

Psychology 452
**Week 13: Spatial Representations
 In PDP Networks**

- Hippocampus As A Cognitive Map
- Networks Learn Metric Spaces
- Networks Learn Nonmetric Spaces

Course Trajectory

When	What
Weeks 1-3	Basics of three architectures (DAM, perceptron, MLP)
Weeks 4-6	Cognitive science of DAMs and perceptrons
Week 7	Music and networks
Weeks 8-10	Interpreting MLPs
Weeks 11-13	Case studies (interpretations, applications, architectures)

Psychology and Space

- Spatial behavior, or spatial reasoning, have long been studied by psychologists
- Some of the earliest cognitive proposals are found in Tolman's studies of spatial behavior
- Tolman introduced the term 'cognitive map' long before the cognitive revolution occurred



Edward Tolman

The Cognitive Map

Vol. 55, No. 4

JULY, 1948

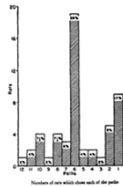
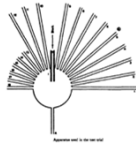
THE PSYCHOLOGICAL REVIEW

COGNITIVE MAPS IN RATS AND MEN
 BY EDWARD C. TOLMAN
University of California

Secondly, we assert that the central office itself is far more like a map control room than it is like an old-fashioned telephone exchange. The stimuli, which are allowed in, are not connected by just simple one-to-one switches to the outgoing responses. Rather, the incoming impulses are usually worked over and elaborated in the central control room into a tentative, cognitive-like map of the environment. And it is this tentative map, indicating routes and paths and environmental relationships, which finally determines what responses, if any, the animal will finally release.

Example Evidence

- Rats, when finding a route to a goal blocked, will find a related route that will take them towards the goal whose learned path is not available
- "As a result of their original training, the rats had, it would seem, acquired not merely a strip-map to the effect that the original specifically trained-on path led to food but, rather, a wider comprehensive map to the effect that food was located in such and such a direction in the room"



Critique

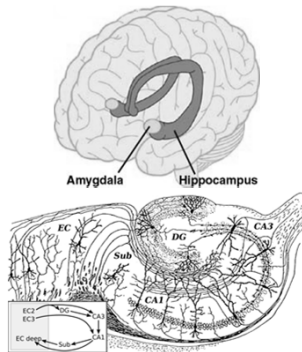
- Tolman's cognitivist musings were critiqued
- In a 1935 text on learning, Guthrie noted: "In his concern with what goes on in the rat's mind, Tolman has neglected to predict what the rat will do. So far as the theory is concerned the rat is left buried in thought; if he gets to the food-box at the end that is [the rat's] concern, not the concern of the theory" (p. 172)
- This is a variant of Ryle's regress being used to attack functionalism and cognitivism
- Needed: an architecture for the cognitive map!



Edwin Ray Guthrie

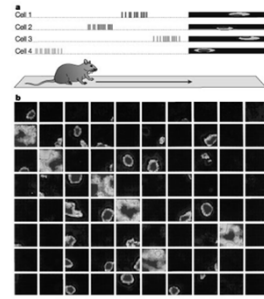
The Hippocampus

- The hippocampus is a component of the limbic system
- Earlier we saw that there are reasons to believe that it is a locus of Hebb-like learning
- There is also substantial evidence to suggest that it provides the cognitive map



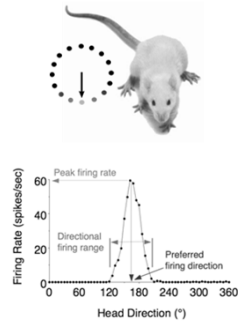
The Hippocampus And Place

- In the 1970s, single-cell recordings from the hippocampus revealed a biological basis for the 'cognitive map'
- Place cells fire when an animal's head is at a particular location in the world
- A variety of other spatial location cells have been found in the hippocampus since the discovery of place cells



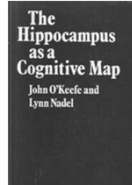
The Brain And Head Direction

- Further explorations of space in the brain discovered other interesting cells that encoded spatial information
- For example, head direction cells spike when an animal's head is pointed in a specific direction



