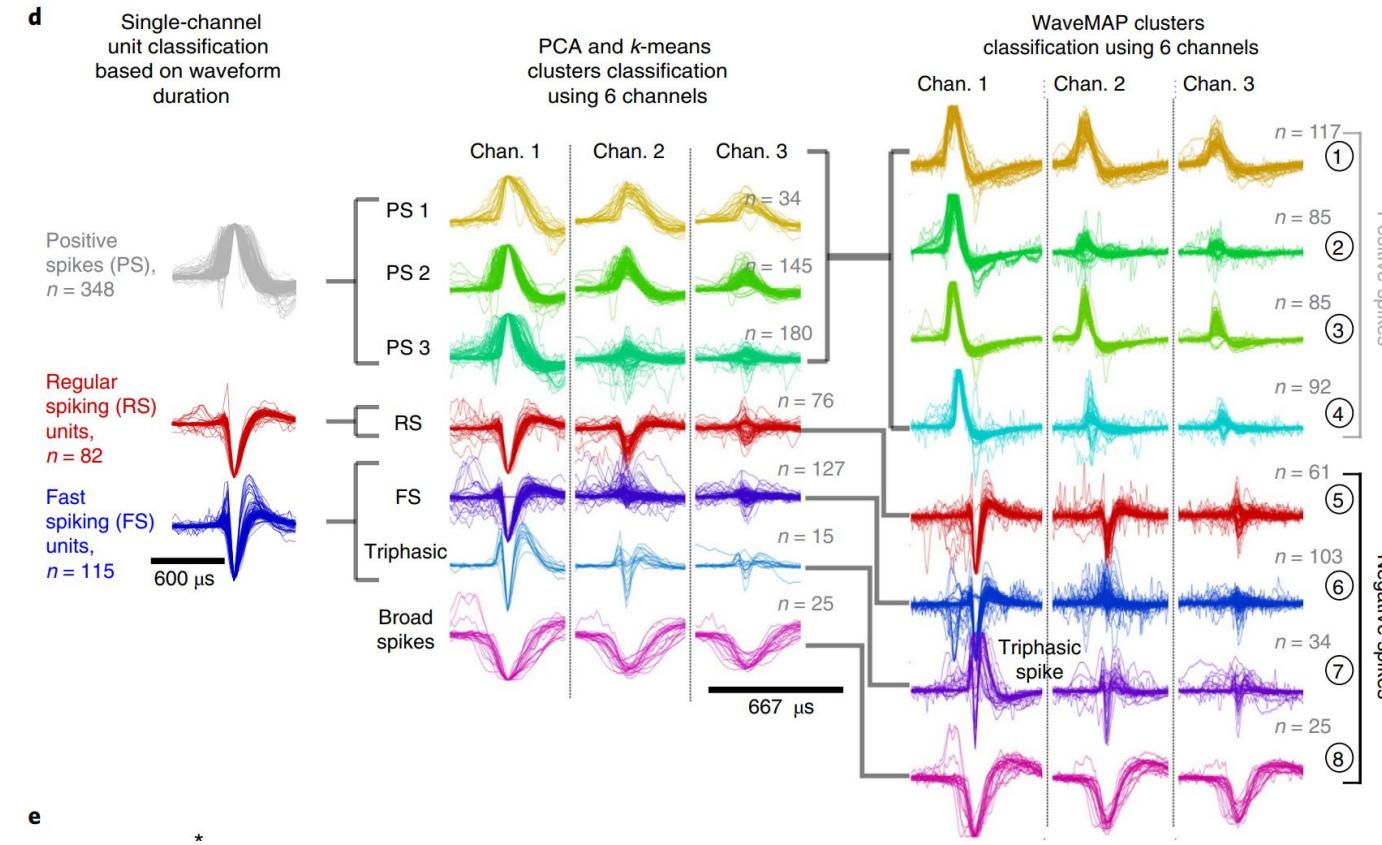
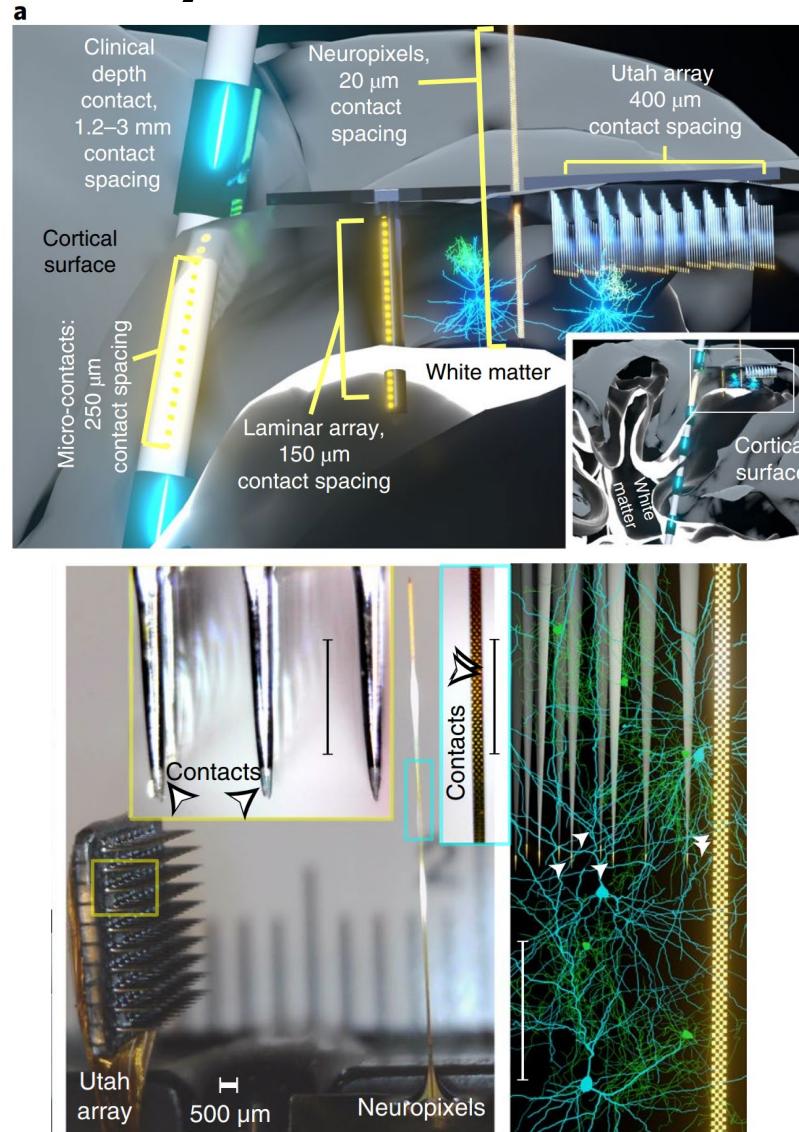


Techniques in Systems Neuroscience – Utah Array versus Silicon Probes



Paul et al. *Nature Neuroscience* 2022c

Techniques in Systems Neuroscience – Utah Array versus Silicon Probes



So, what are we recording? And
how do we make sense of it?

Techniques in Systems Neuroscience – Single cell recording – Hippocampus sleep recording

