

1 **Spatial representation in the hippocampal formation: a history**

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12 Main text (without Box, Figure Legends, and References: 10,818 words

13 8 figures (including timeline), 1 Box

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15 Acknowledgments. The work was supported by two Advanced Investigator Grants from the European
16 Research Council (GRIDCODE – grant no. 338865 to E.I.M.; ‘ENSEMBLE’, Grant Agreement no.
17 268598, to M.-B.M.), the Centre of Excellence scheme of the Research Council of Norway (Centre for
18 Neural Computation, grant number 223262 to M.-B.M. and E.I.M.), the Kavli Foundation (M.-B.M.
19 and E.I.M.), and National Science Foundation Grant 1631465 to B.L.M.

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23 ABSTRACT

24 Since the first place cell was recorded, and the subsequent formulation of the cognitive-map theory,
25 investigation of spatial representation in the hippocampal formation has evolved in stages. Early
26 studies sought to verify the spatial nature of place-cell activity and determine its sensory origin. A
27 new epoch started with the discovery of head direction cells and the realization of the importance of
28 angular and linear movement-integration in the generation of spatial maps. A third epoch began
29 when investigators turned their attention to the entorhinal cortex, which led to the discovery of grid
30 cells and border cells. This review will show how ideas about integration of self-motion cues have
31 shaped our understanding of spatial representation in hippocampal-entorhinal systems, from the
32 1970s until today. How specialized cell types of these systems work together is now investigable and
33 spatial mapping may become one of the first cognitive functions to be understood in mechanistic
34 detail.

35

36

37 Although the study of the cellular and circuit mechanisms of spatial representation in the brain today
38 is centered on the hippocampal and parahippocampal formation, the history of the study of spatial
39 coding did not begin there, but rather with the parietal cortex, in the form of early observations on
40 patients with parietal damage^{1,2}, and in many respects one takes a risk in attempting to limit the
41 discussion to the hippocampal formation³. Nevertheless, in studies of spatial coding, some of the
42 most 'paradigm shifting' discoveries and ideas have come from recordings within the greater network
43 of the hippocampal formation, particularly the dorsal parts of hippocampus, entorhinal cortex and
44 pre- and parasubiculum, where cells exhibit place-dependent activity independently of the animal's
45 behavior or the task that it is performing (Fig. 1). Key among these insights were the discoveries of
46 place cells (Fig. 2)⁴, head-direction cells (Fig. 3)⁵⁻⁷, and grid cells^{8,9}, each of which represent quantum
47 jumps in our realization that there is a system in the brain which has evolved to produce a
48 representation manifold that can be linked to position (grid cells), an inertial compass (head direction
49 cells), and a system for mapping external features and events onto internal, and, at least locally,
50 metric coordinates (place cells). In broad terms, these components, and their interactions, were
51 predicted by O'Keefe in 1976¹⁰.

52 Also key to the emergence of a paradigm for spatial representation was a gradual understanding of
53 the role played by different spatial reference frames and their interactions. There exist three
54 reference frames in which space can be represented: egocentric (defined in relation to a body part
55 axis), allocentric (based on spatial relationships to or among external features), and inertial or
56 idiothetic (relative location and orientation based on direction and distance moved from an arbitrary
57 reference point). Navigation in an idiothetic reference frame is often referred to as 'path integration',
58 a process by which animals use self-motion cues – such as motor efference, optical flow, and
59 vestibular information – to keep track of their own location relative to a starting point¹¹⁻¹⁴. Decades
60 of investigation have shown that egocentric space is not represented primarily in the hippocampal
61 formation but rather in parietal cortex and associated regions¹⁵⁻¹⁷. O'Keefe's studies showed from the
62 outset that instead place cells encode an animal's location in an orientation-independent reference
63 frame¹⁰. Although the term allocentric was applied to place-cell representations, O'Keefe recognized
64 early on that these representations may rely "on the fact that information about changes in position
65 and direction in space could be calculated from the animal's movements"¹⁰. Yet, it was not until the
66 discovery of head direction cells in the 1980s⁵⁻⁷, and the realization that they were indeed performing
67 integration of head angular velocity¹⁸, that the concept emerged, in the 1990s, that the entire
68 hippocampal formation might be using an idiothetic reference frame – or path integration – as a
69 basis for the establishment of its coordinate system¹⁹. The possibility of a path integration
70 mechanism outside the hippocampus proper^{3,20,21} was reinforced at this time by studies showing

71 that, unlike place cells, spatially modulated cells in the entorhinal cortex and subiculum had
72 environment-independent spatial firing patterns^{22,23}. Today it is generally recognized that path
73 integration plays a fundamental role in spatial coding in the hippocampal formation, although there
74 continues to be controversy as to whether path integration is the primary determinant of place and
75 grid cell firing or plays an equal or subordinate role to the integration of information from external
76 stimuli²⁴⁻²⁶.

77

78 Finally, a discussion of paradigm shifts would not be complete without some realization of the role
79 that technology has played (Fig. 4). Key technical advances have been the shift from recording single
80 cells in restrained, usually anesthetized, animals to freely behaving ones^{4,27-29}; the development of
81 quantitative video tracking methods for rodents during hippocampal recording experiments^{30,31}; the
82 invention of stereo (tetrode) recording³² (Fig. 4a), and its extension to large neuronal ensembles³³
83 (Fig. 4b-d); the development of micro-machined silicon electrode arrays³⁴; new cell-type specific
84 optical and chemical methods for stimulation³⁵⁻³⁷; and, most recently, the development of large-scale
85 Ca²⁺ cellular imaging in both freely moving animals³⁸, and restrained animals locomoting in 'virtual-
86 reality' environments^{39,40}. The importance of recording from substantial numbers of cells in
87 interpreting coding dynamics for the hippocampus or any other neural system cannot be
88 overemphasized. Apart from the obvious computational and statistical analysis power enabled by
89 data from large numbers of simultaneously active neurons, it is clear that many results that we now
90 understand as across-trial variations in population dynamics may have been attributed to differences
91 in single neuron classes in early single neuron recording studies.

92

93 We have taken on the task of trying to present, in a relatively small 'space', an historical overview of
94 some of the paradigm-shifting developments that led to our current understanding of spatial coding
95 in the hippocampal formation. This task is daunting for several reasons, not the least of which is that
96 the number of important experimental and theoretical contributions has risen (and continues to rise)
97 almost exponentially since 1971, when O'Keefe and Dostrovsky, after recording in freely behaving
98 rats from what today would be considered a very small sample of CA1 units, made the bold claim
99 that the hippocampus might construct a spatial map⁴ (Fig. 2). Length restrictions have forced us to
100 focus the review on one particular set of ideas that has inspired the investigation of hippocampal
101 representations of space almost since the beginning of studies of place cells, namely that spatially
102 localized firing to a large extent reflects the dynamic integration of self-motion – or path integration
103 – as animals move around in the environment. We shall demonstrate how the idea of a path-
104 integration input explained many fundamental properties of place cells, and how this in turn led

105 investigators in the single-cell recording field to identify a path integration-dependent neural system
106 consisting of multiple functionally specialized cell types in the parahippocampal cortices.

107

108 We shall demonstrate that path integration appears as a leitmotiv that follows the history of spatial
109 representation in the hippocampal formation across generations of investigators. Yet, by directing
110 our spotlight to path integration, we are forced to leave out contributions and research directions
111 that have contributed critically to the broader understanding of place cells and hippocampal-systems
112 function, beyond the representation of self-location. First of all, the more than four decades of
113 hippocampal spatial mapping studies have developed alongside an equally productive line of
114 investigations, using a variety of methodological approaches, into the basis of memory in the same
115 brain system^{3,41-47}. The focus of this review is on the encoding of space, but as we will acknowledge,
116 this does not rule out a broader participation of hippocampal neurons, and place cells, in
117 representation of experience⁴⁸⁻⁵⁰. In shying away from the memory functions of the hippocampus,
118 we shall also pass over the vast and growing literature on how replay and preplay of firing sequences
119 may enable consolidation and storage of hippocampal memory through interactions with neocortical
120 neural networks⁵¹⁻⁵⁴, and we shall not discuss the important but separate question of whether or
121 how place cells are used for goal-directed navigation and route planning⁵⁵⁻⁵⁹. We have also left out
122 dozens of pioneering studies of temporal coding and network oscillations, including theta rhythms,
123 that have shaped our current understanding of hippocampal function beyond the representation of
124 space^{49,60-62}. Finally, the review is dominated by work in rats and mice, reflecting the use of freely-
125 moving rodents as subjects in nearly all studies of spatially modulated cells in the hippocampal
126 formation (see Box 1 for extensions to the primate brain).

127

128 **The Origin of the Spatial Signal**

129 In 1971, O'Keefe and Dostrovsky observed that neurons in the rat hippocampus had what appeared
130 to be "spatial" receptive fields⁴ (Fig. 2ab). In the 1971 paper, the number of place cells, and the
131 evidence for localized firing, was limited, but a lot more substantial data were presented by O'Keefe
132 in 1976¹⁰. By this time, after thorough study of hippocampal activity in unrestrained rats²⁹, Ranck had
133 also seen place cells⁶³. The O'Keefe paper showed that place cells fired whenever the rat was in a
134 certain location in the local environment. Different cells had different place fields, such that, at all
135 locations investigated in the hippocampus, the animal's location could, in principle, be inferred from
136 the joint activity of a fairly small sample of neurons¹⁰ (for direct demonstration, see ref. 33 and Fig.
137 4cd). Based on this observation, and inspired by Tolman's proposal that navigation is guided by
138 internal "cognitive maps"⁶⁴, O'Keefe and Nadel⁶⁵ suggested that place cells are the basic element of a

139 distributed allocentric cognitive map of the animal's environment (Fig. 2c). The spatial relations
140 between landmarks provided by this map was thought to enable animals to find their way
141 independently of local view or movement trajectories, using what O'Keefe and Nadel called a locale
142 strategy. This contrasted with route strategies, which do not take into account the relationship
143 between landmarks. The latter strategies included a spectrum of routines from simple beacon
144 navigation to more complex action sequences. O'Keefe and Nadel's proposal represented a major
145 landmark in the conceptualization of hippocampal function. Their book – "The Hippocampus as a
146 Cognitive Map" – synthesized and re-interpreted decades of discordant experimental studies using a
147 range of experimental approaches, particularly lesions, and put these studies into a coherent
148 theoretical framework organized around the concept of place cells as the cellular basis for
149 representation of space as well as events and experiences associated with space. The book proposed
150 a neural implementation of Tolman's concept of the cognitive map, with visionary perspectives on
151 how such a map might enable a breadth of cognitive functions in higher species including humans.
152 Today, 40 years after its publication, "The Hippocampus as a Cognitive Map" remains the theoretical
153 pillar on which nearly all subsequent study of spatial coding in the hippocampal formation rests.

154

155 The early years of research on place cells, in the late 1970s and 1980s, were dominated by attempts
156 to prove that the place signal was indeed 'spatial' and, given this, to understand what caused place
157 cells to fire where they did, based on the idea that it was some constellation of external sensory
158 cues, rather than a single cue or some other cause (e.g., ref. 66). Two salient observations in this
159 period that both advanced knowledge and increased perplexity were the findings that place cells
160 appeared to be completely direction dependent when animals ran repeatedly on restricted paths³⁰
161 but were unaffected by head direction during free foraging in a large cylinder⁶⁷. Perplexity about the
162 mechanism of place cells was further increased by the fact that place cells had a sort of 'memory';
163 they rotated their fields when external cues were rotated, but continued to fire in relation to the
164 last-seen cue location when the cues were removed^{68,69}. Indeed, early studies indicated that place
165 cells not only continued to fire in the 'correct' location in total darkness; fields could also be formed
166 when animals were introduced to an environment in darkness and they were minimally affected
167 when the lights were subsequently turned on⁷⁰. Nevertheless, place fields became linked to external
168 cues and rotated to maintain registration with them when they were rotated between sessions^{68,71}.