The Hippocampus As A Cognitive Map

- 1978 saw the publication of a seminal book detailing the results of such work
- 2014 saw John O'Keefe, May-Britt Moser and Edvard Moser win the Nobel Prize in Physiology or Medicine for this work



Studying Spatial Learning

- Why is the map in the hippocampus? This permits spatial learning to occur
- Morris (1984) invented the water maze to study spatial learning of rats
- Rats swim to escape the water, and discover a hidden platform
- On repeated trials, the time taken to find the platform decreases, indicating that the agents are learning its position in the maze

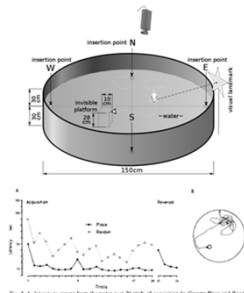
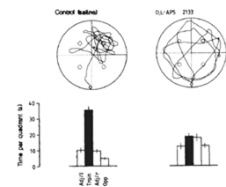


Fig. 4. A. latency to escape from the water over 20 trials of acquisition by Group Platform and Random. The rapid escape by rats of Group Platform cannot be due to head cues from the platform, sufficient water in Group Platform was therefore provided. Note increased latency on trial 21. In particular on trial 21 the rat searched for better platform location in NW prior to finding platform near to SW.

Blocking Spatial Learning

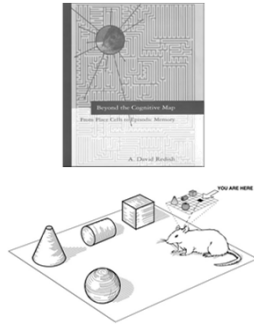
- Morris et al. (1986) studied the role of LTP in spatial learning
- Rats were placed in the Morris water maze
- In the experimental condition, animals were treated with D, L-AP5, known to block NMDA receptors in the hippocampus
- Rats treated in this way were unable to learn the position of the platform in the maze



Paths taken in the maze by control animals on the left and by experimental animals on the right. Lack of preference for a particular quadrant indicates experimental animals did not learn the position of the platform

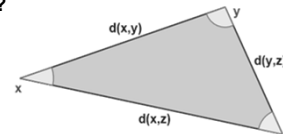
Critiquing The Map

- Some of the ideas about the hippocampus as a cognitive map have been challenged via results from neuroscience
- It is not topographically organized
- It is at best locally metric
- Let us consider whether PDP networks can contribute anything to this discussion



A Metric Space

- A metric space has three general properties true of the relations between points:
 - Minimality**
 - Distance from a point to itself is a minimum
 - Symmetry**
 - $d(x,y) = d(y,x)$
 - Triangle Inequality**
 - $d(x,y) + d(y,z) \geq d(x,z)$
- If a network learned such a space, how would it represent it?



Alberta As A Metric Space

- Dawson, Boechler and Valsangar-Smyth (2000) trained a network to estimate the distances between cities in Alberta
- If it can learn such a task, then it has – by definition – internalized a metric space



Distance Table

- The training set was created by using a table of distances between cities to determine output ratings

Table 1. All distances between cities of Alberta, measured in kilometers. (0) The distance from Table 1a converted into ratings on a 10 scale

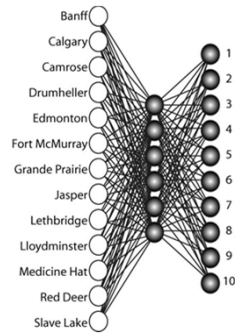
CITY	CALGARY	EDMONTON	REGINA	SASKATOON	LETHBRIDGE	RED DEAR	BLACK HILLS
EDMONTON	0	124	301	451	662	727	828
CALGARY	124	0	274	396	566	621	722
REGINA	301	274	0	181	351	406	507
SASKATOON	451	396	181	0	271	326	427
LETHBRIDGE	662	566	351	271	0	51	152
RED DEAR	727	621	406	326	51	0	102
BLACK HILLS	828	722	507	427	152	102	0

Table 2. Continued

CITY	CALGARY	EDMONTON	REGINA	SASKATOON	LETHBRIDGE	RED DEAR	BLACK HILLS
EDMONTON	0	2	4	7	11	12	14
CALGARY	2	0	3	7	11	12	14
REGINA	4	3	0	5	8	9	11
SASKATOON	7	7	5	0	6	7	9
LETHBRIDGE	11	11	8	6	0	2	5
RED DEAR	12	12	9	7	2	0	3
BLACK HILLS	14	14	11	9	5	3	0

The Alberta Network

- Network of value units
- Local coding of input cities and output ratings
- Smallest network that worked used 6 hidden units
- 169 training patterns



Projected 1D Maps

- Network interpretation was not straightforward
- It was eventually realized that each hidden unit was a one-dimensional map
- A map had a particular orientation
- Connection weights were correlated with distances projected onto this map!

