

The Hippocampus as a Cognitive Map

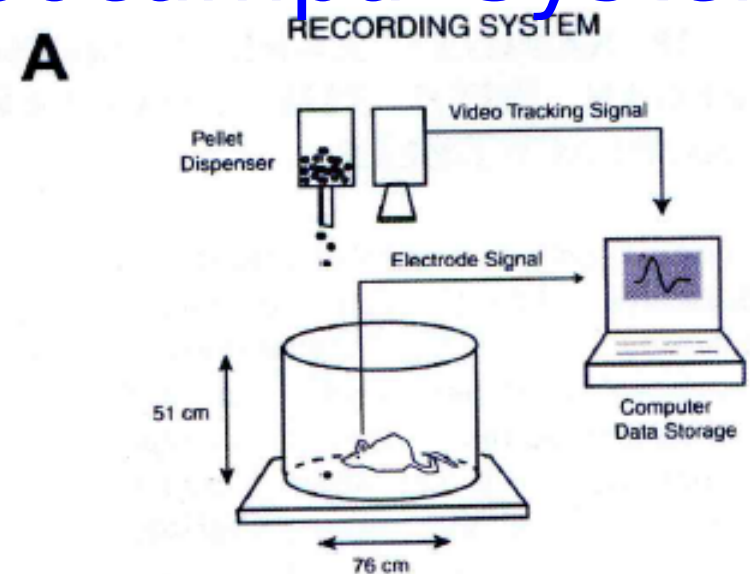
Computational Models of Neural Systems
Lecture 3.5

David S. Touretzky
October, 2025

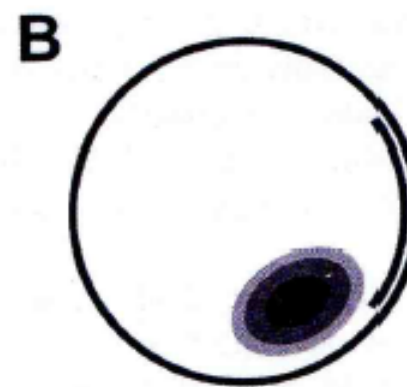
Place Cells Are Found Throughout the Hippocampal System

- Place cells discovered in CA1 in rats by O'Keefe and Dostrovsky (1971)
- Continuous firing fields with gaussian falloff.
- Place fields cover the physical space, forming a “cognitive map” of the environment.

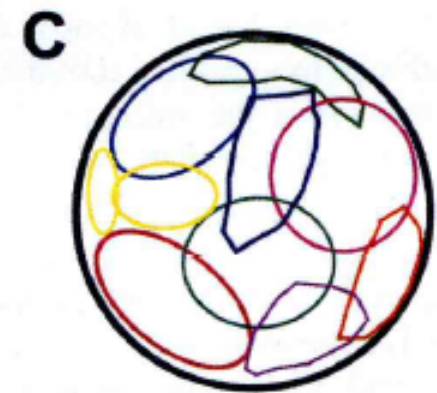
John O'Keefe
2014 Nobel
Laureate in
Physiology or
Medicine



Sharp (2002)



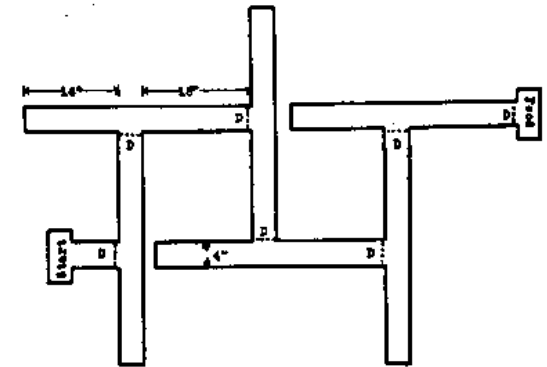
Individual Place Cell



Population of Place Cells

The Hippocampus as a Cognitive Map

- Psychologist E. C. Tolman coined the term “cognitive map” to describe an animal's mental representation of space.
 - Tolman, EC (1948) Cognitive maps in rats and men. *Psych. Review* 55(4):189-208.



6-Unit Alley T-Maze

FIG. 4

(From H. C. Blodgett, The effect of the introduction of reward upon the maze performance of rats. *Univ. Calif. Publ. Psychol.*, 1929, 4, No. 8, p. 117.)

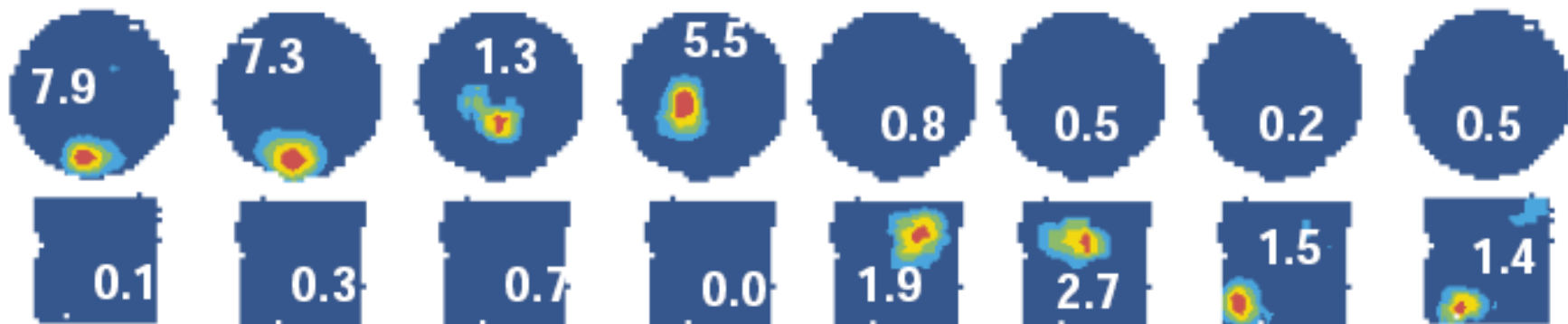
- O'Keefe and Nadel's book about place cells drew its title from Tolman's phrase.
 - O'Keefe, J and Nadel, L. (1978) *The Hippocampus as a Cognitive Map*. Oxford University Press.
 - Now online at <https://discovery.ucl.ac.uk/id/eprint/10103569/>

Properties of Place Fields

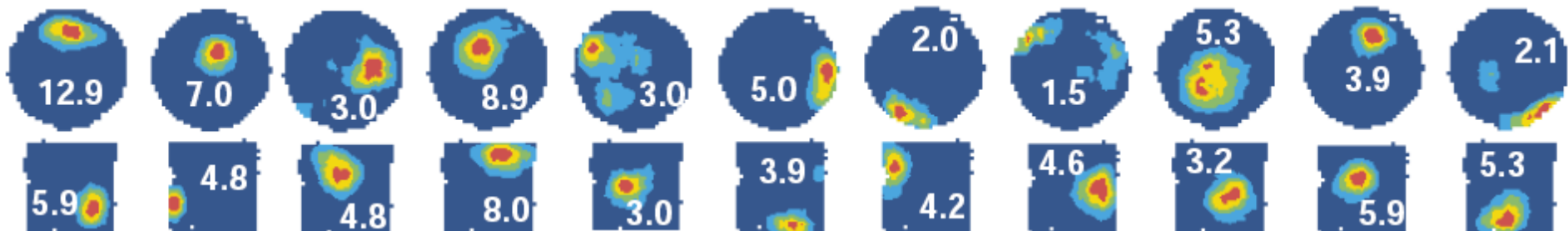
- Appear instantly in a new environment, but take 10-20 minutes to fully develop.
- Can be controlled by distal visual cues. (Rotate the cues and the fields will rotate.)
- Persist in the dark – so not *dependent* on visual input.
 - Driven by path integration?
- Only about 1/3 of place cells have fields in a typical small environment.
- Cells have unrelated fields in different environments.

Place Fields in a Cylindrical and Square Arena

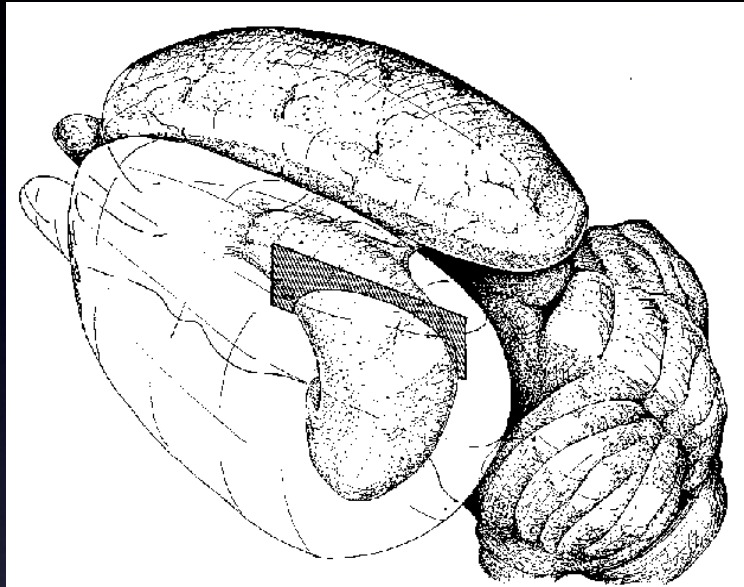
- Roughly gaussian
- Modest peak firing rates (5-10 Hz)
- Largely unrelated fields in the two environments



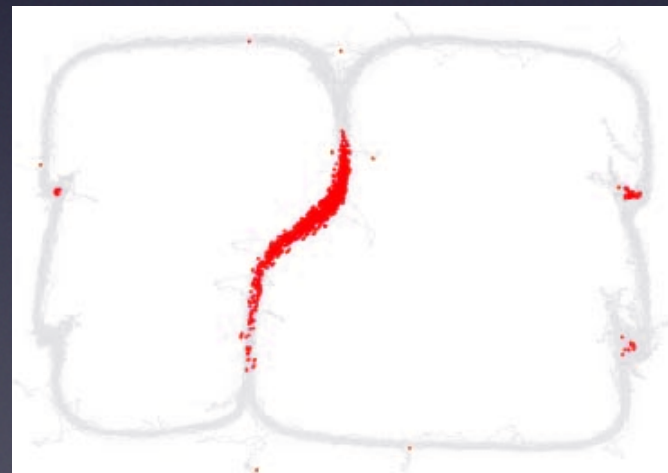
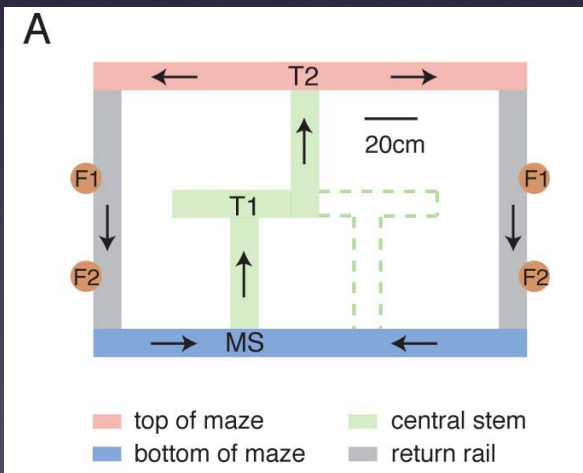
Lever et al., 2002



