# Memories of space

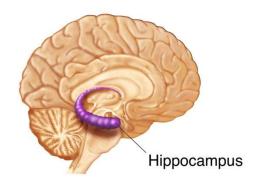
CS786 October 22<sup>nd</sup> 2024

#### A fortuitous discovery

- H.M. suffered from epilepsy from a young age
- Operated upon for treatment
- Post-op presented with a pure case of anterograde amnesia
- Demonstrated criticality of hippocampus for memory formation



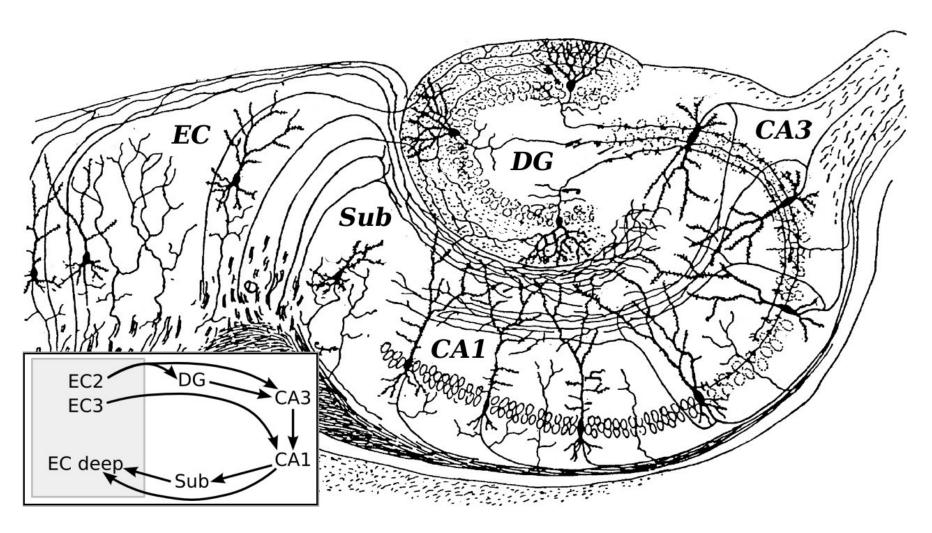
Lateral Brain Anatomy



© 2016 Medimagery - Laura Maaske LLC

https://www.newyorker.com/books/page-turner/the-man-who-forgot-everything

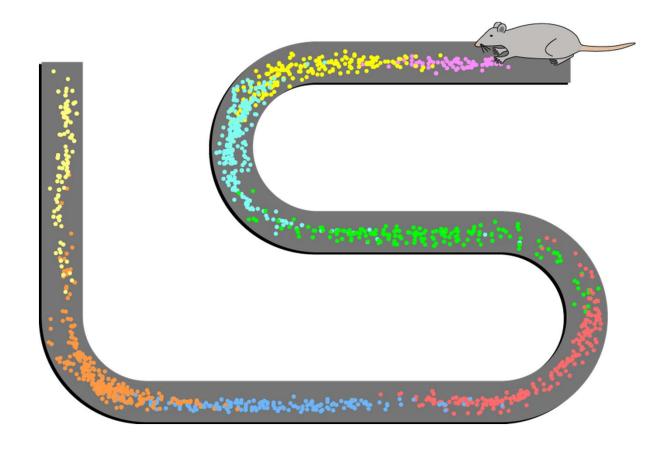
# Hippocampus



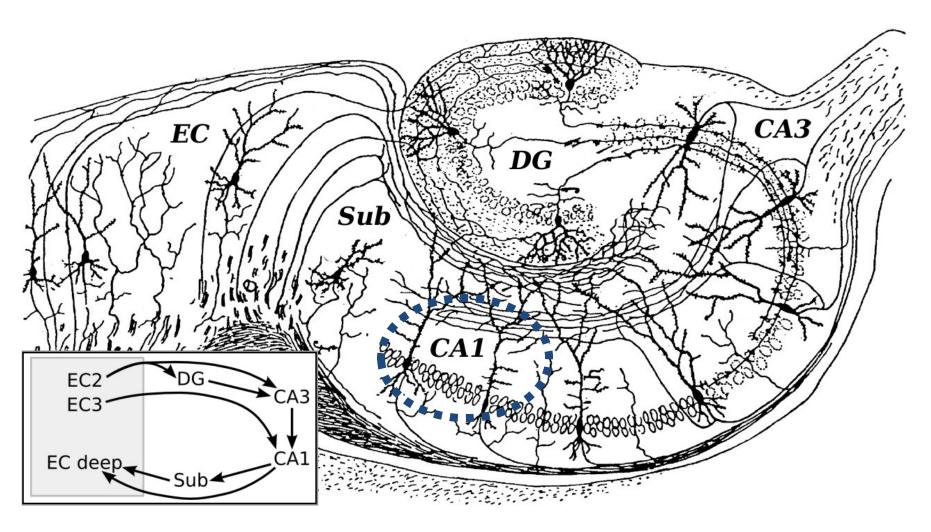