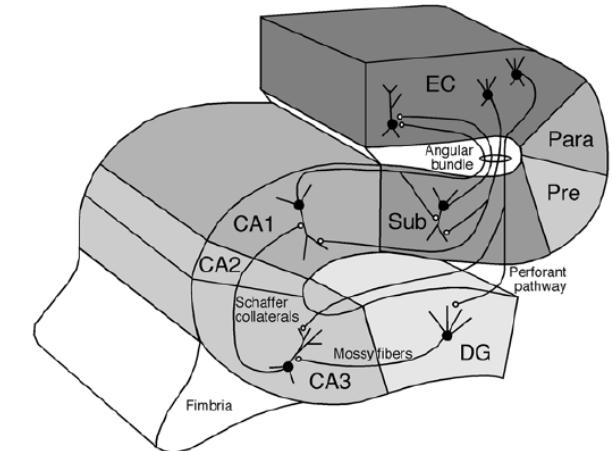
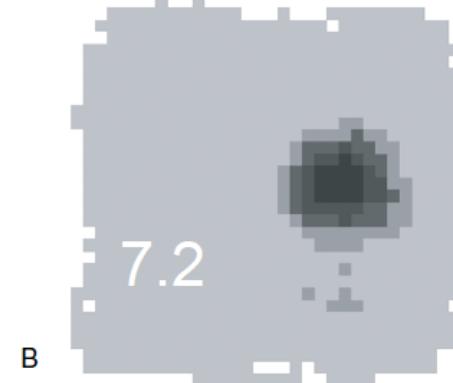
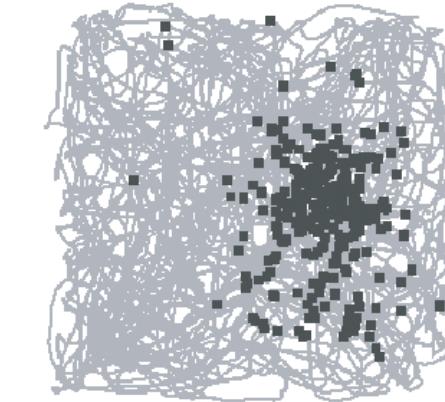
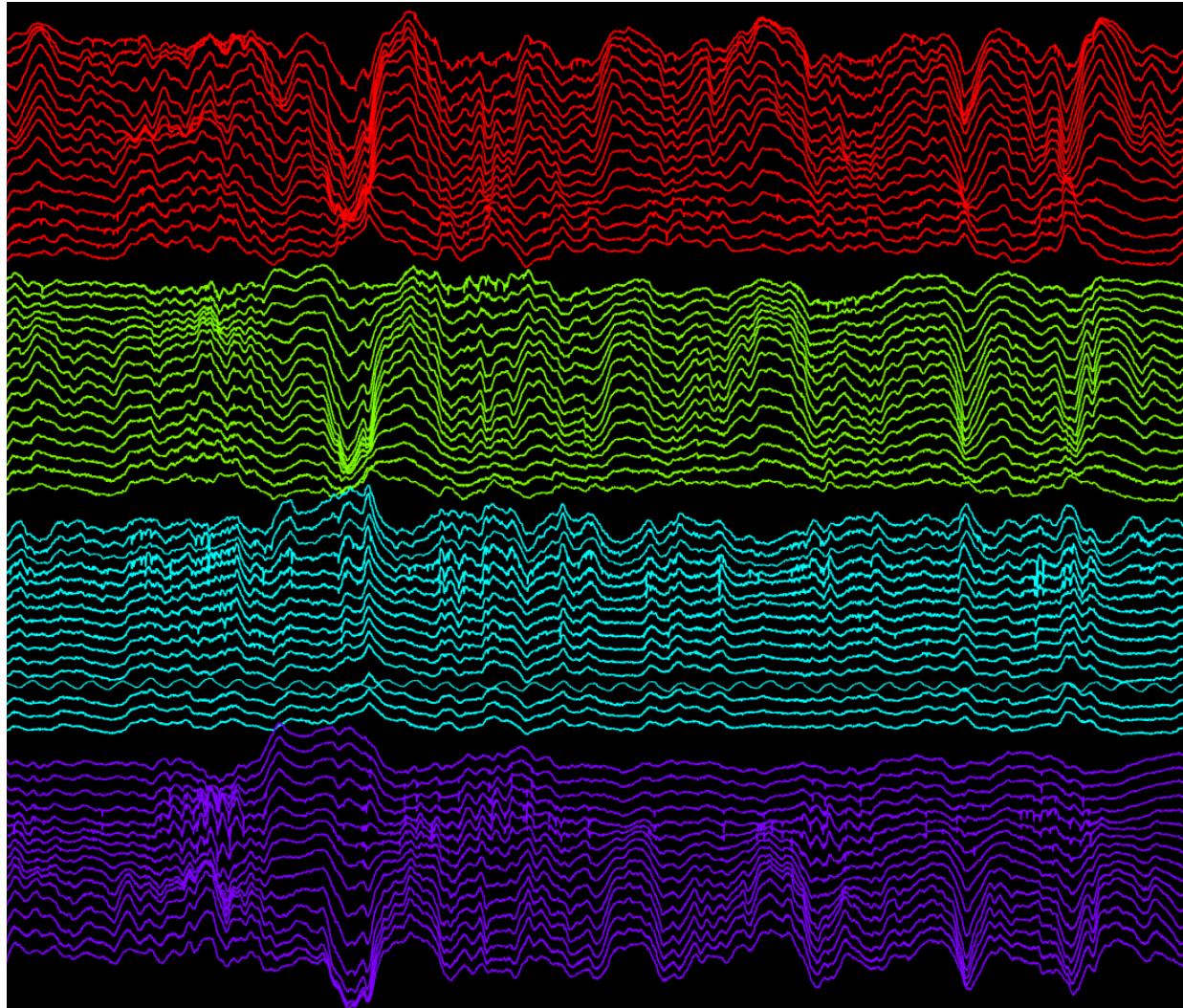
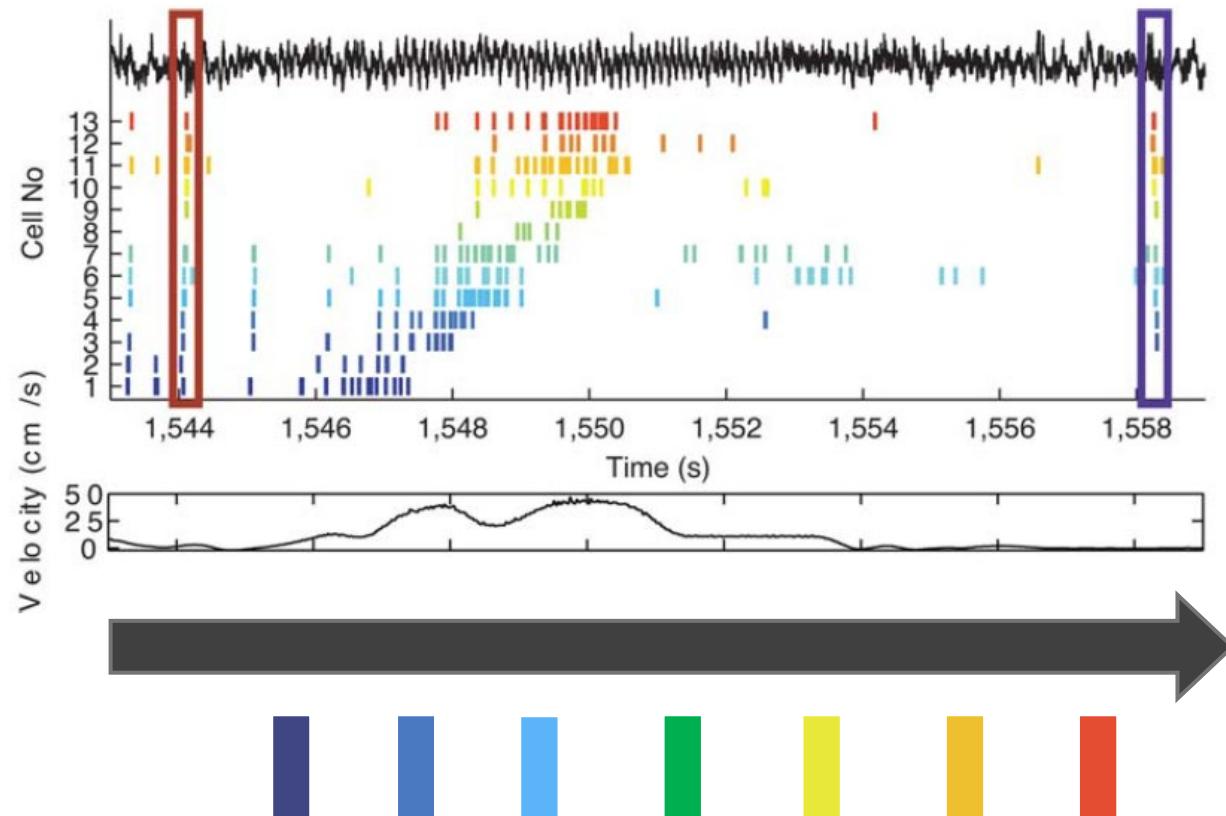


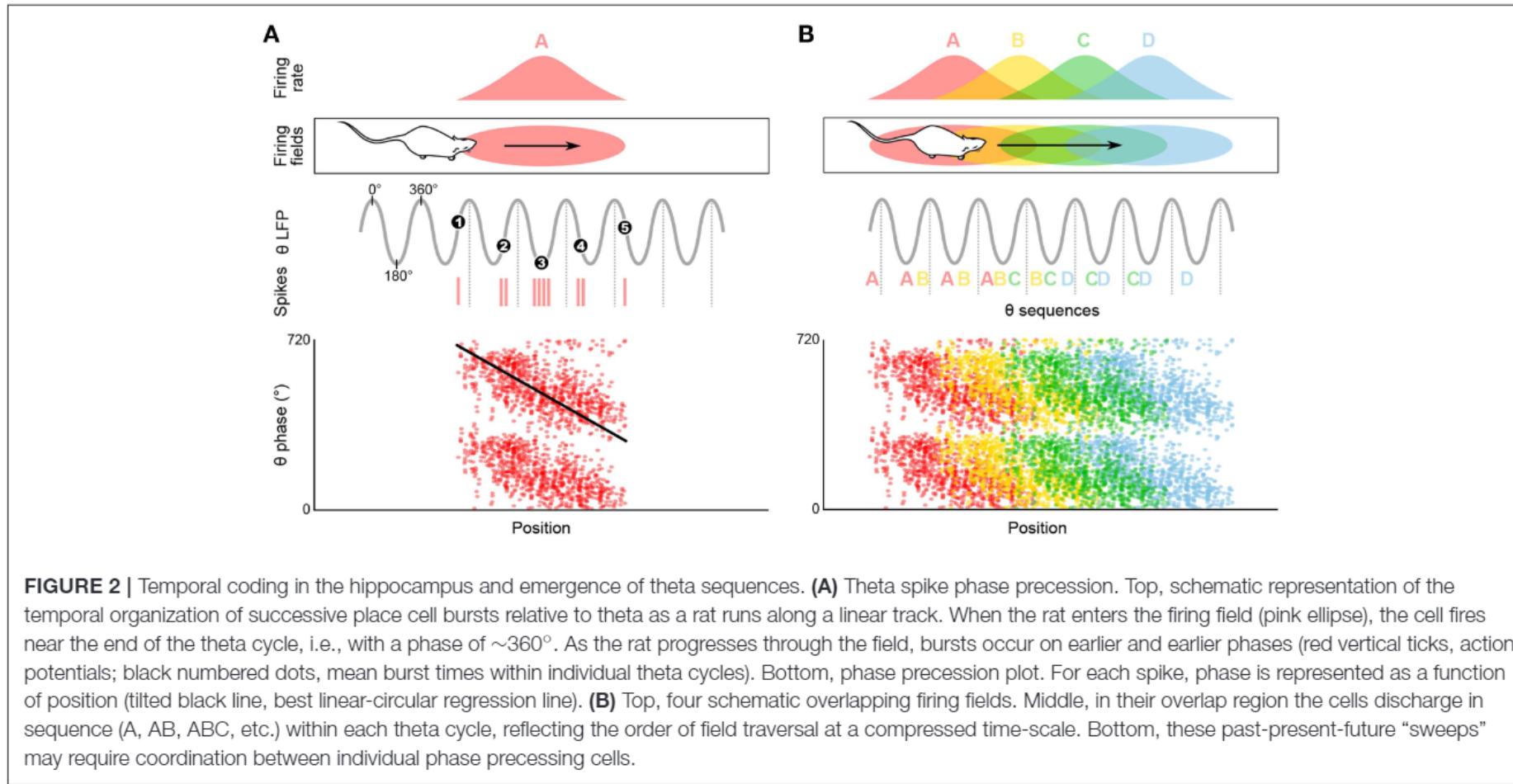
Figure 11-6. Firing field of a CA1 place cell. *A*, Raw data from a rat foraging for food in a square box for 10 minutes. Gray line traces the animal's path through the environment; black boxes show firing of place cell. *B*, Place firing field as a grayscale map where darker colors represent higher firing rates. Inset number in white gives the peak firing rate. (Source: After Wills et al., 2005, with permission.)

