Behavioral/Cognitive

Multimodal Integration and Vividness in the Angular Gyrus During Episodic Encoding and Retrieval

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Much evidence suggests that the angular gyrus (AnG) is involved in episodic memory, but its precise role has yet to be determined. We examined two possible accounts within the same experimental paradigm: the “cortical binding of relational activity” (CoBRA) account (Shimamura, 2011), which suggests that the AnG acts as a convergence zone that binds multimodal episodic features, and the subjectivity account (Yazar et al., 2012), which implicates AnG involvement in subjective mnemonic experience (such as vividness or confidence). fMRI was used during both encoding and retrieval of paired associates. During study, female and male human participants memorized picture-pairs of common objects (in the unimodal task) or of an object-picture and an environmental sound (in the crossmodal task). At test, they performed a cued-recall task and further indicated the vividness of their memory. During retrieval, BOLD activation in the AnG was greatest for vividly remembered associates, consistent with the subjectivity account. During encoding, the same effect of vividness was found, but this was further modulated by task: greater activations were associated with subsequent recall in the crossmodal than the unimodal task. Therefore, encoding data suggest an additional role to the AnG in crossmodal integration, consistent with its role at retrieval proposed by CoBRA. These results resolve some of the puzzles in the literature and indicate that the AnG can play different roles during encoding and retrieval as determined by the cognitive demands posed by different mnemonic tasks.

Key words: angular gyrus; cued recall; episodic memory; multimodal integration; vividness

Significance Statement
We offer new insights into the multiplicity of processes that are associated with angular gyrus (AnG) activation during encoding and retrieval of newly formed memories. We used fMRI while human participants learned and subsequently recalled pairs of objects presented to the same sensory modality or to different modalities. We were able to show that the AnG is involved when vivid memories are created and retrieved, as well as when encoded information is integrated across different sensory modalities. These findings provide novel evidence for the contribution of the AnG to our subjective experience of remembering alongside its role in integrative processes that promote subsequent memory.

Introduction
The ventral posterior parietal cortex (vPPC), and the angular gyrus (AnG) in particular, have been associated with numer-

ous cognitive functions, including episodic memory. Although the vPPC is considered one of the most active regions during successful episodic retrieval (for review, see Wagner et al., 2005; Cabeza et al., 2008, 2012; Vilberg and Rugg, 2008; Shimamura, 2011; Levy, 2012; Rugg and King, 2018; Sestieri et al., 2017), patients with lateral parietal lesions can often successfully retrieve episodic memories and are not usually considered to be amnesic.

Two main explanations have been proposed to account for AnG activation during episodic retrieval. The first derives from findings of fewer “remember” responses, fewer high-confidence responses, and lack of richness, vividness, and specificity of retrieved episodic events in patients with parietal lesions (Berryhill et al., 2007; Simons et al., 2010; Hower et al., 2014). Under this account (Yazar et al., 2012), vPPC involvement is viewed in terms of subjective mnemonic abilities, with the AnG involved in one’s
own experience of episodic memory retrieval rather than in objective memory performance as expressed in response accuracy. This “subjectivity account” has been further supported by several studies in healthy individuals (Qin et al., 2011; Yazar et al., 2014; Richter et al., 2016).

The second account is the “cortical binding of relational activity” (CoBRA) theory (Shimamura, 2011), which suggests that, during retrieval, the vPPC (including the AnG) acts as a convergence zone that becomes increasingly involved after initial encoding to bind together episodic features that are represented in disparate neocortical regions. This account builds on anatomical features of the vPPC—its central location and connections to many neocortical regions (Seghier, 2013)—which make it well situated for binding episodic memories by establishing intermodular links across diverse event features. CoBRA critically predicts that tasks that demand the reinstatement of intermodal features such as voices with faces will depend more on the vPPC than tasks that depend on within-modality associations. This account has received support from several studies (Bonnici et al., 2016; Yazar et al., 2017), but other examinations of CoBRA were inconclusive, with multimodal integration deficits found in parietal lesion patients (Ben-Zvi et al., 2015) but no multimodal versus unimodal associative memory effects on parietal scalp ERPs recorded in healthy volunteers (Tibon and Levy, 2014a).

Even though, theoretically, these two accounts are distinct, in practice, they are not easily dissociated because the multisensory features comprising episodic representations enable us to experience the rich and vivid details that characterize remembering. To tease apart these accounts, we used a pair-associate learning paradigm, based on our prior work (Tibon and Levy, 2014a; Ben-Zvi et al., 2015). MRI of healthy volunteers was used during two learning tasks. In the unimodal task, stimulus pairs were two semantically unrelated object-pictures and, in the crossmodal task, stimulus pairs were an object-picture and a nameable but unrelated sound. At study, participants created a mental image of the two stimuli. At test, an object-picture was presented as the cue and participants recalled the associated target from study, either another picture or a sound, and indicated the vividness of their memory. Therefore, crossmodal “events” involved different modalities for separate objects rather than multimodal representations of the same object. Therefore, even though these events are multimodal, they are not necessarily more vivid.