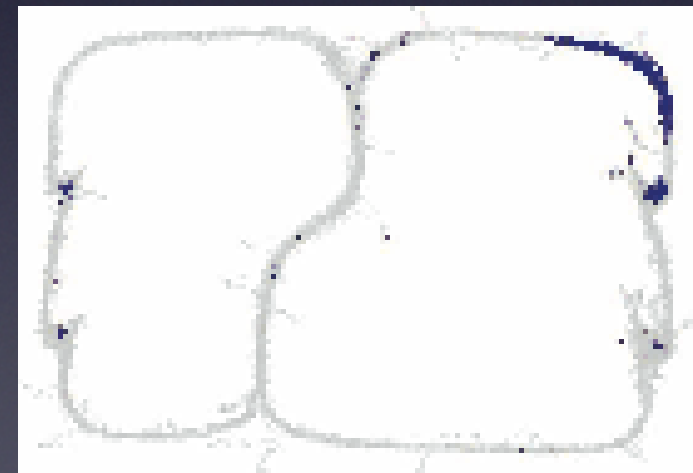
Place Fields On A Maze



Slide courtesy of Anoopum Gupta

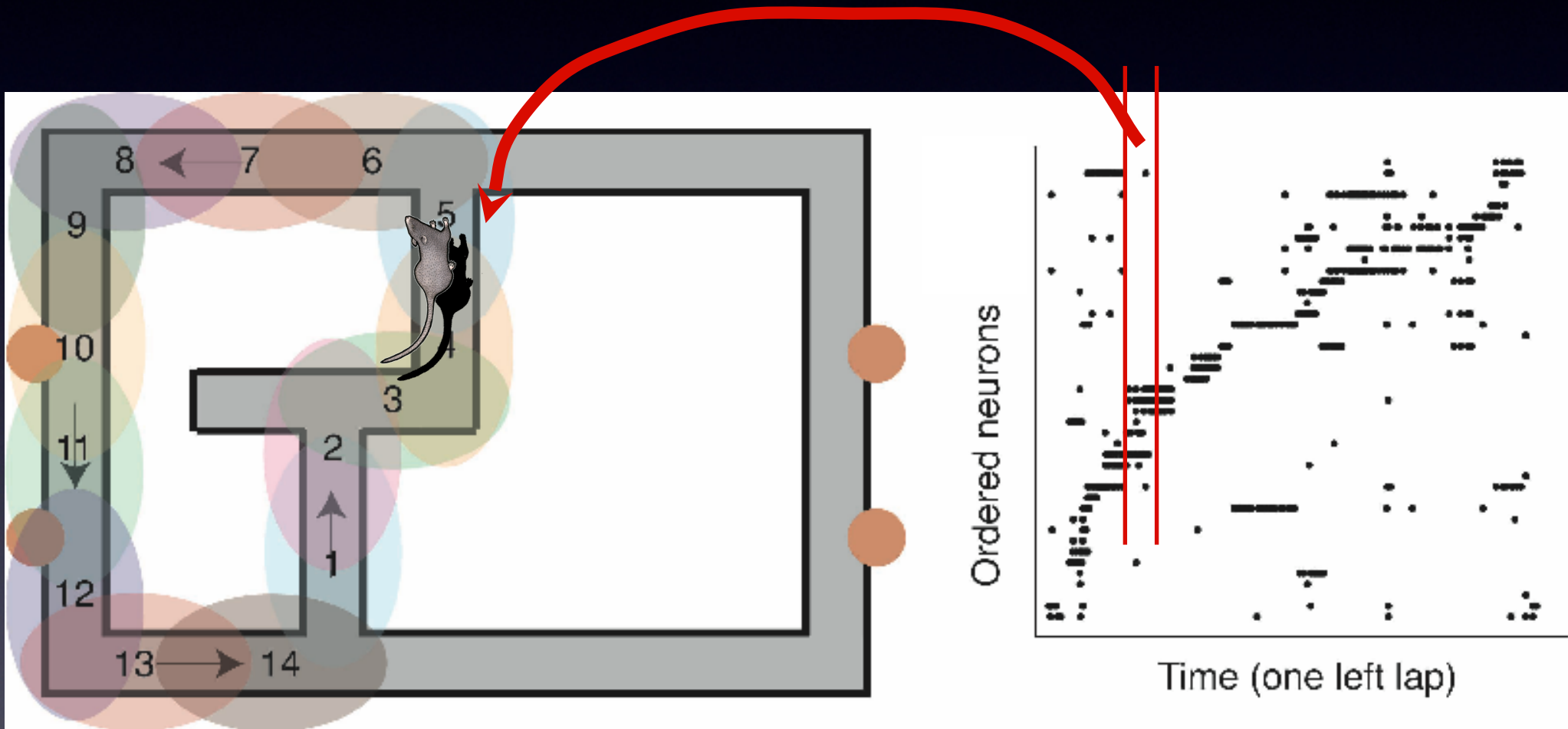


Cell 1

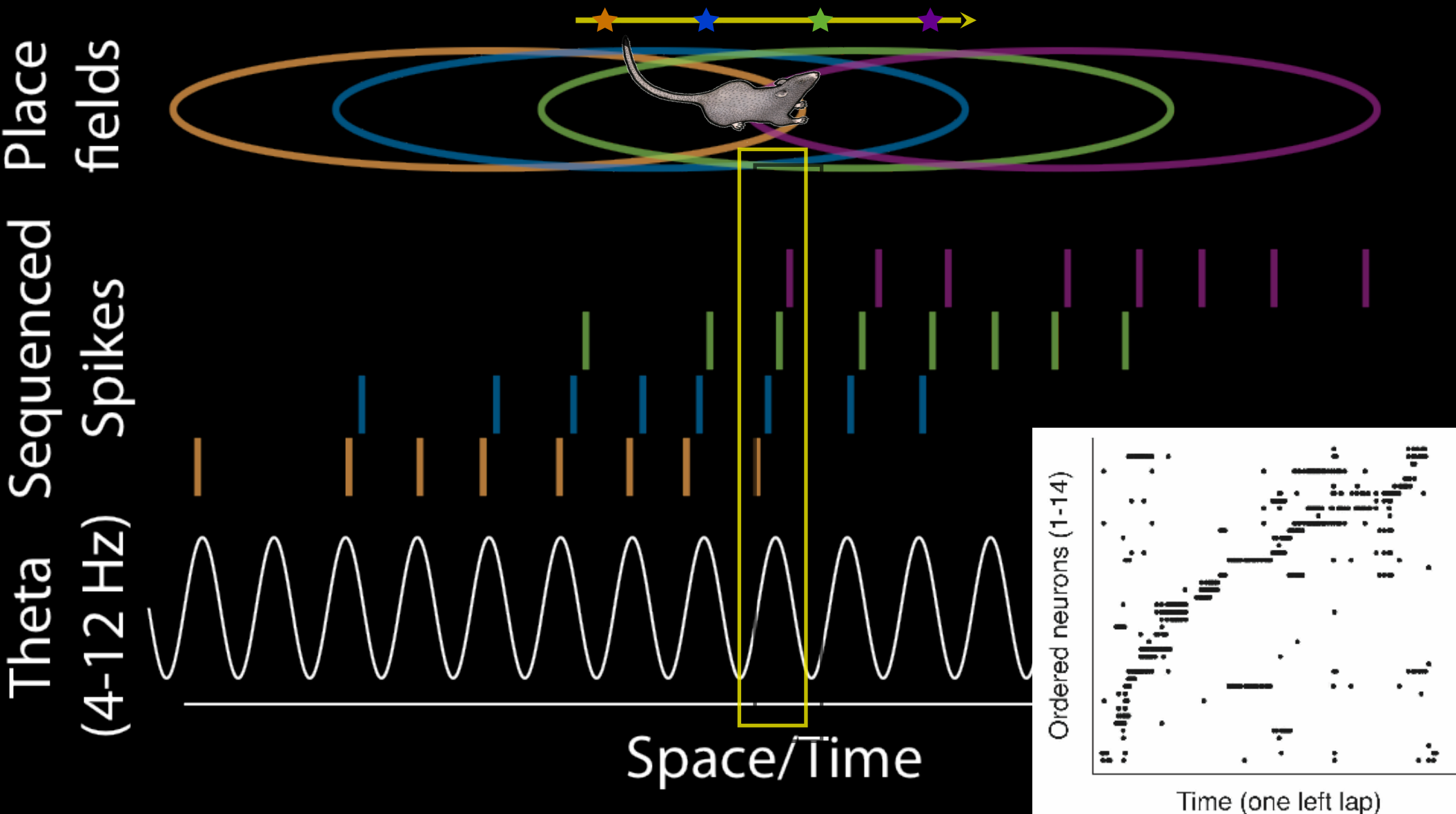


Cell 2

Neural activity during behavior

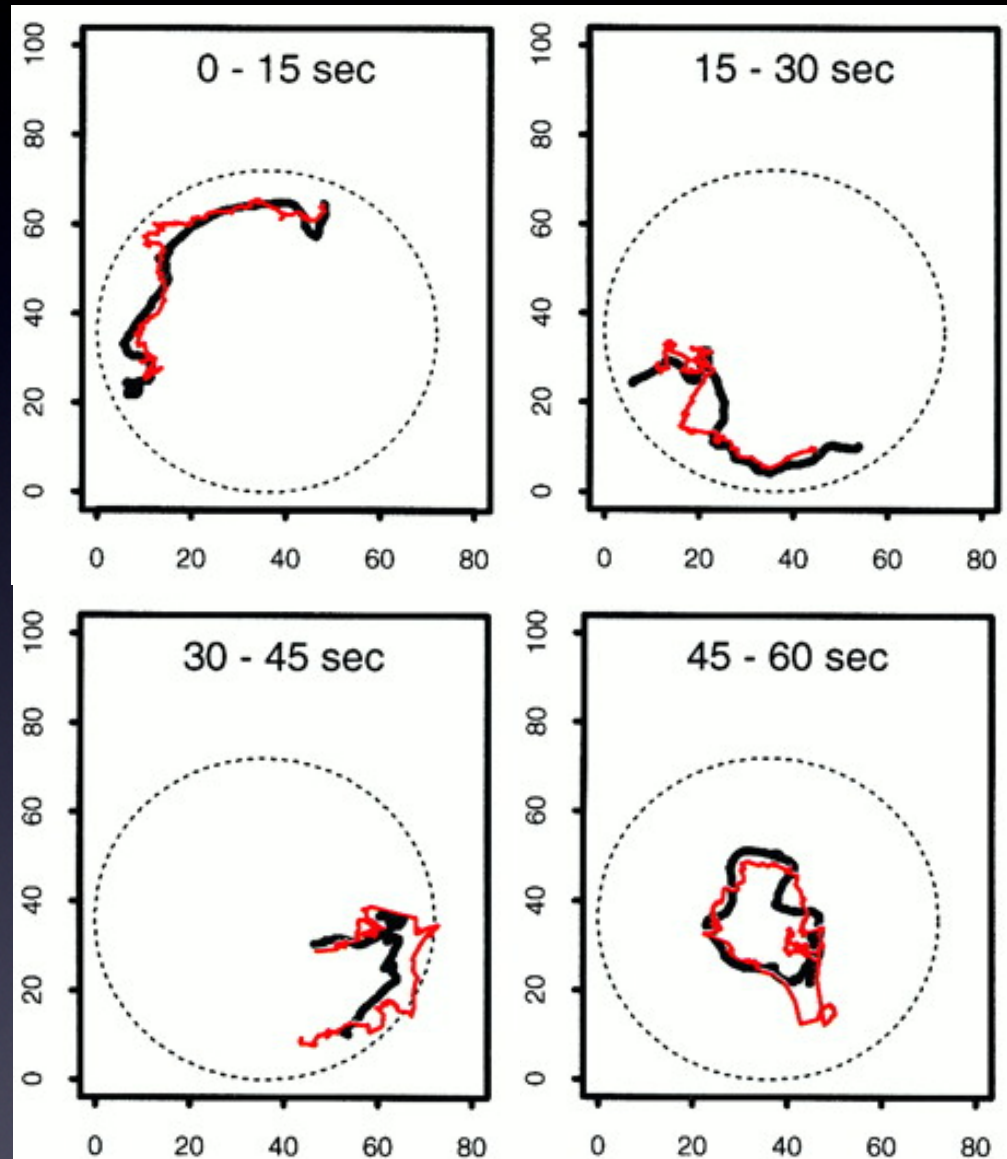


Theta Phase Precession



Slide courtesy of Anoopum Gupta

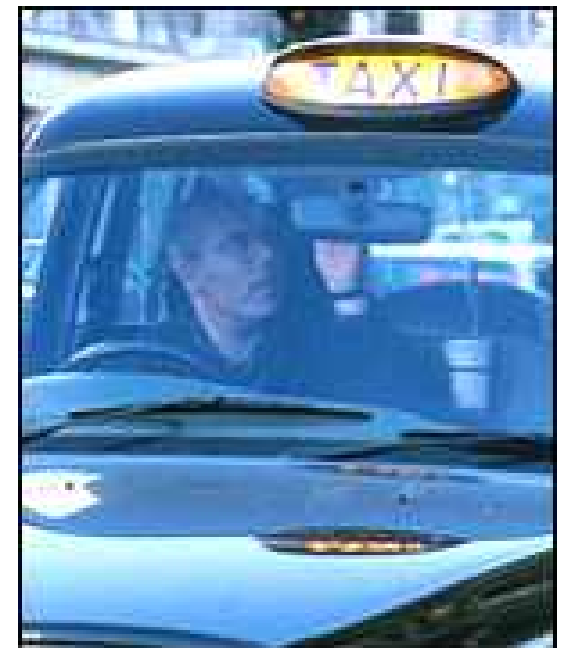
Decoded Paths



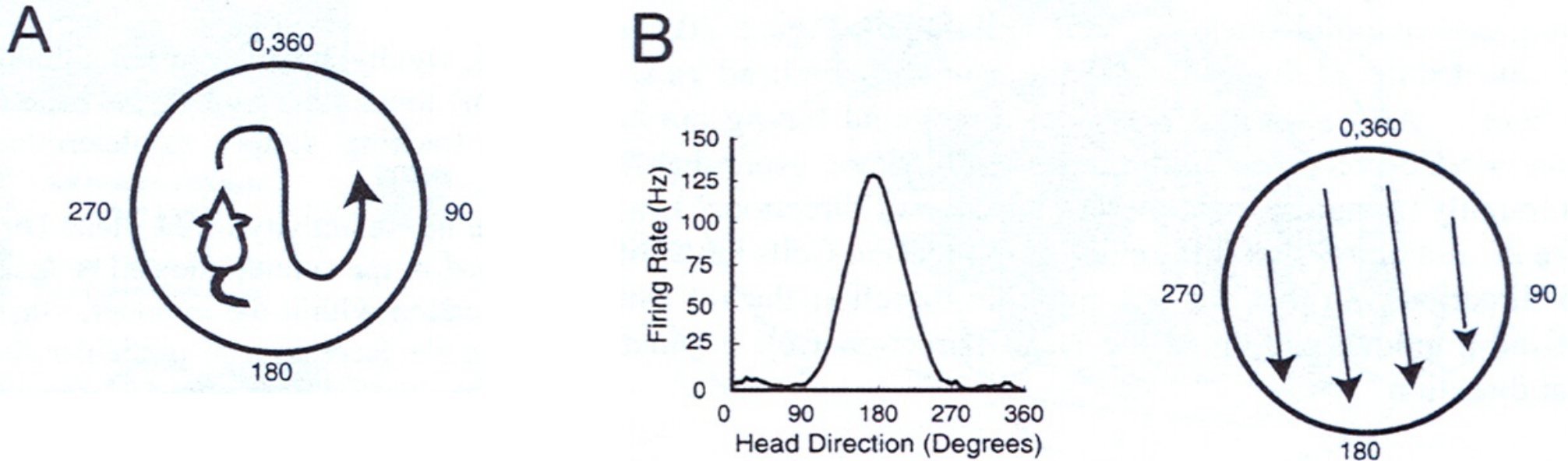
Brown et al., 1998

Eleanor Maguire: Spatial Memory in Humans

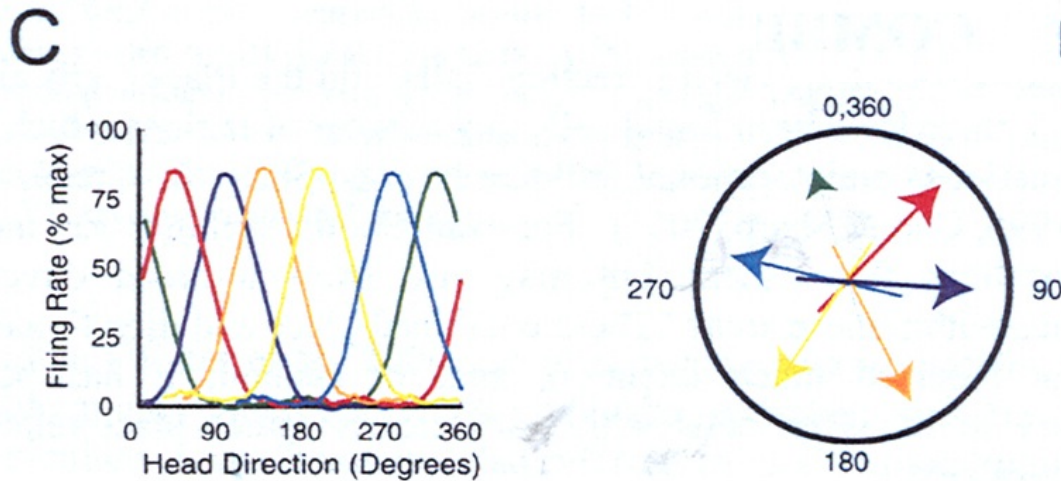
- London cab drivers undergo 2-3 years of training to learn “The Knowledge” of London's complex streets.
- Cab drivers have larger posterior hippocampi than controls. Experienced drivers show greater enlargement than new drivers.
- When remembering complex routes, drivers show elevated activity in right posterior hippocampus; no increase when answering questions about historical landmarks.



Head Direction Cells (Ranck, 1989)