Hippocampus Place Cell Progression through a Linear Maze – rate coding



Diba and Buzsáki *Nature Neuroscience* 2007

Hippocampus Place Cell Progression through a Linear Maze – temporal coding (ie Phase Precession)



Hippocampus Place Cell Progression through a Linear Maze – temporal coding (ie Phase Procession)

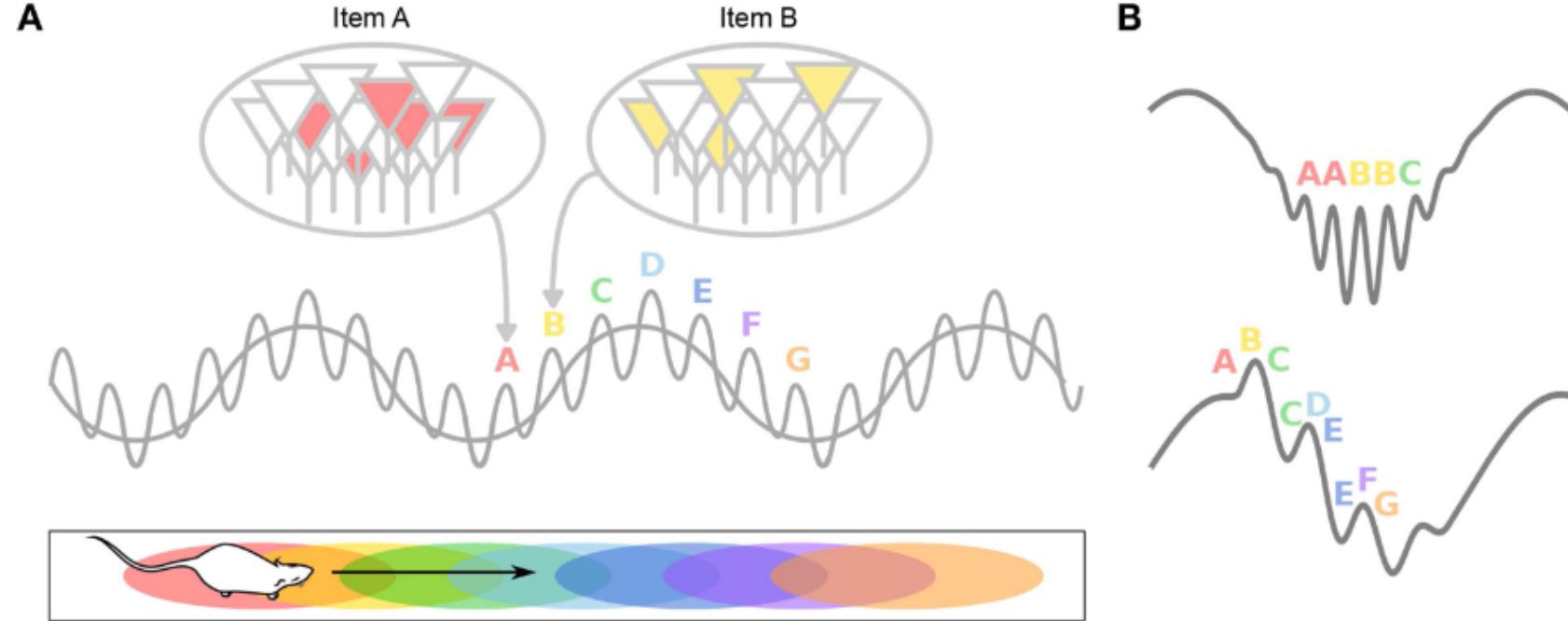
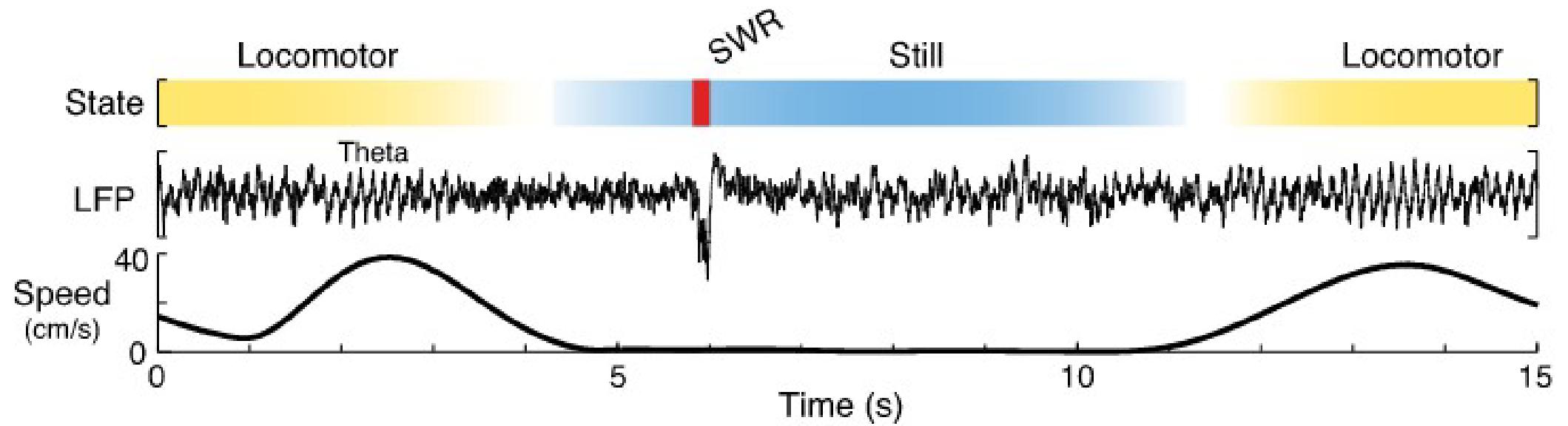
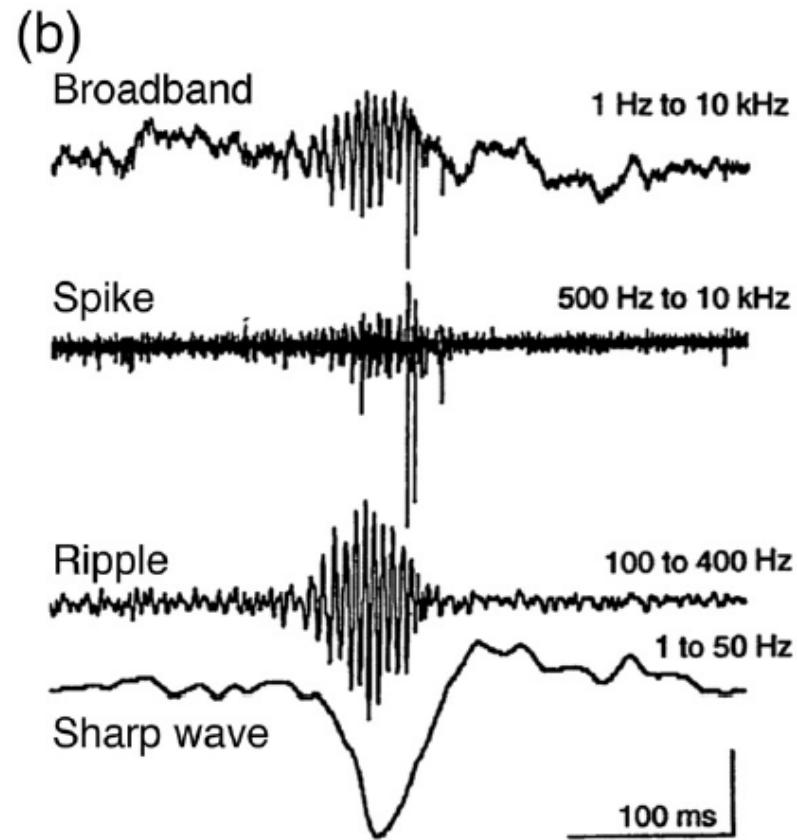
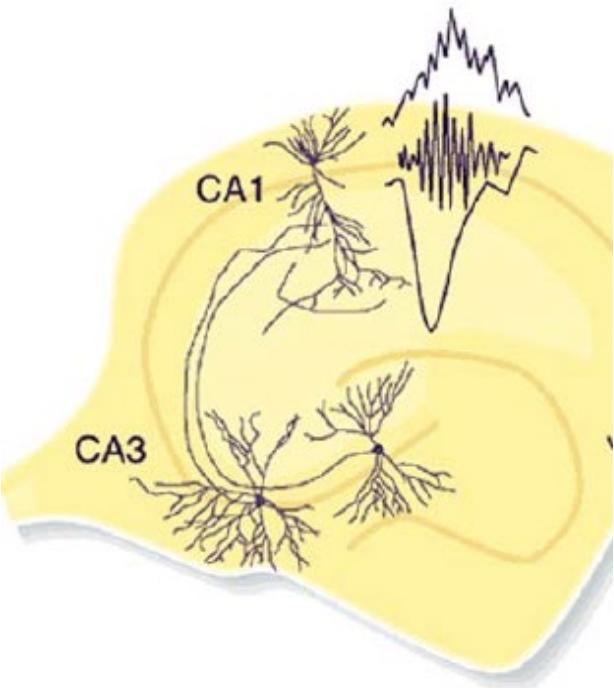


FIGURE 4 | Theta-gamma neural code. **(A)** Individual memory representations (cell assemblies, identified by letters) are repeatedly activated within each new theta cycle. The gamma rhythm separates the respective representations in time. The number of gamma cycles per theta cycle determines the span of working memory. **(B)** If discrete locations (i.e., A–G) are encoded within individual gamma cycles, longer paths would be encoded during fast gamma than during slow gamma. Alternatively, slow (but not fast) gamma cycles encode sequences of locations, rather than single locations. This results in longer paths encoded during slow gamma than during fast gamma.