To test the retrieval-focused predictions of CoBRA and the subjectivity accounts, we examined BOLD activation during test. The CoBRA account predicts greater parietal activity associated with objective retrieval (the difference between recall success and failure trials) in the crossmodal than the unimodal task regardless of the vividness of retrieved memories. Conversely, the subjectivity account predicts a linear pattern of activity, vivid recall > nonvivid recall > recall failure, that is independent of modality. In addition, because some previous findings from patients’ studies might be due to encoding rather than retrieval deficits, we also examined memory effects from the study phase. Finally, given recent claims regarding AnG laterality effects (Bellana et al., 2016), we investigated whether the left and right AnG display different memory-related activation patterns.

Materials and Methods

Participants

Twenty-four adults participated in the experiment (20 females, mean age = 25.83, SD = 4.02) and were reimbursed for their time. One participant who was not fully scanned due to a problem with the equipment and two participants who did not provide any responses in one of the experimental conditions (one with no success responses and another with no nonvivid responses) were excluded, leaving 21 participants (18 females) whose data were analyzed. Participants were recruited from the volunteer panel of the Medical Research Council Cognition and Brain Science Unit in Cambridge, United Kingdom. All were fluent English speakers, MR-compatible, had normal hearing and normal or corrected-to-normal vision, and were never diagnosed with attention deficit hyperactivity disorder, dyslexia, or any other developmental or learning disabilities. Participants provided informed consent for a protocol approved by a local ethics committee (Cambridge Psychological Research Ethics Committee reference PRE2016.055).

Materials

The materials for the experiment were previously used by Tibon and Levy (2014a), although some (~10%) were replaced to improve stimulus quality. Replaced stimuli were verified as recognizable by six pilot participants who did not take part in the main study. Auditory stimuli were 120 environmental sounds (e.g., animal sounds, tools, vehicles, etc.) downloaded from various internet sources and edited using Audacity audio editing software at 44 kHz, 16-bit resolution, and mono mode. They were all adjusted to the same amplitude level and edited to last 2 s. Twelve additional sounds were used for practice trials and examples. Visual stimuli were 360 color drawings of common objects obtained from various internet sources, including fruits and vegetables, tools, sporting goods, electrical and electronic devices, animals, food, vehicles, body parts, furniture, and clothing, each ~6–8 cm in on-screen size.

To form the various experimental conditions, several stimulus lists were created. Two lists of 40 pairs of stimuli were presented as the experimental list. Two additional lists served as target lists, one containing 120 pictures and the other containing 120 sounds. Each entry in the cue lists was pseudo-randomly assigned to an entry in the target lists so that all stimulus-pairs were semantically unrelated. Although, unavoidably, the target lists (one list of pictures and another of sounds) remained fixed for the crossmodal and unimodal conditions, two experimental lists were created to counterbalance the cues across participants: half of the participants viewed the cues from the first cue list with the auditory target and the cues from the second cue list with the visual target and vice versa for the other half. Each experimental list was further divided into three sublists to allow 1/3 of the stimuli to be presented three times (the other 2/3 were presented only once; see more details below), with the sublists used to counterbalance the repeated stimuli across participants.

Procedure

The paradigm used in the experiment is illustrated in Figure 1. The experiment consisted of two tasks, which took place on different days (mean days apart = 7.14, SD = 4.8): one was the crossmodal pair-associate learning and cued recall task and the other was the unimodal pair-associate learning and cued recall task. Task order was counterbalanced across participants. On arrival at the laboratory, participants were seated in a quiet room, where they signed an informed consent form and performed the unscanned part of the relevant task. Participants were told that they would be presented with pairs of stimuli (sound-picture pairs in the crossmodal task and picture-picture pairs in the unimodal task) and were instructed to study those pairs for subsequent retrieval by forming an association between the stimuli. They were told that, after 40 pairs had been presented, a test phase would follow in which cue pictures alone would be presented. They would then need to recall and say the name of the paired object portrayed in the picture (in the unimodal task) or of the paired sound/object making the sound (in the crossmodal task) that had accompanied that cue. Following these instructions, a headset with a microphone was provided, together with a practice block of five trials. During practice, the experimenter ascertained that the participant understood the nature of the associations that were to be generated for the stimulus pairs. After completion of the practice block, participants were told that the unscanned part of the task would be repeated twice with the same 40 pairs. They were further informed that, following the unscanned pair, they would be asked to perform the same task in the fMRI scanner, but for 120 pairs: 40 pairs that had just been studied and recalled at the unscanned part and 80 new pairs. We used this two-stage procedure in
which 1/3 of the stimuli are repeated thrice (twice in the unscanned part and once in the scanned part) and 2/3 of the stimuli are only shown once (in the scanned part) to keep our procedure as close as possible to that of Ben-Zvi et al. (2015), in which all stimuli were repeated several times, while also allowing enough recall failure trials for our contrasts of interest (pilots indicated that performance would be at ceiling if all trials were repeated more than once).