169

170 The foregoing studies were soon followed by a number of observations that cast further doubt on
171 the external sensory origin of place fields: most place fields had asymmetric firing fields in an
172 environment with a symmetric cue configuration⁷²; place fields could dynamically shift between a
173 reference frame defined by a reward box that moved relative to the laboratory reference frame and

174 the lab reference frame itself^{73,74}; the location and orientation of place fields followed the rat when
175 the rat was rotated independently of the environment^{75,76}; place cells and head direction cells
176 exhibited coordinated drift error in a cylindrical environment^{77,78}; the size of place fields was almost
177 completely independent of local cue density, spatial frequency or salience⁷⁹ but varied systematically
178 along the septo-temporal axis of the hippocampus^{80,81}; in rats with age-related memory impairment⁸²
179 or with NMDA receptors blocked⁸³, place fields appeared perfectly normal in a novel environment,
180 but could be completely rearranged when the animals were returned to the same environment after
181 even a short delay; the place field map as a whole dynamically expanded when motor and vestibular
182 information about movement speed was disrupted, in the absence of changes in landmark inputs⁸⁴;
183 place cells shut off completely when animals were restrained from locomotion⁸⁵; and, finally, the
184 variation in scale of place fields along the hippocampal septo-temporal axis was strongly correlated
185 with the gain of physiological speed signals⁸⁶.

186

187 In spite of gradually accumulating evidence for an in many ways 'non-sensory' origin of spatial
188 receptive fields in the hippocampus, the lack of proper quantification prevented a general
189 acceptance of this idea and much of the initial effort was thus spent on proving that the signal was
190 indeed spatial. As this skepticism was gradually overcome, investigators began to focus on how place
191 cells might be synthesized as high-order integrators of sensory data, perhaps endowed with memory
192 properties. This sensory-integration approach changed, literally overnight, however, when James
193 Ranck brought a video of a recorded head direction cell to the 1984 SfN meeting⁸⁷ (Fig. 3). Head
194 direction cells are cells that fire specifically when the animal faces a certain direction⁵⁻⁷ (Fig. 3ab).
195 Ranck first encountered these cells in the dorsal presubiculum – almost by accident in an experiment
196 where electrodes targeted to the subiculum went astray⁸⁷ – but they were later observed across a
197 wide network of cortical and subcortical regions^{88,89}. In the same way that place cells covered all
198 locations of an environment, the preferred firing directions of head direction cells were distributed
199 evenly around angular space, enabling precise read-out of head direction in neural networks
200 downstream of head direction cells. If the brain was endowed so clearly with an internal compass, as
201 suggested by Ranck's 1984 movie, the idea that it also had a map became much more palatable.
202 However, the first full publication on the basic properties of head direction cells did not appear until
203 1990, in joint work by Ranck, Taube, and Muller^{6,7}. At that time, it was already recognized that the
204 basis of the head direction signal was likely integration of head angular velocity, and the outline of a
205 model for how this integration was performed using conjunctive head direction × head angular
206 velocity cells (observed in dorsal presubiculum and parietal cortex) was proposed¹⁸.

207

208 To many investigators, the foregoing observations collectively pointed *almost* inescapably to the
209 hypothesis that the primary determinant of the 'cognitive map' is some form of coordinate system in
210 which head angular velocity and linear velocity are integrated over time to express displacement and
211 orientation from a starting point (path integration)^{19-21,90,91} (Fig. 5). According to this view, the path-
212 integration mechanism assigns place fields based on motion integration. In the absence of external
213 stationary input, errors from noise in the self-motion integration process accumulate, and place
214 fields (and head direction tuning curves) would start to drift. However, in environments with salient
215 cues, rapidly formed associations between cues and place cells enable stabilization of the firing
216 fields, and previously formed 'maps' can be recalled from session to session^{10,19-21,90}, possibly cued by
217 landmark information conveyed through the dorsal presubiculum⁹². Nevertheless, there is also some
218 support for the idea that place cells were formed by integration of salient sensory inputs,
219 independently of movement. One of the main observations presented in favour of this concept was
220 that place fields could be seen to expand⁷¹ or stretch⁹³ in response to corresponding distortions of
221 the enclosure in which recordings took place. However, such distortions do not occur when the
222 animal is introduced *ab initio* into the distorted environment, but only when the animal has first
223 experienced the undistorted version. Stretching or expanding can thus be seen as a result of the
224 external inputs attempting to correct the path integrator based on prior associations⁹⁰.

225
226 During the past decade, virtual environments have enabled investigators to dissociate with increased
227 rigor the relative contributions of self-motion inputs and stationary landmarks. Typically, head-fixed
228 mice or rats run on an air-cushioned ball or a circular treadmill while visual flow is projected onto an
229 immersive screen at a rate that directly reflects the animal's running speed and direction, emulating
230 the sensory-motor coupling of the real world^{39,40}. When the virtual environment is linear, as on a
231 treadmill, hippocampal place cells exhibit firing fields that depend on distance moved^{94,95} or
232 stationary cues on the screen⁹⁴, with some variation between cells⁹⁴. Reducing the gain of ball-to-
233 virtual scene movement causes place fields to move towards the start of the virtual track, as
234 expected if firing locations are determined by self-motion, but the shift is generally smaller than
235 expected from movement distance alone, pointing to an additional role for visual inputs⁹⁴. The dual
236 dependence on self-motion cues and external cues confirms earlier studies where these sets of
237 inputs were disentangled in real environments^{73,74,93}. However, when the virtual environment is
238 made two-dimensional, and movement of the head remains restricted, localized firing breaks down,
239 although a small influence of distance travelled is detectable⁹⁶. In contrast, when body and head
240 rotation is unconstrained, stable position coding persists⁹⁷. Together these studies point to vestibular
241 signals (which are impoverished during head fixation) as a critical source for integrating velocity and

242 direction signals into a coherent two-dimensional representation, in agreement with earlier work
243 showing that place fields are disrupted following inactivation or lesions of the vestibular system^{98,99}.

244

245 **'Remapping': global, partial, local and rate**

246 In the late 1980's, Muller and Kubie began a series of investigations on the effects of changing the
247 most salient visual cues in a cylindrical environment and introducing various local cues^{71,72,100-102} (Fig.
248 6). As alluded to above, cue-card rotations, changes in the size or color of the cue card, or even
249 removal of the cue card altogether, rarely changed the radial coordinate of the field, but could
250 change the angular coordinate, completely unpredictably in the case of complete removal of the cue
251 card when the rat was not present (Fig. 6b). They coined the term "remapping" to describe any
252 manipulation-induced changes in the firing of place cells. These could include mild changes in the
253 firing characteristics in a few cells, such as when new objects or walls were placed in a cell's place
254 field, up to radical changes in the location of firing, including the disappearance of a field altogether,
255 which was sometimes observed when the environmental shape was changed or visual cues radically
256 altered.

257

258 Whether sets of place cells remapped completely or only partially depended on the experimental
259 conditions. The terms "global", "partial" and "local" remapping were introduced by Knierim and
260 McNaughton¹⁰³ in an attempt to distinguish the situations in which only fields near a specific,
261 manipulated, cue changed from the situation in which there was a general (partial or complete)
262 rearrangement of fields throughout the environment, as is often seen when the animal is placed in
263 non-uniform environments^{104,105}, or as in the cases of deficient plasticity discussed above^{82,83}. The
264 concept of remapping was clarified considerably by several experiments that followed. In 2005,
265 Leutgeb et al. showed that, when the cues in the recording chamber or its shape were radically
266 changed between sessions that took place in the same physical location, CA1 and CA3 place cells
267 underwent substantial changes in their firing rates, without changing their firing locations¹⁰⁶ (Fig. 6c).
268 These changes could be sufficient to make a field appear to be present in only one condition, unless
269 the rate map graphs were re-scaled. In contrast, when the recordings took place in identical apparatus
270 located in two separate rooms, the place field distributions became completely uncorrelated.
271 Leutgeb et al. made the distinction between "rate remapping" for the former situation and "global
272 remapping" for the latter. Thus, it appears that, under conditions in which the path-integrator
273 coordinates likely remain consistent, changes in external input or, indeed, internal variables such as
274 motivation, working memory or action plans can result in dramatic changes in firing rate while firing
275 location remains unaltered¹⁰⁷⁻¹¹⁰. Leutgeb et al. suggested that rate remapping might be the cause of
276 apparent "partial" remapping or direction dependency on linear tracks. The role of the path

277 integrator coordinates in governing rate vs. global remapping was fairly decisively demonstrated by
278 Colgin et al.¹¹¹, who showed that when environmental shape was gradually morphed between a
279 circle and a square, abrupt, global remapping only occurred if the rats had previously been allowed
280 to locomote between a circle and a square via a connecting tunnel. When rats were pre-trained on
281 the two shapes in the same location, only rate remapping was observed. Thus, it was the path
282 integrator that determined whether global or rate remapping was observed.

283

284 The presence of a non-spatial code on top of the place code ("rate remapping") is consistent with
285 dozens of studies, starting already in the 1980s, which show that place cells encode more than space.
286 Cells with clear place fields in one task were shown in other tasks to respond in a time-locked
287 manner to various non-spatial features of the environment or the experience, such as odors^{112,114},
288 textures¹¹⁵, conditioned tones^{28,116,117}, or temporal stages of the experiment¹¹⁸. However, in
289 combination with the remapping studies, these observations suggest that hippocampal cells respond
290 conjunctively to spatial and non-spatial variables, with the latter represented as changes in the rate
291 distribution. Experience-related changes in rate distribution can also account for moment-to-
292 moment variability of firing rates within place fields ('overdispersion')¹¹⁹. The conjunctive nature of
293 spatial and event-related firing is demonstrated elegantly in a more recent study of hippocampal
294 activity after systematic variation of location, food cups (objects), and color or pattern of the
295 recording box (context)¹²⁰. The majority of cells in this study fired at specific locations but with rates
296 depending on context and objects. Thus, when location is clamped, unique constellations of cues give
297 rise to unique rate patterns, implying that each experience is characterized by its own hippocampal-
298 neocortical output, even when those experiences occur at a fixed location. This uniqueness is a
299 necessary condition for the widely held view that hippocampus may provide an index that links
300 memory attributes distributed widely over neocortex¹²¹⁻¹²³. The wide range of stimulus
301 configurations that activate hippocampal firing, over and above space, has been taken as evidence
302 for a broad involvement of the hippocampus in episodic memory, where space is just one of several
303 attributes of the encoded representation⁴⁸.

304

305 Lest one conclude from the foregoing that the phenomenon of remapping or the necessity or
306 dominance of path integration is now fully understood, it is necessary to consider some remaining
307 flies in the ointment. First, Tanila, Shapiro and Eichenbaum^{124,125}, and later Knierim¹²⁶, have shown
308 that, when an animal is highly familiarized with the local and distal cues in an environment, rotating
309 these cue sets relative to each other can cause some CA1 cells to follow the local set while others
310 simultaneously follow the distal set (still others may remap). Such discordant responses are stronger
311 in CA1 than CA3¹²⁷. These effects are not inconsistent with a path-integration based *origin* of the

312 place fields if one assumes that the subsequent, plasticity-dependent, association between cues and
313 place cells that leads to robust rate-remapping is also strong enough in some cases to move the fields
314 independently, depending on which type of inputs dominate the synaptic input vector of a given cell.
315 The fact that this effect occurs predominantly in CA1, which lacks the potential stabilizing effects of
316 reciprocal excitatory connections present in CA3, tends to support such a view¹²⁷. A second possible
317 challenge is the fact that place fields can be expressed in CA1 under conditions in which the MEC is
318 effectively silenced¹²⁸⁻³⁰. This suggests that localized firing may itself be generated from alternative
319 inputs, such as from weakly spatially-modulated neurons in the lateral entorhinal cortex (LEC)¹³¹,
320 which may provide hippocampal cells with path integration-independent sensory inputs necessary
321 for efficient rate coding¹³². However, even under conditions where MEC inactivation does not impair
322 hippocampal place selectivity, the intervention causes instant remapping^{133,134}, suggesting that MEC
323 is obligatory for activating the correct place map. This does not preclude, of course, that place maps
324 are stored also in the CA3 network (e.g., the 'charts' of Samsonovich and McNaughton⁹⁰), and that in
325 the absence of a strong MEC input, CA3 attractor dynamics may result in the recall of some
326 previously constructed chart in the novel context.

