Material specificity drives medial temporal lobe familiarity but not hippocampal recollection

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Title:
Material specificity drives medial temporal lobe familiarity but not hippocampal recollection

Running Title: Familiarity-based recognition in the MTL

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ABSTRACT

The specific role of the perirhinal (PRC), entorhinal (ERC) and parahippocampal cortices (PHC) in supporting familiarity-based recognition remains unknown. An fMRI study explored whether these medial temporal lobe (MTL) structures responded in the same way or differentially to familiarity as a function of stimulus type at recognition. A secondary aim was to explore whether the hippocampus responds in the same way to equally strong familiarity and recollection and whether this is influenced by the kind of stimulus involved. Univariate and multivariate analyses revealed that familiarity responses in the PRC, ERC, PHC and the amygdala are material-specific. Specifically, the PRC and ERC selectively responded to object familiarity, while the PHC responded to both object and scene familiarity. The amygdala only responded to familiarity memory for faces. The hippocampus did not respond to stimulus familiarity for any of the three types of stimuli, but it did respond to recollection for all three types of stimuli. This was true even when recollection was contrasted to equally accurate familiarity. Overall, the findings suggest that the role of the MTL neocortices and the amygdala in familiarity-based recognition depends on the kind of stimulus in memory, whereas the role of the hippocampus in recollection is independent of the type of cuing stimulus.
Recognizing a stimulus, such as a face, an object or a scene (e.g., a landscape), as something that has been encountered before involves a pivotal human ability called recognition memory. This can be supported by a feeling of memory that the stimulus has been encountered before and/or by the recall of specific contextual details about a previous encounter with it. These two kinds of memory are called familiarity and recollection, respectively, and have been proposed to depend on partially distinct psychological and neural encoding, storage and retrieval processes (Mandler, 1980; Jacoby, 1991; Aggleton and Brown, 1999; Yonelinas, 2002; Eichenbaum et al., 2007; Montaldi and Mayes, 2010).

Familiarity memory has long been thought to rely on the Medial Temporal Lobe (MTL) and extra-MTL cortical areas. Nevertheless, the specific role of regions such as the perirhinal (PRC), entorhinal (ERC) and parahippocampal (PHC) cortices is still uncertain. In the present study, we used fMRI to explore the interaction between stimulus content and the kind of memory engaged by testing recognition. The primary focus was to determine to what extent the patterns of MTL response found when familiarity for different types of stimuli occurs are shared or distinct. This constitutes a pivotal question relating to the degree of functional specialization for familiarity decisions within the MTL cortices. It should be noted here that we use the term MTL cortices to describe the neocortical PRC, ERC and PHC, but not the archicortical hippocampus or the subcortical amygdala.

Familiarity-based recognition and material-specificity

Material-specific effects have so far been studied almost exclusively for recollection or recollection-related processes, such as in associative tasks requiring the retrieval of different kinds of source information (e.g., Duarte et al., 2011; Staresina et al., 2011, 2013; Hannula et al., 2013). The neural networks that support familiarity memory have been investigated less systematically (but see Montaldi et al. 2006; Kafkas and Montaldi 2012, 2014) and the interaction between familiarity-based recognition and kind of stimulus has received less
attention. This question is particularly important in order to evaluate the role of the different neocortical MTL structures (PRC, PHC and ERC) in supporting familiarity-based recognition.

Most models of the MTL’s role in recognition memory (Aggleton and Brown, 1999; Eichenbaum et al., 2007; Montaldi and Mayes, 2010) agree that the hippocampus plays a special role in mediating recollection, while familiarity-based recognition is mediated by the adjacent neocortical MTL structures, most prominently the PRC (Squire et al., 2007; Wixted et al., 2010). There is less agreement about the role of the PHC with some theories linking its role to recollection (e.g., Diana et al. 2007) although it has also been linked to familiarity for contextual information (e.g., Montaldi and Mayes 2010). Similar uncertainty applies to the role of the ERC. Specifically, it is unclear whether the ERC, which receives inputs from both PHC and PRC, subserves familiarity and, if it does, for what kind of information. Although, there is some evidence linking it to familiarity-based recognition for words (Ranganath et al., 2004; Yonelinas et al., 2007; de Vanssay-Maigne et al., 2011; Brandt et al., 2016).

Indirect neuroanatomical evidence strongly suggests that the PRC, PHC and ERC processing their inputs in a similar way, which is distinct from how the hippocampus processes its inputs. They share similar neocortical cytoarchitectonics, which differs from the archicortical cytoarchitectonics of the hippocampus. Furthermore, the distinct inputs of the PHC and PRC from structures outside the MTL is consistent with the hypothesis that they subserve familiarity for different kinds of information (see Montaldi and Mayes, 2010). However, considering the inputs from within the MTL, particularly the reciprocal connections between PHC and PRC (Suzuki and Amaral, 1994), both structures may mediate familiarity for the same inputs but perhaps at different time points.

It is widely believed that PRC processes object-related inputs whereas PHC processes scene, spatial or context inputs (Davachi, 2006; Eichenbaum et al., 2007). That PRC and PHC process different kinds of information has been supported by fMRI studies, in which the PRC has usually been reported to selectively process and represent high-level object and face information, whereas the PHC has been involved in processing scene
information (e.g., Epstein et al. 1999; Lee et al. 2005, 2008; Awipi and Davachi 2008; Litman et al. 2009; Staresina et al. 2011, 2013). Nevertheless, the evidence so far is not conclusive and findings inconsistent with PRC and PHC processing respectively only object-related versus scene/context-related information, have also been reported (e.g., Bar and Aminoff 2003; Buffalo et al. 2006; Diana et al. 2008; Preston et al. 2010).

For example, Preston et al. (2010), using repeated and novel face and scene stimuli in an fMRI study, found that encoding activation only in PHC selectively related to scenes, whereas activation in PRC related to subsequent memory (old/new recognition judgments) for both faces and scenes. Selective responding in the PHC for scene encoding, but a similar response profile within the PRC for faces and scenes was also reported by Dudukovic et al. (2011). On the other hand, Diana et al. (2008) found that activity within the PHC is sensitive not only to scenes but also to faces and toys, whereas activity in the PRC did not respond to any of the employed stimulus categories. These inconsistencies could be attributed to differences in experimental design in different studies. However, they may also reflect an interaction between stimulus category and the underlying memory process, such as familiarity or recollection, which participants may engage during repeated presentations of stimuli.

As noted above, all the studies that have explored material-specificity in the MTL have to a great extent ignored whether familiarity or recollection are involved in recognition decisions or they have exclusively focused on recollection or other associative tasks. One notable exception is a recent study by Martin et al., (2013; for a neuropsychological study see also Martin et al., 2011) in which familiarity-based recognition for three types of objects (faces, buildings and chairs) was contrasted within the PRC and the PHC. In this study, a preference in the PRC for faces, but not buildings, and in the PHC for buildings but not faces was reported. However, it remains to a great extent unexplored whether the PRC and PHC have a general (i.e., non-specific) role in supporting familiarity memory for every stimulus category (e.g. faces, objects and scenes) or whether they are parts of networks that are
specialized in supporting familiarity-based recognition for specific types of stimuli. This question is directly explored in the present study.

The role of the hippocampus in recollection

In contrast to the MTL cortices, the hippocampus has been proposed to have a general role in recollection and in associative memory (Aggleton and Brown, 1999; Eichenbaum et al., 2007; Mayes et al., 2007; Montaldi and Mayes, 2010) across different domains and stimulus categories (e.g., Davachi 2006; Konkel and Cohen 2009; Duarte et al. 2011; Staresina et al. 2011). Nevertheless, this proposal has been criticized and an important controversy remains as to whether the hippocampus has a selective role in recollection or supports both familiarity and recollection. Recent fMRI and neuropsychological evidence (for reviews see Eichenbaum et al., 2007; Skinner and Fernandes, 2007; Montaldi and Mayes, 2010; Migo et al., 2012; Rugg and Vilberg, 2013) linking recollection to the hippocampus, but not to familiarity, has been challenged on the grounds that these findings are subject to a methodological confound (Squire et al., 2007; Wixted et al., 2010). According to this argument, hippocampal activity and familiarity and recollection decisions systematically co-vary with recognition memory strength, resulting in higher hippocampal activity for recollection, which is usually associated with stronger recognition memories. In contrast, although sometimes familiarity can support strong recognition memory, it is usually associated with weaker recognition memories that fail to noticeably affect the hippocampal activity. The construct of “memory strength” in this argument is operationally defined as recognition memory accuracy and reported confidence.