During the unscanned part of the task, participants viewed two (identical) study-test blocks of 40 stimulus pairs each. Each study trial started with a fixation cross displayed for an exposure time jittered across a range of 1–4 s. Stimulus pairs were then presented for 2 s, followed by a 7 s fixation cross. Participants were instructed to use this time to generate an association. Next, a screen with the text “easy/hard?” was shown for 2 s. During this time, participants were asked to press a green key if they found it easy to come up with an association or a red key if they found it difficult. The right index and middle fingers were used for these key-presses and finger assignment was counterbalanced across participants.

After all 40 study trials had been presented, the test phase started. Test trials also started with a 1–4 s jittered display time fixation cross. The cue picture was then presented alone for 2 s, followed by a 6 s fixation cross. Next, a slide with the text “pair associate?” was shown for 3 s and participants were asked to provide their verbal response while the slide was presented. They were encouraged to formulate their response while the fixation cross was presented so that, once the “pair associate?” slide appeared, they could provide their answer immediately. They were further told that if they could not recall the target object, then they should not try to guess. Instead, they should say “pass” when the “pair associate?” slide was displayed. Finally, a slide with the legend “vivid?” appeared for 2 s.
and participants were asked to press a green key if their memory was vivid or a red key if it was not. Finger assignment was counterbalanced across participants and matched the assignment during the study phase (such that the same green key was used for both “easy” and “vivid” responses, and the same red key was used for both “hard” and “not vivid” responses).

After completing the unscreened part, which included two study-test blocks, participants were provided with instructions for the MRI scan. They were told that, even though they would need to give verbal responses in the scanner, they should try to minimize head movements. They then performed the task in the MRI scanner with one block of 120 pairs (40 pairs that were studied during the unscreened part interleaved with 80 new pairs). Finally, after the scan, participants returned to the laboratory, where they performed a debriefing session. In this session, all target stimuli (either sounds in the crossmodal task or pictures in the unimodal task) were displayed again and participants were asked to type their names. The data from this session were used to subsequently eliminate a (negligible number of) trials in which certain participants were unable to provide a name (even if inaccurate) for the sound/picture that was presented.

fMRI acquisition
The same acquisition and preprocessing protocol was used for both visits (one visit for the crossmodal task and another for the unimodal task). MRI data were collected using a Siemens 3 T TIM TRIO system. Structural images were acquired with a T1-weighted 3D Magnetization Prepared RapId Gradient Echo (MPRAGE) sequence [repetition time (TR) = 2250 ms; echo time (TE) = 3.02 ms; inversion time (TI) = 900 ms; 230 Hz per pixel; flip angle = 9°; field of view (FOV) = 256 × 256 × 192 mm; GRAPPA acceleration factor = 2]. Functional images were acquired using an echoplanar imaging (EPI) sequence with multiband (factor = 2) acquisition. Volumes were acquired over 2 runs, one for the study phase (mean number of volumes = 1095.5, SD = 111.1) and one for the test phase (M = 1263, SD = 6.8). Each volume contained 76 slices (acquired in interleaved order within each excitation band) with a slice thickness of 2 mm and no interslice gap (TR = 1448 ms; TE = 33.4 ms; flip angle = 74°; FOV = 192 mm × 192 mm; voxel-size = 2 mm × 2 mm × 2 mm). Field maps for EPI distortion correction were also collected (TR = 541 ms; TE = 4.92 ms; flip angle = 60°; FOV = 192 × 192 mm).

Data preprocessing
Data were preprocessed using SPM12 (www.fil.ion.ucl.ac.uk/spm), automated in Matlab (version 8.0.0.783 R2012b; The MathWorks) with Automatic Analysis (AA) 5.0 (Cusack et al., 2015; https://github.com/rhodricusack/automaticanalysis). T1 anatomical images for each participant in each visit were coregistered to the Montreal Neurological Institute (MNI) space. For each participant in each visit, we performed coregistration (i.e., whether they were shown in the unscreened part of the task); (1) repeated: trials with correctly recalled targets, which were also shown during the unscreened part; (2) nonrepeated: trials with correctly recalled targets, which were shown for the first time during the scanned part; and (3) failure: all trials with a “pass” response.

The trials classification was corroborated by data obtained in the debriefing session. In cases in which the participants’ response did not match the name of the object, debriefing was used to determine the source of the mismatch. The first possible source of mismatch identification was that the participant erroneously identified the object during the study phase (e.g., an electrical buzz sounding like a buzzing bee) and then provided the same response at debriefing and test, but that response did not match the correct response. Importantly, in this case, although the object was misidentified, memory of the target stimuli was intact. We therefore classified the trial as a success trial (either “vivid” or “nonvivid”). The second possible source of a mismatch is memory failure; that is, the participant failed to retrieve the target and retrieved a different object instead, therefore providing a different response at debriefing and test. In this case, the trial was classified as a “false alarm.” False alarm trials, trials in which no response was collected during the scanning session, and trials in which the target was not named during the debriefing session were marked as “excluded” (see “fMRI acquisition” section).

