From Synaptic Plasticity to Spatial Maps and Sequence Learning
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Abstract: The entorhinal–hippocampal circuit is crucial for several forms of learning and memory, especially sequence learning, including spatial navigation. The challenge is to understand the underlying mechanisms. Pioneering discoveries of spatial selectivity in this circuit, i.e. place cells and grid cells, provided a major step forward in tackling this challenge. Considerable research has also shown that sequence learning relies on synaptic plasticity, especially the Hebbian or the NMDAR-dependent synaptic plasticity. This raises several questions: Are spatial maps plastic? If so, what is the contribution of Hebbian plasticity to spatial map plasticity? How does the spatial map plasticity contribute to sequence learning? A combination of computational and experimental studies has shown that NMDAR-mediated plasticity and theta rhythm can have specific effects on the formation and experiential modification of spatial maps to facilitate predictive coding. Advances in transgenic techniques have provided further support for these mechanisms. Although many exciting challenges remain, these findings have brought us closer to solving the puzzle of how the hippocampal system contributes to spatial memory, and point to a way forward. © 2015 Wiley Periodicals, Inc.

KEY WORDS: place cells; grid cells; NMDAR; synaptic plasticity; STDP

PLACE CELLS

The hippocampal formation has been implicated in a range of behaviors but a cohesive pattern was elusive. The first major breakthrough occurred when careful studies of patients with damage to the hippocampal formation revealed profound deficits in recent memory (Scoville and Milner, 1957) which can be broadly categorized as episodic (Tulving, 1985), declarative (Squire, 1992; Eichenbaum, 2000) or spatial memory (Morris, 1984). Subsequent studies indicate that the hippocampal formation is crucial for various forms of sequence learning, including spatial navigational computation (Abbott and Blum, 1996; Jensen and Lisman, 1996; Mehta et al., 1997; Mehta, 2001; Fortin et al., 2002). Investigation of hippocampal single unit responses in behaving subjects provided the next major breakthrough with the discovery of place cells, namely neurons that fire in a restricted region of space as a function of the rat’s spatial position (O’Keefe and Dostrovsky, 1971), thus providing an allocentric cognitive map of space (O’Keefe and Nadel, 1978). In fact, although there are hundreds of thousands of neurons in the rat hippocampus, the activity of less than hundred dorsal CA1 neurons is sufficient to accurately decode the rat’s spatial position (Wilson and McNaughton, 1993). Neurons in all parts of rat hippocampus show robust spatial selectivity, with interesting interregional differences (Lee et al., 2004; Leutgeb et al., 2004). Navigation not only requires spatial localization but also spatial orientation and the discovery of head direction cells marked a major step forward (Taube et al., 1990).

Place cells have been found in the hippocampus in several other species with unique differences. Hippocampal neurons show robust spatial selectivity in mice, even though it is somewhat weaker than in rats (Cho et al., 1998; Ahmed and Mehta, 2009; Resnik et al., 2012). Spatially selective activity has been found in human patients as well during tasks involving sequential paths (Ekstrom et al., 2003; Miller et al., 2013), although the selectivity is considerably weaker than in rodents (Jacobs et al., 2010). Reports of allocentric representation of space in nonhuman primates are mixed, with one study reporting significant allocentric spatial selectivity during a sequential movement task (Ono et al., 1993), while another study reporting no allocentric spatial selectivity during a random foraging task (Rolls, 1999). Selective hippocampal responses to nontemporal stimuli, such as visual cues, have been reported in human and nonhuman primates (Ono et al., 1993; Rolls, 1999; Quiroga et al., 2005). Recent studies show that visual cues elicit reliable responses in rats as well, to generate head-direction selectivity (Acharya et al., 2015). Olfactory cues too elicit hippocampal responses in rodents (Wood et al., 1999). Nonspatial hippocampal responses have also been found in tasks that include a sequential component (McEchron and Disterholt, 1997; Pastalkova et al., 2008; MacDonald et al., 2011). These findings show that hippocampal neurons show selectivity to a variety of multisensory stimuli and are robustly activated in tasks involving spatial sequences.
How does the hippocampus form spatial maps? Robust spatial maps are found even in young rats that walk in the world for the first time, at age P19, when many parts of the neocortex are not yet fully developed (Langston et al., 2010; Wills et al., 2010), suggesting that innate mechanisms could generate spatial maps. Furthermore, although spatial maps can be robust for many months in familiar environments (Thompson and Best, 1990), place cells sometimes abruptly become active in a novel environment (Hill, 1978), which could arise due to synaptic plasticity or novelty-induced neuremodulation.