A key prediction of this argument, therefore, is that when familiarity and recollection responses are equally strong – as indicated by the response accuracy – the hippocampus should be engaged in both cases. However, recent fMRI studies provide mixed results, either supporting (e.g., Smith et al. 2011) or contradicting (Kafkas and Montaldi, 2012) this key prediction of the strength confound view. However, these studies also had a key difference in
The category of stimuli used. An important factor, therefore, that needs to be addressed in studies of recognition memory is whether the type of information that is being encoded or retrieved has any effect on which neural systems are engaged when familiarity and recollection decisions are taken (for a discussion of this issue see also Kafkas and Montaldi 2012).

The present study
In the present study, we set out to explore MTL responses to familiarity and recollection supporting the recognition of three different types of stimuli: objects, faces and scenes. As outlined above, our primary aim was to explore whether familiarity responses for objects, faces and scenes differentially engage the neocortical MTL structures, the PRC, the ERC and the PHC. A secondary aim of the present experiment was to investigate whether hippocampal activity changes relate to recollection, but not to familiarity, even when familiarity is as strong (i.e., accurate) as recollection and whether this is consistent for object, face and scene recognition memory. In our analyses we combined univariate-GLM and multivariate (pattern recognition) approaches. The combination of the two methods of analysis has the potential to provide complementary findings and bridge inconsistencies reported in previous studies.

MATERIALS AND METHODS

Participants
In total, 20 right-handed healthy volunteers gave informed consent and participated in this experiment. All participants were native English speakers, with no self-reported psychiatric or neurological disorders and normal or corrected-to-normal vision (with contact lenses). Data from three participants were excluded from further analyses; one due to excessive movement during fMRI (more than 3mm), one due to a technical problem affecting the
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The mean age of the remaining 17 participants (12 male) was 23.28 years (SD = 3.40 years). All participants received £20 after completing the testing session. The National Research Ethics Service (North West-GM South) approved all the procedures followed in this experiment.

Stimulus Material

A total of 420 color stimuli (15 for practice) were used in this experiment. As the aim of the study was to explore the brain networks that support familiarity-based and recollection-based recognition for different stimulus types, three stimulus categories were used: outdoor scenes, objects (man-made and natural) and faces. For each stimulus type 140 stimuli (5 for practice) were presented. The scene and the object stimuli (70 man-made and 70 natural) were royalty-free images with transferable copyrights collected from various online databases (e.g., the BOSS object database; Brodeur et al., 2010), whereas the faces (70 male and 70 female) were selected from the Glasgow face database (Burton et al., 2010).

Procedure and Design

Each experimental session comprised an encoding phase (before scanning) and a retrieval phase completed in the MRI scanner (Fig. 1). At encoding, participants were presented with 270 stimuli depicting outdoor scenes, objects and faces (90 stimuli per type), organized in 10 randomly alternating blocks of 9 stimuli each. All three types of stimuli used a perceptual matching-to-sample task. This shallow encoding task (see Montaldi et al., 2006; Kafkas and Montaldi, 2012, 2014) involves taking a matching-to-sample decision based on a perceptual dimension of the presented stimulus. Specifically, each trial comprised image triplets of the same stimulus and participants had to decide which of the two bottom images matched the target image presented on top (see Fig. 1). In each trial, one of the two bottom images had been minimally modified from the original image according to a perceptual characteristic. For the objects and the faces the modified image was slightly smaller or bigger whereas, for the
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scenes, the modified image was slightly shifted horizontally (left or right) by a few mm. In each trial, the modified picture was placed randomly on the left or the right side of the screen and participants had 4s to indicate their response using two keyboard buttons (“1” for left and “0” for right). A practice block, exemplifying this task for the three types of stimuli, preceded the main encoding block. In our previous studies (see e.g., Kafkas & Montaldi, 2012, 2014, 2015a; Montaldi, Spencer, Roberts, & Mayes, 2006) and in pilot work, this encoding task increased reliance on familiarity-based recognition keeping the frequency of the recollection responses at lower levels – without, nevertheless, affecting recollection accuracy, which remains high. This task also ensures an adequate distribution of responses across familiarity levels, which is necessary for the parametric analyses of the brain activation data (see below).

[Figure 1 about here]

After completing the encoding task, participants underwent the retrieval task in the scanner. On seeing each stimulus, participants were required to focus on judging quickly but accurately how familiar each stimulus felt across three levels: weakly, moderately or strongly familiar. If, however, they spontaneously recollected something about having seen the stimulus earlier, then they were required to respond with the ‘recollect’ option. Participants were required not to try to recollect at any time (this ‘familiarity-only’ adaption of the remember/ know procedure is described by Montaldi et al., 2006; Mayes et al., 2007; Kafkas and Montaldi, 2012). Participants were carefully trained beforehand so that they understood the distinction between familiarity and recollection (see Kafkas and Montaldi 2012, 2015b; Migo et al. 2012). Specifically, they were trained to make a familiarity response when they felt that they had seen a stimulus at study, but to make a recollection response when a stimulus brought to mind (albeit spontaneously) information associated with encoding it during its earlier encounter (e.g., that the stimulus was one of the first seen or it triggered a specific thought at encoding). Participants had the opportunity to ask questions about these two kinds of memory and provide examples of each of them from their own past experience.
A practice retrieval block was completed before the participants entered the scanner and another practice block was presented as they lay in the MRI scanner when the structural T1 image was being acquired, before the main retrieval task. The fMRI data were collected and analyzed for this main retrieval phase, which was divided into two functional runs. Participants were presented with 270 stimuli from the encoding task (i.e., target items at encoding) along with 135 new foils (i.e., 405 stimuli across all stimulus types). For each stimulus type, 135 stimuli per category (90 studied) were presented at retrieval while participants provided responses using the three levels of increasing familiarity (F1 = weak, F2 = moderate, F3 = strong familiarity), spontaneous recollection (R) and new (N) options (Fig. 1).

At retrieval, stimuli were presented in a pseudo-random sequence of face, object and scene blocks, separated by short fixation periods (10s). Within this sequence each block was followed by a different stimulus block. Each of these blocks comprised 15 trials intermixed with 3 implicit baseline fixation trials (null events). In total, 81 null events were presented across all the blocks. Each trial (including the null events) lasted for 3s followed by a 1s fixation cross. Participants were instructed to provide a response for each stimulus trial within this period, using a special MR-compatible button box. As the recognition task included 5 possible responses, three buttons on one hand were used for the familiarity responses (F1, F2 and F3) and two buttons on the other hand were used for the extra two responses (R and N). The assignment of these responses to left and right hand was counterbalanced across participants.

fMRI acquisition and analyses

Scanning was conducted on a 3T Philips (Achieva) scanner. A gradient echo-planar pulse sequence was used for the acquisition of the functional data using the blood oxygenation level dependent (BOLD) contrast. In total, 840 volumes were acquired, for each participant, across two sessions with TR = 2.5s, TE = 35ms, 2.5mm x 2.5mm x 3.5mm voxel size and 40
slices per volume covering the whole brain (positioned parallel to the AC-PC). High-
resolution T1 images were also acquired prior to the functional run (180 slices with a voxel
size of 1mm isotropic and matrix size 256 x 256). Soft pads were used to minimize head
motion during scanning and earplugs were provided to reduce MRI scanner noise.

