Top-down modulation of hippocampal encoding activity as measured by high-resolution functional MRI

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1. Introduction

A wealth of evidence indicates that our memories are affected not only by available sensory information, but also the cognitive operations performed on this information. As such, top-down processing is thought to play an important role in influencing what is encoded into memory. Prior studies have implicated prefrontal and parietal regions in mediating such goal-directed control and attentional resources during learning (e.g., Blumenfeld & Ranganath, 2007; Chun & Turk-Browne, 2007; Uncapher & Rugg, 2009). It remains unclear, however, how encoding processes in the medial temporal lobe (MTL) – a region known to play a critical role in episodic memory formation – are modulated by top-down influences.

Early efforts to address this issue focused on how selective attention modulates activity in domain-specific MTL cortex, demonstrating enhanced encoding activity when preferred-category features were selectively attended (Gazzaley, Cooney, McEvoy, Knight, & D’Esposito, 2005; Yi & Chun, 2005). More recently, Uncapher and Rugg (2009) reported subsequent memory effects in the hippocampus (i.e., greater activity for subsequent hits than misses) for features that were selectively attended during learning but not for unattended features. Studies investigating modulatory effects of deep vs. shallow processing, however, have not demonstrated differences in MTL subsequent memory effects according to task demands (Fletcher, Stephenson, Carpenter, Donovan, & Bullmore, 2003; Schott et al., 2013). As such, reports of top-down influences on MTL encoding activity are mixed. Furthermore, the degree to which subregions of the MTL may be differentially affected by top-down influences remains unclear.

To address these questions, we evaluated MTL activity using high-resolution functional MRI (hr-fMRI) as participants engaged in encoding tasks focusing attention on either stimulus distinctiveness or commonalities across stimuli. The phenomenon of better memory for distinctive than common events is well-documented. With respect to word frequency, for example, Glanzer and Adams (1990) found both higher hit rates and lower false alarm rates for rare words like “kumquat” relative to common...
words like “apple”. Similarly, enhanced memory performance has been shown for items within a list that differ from the list context, such as a number within a list of syllables (e.g., Hunt, 1995). Of direct relevance to the current study, researchers have also shown that instructions to participants during learning emphasizing stimulus distinctiveness lead to higher rates of subsequent recollection than those emphasizing similarities among stimuli (Mäntylä, 1997). Taken together, such findings suggest that encoding the distinctive features of an item enhances subsequent memory performance relative to encoding commonalities across items. Given the key role of the MTL in encoding episodic memories, the aim of the current study was to assess whether and how MTL encoding processes are influenced by such differences in task demands at encoding.

Using a modified version of the paradigm employed by Mäntylä (1997), we scanned participants as they attended to either stimulus distinctiveness or similarities across stimuli. We hypothesized that attending to distinctiveness would more effectively engage encoding processes within the hippocampal circuit than attending to similarities across stimuli. Specifically, we predicted that encoding memories with differing degrees of detail would lead to an interaction between encoding task and subsequent memory performance such that subsequent memory effects in the hippocampus would be greater for items encountered in the Distinctiveness task than in the Similarity task. Critically, our use of hr-fMRI allowed us to evaluate whether potential task-related differences in subsequent memory effects were subfield-specific or evident across the entirety of the hippocampal circuit. Prior hr-fMRI studies suggest that input structures of the hippocampal circuit, namely the dentate gyrus and CA2/3 region (DG/CA2/3), are selectively active during encoding and show a bias toward pattern separation, whereas output structures (CA1 and the subiculum) have been shown to be differentially active during retrieval and pattern completion (Bakker, Kirwan, Miller, & Stark, 2008; Carr, Viskontas, Engel, & Knowlton, 2010b; Eldridge, Engel, Zeineh, Bookheimer, & Knowlton, 2005; Olsen et al., 2009; Viskontas, Carr, 2003). Other hr-fMRI data, however, have demonstrated hippocampal encoding or novelty effects not just in DG/CA2/3 but across all subfields of the hippocampus (Chen, Olsen, Preston, Glover, & Wagner, 2011; Duncan, Ket, Inati, & Davachi, 2012; Preston et al., 2010; Suthana, Ekstrom, Moshirvaziri, Knowlton, & Bookheimer, 2009; Wolosin, Zeithamova, & Preston, 2012; Zeineh, Engel, & Bookheimer, 2000), and are thus more consistent with extant models of hippocampal function which vest encoding and retrieval mechanisms within the same hippocampal subfields (e.g., O’Reilly and Rudy, 2001; Rolls, 1996). Given these latter findings, we hypothesized that attention to distinctiveness would lead to more effective hippocampal encoding than attention to similarity, as evidenced by a task by subsequent memory interaction across the entirety of the hippocampus.