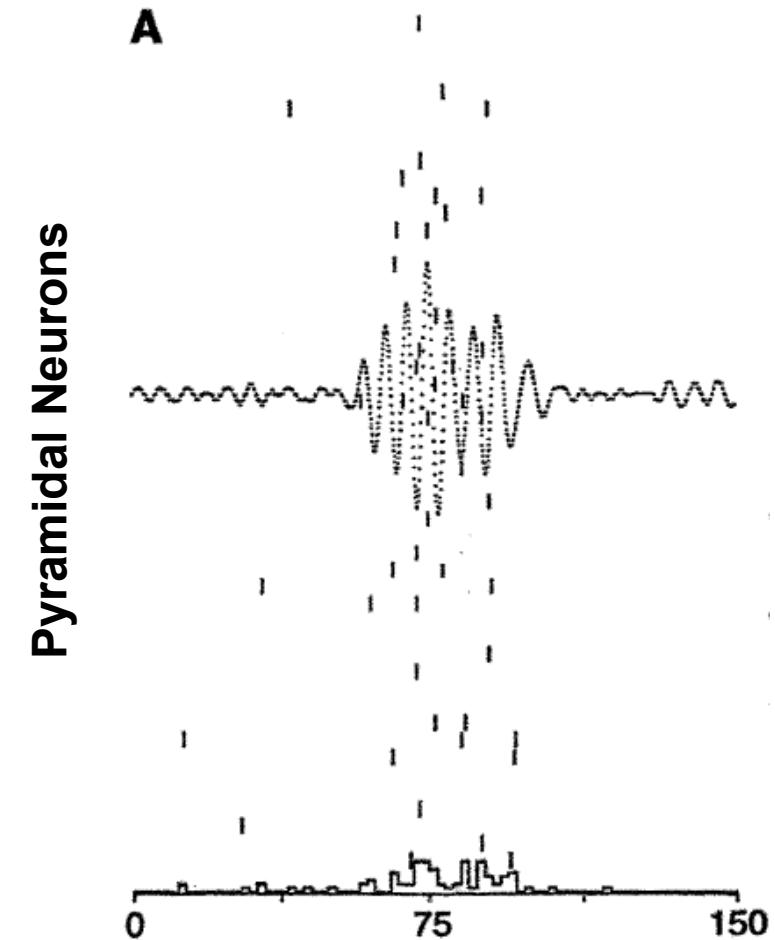
Hippocampus States – Theta and SWRs



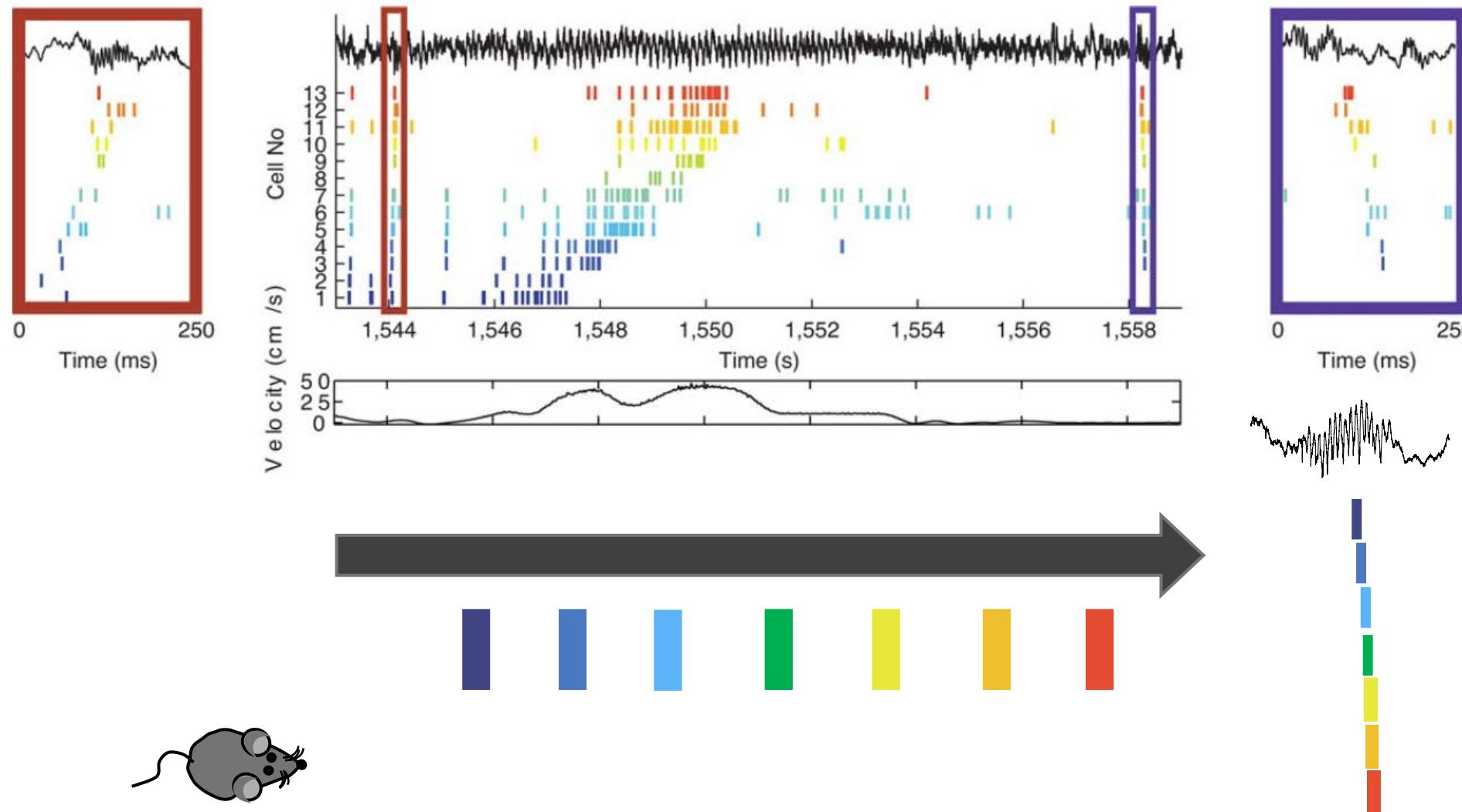
Hippocampus Place Cell Progression through a Linear Maze – Sharp Wave Ripples reactivation state



