

Memory integration constructs maps of space, time, and concepts

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Recent evidence demonstrates that new events are learned in the context of their relationships to existing memories. Within the hippocampus and medial prefrontal cortex, related memories are represented by integrated codes that connect events experienced at different times and places. Integrated codes form the basis of spatial, temporal, and conceptual maps of experience. These maps represent information that goes beyond direct experience and support generalization behaviors that require knowledge be used in new ways. The degree to which an individual memory is integrated into a coherent map is determined by its spatial, temporal, and conceptual proximity to existing knowledge. Integration is observed over a wide range of scales, suggesting that memories contain information about both broad and fine-grained contexts.

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Introduction

Although episodic memory research often focuses on representation of discrete events, memories can extend beyond direct experience by connecting information encountered at different times or places [1,2]. For instance, learning your way around a new town involves gradually learning about landmarks and their relative positions. You might learn about one set of landmarks on one trip through the town and an overlapping set on a different trip. An effective way to represent the paths

taken on both trips is through a common map that represents the relationships between landmarks experienced at different times. A similar process can support forming maps of social relationships [3,4]; for example, meeting a woman and her son, then later meeting the woman's husband, one can infer that the man and the boy are father and son.

These types of knowledge structures, known as cognitive maps [5], are thought to rely on memory integration. Memory integration is a dynamic process wherein new events interact with existing knowledge [1]. For memory integration to occur, a new event that overlaps with prior experience (e.g., meeting the husband) must trigger reactivation of a prior related episode (initially meeting the woman and her son); new information may then be integrated into the reactivated memory (connecting the boy with his father in memory) [1,6]. As a result of integration, the two events are represented with overlapping neural codes, in which elements common to both events act as nodes linking the two memories [7,8]. Such links allow representation of information beyond direct experience and can support flexible behaviors [1], such as taking a shortcut between places in a new town.

Memory integration is thought to be supported by bidirectional interactions between hippocampus and medial prefrontal cortex (mPFC) [9,10]. Reactivation of related memories during new events is mediated by hippocampal pattern completion processes that allow for reinstatement of entire memory traces from overlapping input [11,12]. Medial PFC, which is thought to represent mental models that guide behavior [13,14], may further bias hippocampal pattern completion to the most relevant prior knowledge [9]. Hippocampus then signals deviations between current events and reactivated memories, triggering memory updating [1,15,16]. Finally, new content is integrated with existing mental models via hippocampal — mPFC interactions [17–19]. During integration, mPFC biases hippocampal encoding processes to emphasize representation of features common to multiple events [20,21], resulting in cognitive maps that use overlapping neural codes to represent the relationships among discrete learning episodes [22,23]. Here, we review recent evidence that hippocampal — mPFC memory integration mechanisms support the formation of cognitive maps across different domains of experience, including space, time, and concepts. Hippocampus and mPFC may organize information about spatial, temporal,

and conceptual relationships in similar ways, allowing for the formation of flexible knowledge about many different aspects of the world. Furthermore, memory integration occurs simultaneously at different spatial, temporal, and conceptual scales, which may facilitate learning about both detailed features and broader contextual attributes that are shared across events.