2. Materials and methods

Having modified the encoding tasks originally developed by Mäntylä (1997), we first conducted a behavioral experiment (Experiment 1) to ensure that attending to stimulus distinctiveness at encoding led to greater subsequent memory performance than attending to similarities across stimuli. A separate hr-fMRI experiment (Experiment 2) was then conducted in a different group of participants.

2.1. Experiment 1

2.1.1. Participants

Thirty two healthy young adults (24 female; 18–29 years of age, mean age = 20.87 ± 2.78 years) participated in Experiment 1 and were paid $8/h for their participation. All participants were fluent English speakers. The study was performed under a protocol approved by the UCLA Office for Protection of Research Participants.

2.1.2. Materials

All stimuli were presented via a Macintosh laptop computer using the MATLAB Psychophysics Toolbox. Stimuli consisted of 99 black and white photographs of young adults, 45 male and 44 female, with neutral facial expressions (Psychological Image Collection at Stirling, http://pics.psych.stir.ac.uk). Faces were divided into three sets of 33, such that 33 served as stimuli for the Distinctiveness task, 33 for the Similarity task, and the remaining 33 faces served as lures for a surprise recognition test. Sets were counterbalanced across participants such that each set was assigned to the distinctiveness, similarity, or lure condition an equal number of times.

2.1.3. Procedure

Having been recruited to perform a study of “face processing,” participants were unaware of any memory-related components of the experiment until performing a surprise recognition test. For the Distinctiveness, or “Taxi Driver”, task (Fig. 1A), participants were told to imagine that they were chauffeurs picking up specific customers from the airport. As each customer’s photo appeared on the screen, participants were asked to make a subjective judgment regarding how easy it would be to find that particular customer in the crowd on a scale from 1 to 4, ranging from 1—hard (customer has an average face; will blend in with the crowd), to 4—easy (customer is very unique looking; will stand out of the crowd). In the Similarity, or “Casting Director”, task (Fig. 1B), participants were told to imagine themselves as casting directors whose job it was to audition the lead male and female roles for a new film. They were informed that the director desired a particular look for each role, and that they should select actors matching this look for their current audition. To this end, participants were first shown a sample male and female face fitting the desired look, and were then asked to rate a series of photos according to how well each actor matched the desired look on a scale from 1 to 4, ranging from 1—poor match to 4—excellent match.

The study employed a within-subjects design such that all participants completed both the Distinctiveness and Similarity tasks prior to performing the recognition memory test. Participants received instructions for both encoding tasks prior to beginning the experiment, and task order was counterbalanced across participants. For both tasks, trials lasted 6 s each, beginning with a brief fixation cross (0.75 s) followed by presentation of the face for 5.25 s. Participants were asked to rate each face across each task independently, to task instructions within this 5.25 s period. Following completion of the second encoding task, participants performed a difficult maze for 10 min followed by a surprise recognition test inclusive of faces from both encoding tasks. Participants first made studied/new judgments followed by remember/know judgments for those faces deemed studied. Participants were told to give a “remember” response when they remembered contextual details pertaining to when they originally viewed the face, and to give a “know” response when they confidently recognized a face in the absence of such details. This test was used to assess the extent to which successful recognition was based on episodic recollection or feelings of familiarity.

Results from each participant’s recognition test were used to sort trials into hits (studied faces that were successfully recognized), misses (studied faces that were forgotten), correct rejections (lure faces that were correctly judged as new) and false alarms (lure faces incorrectly judged as old). Hits and false alarms were further broken down according to “remember” and “know” responses. To examine whether task demands at encoding influenced subsequent memory performance, we conducted a 2-way repeated measures ANOVA examining hit rates for the two encoding tasks (Distinctiveness, Similarity) broken down according to memory quality (remember, know). Note that, because the recognition test included faces from both encoding tasks, participants first made studied/new judgments followed by remember/know judgments for those faces deemed studied. Participants were told to give a “remember” response when they remembered contextual details pertaining to when they originally viewed the face, and to give a “know” response when they confidently recognized a face in the absence of such details. This test was used to assess the extent to which successful recognition was based on episodic recollection or feelings of familiarity.