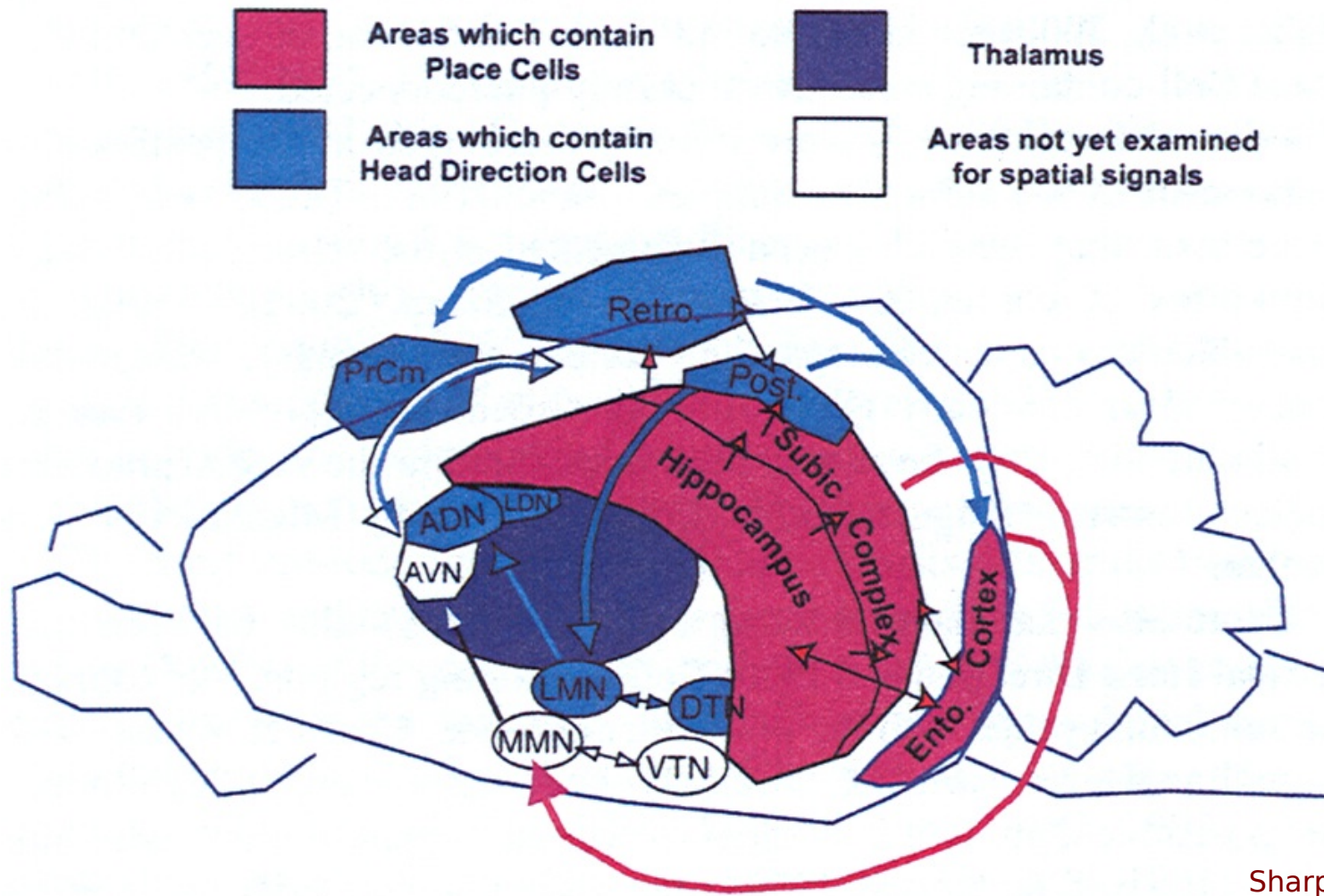
Individual Head Direction Cell



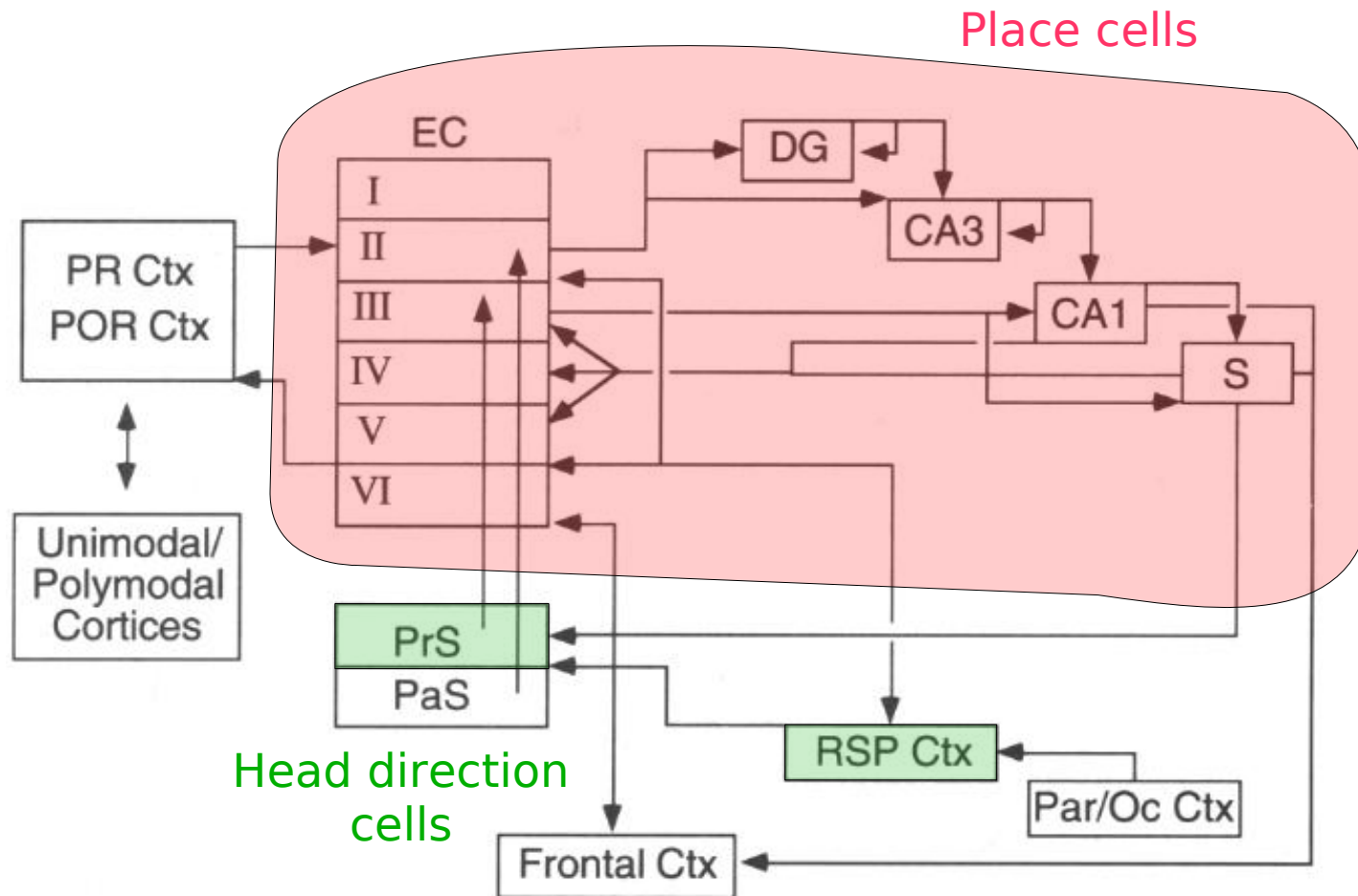
Population of Head Direction Cells

Figures from Sharp (2002)

Place and Head Direction Systems



Rodent Navigation Circuit

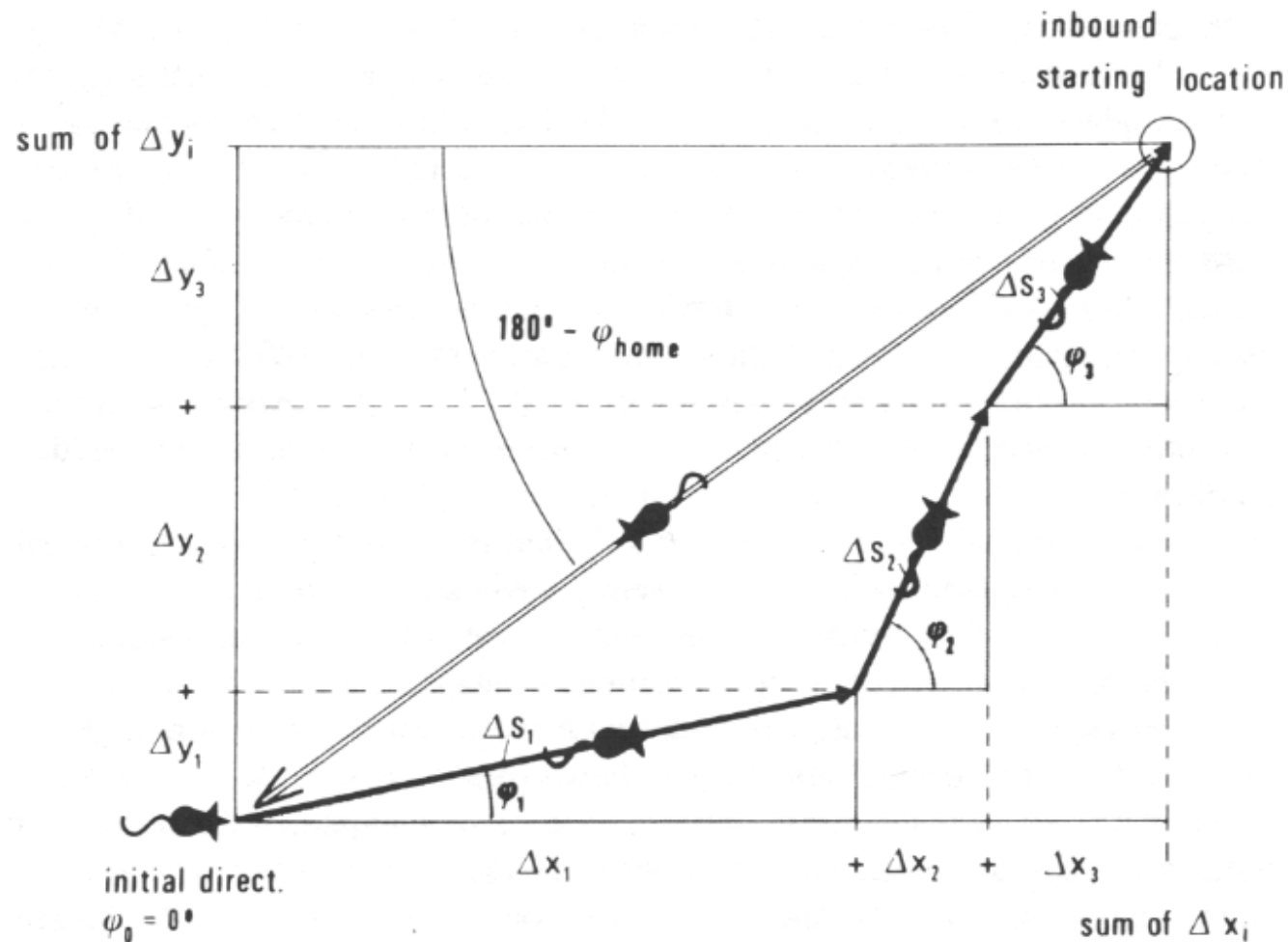


From (Johnston & Amaral, 1998)

PR: perirhinal cortex; POR: postrhinal cortex; EC: entorhinal cortex; PrS: presubiculum; PaS: parasubiculum; DG: dentate gyrus; CA: Cornu amonis; S: subiculum; RSP: retrosplenial cortex; Par/Oc: parietal/occipital cortex

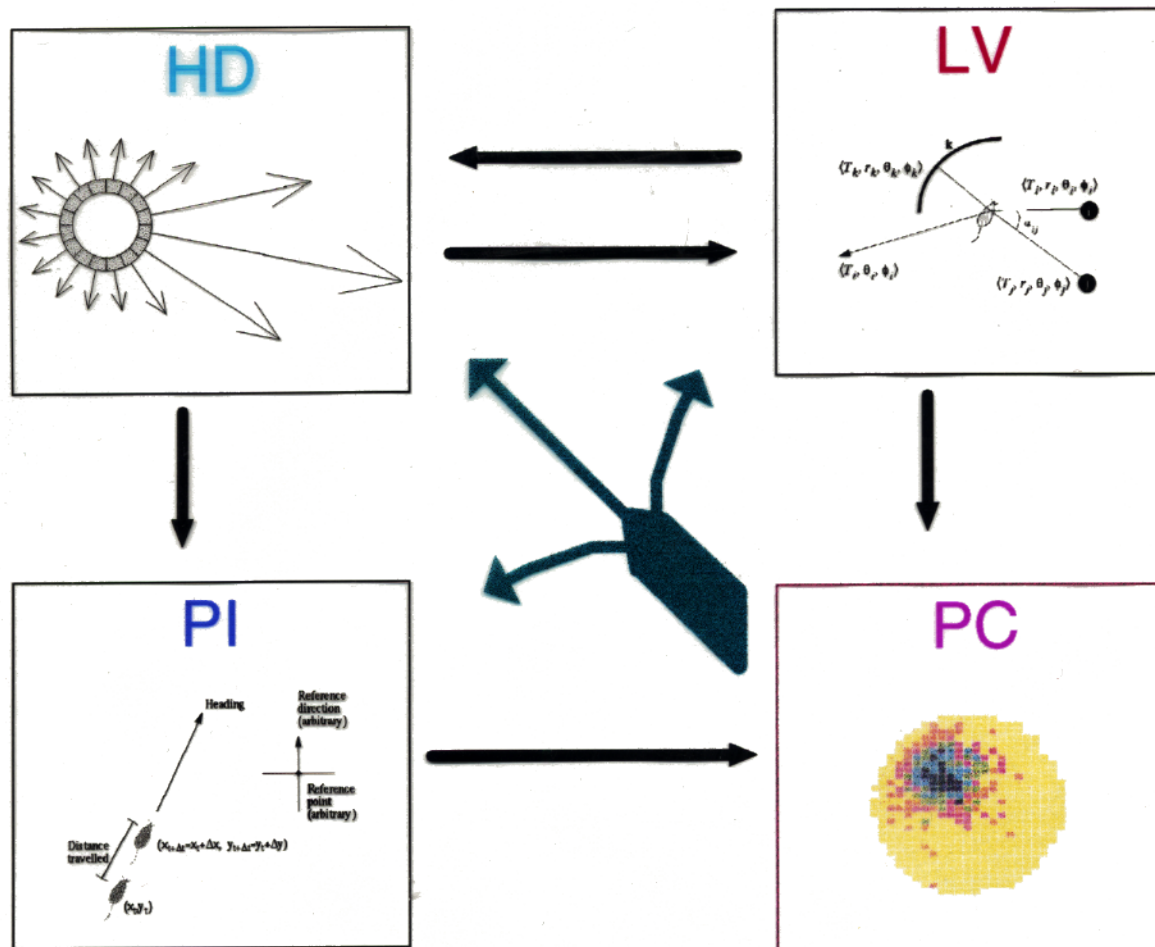
Path Integration in Rodents

Mittelstaedt & Mittelstaedt (1980): gerbil pup retrieval



Redish & Touretzky Model of Rodent Navigation

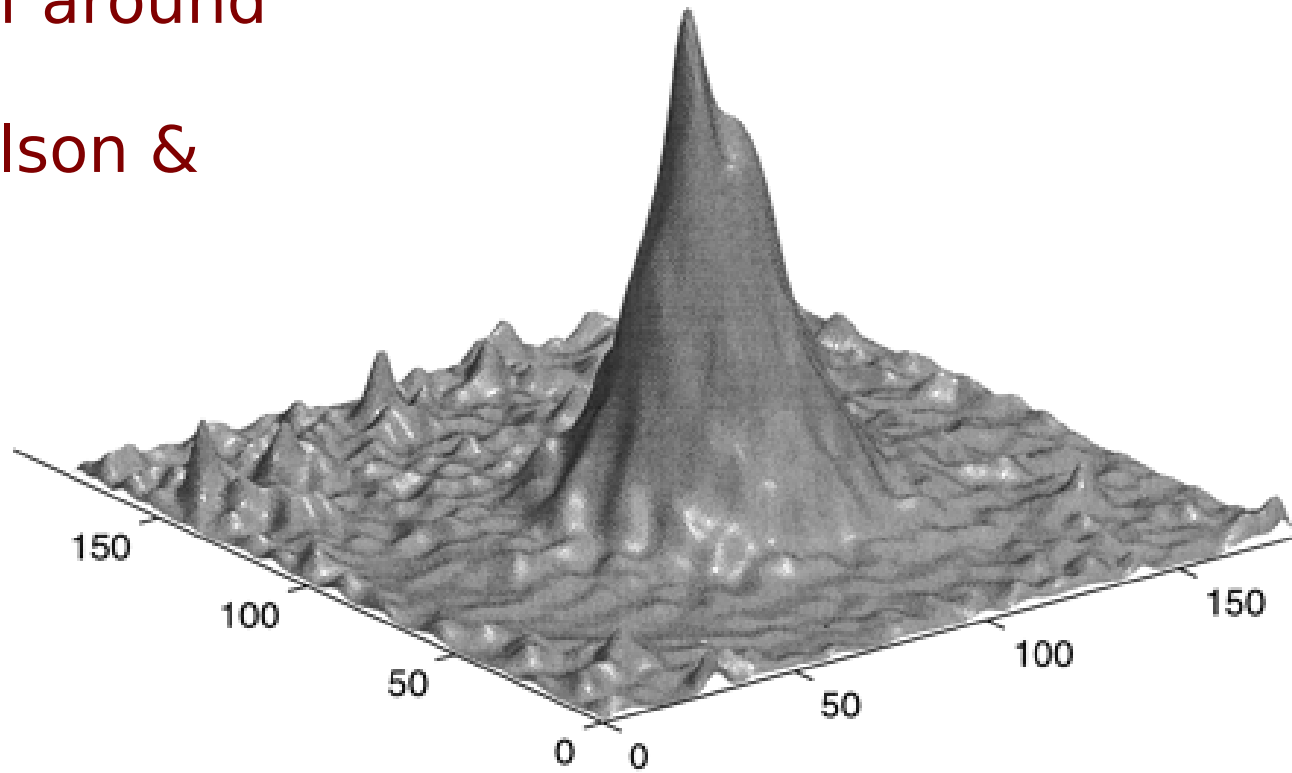
Place cells *learn* and *maintain* the correspondence between local view representations and path integrator coordinates.



Redish (1997)

Hippocampal State: A Moving Bump of Activity

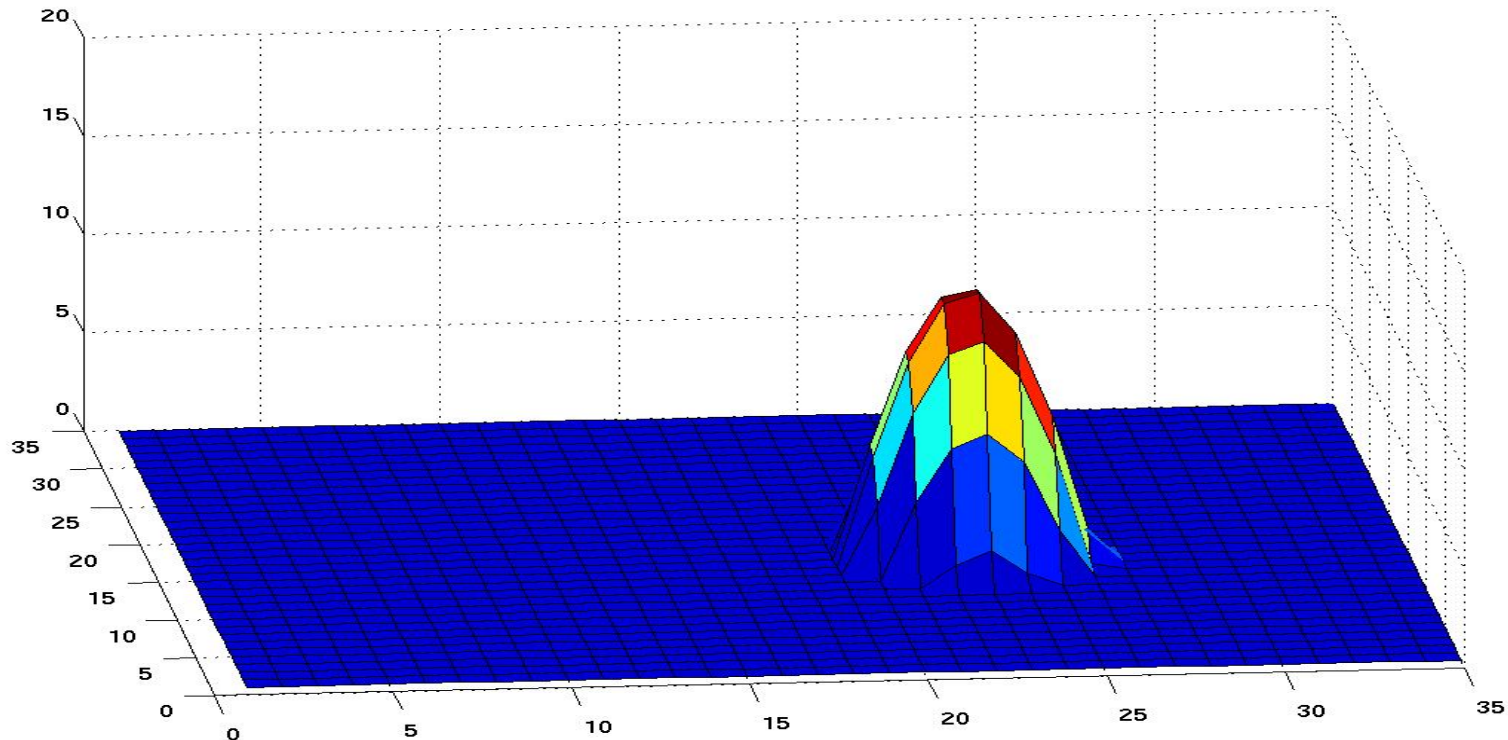
Activity packet reconstructed
from firing patterns of around
100 cells recorded
simultaneously by Wilson &
McNaughton (1993)