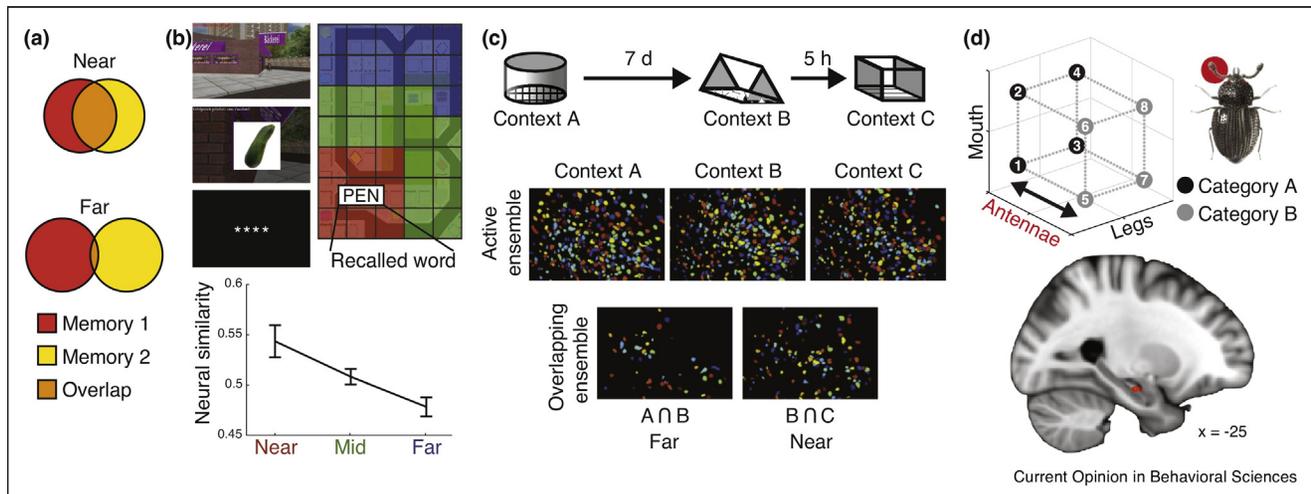
Integration of spatial experience

The seminal work of Tolman first showed that memories of recently traveled routes are combined with memories of previously traveled routes to create an integrated map of the environment [5]. More recent evidence indicates that representations of past and present spatial trajectories are simultaneously active within hippocampus during learning [24–26], providing an opportunity for links to be formed among different spatial experiences in an environment. Reinstatement of behaviorally relevant memories within hippocampus during spatial learning may be particularly dependent on top-down influence from mPFC [10]. When rodents make decisions that depend on spatial context, mPFC drives hippocampal responses [9^{*}], suggesting that mPFC coordinates reinstatement of hippocampal memory representations that are most relevant to ongoing experience [27]. Increased hippocampal—mPFC coupling is also observed during spatial memory retrieval in humans [28,29], and such coupling

has been further shown to support individuals' ability to connect past and present experience [17,18,30].

Memory integration mechanisms may cause events (e.g., encountering landmarks) experienced at different times within the same spatial environment to be represented by overlapping neural responses within hippocampus. Evidence from rodents has revealed highly structured hippocampal population codes, wherein responses are similar for objects that share spatial context or position information [33] as well as for locations that share relationships [34]. To represent cognitive maps, the similarity between hippocampal representations of events experienced within an environment should scale with their distance from one another in the environment. Events experienced close together in space should be represented by highly overlapping hippocampal populations, while events experienced in more distant locations should evoke less similar hippocampal responses (Figure 1a). Human electrophysiological data shows that when participants recall objects learned within a virtual environment, hippocampal place-cell activity is highest for locations closest to an object's learned location, dropping off with increasing distance from the object (Figure 1b) [31]. Moreover, human neuroimaging data indicate that hippocampal patterns for individual objects experienced in different locations within a virtual town scale with

Figure 1



(a) The overlap in representations of different memories reflects the distance between their spatial, temporal, and conceptual attributes, with memories nearer to one another resulting in more overlapping representations. (b) Ensemble place-cell activity in human hippocampus reflects the spatial distance between memories. Participants learned the locations of objects in a virtual environment, then verbally recalled the objects. Ensemble activity during object recall was most similar to place-cell activity for locations near to where the object was learned. Adapted from Ref. [31]. (c) In mice, memories of different cage contexts are represented by overlapping neural ensembles in hippocampal subfield CA₁ when exposure to the cages is separated by a shorter delay of 5 hours, but do not show significant overlap when separated by a longer interval of 7 days. Adapted from Ref. [32^{**}]. (d) Model-based fMRI data in humans demonstrates that item representations within hippocampus reflect conceptual similarity. The representation of three distinct features of categorized items was estimated based on patterns of activity. After learning, hippocampal representations emphasize features that are diagnostic for the category, causing items from the same conceptual category (e.g., the beetles with thin antennae represented by black dots) to become representationally more similar, and dissimilar to items from a different category (the beetles with thick antennae represented by the gray dots). Adapted from Ref. [20^{*}].

individual's subjective memory for their distance in space [35[•]]; objects judged as more spatially proximal evoke more similar hippocampal activation patterns after learning.

Collectively, these findings indicate that reactivation of prior spatial experience during new learning results in the formation of integrated spatial maps within the hippocampus. Importantly, such maps include information not only about directly traveled paths, but also routes through the environment that can be inferred from relationships among those traveled [36–38]. When directly experienced routes to a goal are blocked, the integrated hippocampal map can thus support flexible navigation via novel, alternate paths [39–41]. The mPFC may resolve interference by biasing the hippocampus to retrieve the behaviorally relevant route [42,43]. Reactivation of integrated spatial maps also facilitates new learning. Work in rodents has shown that integrated spatial maps provide a framework for rapid learning of new object-location associations in the same environment [44,45]. Reactivation of hippocampal spatial maps also facilitates learning in new environments that share properties with existing maps [24], thus allowing for generalization of spatial experience to novel contexts.