A large body of studies have implicated Hebbian (Hebb, 1949) synaptic plasticity in hippocampus-dependent learning and memory (Bliss and Lomo, 1973). In particular, the contribution of NMDA-receptor mediated synaptic plasticity (Bliss and Collingridge, 1993) to hippocampal function has received much attention (Morris, 1984; Bannerman et al., 1995). NMDAR-mediated plasticity between hippocampal areas CA3 and CA1 has been most extensively investigated and it is required for spatial learning (Tsien et al., 1996). Depending on the nature of stimulation NMDAR-dependent synapses show long-term potentiation (LTP) and depression (LTD) (Shouval et al., 2002; Malenka and Bear, 2004; Kumar and Mehta, 2011). Both the rate of stimulation and the precise timing of the stimulation determine LTP and LTD. Due to their voltage-dependent magnesium block, NMDAR-dependent synaptic plasticity requires coincident activation of the presynaptic terminal and postsynaptic spine and it therefore critically depends on precise spiking timing of the presynaptic and postsynaptic neurons hence it is termed spike timing-dependent plasticity (STDP) (Magee hence it Johnston, 1997; Markram et al., 1997; Bi and Poo, 2001). However, NMDAR-mediated synaptic plasticity can also occur without postsynaptic spiking. Instead, it is mediated by cooperative mechanisms within dendrites that provide sufficient depolarization for plasticity induction (Golding et al., 2002; Mehta, 2004; Kumar and Mehta, 2011).

How does Hebbian plasticity alter hippocampal spatial maps during sequential tasks and facilitate predictive coding? Computational models based on attractor networks show that Hebbian plasticity within the recurrent CA3 network should make place cells more anticipatory, therefore enabling the animals to predict the upcoming location based on past experience, i.e. navigate (Blum and Abbott, 1996; Gerstner and Abbott, 1997). Experimental measurements show that place fields in CA1 indeed become increasingly more anticipatory with experience (Mehta and McNaughton, 1997a; Mehta et al., 1997). Place fields also fire more robustly with experience. These changes occur rapidly, within a couple of sequential traversals of a linear track. This experiential place field plasticity is unlikely to arise from nonspecific effects such as novelty or neuremodulation because similar levels of place field plasticity occur in both novel and familiar environments (Mehta et al., 1997, 2000) and administration of NMDAR-antagonists blocks place field plasticity (Ekstrom et al., 2001). These results therefore support the hypothesis that anticipatory place field plasticity is a result of Hebbian synaptic plasticity within recurrent CA3 network.

However, the experiential place field plasticity is observed in area CA1, which has little excitatory recurrence, a necessary component of the attractor network models. Can CA1 place field plasticity be inherited from CA3 or instead arise from plasticity within CA3-CA1 feed-forward network? CA3 place fields do not show significant experiential anticipatory shift (Roth et al., 2012a), which implicates CA3-CA1 network for the observed CA1 place field plasticity. Robust plasticity, including STDP is indeed found in NMDAR-mediated synapses from CA3 to CA1 (Wittenberg and Wang, 2006), and this plasticity is involved in spatial learning (Tsien et al., 1996). What is the effect of STDP in the feed-forward, CA3-CA1 network on CA1 place fields? Computational models show that STDP in a feed forward network, such as CA3-CA1, can also make CA1 place fields more robust and anticipatory. In addition, the model predicts that CA1 place fields should have an asymmetric, ramping shape (Mehta et al., 2000). This prediction is supported by several subsequent experiments (Mehta et al., 2000; Roth et al., 2012b). Furthermore, in transgenic mice that lack NMDAR-dependent plasticity between CA3-CA1, CA1 place fields exhibit weaker ramping asymmetry and lesser experiential plasticity (Cabral et al., 2014) than control animals. The feed-forward model (Mehta et al., 2000; Mehta, 2001) also predicted that the subthreshold membrane potential of CA1 neurons should produce ramping, asymmetric shape, and the asymmetry of the membrane potential should be greater than the observed ramping in the spiking output of the neuron, due to recurrent inhibition and spike-frequency adaptation. Recent experiments in virtual reality support this hypothesis (Harvey et al., 2009) (Fig. 1).