327

328 **Moving from hippocampus to entorhinal cortex**

329 Until the 1990's, for primarily technical reasons, most recording studies had been confined to CA1 of
330 the dorsal hippocampus, in spite of the fact that hippocampal subfields may have distinct
331 computational functions. David Marr, had, in the early 1970s, already pointed to the unique
332 properties of area CA3 as a recurrent network capable of autoassociation, pattern formation and
333 pattern completion¹³⁵. His work was followed by theoretical investigations pointing to the possible
334 role of the dentate gyrus in pattern separation processes needed to counteract memory interference
335 at subsequent stages of the hippocampal circuit¹³⁶⁻¹³⁸. An additional, striking property that was
336 discovered to differentiate between hippocampal subfields was coding sparsity. Contrary to some
337 expectations, in the successive transformations from CA3 to CA1 to subiculum, mean firing rates
338 increased and coding became less sparse and less spatially selective^{139,140}. This observation led
339 Barnes et al. to conclude that "discrete spatial representations are constructed within early stages of
340 the process, for some purpose intrinsic to the hippocampus itself, possibly that of rapid information
341 storage" and that "the information leaving the hippocampus through the subiculum seems to consist
342 of much more highly distributed representations, constructed perhaps through the convergence and
343 disjunction of a number of unrelated hippocampal place cells"¹³⁹. For a long time, however, these
344 pioneering ideas did not fully catch the attention of the place-cell community, which, with few
345 exceptions, retained its focus on the readily accessible CA1 area.

346

347 In a similar manner, until the 1990s, there was minimal focus on computational operations outside
348 the hippocampus and computations underlying place-field formation were at risk of being
349 erroneously be attributed to the hippocampus itself. The focus on a hippocampal origin of the place-
350 cell signal was further influenced by the observations of a relatively small set of tetrode studies in the
351 entorhinal cortex, the major cortical input to the hippocampus. These studies showed that entorhinal
352 cells were spatially modulated, but their firing fields were broad and dispersed, with little spatial
353 selectivity in standard laboratory environments, and the fields seemed not to remap between
354 environments^{22,139,141}. This, together with the observation that CA1 place fields persisted following
355 large lesions of the dentate gyrus¹⁴², pointed to the remaining associative networks of CA3 as one
356 possible origin for the formation or learning of the sharply localized place signals seen in CA1. The
357 validity of this interpretation was questioned, however, by the fact that partial inactivation of CA3
358 cells, following inhibition of septal inputs, failed to remove spatial firing in CA1¹⁴³.

359
360 Given the uncertainty about how CA3 contributed to the CA1 place signal, Brun and colleagues¹⁴⁴
361 decided to record place cells in CA1 after the CA3 input to these cells had been entirely removed by
362 excitotoxins or by knife cuts that completely separated CA1 from CA3 as well as dentate gyrus and
363 subcortical afferent regions. Retrograde tracer injections in CA1 verified that no input was spared.
364 Confirming the interpretation of the septal-inactivation work¹⁴³, the study found, in 2002, that CA1
365 place cells do not require input from CA3 to maintain reasonably selective spatial firing. This
366 suggested that place fields were either generated within the limited circuitry of the CA1 itself, or
367 place cells in CA1 received spatial input from the entorhinal cortex via temporo-ammonic projections
368 that survived the CA3-CA1 transection. These observations were made only a few years after
369 theoretical studies^{3,21,90,145} proposed that the path integrator might located outside the hippocampus
370 – in the subiculum, the entorhinal cortex, or both – because correlations between firing fields in
371 these regions appeared to be invariant across contexts^{22,23}, as might be expected for a path
372 integration-based representation. At this time it was clear that the entorhinal cortex, the main
373 cortical input to the hippocampus, was worth a re-visit.

374
375 An important additional inspiration for the renewed interest in entorhinal cortex was Menno Witter's
376 extensive review of entorhinal-hippocampal systems¹⁴⁶. Witter pointed out that dorsal and ventral
377 regions of the hippocampus receive inputs from, and project back to, different regions of the
378 entorhinal cortex, in a topographical manner, with increasingly dorsal hippocampal regions mapping
379 onto areas that were increasingly closer to the rhinal sulcus, or increasingly more dorsal within the
380 medial entorhinal cortex (MEC). In 1990, based on his review, and after direct consultation with
381 Witter, two of us (M.-B.M. and E.I.M.) realized that in earlier MEC recordings where histology was

382 available^{22,141}, cells had been recorded quite far outside the area of MEC that receives most visual-
383 tactile information and projects most extensively to the dorsal hippocampus, where the most
384 sharply-tuned place cells of the hippocampus are located^{80,81}. This led us eventually, after the turn of
385 the millennium, to target tetrodes to the dorsal MEC, the origin of the majority of inputs to the
386 dorsal hippocampus^{8,146}, a region of MEC so far not touched by electrodes *in vivo*.

387

388 **Grid cells – a metric for space?**

389 Recordings in dorsal MEC soon showed that cells in this region have sharply defined firing fields,
390 much like in CA1 of the dorsal hippocampus, except that each cell had multiple firing fields,
391 distributed all over the environment⁸. These findings, reported in 2004, pointed to the MEC as a key
392 element of a circuit for space but the nature of the entorhinal representation remained elusive.

393

394 A striking characteristic of many spatially modulated MEC cells was that the distribution of the
395 multiple firing fields of each cell was more regular than expected by chance⁸. When the data from
396 MEC were presented at the 2004 Society for Neuroscience meeting, they created considerable
397 excitement. Among those who were most excited were Bill Skaggs, who thought he saw hexagonal
398 symmetry, inspiring the Mosers, and their students Hafting, Fyhn and Molden, to increase the size of
399 the recording arena and visualize the firing pattern once and for all. Using a newly constructed 2-m
400 wide circular recording cylinder, these authors found, in a substantial fraction of MEC superficial-
401 layer cells, that the firing fields of individual cells created a grid-like periodic hexagonal pattern tiling
402 the entire space available to the animal⁹ (Fig. 7a). The cells were coined grid cells. For each cell, the
403 grid could be assigned a phase (the *xy* locations of the grid vertices), a wavelength or spacing (the
404 distance between the vertices), and an orientation (how much the axes through the vertices were
405 tilted compared to an external reference line). In addition, the peak firing rates varied between
406 fields^{9,147}. The spatial periodicity of the pattern⁹ was so striking that the authors were concerned,
407 initially, that it was some sort of artefact. However, the grid pattern soon appeared in other labs
408 too^{131,148}.

409

410 One of the most striking aspects of the grid-cell finding was that the spatial periodicity was
411 maintained despite constant changes in the animal's running speed and running direction. The cells
412 fired at the same vertices regardless of how much time and space the rat had travelled between each
413 crossing, implying that grid cells had continuous access to information about distance and direction
414 moved. The persistence of grid fields⁹ and place fields⁷⁰ when rats run in darkness is consistent with
415 the primary role that such self-motion information might have in determining firing locations, as is
416 the fact that grid patterns unfold immediately in new environments⁹ and are expressed with similar

417 phase relationships between cell pairs in all environments tested¹⁴⁷. It should be added, for the sake
418 of balance, that stable grid fields have not yet been identified in darkness in mice^{149,150}. The reason
419 for the possible species difference is not known. Associations between path integration coordinates
420 and stationary cues may be weaker in mice¹⁵¹, or grid fields of mice may simply be harder to visualize
421 at times of increased jitter, given their smaller field size and shorter grid spacing compared to rats¹⁵².

422

423 Based on the possible role of self-motion information in the formation of grid patterns, the three of
424 us suggested, in 2006, that grid cells are part of an intrinsic path integration-based metric for space⁹¹.
425 A similar proposal was made the same year by a different group of investigators¹⁵³. Both concepts
426 bore similarities to the mechanism proposed a decade earlier from studies of place cells^{19,90}. In fact,
427 by implementing their attractor map model for path integration on a torus, Samsonovich and
428 McNaughton⁹⁰ indirectly predicted periodic place fields, although, at the time, the idea seemed to
429 them too preposterous to publish, and an attempt to discover such periodicity in CA1 by running rats
430 down a long hallway concluded that "place field distributions can best be described by a random
431 selection with replacement"¹⁵⁴. A decade later, with the new data from the entorhinal cortex, it was
432 clear that grid cells may supply the brain's spatial map with a coordinate system not available from
433 place cells in the hippocampus, given the apparently random allocation of place fields to position¹⁵⁵
434 and the related extreme remapping across environments.

435

436 It soon turned out that if grid cells supply a metric, this metric is not always constant over time or
437 locations. Experiments showed that when environments were stretched or rescaled, the spacing of
438 the grid increased in the extended direction^{148,156}, in concert with either scaling or remapping in
439 hippocampal place cells¹⁵⁷. However, these distortions of the grid pattern were recorded when the
440 environment was changed after the animal was already familiarized with it, suggesting that grid maps
441 might be formed by path integration but linked to external cues in such a way that the latter can
442 override the path integration dynamics⁹⁰. Yet, under some conditions, grid cells appear to be
443 fragmented or distorted even after extended training in a constantly shaped environment. When rats
444 are tested in environments with discrete compartments¹⁵⁸ or in irregular geometric shapes¹⁵⁹, the
445 strict periodicity of the grid pattern is often gone. In particular, it has been shown that walls exert
446 strong local influences on the grid pattern^{159,160}, causing distortions and rotations that can be
447 described effectively as a shearing process¹⁶⁰. The common existence of fragmented and distorted
448 grids have raised questions about whether grid cells are useful as a source of metric information¹⁵⁹.
449 Speaking against these doubts, theoretical analyses have shown that precise symmetry may not be
450 necessary for accurate population-based decoding of position, distance and direction if the grid cells

451 are all distorted in the same way¹⁶¹. Direct behavioral evidence is needed, however, to establish how
452 well spatial metrics can be decoded from distorted grid patterns.

453

454

455 **Network properties of grid cells**

456 Grid cells differed from place cells in more than one way. Not only did they have periodic firing fields
457 but also the relationship between the firing fields of different cells followed a different rule. Whereas
458 place cells often remap completely between environments and multiple fields can appear in large
459 environments, with no more overlap in the subset of active cells than expected by chance^{106,155,162-164},
460 the ensemble activity of grid cells is normally maintained coherently from one environment to the
461 next, without changing phase or orientation relationships between cells^{147,165}, much like in early
462 recordings from MEC cells before grid cells were discovered²². The coherence of the grid map is
463 particularly strong within ensembles, or modules, of similarly scaled grid cells¹⁵⁶. A similar degree of
464 coherence is present among head direction cells^{6,7,77,78,166} as well as in the more recently discovered
465 populations of entorhinal border cells and speed cells^{167,168}. The coherence of grid cells and head
466 direction cells is state-independent and persists during sleep¹⁶⁹⁻¹⁷¹. Collectively, these findings point
467 to a fundamental difference between hippocampal and entorhinal spatial maps: hippocampal circuits
468 are high-dimensional, capable of storing a very large number of patterns, while MEC maps are low-
469 dimensional and rigid, expressing the same intrinsic structure in all behavioral contexts, as would be
470 expected for a path integration-based map that keeps metric properties constant across contexts and
471 environments.

472

473 It was clear from the outset that grid cells come in different varieties – with different phases,
474 wavelengths, orientations and field amplitudes – and that the network of grid cells is anatomically
475 organized according to some but not all of these variables^{8,9}. While the phase of the grid pattern
476 appeared to be distributed randomly among cells on the same tetrode, the scale of the grid showed a
477 striking increase from dorsal to ventral recording locations in the MEC (Fig. 7b). In both respects, the
478 organization of grid cells was reminiscent of that of place cells, which also appear to have random
479 spatial relationships^{162,172,173} but show an increase in scale from dorsal to ventral^{80,81}. In the
480 hippocampus, the scale increase is strongly coupled with decreasing gain of self-motion
481 parameters^{84,86}. A similar gain change may underlie the scale change in MEC, consistent with the
482 hypothesis that the overall system parameters are dominated by path integration mechanisms.

483

484 One question that was not settled by the earliest grid-cell recordings was whether the scale gradients
485 were smooth and gradual or instead consisted of multiple discrete maps with distinguishable scale

486 and self-motion gain, the latter being a necessary prediction of attractor map based models^{91,174}. In
487 2007, Barry and colleagues showed with a small cell sample that values of grid spacing were not
488 evenly distributed¹⁴⁸. In 2012, Stensola and colleagues were able to record activity from up to 180
489 grid cells in the same animal – enough to determine once and for all if grid cells clustered in groups
490 with similar properties¹⁵⁶. Stensola et al. found that grid cells were organized in at least four
491 modules, each with their own scale, orientation and asymmetric distortions (Fig. 7c). The scale
492 change across successive grid modules could be described as a geometric progression with a
493 constant scale factor¹⁵⁶, confirming the prior predictions^{91,174}, as well as theoretical analyses pointing
494 to nested and modular organizations as the most efficient code for representing space at the highest-
495 possible resolution with the lowest-possible cell number^{175,176}.

