When Does Sleep Affect Veridical and False Memory Consolidation? A Meta-Analysis

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Abstract

It is widely accepted that sleep aids in the encoding, consolidation and retrieval processes involved in memory processing, however, the conditions under which sleep influences memory may be substantially constrained. In a meta-analysis, we examined the effect that sleep has on both veridical (accurate) and false memory consolidation, in studies using the Deese-Roediger-McDermott (DRM) paradigm for memory of thematically-related words. The meta-analysis revealed that, whereas there was no overall effect of sleep on either accurate or false memories, the effect of sleep on memories was moderated by two constraints. First, sleep effects were influenced by the number of words within each themed word list, relating to differences in processing the associative network of related words. Second, sleep effects were greater in recall than recognition tests. Thus, whether sleep consolidation increased or decreased DRM veridical or false memory effects depended on specific features of the memory task.
When Does Sleep Affect Veridical and False Memory Consolidation? A Meta-Analysis

Sleep benefits both the encoding and retrieval processes involved in memory consolidation, improving both declarative and procedural memory compared to the same time spent awake (Rasch & Born, 2007; Rasch & Born, 2013; Stickgold, 2005; Walker & Stickgold, 2006). The Active Systems Consolidation hypothesis (Diekelmann & Born, 2010; Marshall & Born, 2007; Rasch & Born, 2013), suggests that information and events that we are exposed to during wakefulness are encoded initially in the hippocampus and neocortical systems. Consolidation during sleep then leads to repeated reactivation of these encoded memory representations, leading to an integration of selective information to the neocortex, where the memory is established in the long-term store (Lewis & Durrant, 2011). Substantial evidence supports this theory, for example declarative memory for word pairs has been found to be greater after a delay including a period of sleep compared to wake (Gais & Born, 2004; Plihal & Born, 1997; Wilson, Baran, Pace-Schott, Ivry, & Spencer, 2012).

Several studies have also tested the hypothesis that sleep not only affects processing and consolidation of previously experienced material, but also impacts the formation of false memories. The Deese-Roediger-McDermott (DRM) paradigm (Roediger & McDermott, 1995) has been extensively used to test when unseen, related information, termed false memories, is activated in memory. In this paradigm, participants are exposed to lists of semantically related words (eg., bed, dream, tired, snooze, yawn, etc.), and are asked to recall or recognise words previously seen in the initial lists. Words are categorized as either those that appeared in the initial lists (old words), words that did not appear in the lists but were closely related, known as lure
words (e.g., sleep in the above list), or unseen, unrelated words (new words). Participants are more likely to recall, or identify as previously seen, lure words than new words, demonstrating the false memory effect (McDermott, 1996; Roediger & McDermott, 1995; Roediger, Watson, McDermott, & Gallo, 2001).

Whilst evidence for sleep’s effect on veridical memory performance has been widely replicated, the question of whether sleep has an effect on DRM false memories remains. Potential inconsistencies in results emerge between tests of recall, where false memories seem to be enhanced by sleep (Diekelmann, Born, & Wagner, 2010; Payne et al., 2009), and tests of recognition, where sleep has been observed to either enhance, have no effect, or even reduce false memories (Diekelmann, Landolt, Lahl, Born, & Wagner, 2008; Fenn, Gallo, Margoliash, Roediger, & Nusbaum, 2009; Monaghan, Shaw, Ashworth-Lord, & Newbury, 2017). The activation/monitoring framework (Collins & Loftus, 1975) provides one possible explanation for the differences found between DRM tests of recall and recognition. The framework proposes that during tests of recognition, monitoring cues are activated when the words are presented to participants, allowing for the suppression of related but unseen words (Watson, McDermott, & Balota, 2004). During tests of recall, these monitoring cues are not available, and so a greater number of associated words are activated. This leads to greater false memory in tests of recall than recognition. Sleep has been found to improve source-monitoring abilities (Johnson, Hashtroudi, & Lindsay, 1993), and therefore improves the ability to reject unseen, related items during tests of recognition to a greater extent than during recall.