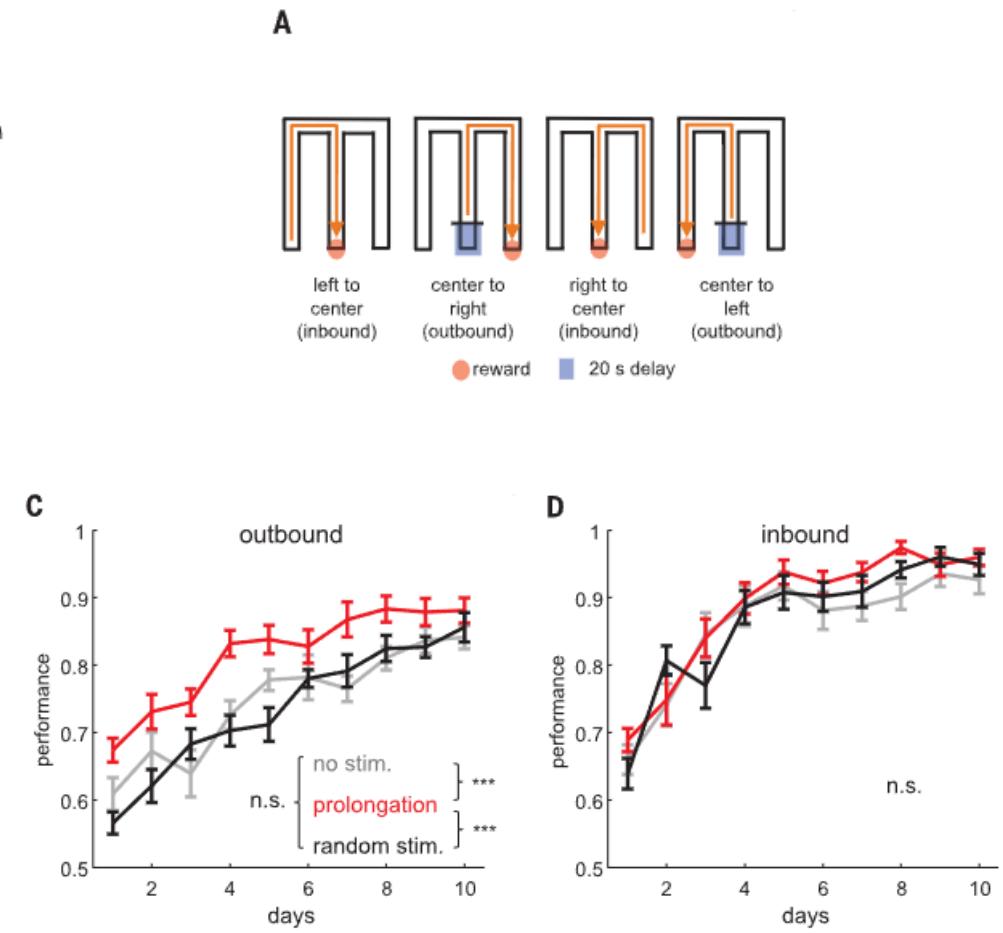
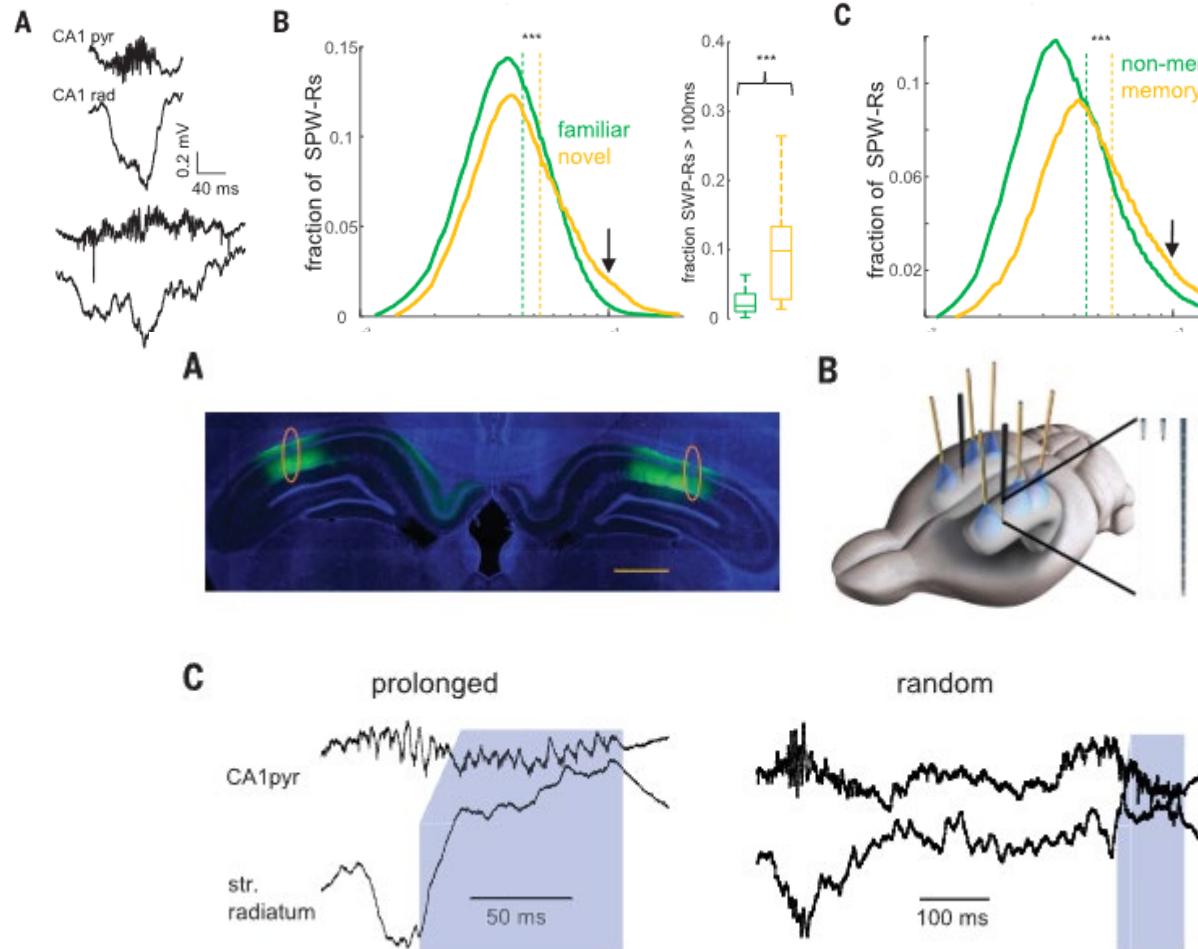
Buzsáki and Chobrak *Nat. Neuroscience* 2005
Buzsáki *Progress in Neurobiology* 1984
Buzsáki *Brain Research* 1986
Buzsáki...Wise *Science* 1992



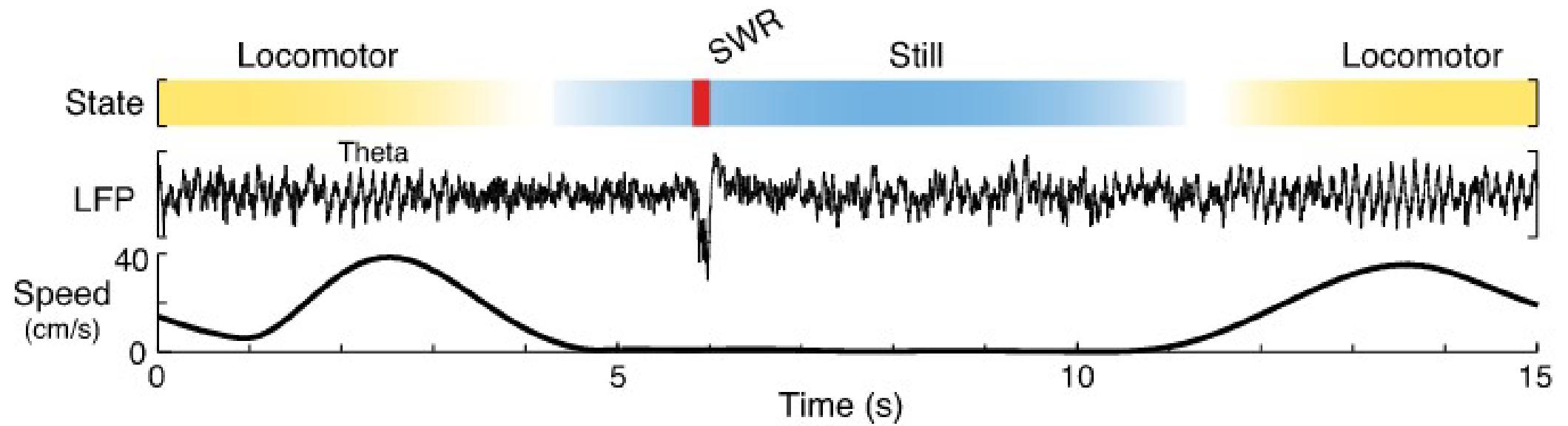
Hippocampus Place Cell Progression through a Linear Maze – Sharp Wave Ripples reactivate place cells



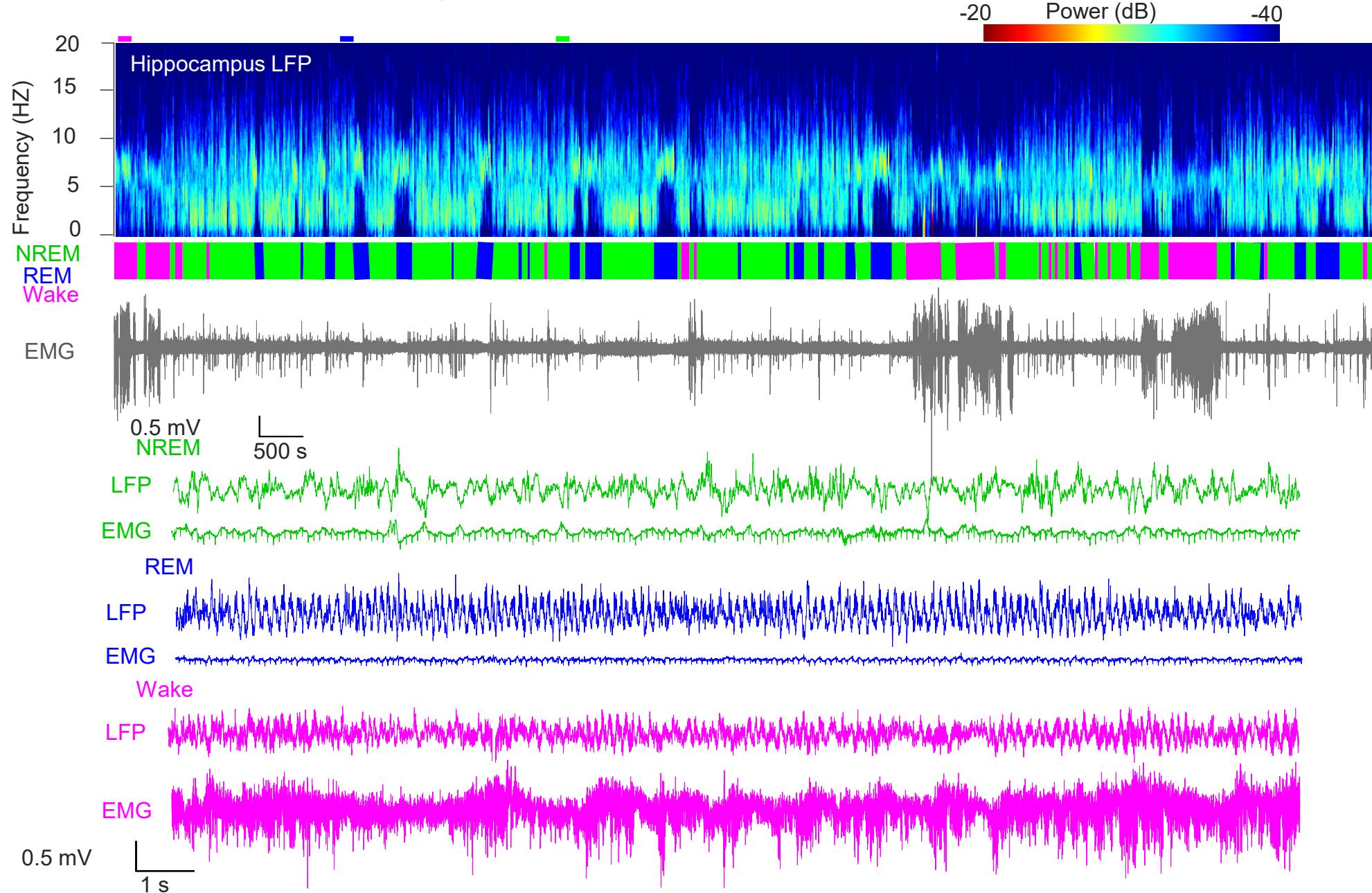
Increasing the Length of Awake SWRs Artificially Increases Spatial Memory