Phenomenally interesting anatomical structure; crucial for forming representations

#### Place cells

 Firing patterns of 8 place cells recorded from CA1 in a rat



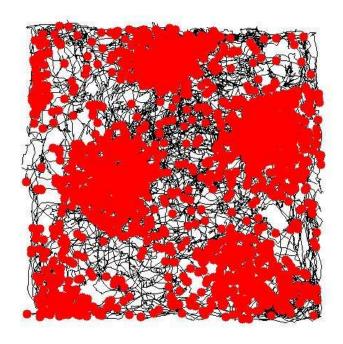
# Hippocampus



Phenomenally interesting anatomical structure; crucial for forming representations

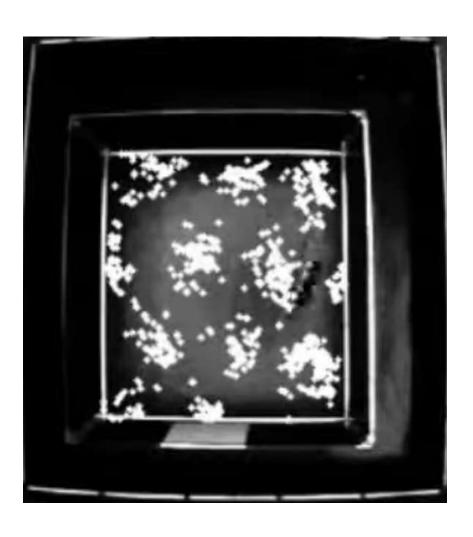
#### Grid cells

- Like place cells, but embed Euclidean space assumptions
- Encode spatial firing fields at equal distances from neighbors
- As if neurons are sensitive to an underlying triangular coordinate system



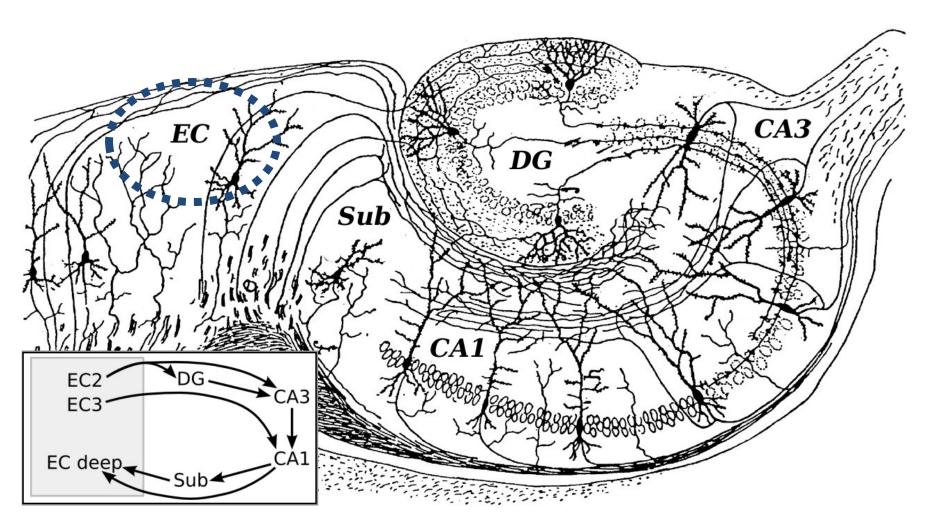
Red dots indicate location of rat in physical space when the grid cell fires