This difference in memory performance between tests of recall and recognition has been suggested in a meta-analysis of only a small number of studies that were published at the time (Chatburn, Lushington, & Kohler, 2014). A small non-
significant effect of sleep on false recognition was found, whereas false recall led to a large significant increase in false memory. However, this study examined the overall effect of only four studies in total, two studies on false recognition, and two on false recall. Therefore, the reliability of the effect of sleep on both false recall and recognition is still under review. To address this, the current meta-analysis includes a larger sample of DRM studies, with five individual experiments examining the effect of sleep on false recall, and eight experiments investigating false recognition. This allows for a more detailed exploration as to the effects of the two methods of testing, and a greater understanding as to whether the effect of sleep on false memories does in fact reliably differ between tests of recall and recognition.

The larger number of experiments analysed in this meta-analysis also permits investigation of other potential moderator variables that may contribute to the effect of sleep on memory consolidation and the production of false memories within the DRM paradigm. In particular, we can determine whether the number of words in each list and the total number of lists that participants are required to remember influences the false memory effect. Using the DRM paradigm in a standard memory test (so not testing the effect of sleep), Robinson and Roediger (1997) investigated the effect of varying list lengths on false recall and recognition. They found that increasing list length led to increases in both false recall and false recognition. Robinson and Roediger (1997) suggested that a larger number of words in each list increases the opportunity for participants to develop associations between the words, and therefore primes a larger number of unseen, related words during testing.

A possible explanation for generation of false memories in DRM tests is that of spreading activation (Collins & Loftus, 1975). Word lists that participants are exposed to can activate unseen words that are similar in meaning to previously seen
words. The Associative Activation Theory (AAT) of false memories suggests that these lure words are activated due to their similarity or association with the seen words (Howe, Wimmer, Gagnon, & Plumpton, 2009; Roediger et al., 2001). Lists with greater strength of semantic association with the critical lure elicit increased false memories than those with weaker associations, due to spreading activation among associates within semantic memory (Gallo & Roediger, 2002). Alternatively, Fuzzy Trace Theory (Payne et al., 2009) argues that false memories are a consequence of participants determining the gist or general theme of a list, and then activating all words related to that general meaning (Howe & Wilkinson, 2011). The mechanism of gist generation could again be due to spreading activation, with the theme generated as a consequence of interactive activation among associated words.

If sleep leads to greater spread of activation of previously seen word lists due to AAT or FTT (as proposed by Cai, Mednick, Harrison, Kanady, Mednick & Albright, 2009; Sio, Monaghan & Ormerod, 2013), then we should expect to see an increase in false recall and recognition of lure words after sleep in comparison to wakefulness. These theories raise predictions about the extent to which manipulating the density of inter-relations between words in a thematically-related list affects the role of sleep in consolidation. A longer list of related words is more densely interconnected (Robinson & Roediger, 1997), and so spreading activation will occur to a greater extent for both sleep and wake groups equally. Thus, the benefit of sleep-related spreading activation is less likely to be detected than in a shorter list of related words, where the lure word concept receives only weak activation from a small set of related words within semantic associative memory (see Shaw & Monaghan, 2017, for a similar argument related to hemispheric processing). Hence, list length may be a critical factor in determining whether veridical and false memories are promoted by
sleep. Indeed, previous research suggests that sleep is more beneficial when task difficulty increases, for both motor skills tasks (Stickgold & Walker, 2004), and problem solving tasks (Sio et al., 2013). If increasing the number of words in each list leads to closer associations and so easier access to semantically similar lure words, then we would expect sleep to increase false memories for studies with fewer words in each list, as activation of the lure word, or the theme, is more difficult to accomplish, so greater spreading activation is required across semantic networks.

Similar principles could also be expected to apply to the number of different lists that participants are exposed to. For instance, source monitoring is likely to become more difficult with larger numbers of lists, thereby increasing the likelihood of false memories, and decreasing veridical memory. Spreading activation across a large number of distinct thematic lists may also mitigate the potential effect of sleep on the generation of false memories.