Data quality of the fMRI time-series was examined using the ArtRepair software
(http://cibsr.stanford.edu/tools/human-brain-project/artrepair-software.html). Fewer than 5%
of the slices from two subjects were repaired using ArtRepair algorithms implementing linear
interpolation of the adjacent (preceding and following) slices within the time-series. The other
15 participants did not have any major artifacts in their functional data and no repair
algorithms were applied. The EPI data from each participant were pre-processed and
analysed using SPM8 software (Statistical Parametric Mapping, Wellcome Trust Centre for
Neuroimaging, http://www.fil.ion.ucl.ac.uk/spm/). The fMRI time-series were realigned to the
mean image (registration) using a six-parameter rigid body transformation, resliced using
sinc interpolation and slice-time corrected (to the middle slice) to account for differences in
slice acquisition times. Residual movement artefacts were also obtained from the ArtRepair
toolbox and used as nuisance regressors in the first-level analysis (see below). Individual T1
images were coregistered to the corresponding mean EPI image. Spatial normalization of the
EPI and T1 images to the MNI template was performed using the DARTEL toolbox
implemented in SPM8 (Ashburner, 2007). Finally, the spatially normalized EPI data were
resliced to 3mm isotropic and spatially smoothed using an isotropic 6mm full width half
maximum (FWHM) Gaussian kernel.

Univariate Analyses

Pre-processing

The pre-processed individual EPI data were further analyzed in SPM8, first at a single
subject level using the general linear model (GLM analysis). Specifically, the event-related
functional data were modelled separately for each participant using a canonical
hemodynamic response function (Friston et al., 1998). In this model a series of delta (stick)
functions corresponding to the onset of each event were convolved with the canonical
hemodynamic response. Two models, one parametric and one categorical were specified for
each participant (see below) and included all response outcomes for each stimulus type
(objects, faces and scenes) modeled as conditions of interest. Nuisance regressors were
also modeled and included trials with no behavioral response, the six movement parameters,
produced at realignment for each of the two functional runs and residual movement artefacts
obtained from the ArtRepair toolbox. To remove low-frequency noise the data were high-
pass filtered using a cut-off of 128s.

**Parametric analyses**

Brain activity modulations by familiarity strength were investigated using the parametric
model and involved three parametric analyses, exploring monotonic increases or decreases
in activity across familiarity, separately for scenes, objects and faces (Büchel et al., 1998).
Specifically, at the first (subject) level, familiarity hits (i.e., old stimuli reported as familiar), for
each stimulus type were specified as separate conditions, while the reported strength
accompanying each familiarity response was used as a covariate and was convolved with
the stimulus-specific HRF. Three parametric conditions were specified in this way; one for
objects, one for scenes and one for faces, each of them comprising the four levels of
familiarity strength (F0, F1, F2 and F3), with misses used as the level reflecting zero
familiarity (F0). All participants had a minimum of 9 trials in each response category for each
stimulus type enabling reliable parametric analyses (for the mean number of trials across the
four response categories for the three types of stimuli see Supplementary Table 1).

Parametric \( t \) contrasts were created at the first-level for monotonic increases or decreases in
activity across familiarity strength for each stimulus type. Non-linear quadratic effects were
also modeled to capture residual variance not explained by the linear function; however,
these produced no significant additional activations and are not reported separately. Finally,
the parametric analyses were also conducted using 3 levels of strength (F1, F2 and F3) for each type of stimulus, but these produced very similar results to the main analyses and are not presented separately.

In the parametric analyses both common (i.e., shared) activations across all stimulus types as well as material-specific familiarity activations for each type of stimulus were examined. Specifically, shared familiarity activity was explored by means of a conjunction analysis (Friston et al., 2005) testing for consistent parametric effects in the whole brain across scenes, objects and faces. Unique parametric activation and deactivation patterns across familiarity strength were explored for each stimulus type (scenes, objects and faces) by applying a series of exclusive masks (at $P < 0.05$) to each parametric contrast. In the case of scene familiarity, the parametric responses were exclusively masked by object and face familiarity activations. Similarly, object familiarity activations and deactivations were exclusively masked by scene and face familiarity activation and deactivation patterns. Finally, parametric responses to familiarity for faces were exclusively masked by familiarity responses to scenes and objects.

**Familiarity versus recollection contrasts**

To explore whether the hippocampus preferentially responds to recollection or to both recollection and strong familiarity (F3), a direct contrast between these two conditions was run for objects ($R_{\text{objects}} > F3_{\text{objects}}$) and scenes ($R_{\text{scenes}} > F3_{\text{scenes}}$) and collapsed across the three types of stimuli ($R > F3$) using the categorical model. Furthermore, two conjunction analyses were conducted to assess the overlap of the hippocampal responses for scene and object stimuli, one using the $R > F3$ contrast and the other using the $R > M$ contrast for the two types of stimuli. As faces produced very few R responses, these were not analysed separately or in the conjunction analyses, but were included in the analysis with the collapsed R responses. Furthermore, F3 and R responses were also contrasted versus misses (M) again for each stimulus type ($F3_{\text{objects}} > M_{\text{objects}}$; $F3_{\text{scenes}} > M_{\text{scenes}}$; $F3_{\text{faces}} > M_{\text{faces}}$).
R_{objects} > M_{objects}; \ R_{scenes} > M_{scenes}) \text{ and collapsed across type (} F3 > M \text{ and } R > M). \ For the recollection analyses, data from 16 participants were used as 1 participant did not give enough recollection responses at retrieval. The individual parametric and categorical contrasts of interest were entered into a one-sample $t$-test in the second-level analysis treating participants as a random effect. All the produced SPM(t) maps, for all the analyses reported here, were initially threshold at an uncorrected voxel level of $P < 0.001$ and clusters are reported as significant when at least 8 contiguous voxels were active unless noted differently. In the case of activations surviving a cluster-wise FWE-correction for multiple comparisons these are denoted separately.

**Multivariate analyses**

The multivariate analysis within the whole brain and the a priori ROIs were conducted using *Pattern Recognition for Neuroimaging Toolbox* (PRONTO, http://www.mlnl.cs.ucl.ac.uk/pronto/; Schrouff et al. 2013). In this multi-voxel pattern analysis (MVPA) method the aim is to classify different experimental conditions within specified regions of interest, based on distributed patterns of activity across voxels. The a priori anatomical ROIs included the hippocampus, the perirhinal cortex (PRC), the entorhinal cortex (ERC), the parahippocampal cortex (PHC) and the amygdala. The PickAtlas Toolbox was used for the definition of each of these ROIs (Maldjian et al., 2003, 2004), with the exception of the PRC and ERC, which were identified individually for each participant using the probabilistic map created by Devlin and Price (2007) in combination with previously published anatomical criteria (Insausti et al., 1998). Each classification, as described below, was run separately for left and right structures as well as for the bilateral mask of each structure.

To investigate the two main questions of the present study, two sets of classification analyses were performed using in each case the condition-specific fMRI beta images generated for each subject in the GLM analysis. First, to explore whether familiarity-related
brain responses within the different ROIs vary as a function of stimulus type, we used a Multiclass Gaussian Process Classification (GPC) (Rasmussen and Williams, 2006). This algorithm was applied to the F3 (strongly familiar) responses across the three types of stimuli (scenes vs objects vs faces). The same analysis was also conducted for the collapsed familiarity responses and is presented in the supplement (Supplementary Fig. 1). To control for any potential systematic bias favoring the classification of one stimulus type over the other two types of stimuli in the multiclass GPC classification model, separate binary SVM (Support Vector Machine) classifications were also performed between all the possible stimulus combinations (i.e., scenes vs objects; scenes versus faces; objects versus faces) and an aggregate classification accuracy (i.e., mean accuracy across all combinations) for each stimulus type was calculated. As the results produced from this approach are very similar to those from the main Multiclass GPC, they are not reported separately in the Results but can be found in Supplementary Table 2. Finally, a separate analysis examined pattern classification for familiar stimuli using binary SVM between F3 responses and misses (M) separately for each stimulus type. This analysis, which is presented in Supplementary Table 5, complemented the main GPC classification reported in the Results below and produced similar findings to it. However, any discrepancies are noted and discussed (see Results and Discussion).

To approach the second aim of this study regarding the role of the hippocampus in coding for recollection and/or equally strong familiarity, a second set of classification analyses compared classification performance for F3 and R responses versus misses (F3 vs M and R vs M) collapsed for the three types of stimuli using a binary SVM (Support Vector Machine) algorithm. Finally, to explore any differences in the classification outcome for Rscenes and Robjects in the hippocampus, a separate binary SVM model was also run. Each classification was performed separately for each ROI and included all voxels within each ROI (no feature selection was used). In these classification analyses the data were mean centered and a leave-one-subject-out (LOSO) cross-validation procedure was performed to
ensure independence between training and test data sets. Specifically, this cross-validation
procedure involves repeated repartitioning of the training and test data to derive an estimate
of the accuracy and the generalization error of the model across subjects. This group
analysis, treating subjects as a random factor, enables the generalization of the classification
results to the population. Statistical significance of the classifications within each ROI was
tested using permutation testing (1000 permutations) for each model (classifier).