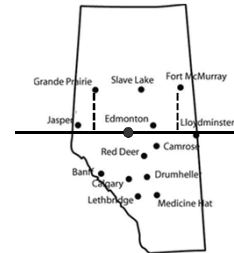
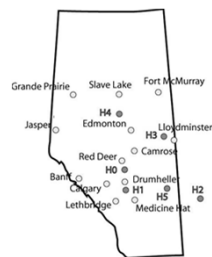


Table 2. Results of relating Alberta map distances between cities and hidden units (the values of the connection weights leading into the hidden units in Sumstat-1)

Hidden Unit	Latitude	Longitude	Correlation Between Map Distance and Learning Weights
H0	51.72	113.55	0.88
H1	50.84	113.63	0.90
H2	50.88	113.79	0.72
H3	51.30	113.68	-0.54
H4	51.01	113.42	0.79
H5	51.17	113.57	-0.68

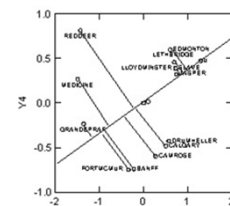
Positioning Hidden Units

- Excel Solver was used to place each hidden unit on the map
- A hidden unit is the origin of a 1D map
- Solver also oriented the hidden unit's map
- Solver maximized correlation between weight and distance
- High correlations indicates that this is a plausible account of hidden unit function



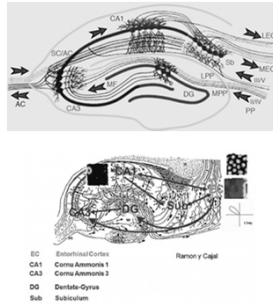
Hidden Unit Properties

- Hidden units were metric
- Individual hidden units, though, had a very inaccurate internal map
- Accuracy of space came from coarse allocentric coding!



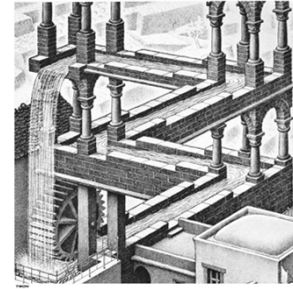
Place Cells And Coarse Allocentric Coding

- Place cells have been criticized as not being 'map-like'
- Our network is not map-like either, but has internalized a map of Alberta
- Hidden units are like place cells
- Perhaps the hippocampus is a PDP map, using coarse allocentric coding – which acts like a map, but doesn't look like one!



Advantages Of Non-Map-Like Maps

- Why might it be advantageous to have a cognitive map that is not map-like in structure?
- Some spatial reasoning may involve spaces that are not metric
- Map-like representations might fail on such tasks



Antisymmetry And Direction

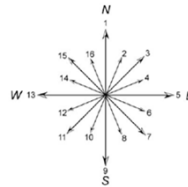
- One can violate metric space property by using antisymmetric relations instead of symmetric ones
- $d(x,y) = -d(y,x)$
- Compass directions are antisymmetric
- The direction from Edmonton to Calgary is the opposite of the direction from Calgary to Edmonton

$$A = \begin{bmatrix} 0 & -c & b \\ c & 0 & -a \\ -b & a & 0 \end{bmatrix}$$



A Direction Task

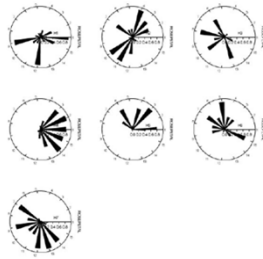
- Dawson and Boehler (2007) trained a network to judge the direction between pairs of Albertan cities
- This task is intrinsically nonmetric