2.2. Experiment 2

2.2.1. Participants

Fifteen healthy, right handed individuals (8 female; 18–34 years of age, mean age = 23.67 ± 2.14 years) participated in Experiment 2. Imaging data from one participant was excluded due to excessive motion. Participants were fluent English speakers, and were paid $25/h for participation in the study. The study was performed under a protocol approved by the UCLA Office for Protection of Research Participants.

2.2.2. Materials

Stimuli were presented using the MATLAB Psychophysics Toolbox and were viewed via magnet-compatible goggles placed directly over the eyes (Resonance Technologies, Inc). Face stimuli were drawn from the same database used in Experiment 1, but a larger number of stimuli was used in Experiment 2 to allow for sufficient trial numbers for imaging analyses. To this end, stimuli consisted of 150 faces divided into three sets of 50, such that 50 served as stimuli for the
Distinctiveness task, 50 for the Similarity task, and the remaining 50 faces served as lures for the surprise recognition test. Sets were counterbalanced across participants such that each set was assigned to the distinctiveness, similarity, or lure condition an equal number of times. Instructions to participants were identical to those of Experiment 1.

2.2.3. Procedure
As in Experiment 1, Experiment 2 employed a within-subjects design such that all participants completed both the Distinctiveness and Similarity tasks prior to the surprise recognition memory test, with task order counterbalanced across participants. Each task consisted of 50 face trials with odd-even baseline trials interspersed. During low-level tasks like passive fixation, participants may spontaneously engage in mnemonic processing, leading to hippocampal activation. The odd-even task has been shown to minimize hippocampal activity in comparison to passive fixation (Stark & Squire, 2001), and thus was chosen as the baseline task for the current study.

Each of the two encoding tasks was divided into two shorter runs of 25 face trials, such that participants performed both runs of one task followed by both runs of the other. Face trials lasted 6 s each, beginning with a brief fixation cross (0.75 s) followed by presentation of the face for 5.25 s. Baseline trials were of variable duration, ranging from 3 to 18 s in multiples of 3 s, and consisted of a series of single digits presented for 600 ms each, with participants responding whether digits were odd or even. The order of face and baseline trials within each run was determined using a genetic algorithm (Wager & Nichols, 2003). Approximately 10 min following the last encoding run, participants performed a surprise recognition test inclusive of faces from both encoding tasks in which they made studied/unstudied judgments.

2.2.4. FMRI Data acquisition and pre-processing
Structural and functional imaging for Experiment 2 were performed using a 3 T Siemens Allegra scanner. Structural images included (1) a sagittal localizer to identify the long axis of the hippocampus, (2) high-resolution T2 hippocampal images perpendicular to the long axis of the hippocampus, and (3) high-resolution gradient EPI sequences coplanar with the functional images. Functional imaging of the MTL was conducted with high-resolution gradient echo EPI sequences, consisting of 19 slices perpendicular to the long axis of the hippocampus. Pre-processing was performed using the FSL toolbox. Skulls were stripped using the Brain Extraction Tool (Smith, 2002), and functional images were realigned using FMRIB to compensate for small head movement (Jenkinson, Bannister, Brady, & Smith, 2002). For participants with translational motion over 1 mm, images were denoised using MELODIC (Beckmann & Smith, 2004). One subject displayed motion over 3 mm, and thus was not included in any imaging analyses. Data were filtered with a high-pass cutoff of 75 s, and were smoothed using a 2 mm FWHM smoothing kernel.

2.2.5. Response classification
To examine the degree to which MTL encoding activity was modulated by goal state, we used results from each participant’s recognition test to back sort encoding trials into those that were subsequently recognized (hits) and those that were subsequently forgotten (misses). Importantly, this allowed for comparison of activity associated with successful memory formation in each of the two tasks in each region of interest. Furthermore, given that behavioral analyses for Experiment 2 revealed no significant difference in overall hit rate across the two tasks, subsequent memory analyses were not biased by unequal trial numbers in the two tasks. Finally, remember/know data from the recognition test were used to examine whether quality of memory for subsequent hits differed between the two tasks. Note, however, that there were insufficient trial numbers to perform additional analyses comparing subsequent remember and know hits for each task (specifically, 10 participants had fewer than 10 trials for one or more relevant conditions).