## Grid cells are amazing



- https://www.youtube.c om/watch?v=dgN5j16sc j4
- Reflective of intrinsic coordinate geometry embedded in our hippocampus

# Hippocampus

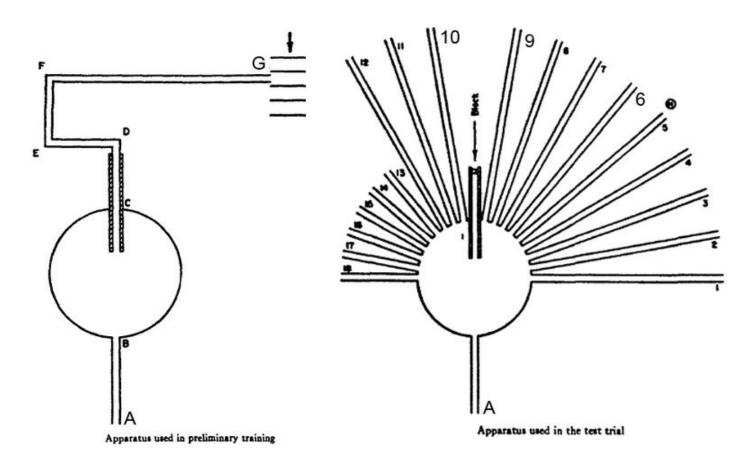


Phenomenally interesting anatomical structure; crucial for forming representations

#### Head direction cells

- Fire when animal's head turns in a particular direction
  - Tend to lead the actual head movement by about
    100 ms
- HD system interacts with place cells to generate spatial map of environment?
- Operates coherently during REM sleep

### Cognitive maps in rats and men



(From E. C. Tolman, B. F. Ritchie and D. Kalish, Studies in spatial learning. I. Orientation and short-cut. J. exp. Psychol., 1946, 36, p. 17.)

#### Revisiting an old debate

- Model free learning → learn stimulus-response mappings = habits
- What about goal-based decision-making?
  - Do animals not learn the physics of the world in making decisions?
- Model-based learning → learn what to do based on the way the world is currently set up = thoughtful responding?
- People have argued for two systems
  - Model-based systems are difficult to learn
  - Model-free systems don't generalize well

### The successor representation

Remember the value iteration equation?

$$V^{\pi}(s) = \sum_{a} \pi(a|s) [R(s,a) + \sum_{s'} P(s'|s,a) \gamma V^{\pi}(s')]$$

Peter Dayan showed long ago that it could be rewritten as

$$V^{\pi}(s) = \sum_{s'} M^{\pi}(s, s') \sum_{a} \pi(a|s') R(s', a)$$

Where

$$M^{\pi}(s, s') = \mathbb{E}\left[\sum_{t=0}^{\infty} \gamma^{t} \mathbb{I}(s_{t} = s') | S_{0} = s\right],$$

 The successor representation offers one <u>explanation</u> for how the TD signal could yield model-based policies

### The successor representation

The big problem with model-free RL

$$\delta_t = r_{t+1} + \gamma \, V^\pi(s_{t+1}) - V^\pi(s_t)$$
  $\Delta V^\pi(s_t) = lpha \, \delta_t$ 

- Changes in reward distributions entail propagate very slowly across action chains
- SR-based RL tries to fix this

$$egin{align} \delta_t^{ ext{SR}} &= \mathbb{I}(s_t = s') + \gamma \, M(s_{t+1}, s') - M(s_t, s') \ & \Delta M(s_t, s') = lpha \, \delta_t^{ ext{SR}} \ \end{aligned}$$

#### Benefits over model-free learning

- SR based learning uses a sensory prediction error to guide learning
  - Rather than look at the discrepancy between how rewarding I find a state, and how much I expected it to be
  - I look at whether I am arriving in the state more frequently than I expected, while trying to maximizing reward
- Changes in the reward distribution quickly update the state visitation matrix for all states from this state