Samsonovich & McNaughton (1997)

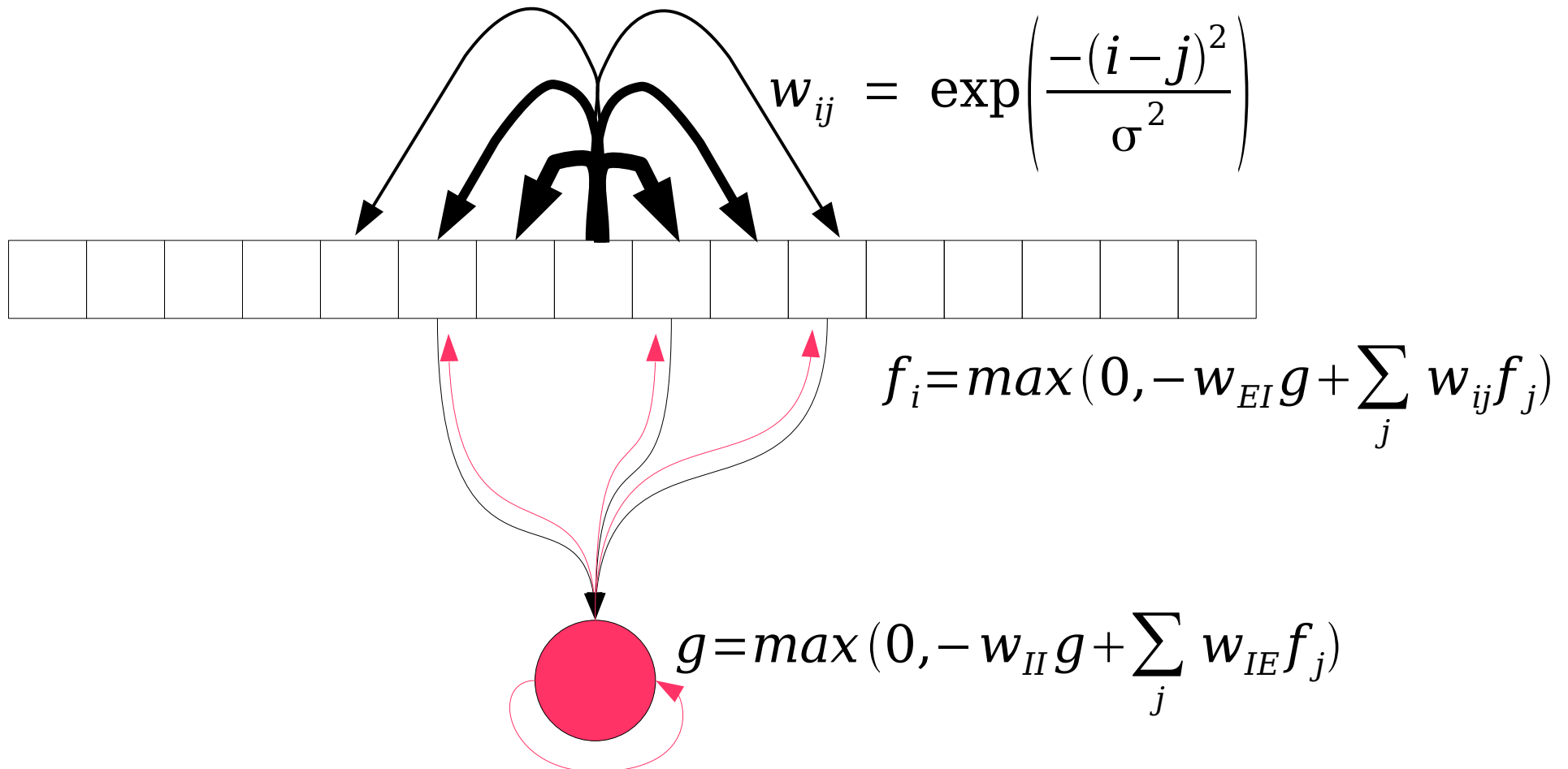
2D Attractor Bump Simulation

- In 1972, Amari, and Wilson & Cowan demonstrated continuous attractor bumps in a recurrent network.
- 25 years later: Samsonovich & McNaughton (1997): 2D attractor bump model of place cells.
- Bumps are easy to simulate and visualize in MATLAB.



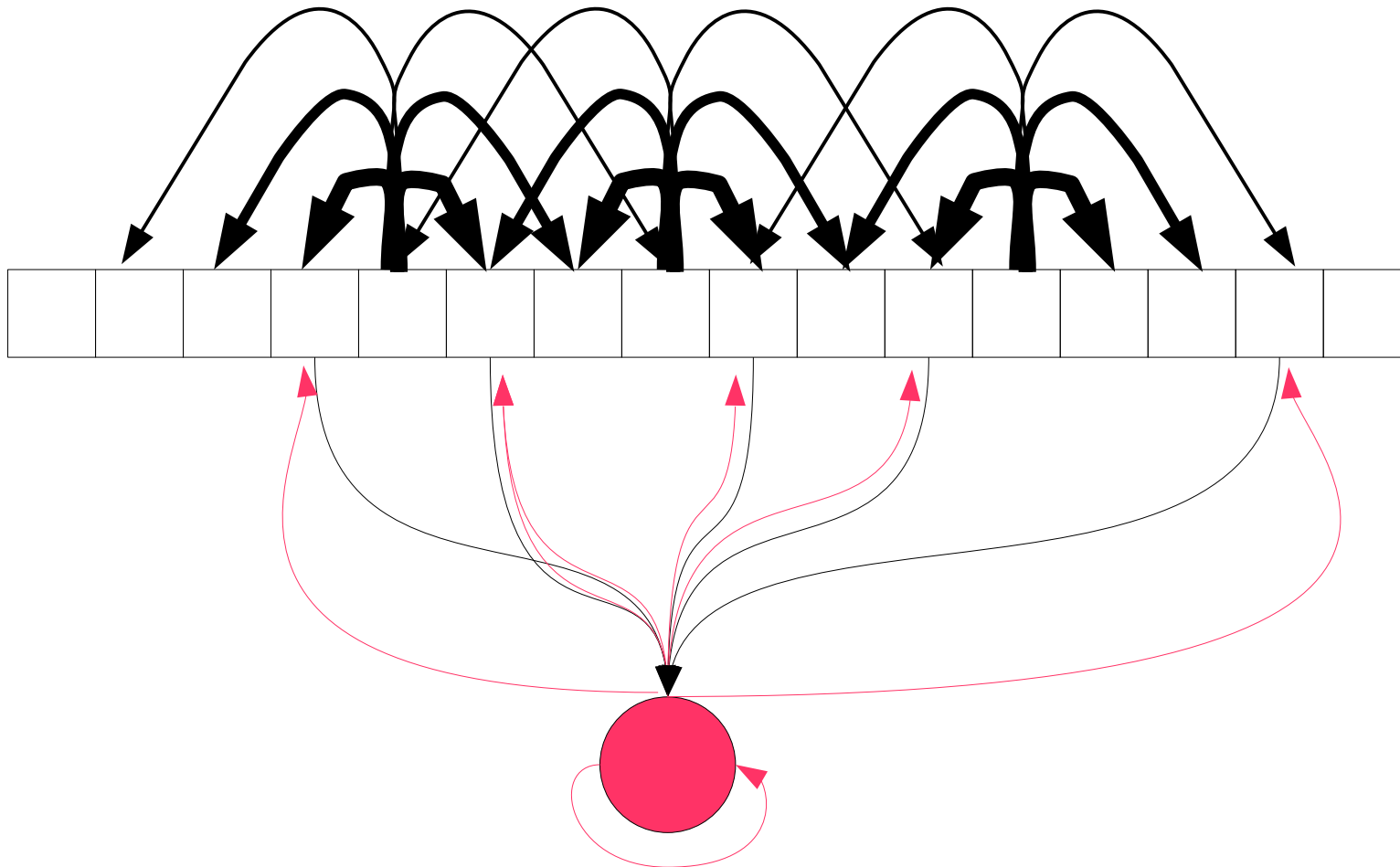
How to make a bump (1D version)

Local excitation plus global inhibition:

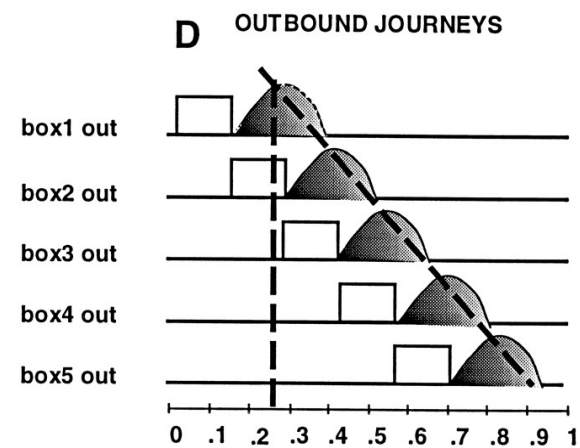
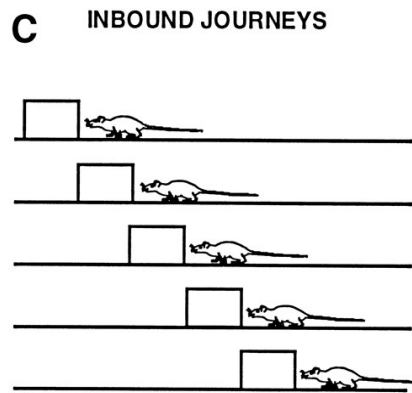
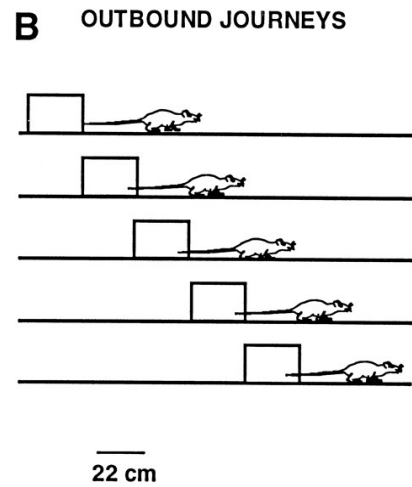
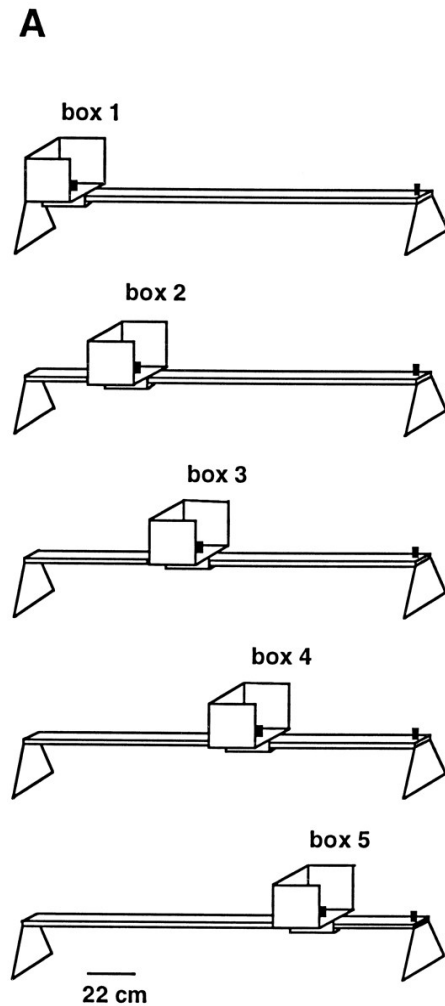


How to make a bump (1D version)

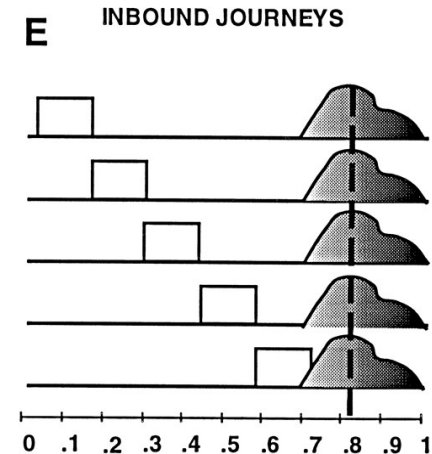
Same weights for every unit (shifted):



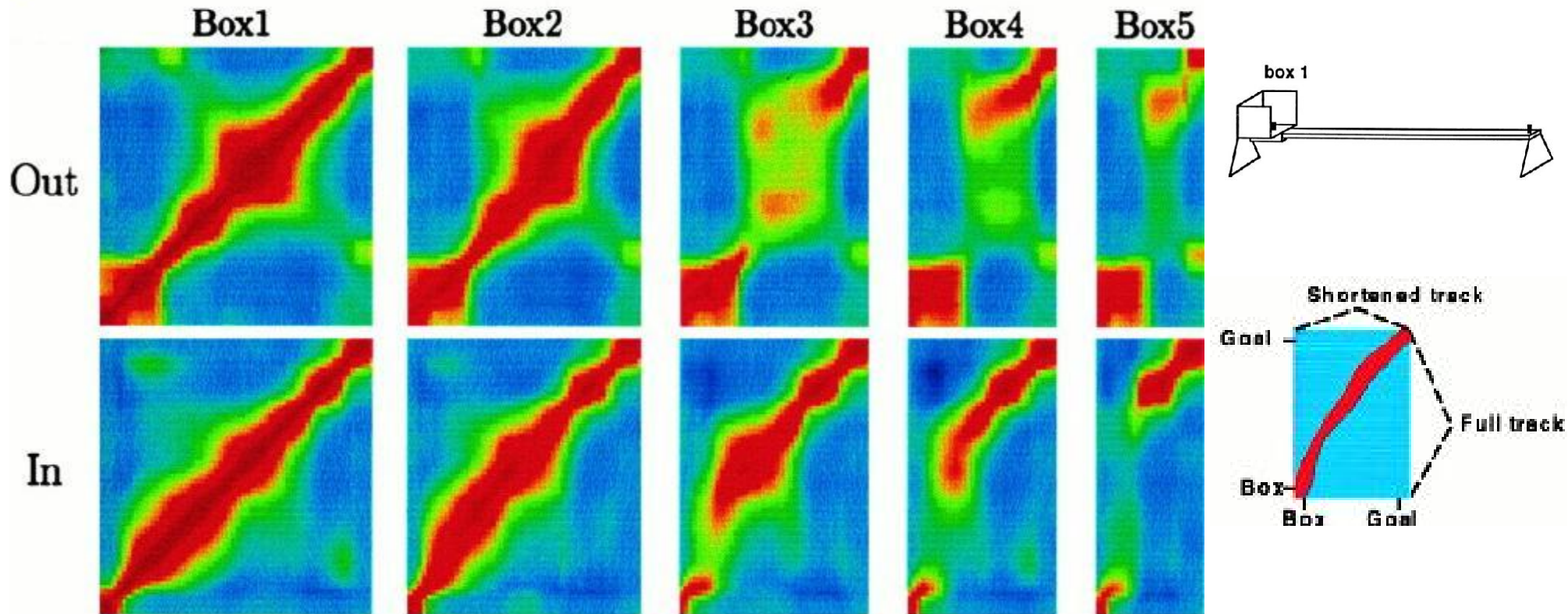
Gothard et al. (1996): bump jumps



From (Gothard et al., 1996)