	EDMONTON	CALGARY	WINNIPEG	REGINA	SASKATOON	VANCOUVER	OTTAWA	ST. CATHARINES	WINDSOR	EDMONTON	WINNIPEG	REGINA	SASKATOON	VANCOUVER	OTTAWA	ST. CATHARINES	WINDSOR
EDMONTON	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
CALGARY	1	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
WINNIPEG	2	1	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
REGINA	3	2	1	0	1	2	3	4	5	6	7	8	9	10	11	12	13
SASKATOON	4	3	2	1	0	1	2	3	4	5	6	7	8	9	10	11	12
VANCOUVER	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	10	11
OTTAWA	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9	10
ST. CATHARINES	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8	9
WINDSOR	8	7	6	5	4	3	2	1	0	1	2	3	4	5	6	7	8

Coarse Coding

- A seven hidden unit network of value units learned the task
- Its hidden units coarse coded directions
- Not a topographic map of 'head direction'!



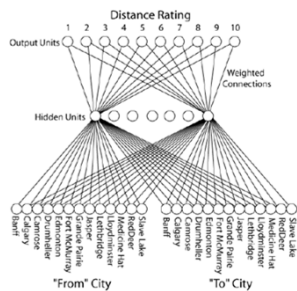
Asymmetry Of Wiretaps

- Dawson and Boechler wiretapped each hidden unit for each pair of input stimuli
- The resulting matrices of activity were highly asymmetric

Hidden Unit	Proportion Asymmetry Of Activation Matrix	Proportion Asymmetry Of Net Input Matrix	Correlation Between "From" Weights and "To" Weights
H11	0.47	0.63	-0.27
H12	0.36	0.36	0.28
H13	0.51	0.49	0.03
H14	0.92	0.95	-0.91
H15	0.72	0.86	-0.76
H16	0.45	0.50	-0.01
H17	0.81	0.92	-0.86

Another Distance Network

- Dawson, Boechler and Orsten (2005) trained a different value unit network on the city distance problem



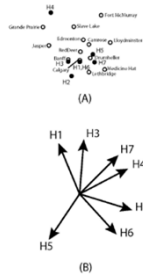
Weights Do Not Code Distance!

- The weights feeding into each hidden unit were highly systematic
- But weights could not be related to distances between cities!
- What systematic feature were hidden units detecting?



Hidden Units Encode Direction

- Hidden unit weights encoded **direction!**
- When placed by Solver on the map, weights were correlated with direction from the origin and plane of the hidden unit's 1D map



Hidden Sextants

- Each hidden unit provided distance information by acting like a sextant

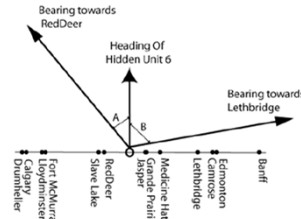


Figure 4. A graphical interpretation of the operation of hidden units in the network, using Hidden Unit 6 as an example. The horizontal line represents the positions of the thirteen cities on the unit's weight space. The circle represents the origin of this space. The unit's heading is given from this origin. Angle A is the bearing towards RedDeer relative to this heading, and Angle B is the bearing towards Lethbridge. The position of a city on the weight space is highly correlated with the cosine of the city's bearing.

Coarse Coding Again

- As a sextant, each hidden unit is a highly inaccurate measure of distance
- Solution: coarse coding
- Seven sextant readings from different perspectives combine to measure distance accurately

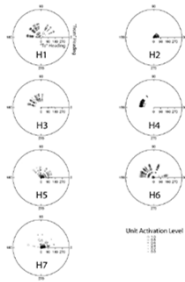


Figure 5. Bubble plots of the activity of each hidden unit as a function of the heading of the origin to from city and to city for each unit of cities in the training set. The size of the bubble reflects the degree of hidden unit activity. The coordinates of the bubble indicate the bearing of the two cities. The circular axis provides the bearing of the from city, and the horizontal axis provides the bearing of the to city.

Functional Maps

- "The cognitive map is not a picture or image which 'looks like' what it represents; rather, it is an information structure from which map-like images can be reconstructed and from which behavior dependent upon place information can be generated" (O'Keefe & Nadel, 1978)
- Our network simulations provide several illustrations of this point: three very different representations that are not map-like, but which function like maps on spatial reasoning tasks of known structure



John O'Keefe



Lynn Nadel