496

497 The discovery of grid cells cast new light on the mechanisms underlying formation of place cells – the
498 very question that motivated the search for spatially modulated cells in the entorhinal cortex. The
499 periodicity of the firing pattern, and the variability of the grid scale, suggested early on that place
500 cells may emerge by a Fourier-like linear summation of output from grid cells with similar phase
501 throughout the environment over a range of spatial scales^{91,177}. This summation mechanism might be
502 facilitated further by coordinated gamma-frequency oscillations in MEC and CA1 cells¹⁷⁸.

503 Alternatively, and more in line with the sensory-integration ideas of the 1980s, place fields might be
504 generated from any weak spatial input, so long as the hippocampal circuit contains mechanisms for
505 amplifying a subset of these inputs, either through Hebbian plasticity or through local recurrent
506 networks¹⁷⁹⁻¹⁸². The merits of these two classes of models remain to be determined. Experimental
507 studies have shown that MEC grid cells are not necessary for the emergence of spatially tuned firing
508 in place cells. Place fields have been reported to persist when the spatially periodic firing pattern of
509 MEC grid cells is compromised by inactivation of septal inputs^{128,129}, and in young animals, place cells
510 acquire stable firing fields before sharp periodic firing patterns emerges in grid cells^{183,184}.

511 Inactivation or damage of the MEC is not sufficient to disrupt place-cell firing in the
512 hippocampus^{130,133,134,185}. However, neither of these observations rule out grid cells as a key
513 determinant of spatially selective firing in the hippocampus. The hippocampus receives input from
514 multiple spatially tuned entorhinal cell types, including not only grid cells but also border cells and
515 spatially modulated cells with non-periodic firing patterns¹⁸⁶, as well as weakly place-tuned cells in
516 the LEC¹³¹. Place fields may be formed from any of these inputs, by more than a single mechanism.
517 Even pure rate changes among the MEC inputs are sufficient to completely alter the activity
518 distribution among place cells in the hippocampus¹⁸⁵. The mechanism for grid to place or place to
519 grid transformation may have many faces, and understanding it may require that circuitry is
520 disentangled at a higher level of detail, possibly in terms of inputs and outputs of individual cells.

521

522 **A zoo of cell types**

523 Grid cells are abundant especially in the superficial layers of the MEC but not all cells are grid cells. As
524 early as 2006, it was clear that in layers III-VI of the rat MEC, a number of cells respond to head
525 direction¹⁸⁷ (Fig. 7d), very much like the head-direction cells reported in the neighboring pre- and
526 parasubiculum years before^{5-7,188}. The directional tuning curves of many entorhinal head direction
527 cells were found to be broader than in pre- and parasubiculum and many head direction cells
528 responded conjunctively to location, expressing grid-like firing fields but discharging within each grid
529 field only when the rat's face pointed in a certain direction¹⁸⁷. Head direction cells intermingled with
530 grid cells and conjunctive grid × head direction cells (Fig. 7e) throughout MEC layers III-VI as well as in
531 pre- and parasubiculum¹⁸⁹, pointing to a computational mechanism for imposing the angular
532 component of path integration on grid cells^{19,91}.

533

534 Shortly after head direction cells were observed in recordings from the MEC, another cell type
535 appeared on the entorhinal stage. These cells – named border cells – fired exclusively along
536 geometric borders of the local environment: along one or sometimes several walls of the recording
537 enclosure or along the edges of a platform^{167,190} (Fig. 7f). Border cells were distinct from grid cells – a
538 border cell could never be transformed to a grid cell or vice versa – but there was overlap between
539 border cells and head direction cells, i.e. some (conjunctive) border cells fired within their border
540 fields only when the animal was running in one direction¹⁶⁷. Border cells intermingled with grid cells
541 and head direction cells, particularly in layers II and III of MEC¹⁶⁷, suggesting that the three types of
542 cells interact. However, while grid cells and head direction cells seemed to be confined to
543 parahippocampal – and not hippocampal – regions, cells with border-like firing fields were observed
544 also in the hippocampus¹⁹¹ and the subiculum^{192,193}, raising the possibility that firing patterns of
545 entorhinal border cells are inherited by at least subsets of neurons in the hippocampus and
546 subiculum^{93,194}, or vice versa.

547

548 Border cells are more sparse than grid cells and head direction modulated cells, and may comprise
549 less than 10% of the local principal cell population¹⁶⁷, but this does not negate a significant role in
550 shaping hippocampal-entorhinal representations. The discovery of border-like properties in several
551 regions of the hippocampal formation confirmed, to some extent, predictions from computational
552 models dating back to the observation that the location and shape of place fields are determined by
553 local boundaries of the recording environment⁹³. Based on this observation, O'Keefe and Burgess and
554 their colleagues proposed a model in which place fields are formed by summation of tuning curves
555 from upstream 'boundary vector cells', cells with firing fields tuned to the animal's distance from a

556 particular wall or boundary in the environment^{93,192,194}. Boundary-vector-like cells, with distance-
557 dependent tuning curves, were reported in the subiculum¹⁹³, but given the unidirectional wiring of
558 the hippocampal circuit, these cells are unlikely to provide major input to hippocampal place cells.
559 Such inputs might instead come from border cells in the MEC. On the other hand, the latter lack
560 distance tuning, firing only along the borders and not away from them. If border cells provide input
561 to place cells, their influence might be limited to cells with firing fields in the periphery of the
562 recording enclosure, near boundaries and not in open spaces. There is some indirect evidence for
563 this possibility as, in juvenile rats, place cells with fields in the centre of an open recording
564 environment mature at the same slow rate as grid cells¹⁹⁵, which acquire adult-like hexagonal
565 symmetry only late in juvenile development^{183,184}. Place cells near the borders of the recording box
566 appear at an earlier age, similar to entorhinal border cells¹⁹⁶. Regardless of whether border cells fulfil
567 criteria for boundary vector cells or not, the existence of border cells, as well as the strong
568 asymmetries in grid patterns caused by environmental boundaries^{159,160}, point to a significant role for
569 boundaries in defining the location of firing in place cells and grid cells, consistent with behavioral
570 studies identifying geometry of the environment as a determinant of the animal's perception of self-
571 location^{13,197,198}. However, these observations are not at variance with a path integration-based
572 account of spatial firing of grid cells. Boundaries may serve as references for path integration-based
573 position estimates, with resetting of the path integrator, and subsequent reduction of error, taking
574 place regularly near major boundaries or landmarks^{19-21,90}. The increased variability of grid field
575 locations in open spaces compared to locations near the walls¹⁹⁹, as well as the instability of place
576 fields in open spaces when spatially stable information is available only from border cells¹⁹⁵, speak in
577 favor of a reference function for environmental boundaries, where grid and place representations
578 are reset and corrected from drift each time the animal encounters a salient boundary.

579

580 With the presence of head direction cells and border cells, it became clear that grid cells have local
581 access to information about direction, needed for the angular component of path integration, as well
582 as information about the geometry of the environment, needed to prevent drift in the path-
583 integrator coordinates. Head velocity signals upstream of head-direction cells – in the lateral
584 mammillary nuclei²⁰⁰ and further upstream in the dorsal tegmental nuclei^{201,202} – might enable head
585 direction cells to infer direction at the time scale of behavior. However, if grid cells express path
586 integration, they must also have access to information about moment-to-moment changes in the
587 animal's speed. Such information was known from early on to be present in the hippocampus, where
588 both place cells and fast-spiking interneurons exhibit speed tuning^{30,86,203}. Speed-responsive cells
589 have similarly been observed in subcortical areas directly or indirectly connected with hippocampal
590 and parahippocampal regions²⁰⁴⁻²⁰⁷. These cells might feed into the brain's path integration system.

591 Speed tuning of hippocampal theta rhythm amplitude is sufficient to enable accurate reconstruction
592 of distance travelled²⁰⁸, and distance travelled might be decoded by integrating the net discharge
593 rate of a population of hippocampal cells or afferents of the hippocampus.

594

595 The observation of speed coding in the hippocampus and subcortical areas motivated the search for
596 speed information locally within the MEC circuit. Already in 2006, it was observed that some
597 information about speed is present in a subset of grid cells, especially in layer III and deeper¹⁸⁷, but
598 the correlations between firing rate and speed in these cells were weak and would require decoding
599 from large cell numbers to yield a reliable momentary speed signal¹⁶⁸. We now know that the
600 entorhinal cortex has a distinct population of cells whose firing rates increase linearly with
601 speed^{168,209}. In the large majority of speed-tuned MEC cells¹⁶⁸, firing rates increase linearly as a
602 function of speed up to 30-40 cm/s. A small but significant number of cells have negative speed-rate
603 relationships¹⁶⁸. As in the hippocampus, many of these are fast-spiking cells²¹⁰. The rates of these
604 cells are tuned so strongly to running speed that speed can be decoded with extreme accuracy from
605 just half a dozen cells¹⁶⁸. Tuning profiles (slope and y-intercept of the speed-rate relationship) vary
606 between speed cells but remain constant across environments, and persist in the absence of visual
607 cues, pointing to speed cells as yet another component of a low-dimensional path integration-based
608 position map in the MEC¹⁶⁸. In CA1, the gain of speed tuning varies systematically along the septo-
609 temporal axis in register with the change in spatial scale⁸⁶. This has yet to be confirmed in MEC, but if
610 verified it would strongly support the idea that speed cells convey the necessary information to set
611 the grid scale.

612

613 Taken together, these observations point to a network of entorhinal and hippocampal neurons
614 where position, direction and distance are encoded at sufficient accuracy to enable dynamic
615 representation of the animal's location in an empty enclosure. However, most real-world
616 environments differ from experimental settings in that the available space is cluttered with objects.
617 Salient objects may serve as references for navigation but little is known to date about whether and
618 how objects are included in the representation of self-position in the MEC. It has been shown that a
619 subset of neurons in the LEC respond specifically at the location of discrete objects in the recording
620 enclosure^{211,212}. These neurons increase firing whenever the animal encounters an object at a certain
621 location, regardless of the exact identity of the object. In a subset of these object cells, firing even
622 persists for minutes, days or weeks after the object is removed²¹². Whether and how these cells
623 contribute to representation of the animal's own location has remained elusive. Theoretical models
624 from the 1990s postulated the existence of cells with place fields defined by the animal's vectorial
625 relationship to salient landmarks in allocentric coordinates²¹³ and such cells are indeed found in small

626 numbers in the hippocampus²¹⁴. These cells encode direction and distance from one or a small
627 number of discrete objects placed at different locations in the recording arena. Now new data
628 suggest that a class of MEC cells has more general vectorial properties. These 'object vector cells'
629 have firing fields defined by distance and direction from an object, regardless of the object's location
630 in the environment and regardless of what the object is²¹⁵. Thus, the main difference between object
631 vector cells in MEC and in CA1 appears to lie in their object specificity. Perhaps, like rate remapping
632 of hippocampal place cells, the coordinate information in CA1 is inherited from MEC whereas the
633 identity information is added after the fact, possibly from LEC^{131,132,211,212}. Like rate remapping in
634 place cells²¹⁶, at least some of the CA1 object vector cells appear to require extended experience²¹⁴.

635 Finally, investigators have identified a population of hippocampal cells with activity defined by the
636 animal's egocentric orientation to a goal location. Sarel et al.²¹⁷ recorded from the CA1 region of
637 flying bats, which have hippocampal-parahippocampal spatial representations similar to that of
638 rodents²¹⁸⁻²²⁰. The investigators identified a set of cells that responded as a function of the animal's
639 orientation towards a salient goal positioned centrally in the environment. Although the preferred
640 orientation of the cells spanned the full 360-degree range relative to the direction to the goal, a large
641 proportion of the cells in this category fired when the animal was heading directly towards the goal,
642 ramping up their firing as the bat approached the goal. A little more than half of the cells were also
643 place cells, but a substantial fraction did not have any significant tuning to place. Cells with
644 essentially the same characteristics were recently reported in posterior parietal cortex²²¹. Goal-
645 vector cells are reminiscent of cells reported in rats in earlier hippocampal studies, in which neural
646 firing increased in the proximity of a goal^{173,222-226}, and the finding of goal orientation cells in both
647 parietal cortex and hippocampus begs the question of which region is 'copying' which. It remains for
648 the future to determine if similar cells are also present in the MEC circuit and whether they remap
649 between goals and environments, like place cells, or maintain intrinsic spatial and directional
650 relationships, like all medial entorhinal functional cell types characterized so far.