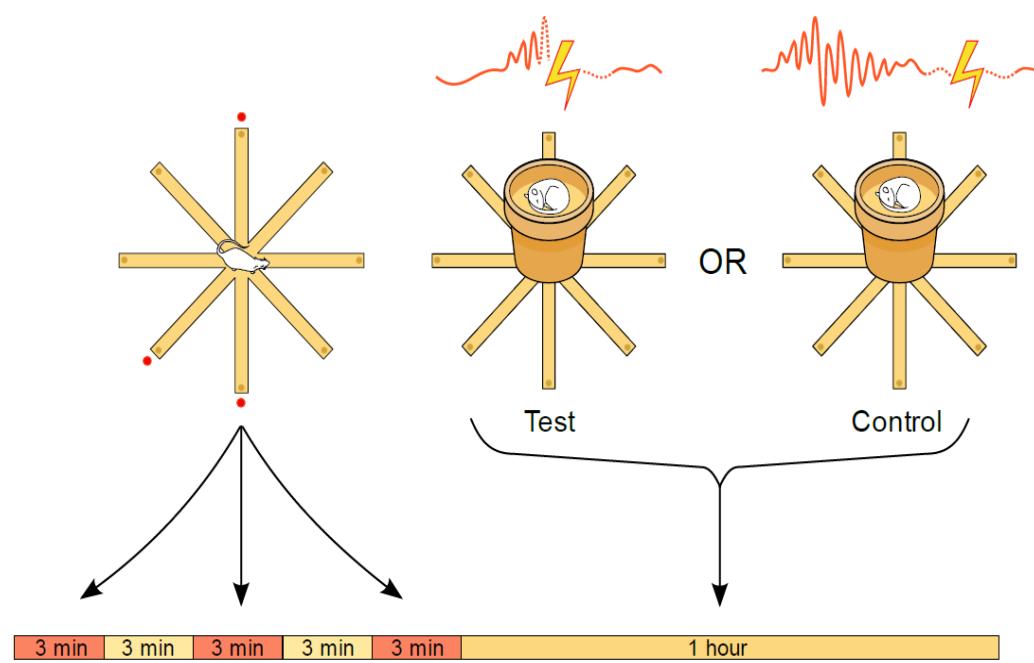
Hippocampus States – Theta and SWRs



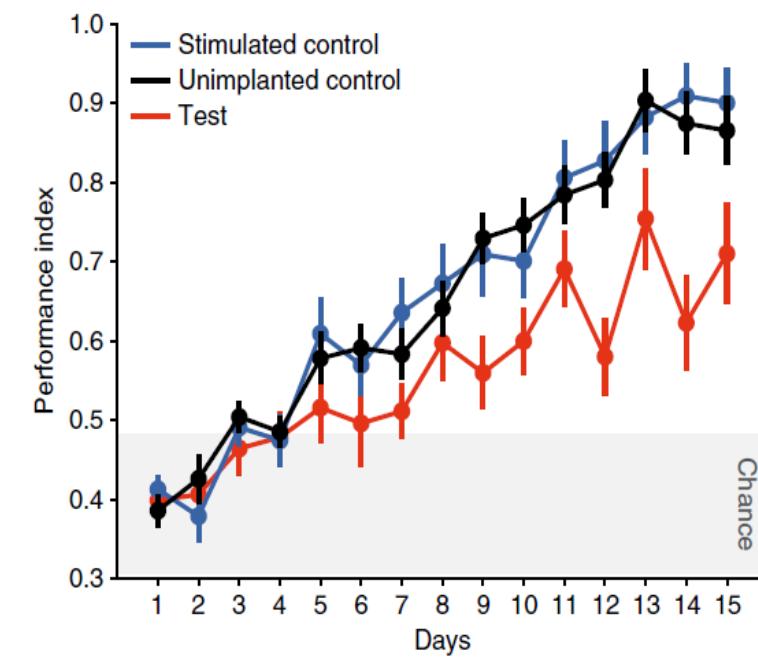
Sleep States – Wake, REM, NREM



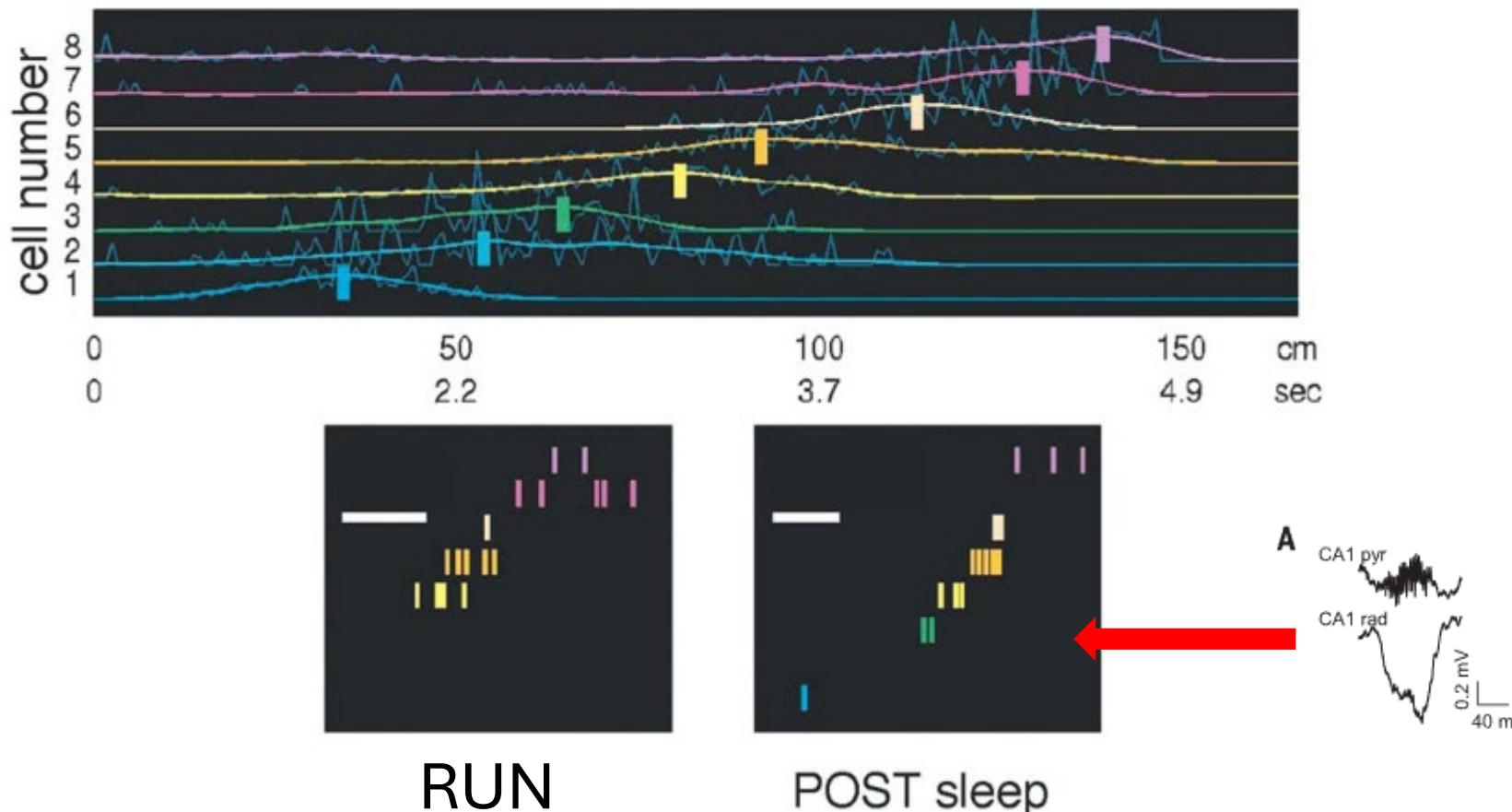
Impairing Sleep SWRs Impairs Spatial Memory