These computational and experimental studies thus confirm that NMDAR-mediated plasticity in the CA3-CA1 network plays a crucial role in the experiential dynamics of CA1 neurons during sequential tasks, thereby making the receptive fields more robust, more anticipatory and more spatially asymmetric. This underlying plasticity is environment specific (Mehta et al., 2000), which could be responsible for the rapid, experience-dependent and environment-specific expression of immediate early gene (Guzowski et al., 1999).

The feed-forward model (Mehta, 2001) is fairly general and applies equally well to commonly found feed-forward networks in other brain areas. For example, ramping, asymmetric membrane potential is found in the entorhinal grid cells in experiments using virtual reality (Schmidt-Hieber and Häusser, 2013) (Fig. 1). Anticipatory shifts in receptive fields have been found in many other systems including rodent head direction
neurons (Yu et al., 2006) and the tadpole optic tectum (Engert et al., 2002). The asymmetric ramping excitation within a receptive field may interact with theta rhythmic inhibition to generate the precise spike-timing needed for inducing STDP (Mehta et al., 2000, 2002; Mehta, 2001).

DENDRITIC CONTRIBUTION TO SYNAPTIC PLASTICITY AND PLACE CELLS

The feed-forward STDP model can explain the experiential modification of place fields, but how do place fields arise? Indeed, place fields often appear abruptly not only in novel environments (Hill, 1978) but also in familiar environments (Mehta and McNaughton, 1997a) (Fig. 2). While this abrupt onset in CA1 could be the result of STDP mediated changes upstream, another possibility is that the CA1 dendrites, where the excitatory synapses are intricately arranged, play a crucial role because they are electrically active. STDP requires a coincidence between presynaptic spike and the postsynaptic backpropagating action potential (bAP), but its amplitude quickly decays with dendritic distance and may never reach the distal dendrites. Computational models show that this dendritic attenuation of bAP can determine both the magnitude and direction of synaptic plasticity such that different segments of dendrites are tuned to different stimulation frequency for inducing maximal LTP (Kumar and Mehta, 2011). Further, activation of a single inhibitory synapse on a dendrite could interfere with the propagation of bAP and thus interfere with STDP.

FIGURE 1. Asymmetric, ramping shape of place fields on linear tracks. A) Firing rate of a single place field as a function of rat’s position is spatially asymmetric, or negatively skewed (inset), with ramping increase in firing rate, which is consistent the predictions of computational model of STDP within CA3-CA1 synapses (Mehta et al., 2000). B) Subthreshold membrane potential of CA1 place field is spatially asymmetric, showing ramping depolarization (Harvey et al., 2009). C) Both firing rate and subthreshold membrane potential are spatially asymmetric in the entorhinal grid cells (Schmidt-Hieber and Hauser, 2013). [Color figure can be viewed in the online issue, which is available at wileyonlinelibrary.com.]