The physical similarity between environments or events within an environment may be an important boundary condition that determines which events are combined into a coherent cognitive map of space. A new spatial experience that shares many features with an existing memory may be more likely to trigger reactivation of the related memory than an event that shares fewer overlapping features. As a result, increased physical similarity among spatial experiences may be associated with an increased likelihood of integration. Consistent with this idea, population responses within the CA₁ subfield of the rodent hippocampus scale linearly with the number of spatial features shared across environments [46]; the greater the number of shared features, the greater the overlap in the CA₁ population response. However, the integration boundaries may be different across hippocampal subfields [47]. In contrast to CA₁, CA₃ population responses overlap only when environments are highly similar, suggesting that the CA₃ subfield has a higher threshold for integration [46]. One possibility for future study is to explore whether different hippocampal subregions support different levels of memory integration that represent the similarities among events in terms of whether they share fine details or global characteristics.

Representation of temporal context through memory integration

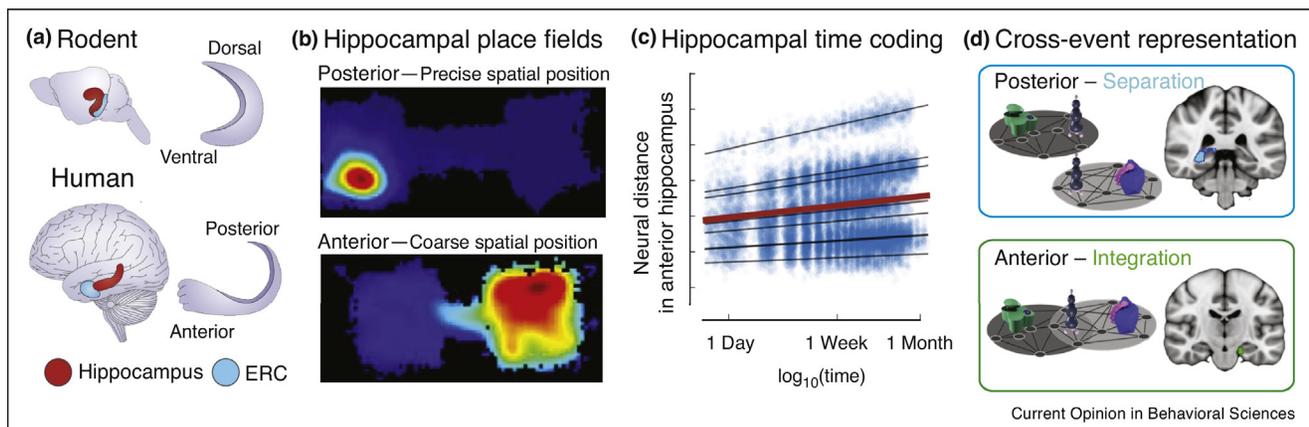
Similar to maps of space, memory integration in hippocampal and mPFC networks may code the temporal context of events by representing temporally proximal events with overlapping neural ensembles (Figure 1c). Lesions to the hippocampus in rodents and humans

impair the ability to extract temporal regularities from the environment and remember the sequential order of events [48,49], suggesting a critical role for the hippocampus in temporal integration. Recent evidence suggests that temporal proximity affects memory integration at a range of timescales, which involve distinct mechanisms [50]. Memory integration at time scales on the order of seconds may be organized by 'time cells' in the hippocampus and mPFC, which respond at specific temporal intervals during task performance [51,52]. Computational modeling suggests that, like place cells, time cells may provide a map for organizing sequences of memories [53]. Integration of new memories into an existing temporal map may facilitate learning about the relative timing of overlapping sequences of events [54]. For example, if a musician learns to play a song, then later learns a new introduction to the song, memories of these subsequences may be integrated to form a complete song timeline.

Human neuroimaging data has shown that hippocampal [55–57] and mPFC [58] activation patterns reflect the temporal organization of events that occur seconds apart. In these studies, hippocampal and mPFC activation patterns are more similar for items that are seen in immediate succession to one another [55,56,59]. Similar to maps of space, the degree of similarity in these regions scales with the temporal distance among items [35[•],56,58] and, in the case of hippocampus, is related to individuals' subjective estimates of temporal proximity [35[•],56]. In rodents, the temporal context of individual items in a sequence can be decoded from CA₁ ensembles, and sequence coding at the single cell and ensemble level is correlated with sequence memory [60[•]]. Neuroimaging has also revealed coding of sequential position information within human hippocampus that relates to memory performance [61].