Girardeau...Zugaro *Nat. Neuroscience* 2009



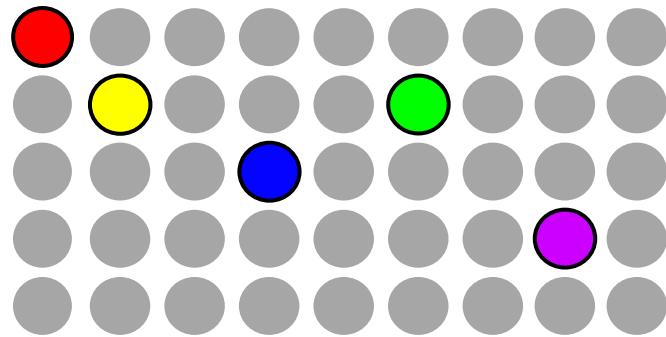
Sleep SWRs Reactivate Trajectories from Previous Experience



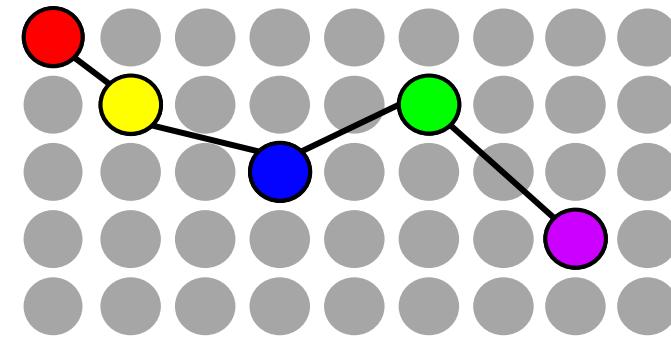
Two stage Model of Memory in Hippocampus

- Encode, then Consolidate

1. Encoding



2. Consolidation



How does the brain-firing activity present itself?

How Does the Brain Present Neural Patterns – Groups of Assemblies and Hebbian Plasticity

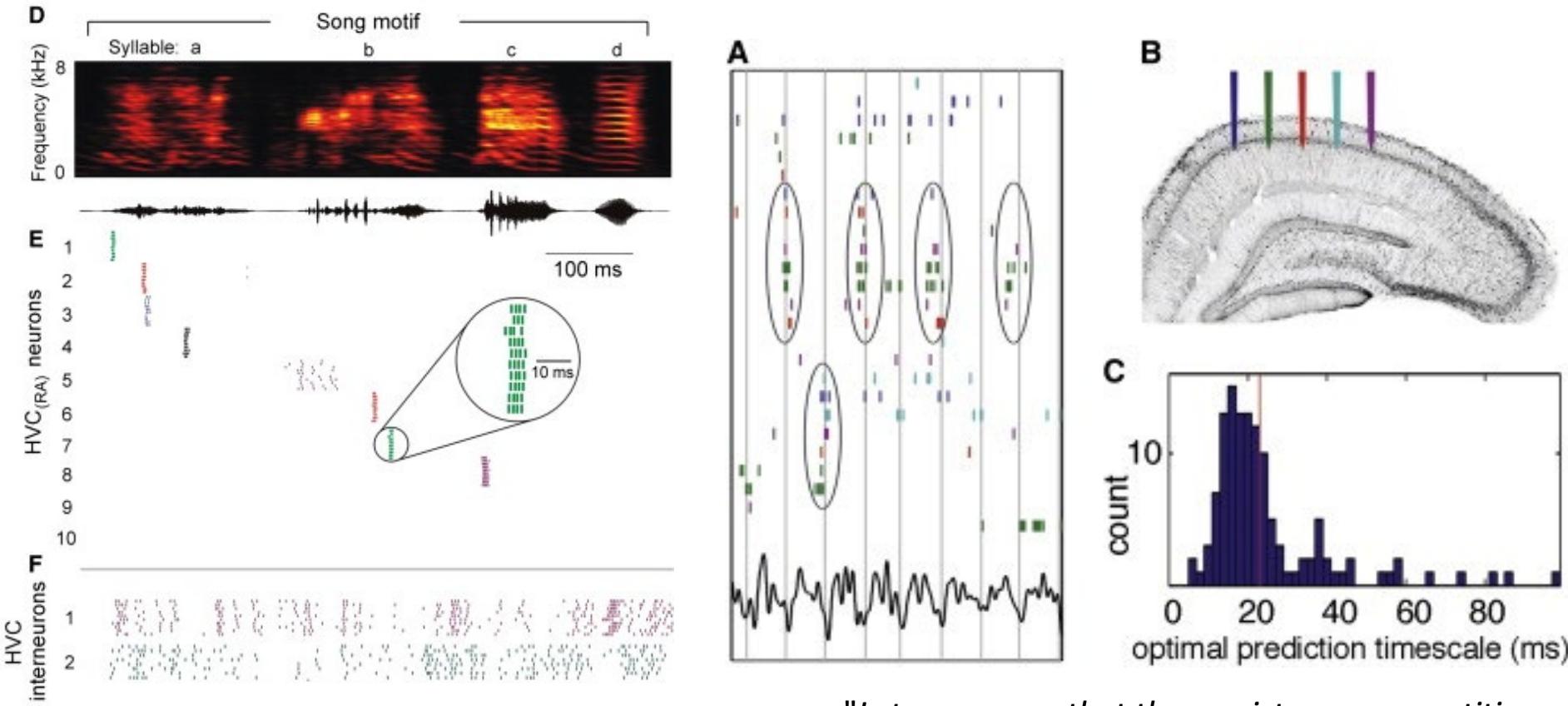


"Let us assume that the persistence or repetition of a reverberatory activity (or "trace") tends to induce lasting cellular changes that add to its stability.... When an axon of cell A is near enough to excite a cell B and repeatedly or persistently takes part in firing it, some growth process or metabolic change takes place in one or both cells such that A's efficiency, as one of the cells firing B, is increased."

-Donald Hebb *The Organization of Behavior: A Neuropsychological Theory*. 1949

“Neurons that fire together wire together”

How Does the Brain Present Neural Patterns – Groups of Assemblies and Hebbian Plasticity



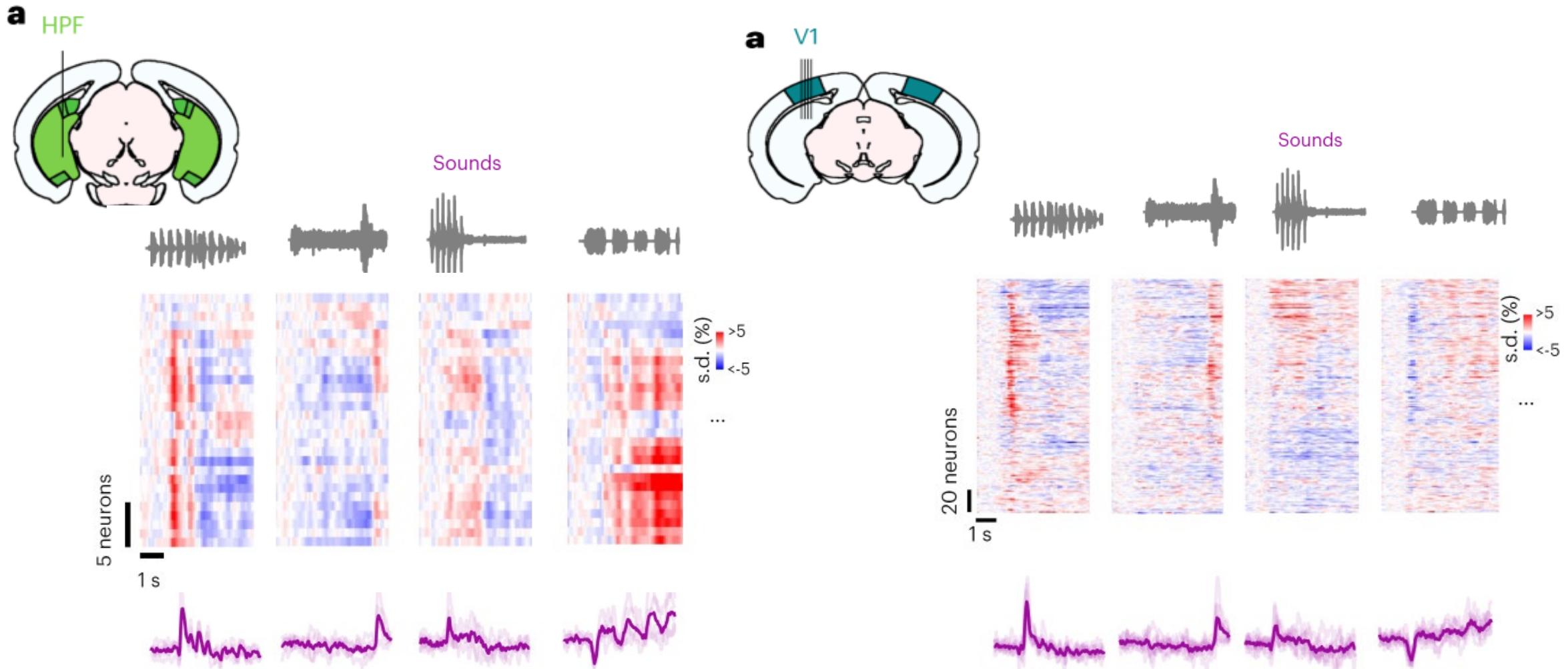
"Let us assume that the persistence or repetition of a reverberatory activity (or "trace") tends to induce lasting cellular changes that add to its stability.... When an axon of cell A is near enough to excite a cell B and repeatedly or persistently takes part in firing it, some growth process or metabolic change takes place in one or both cells such that A's efficiency, as one of the cells firing B, is increased."