2.2.6. Regions of interest and time course analyses
Anatomically defined regions of interest (ROIs) were created for each individual participant. Specifically, ROIs for hippocampal subfields (DG/CA3, CA1, and subiculum) and MTL cortical areas (entorhinal, perirhinal and parahippocampal) were defined in each participant according to landmarks visible on his or her high-resolution structural scan. The DG and CA1 fields are difficult to unambiguously separate given current functional resolution, and therefore these regions were collapsed into a single ROI (DG/CA3). Boundary definitions were guided by MTL atlases (Duzel & Tewani, 2005; Insuastri & Amaral, 2004) and specifications defined in structural MRI studies of the MTL (Insuastri et al., 1998; Pruessner et al., 2002, 2000). Furthermore, because susceptibility artifacts may be seen in anterior regions of the MTL during EPI sequences, voxels that were not visible in the mean functional image were excluded from ROI analyses (Zeineh et al., 2000). All ROIs were prepared by the same observer (VAC) to maintain consistency.

Group timecourses were created using the summary statistics approach to the mixed effects model (Mumford & Nichols, 2006). First, timecourses for each condition were extracted from each anatomical ROI in each participant using a Finite Impulse Response (FIR) model, allowing for examination of differences across
conditions in timecourse amplitude as well as shape. The design matrix in this model contained entries for subsequent hits and misses; odd-even baseline trials were left un-modeled. The underlying hemodynamic response for each trial type was estimated by averaging the signal at 3 s bins beginning 3 s prior to stimulus onset and ending 15 s after stimulus onset. For a given task, weighted least squares was used to combine activation across the two encoding runs. Finally, timecourses across all participants were averaged to create a group timecourse for each ROI.

To evaluate subsequent memory effects in the MTL under differing goal states, we conducted a 3-way repeated measures ANOVA examining encoding activity related to encoding task (Distinctiveness, Similarity), subsequent memory performance (hit, miss), and timecourse bin (3–6 s, 9–12 s). The latter factor was included to allow for examination of differences in timecourse shape, given that a number of studies have shown that shape can vary considerably across regions (see, e.g., Buckner, Rountaal, Schacter, Wagner, & Rosen, 1998). Specifically, we were interested in whether subsequent memory effects would be seen (a) early in the timecourse, (b) late in the timecourse, or (c) throughout the entirety of the timecourse. Based upon a generic timecourse created by averaging across all participants, ROIs, and conditions, we selected 3–6 s to represent early timecourse activity, and 9–12 s to represent late timecourse activity. Finally, planned comparisons were conducted using paired, two-tailed t-tests (alpha level .05).

3. Results

3.1. Experiment 1

3.1.1. Encoding task ratings

To verify that participants engaged in different encoding processes under differing goal states, we compared response rates along each task’s respective 4-point scale. A repeated measures ANOVA revealed a significant main effect of response option (1 through 4) (F(1,31) = 34.05, p < .001) but no effect of task (Distinctiveness, Similarity) (p > .3). A significant two-way interaction between encoding task and response option (F(3,93) = 36.23, p < .001) was found such that participants’ patterns of responding differed between the two encoding tasks (Fig. 2). When judging how easy it would be to pick a given customer from the crowd in the Distinctiveness task, participants’ responses clustered around 2 (fairly hard) and 3 (fairly easy), with few faces rated at the extremes of the scale. When asked how well each actor matched the desired look in the Similarity task, participants’ responses clustered around 1 (poor) and 2 (average) for approximately 75% of trials, with few faces rated at the upper end of the scale. Such response patterns suggest that subjects processed faces differently in the Distinctiveness and Similarity tasks.