FIGURE 2. Abrupt appearance of place fields. A) Spikes (blue dots) within a CA1 place field as a function of spatial position and experience (lap number). The neuron spikes rarely for the first few trials and then abruptly begins to spike robustly, subsequently showing experiential anticipatory shift of the place field center (red line) (Mehta et al. 1997a). B) Small depolarizing current injection (bottom, blue) does not generate a significant response in a CA1 neuron (bottom, black), but a slightly stronger depolarizing current injection (top, red) results in the abrupt appearance of a place field (Lee et al., 2012). [Color figure can be viewed in the online issue, which is available at wileyonlinelibrary.com.]
Thus, despite many successes of Hebbian synaptic plasticity rule, a major limitation is that it does not take into account the nonlinear processes of the neuron's extensive dendritic arbor. However, recent studies show that upon sufficient depolarization, dendrites can generate their own dendritic spikes, which can remove the magnesium block to induce NMDAR-mediated synaptic plasticity, in complete absence of somatic spikes (Golding et al., 2002). Hence a comprehensive model of synaptic plasticity needs to take into account both active and passive dendritic processing (Mehta, 2004; Kumar and Mehta, 2011). According to this view the unit of memory is not the synapse, but instead, an entire cluster of adjacent synapses on a dendrite that can cooperate to generate a local dendritic spike, which in turn can induce NMDAR-mediated plasticity locally and represent a brief temporal sequence of events. Inhibitory synapses flanking the cluster of excitatory synapses may regulate the spread of this synaptic cluster. Unlike Hebbian learning rule or STDP, such dendritic spike mediated cooperative LTP does not require postsynaptic somatic spiking and hence could be responsible for the abrupt appearance of a place field after sequential experience (Fig. 2) (Mehta and McNaughton, 1997a; Mehta, 2004). Small changes in the strength of coactivated and clustered synapses on a dendrite could abruptly generate dendritic spikes resulting in abrupt appearance of place fields. This hypothesis is consistent with experiments showing that small changes in subthreshold depolarization of CA1 neurons results in the abrupt appearance of a place field (Lee et al., 2012) (Fig. 2).

In fact, dendritic processing could play a key role in the formation of hippocampal spatial maps. For example, CA1 dendrites show dendritic distance-dependent synaptic scaling such that synaptic strengths increase with increasing dendritic distance (Magee and Cook, 2000). Abolishment of the dendritic synaptic scaling may thus interfere with synaptic integration and spatial map formation. Indeed, GluR1-deficient mice have impaired dendritic synaptic scaling (Andrasfalvy et al., 2003) and almost complete absence of spatial selectivity in CA1 (Resnik et al., 2012). Additional experiments are needed that measure dendritic potentials from hippocampal neurons in freely behaving rodents to determine their role in spatial map formation and learning.

**ENTORHINAL GRID CELLS AND PLASTICITY**

In addition to CA3, the entorhinal cortex is a major source of input to CA1, with entorhinal synapses arising on the distal most CA1 dendrites. As a consequence it was thought that the entorhinal influence on CA1 somatic spiking should be minimal while CA1 is largely driven by CA3. However, surgical removal of CA3 afferents leaves CA1 spatial selectivity and spatial memory relatively intact, suggesting that entorhinal inputs are in fact sufficient to drive CA1 (Brun et al., 2002). Furthermore, the medial entorhinal cortex (MEC) provides strong inputs to dorsal CA1, and MEC neurons also show spatial selectivity. Surprisingly, however, MEC neurons have multiple place fields that are evenly distributed in space along the vertices of a triangular lattice, hence termed grid cells (Hafting et al., 2005). This was a seminal discovery not only because it provided evidence that spatial selectivity existed outside hippocampus proper, but also posed a puzzle: What mechanism could generate this spatially periodic activity although there is no such periodicity in the rat’s behavior, or the environment?

One of the first models to explain the formation of grid cells was the oscillatory interference model (O’Keefe and Recce, 1993; Hasselmo et al., 2007; Jeejeebhoy et al., 2008). It posits that in addition to the 6–12 Hz theta rhythm (Green and Arduini, 1954) there must be another source of theta rhythm. The sum of these two slightly different oscillators could cause an interference pattern, akin to beats produced by two similar sound sources. The low frequency component of the interference pattern was suggested to be the grid fields (Hafting et al., 2005) while the high frequency component can explain the theta phase-precession (O’Keefe and Recce, 1993; Skaggs et al., 1996; Harris et al., 2002; Mehta et al., 2002), which supports the interference model.