RESULTS

Behavioral results

Memory accuracy \[\text{Hit rate} / (\text{Hit rate} + \text{FA rate})\] for familiarity and recollection responses is
presented in Figure 1B. The 3 x 3 ANOVA with stimulus type (objects, faces, scenes) and
familiarity strength (F1, F2, F3) as the within-subjects factors showed significant main effects
of type \(F_{2,30} = 3.44, P = 0.045\) and strength \(F_{2,30} = 150.99, P < 0.001\). Pairwise
comparisons revealed higher familiarity accuracy for objects than faces \(P = 0.007\), but no
other difference across the three types, and increased memory accuracy with increased
familiarity strength \(F3 > F2 > F1; \text{all } Ps < 0.001\). Importantly, as presented in Figure 1B, F3
responses were characterized by matched memory accuracy with R responses, across all
stimuli types collapsed \(M_{F3} = 0.90, SD_{F3} = 0.09\) and \(M_{R} = 0.90, SD_{R} = 0.13; t < 1\) and
separately for scenes \(M_{F3} = 0.91, SD_{F3} = 0.08\) and \(M_{R} = 0.90, SD_{R} = 0.26; t < 1\) and objects
\(M_{F3} = 0.93, SD_{F3} = 0.10\) and \(M_{R} = 0.90, SD_{R} = 0.18; t < 1\). Recollection responses for faces
were very rare and therefore are not evaluated separately.

The 3 x 3 ANOVA on the RTs showed a significant main effect of stimulus type \(F_{2,30} = 54.31, P < 0.001\), with longer latencies for scenes (1943 ms; SD = 42.61 ms) than both
objects (1620 ms; SD = 39.52 ms) and faces (1656 ms; SD = 56.43 ms; both \(P < 0.001\)) and
a significant effect of strength \(F_{2,30} = 8.56, P = 0.001\), indicating shorter latencies for more
confident responses \(F1 > F2 > F3\). Matched RTs characterized recollection and F3
responses across all stimulus types collapsed (F3 = 1577 ms, SD = 213; R = 1653 ms, SD = 284 ms) and separately for scenes (F3\textsubscript{scenes} = 1807 ms, SD = 281 ms; R\textsubscript{scenes} = 1932 ms, SD = 421 ms) and objects (F3\textsubscript{objects} = 1467 ms, SD = 230 ms; R\textsubscript{objects} = 1537 ms, SD = 240 ms; all t < 1).

fMRI results

\textbf{Material-specific familiarity effects within the Medial Temporal Lobes}

In the univariate analysis, parametric increases or decreases across familiarity strength (from F0 to F3) were analyzed separately for scene, face and object stimuli (see Methods). Shared activation patterns within the MTL across all stimulus types as well as material-specific familiarity activations for each type of stimulus were explored. In the conjunction analysis, no overlapping MTL region was found to respond to familiarity for the three types of stimuli. Instead, only material-specific responses in the MTL were found during familiarity decisions. This means that structures of the MTL only provide material-specific support when involved in familiarity decisions.

Specifically, a variety of familiarity responses across the three stimulus types emerged within the MTL. As shown in Figure 2, the right PHC (BA 35; x = 24, y = -30, z = -12 and x = 21, y = -33, z = -12, 21 voxels; Fig. 2A) and a cluster within the bilateral ERC and PRC (BA 28/35; only the right cluster survived the exclusive mask; x = 21, y = -6, z = -30, 9 voxels) increased their activity across familiarity strength selectively for the scene stimuli. In contrast, unique parametric deactivation patterns for object stimuli (Fig. 2C) were found within the left PHC (BA 36; x = -27, y = -30, z = -18, 5 voxels) as well as in an area of the anterior PRC/ERC (x = 21 y = -6 z = -27) at a lower threshold ($P < 0.001$, uncorrected; 2 voxels). Finally, a cluster within the left amygdala (including the dorsal ERC; BA 34; x = -21, y = 0, z = -24; 35 voxels) and another cluster within the fusiform gyrus (BA 20; x = 33, y = -36, z = -24; 17 voxels) uniquely responded to face familiarity (Fig. 2B). Importantly, no shared or
selective familiarity response was found in the hippocampus for any of the 3 types of stimuli even at a lower threshold ($P < 0.01$, uncorrected).

[Figure 2 about here]

**MVPA analysis**

Multivariate classification was conducted for the whole brain and a priori ROIs including the hippocampus, the PRC, the ERC, the PHC, and the amygdala. A multi-class Gaussian Process Classification (GPC) algorithm was trained to classify F3 (strongly familiar) responses for scenes, objects and faces across participants (see Methods; see also Supplementary Table 2 for an alternative analysis). The classification profile, in terms of classification accuracy and classification errors, for each anatomical ROI is summarized in Figure 3 and in Supplementary Table 3. The whole brain analysis yielded a significant accuracy of 60.4% ($P = 0.01$) and significant classification accuracy for scenes (accuracy = 56.3%, $P = 0.05$), objects (accuracy = 68.8%, $P = 0.05$) and faces (accuracy = 56.3%, $P = 0.03$). This indicates that activation patterns within the whole brain successfully discriminated F3 responses across the three types of stimuli. In the hippocampus, classification accuracy for F3 responses to scenes, objects and faces was low and not significant (Fig. 3), whereas in the PRC there was a selective significant classification for objects versus faces and scenes (whole PRC: accuracy = 75%, $P = 0.01$; Left PRC = 81.3%, $P = 0.01$ and Right PRC = 81.3%, $P = 0.01$; Fig. 3). Similar to the PRC, familiar objects, versus the other two stimulus types, were significantly classified in the ERC (whole ERC: accuracy = 75%, $P = 0.02$; Left PRC = 75%, $P = 0.01$ and Right PRC = 75%, $P = 0.01$; Fig. 3). In the PHC, as shown in Fig. 3, significant classification accuracy was found for scenes in the whole PHC (accuracy = 68.8%, $P = 0.02$) and the left PHC (accuracy 75%, $P = 0.02$) and for objects in the right PHC (accuracy = 75%, $P = 0.01$). Similar results were obtained when partitioning the parahippocampal gyrus (including both PRC and PHC) into anterior, middle and posterior portions. The anterior aspect (corresponding to the PRC/ERC) showed selective
classification sensitivity to object familiarity (accuracy = 68.8, \( P = 0.02 \)). The posterior aspect (corresponding to the PHC) showed significant classification accuracy for both objects (accuracy = 81.3, \( P = 0.01 \)) and scenes (accuracy = 68.8, \( P = 0.02 \)), while the middle aspect of the gyrus, corresponding to the transitional zone between PRC and PHC, was also found to code for both object (accuracy = 68.8, \( P = 0.04 \)) and scene familiarity (accuracy = 68.8, \( P = 0.04 \)). In the amygdala (Fig. 3), classification accuracy was significant for faces within the left (accuracy = 56.3, \( P = 0.05 \)) and the right amygdala (accuracy = 62.5\%, \( P = 0.04 \)) but not for the other two stimulus types. Similar multivariate classification findings were obtained when the same analysis was conducted for the familiarity responses collapsed across the three levels of strength (Supplementary Fig. 1) and when F3 responses were classified relative to misses separately for scenes, objects and faces (Supplementary Table 5). The only difference between the main analysis, as reported above, and the one classifying familiar (F3) versus missed stimuli (reported in Supplementary Table 5), was the accurate classification of familiar faces in the left ERC (accuracy = 68.75, \( P = 0.02 \)).