Buzsáki *Neuron* 2010

Hahnloser, Kozhevnikov, Fee *Nature* 2002

Harris... Buzsáki *Nature* 2003

How Does the Brain Present Neural Patterns – Groups of Assemblies and Hebbian Plasticity



The current state of Neuroscience – Understanding the Neurophysiology Underlying Behavior



From Netflix “Birds of Paradise”

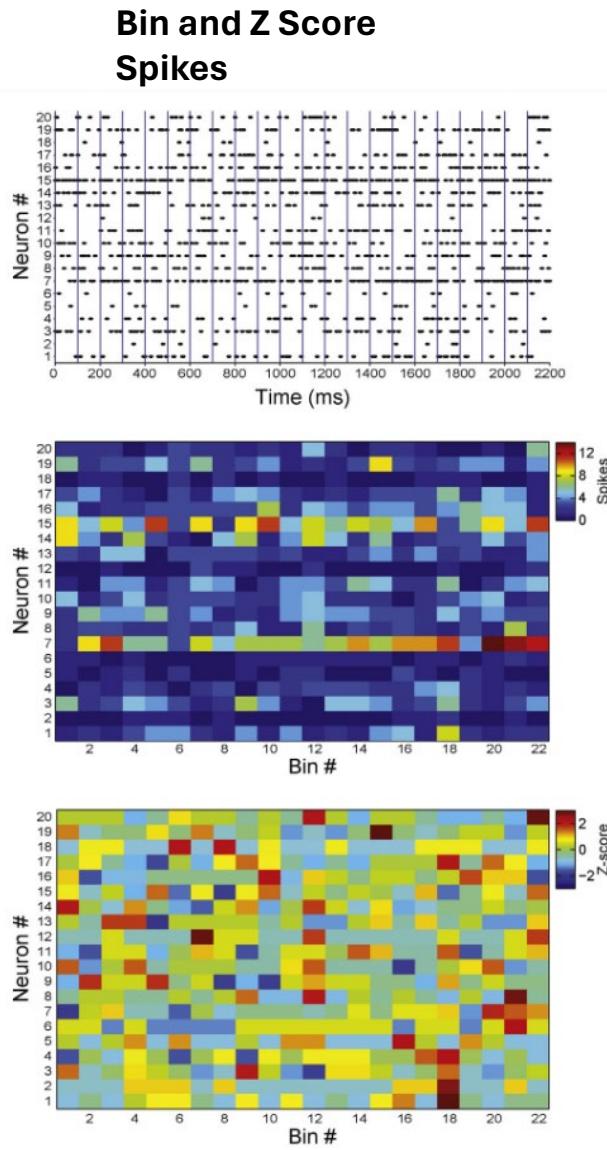
"The problem of understanding behavior is the problem of understanding the total action of the nervous system, and vice versa.

This has not always been a welcome proposition, either to psychologist or to physiologist"

-D.O. Hebb

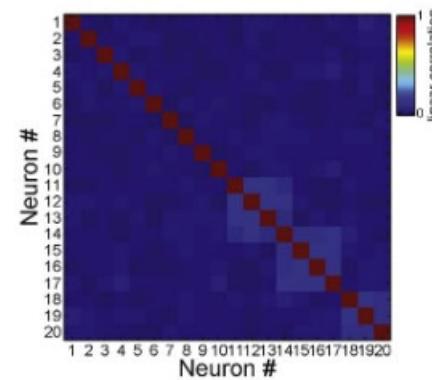
Thanks!

Bonus - Neuron Cell Assembly Detection

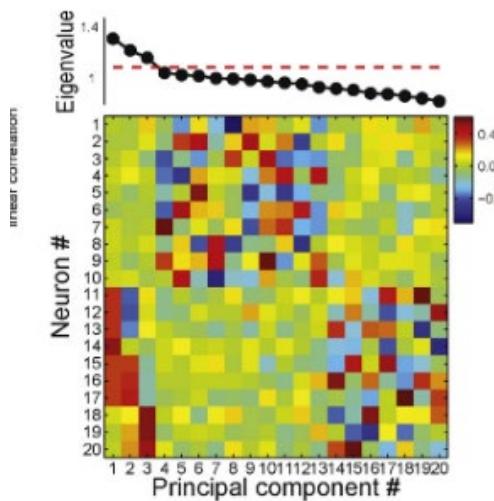


Correlation Matrix

A

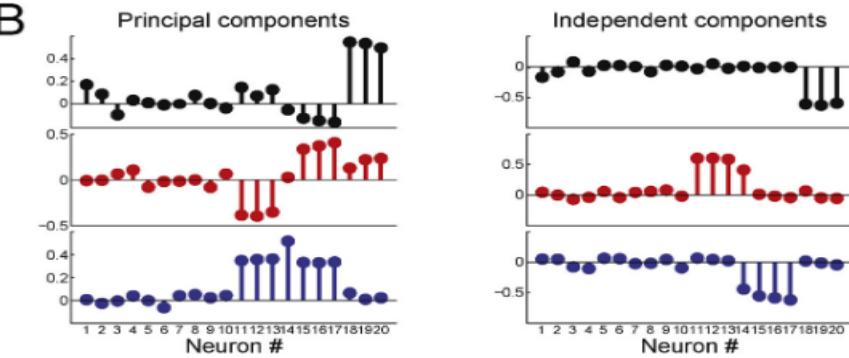


PCA and Determination of components (Marchenko -Pastur law)



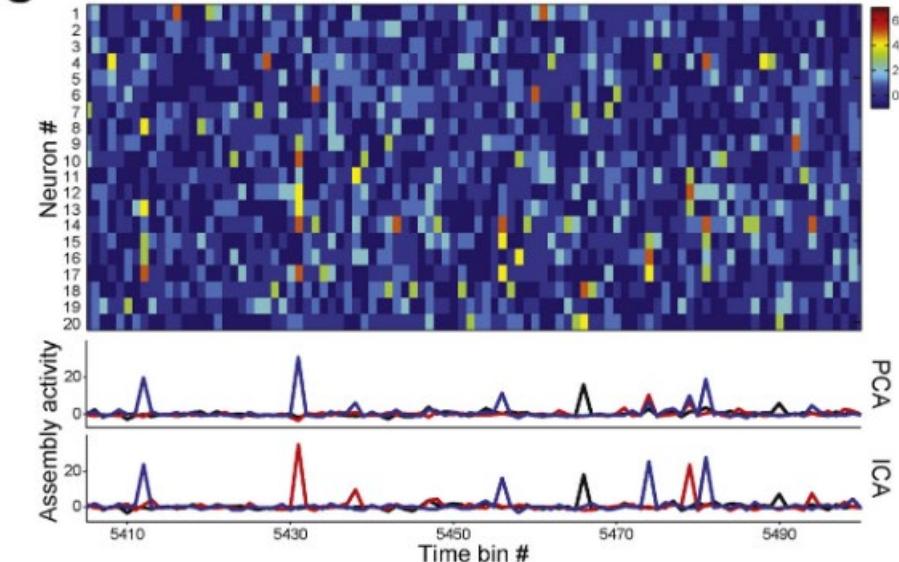
ICA to determine the contribution (i.e. weight) of a neuron to each assembly (component)

B

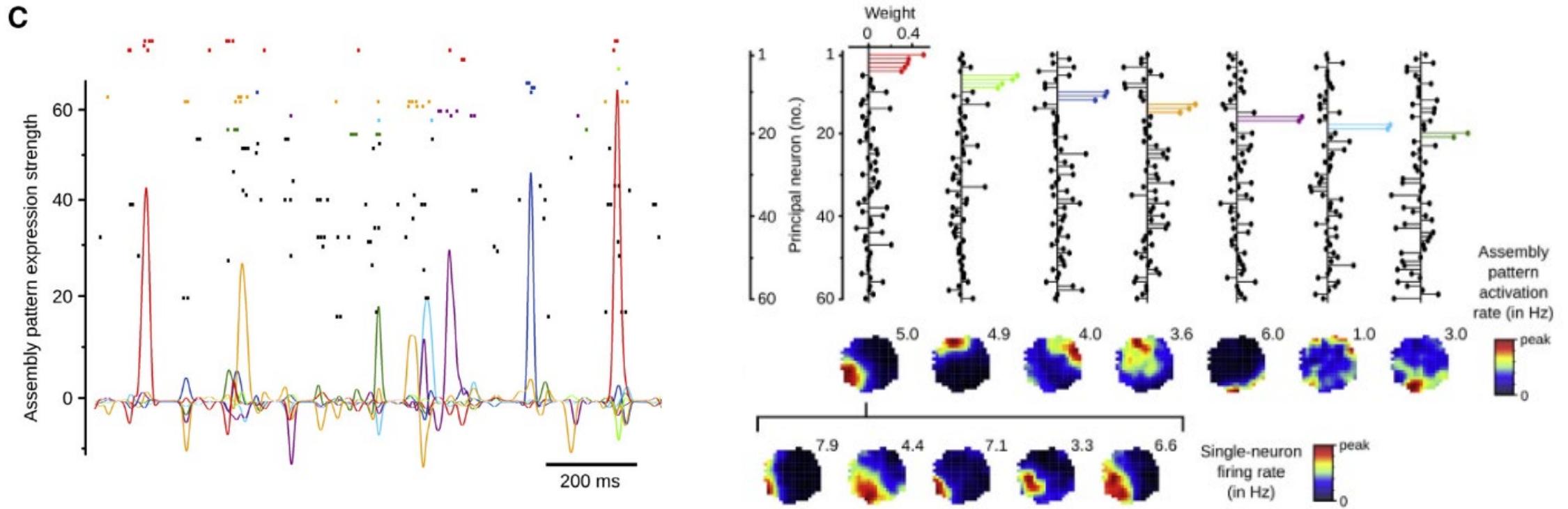


Project the Weight Matrix back onto the z-score binned data = Assembly Activity of each assembly over time

C



Bonus - Neuron Cell Assembly Detection



Van de Ven...Dupret *Neuron* 2016