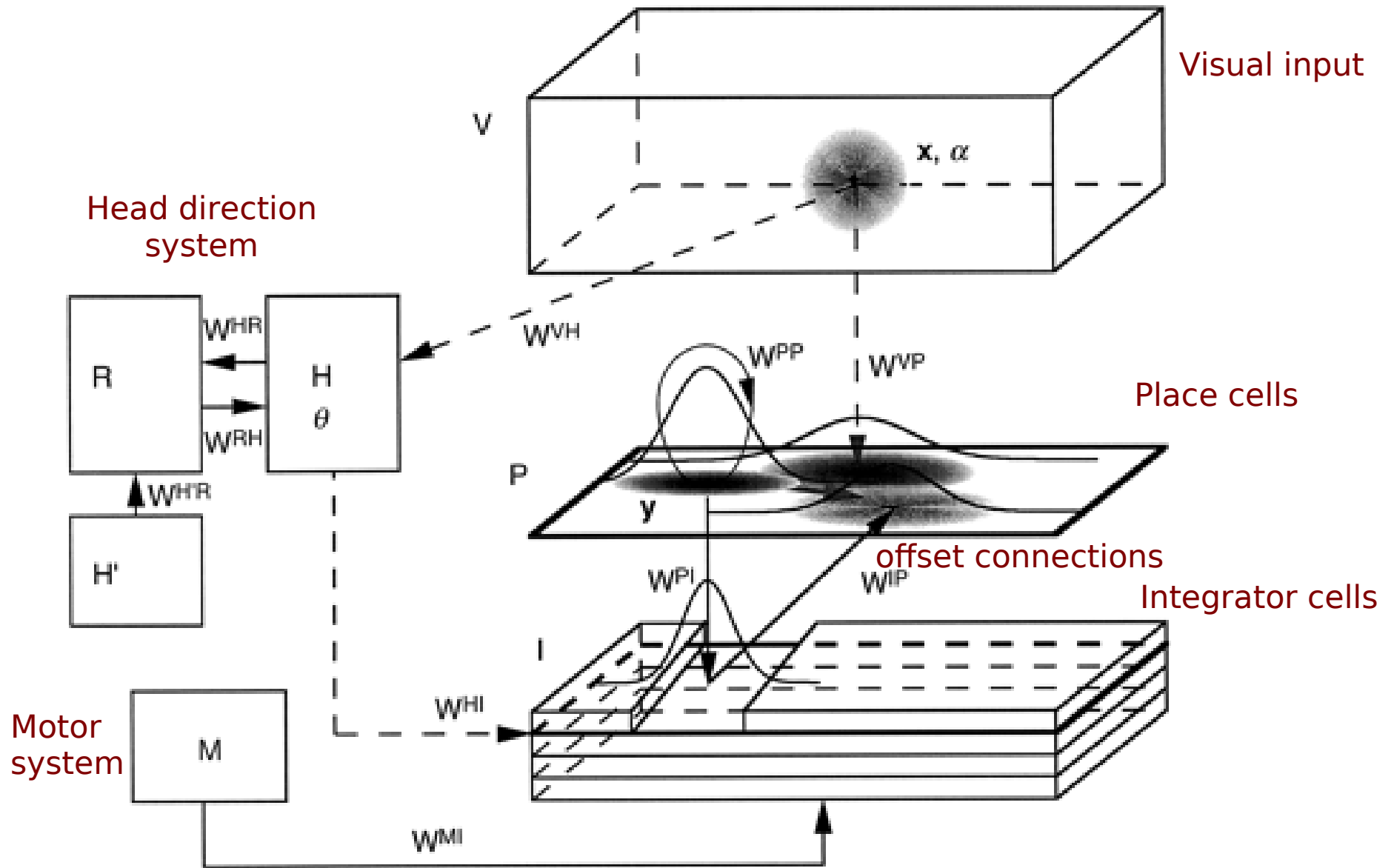
Watch the bump jump!



From (Gothard et al., 1996)

Cross-correlation plots of the ensemble activity patterns show a “jump” on the map as a discontinuity.

Samsonovich & McNaughton Model

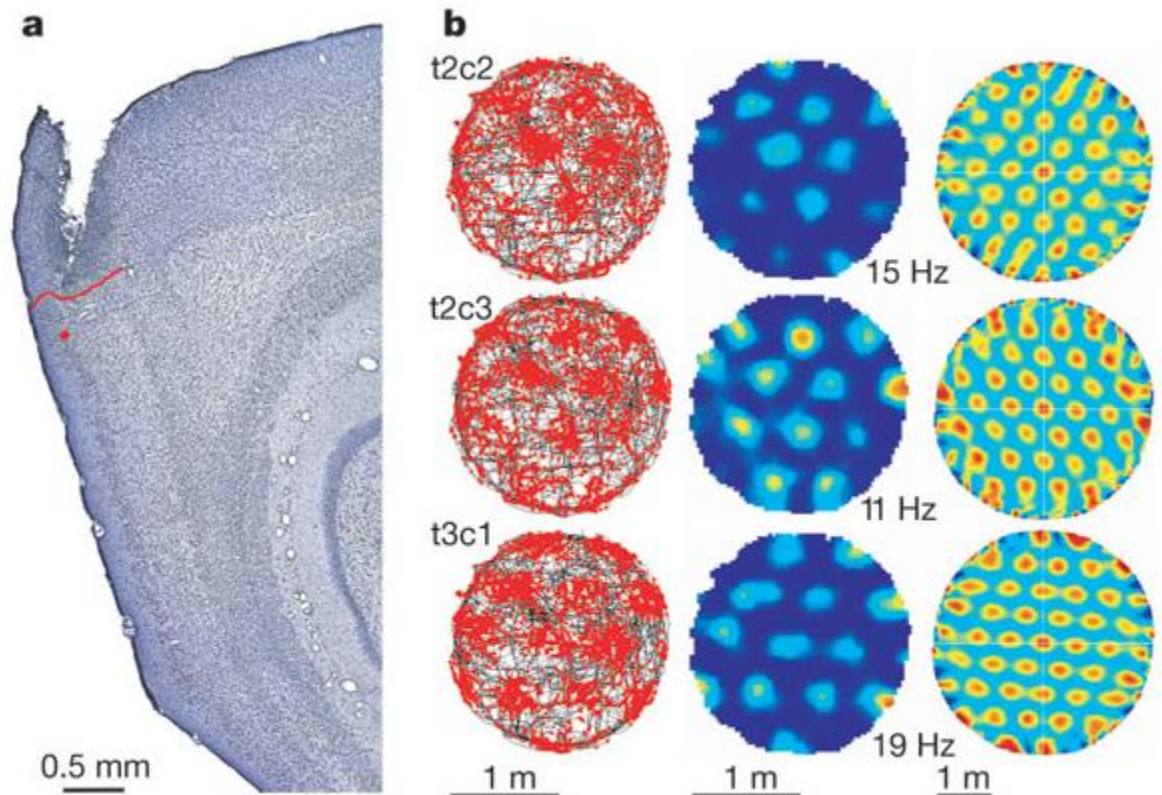


Where is the Path Integrator?

- Early theories (McNaughton) placed it in hippocampus.
- Redish & Touretzky: it can't go there, because multiple maps make it too hard to update position.
- Fyhn et al. (Science, 2004) found the PI in medial entorhinal cortex: “grid” cells.

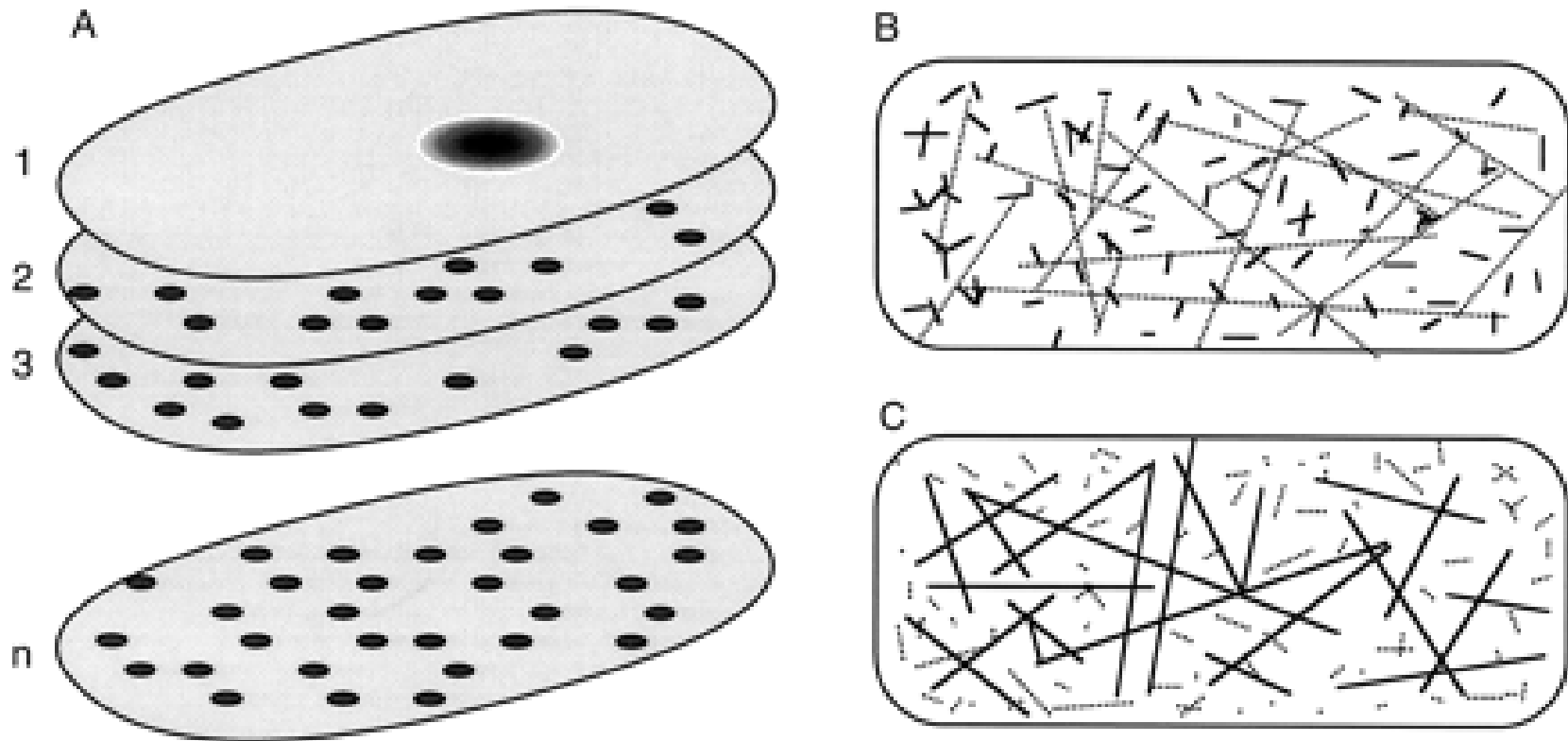


May-Britt and Edvard Moser,
2014 Nobel Laureates in
Physiology or Medicine

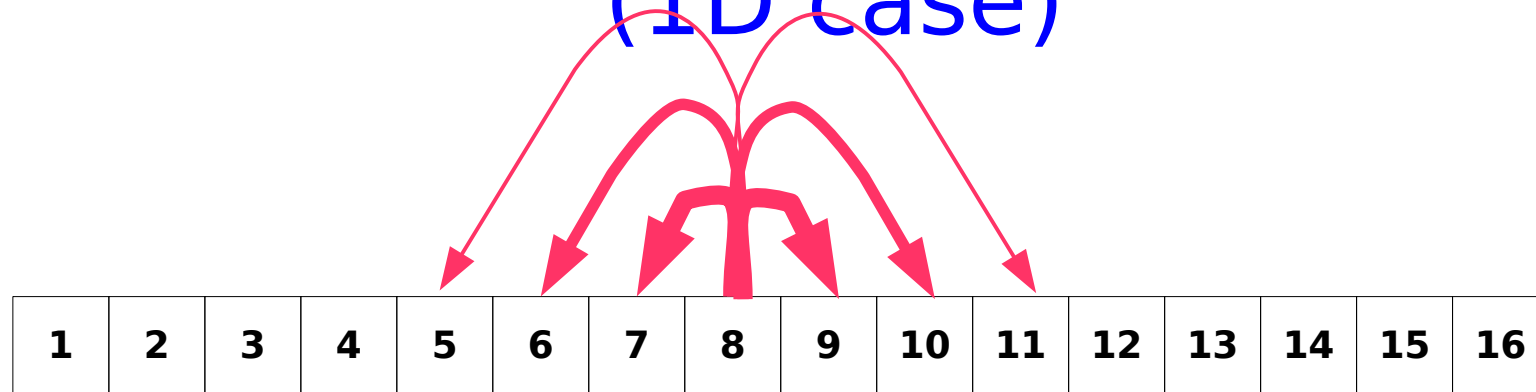


Multiple Maps in Hippocampus

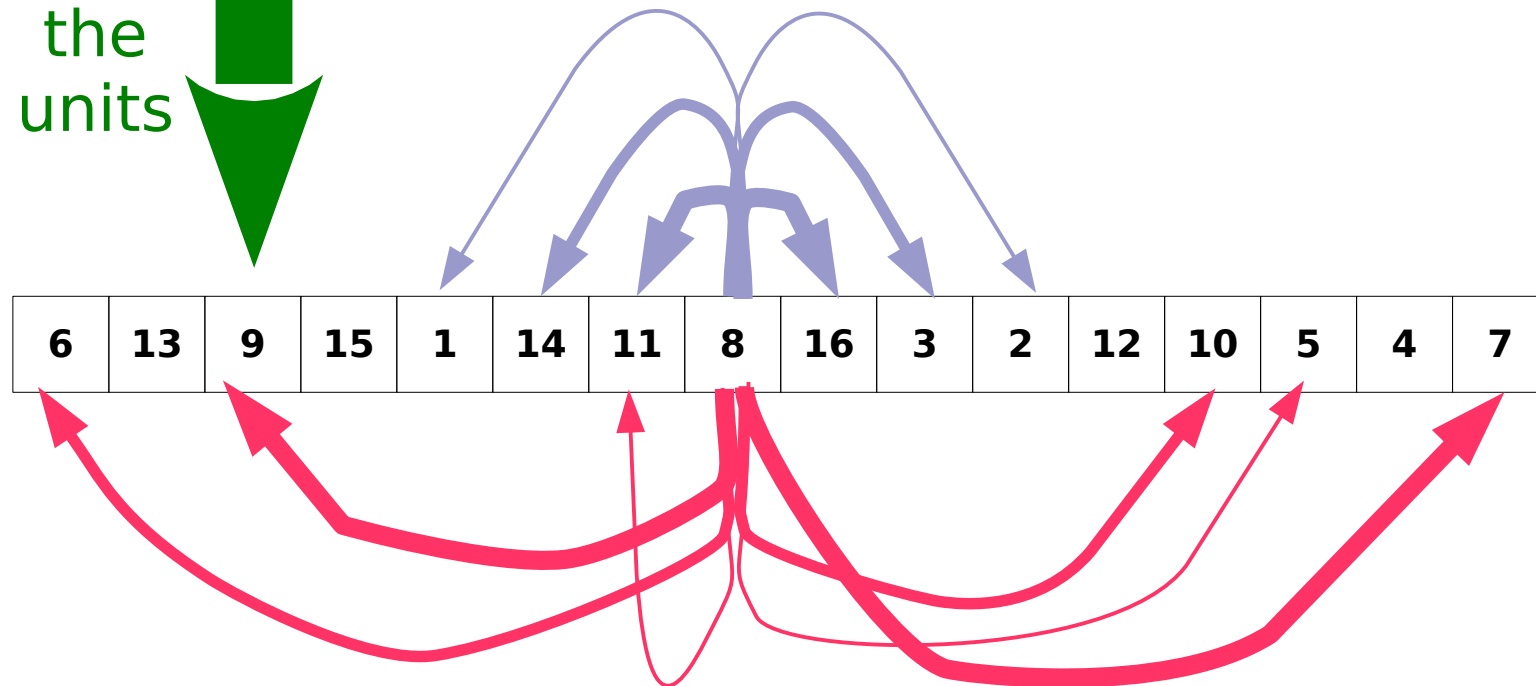
Samsonovich & McNaughton's "charts" proposal:



How to make multiple maps (1D case)