[Figure 3 about here]

**Selective hippocampal response to recollection**

In the univariate analyses, as summarized in Figure 4, selective responses to R were found in the hippocampus in a series of contrasts between R versus accuracy-matched F3 responses and R versus misses (M). Specifically, the bilateral hippocampus (Left: \( x = -15, y = -30, z = -6, P < 0.001, 21 \) voxels; Right: \( x = 18, y = -15, z = -15, P < 0.001, 9 \) voxels) selectively responded to R versus F3 responses collapsed across the three stimulus types (Fig. 4C). Hippocampal activation was also found when R > F3 contrast was conducted separately for objects (\( R_{\text{Objects}} > F3_{\text{Objects}} \): \( x = -21, y = -27, z = -6 \) and \( x = -24, y = -18, z = -15, P < 0.05 \) FWE-corrected, 18 voxels; Fig. 4B) and scenes (\( R_{\text{Scenes}} > F3_{\text{Scenes}} \): \( x = -27, y = -9, z = -21, P < 0.001, 14 \) voxels and \( x = 18, y = -9, z = -15, P < 0.001, 10 \) voxels; Fig. 4A) denoting that the hippocampus has a material-independent role in recollection. A direct
contrast between $R_{\text{Objects}}$ and $R_{\text{Scenes}}$ (both $R_{\text{Objects}} > R_{\text{Scenes}}$ and $R_{\text{Scenes}} > R_{\text{Objects}}$) yielded only a significant cluster in the PHC ($x = 27, y = -42, z = -15, P < 0.001, 25$ voxels) responding more to $R_{\text{scenes}}$ than $R_{\text{Objects}}$ (Fig. 4D), while no differential activation for either $R_{\text{Objects}}$ or $R_{\text{Scenes}}$ was found in the hippocampus.

[Figure 4 about here]

To further validate the selectiveness of the hippocampal response to R relative to strength-matched familiarity (F3), separate analyses were run contrasting R versus misses (M) and F3 versus M, both collapsed across stimulus type and separately for objects and scenes (and faces for F3$\_\text{faces}$ versus M$\_\text{faces}$). As shown in Figure 4, recollection responses to scenes, objects and collapsed across scenes, objects and faces, resulted in greater activations within the hippocampus relative to misses ($R_{\text{scenes}} > M_{\text{scenes}}$: $x = -27, y = -9, z = -21, P < 0.05$, FWE-corrected, 20 voxels; $R_{\text{Objects}} > M_{\text{objects}}$: $x = -15, y = -36, z = 0$ and $x = -24, y = -27, z = -6, P < 0.001$, 18 voxels; $R_{\text{all}} > M_{\text{all}}$: $x = -18, y = -27, z = -9$ and $x = -24, y = -6, z = -21, P < 0.05$ FWE-corrected, 60 voxels). In contrast, none of the contrasts between F3 and misses for each stimulus type, separately and collapsed, yielded any significant activation within the hippocampus even at a considerably lower threshold ($P < 0.01$, uncorrected).

Overall, these findings stress that the hippocampus selectively supports recollection (versus familiarity) and it does so in a non-material specific fashion. To further explore the degree of overlap of the recollection activations in the hippocampus for the two types of stimuli (i.e., objects and scenes), two conjunction analyses were further conducted. The first one was performed between $R_{\text{Objects}} > F3_{\text{Objects}}$ and $R_{\text{Scenes}} > F3_{\text{Scenes}}$, while the second one between $R_{\text{Objects}} > M_{\text{objects}}$ and $R_{\text{Scenes}} > M_{\text{scenes}}$. Both conjunction analyses (see Supplementary Fig. 2) confirmed that overlapping areas within the hippocampus respond to recollection for objects and scenes ($R > F3$ conjunction: left hippocampus $x = -24, y = -18, z = -21, P < 0.05$, FWE-corrected, 20 voxels and right hippocampus $x = 24, y = -21, z = -15, P < 0.001$, 8 voxels; $R > M$ conjunction: right hippocampus $x = 24, y = -27, z = -12, P < 0.05$, FWE-corrected, 13 voxels).
**MVPA analysis**

To compare classification performance for F3 and R responses within each ROI and specifically within the hippocampus, binary SVM analyses were used (see Methods) to classify F3 and R responses versus misses collapsed across the three types of stimuli (Fig. 5 and Supplementary Table 4). As shown in Figure 5, classification accuracy for R responses was significantly above chance within both the left (accuracy = 81.3%, \( P = 0.02 \)) and the right hippocampus (accuracy = 87.5%, \( P = 0.001 \)). In contrast, F3 responses were classified very poorly in the hippocampus producing low and non-significant classification accuracy (Fig. 5).

Classification within the PRC and the PHC was significant for both F3 and R responses. Specifically, classification accuracy was significant in the whole PRC for F3 responses (accuracy = 75%, \( P = 0.03 \)), whereas R responses were reliably classified only in the left PRC (accuracy = 81.3%, \( P = 0.02 \)). In the PHC, classification within the left PHC yielded significant classification accuracy for both F3 (accuracy = 81.3%, \( P = 0.007 \)) and R (accuracy = 81.3%, \( P = 0.012 \)), while within the whole PHC significant classification accuracy was found only for R responses (accuracy = 87.5%, \( P = 0.003 \)). In the ERC, classification within the whole ERC mask was significant for R (accuracy = 84.4, \( P = 0.001 \)), but no other significant classification outcome for R or F3 was found either in the left or the right ERC.

Finally, classification accuracy in the amygdala was low and non-significant for both F3 and R responses.

An SVM analysis was also performed to compare \( R_{\text{objects}} \) and \( R_{\text{scenes}} \) specifically in the hippocampus. Consistent with the conjunction analyses reported above for scene and object recollection and the lack of any difference in the hippocampus when contrasting \( R_{\text{scenes}} \) and \( R_{\text{objects}} \), this analysis yielded a very low overall model accuracy in the hippocampus (accuracy = 56.7%, \( P = 0.26 \)) denoting the very poor discrimination between recollection for objects and scenes within the hippocampus. This is consistent with the hippocampus supporting recollection in a non-material specific way. Indeed, when \( R_{\text{objects}} \) and \( R_{\text{scenes}} \) were examined...
separately, relative to misses, they produced equally good classification outcomes in the hippocampus ($R_{\text{objects}}$: accuracy = 62.5%, $P = 0.05$; $R_{\text{scenes}}$: accuracy = 73.3%, $P = 0.02$)."

[Figure 5 about here]

**DISCUSSION**

In this fMRI study we employed a recognition memory paradigm to explore the interaction between stimulus content and the kind of memory engaged at retrieval. Our primary aim was to investigate whether familiarity for different kinds of visual stimulus is mediated by a common MTL neural network or whether there are at least partially non-overlapping MTL networks mediating familiarity for different kinds of stimulus. This question is particularly important because different models of MTL functional organization currently make different predictions about the contribution of MTL structures in familiarity-based recognition for different stimulus types (see Introduction). The findings presented here point towards a degree of specialization, with respect to stimulus type, within the MTL structures (including the neocortical structures and the amygdala) for familiarity. On the other hand, overlapping regions of the hippocampus mediated recollection, regardless of the kind of visual stimulus to which it related. However, no response to familiarity was found in the hippocampus, even when recollection was contrasted to equally accurate familiarity. These findings have important implications for current theories of recognition memory and for evaluating the extent of functional specialization within the MTL. These are further discussed below.

**Stimulus content and familiarity-based recognition: Selectivity within the MTL**

As discussed in the Introduction, different models of recognition memory make different predictions with respect to the role of PHC in supporting familiarity memory with some models stressing its role in supporting recollection (e.g., Diana et al., 2007) and others stressing its role in familiarity-based recognition (e.g., Montaldi and Mayes, 2010). The MTL cortices and specifically PRC, ERC and PHC were found in the present study to respond to
stimulus familiarity for object and scene stimuli but not for faces. The parametric analyses identified regions that increased or decreased their activity as a function of the reported familiarity strength. The identified regions are assumed to have a special role in familiarity-based recognition, as their activity systematically correlates with the level of reported familiarity (Montaldi et al., 2006; Kafkas and Montaldi, 2014). The PHC showed significant increases in activity with increased familiarity strength for scenes, but decreased activity with increased familiarity strength for objects. The same pattern was observed in the PRC/ERC with increases in activity tracking reported familiarity strength for scenes, but decreases in activity accompanying increased familiarity strength for objects. These findings indicate that both PHC and PRC/ERC have a role in familiarity-based recognition at least for objects and scenes, but not for faces. Nevertheless, the differential direction of the activation modulation with familiarity strength, within the PHC and PRC/ERC depending on the type of stimulus – activation pattern for scenes and deactivating patterns for objects – may suggest that familiarity-based recognition for both objects and scenes triggers different computations within the MTL cortices. Indeed, to the extent that feelings of familiarity can be driven by a number of underlying processes (for a discussion of this see Montaldi and Mayes, 2010; Kafkas and Montaldi, 2014), then the computations performed by the MTL cortices could differ; for example, relating to repetition suppression, familiarity strength or evaluation of this strength.