3.1.2. Effect of goal on subsequent memory

To determine whether the two encoding strategies led to differences in subsequent episodic recollection, a repeated measures ANOVA examining the relationship between encoding task (Distinctiveness, Similarity) and quality of memory (remember, know) was performed. This analysis revealed a significant main effect of task (F(1,31) = 4.70, p = .038) such that, collapsed across memory quality, the hit rate for the Distinctiveness task was higher than that of the Similarity task. No main effect of memory quality was found (p > .3). Analyses further revealed a significant interaction between task and memory quality (F(1,31) = 5.07, p = .032) such that a higher percentage of remember hits was found for the Distinctiveness than Similarity task (t(31) = 2.92, p = .007); there was no significant difference in the percentage of know hits for the two tasks (p > .5) (Fig. 3). Importantly, the order in which the two tasks were performed did not significantly influence subsequent memory performance (no main effect of task order, p > .5, nor interactions involving task order, p values > .5). As noted in Section 2.1.3, corrected hit rates were not used given that false alarms were not specific to a given task. Nonetheless, we include false alarm rates here for completeness: Participants responded to foils significantly more often with a “Know” response (10.17% ± 0.48%) than a “Remember” response (3.32% ± 0.23%) (t(31) = 5.44, p < .001).

To further explore the relationship between encoding task and subsequent memory, we next examined the extent to which Distinctiveness and Similarity ratings at encoding related to subsequent memory performance. For a given encoding task, faces were first binned according to rating and then further subdivided into subsequent remember, know, and miss trials. In this way, the proportion of subsequently recollected faces for each rating bin could be calculated. Given that not all participants used each rating option during encoding—leading to an unbalanced data set, linear mixed modeling was used to evaluate the relationship between encoding task, task ratings, and subsequent rates of recollection. Specifically, this analysis tested whether increasing rates of subsequent recollection were found with increasing Distinctiveness or Similarity ratings. Results revealed a main effect of rating (X^2(1) = 14.758, p < .001) as well as a task x rating interaction (X^2(1) = 6.293, p = .012), such that a linear effect was found for the Distinctiveness task (X^2(1) = 21.644, p < .001) but not the Similarity task (p > .5) (Fig. 4). That is, faces judged as more distinctive at encoding were associated with higher subsequent rates of recollection, whereas in the Similarity task increasing similarity ratings were not associated with higher subsequent rates of recollection.

3.2. Experiment 2

3.2.1. Behavioral results

As in Experiment 1, a repeated measures ANOVA revealed a significant main effect of response option (1 through 4) (F(1,14) = 19.96, p < .001), but no effect of task (Distinctiveness, Similarity) (p > .1). A significant two-way interaction was found between encoding task and response option (F(3,42) = 23.55, p < .001), such that participants’ patterns of responding differed between the two encoding tasks.
encoding tasks in manner similar to that of Experiment 1 (Fig. 2). To determine whether the two encoding strategies led to differences in subsequent episodic recollection as in Experiment 1, a repeated measures ANOVA examining the relationship between encoding task (Distinctiveness, Similarity) and quality of memory (remember, know) was performed. This analysis revealed no main effects of task (p > .1) or memory quality (p > .8), but did reveal a trend toward a significant interaction between task and memory quality (F(1,14) = 3.05, p = .10) such that a higher percentage of remember hits was found for faces studied under the Distinctiveness than Similarity task (t(14) = 2.20, p < .05). There was no significant difference in the percentage of know hits for the two tasks (p > .6) (Fig. 3). As in Experiment 1, participants responded to foils significantly more often with a “Know” response (11.47% ± 1.25%) than a “Remember” response (1.87% ± 0.30%) (t(14) = 3.97, p = .001). No effects of task order on subsequent memory performance were found (no main effect of task order, p > .1, nor interactions involving task order, p values > .5).

Next, mixed modeling was used to examine the extent to which Distinctiveness and Similarity ratings at encoding related to subsequent memory performance. Results revealed no main effect of rating on subsequent rates of recollection (p > .2), nor an interaction between task and rating (p > .4) (Fig. 4).

Finally, to assess whether our fMRI results could be attributed to differences in time on task, encoding trials for each task were back sorted into hits and misses, and reaction times for each condition (see Table 1) were subjected to a repeated measures ANOVA. Results revealed no main effect of task (p > .2) or memory performance (p > .5) on reaction time, nor was there an interaction between task and memory performance (p > .3).