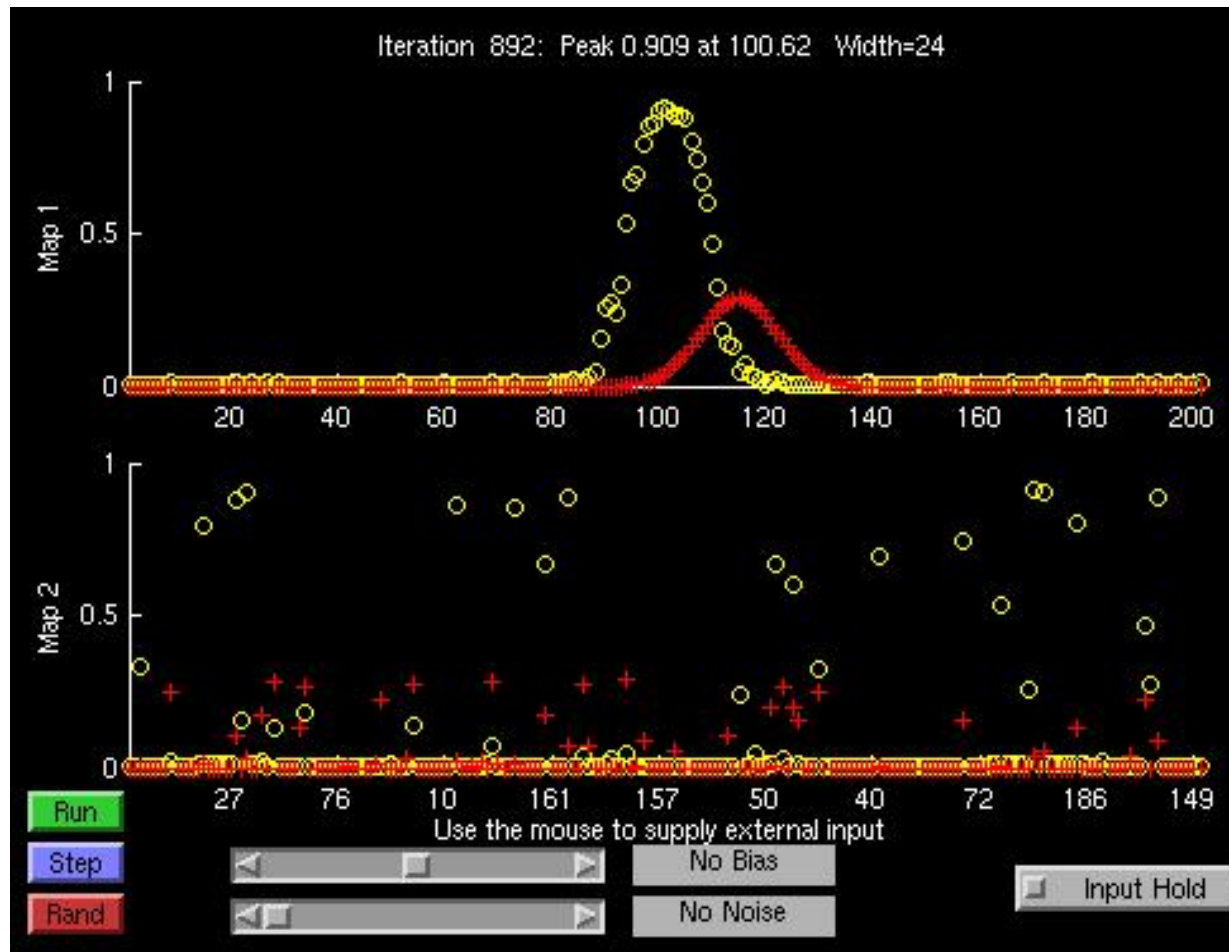


Shuffle
the
units

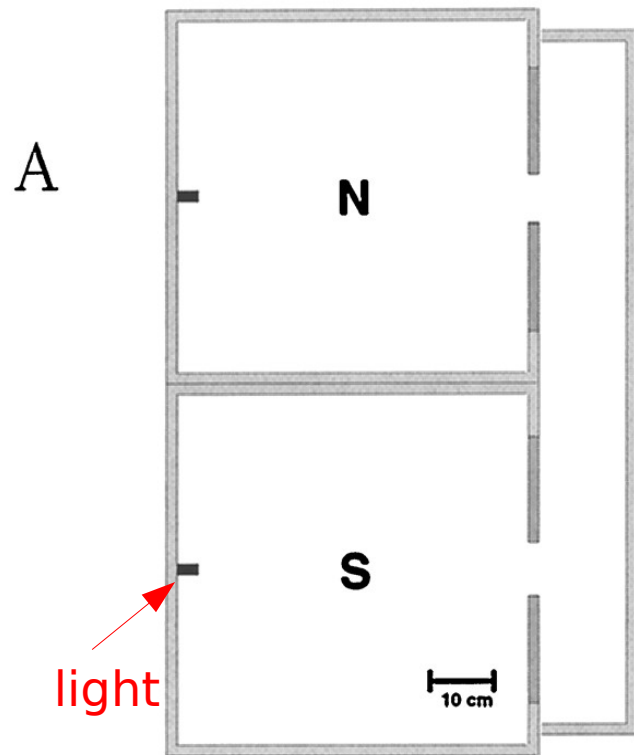


Multiple Maps Can Co-Exist In An Attractor Network

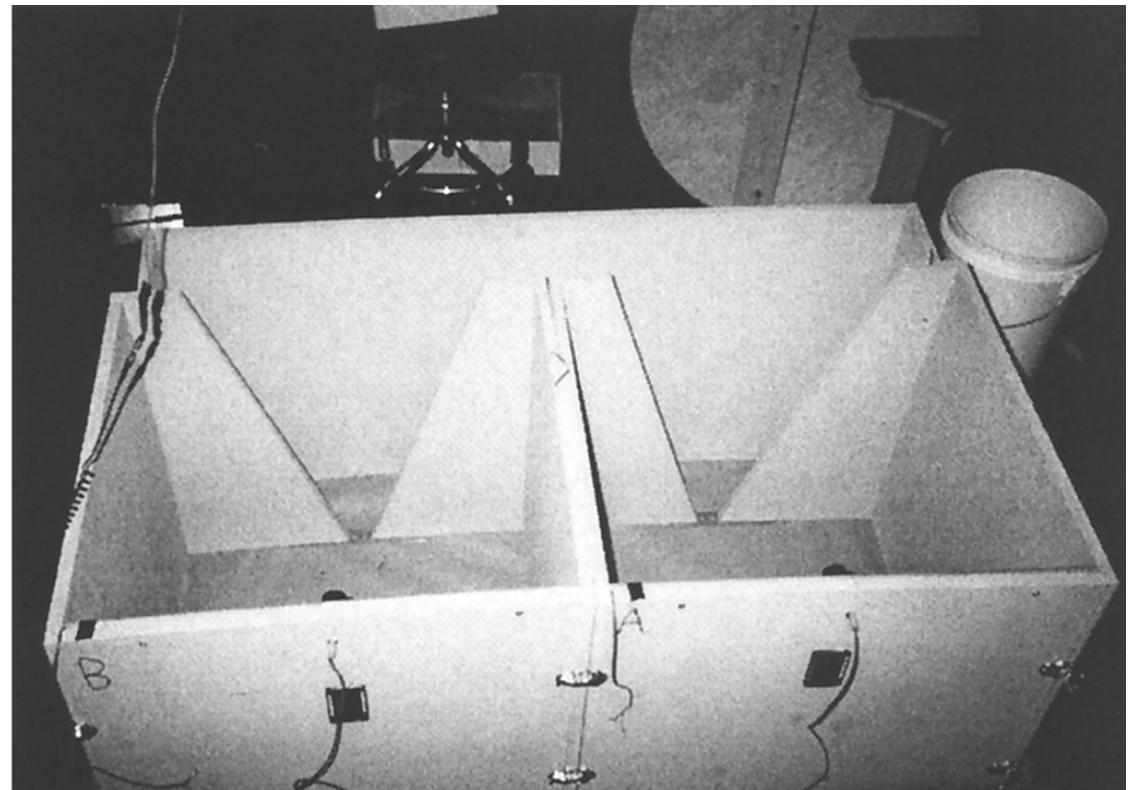
Because activity patterns are sparse, the weight matrix is also sparse. Interference isn't too bad.



Skaggs & McNaughton (1998): Partial Remapping in Identical Environments



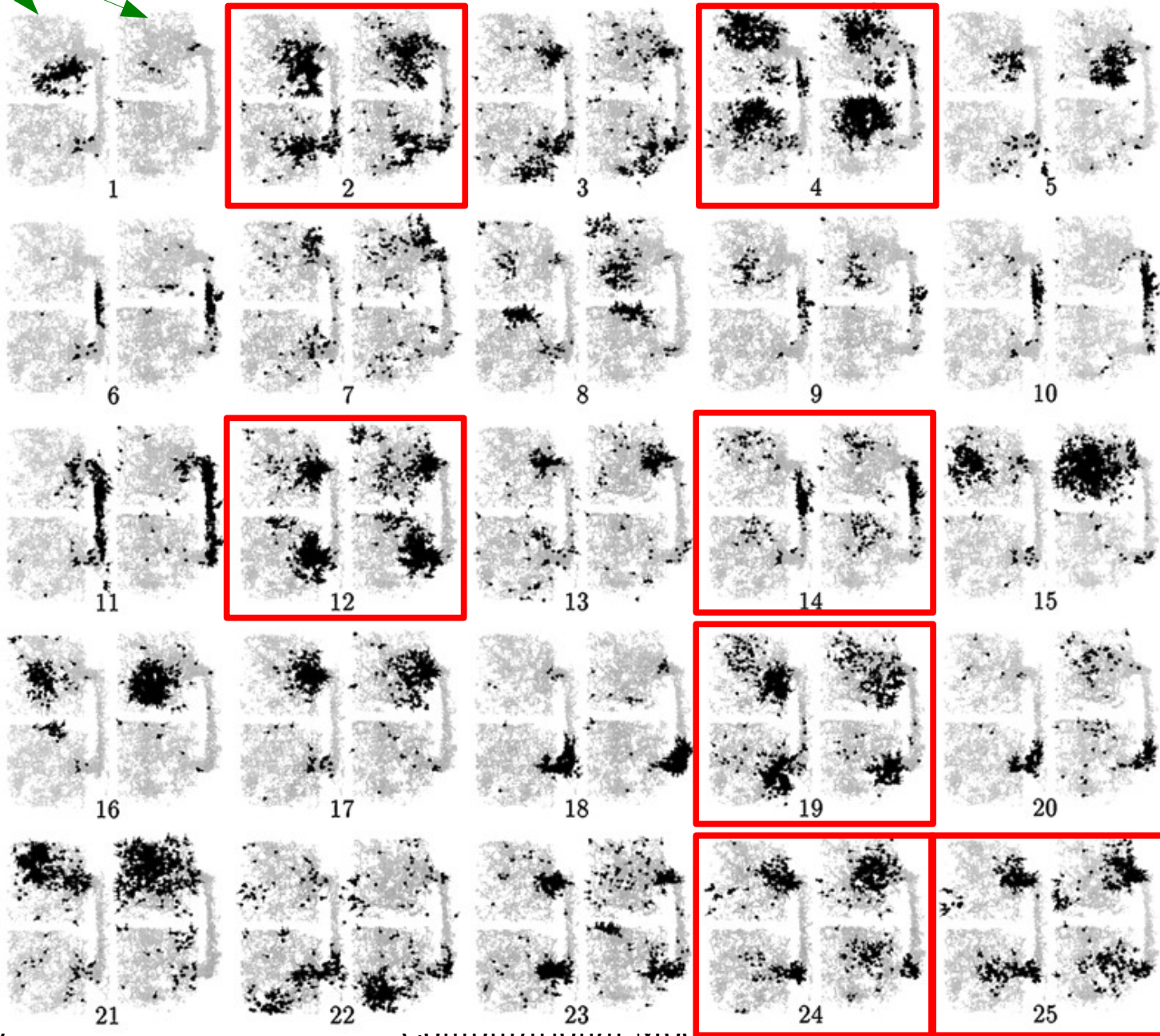
B



(Skaggs & McNaughton, 1998)

Identical Environments, Similar Fields in Both Boxes

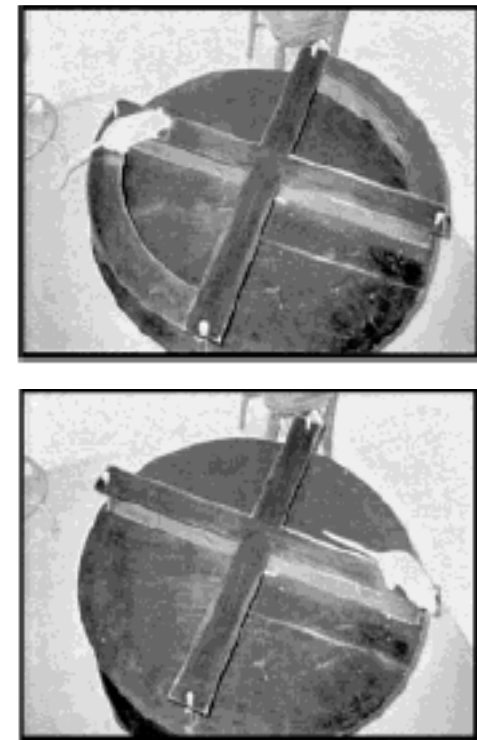
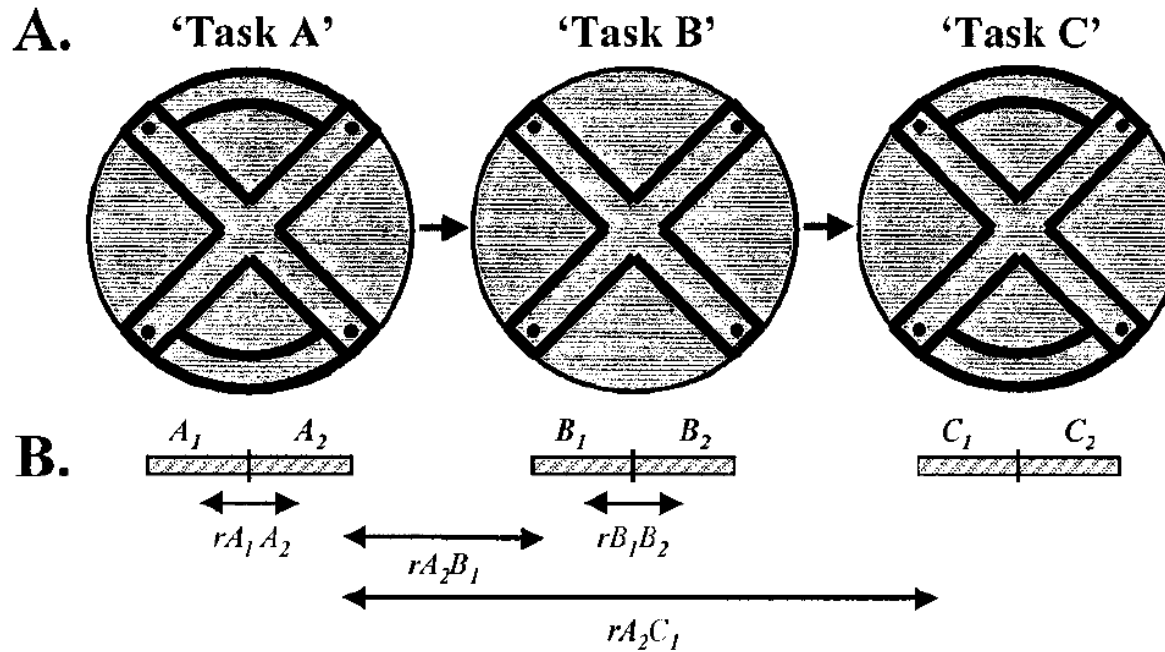
Same cell;
two sessions



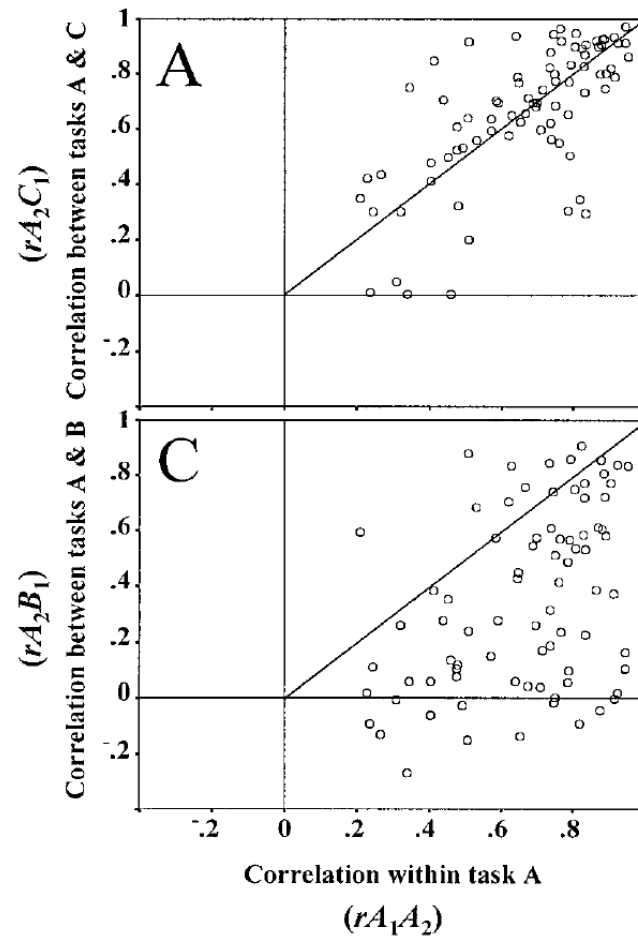
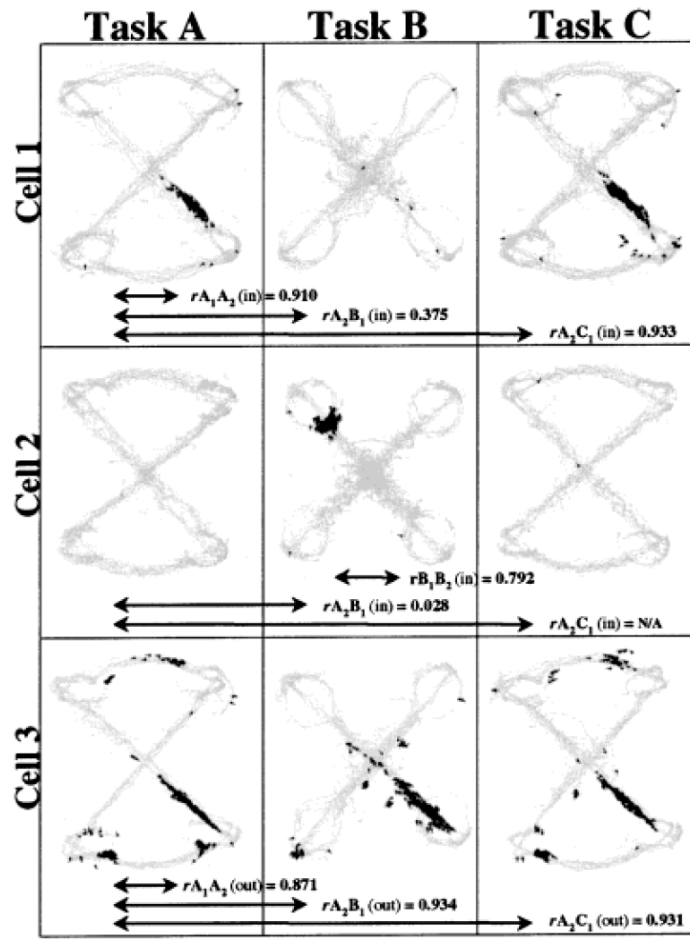
Skaggs &
McNaughton
(1998), Fig. 2.