#### Are humans SR RL learners?

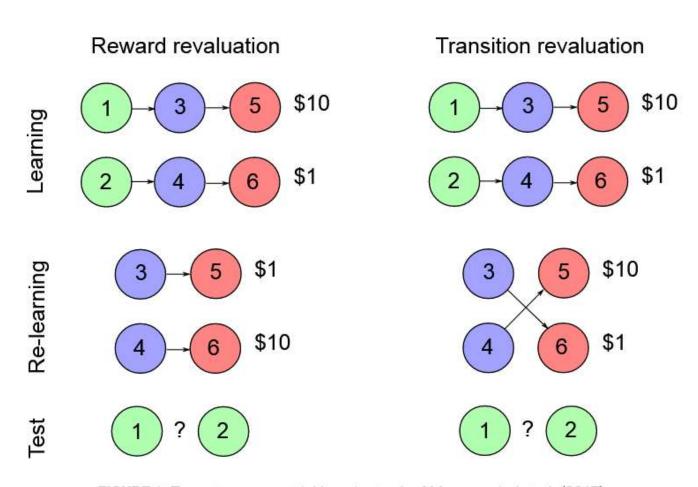


FIGURE 1: Two-steps sequential learning task of Momennejad et al. (2017).

#### What should happen?

- Model-free RL would predict no change after both reward and transition re-learning
- Model-based RL would predict a change in both conditions
- SR-based RL would predict a change in the reward relearning condition, but not in the transition relearning condition

## What actually happens

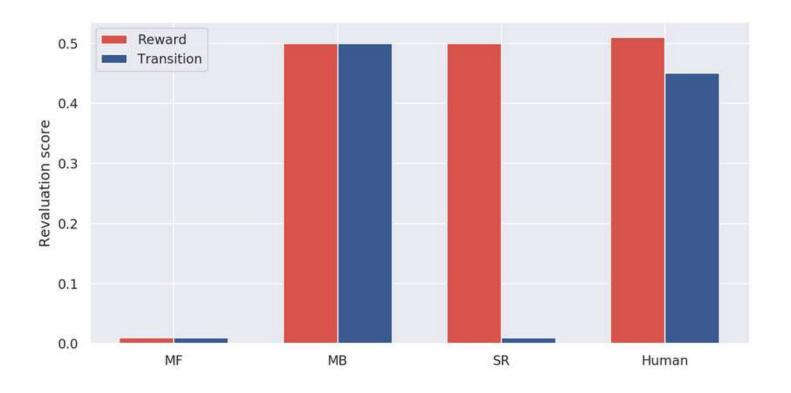
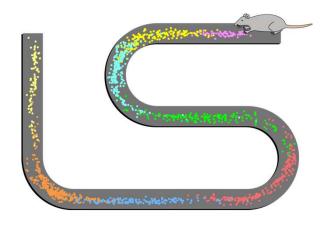


FIGURE 2: Revaluation score in the reward (red) and transition (blue) revaluation conditions for the model-free (MF), model-based (MB), successor representation (SR) and human data as reported in Momennejad et al. (2017).

Humans seem pretty model-based in this task, though SR seems to be mildly implicated