The multivoxel classification analyses, a more sensitive way of capturing activity changes across multiple voxels within a region (Norman et al., 2006), indicated a further degree of specialization, at least in the PRC and ERC. Specifically, the PRC and ERC coded familiarity-based recognition selectively for objects, whereas the PHC coded familiarity signals for both objects and scenes. Partitioning the MTL cortex (i.e., dividing parahippocampal gyrus) into anterior, middle and posterior aspects revealed consistent results, with the anterior aspect (corresponding to the ERC and PRC) showing selective classification sensitivity to object familiarity and the posterior aspect (corresponding to the
showing accurate classification for both objects and scenes. The middle aspect of the gyrus, corresponding to the transitional zone between PRC and PHC, was also found to code for both object and scene familiarity. This finding partially agrees with previous evidence showing a gradient in functional specialization within the MTL cortex with respect to object and scene information, which does not follow strict boundaries between PHC and PRC (Litman et al., 2009; Staresina et al., 2011). In these studies, as in our findings, the most anterior aspect of the MTL cortex responded to object stimuli, whereas the middle transitional zone of the cortex responded to both object and scene information. This transitional zone between PRC and PHC has also recently been found to have unique functional and anatomical characteristics in relation to the more anterior or posterior portions of the parahippocampal gyrus (Zhuo et al., 2016). Nevertheless, contrary to the selective response to scene stimuli in the posterior MTL cortex in previous studies (Litman et al., 2009; Staresina et al., 2011), our findings show that even in the posterior aspect of the gyrus (and indeed even in the posterior PHC), the activation patterns were sensitive to both object and scene familiarity.

Face familiarity, on the other hand, did not lead to univariate parametric activity, and did not result in an accurate classification outcome, in the MTL cortices or the hippocampus. However, the amygdala was found in both analyses (univariate and multivariate classification) to selectively code for face familiarity. The amygdala cluster in the univariate analysis also included voxels (9 out of 35 active voxels) falling within the ERC (BA 34). This finding may agree with recent evidence that areas within the anterior temporal lobe (including portions of the ERC and PRC) have a role in face discrimination and identification (e.g., O'Neil et al., 2009; Nestor et al., 2011; Nasr and Tootell, 2012; Rossion et al., 2012; Von Der Heide et al., 2013; but see Axelrod and Yovel, 2015). However, when examined separately, the multivariate analysis within the ERC favored the classification of object, but not faces, similar to the pattern observed in the PRC, whereas the amygdala ROI clearly favored the classification of familiar face stimuli. It is possible, therefore, considering the proximity of the
two structures, that the observed univariate parametric activation in the ERC is driven predominantly by the amygdala activation. However, in one alternative MVPA analysis (Supplementary Table 5), as noted in the Results, when classifying F3 relative to misses for each stimulus type left ERC was found to accurately and significantly discriminate familiar faces. Therefore, these findings cannot exclude the possibility that a portion of the ERC – especially the dorsal BA 34 area – has a role in face familiarity along with the amygdala.

It has been reported before that the amygdala interacts with the hippocampus to promote encoding of emotional (Dolcos et al., 2005; LaBar and Cabeza, 2006), as well as non-emotional information (Babiloni et al., 2009), and that successful memory encoding of item information activates the amygdala (Kensinger and Schacter, 2006). Selective amygdala activation to nominally neutral faces has also been reported in novelty detection tasks (Schwartz et al., 2003; Wright et al., 2003) and when faces and scenes are contrasted (Balderston et al. 2011). Our findings further suggest that the amygdala has a selective role in supporting familiarity-based recognition for faces, even when the faces were not overtly emotionally arousing. Therefore, contrary to the view that the neocortical MTL structures (and especially the PRC) supports familiarity-based recognition for every stimulus type (Aggleton and Brown, 1999; Montaldi and Mayes, 2010), our findings show that familiarity decisions for faces is not accomplished in the MTL cortices but predominantly in the subcortical amygdala, although the dorsal ERC may also have a role.

One contentious issue with respect to this proposal, is that previous neuropsychological evidence (e.g., Taylor et al., 2007; Mundy et al., 2013) suggests that face learning and recognition should rely on the integrity of the MTL cortices. However, in these studies, lesions within the MTL were not confined to the parahippocampal gyrus but normally extended into the amygdala. Furthermore, sparing of the MTL neocortical regions and the amygdala when the hippocampus was damaged, resulted in normal recognition performance when faces were involved (e.g., Mayes et al., 2002; Bird et al., 2007). Consistent with this suggestion, lesions of the amygdala in rats have been reported to selectively result in
impaired familiarity for odors (Farovik et al., 2011). Odors can be seen as carrying important social cues for rats in the same way as faces signify an important social cue to humans. Indeed, in humans, selective amygdala damage has not only been associated with deficits in facial expression recognition (especially fear; Adolphs et al., 1999; Mattavelli et al., 2014), but also with impaired learning of new faces (Young et al., 1995). Therefore, a critical role of the amygdala in familiarity-based recognition for faces is convergently supported by both lesion and fMRI evidence.

The lack of PRC sensitivity to face familiarity (in both univariate and multivariate analyses) may indicate that the role of PRC in face processing is limited to face discrimination and identification (as noted in previous studies; e.g., Rossion et al., 2012; Olson et al., 2015) without, nevertheless, contributing to familiarity assessment. However, although the present study was not designed to explore face perception or discrimination, even when all the face stimuli were contrasted with either objects or scenes, irrespective of their memory status, no noticeable activation in PRC for faces even at a very low threshold ($P < 0.01$, uncorrected) was found. Instead, other, more posterior areas of the inferior temporal lobe (including the fusiform gyrus), were activated by faces (for a similar lack of activations in the anterior temporal areas for faces see Axelrod and Yovel, 2015). Therefore, these findings are not consistent with a role of the PRC in face discrimination.

It is possible, however, that any potential role of the PRC, as well as the rest of the anterior temporal lobe, in face discrimination, as has been proposed in a few recent studies (e.g., O’Neil et al., 2009; Nestor et al., 2011; Nasr and Tootell, 2012; Rossion et al., 2012; Von Der Heide et al., 2013), is contingent on the nature of the perceptual discrimination tasks that have been used in previous studies as opposed to a memory task as employed here. However, O’Neil et al., (2009) and Martin et al. (2013; see also Martin et al., 2016) also using memory tasks (memory discrimination and familiarity-based recognition tasks respectively), reported PRC activations for faces. As noted above, although no face familiarity signal was identified in the PRC in the present study, another anterior temporal area (the left ERC) was
found to have a potential role in face familiarity, as indicated in the univariate activation patterns (clustered with the amygdala activation) and in one of the additional MVPA analyses reported in Supplementary Table 5. Could this discrepancy between previous studies that have identified face familiarity signal in the PRC (O’Neil et al., 2009; Martin et al., 2013) and the present findings be explained by differences in the anatomical definition of PRC and ERC ROIs? We do not consider this to be a valid explanation as an ROI-led analysis in the present study was used in the MVPA analysis and not in the univariate analysis, where the ERC activation to face familiarity was identified too (within the amygdala cluster). Also, although the ROI definition in the present study was based on probabilistic maps, the separation of the ERC and PRC was performed at the individual levels following established landmark-based criteria (see Methods). Therefore, our findings show that areas within the anterior temporal lobe (ERC and amygdala) carry face familiarity information but such signal was not identified in the PRC. Further studies, therefore are needed to evaluate the contribution of different tasks in modulating the activity within areas along the tip of the collateral sulcus in face familiarity.