Task-Dependent Hippocampal Remapping

Oler and Markus (2000) recorded from DG, CA3, and CA1 while animals ran either on a Figure-8 or Plus maze.



Task-Dependent Remapping



Some but not all fields remapped depending on which task was being performed

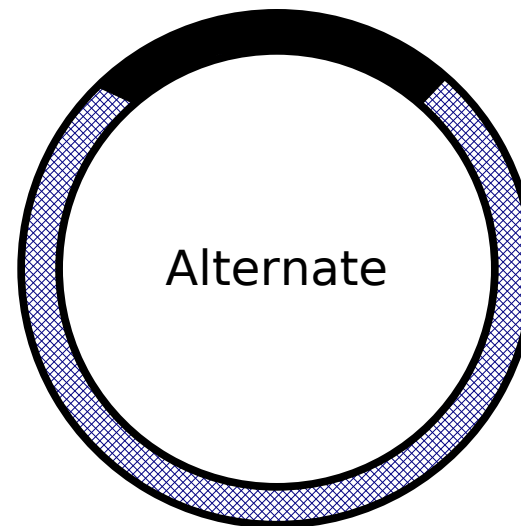
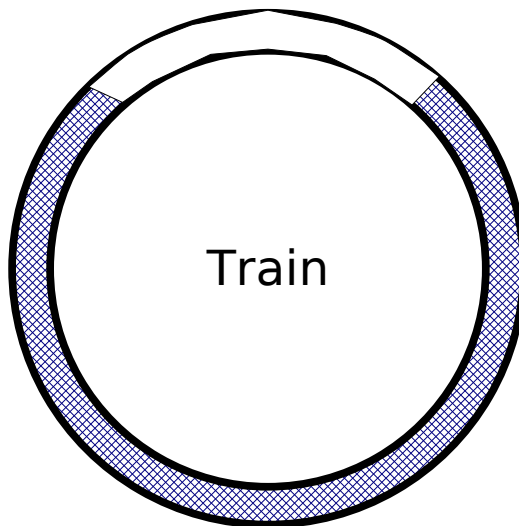
Experience-Dependent Remapping

In some circumstances, rats don't remap right away:

- **Onset** may be delayed.
 - So cannot be direct result of a sensory change.
 - Must reflect some internal change in the rat's representation of its environment: learning.
- **Rate** may be gradual.
 - The time course of remapping tells us something about the experience-dependent learning process.
- **Extent** may be partial or complete.
- What **learning algorithm** is responsible for these experience-dependent changes?

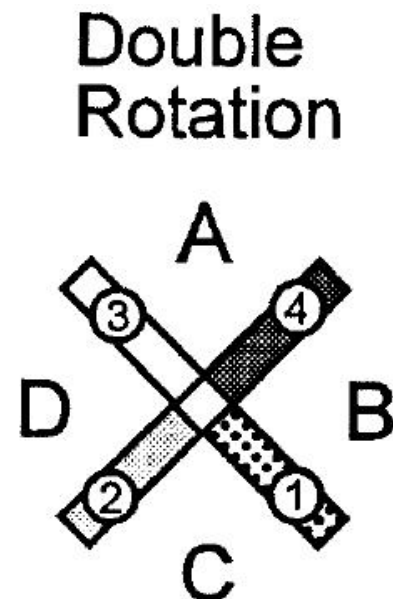
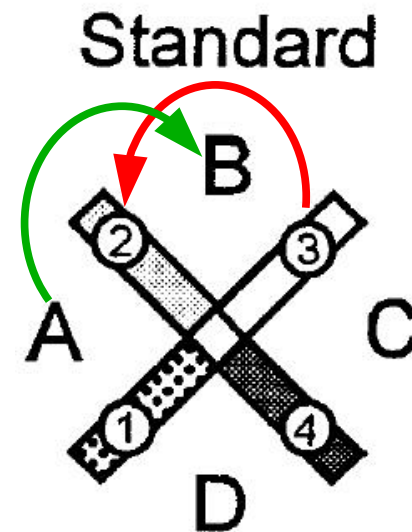
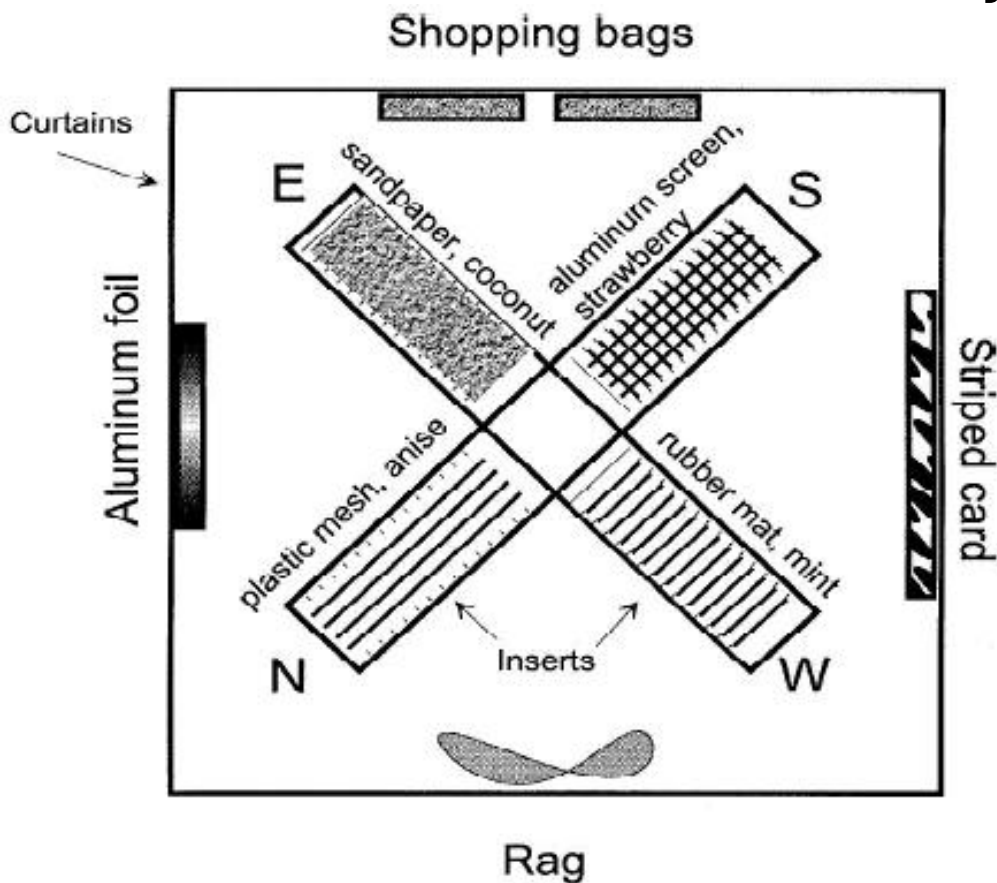
Bostock et al. (1991): Delayed Abrupt Complete Remapping

- Train in cylinder with white card, then alternate exposure to white and black cards.
- Most rats did not remap upon first exposure to black card.
- But once a rat remapped, it continued to do so.



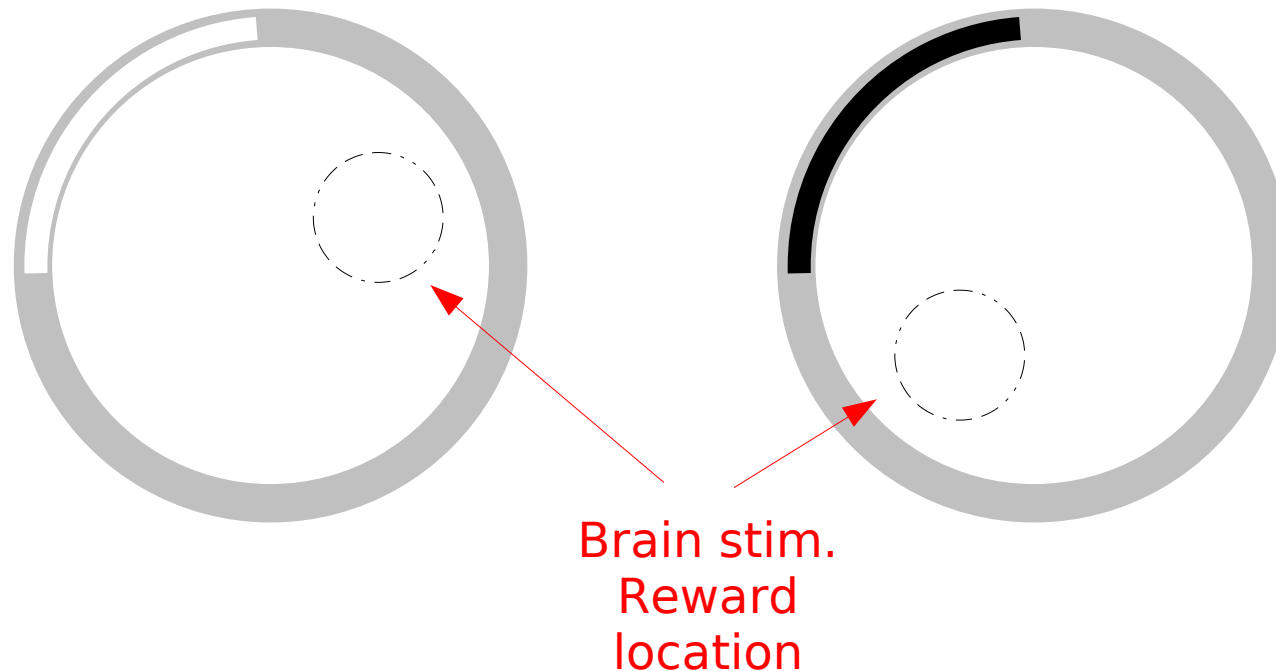
Tanila et al. (1997): Gradual Remapping

- Discordant responses: some cells followed local cues, some followed distal, some remapped. The extent of remapping appeared to increase over several days. (Based on data summed over rats.)
- Is the rat becoming more *certain* that the two environments are reliably different?



Does Remapping Matter?

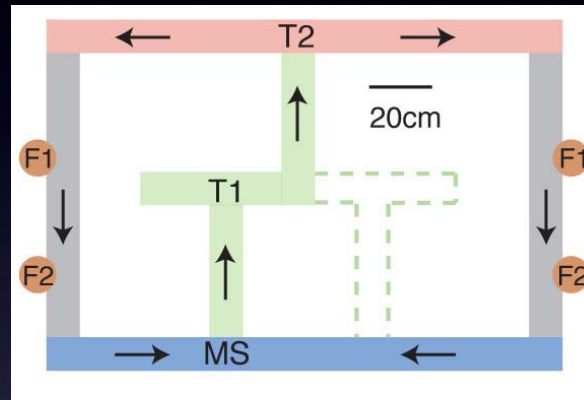
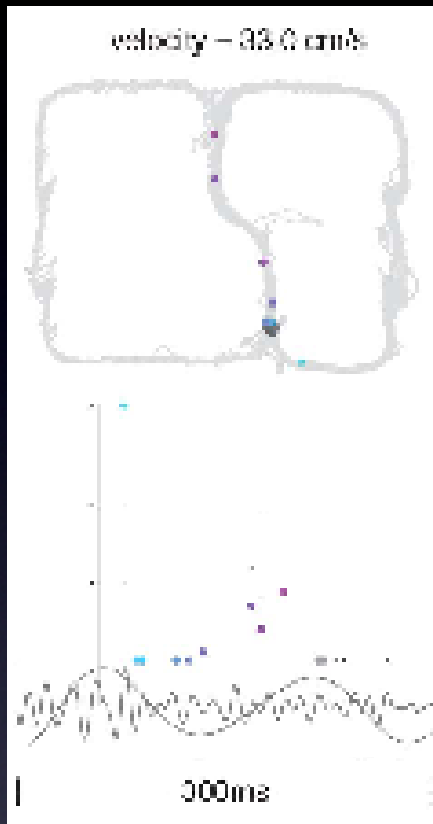
- Masters & Skaggs: remapping coincides with insight into a task:



- One rat quickly remapped & learned the task; one never did. One rat didn't remap until day 11, when it suddenly “got” the task. Cause or effect?

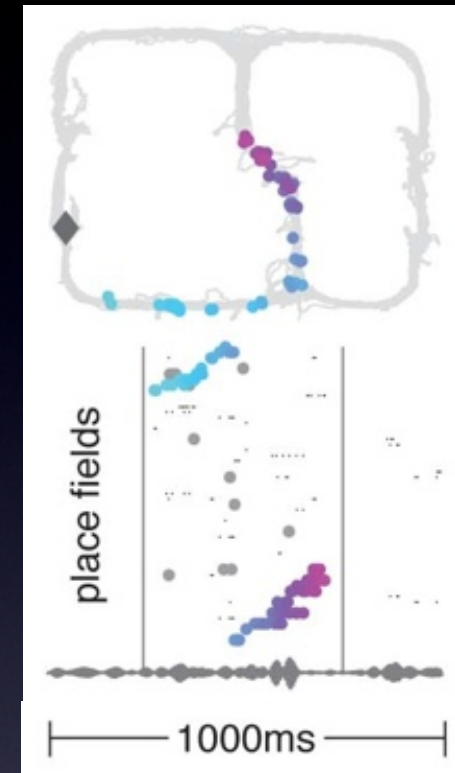
Theta vs Replay Sequences

Theta



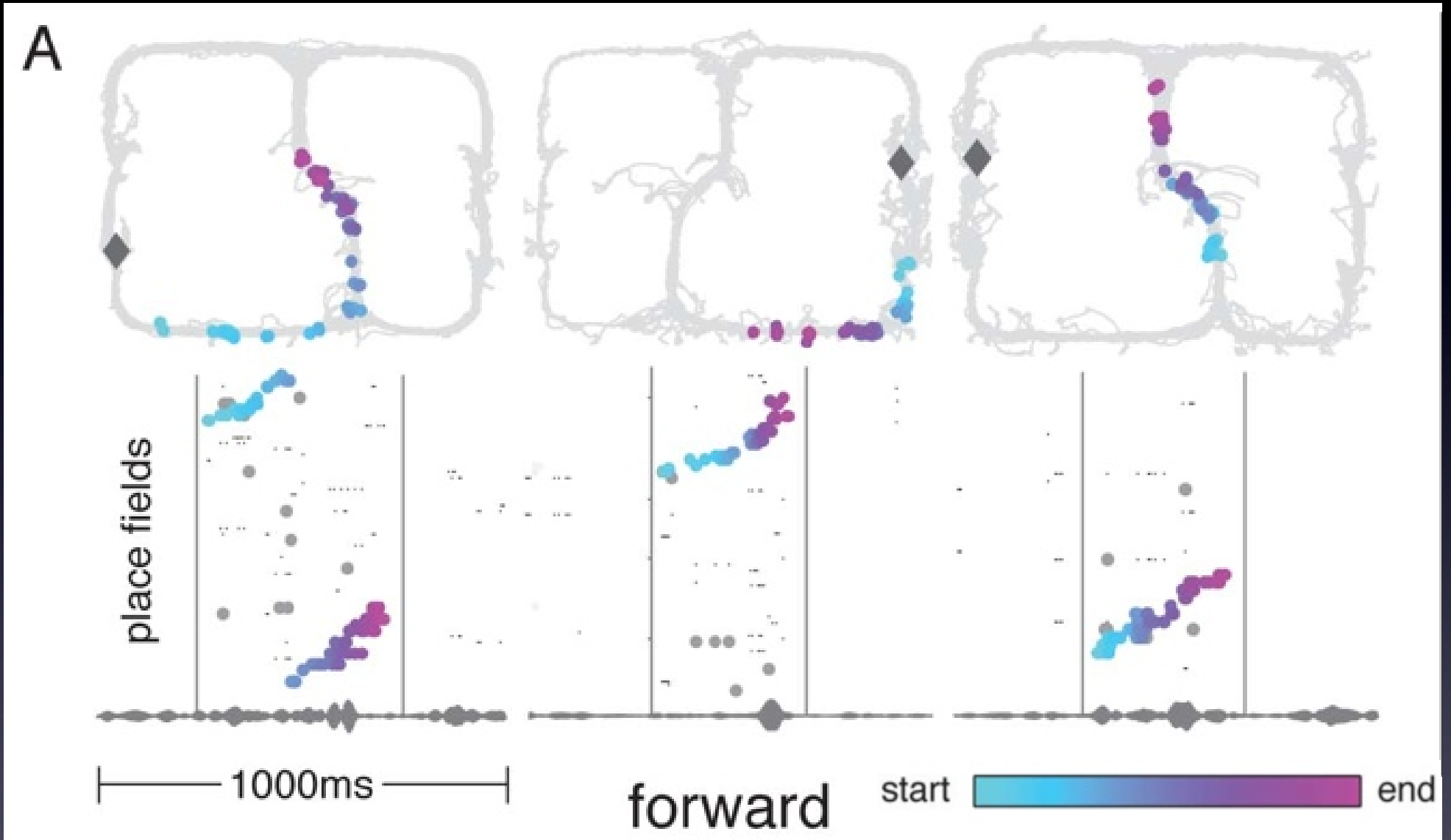
- Occur during attentive behavior
- Theta oscillation is present
- Tied to the animal's location
- Forward sequence
- Few neurons are active
- Relatively short paths represented
- Experience encoding and recall