Behavioral analyses. All statistical analyses were performed in R version 3.4.1 and RStudio version 1.0.153. After classifying the trials according to their responses, we analyzed the behavioral data from study and test. To test for task effects (differences between the unimodal and the crossmodal task) during the study phase, data were analyzed using a paired-samples t test to compare encoding difficulty (calculated as percentage of pairs for which an “easy” response was recorded at encoding) between the unimodal and the crossmodal tasks. To test for memory effects at test, accuracy data (number of trials for which vivid, nonvivid, and failure responses were obtained) were analyzed with a repeated-measures ANOVA, which included task (crossmodal, unimodal) and response type (vivid, nonvivid, failure) as within-subject factors, using the ez package (Lawrence, 2016). Whenever sphericity assumptions were violated, Greenhouse–Geisser-corrected p-values are reported. Main effects and interactions were decomposed with Bonferroni-corrected pairwise comparisons using the lsmeans package (Lenth, 2016). Degrees of freedom were corrected using the Satterthwaite method as implemented in the lmerTest package (Kuznetsova et al., 2017).

ROI analyses. To analyze ROI data, we used a mixed-models analysis that accommodates both within- and between-subject variability. This approach is particularly recommended for unbalanced data (an unequal number of trials in each condition; Tibon and Levy, 2015), which we had here due to the post hoc division of trials into vivid, nonvivid, and failure trials (see Tibon and Levy, 2014a,b, 2015 for a similar use of this approach). Importantly, rather than averaged estimates across participant/condition, the mixed-models used here require estimation of the BOLD response for each trial.

To estimate the BOLD response for each trial in the extracted time series data, these ROIs were defined based on the coordinates of peak left ANG activation from a previous univariate meta-analysis of episodic memory by Vilberg and Rugg (2008), −43, −66, 38, and its homologous location in the right hemisphere, 43, −66, 38. These coordinates were also used more recently in a study by Bonnici et al. (2016). For each participant, we extracted time series data from these ROIs by taking the first eigenvariate across voxels within a 6-mm-radius sphere centered on these coordinates and removing effects of no interest such as those captured by the motion regressors (see “Whole-brain univariate analysis” section below). The first eigenvariate captures the dominant time series across voxels in an ROI without assuming that all voxels express that time series equally (as assumed when simply averaging across voxels within an ROI).

To estimate the BOLD response for each trial in the extracted time series separately for study and test, we used the least-squares separate (LSS-N) approach (Mumford et al., 2014; Abdulrahman and Henson, 2016), where “N” is the number of conditions. LSS-N fits a separate general linear model (GLM) for each trial, with one regressor for the trial of interest and one regressor for all other trials of each condition. This implements a form of temporal smoothness regularization on the parameter estimates (Abdulrahman and Henson, 2016). The regressors were created by convolving a delta function at the onset of each stimulus with a canonical hemodynamic response function (HRF).
set of cosines that functioned to high-pass the model and data to 1/128 Hz (implemented to detrend the BOLD signal and to correct physiological noises) fit using restricted maximum likelihood. The estimated error autocorrelation was then used to “prewhiten” the model and data and ordinary least squares was used to estimate the model parameters.

The images for the parameter estimates for each of the three response types for each task and each phase were then entered into second-level GLM corresponding to a repeated-measures ANOVA, which treated subject as a random effect, and with phase (study, test), task (unimodal, crossmodal), and response type (vivid, nonvivid, failure) as repeated factors. SPMs were created of the T-statistic for the “recall success” contrast of interest [success (vivid + nonvivid) − failure] at study and test. The statistical threshold was set to $p < 0.05$ (FWE corrected for multiple comparisons across the whole brain) at the cluster level when the voxel-level threshold was set at $p < 0.0001$.

Results

Behavioral results

**Study phase**

A paired-samples $t$ test revealed a significant difference in task difficulty during study ($t_{(40)} = 3.63$, $p = 0.0017$, CI [.03 .0313]), with a greater proportion of easily formed associations in the unimodal task ($M = 0.69, SD = 0.19$) than in the crossmodal task ($M = 0.61, SD = 0.23$). This difference was accounted for in our subsequent ROI analyses (see below).

**Test phase**

The number of responses for each response type in the crossmodal and unimodal tasks are shown in Figure 2. Analysis of these data showed a significant main effect of task ($F_{(1,20)} = 31.89$, $p < 0.0001$, $\eta^2 = 0.65$), a significant main effect of response type ($F_{(2,40)} = 46.48, p < 0.0001, \eta^2 = 0.65$), and a significant interaction between these factors ($F_{(2,40)} = 5.11, p = 0.01, \eta^2 = 0.04$). Bonferroni-corrected pairwise comparisons showed greater number of vivid responses in the unimodal than in the crossmodal task ($t_{(42,67)} = 4.05, p = 0.0002$), but no difference between the tasks in the amount of nonvivid or failure responses. These results support our expectation that, when different modalities are associated with separate objects within the event rather than with the same object, crossmodal events are not necessarily remembered more vividly than unimodal events.

ROI mixed-effects analyses

For the main hypotheses under investigation, we focused on left and right AnG ROIs defined *a priori* from independent data. Figure 3 depicts contrasts of the beta values according to our *a priori* contrasts of interest averaged across left and right AnG. Table 1 shows the betas for each condition and ROI separately.