Our findings, only partially support the previously proposed functional division between PRC and PHC as specializing in processing object and scene information, respectively (e.g., Lee et al., 2005; Davachi, 2006; Eichenbaum et al., 2007; Montaldi and Mayes, 2010; Staresina et al., 2011; Watson et al., 2012). Rather the PHC appears to have a more general role in coding familiarity, at least for objects and scenes. In a previous MVPA study, Diana et al. (2008) reported non-stimulus selective classification of pictorial stimuli (scenes, objects, faces and toys) in the PHC and Liang et al. (2012), reported significant classification of various stimulus types (faces, scenes and words) within both PHC and PRC. Furthermore, Bar et al. (2008; see also Aminoff et al. 2007, 2013) have proposed that the role of the PHC is not limited to processing scene or place information but in coding contextual associations for spatial and non-spatial stimuli (for a similar discussion see also Montaldi and Mayes,
It is, therefore, reasonable to propose that functional divisions within the MTL strictly limited to a division between scenes and objects cannot explain the breadth of findings.

One unique aspect of the present study is the emphasis on investigating the role of the PRC, ERC and PHC selectively in familiarity-based recognition and not on visual discrimination or recollection memory retrieval as in all previous studies that have explored material-specificity effects in the MTL (Awipi and Davachi, 2008; Diana et al., 2008; Litman et al., 2009; Preston et al., 2010; Duarte et al., 2011; Staresina et al., 2011, 2013; Liang et al., 2012). However, as perception and recognition memory coincide and, in many cases, it is difficult to experimentally separate them, one important question in relation to our findings is to what extent the material specificity found for familiar stimuli in the MTL is driven by perception more than memory for the items. This is more relevant in the case of the MVPA approach as, in this case, a classification analysis across the three types of familiar stimuli may be attributed to perceptual factors and not necessarily to the sensitivity of these structures to the type of memory (i.e., familiarity). There are, however, two important points that speak against this interpretation. First of all, there is a striking agreement between the univariate and the multivariate effects in the MTL. Taking this into account, a perceptual explanation cannot explain the fact that the areas that are reported as selectively coding for familiarity for the different types of stimuli, also responded in a parametric way tracking increases in reported familiarity strength. In this case, an area that increases its activity from weak to strong familiarity for a type of stimulus (e.g., objects) does so because it detects changes in the reported familiarity and not because it responds to the perceptual characteristics of objects. Secondly, when running the classification analysis to compare strong familiarity (F3) to misses (M) for each type of stimulus separately (Supplementary Table 5), similar findings to those reported in the main analysis are revealed. In this case the type of stimulus is the same between F3 and M and therefore the successful classification and discrimination of these two responses (F3 vs M) cannot be attributed to the stimulus type, but only to the existence or not of familiarity for the stimulus.
The findings presented here provide direct evidence that the PHC supports familiarity-based recognition at least for objects and scenes, along with a more specialized role of the PRC and ERC in object familiarity. These findings challenge the traditional view of a specialized role of PHC in supporting memory for spatiotemporal types of contexts, such as scene stimuli. That said, one possibility, which is worth considering, is that the MTL cortices may engage in familiarity-based recognition for different types of object stimuli or different object properties (such as size, contextual reference etc.). Indeed, some evidence for this comes from a recent study (Martin et al., 2013) in which familiarity responses in the PHC were selectively identified for buildings, but also from studies linking PHC to specific properties of scene stimuli which may also contain objects (e.g., Bar et al., 2008). This issue merits further investigation in order to understand which specific object dimensions/properties during familiarity decisions may be selectively processed in the PHC.

Related to this point, in one of our previous studies (Montaldi et al., 2006) in which familiarity memory for scenes was explored, PRC (but not PHC) was found to respond to scene familiarity. The difference in relation to the role of PHC in scene familiarity between the present findings and the ones reported in Montaldi et al., (2006), may be attributed to the characteristics of the scene stimuli used in the two studies. In the present study landscape images with very few central objects were used promoting, therefore, processing of broader visuospatial contexts. In contrast, Montaldi et al., (2006) used scene stimuli that constituted unique events incorporating object information too. Therefore, the role of the PRC in the earlier study may have been driven predominantly by object information within the scenes or by processing scenes more as objects and not as complex visuospatial arrays. Indeed, the direction of the familiarity effects in the prior study (Montaldi et al., 2006) is the same as the direction of the deactivation patterns in the PRC for object stimuli as found in the present experiment.

The findings also provide clear evidence for the role of ERC in familiarity-based recognition, at least for object stimuli and possibly for faces. The role of this structure
appears to be very similar to that of the PRC in familiarity discrimination as evident from the very similar classification outcomes within the two structures (see Fig. 3). Although, most models of recognition memory do not make any explicit predictions about the role of ERC in familiarity-based recognition, our findings, along with other fMRI and lesion evidence (Ranganath et al., 2004; Yonelinas et al., 2007; de Vanssay-Maigne et al., 2011; Brandt et al., 2016), indicate that its role in mediating familiarity memory for at least some kinds of stimulus information may be more important than previously appreciated.

One final note is that the degree of functional specialization within the MTL cortices and the amygdala in supporting familiarity memory, as found here, should be consistent with the cytoarchitectural profiles of these regions and the inputs they receive. Indeed, as described in the Introduction, the MTL cortices are characterized by similar cytoarchitecture and therefore, it is plausible that they perform similar computations with respect to stimulus familiarity. However, they also receive different inputs from other cortical areas, which may constrain the type of stimuli they process, consistent with the findings from the present study with the more selective role of the PRC and ERC in processing familiarity for object stimuli. On the other hand, the subcortical amygdala has a unique and complex cytoarchitecture (Pitkänen, 2000; Pitkänen and Kemppainen, 2002), radically different from the adjacent MTL cortices. It is, therefore, plausible that the type of familiarity processing that the amygdala applies to the stimuli it processes (e.g., faces), is also radically different from the kind of familiarity computed in the neocortical MTL regions. Such a discrimination implies that familiarity may be a multi-process kind of memory, with differences expressed within the MTL or even in extra-MTL regions, a suggestion that merits further investigation.

The hippocampus supports recollection, but not familiarity, irrespective of stimulus type
A secondary aim of the present study was to explore more closely whether the hippocampus has a selective role in recollection, when compared to equally accurate familiarity, and
whether this selective role applies to all three different types of cuing items (objects, faces and scenes). As described in the Introduction, although strong neuroimaging evidence supports the selective role of the hippocampus in recollection, but not in familiarity (for reviews see Eichenbaum et al., 2007; Skinner and Fernandes, 2007; Montaldi and Mayes, 2010; Rugg and Vilberg, 2013), a counterproposal has challenged this evidence on the basis of a methodological confound (Squire et al., 2007; Wixted et al., 2010). According to this argument, when strong memories are produced (in terms of reported confidence and accuracy), irrespective of whether the basis is familiarity, recollection or a combination of the two, the hippocampus will be active.

The results presented here are inconsistent with this prediction. First, activity in the hippocampus was not modulated by familiarity strength for any of the three kinds of stimulus used in the present study. Second, the hippocampus responded to recollection for all stimulus types collapsed and separately for objects and scenes when compared to misses and equally accurate familiarity. This was also complemented by a reliable selective classification of recollection responses in the hippocampus, but not F3 (strongly familiar) responses. These patterns of results are not compatible with the predictions suggested by the strength confound view, but join the previous evidence for a selective role of the hippocampus in supporting recollection (e.g., Yonelinas et al., 2005; Montaldi et al., 2006; Cohn et al., 2009; Montaldi and Mayes, 2010; Kafkas and Montaldi, 2012; Rugg et al., 2012; for a recent review see Rugg and Vilberg 2013).

The findings further stress that this contribution of the hippocampus in supporting recollection is not material-specific (for comparable results see Konkel and Cohen, 2009; Duarte et al., 2011; Staresina et al., 2011, 2013) but applies to at least three types of visual stimuli – scenes, faces and objects. With respect to face recollection, we were unable to explore the hippocampal response selectively for faces because of the small number of recollection responses to faces. This may stem from the high degree of similarity characterizing face stimuli, reaching the limits of pattern separation computation in the
hippocampus (Kim and Yassa, 2013), coupled with the use of a shallow encoding task, which
may have further hindered the formation of contextual associations at encoding. Nevertheless, hippocampal activity related to recollection relative to equally accurate familiarity was also found in the collapsed analyses across the three types of stimuli. However, this finding cannot exclude the possibility that the hippocampal activity is driven to a greater extent by recollection in the case of objects and scenes. Therefore, the proposed role of the hippocampus in supporting face recollection requires further investigation.