Table 1

<table>
<thead>
<tr>
<th>Condition</th>
<th>Subsequent hit (ms)</th>
<th>Subsequent miss (ms)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Distinctiveness</td>
<td>2798 ± 126</td>
<td>2805 ± 132</td>
</tr>
<tr>
<td>Similarity</td>
<td>2843 ± 149</td>
<td>2959 ± 166</td>
</tr>
</tbody>
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* Reaction times during encoding as a function of task and subsequent memory performance. Note: Data reflect reaction times and standard error of the mean for Experiment 2.

### 3.2.2. Imaging results
A repeated measures ANOVA was conducted in each ROI to examine encoding activity associated with encoding task (Distinctiveness, Similarity), subsequent memory performance (hit, miss), and timecourse bin (3–6 s, 9–12 s). Analyses of DG/CA2/3 and PHC data revealed no main effect of timecourse bin nor any two or three-way interactions involving timecourse bin (DG/CA2/3: p values > .4; PHC: p values > .2); thus, analyses for both regions were calculated using integrated signal change across the entire 3–12 s period. Findings in left DG/CA2/3 revealed a significant main effect of memory performance (F(1,13) = 12.60, p = .004) such that collapsed across task, activity associated with subsequent hits was greater than that for misses (Fig. 5A). However, no significant interaction between encoding task and memory performance (p > .9), nor a main effect of task (p > .3) was seen. Similarly, in right PHC, analyses revealed a main effect of memory performance (F(1,13) = 15.64, p = .002) such that activity associated with subsequent hits was greater than that for misses for both tasks (Fig. 5B). No significant interaction between encoding task and
memory performance ($p > .2$) nor a main effect of task was found ($p > .8$).

A different pattern of results was found in right CA1, which demonstrated no main effects of encoding task, memory performance, or timecourse bin ($p$ values $> .2$), but did reveal a significant three-way interaction between these factors ($F(1,13)=4.99$, $p = .044$). Specifically, no significant main effects or interactions were found early in the timecourse ($3–6$ s; $p$ values $> .1$). Later in the timecourse (9–12 s), however, a two-way interaction was found between task and memory performance ($F(1,13)=7.73$, $p = .016$) such that activity associated with subsequent hits was greater than that for misses in the Distinctiveness task ($t(13)=2.11$, $p = .05$), whereas in the Similarity task activity associated with subsequent misses was numerically but not significantly greater than that for hits ($p > .2$) (Fig. 6A). Stated more succinctly, late in the timecourse right CA1 activity revealed a greater subsequent memory effect for the Distinctiveness than the Similarity task. When this analysis was repeated using the entire timecourse (3–12 s), the two-way interaction between task and memory performance was no longer significant ($p > .1$). Thus, the influence of the Distinctiveness task on encoding activity appeared to be driven by activation late in the timecourse in CA1.

Similarly, in the left subiculum, no main effects of encoding task, memory performance, or timecourse bin ($p$ values $> .2$) were found, but results revealed a significant three-way interaction between these factors ($F(1,13)=4.79$, $p = .047$). Specifically, no significant results were found early in the timecourse ($p$ values $> .1$), but late in the timecourse a two-way interaction was found between task and memory performance ($F(1,13)=5.70$, $p = .033$; see Fig. 6B). As in CA1, activity associated with subsequent hits was greater than that for misses in the Distinctiveness task ($t(13)=2.17$, $p < .05$), but not the Similarity task ($p > .5$). When this analysis was repeated using the entire timecourse, the two-way interaction between task and memory performance was no longer significant ($p > .6$). Finally, no significant results were found in perirhinal or entorhinal cortices.

To directly evaluate whether top-down influences on MTL activity differed between regions, we conducted three-way repeated measures ANOVAs for each pairwise combination of regions, with factors region, encoding task (Distinctiveness, Similarity), and subsequent memory performance (hit, miss), focusing specifically on activity late in the timecourse. As expected based on the results reported above, analyses revealed a main effect of memory when comparing left DG/CA2/3 and right PHC ($F(1,13)=12.46$, $p = .004$), but no task x memory interaction ($p > .6$) nor any significant interactions involving region ($p$ values $> .4$). Conversely, comparisons between right CA1 and left subiculum revealed a significant task x memory interaction ($F(1,13)=10.72$, $p = .006$) but no main effect of memory ($p > .3$) nor any significant interactions involving region ($p$ values $> .2$). Moreover, a trend towards a significant region x task x memory interaction was found for comparisons between left DG/CA2/3 and right CA1 ($F(1,13)=4.32$, $p = .058$), left DG/CA2/3 and left subiculum ($F(1,13)=3.26$, $p = .094$), and right PHC and right CA1 ($F(1,13)=4.39$, $p = .056$). These results raise the possibility that encoding activity in MTL subregions is differentially affected by task demands, such that subicular and CA1 encoding activity may be influenced more strongly by top-down goal states than DG/CA2/3 or PHC.