The estimated model showed a significant main effect of response type ($\chi^2_{(2)} = 64.2, p < 0.0001$), a significant interaction between phase and response type ($\chi^2_{(2)} = 18.1, p = 0.0001$), a significant interaction between phase and laterality ($\chi^2_{(1)} = 21.33, p < 0.0001$), a significant interaction between task and laterality ($\chi^2_{(1)} = 8.14, p = 0.004$), and a significant 3-way interaction among phase, task, and response type ($\chi^2_{(2)} = 11.27, p = 0.004$).

Because laterality did not interact with response type ($\chi^2_{(2)} = 0$), the interaction term was removed from the model. For the remaining effects, separate GLMs were estimated for each AnG ROI, each task, and each phase. Each of these regressors was generated with a delta function convolved with a canonical HRF. Six subject-specific movement parameters were also included to capture residual movement-related artifacts. The GLM was fit to the data in each voxel. The autocorrelation of the error was estimated using an AR(1)-plus-white-noise model, together with all possible interactions between these factors as the fixed part of the model and subject-specific slopes for each factor (see discussion by Barr et al., 2013 and Bates et al., 2015 on estimation and convergence problems of the maximal model effects model) plus a subject-specific intercept as the random part of the model. We fitted the model using restricted maximum likelihood in R package lme4 (Kuznetsova et al., 2017) with the following formula where the asterisk indicates all possible interactions and “(. . . | x)” indicates random effects:

\[
\text{betas} \sim \text{phase} * \text{task} * \text{response type} * \text{laterality} + (1 + \text{phase} + \text{task} + \text{response type} + \text{laterality} | \text{subject})
\]

To test the predictions of the CoBRA and the subjective accounts, we used two predefined contrasts, which we applied separately for each phase and ROI mixed-effects analyses. Whole-brain univariate analyses. We further conducted whole-brain analyses to ensure that we replicated the basic encoding and retrieval mnemonic effects found in previous studies. SPM12 and AA were used to construct GLMs for each participant separately for each run (study, test) and for each task (crossmodal, unimodal; which took place at different visits). These first-level GLMs included three separate regressors for each response type of interest (vivid, nonvivid, failure) and a regressor for excluded responses (either false alarms or unnameable targets, see above). These regressors were modeled at the onset of stimulus presentation. For the study run, the GLM further included a regressor for the motor response to the “easy/hard” slide (locked to the motor response). For the test run, the GLM further included a regressor for the motor response to the “vivid” slide (locked to the motor response) and a regressor for the verbal response (at the onset of the “pair-associate” slide). Each of these regressors was generated with a delta function convolved with a canonical HRF. Six subject-specific movement parameters were also included to capture residual movement-related artifacts. The GLM was fit to the data in each voxel. The autocorrelation of the error was estimated using an AR(1)-plus-white-noise model, together with all possible interactions between these factors as the fixed part of the model and subject-specific slopes for each factor (see discussion by Barr et al., 2013 and Bates et al., 2015 on estimation and convergence problems of the maximal model effects model) plus a subject-specific intercept as the random part of the model. We fitted the model using restricted maximum likelihood in R package lme4 (Kuznetsova et al., 2017) with the following formula where the asterisk indicates all possible interactions and “(. . . | x)” indicates random effects:

**betas** ~ **phase** * **task** * **response type** * **laterality** + (1 + **phase** + **task** + **response type** + **laterality** | **subject**)

Figure 2. Number of responses for each response type in the crossmodal task (dark gray) and the unimodal task (light gray). Error bars indicate SEs for each condition separately. ***$p < 0.001$.**
1.76, $p = 0.42$), we collapsed across this factor when computing our planned comparisons.

The first planned comparison confirmed a greater recall success effect for crossmodal than unimodal memories during the study phase ($\chi^2(1) = 8.77, p = 0.006$), but not during the test phase ($\chi^2(1) = 0.19, p = 1$). The second planned comparison confirmed a linear trend whereby vivid > nonvivid > fail during both study ($\chi^2(1) = 29.14, p < 0.0001$) and test ($\chi^2(1) = 72.07, p < 0.0001$).

The following analyses reported in this section were conducted to further explore these results and to rule out alternative explanations. First, we investigated whether these results are also observable in the broader AnG beyond the 6 mm sphere based on the functional activation of Vilberg and Rugg (2008). To this end, we repeated the analysis on a larger, anatomical ROI from the Anatomical Automatic Labeling atlas. This analysis revealed the same pattern as before: a significant 3-way interaction among phase, task, and response type ($\chi^2(2) = 10.65, p = 0.005$) coupled with a greater recall success effect for crossmodal versus unimodal pairs at study ($\chi^2(1) = 29.14, p < 0.0001$) and test ($\chi^2(1) = 72.07, p < 0.0001$). Second, to ensure that the interaction between modality and response type at study cannot be explained by difficulty in forming associations (as our behavioral analysis indicated greater difficulty in the crossmodal vs the unimodal task), we ran an additional analysis, which included the difficulty rating provided for each trial during study as a covariate in our model. Despite the addition of the covariate, the 3-way interaction remained signifi-
icant ($\chi^2_{(2)} = 11.28, p = 0.004$) and the same pattern as before was confirmed by our planned comparisons; that is, a greater recall success effect for crossmodal than unimodal memories at study phase ($\chi^2_{(1)} = 8.22, p = 0.008$), but not at test ($\chi^2_{(1)} = 0.58, p = 0.9$), as well as a linear trend during both study ($\chi^2_{(1)} = 25.46, p < 0.0001$) and test ($\chi^2_{(1)} = 65.28, p < 0.0001$).