651 The multitude of functionally specialized cell types in the entorhinal-hippocampal space circuit is
652 striking; however, equally striking is that many cells still express more than one type of information,
653 particularly in the intermediate and deep layers of MEC, where many grid cells fire conjunctively for
654 position and head direction, or position and speed, and many border cells are direction-
655 selective^{167,168,187,227}. Conjunctive cells are recognized to be essential ingredients of the 'hidden layer'
656 for almost any type of coordinate transformation or conditional association network^{18,228-230}. A
657 challenge for future work will be to determine how this variety and mixture of differently tuned cell
658 types enable a dynamic representation of self-position that can be read out to guide navigation and
659 memory for a wide variety of environments.

660

661 **The role of theory: Mechanisms of place cells, head direction cells, and grid cells**

662 The abundance of functionally dedicated cell types in the entorhinal-hippocampal system has
663 prompted investigators to look for the neural mechanisms that enable their characteristic firing
664 patterns. Mechanisms have been sought in the properties of single cells as well as neural networks.
665 While details remain elusive, the preceding sections of this review have already emphasized how
666 circumstantial evidence points to path integration-based attractor network properties as a key
667 contributor to pattern formation in the entorhinal-hippocampal space system.

668

669 Attractor networks have provided starting points for models of localized firing since the earliest
670 studies of hippocampal function. Already in 1949, Hebb proposed that activity may self-sustain in
671 networks of recurrently connected neurons²³¹. In 1977, Amari took a giant step by showing that
672 localized firing can be maintained in networks of neurons arranged conceptually on a ring with
673 Mexican-hat connectivity²³². In such an architecture, each neuron has strong excitatory connections
674 to its nearest neighbors, with excitation decreasing with distance along the ring, in contrast to
675 inhibition, which is maintained at longer distances. Almost 20 years later, Skaggs and McNaughton
676 and colleagues²³³, Zhang²³⁴, and Redish and Touretzky and colleagues²³⁵, showed, independently,
677 how the concept of a ring attractor with local (Gaussian) connectivity and global recurrent inhibition
678 could be used to explain the emergence of directionally specific firing in head direction cells (Fig. 5b).
679 The connectivity created a self-maintained activity bump, which could be induced to move around
680 the ring in accordance with external angular velocity signals that were transmitted through a hidden
681 layer of conjunctive, head direction \times angular velocity cells¹⁸. The model explained a number of
682 features of head direction cells, including the persistence of directional phase relationships across
683 conditions and environments. Today – more than 20 years after its proposal – the key concepts of
684 the ring-attractor model for head direction cells remain unchallenged, which is remarkable for
685 theoretical models in systems neuroscience, and competing models have not surfaced. In mammals,
686 the reciprocally connected network of the dorsal tegmental nucleus and the lateral mammillary area
687 has been proposed as a location for the ring attractor²³⁶, and in *Drosophila*, the concept of a ring
688 attractor for directional tuning has received its first experimental support in studies of central body
689 neurons, where a circular anatomical arrangement has been shown to underlie firing in neurons that
690 represent orientation relative to landmarks^{237,238}.

691

692 Only a year after the introduction of velocity-driven ring attractors to models of head direction cells,
693 it was acknowledged that a similar integration mechanism might apply for position mapping in two
694 dimensions, as expressed in hippocampal place cells^{19,90,234,239,240} (Fig. 5cd). In the position version of

695 the model, neurons were arranged conceptually according to their location of firing in two-
696 dimensional space. A matrix of recurrent connections was generated in which excitation decreased
697 with the distance between neurons on the sheet. In combination with global inhibition, self-
698 excitation between similarly tuned cells maintained localized firing. A path-integration mechanism
699 moved the activity bump across the network in accordance with the animal's position in the
700 environment, using conjunctive head direction \times place cells, in the same way that angular velocity
701 inputs moved the bump in the ring attractor for head direction cells. The model was proposed to
702 apply for any neural architecture of the hippocampal system, but with the knowledge that existed in
703 the 1990s, the implementation was focused on area CA3 of the hippocampus. This explained a
704 number of properties of place cells but faced one major challenge – that the subset of active
705 hippocampal neurons remaps across environments and circumstances^{71,100-102}. For position to be
706 computed in place cells, some sort of independent architecture for each environment would then be
707 required. This is computationally possible^{90,241} but nonetheless raises the question of whether a
708 single network matrix, expressed in all environments, would not be more efficient^{21,240}. A few years
709 later it became apparent that such a low-dimensional architecture exists in the entorhinal cortex.

710

711 When grid cells entered the research arena in 2005⁹, it was quite obvious that the dynamics
712 proposed for localized firing in place cells might take place also in parahippocampal regions^{91,153,240},
713 as alluded to already by Samsonovitch and McNaughton⁹⁰. In the first models proposed after the
714 discovery of grid cells^{91,153}, cells were arranged on a matrix according to the phase of the grid. A
715 bump of activity was formed when cells with similar phases were connected through excitatory
716 connections, in the presence of global inhibition. Competitive network interactions led to multiple
717 activity bumps¹⁵³, or toroidal connectivity caused a single bump that returned periodically to the
718 same location⁹¹. Under certain conditions, in the presence of tonic excitatory input, a radius of
719 inhibitory connectivity was sufficient to generate hexagonally patterned firing, without intrinsic
720 excitatory connections²⁴²⁻²⁴⁵.

721

722 Whether a path integration-based attractor network architecture exists in MEC remains to be
723 determined but there is indirect evidence for this possibility. First, correspondence between
724 movement and displacement on the neural sheet can only be maintained so long as the participating
725 grid cells have a common scale and orientation. Grid cells exist at a range of scales, suggesting that to
726 maintain the correspondence, grid cells must be organized in functionally independent grid modules,
727 all with their own spacing and orientation^{91,174}. Experimental evidence suggests that such a modular
728 functional organization is indeed present^{148,156}. A second observation consistent with a path
729 integration-dependent attractor architecture is the maintenance of a single grid-phase structure

730 across environments and tasks^{147,165}, which would be expected if MEC neurons are organized as
731 strongly interconnected networks where external inputs recruit the same subset of neurons under a
732 wide range of starting conditions. The strongest prediction of the attractor models, however, is
733 perhaps that grid cells with similar grid phases have enhanced connectivity. Statistical analysis of
734 firing patterns in simultaneously recorded grid cells confirm this prediction^{246,247} but direct
735 measurements of connections between functionally verified cell types are still missing.

736

737 Attractor models do not provide the only possible explanation of how grid patterns might be created.
738 For several years, a competing class of models, based on properties of the hippocampal theta-
739 frequency network rhythm⁶⁰⁻⁶², suggested that grid patterns were generated as a result of wave
740 interference between a constant global theta oscillation and a velocity-controlled cell-specific theta
741 oscillation²⁴⁸⁻²⁵¹. The model can be traced back to O'Keefe and Recce's observation, in the early
742 1990s, that, as animals move through the place field of a place cell on a linear track, the spike times
743 of the cell move forward across the cycle of background theta oscillations²⁵². As the animal moves
744 through the field, the theta phase moves progressively forward also in space, in fact with a stronger
745 correlation to location than time^{252,253}. This observation suggested to O'Keefe and colleagues that
746 position could be calculated from the interference pattern between the global theta rhythm and a
747 velocity-dependent oscillator specific to the cell. If position reflected peaks of the interference
748 pattern, however, the firing positions should be periodic, which, for place cells, they were not. With
749 the discovery of grid cells, the model was instantly revised and grid patterns were suggested to
750 emerge from interference with velocity-controlled oscillators controlled by the projection of velocity
751 in three directions separated by 60-degree intervals onto three separate dendrites²⁴⁸⁻²⁵⁰. Interference
752 with the global oscillator led to a band-like spatial activity pattern along each orientation, and the
753 combination of bands led to a hexagonal pattern. The oscillatory interference models guided some of
754 the most influential studies of grid formation, but in the end accumulating evidence, such as the
755 biophysical implausibility of independent dendritic oscillations²⁵⁴, the sensitivity to period
756 irregularity²⁵⁵, the persistence of grid patterns in the absence of theta oscillations^{219,256}, as well as the
757 presence of a ramping depolarization and the absence of a theta interference oscillation in
758 intracellular recordings from MEC cells^{257,258}, suggested that oscillatory interference is not the
759 mechanism of the grid pattern. Yet phase precession is a reliable observation. Although it may not
760 explain periodicity in grid cells, phase precession causes sequences of place-cell activation to be
761 replicated, in compressed format, within individual theta cycles – an effect that may be used by
762 hippocampal circuits to store temporal sequences in addition to mere locations²⁵³. Indeed, as
763 recognized by several investigators soon after phase precession was discovered^{253,259,260}, theta

764 rhythm and phase precession may exist precisely to enable memory for spatial and temporal
765 sequences.

766

767 The evidence against the oscillatory-interference model did not, however, rule out single-cell
768 properties as determinants of the grid pattern. Kropff and Treves²⁴ showed how hexagonally
769 patterned firing may arise through competitive Hebbian plasticity in a path integration-independent
770 manner in feed-forward networks where neurons undergo neuronal fatigue or adaptation. Because
771 the emergence of grids in this model required many iterations, it was proposed that the adaptation
772 mechanism contributed particularly to development of the network in young animals, and that the
773 coherence of phase and orientation relationships across environments was the result of recurrent
774 connections that were added as the cortex matured²⁶¹. Thus, competitive Hebbian plasticity offers an
775 alternative mechanism for grid formation, although this mechanism may co-exist with attractor-
776 network architectures²⁶². Regardless of mechanism, accounts of grid formation must consider not
777 only intrinsic MEC dynamics but also how external inputs from the hippocampus²⁴³, the medial
778 septum^{128,129}, and locomotor^{204-207,263} and head-direction circuits²⁶⁴ contribute to the emergence of
779 grid patterns (Box 1).

780

781

782 **Perspective**

783 The search for a hippocampal positioning system began with the discovery of place cells in 1971. We
784 have illustrated how the first subsequent decades were characterized by attempts to find the
785 determinants of spatially localized firing, with a focus on the sensory sources. As we entered the
786 1990s, the discovery of head direction cells, and the turn to population dynamics, prepared the field
787 for more targeted investigation of the circuit operations underlying place field formation and spatial
788 mapping. The 1990s showed how ensembles of simultaneously recorded hippocampal neurons
789 encoded functions that could not be read out from the activity of individual neurons. From around
790 2000, with increasing awareness that these ensembles likely extended beyond the hippocampus,
791 investigators entered the entorhinal cortex, and an intricate circuit of grid cells and other specialized
792 cell types was discovered there. The investigation of space has been brought to a new level, where it
793 is possible to ask questions about how functions emerge through interactions within extended
794 networks of heterogeneously connected cell types and subsystems.

795

796 While we will certainly learn more about the neural origins of spatial cognition during the years to
797 come (Box 1), studies of spatial representation and navigation are informative about cortical
798 functions in a wider sense. The ease at which spatial functions can be studied in the hippocampal

799 formation of a number of mammals has made the study of the positioning system pioneer the
800 development and testing of sophisticated computational neural-network models. Few other areas of
801 systems neuroscience have benefited so strongly from the interplay between computational and
802 experimental neuroscience. Place cells and their entorhinal counterparts have helped to open the
803 cortex to studies of neural computation, allowing researchers to identify generic circuit motifs that
804 may be expressed not only in the spatial circuits of the hippocampus and entorhinal cortex but across
805 widespread regions of the brain. Almost 50 years after place cells were discovered, place cells and
806 their parahippocampal counterparts have become one of the most powerful tools we have for
807 understanding cortical computation and spatial mapping, and navigation may become one of the first
808 cognitive functions to be understood in mechanistic terms.