However, other observations challenge the interference model. For example, bats have clear grid fields although they lack theta rhythm during locomotion (Yartsev et al., 2011) which is not compatible with the interference model. The absence of theta rhythm in bats is also at apparent odds with extensive literature showing that theta-burst plasticity induces robust NMDAR-mediated synaptic plasticity, especially STDP. How can bats learn spatial maps without theta rhythm? An alternate model suggests that to induce robust NMDAR-mediated plasticity, including STDP, low-frequency correlated noise that modulates the excitability of a neural ensemble is sufficient, even if there is no clear rhythmic modulation (Mehta, 2001; Mehta et al., 2002; Kumar and Mehta, 2011). Further an increase in the rhythmicity of the correlated noise can significantly enhance NMDAR-mediated plasticity (Kumar and Mehta, 2011).

While the theta-interference model is not valid for bats, could it still apply in rodents? Intact hippocampal phase precession is observed when rats navigate in virtual reality without any speed-dependence of theta-frequency (Ravassard et al., 2013), a necessary requirement of the interference model. This challenges the interference model in rodent hippocampus, but the model may apply in the rodent entorhinal cortex. A prominent model suggested that the second source of entorhinal-specific theta rhythm is the h-current (Mehta, 2011). This model can explain several experimental findings, including the existence of phase-precession within each grid field, and the dorso-ventral gradient of the entorhinal grid field size (Hafting et al., 2008). To test the h-current mediated oscillatory interference model, entorhinal activity was measured in knockout mice lacking the h-current. Contrary to the prediction of the oscillatory interference model that both phase precession and grid pattern should be abolished, grid cells were intact (Giocomo et al., 2011). Interestingly the grid fields were somewhat larger.

Hence, an alternate model was proposed to explain the contribution of h-current to grid fields, based on the observation that temporal integration is reduced by h-current (Mehta, 2011). This
model explains the increased grid field size (Giocomo et al., 2011), and place field size (Hussaini et al., 2011) in HCN1 knockout mice, as well as the dorso-ventral gradient of grid field sizes in normal mice, all field without invoking theta-interference. This temporal-integration based model also explains the intact dorso-ventral gradient of grid field sizes in primates regardless of the presence or absence of theta rhythm (Killian et al., 2012). Furthermore, the model (Mehta, 2011) additionally predicts enhanced NMDAR-mediated LTP in HCN1 knockout mice, which results in greater experiential plasticity of grid fields during sequential tasks, including greater asymmetric ramping shape of place fields and better phase precession than in wild type mice. These predictions too have been recently confirmed (Eggink et al., 2014). The better phase precession in HCN1 knockout mice can enhance the precise spike-timing needed for induction STDP (Mehta et al., 2002, Mehta, 2011) to improve performance on sequence learning tasks such as spatial navigation, which too has been observed (Nolan et al., 2004). Recent modeling studies suggests that STDP could also play a role in the formation of grid pattern (Widloski and Fiete, 2014).

Although these studies elucidate the mechanisms by which synaptic plasticity could play a role in spatial map plasticity in the hippocampal formation, many exciting questions remain unsolved. For example, how the environmental and biophysical mechanisms interact to form of grid fields, place fields and head direction selectivity are still not fully understood. The intrinsic mechanisms of persistent activity, frequently found in the entorhinal cortex in vitro (Egorov et al., 2002; Hasselmo, 2008) and in vivo (Hahn et al., 2012) could drive hippocampal responses and shape entorhinal-hippocampal response properties. It is hypothesized that persistent activity could facilitate rapid induction of synaptic plasticity by consistent pairing of multisensory inputs and locomotion cues to generate a diversity of episodic neural codes including place-code (O’Keefe and Dostrovsky, 1971), time-cells (Pastalkova et al., 2008; MacDonald et al., 2011), head-direction code (Acharya et al., 2015), and disto-code in hippocampus (Ravassard et al., 2013; Aghajani et al., 2014) and entorhinal cortex (Derdikman et al., 2009). Advances in virtual reality techniques (Holscher et al., 2009; Harvey et al., 2009; Chen et al., 2013; Ravassard et al., 2013; Aghajani et al., 2014; Aronov and Tank, 2014) make it possible to measure hippocampal responses directly during multisensory virtual navigation tasks (Cushman et al., 2013) to determine the contribution of hippocampal map plasticity to sequence learning.

REFERENCES


Hippocampus
PLACE FIELD PLASTICITY


Hippocampus