Third, as was mentioned before, in the current experiment, some of the trials were repeated several times whereas others were only shown once. This was done to keep the procedure similar to that of Ben-Zvi et al. (2015) while also allowing for a sufficient number of failure trials. This particular design raises the possibility that the reported effects are driven by repetition rather than by the proposed cognitive processes (i.e., multimodal integration and vividness). To address this potential confound, we ran the model again after excluding trials that were also shown during the unscanned part of the task, thus limiting our analysis to the 2/3 of the trials that were only presented once. Importantly, this analysis confirmed our previous results: the 3-way interaction remained significant ($\chi^2_{(2)} = 9.36, p = 0.009$), a greater recall success effect for crossmodal than unimodal memories was observed at the study phase ($\chi^2_{(1)} = 8.24, p = 0.008$), but not at test ($\chi^2_{(1)} = 0.12, p = 1$), and a linear trend was observed during both study ($\chi^2_{(1)} = 12.58, p < 0.001$) and test ($\chi^2_{(1)} = 67.45, p < 0.0001$). This indicates that repetition cannot account for the effects observed in the current study.

Finally, the repetition embedded in current design provides an opportunity to explore repetition effects in the AnG, thus tapping into CoBRA’s suggestion of greater vPPC involvement in episodic binding with the passage of time. One drawback of our design in addressing this issue is that repetition and vividness are highly correlated (inevitably, memory for repeated stimuli tends to be more vivid than memory for stimuli that were only experienced once). Nevertheless, we still sought to determine whether there are any residual repetition effects that cannot be explained by vividness. Given that the first study episode of repeated items occurred earlier than the first study episode of nonrepeated ones according to the view that consolidation can begin immediately following initial encoding (Liu et al., 2018), then CoBRA would predict greater AnG activation for repeated versus nonrepeated items. To explore this potential effect, we added a “repetition” factor to our model and ran a model with repetition (nonrepeated, repeated), response type (vivid, nonvivid), phase (study, test), task (crossmodal, unimodal), laterality (left, right), and all possible interactions between these factors as the fixed part of the model and with subject-specific intercept and slope for each factor as the random part of the model (because “failure” trials overlap for the “vividness” and the “repetition” factor, these trials had to be excluded to achieve model convergence). In general, as predicted by CoBRA, repeated trials were associated with greater AnG activation than nonrepeated trials ($\chi^2_{(1)} = 34.15, p < 0.0001$). However, the repetition effect interacted with other factors in our design, including a 3-way interaction among phase, repetition, and lateralization ($\chi^2_{(1)} = 10.51, p = 0.001$) and a 4-way interaction among repetition, response type, phase, and task ($\chi^2_{(1)} = 5.03, p = 0.024$). Given that the latter interaction is of theoretical interest, a contrast of repeated versus nonrepeated trials was explored separately for each response type, phase, and task. During study, this revealed a significant repetition effect for nonvivid and vivid unimodal memories ($\chi^2_{(1)} = 11.99, p = 0.004$; $\chi^2_{(1)} = 20.5, p < 0.0001$, respectively) and for vivid unimodal memories ($\chi^2_{(1)} = 43.6, p < 0.0001$). During test, this analysis revealed a significant repetition effect for nonvivid and vivid unimodal memories ($\chi^2_{(1)} = 20.01, p < 0.0001; \chi^2_{(1)} = 9.87, p = 0.013$, respectively).

**Whole-brain univariate analyses**

Looking beyond the AnG, to confirm that our paradigm replicated the typical engagement of regions across the brain during episodic memory encoding and retrieval, we conducted a whole-brain analysis (results are shown in Fig. 4). We examined the activation to successful memory trials (collapsed across vividness) relative to failure trials at both study and test. For both phases, this contrast yielded increased BOLD response in multiple brain regions that have previously been linked with episodic memory (see Table 2 for full results).