809

810 **In memoriam**

811 In memoriam, Howard B. Eichenbaum (1947-2017; Fig. 8). The field of hippocampal and memory
812 research mourns the loss of our friend and colleague Howard, who passed away unexpectedly
813 recently. Howard's contribution to the field was immense both scientifically and in service. His
814 research was mostly focussed on one of the major aspects that we have explicitly not covered in this
815 review: the role of the hippocampus in memory. Over the years, his position evolved from that of an
816 unafraid and much-needed devil's advocate against the pure spatial map hypothesis towards what is
817 now the general consensus view that spatial coding provides a foundation on top of which sensory
818 and event-specific memory is superimposed, and he became a pioneer in the study of how time and
819 temporal order also play a role. His thinking on hippocampal-cortical interactions in memory
820 organization and control is beautifully summarized in his 2017 Annual Review of Psychology article⁴⁷.

821

822

823 **Figure Captions**

824

825 **Figure 1:** Some historical milestones in the study of spatial coding in the hippocampal formation.

826

827 **Figure 2:** Place cells.

828 a) First place cell described⁴. Arrows and letters mark positions at which the animal was restrained as
829 it was pushed or coaxed around the test platform. Firing rate of the unit is illustrated by the
830 frequency histograms in the middle of the figure. Letters correspond to positions and lines indicate
831 periods of restraint. Bottom lines show spikes at the onset of the unit response at A (1) and during
832 the absence of a response at D (2). Calibration bar 400 msec. Note that the cell responds selectively

833 at only a few positions. O'Keefe and Dostrovsky reported 8 units out of 76 recorded hippocampal
834 cells that responded solely or maximally when the rat was situated in a particular part of the testing
835 platform facing in a particular direction. It should be noted that the single electrode technology
836 available to the authors at the likely precluded regular good isolation of cells, which may have limited
837 the number of clear 'place' responses observed. Reproduced from ref. 4, with permission.

838 b) A place field as typically displayed today. Top rat's trajectory in grey; spike locations superimposed
839 as black dots. Bottom: Color-coded rate map; dark red is maximum rate, blue is silence. Regions not
840 visited in black.

841 c) Left: The book by John O'Keefe and Lynn Nadel was long a "Bible" in the study of spatial coding in
842 the hippocampal formation. Right: Nadel (left) and O'Keefe (right) during preparation of the book.
843 Picture taken by Dulcie Conway around 1975. Picture courtesy of John O'Keefe²⁶⁵

844

845 **Figure 3:** Ensemble recording technology.

846 a) The principle of tetrode recording proposed by McNaughton et al.³² exploits the variation in
847 extracellular spike height as a function of distance to the recording site to resolve multiple single
848 units in structures such as hippocampus, where the neurons are fairly tightly packed. Example of
849 spike amplitude clusters from a tetrode recording showing two of the 4 spike-amplitude dimensions.
850 The corresponding spike waveforms are shown on the right. Reproduced from ref. 80.

851 b) A 48 channel, 12 tetrode probe array (hyperdrive) from ca. 1995. This system exploited the
852 flexibility of wire tetrodes to enable them to be advanced by pushing them through gently curving
853 tubes (like a mosquito proboscis).

854 c) Multi-tetrode recording made it possible to record from more than 100 hippocampal neurons
855 simultaneously. Here is shown 80 firing rate maps from simultaneously recorded CA1 cells as the rat
856 ran in a 70x70 cm arena³³. Firing rate is color-coded from blue (silent) to red (maximum rate). Note,
857 many CA1 cells were virtually silent in this particular arena, whereas about 40% had place fields. Six
858 of the recorded cells correspond to fast-spiking cells (interneurons) which have much less spatial
859 selectivity. Reproduced from ref. 33.

860 d) Examples of the actual (blue) spatial trajectory of the rat and the trajectory reconstructed from
861 the population firing rate vector (red). Reproduced from ref. 33.

862

863 **Figure 4:** Head direction cells⁶. a) Firing rate as a function of head direction for 2 representative cells
864 from 2 different animals. b) A head direction cell firing rate in polar coordinates. Peak firing rate, in
865 the left orientation, is 6 Hz. c) Jeffrey Taube (left) and James B. Ranck Jr. (right), SUNY Downstate
866 Medical Center in Brooklyn., N.Y., in 1987. Panel a reproduced from ref. 6, with permission. Picture
867 courtesy of Jeffrey Taube.

868

869 **Figure 5: Path integration.**

870 a) Illustration of Mittlestaedt & Mittlestaedt 1980 experiment¹². This experiment showed that
871 rodents can perform angular and linear path integration. A female mouse returns directly to her nest
872 after finding a lost pup in total darkness, but makes a heading error if she is rotated below vestibular
873 threshold prior to starting the inbound journey. Reproduced from ref. 91.

874 b) The Skaggs et al. continuous attractor model from 1995 proposed to explain how head direction
875 cells arise through 'integration' of head angular velocity signals from the vestibular system^{18,228}.
876 Updating in the head direction (attractor) layer was performed by a 'hidden layer' of cells conjunctive
877 for head angular velocity and starting head direction whose return projections to the head direction
878 layer are offset according to the sign of rotation. Such conjunctive cells have been found in several
879 regions of the brain. Reproduced from ref. 233.

880 c & d) The continuous attractor model for path integration in 2-D proposed by McNaughton et al. in
881 1996¹⁹ and simulated by Samsonovich and McNaughton in 1997⁹⁰. H'=head angular velocity; H'H =
882 conjunctive cells; H = head direction; P = place cells; M = speed cells; PH x M = cells conjunctive for
883 place and head direction and modulated by speed; V = external sensory inputs that were assumed to
884 associatively bind to both H cells and P cells to enable correction of drift error in the path integrator
885 and resetting of the integrator upon entry to a familiar environment. Reproduced from ref. 90.

886

887 **Figure 6: Remapping**

888 a) John Kubie and Robert Muller from SUNY Downstate Medical Center., NY. Picture courtesy of John
889 Kubie.

890 b) Global remapping apparently induced by only changing the color of the recording environment¹⁰⁰.
891 Rate maps are shown for the same place cell recorded in a white cylinder (left) and a black cylinder
892 (right). Firing rate is color-coded from yellow (no firing) to dark blue or black (high rate). The cell fires
893 in different regions of the cylinder (and other cells are active in only one cylinder) despite changing
894 only the color of the box. It is interesting to know that the authors later confirmed anecdotally that
895 they pretrained the animals in the white and black cylinder in two different rooms, which would have
896 allowed differences in path integrator coordinates to control the global remapping, as later shown by
897 Colgin et al.¹¹¹. Rate maps adapted from ref. 100, with permission.

898 c) Rate remapping induced by changing the color of the recording environment while keeping its
899 location constant¹⁰⁶. The rat's trajectory in a white and a black box is shown for three cells, with
900 spikes superimposed as red dots. Note that changing only the color of the box causes substantial
901 change in the distribution of firing rates across cells, but firing locations are retained. Adapted from
902 ref. 106.

903

904 **Figure 7:** Grid cells and other functional cell types of the MEC.

905 a) Firing fields of one of the first grid cells reported in 2005⁹. Left, trajectory of the rat (black) with
906 superimposed spike locations (red). Middle: color-coded rate map with peak rate indicated. Right:

907 spatial autocorrelogram, color-coded from blue ($r = -1$) through green ($r = 0$) to red ($r = 1$).

908 Reproduced from ref. 9.

909 b) Sagittal section of the rat brain showing the hippocampus and the MEC (red area) and grid cells of
910 different scales recorded at three locations on the dorsal – ventral axis (trajectories with spike
911 locations as in a). Note expansion of grid cell from dorsal to ventral MEC. Adapted from ref. 91.

912 c) Grid cell modules¹⁵⁶. Top: autocorrelation plots showing grid patterns at successive positions along
913 the dorsoventral axis of MEC. Bottom: grid size, defined as distance between grid vertices, as a
914 function of position along the dorsoventral MEC axis (positions rank-ordered). Note that the increase
915 in grid size is not linear but discretized, following a geometric order with a factor of approximately
916 square root of two. Mean grid size for each module is indicated by stippled lines. Such
917 modularization is an essential prediction of the attractor map theory if it is to account for variable
918 spatial scaling⁹¹. Adapted from ref. 156.

919 d) Head-direction cell in layer V of MEC. Adapted from ref. 187.

920 e) Conjunctive grid \times head-direction cell in layer III of MEC. Adapted from ref. 187.

921 f) Border cell¹⁶⁷. Color-coded rate maps showing cell with selective firing along one of the walls of the
922 recording environment. Top: open environment. Bottom: rate map following insertion of a wall. Note
923 that the border cell responds to the same side of the wall insert as the main wall in the environment.
924 Adapted from ref. 167.

925

926 **Figure 8:** Howard Eichenbaum (1947-2017). Few have contributed more to the modern
927 understanding of hippocampal memory function, with place cells as a key component, than Howard
928 Eichenbaum, who sadly passed away, far too early, before the publication of this article. Photo
929 credit: Photographer Dan Kirksey, KDKC Photos, Escondido CA.

930

931 **Box 1**

932

933 We have listed some outstanding problems in entorhinal-hippocampal space circuits that we believe
934 can be addressed with state-of-the-art systems neuroscience tools:

935

936 **1. Path-integration networks and mechanisms of grid cells and head direction cells**

937 The performance of attractor network models for space relies on a unique and testable connectivity
938 between functionally similar cells. With state-of-the-art tools for neural imaging, genetic tagging and
939 structural analysis, it may soon be possible to examine directly, in large MEC populations, the
940 probability of connections between functionally identified neurons with various degrees of feature
941 similarity and dissimilarity. On a longer time scale, one may hope for a direct visualization, with in
942 vivo microscopy, of activity flow between connected mammalian neurons in a way that matches the
943 animal's movement in space (similar to ref. 237 and 238 in flies).

944

945 **2. Development of spatial network architectures**

946 How is the specificity of the hippocampal-entorhinal spatial neural network architectures achieved
947 during development of the nervous system? Excitatory neurons from the same radial glial progenitor
948 are known to have stronger interconnections than other cells^{266,267}. Might such connectivity between
949 clonally related cells underlie a possible preferential coupling between similarly tuned MEC cells, in
950 the same way that cells from the same clone exhibit similarities in orientation preferences, and
951 possibly preferential coupling, in the visual cortex^{268,269}? Does the young MEC have a topographically
952 arranged teaching layer, with connections between clonally related cells, that during early postnatal
953 development gives way to the largely non-topographical^{9,270} grid-cell network of the adult MEC (Fig. 8
954 of ref. 91)? Tools for targeted analysis of functional identity and connectivity of discrete
955 developmental cell populations have been developed, allowing these questions to be resolved in the
956 near future²⁷¹.

957

958 **3. Including the entire entorhinal-hippocampal circuit**

959 A key objective for a more complete understanding of entorhinal-hippocampal function will be to
960 determine how cell types with different functional correlates map onto the variety of morphological
961 or neurochemical cell types and their unique connectivity patterns. Recent data suggest that, in layer
962 II of MEC, both stellate and pyramidal cells can be grid cells, although stellate cells may contain the
963 majority of them^{257,258,272-275}. If so, are grid patterns created independently in these two cell classes,
964 or does one of them inherit the grid from the other?

965

966 **4. Read-out**

967 Position can be decoded from grid cells and place cells, with greater accuracy in grid cells than place
968 cells if the population is multimodular and scaled in particular ways^{161,175,176,276}. Whether neural
969 circuits decode information in the same way remains to be determined, however. Do neurons have
970 access to grid cells with different phase relationships or different spacing; do they integrate
971 information from grid cells with information from border cells or head direction cells? If so, where

972 are these neurons and how do they communicate with neocortical regions involved in strategy
973 formation and decision making? Most research on the mechanisms of spatial coding in hippocampus
974 has focused on the nature of the *inputs* that contribute to it, and less is known about the impact of
975 hippocampal *output* on coding dynamics in the widespread regions²⁷⁷ of neocortex and other areas
976 to which the hippocampal formation projects. The impact of outputs from the entorhinal-
977 hippocampal circuit will perhaps constitute a new frontier in the study of this system.