4. Discussion

The current study examined the manner in which task demands during learning affect MTL activity and its relation to subsequent memory performance. Using a paradigm in which participants’ attention was directed toward either item distinctiveness or similarities among items, we found that attending to distinctiveness was associated with enhanced subsequent recollection relative to attending to similarity. Imaging results demonstrated robust subsequent memory effects across the MTL. Critically, in CA1 and the subiculum, significant subsequent memory effects were only observed when task goals required attention to stimulus
4.1. Distinctiveness and subsequent memory

Consistent with previous studies (e.g., Glanzer & Adams, 1990; Hunt, 1995), we found that attending to distinctiveness plays a key role in the formation of rich episodic memories, and that stimuli judged as more distinctive are better remembered than those judged as more typical.

4.2. Mechanisms of hippocampal encoding

Results from Experiment 2 revealed increased activity for subsequent hits relative to misses in several regions throughout the MTL: Left DG/CA2/3, right CA1, left subiculum and right PHC. Subsequent memory effects in DG/CA2/3 and PHC did not differ according to task demands. One exception was activity patterns in DG/CA2/3, which were less susceptible to top-down influences than in CA1 or the subiculum. It bears noting, however, that when the pattern of activity in DG/CA2/3 was directly compared to that of CA1 and the subiculum, differences between subfields only reached trend levels of significance. Furthermore, given that anatomical ROIs were used, it is possible that portions of DG/CA2/3 may have exhibited a task × subsequent memory interaction as well, but that such effects were not detected when averaging across the entirety of the ROI. To this end, caution is warranted in interpreting regional differences among hippocampal subfields in the degree to which top-down goal states affect encoding activity. It may be that the behavioral benefits of attending to distinctiveness depend upon robust engagement of the entire hippocampal circuit such that each region plays a critical role in supporting the formation of rich, vivid memories.

Supporting this view, computational models of MTL function posit that during encoding, overlapping activity patterns from entorhinal cortex are passed to CA3 as separable, distinct patterns via the DG (e.g., Norman & O’Reilly, 2003). These disparate patterns are then thought to be bound together into a conjunctive representation within recurrent connections of CA3, while simultaneously, another representation of entorhinal activity is formed in CA1 via direct perforant path connections. This CA1 representation is thought to be invertible in that CA1 both receives feedforward projections from entorhinal cortex and projects back to entorhinal cortex either directly or via the subiculum. Importantly, during learning CA1 receives input not only from entorhinal cortex, but from CA3 as well. This co-activation is thought to create a key link between the pattern-separated representations of CA3 and overlapping entorhinal representations, allowing CA1 to serve as a “translator” during subsequent retrieval processes (Hasselmo, Wyble, & Wallenstein, 1996; Norman & O’Reilly, 2003; O’Reilly & McClelland, 1994; Rolls & Kesner, 2006). Thus, such models suggest that successful encoding depends upon linking the pattern-separated representation in CA3 to the invertible representation of entorhinal input within CA1 and the subiculum, facilitating future retrieval processes.

Given that the current results demonstrate robust hippocampal engagement across the entire circuit during the Distinctiveness but not Similarity task, and that subsequent memory performance...
was superior for the Distinctiveness task, our findings are in agreement with computational models of hippocampal function highlighting the importance of the entire hippocampal circuit in successful memory formation. Our findings in CA1 and the subiculum also lend support to prior hr-fMRI studies suggesting a role for these regions in memory formation and not simply memory retrieval. Such studies have demonstrated mismatch detection in CA1 (Chen et al., 2011; Duncan et al., 2012), performance-related activity during spatial learning in CA1 (Suthana et al., 2009), novelty encoding in the subiculum (Bakker et al., 2008; Preston et al., 2010; Zeineh et al., 2000), and subsequent memory effects in the subiculum (Wolosin et al., 2012).