Notably, integration of memories at longer timescales, on the order of minutes [62] as well as days and weeks (Figure 2c) [63], is also observed within hippocampus. Memory integration across longer temporal intervals may result from a different mechanism than that supported by time cells. Temporal proximity at longer time scales may promote integration through a memory tagging and allocation mechanism, whereby neurons and synapses recruited to represent a recent episode are more readily reactivated during new events that occur within hours of the original episode [32^{••},50,64]. Recruitment of the same neural ensembles through tagging and allocation would thus result in overlapping population codes for events that occur within minutes and hours of one another as observed in CA₁ of rodents (Figure 1c) [32^{••}] and CA₁ and mPFC in humans [16,22^{••}]. Similarly, neurogenesis in the dentate gyrus of hippocampus may affect memory allocation on longer scales of days and weeks, linking memories encoded at similar times while avoiding

Figure 2



(a) Illustration of the hippocampal long axis in rodent and human. Ventral and dorsal hippocampus in rodent correspond to anterior and posterior hippocampus, respectively, in human. ERC: entorhinal cortex. Adapted from Ref. [78]. (b) Hippocampal place fields differ in size along the hippocampal long axis; place fields in anterior hippocampus are larger than those observed in posterior hippocampus. Smaller place fields in posterior hippocampus may support coding of detailed event information (such as precise spatial position), whereas larger place fields in anterior hippocampus may integrate information across spatial positions to support coding of global context information. Adapted from Ref. [79]. (c) Anterior hippocampus represents memories for real-world events more similarly when they occur closer together in time. The temporal organization of memories within anterior hippocampus is even apparent at long time scales, such as one week and one month. Adapted from Ref. [63]. (d) Memory elements that share a common association with an overlapping item are represented differently within anterior and posterior hippocampus. Anterior hippocampal activation patterns become more similar for elements that share relationships, reflecting integration of overlapping events. In contrast, posterior hippocampal activation patterns for those same elements become more dissimilar after learning, reflecting the formation of orthogonal representations for overlapping events. Adapted from Ref. [22**].

interference between temporally distant events in memory [65]. Through these different mechanisms, hippocampus and mPFC may come to represent the temporal context of individual events at different time scales, from fine (e.g., position in a short sequence of events) to broad (e.g., whether an event happened one week or four weeks ago).

In addition to providing a map of temporal context, integration may support flexible generalization of information across episodes that share temporal context. For instance, a study examining hippocampal ensemble responses in rodents found that memories that are encoded close together in time (within 5 hours) are represented with overlapping populations of CA₁ neurons, with the amount of overlap predicting the degree of fear responding that generalized across the two memories (Figure 1c) [32**]. However, there may be important boundary conditions on the degree of integration and generalization observed across longer time intervals. Although reactivation of related memories has been found to promote integration [17], a recent study found that memories encoded on different days were less likely to be integrated compared to memories encoded on the same day, even when controlling for the amount of reactivation [66]. Future work is needed to determine precisely how different mechanisms for temporal coding facilitate generalization of information across episodes.

Integrated maps of conceptual space

Emerging evidence suggests that the same mechanisms that support integration of spatial and temporal information may also play a key role in our ability to acquire concepts [67]. Concepts organize our experiences by highlighting shared features and allow for meaningful generalizations in novel situations. Acquiring new concepts requires extracting information across multiple individual learning experiences to learn both what features are common to concept exemplars and what features differentiate between concepts. As a result of learning, concept representations emphasize diagnostic features for a given concept, rather than the overall perceptual similarity of individual stimuli (Figure 1d). When presented with a stimulus to be assigned to a concept, hippocampal — mPFC interactions may trigger reactivation of similar, concept-relevant learning experiences. Medial PFC also influences hippocampal encoding by compressing memory representations to emphasize features that capture commonalities across events [21], consistent with its hypothesized role in forming mental models of latent structures that are not directly observable [13,14,68]. The retrieved memories may then be used to predict a concept label for the current experience. The outcome of that prediction may then lead to integration of the new exemplar into an existing concept or formation of a new concept representation. In this way, learning concepts requires similar computations and

representations as learning about the spatial and temporal regularities of the environment [67].