Replay



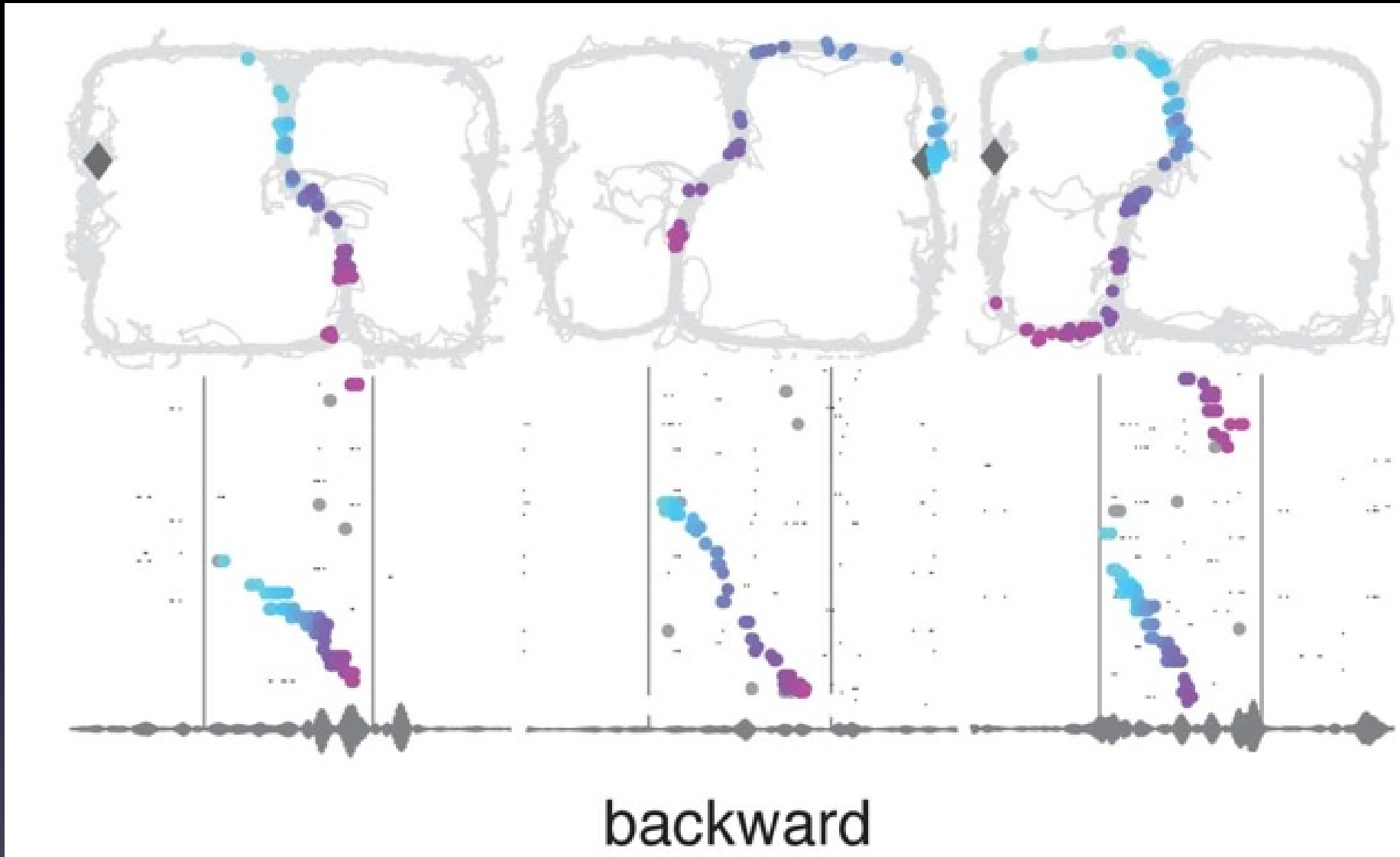
- Occur during awake rest
- Sharp wave ripples present
- Not always tied to the animal's location
- Forward or backward sequence
- Many neurons are often active
- Highly variable path lengths represented
- Memory consolidation, learning of cognitive maps

Forward Replay



Gupta, van der Meer, Touretzky, Redish, 2010

Backward Replay

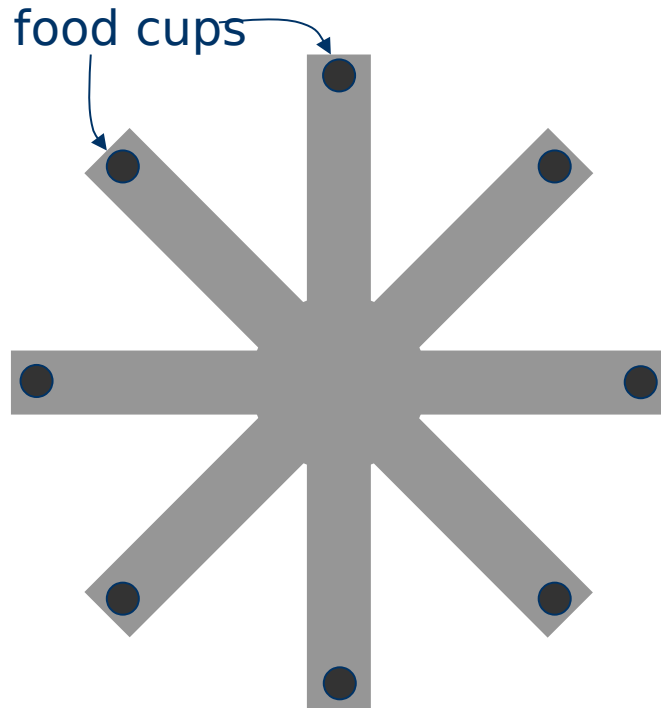


Gupta, van der Meer, Touretzky, Redish, 2010

Configural Learning

- Sutherland and Rudy suggested that hippocampus learns complex configurations of cues.
- After lesion, animals can still do tasks that depend on only one cue at a time.
- But tasks that depend on *relationships* among cues are impaired. Examples:
 - eight-arm radial maze
 - Morris water maze

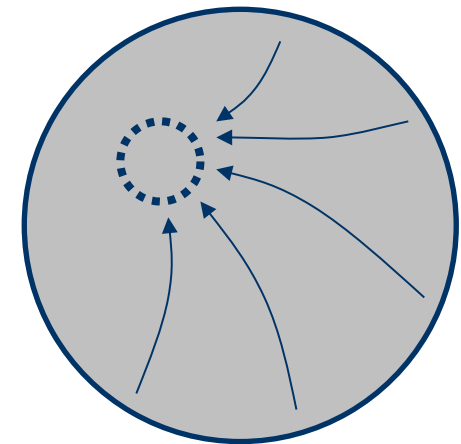
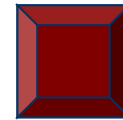
Spatial Working Memory



- Apparatus: 8-arm radial maze with food cups at each arm end
- All food cups are baited at the beginning of each trial
- During each trial, rats must remember which arms have already been visited. *A second arm visit provides no reward.*
- Rats with hippocampal lesions are severely impaired at this task (Neave et al., 1997)

Morris Water Maze

- Large pool filled with milky (opaque), cold water.
- A submerged platform is located at a fixed position in the pool.
- Distal landmarks outside the pool are located around the room; they never move.
- The rat is released from a random starting position and must swim to the platform to get out of the water.



Morris Water Maze

Sutherland and Rudy
(1988):

- Rats with fornix lesions can still navigate to a visible platform.
- But they are impaired at learning to find the hidden platform.
- Finding the hidden platform presumably requires recognizing a *configuration* of cues.



Morris Water Maze Revisited

- Rats with 48 training trials prior to lesioning the hippocampus showed no deficit (Morris *et al.*, 1990).

Hippocampal lesion causes a *learning* deficit!

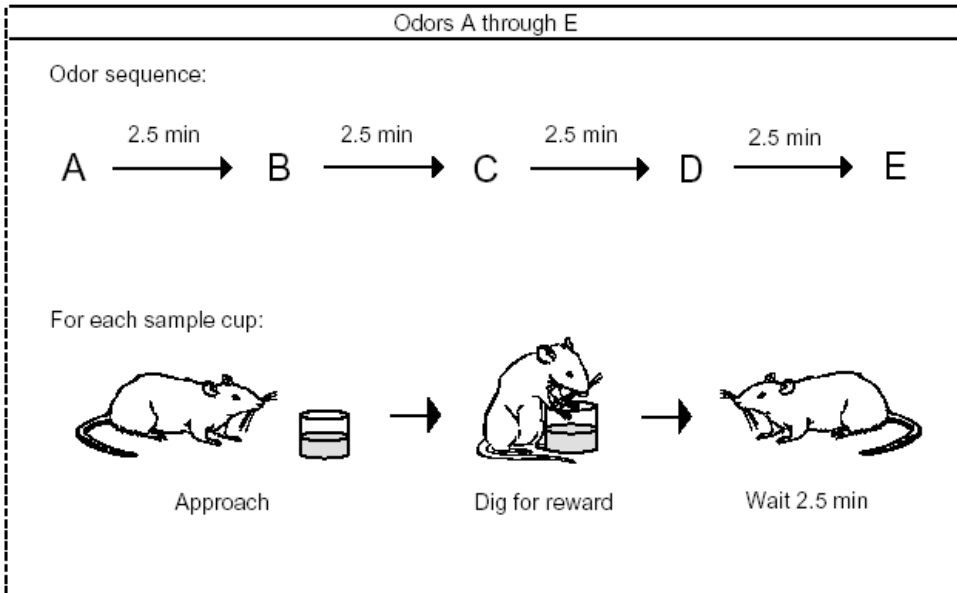
- Lesioned rats can gradually learn to find a hidden platform using successively smaller platforms (Schallert *et al.*, 1996):



**Hippocampal lesions cause impairment
only when learning the whole path at once!**

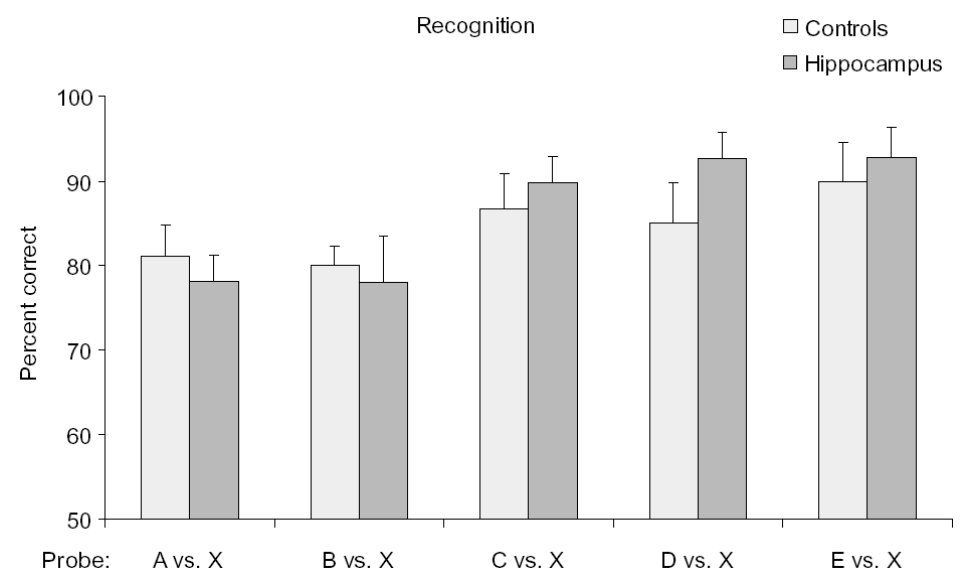
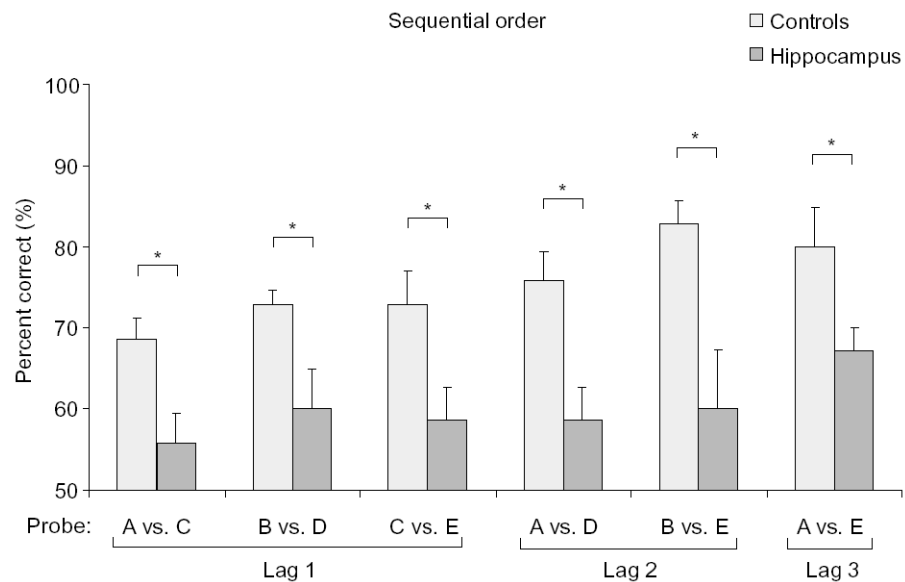
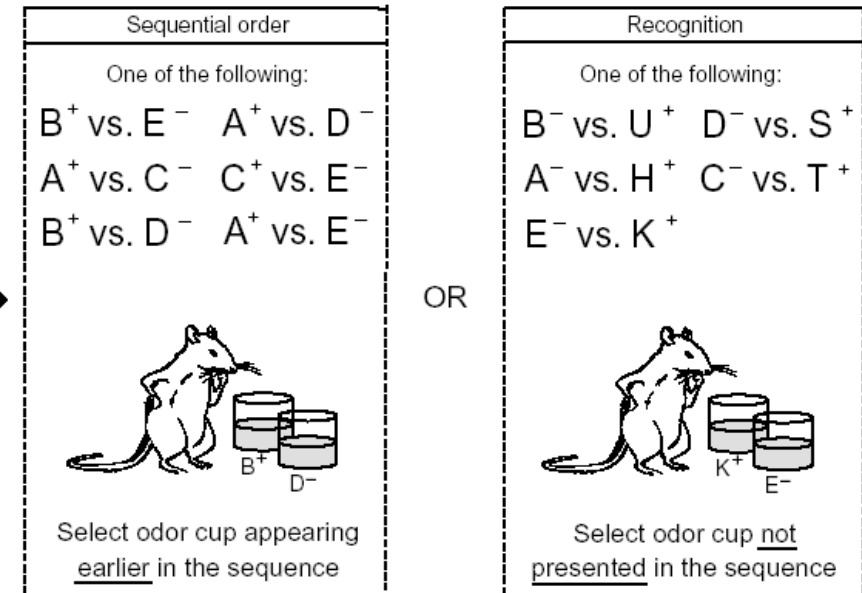
Sequence Learning

Sequence presentation



3 min

Probe



What Does the Hippocampus Do?

- Builds sparse random representations of complex configurations of sensory and behavioral information.
- Learns spatiotemporal associations between these, within appropriate context, e.g., for:
 - Learning paths to a goal
 - Learning odor sequences
- Retains representations for later use / consolidation.
 - Replay of paths during sleep
 - Recall of task state after delay:
 - DMS and DNMS tasks
 - Trace conditioning