It is worth noting here that despite the aforementioned sensitivity of the PHC to familiarity-based recognition for both scenes and objects, a recollection effect within the posterior PHC for scene stimuli was also isolated when contrasting scene recollection to object recollection (Fig. 4D). Considering the dense anatomical connectivity between the hippocampus and the posterior PHC (Suzuki and Amaral, 1994) and the selective recollection response within the hippocampus, one possible interpretation is that scene/context-selective areas in the posterior PHC provide scene-selective input to the recollection signals computed in the hippocampus. The proposal that the recollection effect in the posterior PHC is probably stimulus-driven is further reinforced by the fact that this effect vanishes when contrasting scene recollection with scene familiarity. This means that the activation in the posterior PHC is not critical to recollection but stems from the presentation of familiar scene stimuli. Therefore, although this finding is compatible with the view that the PHC supports recollection of contextual details (Diana et al., 2007), this probably only occurs when familiar scene stimuli act as cues for recollection. Related to this finding, recollected stimuli were classified reliably in the MTL cortical areas (PRC, PHC and ERC) in the pattern classification analysis. As noted above for PHC, this finding may relate to the fact that most (possibly even all) recollected stimuli are potentially familiar too.
Conclusions

Taken together, familiarity-based recognition responses within the MTL were found to be material-specific, with PRC and ERC responding to object familiarity and the PHC to both object and scene familiarity. The amygdala was found to have a selective role in familiarity-based recognition for faces, whereas the adjacent hippocampus did not respond to stimulus familiarity for any of the three types of stimuli employed in the present study, in either the univariate and multivariate analysis. In contrast, the hippocampus was found to have a non-material specific role in recollection, even when compared to strength-matched familiarity. These findings illustrate the prominent role of the MTL neocortical areas and the amygdala in supporting familiarity-based recognition and show that this role is constrained by the inputs these structures receive and the content of stimulus category they process. Overall, the findings point to a degree of specialization within the MTL that respects whether familiarity or recollection is active and (secondarily) the kind of stimulus that is recognized as familiar.
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FIGURE LEGENDS

Figure 1. Experimental design and behavioral data. A) Design of the fMRI experiment and sequence of trials at encoding and retrieval (scanned). Three types of stimuli (scenes, objects and faces) were studied at encoding and were later presented at retrieval (along with unstudied stimuli). Participants were asked to provide familiarity (F1, F2, F3), new (N) and recollection (R) responses for each stimulus. B) Accuracy [Hits / (Hits + FAs)] collapsed across the three types of stimuli and separately for scenes, objects and faces. Recollection responses to faces were very rare and therefore are not reported separately. Error bars show the standard error of the mean.

Figure 2. Material-specific familiarity effects for (A) scenes, (B) faces and (C) objects in the MTL, the amygdala and the fusiform gyrus. Error bars show the standard error of the mean.

Figure 3. A) Multivariate pattern recognition classification accuracy and B) confusion matrices for the classification of strong familiarity (F3) responses within the perirhinal cortex (PRC), the entorhinal cortex (ERC), the parahippocampal cortex (PHC), the amygdala and the hippocampus for each stimulus type. Dashed lines in the graphs mark chance classification. Error bars indicate the standard error of the mean across participants. * P < 0.05; ** P < 0.01. P-values were obtained through permutation testing with 1000 permutations.

Figure 4. Recollection selective response in the hippocampus for A) scenes, B) objects and C) collapsed across scenes, objects and faces. D) Selective parahippocampal cortex effect for $R_{\text{scenes}}$ versus $R_{\text{objects}}$. Error bars show the standard error of the mean.
Figure 5. Multivariate pattern recognition classification accuracy for R and F3 responses in the hippocampus, the perirhinal cortex (PRC), the entorhinal cortex, the parahippocampal cortex (PHC) and the amygdala. Error bars indicate the standard error of the mean across participants. * $P < 0.05$; ** $P < 0.01$; *** $P < 0.001$; † $P = 0.08$ (trend). $P$-values were obtained through permutation testing with 1000 permutations.

Supplementary Figure 1. Multivariate pattern recognition classification accuracy for the classification of familiarity responses (collapsed across F1, F2 and F3) within the perirhinal cortex (PRC), the parahippocampal cortex (PHC), the amygdala and the hippocampus for each stimulus type. Dashed lines in the graphs mark chance classification. Error bars indicate the standard error of the mean across participants. * $P < 0.05$; ** $P < 0.01$. $P$-values were obtained through permutation testing with 1000 permutations.

Supplementary Figure 2. Overlap of the hippocampal activations for object and scene recollection assessed with two conjunction analyses. A) Conjunction between $R_{\text{Objects}} > F3_{\text{Objects}}$ and $R_{\text{Scenes}} > F3_{\text{Scenes}}$. B) Conjunction between $R_{\text{Objects}} > M_{\text{objects}}$ and $R_{\text{Scenes}} > M_{\text{scenes}}$. 
Figure 1. Experimental design and behavioral data. A) Design of the fMRI experiment and sequence of trials at encoding and retrieval (scanned). Three types of stimuli (scenes, objects and faces) were studied at encoding and were later presented at retrieval (along with unstudied stimuli). Participants were asked to provide familiarity (F1, F2, F3), new (N) and recollection (R) responses for each stimulus. B) Accuracy [Hits / (Hits + FAs)] collapsed across the three types of stimuli and separately for scenes, objects and faces. Recollection responses to faces were very rare and therefore are not reported separately. Error bars show the standard error of the mean.

Figure 1
170x132mm (300 x 300 DPI)
Figure 2. Material-specific familiarity effects for (A) scenes, (B) faces and (C) objects in the MTL, the amygdala and the fusiform gyrus. Error bars show the standard error of the mean.

Figure 2
206x94mm (300 x 300 DPI)
Figure 3. A) Multivariate pattern recognition classification accuracy and B) confusion matrices for the classification of strong familiarity (F3) responses within the perirhinal cortex (PRC), the entorhinal cortex (ERC), the parahippocampal cortex (PHC), the amygdala and the hippocampus for each stimulus type. Dashed lines in the graphs mark chance classification. Error bars indicate the standard error of the mean across participants. * P < 0.05; ** P < 0.01. P-values were obtained through permutation testing with 1000 permutations.
Figure 4. Recollection selective response in the hippocampus for A) scenes, B) objects and C) collapsed across scenes, objects and faces. D) Selective parahippocampal cortex effect for Rscenes versus Robjects. Error bars show the standard error of the mean.

Figure 4

209x132mm (300 x 300 DPI)
Figure 5. Multivariate pattern recognition classification accuracy for R and F3 responses in the hippocampus, the perirhinal cortex (PRC), the entorhinal cortex, the parahippocampal cortex (PHC) and the amygdala. Error bars indicate the standard error of the mean across participants. * P < 0.05; ** P < 0.01; *** P < 0.001; † P = 0.08 (trend). P-values were obtained through permutation testing with 1000 permutations.
Supplementary Figure 1. Multivariate pattern recognition classification accuracy for the classification of familiarity responses (collapsed across F1, F2 and F3) within the perirhinal cortex (PRC), the parahippocampal cortex (PHC), the amygdala and the hippocampus for each stimulus type. Dashed lines in the graphs mark chance classification. Error bars indicate the standard error of the mean across participants. * P < 0.05; ** P < 0.01. P-values were obtained through permutation testing with 1000 permutations.

Supplementary Figure 1
51x12mm (300 x 300 DPI)
Supplementary Figure 2. Overlap of the hippocampal activations for object and scene recollection assessed with two conjunction analyses. A) Conjunction between RObjects > F3Objects and RScenes > F3Scenes. B) Conjunction between RObjects > Mojects and Rscenes > Mscenes.

Supplementary Figure 2
128x61mm (300 x 300 DPI)