Electrophysiological data from human patients indicates that hippocampal neurons show high selectivity to the conceptual rather than the perceptual features of events [69]. Moreover, hippocampal responses scale with the conceptual novelty of experiences [70,71], and such signals may be critical to concept formation and updating [72,73]. Further evidence for the role of the hippocampus and mPFC in formation of conceptual maps comes from a set of recent studies that use representational approaches to neuroimaging data analysis [20*,74*,75]. These studies have shown that the similarity of hippocampal [20*,75] and mPFC [74*] activation patterns between individual concept exemplars scales with their distance in a learned conceptual space; exemplars of the same concept evoke more similar activation patterns in hippocampus and mPFC and are distinct from exemplars of different concepts (Figure 1d). In one case, the conceptual organization observed in medial temporal lobe and mPFC regions followed a grid-like organization that is commonly observed during spatial navigation and imagery [74*], providing a direct link between spatial and conceptual representation within these regions. Additional evidence suggests that hippocampal — mPFC interactions are critical when conceptual knowledge is applied to new situations with the same underlying conceptual structure [9*,76]. Collectively, these results indicate that individual memories are integrated through hippocampal — mPFC interactions to create conceptual knowledge that supports flexible decisions when confronted with new stimuli or environments.

The scale of integration differs between hippocampal subregions

The evidence reviewed thus far indicates that integration of spatial, temporal, and conceptual relationships may occur at a variety of scales from fine-grained to broad. Different scales of integration may be linked to the function of different hippocampal subregions. For instance, the spatial remapping properties of CA₁ and CA₃ noted above [46] suggest that CA₁ may integrate information across a broader set of contexts than CA₃, which forms overlapping representations for only highly similar environments. Computational modeling further indicates that CA₁, but not CA₃, representations support extraction of temporal and associative regularities from the environment and enable inference about the relationships among episodes [47]. Human and rodent data have also exclusively linked memory integration processes to CA₁ [16,32**]. The connectivity of CA₁ may explain its privileged role in memory integration. It receives input about incoming sensory information from entorhinal cortex simultaneously with memory-driven expectations derived from CA₃ pattern completion processes [77]; networks of CA₁ cells that represent new content and

existing memories may therefore become linked due to their coactivation within CA₁.

Differences in representational scaling have also been observed along the long axis (ventral–dorsal in rodents, anterior–posterior in humans) of hippocampus (Figure 2a) [22**,23]. Place cells in anterior hippocampus have larger receptive fields than in posterior hippocampus (Figure 2b) [78–80]. Place field gradients along the hippocampus axis may thus support integration of experiences at different spatial, temporal, and conceptual scales. For instance, large place fields in anterior hippocampus are well suited for integrating events that occur at different positions within the same environment, while smaller place fields in posterior hippocampus allow distinct coding of events experienced in the same environment [81]. It is possible that temporal representations may vary in scale along the hippocampal long axis in a similar manner to spatial representations, with larger timescales represented in anterior hippocampus and smaller timescales in posterior hippocampus [82]. Consistent with this hypothesis, anterior hippocampus is more likely to demonstrate integrated codes for overlapping events that are separated by long time intervals than posterior hippocampus (Figure 2c, d) [22**,23,63]. The different representational capacities of anterior and posterior hippocampus may thus simultaneously support behaviors that rely on different levels of memory detail [83], with posterior hippocampus supporting fine judgments relating individual memory elements and anterior hippocampus supporting abstractions across broader experiential scales. There is also evidence that anterior and posterior mPFC subregions exhibit different levels of integration that mirror functional differences along the hippocampal long axis [22**], but more work is needed to determine whether human mPFC subregions are functionally distinct.

Conclusions

Much of our knowledge is derived by forming links across the individual events we experience. The findings reviewed here indicate that structured knowledge about the spatial, temporal, and conceptual relationships among events arises from memory integration processes supported by the hippocampus and mPFC. By forming overlapping memory codes that integrate information acquired at different times and places, our memories extend beyond what we directly experience and can be deployed flexibly to support behavior in new situations.

However, several open questions remain for future research on memory integration. For instance, what are the boundary conditions that determine when representations of overlapping memories are integrated rather than separated, and how might we manipulate the formation of integrated codes of experience? While initial findings suggest that the strength of existing knowledge [18], the degree of memory reactivation during encoding

[17], the magnitude of memory-based prediction errors [15], and task demands [84] all impact the likelihood of integration, more work is needed to understand the full complement of conditions that influence how we represent related episodes in memory. As noted above, different memory circuits (e.g., anterior and posterior hippocampus) may integrate information at different scales. It remains to be tested whether these circuits play functionally different roles in memory, with fine-scaled representation supporting decisions about relationships among detailed event elements and integration at broader scales supporting more abstract knowledge about the conceptual relationships among events. The fact that hippocampus and mPFC continue to mature through the third decade of life [85,86] further suggests that memory integration may have a prolonged developmental timecourse that extends through adolescence, with corresponding impacts on spatial, temporal, and conceptual learning abilities. In reviewing the current state of knowledge on memory integration, we hope to inspire future work on the rich set of questions that remain.

Conflict of interest statement

Nothing declared.

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