978

979 **5. Moving towards naturalistic environments**

980 Natural environments are large, three-dimensional, compartmentalized, nested and full of objects.
981 Ultimately studies of the hippocampal-entorhinal circuit should explore how cells map environments
982 of shapes, sizes and content more comparable to the animal's natural habitat²⁷⁸. Are grid cells, head
983 direction cells and place cells used only for local mapping, in the range of a few meters, or is the
984 entorhinal-hippocampal network used also for extended spaces, and if so, how? Is there a single
985 continuous map, or are there different maps for different local spaces, as proposed by theoretical
986 studies²⁷⁹ as well as observations in compartmentalized laboratory environments¹⁵⁸? If the latter is
987 true, how are they connected? And how is space coded in large and three-dimensional
988 environments²⁷⁸? In flying bats, place cells have spherical firing fields²⁸⁰ and head-direction cells are
989 tuned to all three axes of orientation²²⁰. Whether such volumetric coding extends to terrestrial
990 animals remains unsettled, although experimental data suggest that, in rats, head direction is
991 encoded not only by classical azimuth-sensitive head direction cells but also by cells, in the lateral
992 mammillary bodies, that respond to head pitch²⁰⁰. Observations in rats also suggest that the tilt of a
993 surface is factored into hippocampal and entorhinal representations of space^{281,282}.

994

995 **6. Representation of time**

996 Understanding space requires understanding time. Direct representation of the passage of time was
997 not observed in hippocampal neurons until the Buzsáki and Eichenbaum groups showed that when
998 animals run for a known interval at a steady location, in a running wheel²⁸³ or on a treadmill²⁸⁴,
999 hippocampal neurons fire successively at distinct times during the interval, following the same order
1000 on each trial. Cells with similar properties are present in the medial entorhinal cortex²⁸⁵. Most of
1001 these 'time cells' have discrete place or grid fields in standard spatial foraging tasks. Different
1002 assemblies and sequences of hippocampal time cells are active in tasks with different delays²⁸⁴,
1003 suggesting that hippocampal ensembles encode temporally organized information much the same
1004 way as they represent space. The observation of time cells is a provocative finding that may share
1005 properties with mechanisms underlying path integration-based representation of location, but the
1006 temporally confined firing fields of time cells do not disappear when time and distance are decoupled

1007 by restraining the animal²⁸⁶ or changing the speed of the treadmill²⁸⁷, suggesting that sequences do
1008 not exclusively reflect the number of steps at the task location. Certainly the relationship between
1009 representations of space and time and the role of time cells in perception and recall of time require
1010 further study. While time cells have firing fields in the order of a few seconds, and assemblies of time
1011 cells can represent events at the scale of tens of seconds, encoding of longer temporal distances may
1012 require different mechanisms. One may speculate that the spontaneous drift over hours and days in
1013 the firing properties of place cells in CA2 and to a lesser extent CA1²⁸⁸⁻²⁹⁰, as well as cell populations
1014 in LEC²⁹¹, may possess the power to encode temporally distant events as distinguishable memories.

1015

1016 **7. Beyond physical space**

1017 Do grid cells and other spatially modulated cells encode information beyond physical space, as
1018 suggested already by O'Keefe and Nadel⁶⁵? Evidence for such an extension of functions has been
1019 reported recently in a task where rats press a lever to alter the frequency of a sound on a continuous
1020 scale, hippocampal and entorhinal cells display frequency fields resembling place fields during
1021 navigation of physical space²⁹². Further functional expansion might be expected in primates. Indeed,
1022 in monkeys, hippocampal and entorhinal cells fire in patterns defined not by the animal's location in
1023 space but by where it moves its eyes on a visual scene^{256,293,294}. This observation raises the possibility
1024 that place and grid cells create a map of visual space using eye movement signals instead of
1025 locomotor information to support coordinate transformation, without having to change any other
1026 computational elements of the circuit. In humans^{295,296}, grid cells may take on functions in conceptual
1027 mapping²⁹⁷. The possible adoption of grid cells as a metric for navigating abstract spaces would be
1028 consistent with the idea that hippocampal circuits first evolved for representation of space and later
1029 acquired capacity for imaginary navigation^{49,65,298,299}. This expansion of functions would be
1030 reminiscent of the way cortices originally involved in object recognition formed the basis for a visual
1031 word form area during the evolution of written language processing in the human cortex³⁰⁰.

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Figure 1

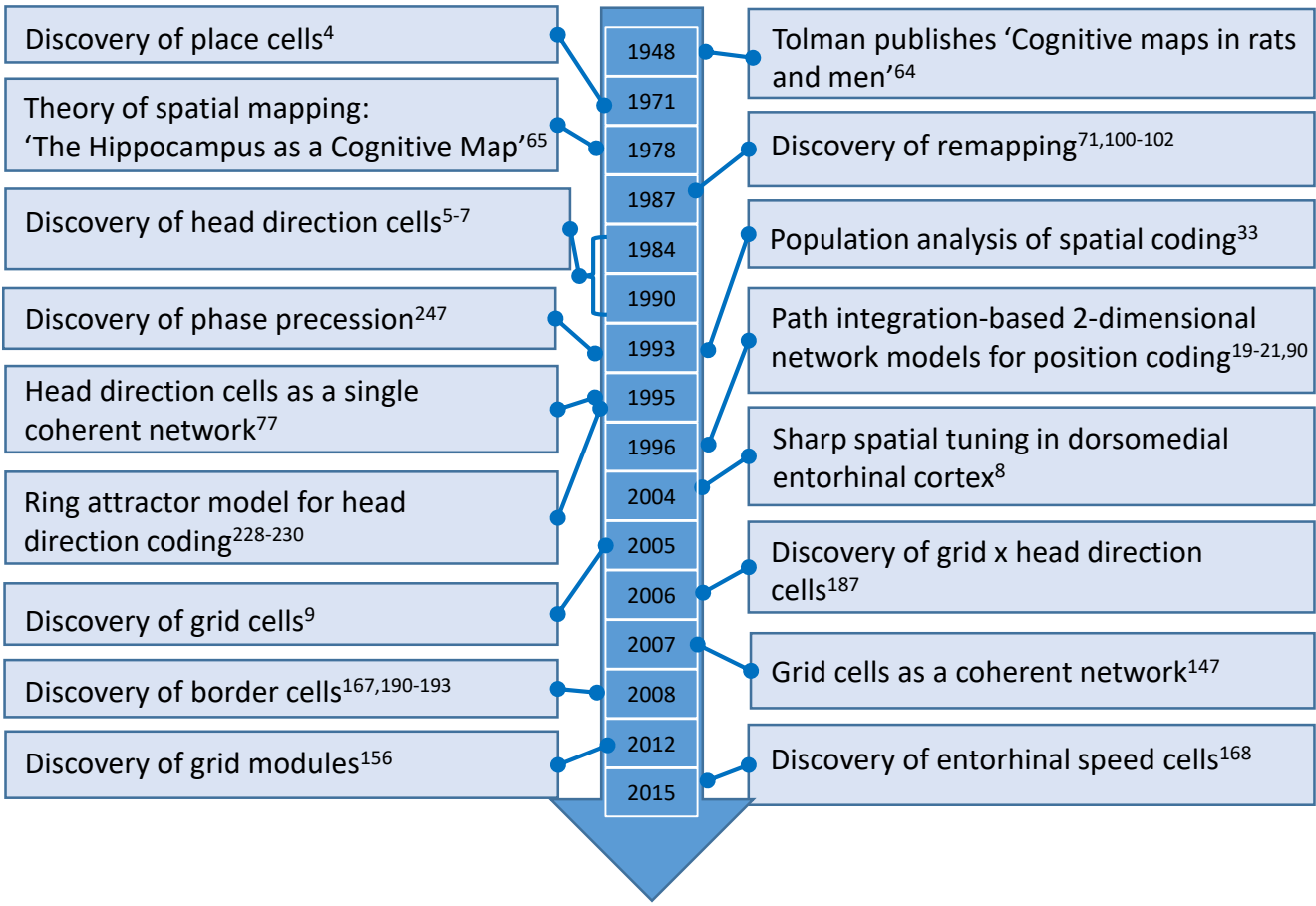


Figure 2

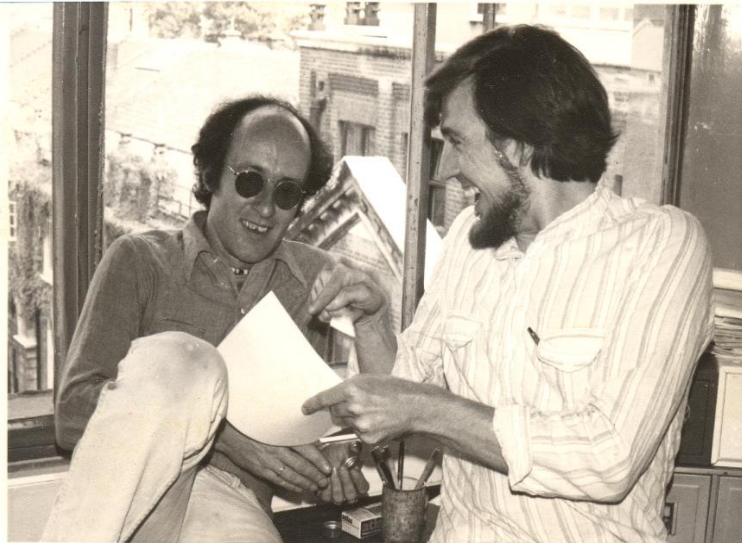
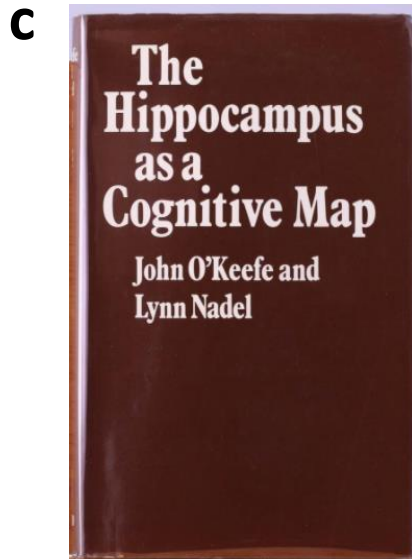
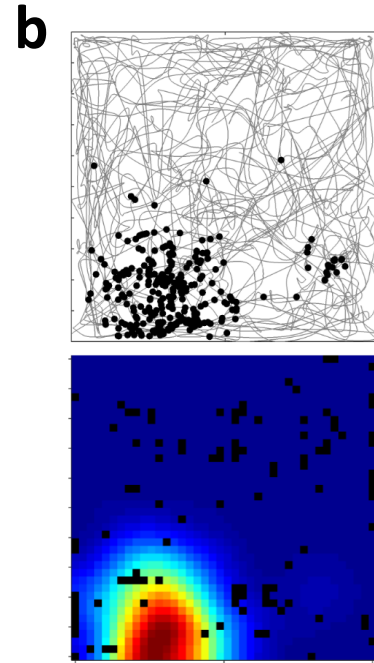
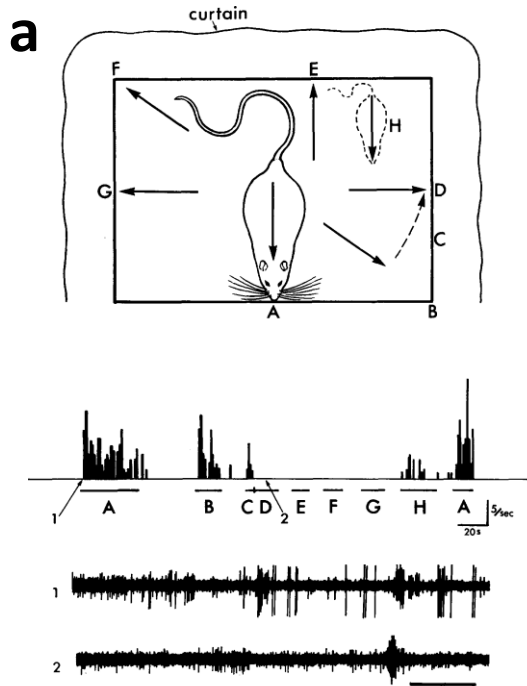
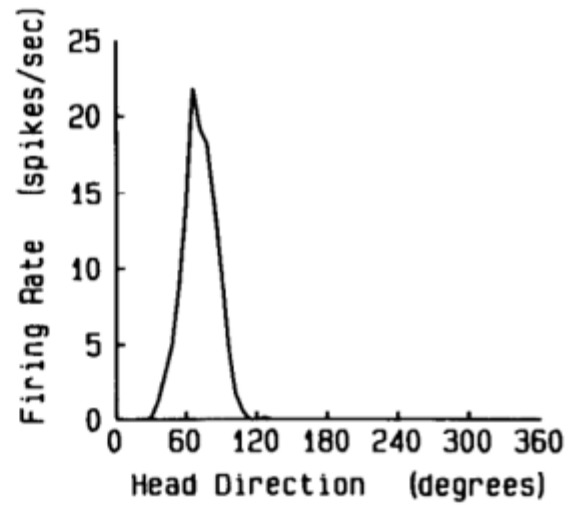
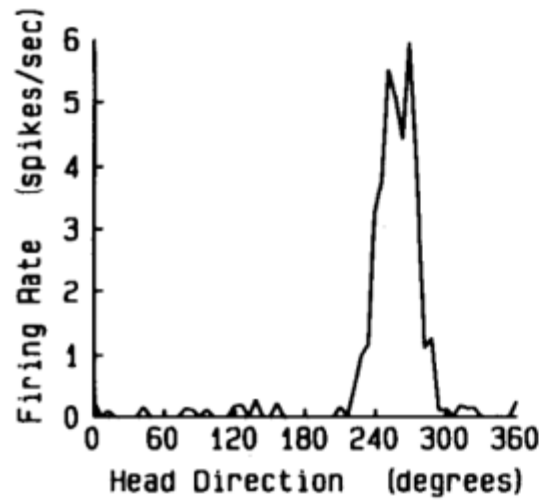
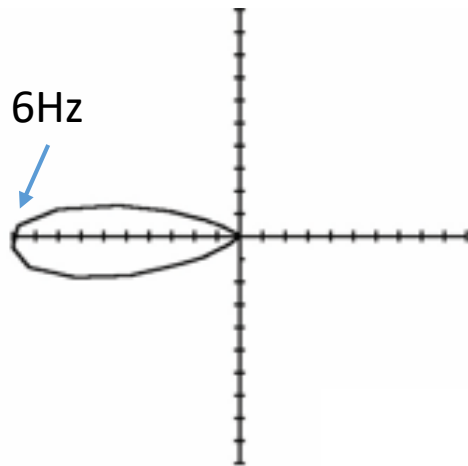