**Discussion**

The current study tested two dominant accounts of the activation in parietal cortex consistently found in neuroimaging studies of episodic memory, the CoBRA account and the subjectivity account, and revealed a dissociation between encoding and retrieval of memory-related activations in the AnG. During retrieval, we found a linear trend as predicted by the subjectivity account: BOLD activation was greatest for vividly remembered associates and least for associates that failed to be recalled. Contrary to the prediction of the CoBRA account, however, which predicts greater activation when information from multiple modalities is reinstated at retrieval, the magnitude of the objective recall success effect did not differ between the unimodal and crossmodal tasks. During encoding, activation associated with subsequent memory showed the same linear pattern predicted by the subjectivity account, but in addition, this pattern was now moderated by task, with greater activations associated with subsequent recall success in the crossmodal than unimodal task. Memory effects at encoding and retrieval did not interact with lateralization, suggesting that, at least with the current task, the left and right AnG play a similar role. Together, these results support the subjectivity account, but not the CoBRA account as currently specified, which implicates the involvement of the AnG in multimodal reinstatement specifically at retrieval; rather, in our current data, the involvement of AnG in multimodal processing was restricted to encoding.

Our finding that AnG activity scales with ratings of vividness during retrieval is consistent with prior studies that have found AnG to be sensitive to qualitative characteristics of memory such as vividness, confidence, and precision (Yazar et al., 2014; Kuhl and Chun, 2014; Richter et al., 2016), supporting the notion that the AnG is involved in the subjective experience of remembering. We show further here that the linear pattern associated with vivid remembering in the AnG also occurs at encoding, suggesting that it is also involved in the construction of representations that enable vivid subsequent memories. We note that this pattern of encoding-related activity, greater AnG activation for subsequently remembered versus forgotten memories, is inconsistent with some previous studies (Daselaar et al., 2009; Lee et al., 2017), which reported negative subsequent memory effects in the AnG (i.e., greater activation for subsequently forgotten vs remembered memories). Nevertheless, an extensive meta-analysis (Uncapher and Wagner, 2009) indicated that both positive and negative subsequent memory effects are observed in the vPPC. Although a comprehensive explanation of this discrepancy is not yet available, prior results point to two possible accounts for the positive subsequent memory effect found in the current study. First, Uncapher and Wagner (2009) suggest that retention interval is a crucial predictor of the directionality of subsequent mem-
Figure 4. Brain regions showing stronger BOLD response for successful versus failure recall at study (left) and test (right). $p < 0.05$ FWE cluster-level corrected with voxel-level threshold at $p < 0.0001$ (uncorrected). Data are shown on sagittal, coronal, and axial slices of the group-averaged brain ($n = 21$).

Table 2. Regions of increased BOLD activation during study and test in the recall success contrast of interest

| Lat | Region                      | Peak x, y, z | Cluster size (voxels) | T value |
|-----|-----------------------------|-------------|----------------------|---------|
| Study |                               |             |                      |         |
| R   | Precuneus                   | 6, −68, 46  | 1242                 | 6.19    |
| R   | Angular gyrus               | 44, −54, 54 | 515                  | 5.01    |
| R   | Middle temporal gyrus       | 66, −24, −8 | 80                   | 5.02    |
| R   | Middle cingulate gyrus      | 4, −18, 30  | 214                  | 4.87    |
| R   | Middle frontal gyrus        | 32, 66, 2   | 225                  | 4.52    |
| L   | Angular gyrus               | −38, −62, 42| 955                  | 5.9     |
| L   | Middle temporal gyrus       | −64, −38, −8| 348                  | 5.63    |
| L   | Middle frontal gyrus        | −30, 58, 0  | 216                  | 4.65    |
| Test |                               |             |                      |         |
| R   | Cerebellum                  | 38, −56, −38| 1376                 | 7.13    |
| R   | Angular gyrus               | 42, −50, 20 | 1648                 | 5.62    |
| R   | Superior frontal gyrus      | 22, 30, 48  | 482                  | 5.19    |
| R   | Caudate                     | 16, 2, 26   | 172                  | 5.1     |
| R   | Postcentral gyrus           | 50, −10, 26 | 247                  | 5.05    |
| R   | Hippocampus                 | −28, −34, −4| 74                   | 4.99    |
| R   | Middle frontal gyrus        | 58, −40, −14| 98                   | 4.7     |
| L   | Supramarginal gyrus<sup>a</sup> | −50, −40, 42| 5356                 | 7.9     |
| L   | Angular gyrus               | −50, −62, 32|                      |         |
| L   | Superior frontal gyrus      | −20, 26, 56 | 5047                 | 7.76    |
| L   | Precuneus                   | −6, −52, 30 | 4494                 | 7.69    |
| L   | Inferior frontal gyrus      | −54, 12, 8  | 1882                 | 6.96    |
| L   | Hippocampus                 | −26, −28, −10| 321                 | 6.57    |
| L   | Supplementary motor cortex  | −4, 2, 62   | 123                  | 5.95    |
| L   | Putamen                     | −12, 8, −12 | 135                  | 5.88    |
| L   | Thalamus                    | −8, −28, 8  | 84                   | 5.04    |
| L   | Cerebellum<sup>b</sup>      | −18, −64, −26| 75                  | 4.99    |
| L   |                             | −36, −74, −42| 236                 | 4.89    |

<sup>a</sup>Cluster included local maxima in the supramarginal gyrus and the angular gyrus.