In Experiment 1, faces encountered in the Distinctiveness task were more often subsequently recollected than those encountered in the Similarity task, an effect which reached trend levels of significance in Experiment 2. Previous work employing explicit encoding tasks has shown greater hippocampal encoding activity for subsequently recollected than familiar items (e.g., Eldridge et al., 2005; Carr et al., 2010b), a finding thought to reflect differences in quality among memory representations. Here, we used different incidental encoding tasks to influence the quality of encoded memory representations. It is possible that when participants attend to distinctiveness under incidental learning conditions, they are engaging similar neural processes as when intentionally and effectively memorizing items. As such, encoding conditions that enhance subsequent recollection may engage the hippocampus in a similar manner.

Interestingly, our observed pattern of results in CA1 and the subiculum was due to a late hit–miss difference in activation specific to the Distinctiveness task. When attending to distinctiveness, CA1 and subicular activity associated with subsequent hits returned to baseline more slowly than for subsequent misses, whereas attention to similarity led to an equivalently rapid return to baseline for both subsequent hits and misses. These results raise the possibility that successful memory formation when attending to distinctiveness resulted in more sustained neural activity in CA1 and the subiculum. However, given that the blood oxygenation level dependent (BOLD) signal measured by fMRI is an indirect measure of neuronal function, it is important to note that regional differences in timecourse shape may also reflect differences in the vascular networks supporting each region, or regional differences in metabolism.

4.3. Top-down modulation of hippocampal activity

There have been several studies exploring the influence of top-down goal states on hippocampal activity during memory retrieval. In a study of recognition memory in which task demands were manipulated, Dudukovic and Wagner (2007) demonstrated that hippocampal activation was influenced by retrieval goals rather than the perceptual novelty of test probes. Similarly, when evaluating hippocampal responses to test probes that either matched or did not match previous targets, Duncan, Curtis, and Davachi (2009) demonstrated that hippocampal sensitivity to perceptual mismatch was not obligatory but rather modulated by task goals. Of particular relevance to the current study, Hashimoto et al. (2012) demonstrated that when recognition judgments were based upon perceptual novelty, hippocampal responses to foils were larger than when recognition judgments were based upon semantic novelty, such that attending to perceptual differences led to greater novelty responses than generalizing across differences.

The current study extends such work by exploring top-down influences on hippocampal processes during encoding, enabling us to evaluate the relationship between task demands, hippocampal activity, and subsequent memory. Our findings provide novel evidence that focusing attention on stimulus distinctiveness during learning modulates the degree to which hippocampal subfield activity is associated with subsequent memory success.

4.4. Top-down modulation of MTL cortical activity

Analyses of MTL cortical activity revealed a main effect of subsequent memory in PHC, but no task x subsequent memory interaction, indicating that top-down goal states did not significantly affect encoding activity in this region. One interpretation of these findings is that the cortex is specialized to encode regularities in the environment by assigning them overlapping representations (Norman & O’Reilly, 2003; O’Reilly & McClelland, 1994), and as such is biased to form non-specific memories regardless of top-down influences. It bears noting, however, that whereas the pattern of activity in PHC was directly compared to that of CA1 and the subiculum, differences between subfields only reached trend levels of significance. Finally, it is likely that the observed influence of task demands on hippocampal encoding activity likely resulted from interactions with a variety of cortical areas outside the MTL. Thus, an important topic for future study is to understand the nature of such hippocampal-cortical interactions by evaluating, for example, functional connectivity between fronto-parietal areas and the hippocampus.

4.5. Conclusions

Using hr-fMRI and a paradigm in which participants’ attention was directed toward either item distinctiveness or similarities among items, we demonstrated that hippocampal encoding activity was better predictive of subsequent memory success when attending to distinctiveness than similarity. Such findings provide evidence that goal states differentially modulate hippocampal activity during learning, and that recruitment of the entire hippocampal circuit during successful encoding may underlie the behavioral benefits of attending to distinctiveness. Taken together, these findings advance understanding of the effects of top-down goal states on MTL processing during learning.

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