## A neuroscientific hypothesis

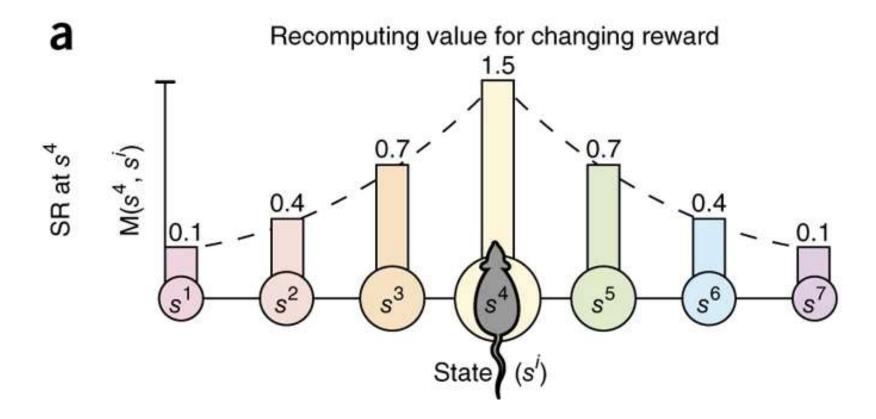


- Place cells have historically been seen to represent an animal's sense of place
- Stachenfeld et al (2017)
   propose that they
   encode the SR matrix
   for a task
- Representing a sense of where they are likely to go next

https://www.nature.com/articles/nn.4650

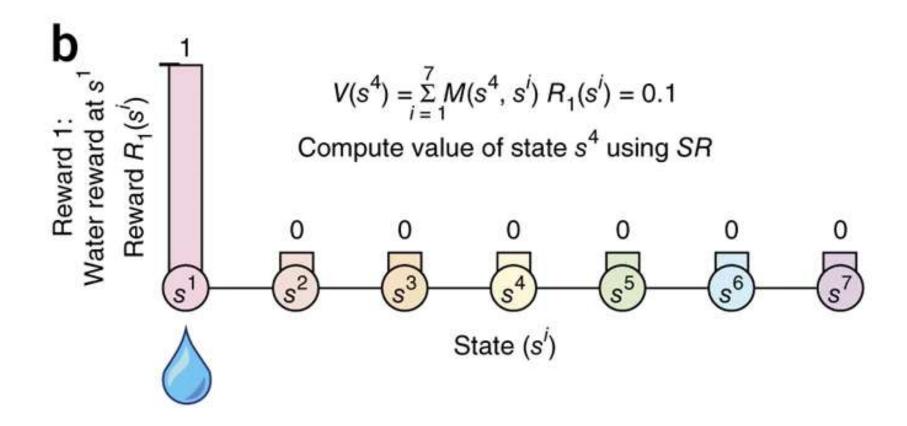
# The model

We assume that value is updated with the SR when reward changes



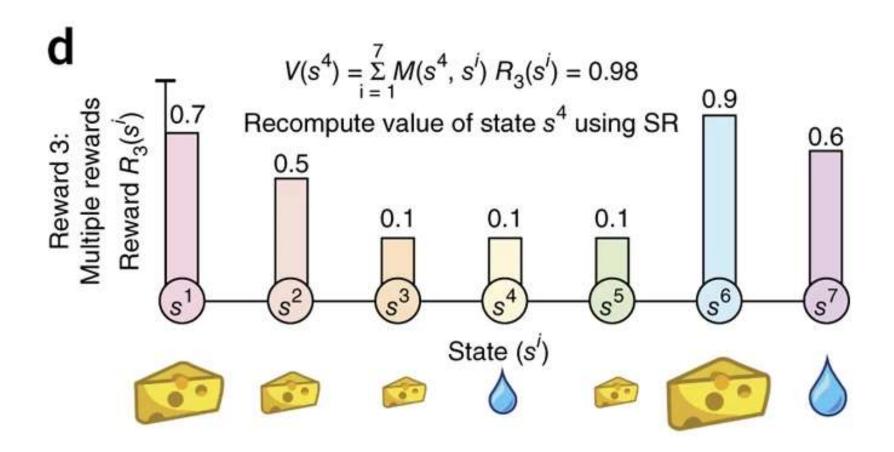
# The model

 Halfway house between model free and model-based learning

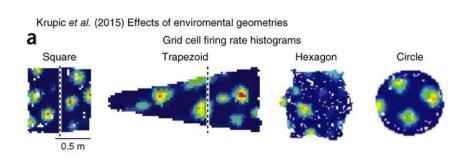


## The model

 Value functions can be recomputed without changing the SR



# Consistent with existing data on place cell behavior



- Model accommodates deviations from Euclidean geometry seen in actual rat experiments
- Also explains several nonspatial task results drawn from fMRI imaging
- A subtle hypothesis that is consistent with a primarily predictive role for hippocampus