<sup>b</sup>Two local maxima were observed in the left cerebellum at test.
and were subsequently asked to retrieve unimodal sources (e.g., which side + which location) or crossmodal sources (e.g., which side + male or female voice) of the studied objects. Following stimulation, participants’ ability to retrieve information was reduced for crossmodal but not for unimodal sources. In the second study, by Bonnici et al. (2016), participants memorized audio clips, visual clips, and audio-visual clips presenting different objects (e.g., a train). At test, they were given a verbal cue for the to-be-remembered object (e.g., the word “train”) and were asked to recall the associated clip as vividly as possible. This study showed greater AnG activation when crossmodal than unimodal memories were retrieved coupled with above-chance classification accuracy of individual crossmodal (but not unimodal) memories in the AnG. These previous evidence of AnG involvement in multimodal retrieval effects suggest that, although our current results seem to dissociate between encoding and retrieval of memory-related activations in AnG, a simple distinction between memory stages might be too simplistic. Instead, we suggest that what determines AnG involvement is not the memory stage per se (encoding vs retrieval), but rather the specific combination of cognitive demands posed by the task. Therefore, unlike CoBRA, which predicts AnG involvement whenever multimodal elements are reinstated, we suggest that such activation might be further constrained by other task characteristics, such as the type of information that is retrieved and the mnemonic cues that are provided (e.g., it might be that AnG activation is only triggered when the retrieved item itself contains multimodal information or when source/contextual information rather than core item information is being retrieved). This suggestion is consistent with the component process model of memory (Witherspoon and Moscovitch, 1989; Moscovitch, 1992), which posits that numerous different processing components associated with distinct brain regions are recruited in various combinations by different memory tasks. Therefore, different task demands would involve distinct process-specific alliances (i.e., transient interactions between several brain regions; Cabeza and Moscovitch, 2013; Moscovitch et al., 2016). Even though some aspects of our results seem to be inconsistent with previous studies that used different tasks (Bonnici et al., 2016; Yazar et al., 2017), our findings do converge across the three studies that used the same paradigm (i.e., the current study, Ben-Zvi et al., 2015, and Tibon and Levy, 2014a), where task demands were kept similar. Importantly, by elucidating the nature of process-specific alliances, future studies can provide a more decisive and fine-grained account of AnG involvement in multimodal processing.

In addition to the main processes investigated here, our study offers some ancillary insights regarding repetition effects in the AnG. Greater activity for repeated versus nonrepeated items was consistently observed during the study phase. This finding is consistent with CoBRA’s suggestion that vPPC involvement begins after initial encoding and increases as time passes. One exception in which this repetition effect was not significant at study was the case of nonvivid unimodal memories. We speculate that, in this case, repeated trials (subsequently classified as nonvivid) were forgotten following their initial presentation and were therefore experienced as new during the scanned part of the task. This would make them more similar to nonrepeated trials and possibly eliminate the repetition effect (why the same pattern was not observed in the crossmodal task is unclear, but might be related to a difference in the criterion set for trial classification as vivid or nonvivid in this task). During retrieval, repetition effects were observed in the unimodal task, but not in the crossmodal task. CoBRA suggests that vPPC acts as a way of offloading reliance on MTL bindings. The retrieval repetition effect might therefore indicate that responsibility for unimodal memories rapidly shifts from MTL to vPPC, whereas crossmodal memories require prolong MTL binding before vPPC involvement. Importantly, however, the exploration of repetition effects goes beyond the original purpose of our design and therefore these suggested interpretations of repetition effects should be treated with caution.

In the current study, unimodal associations were composed of two visual stimuli and crossmodal associations were composed of a visual and an auditory stimulus. We chose these materials to allow simultaneous presentation of the stimuli (given that simultaneous processing of two auditory stimuli, for example, is perceptually challenging). However, this means that a limitation of the current study is that we cannot be certain that our conclusions generalize to other kinds of materials. It could be, for example, that the AnG is selectively involved in multimodal integration of audio-visual associations, but not in the integration of information deriving from other sensory modalities. Although we have no reason to assume that audio-visual associations are unique in this sense, future studies can use a similar paradigm to test other forms of within- and between-modality combinations. Another potential limitation of the current study is the uneven distribution of female (n = 18) and male (n = 3) participants. Although we do not expect the mnemonic processes addressed in this study to differ between genders, generalization of our findings to males should be done cautiously.

In summary, the results of the current study show that the AnG is involved in multiple mnemonic processes during both encoding and retrieval. They provide a straightforward answer to the puzzle arising from our previous findings (Ben-Zvi et al., 2015 versus Tibon and Levy, 2014a) by suggesting that the AnG is involved in (at least) two memory-related processes: multimodal integration at encoding and construction of representations that enable vivid recall during both encoding or during retrieval. Based on current and prior evidence (Bonnici et al., 2016; Yazar et al., 2017), it is plausible that the extent of AnG activation in mnemonic processes is determined by specific cognitive demands posed by the task (Moscovitch et al., 2016). Future studies might systematically manipulate such demands within the same experimental paradigm and examine activation during encoding and retrieval possibly using a more fine-grained vividness rating (see also Richter et al., 2016). Nonetheless, our current results provide an important step toward clarification of the complexities regarding AnG involvement in episodic memories.

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