Figure 3

a



b



c



Figure 4

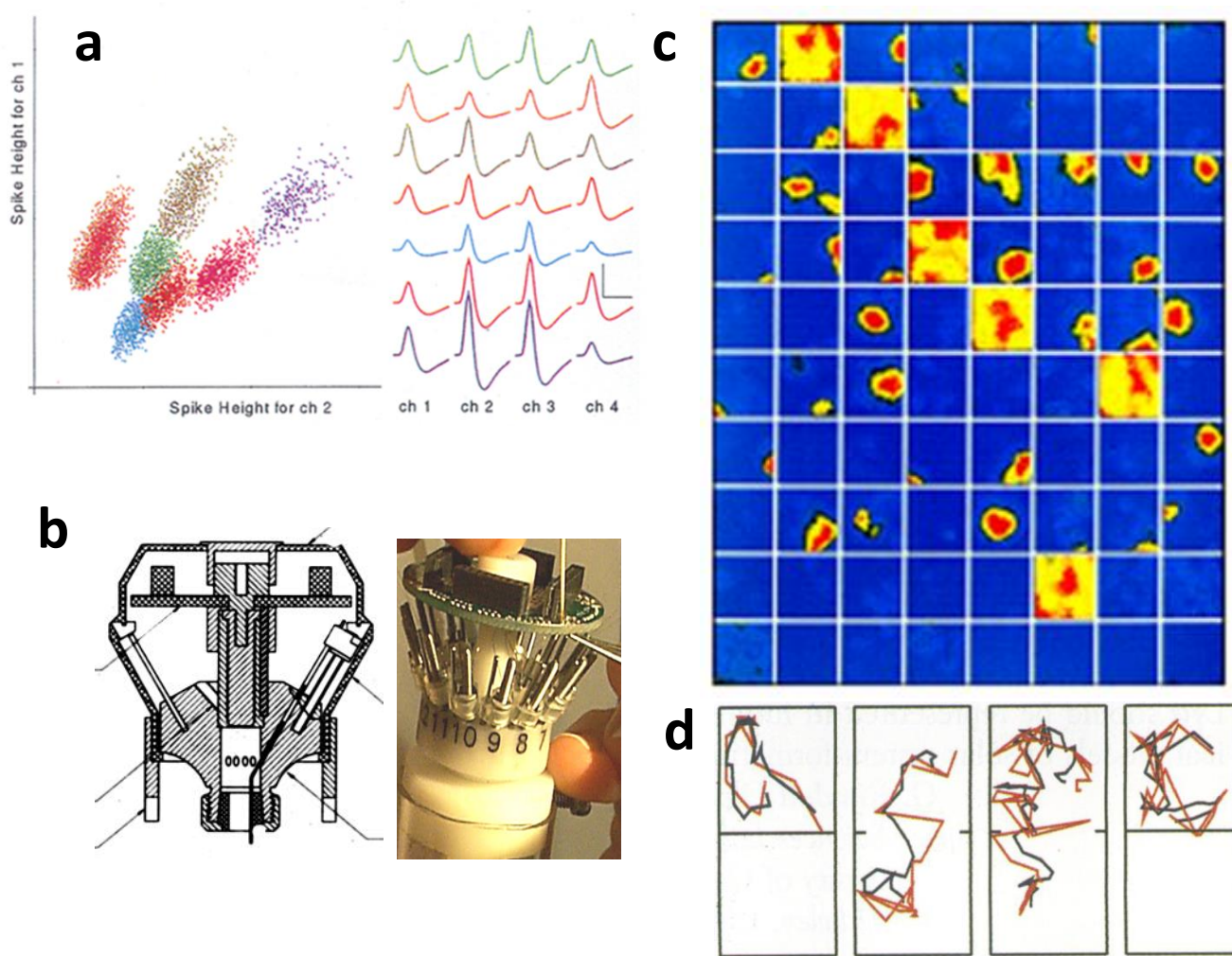


Figure 5

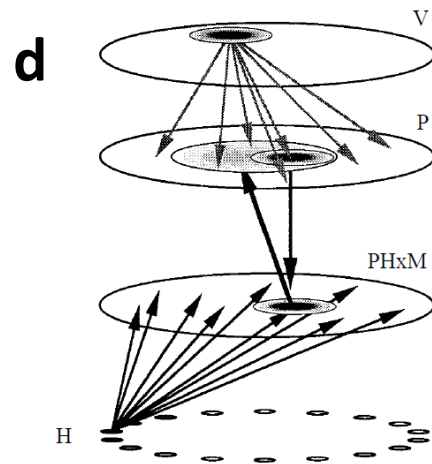
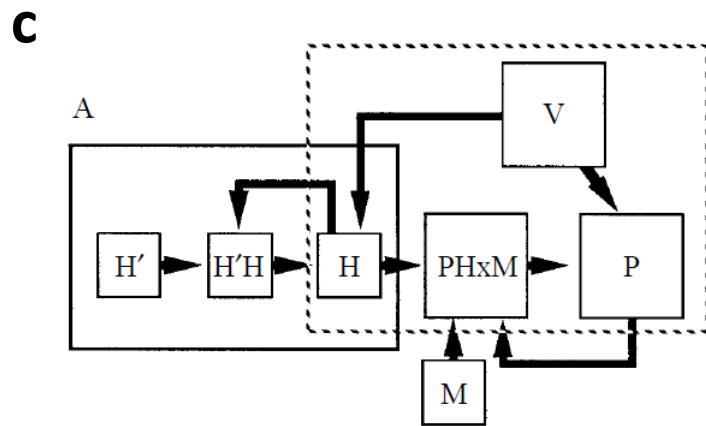
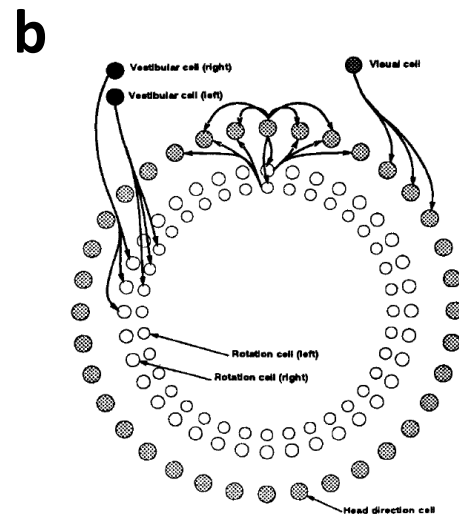
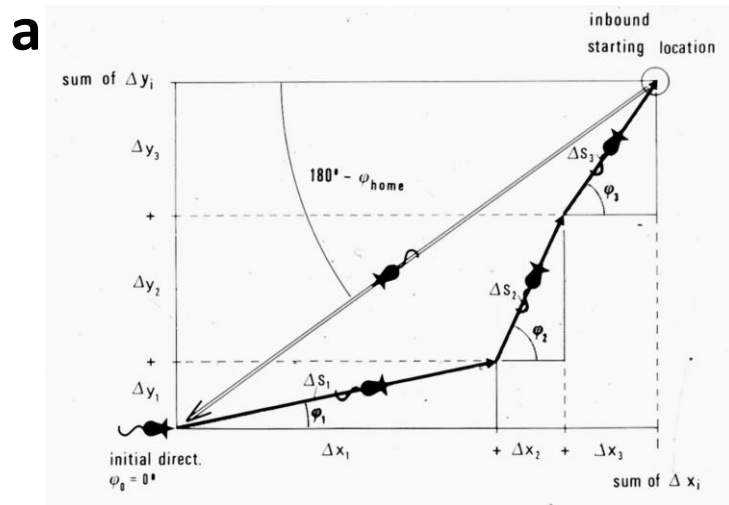


Figure 6

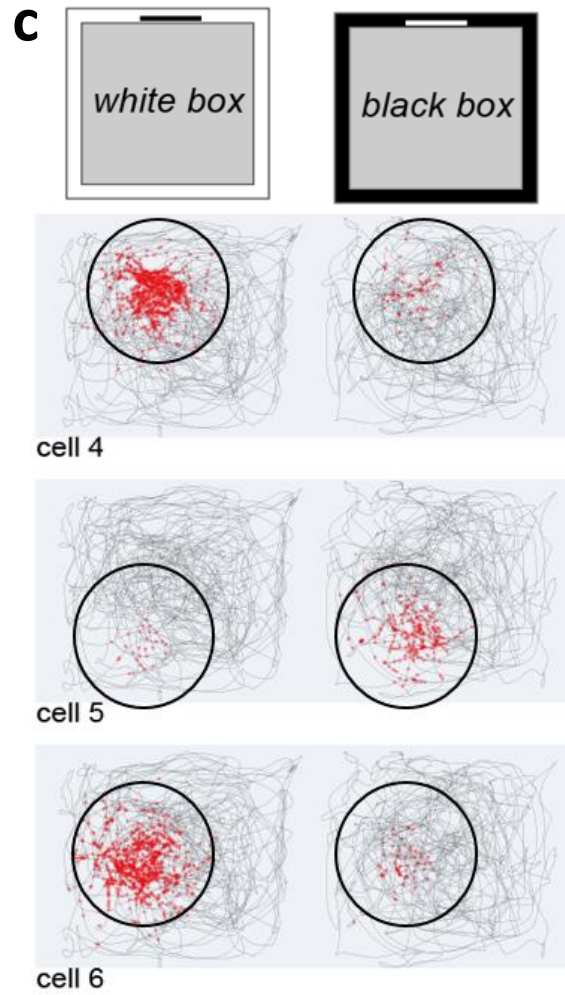
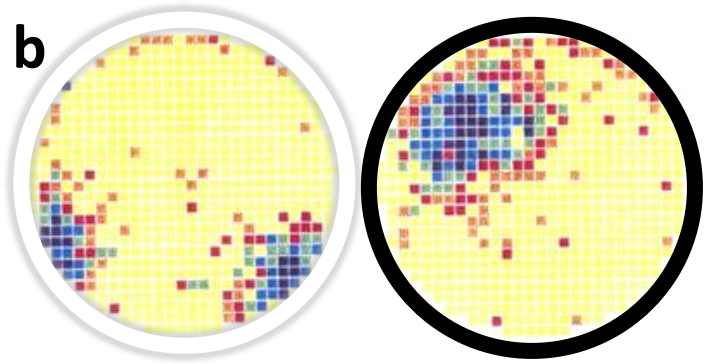


Figure 7