The modality of presentation of word lists has also been found to affect the formation of DRM false memories. Previous research suggests differences in performance following visual compared to auditory presentation of word lists. For both tests of recall and recognition, research indicates a significant reduction in the false memory effect when words are presented visually (Kellogg, 2001; Smith & Hunt, 1998). However, this difference in the effect of modality on false memory performance has been found to only be significant in those participants with higher working memory capacity (Smith & Engle, 2011). This difference in performance between visually and auditorially presented word lists was not found for veridical memory (Smith & Engle, 2011; Smith & Hunt, 1998). It is therefore of interest to assess modality as a potential moderator in the current meta-analysis.
The emotionality of to-be-remembered word lists may also influence the size of the effect of sleep on both accurate and false memories. Research indicates an increase in overall memory performance for information with positive or negative emotional valence (Adelmann & Estes, 2013; Kensinger & Corkin, 2003). Further, emotionality of word lists has also been found to increase false recognition in DRM tests (Howe, Candel, Otgaar, Malone & Wimmer, 2010; Sharkawy, Groth, Vetter, Beraldi & Fast, 2008), however possible differences arise between lists of negative and positive valence, with an increase in false recognition of negative word lists, and a decrease in false recognition of positive lists, compared to lists rated as neutral (Brainerd, Stein, Silveira, Rohenkohl & Reyna, 2008). The effect of emotionality on false recall is less clear. Bauer, Olheiser, Altarriba and Landi (2009) suggest an increase in false recall for emotional word lists, whereas Howe et al., (2010) suggest a reduction in false recall for emotional compared to neutral word lists.

Sleep is suggested to further enhance this bias for the consolidation of emotional information, with studies indicating a role of rapid eye movement (REM) sleep specifically in the processing of emotional memories (Carr & Nielsen, 2015; Goldstein & Walker, 2014). Cai et al., (2009), and Carr and Nielsen (2015) suggest that REM sleep increases spreading activation, hence performance differences may be evident between emotional and neutral word lists after sleep. We thus tested emotionality of word lists as a potential moderator in the current meta-analysis, to assess whether emotionality leads to an enhanced effect of sleep compared to wake on both veridical and false memories.

This larger set of studies included in a meta-analysis of DRM sleep-related effects means we can also assess daytime nap versus overnight sleep effects on veridical and false memories. If sleep leads to greater spreading activation to semantic
associates (Collins & Loftus, 1975), then we would expect an increase in time spent asleep to result in improved veridical performance as well as enhanced false memories. Since Cai et al., (2009) suggest that REM sleep, which occurs to a greater degree in the latter half of a night’s sleep, increases spreading activation, performance differences between sleep and wake groups may be more significant with overnight sleep than a short nap. Furthermore, Payne et al. (2009) found a negative correlation between veridical recall and slow wave sleep (SWS), indicating reduced veridical recall performance with increasing SWS, again suggesting that differences between the sleep and wake groups may be more significant during a longer period of sleep compared to a daytime nap.

In this current meta-analysis, we therefore aimed to analyse what effect sleep has on both accurate and false memory in DRM tests. We included six potential moderator variables, and analysed the possible effect that these may have as constraints on effects of sleep on memory consolidation: (1) whether the memory task is recall or recognition testing, (2) the number of words in each list, (3) the number of different lists learned, (4) whether words were presented auditorially or visually, (5) emotionality of the lists, and (6) whether the study was an overnight or nap study.

Analysing sleep effects on old, new, and lure words individually is useful for formulating comparisons between recall and recognition tests. However, in recognition tests, any observed changes in accuracy as a consequence of sleep could be due to changes in discriminability between word types or changes in response biases to respond yes more or less often. We therefore also used Signal Detection measures to distinguish the overall sensitivity or discriminability (d’) and response bias (C) between sleep and wake groups for the studies testing recognition memory. We distinguished true recognition, defined as differences in responses to old words
and new (unrelated) words, and false recognition, defined as differences in responses to lure words and new words. We hypothesised that sleep groups would have a larger discriminability and response bias score than wake groups for true recognition, which would indicate that the sleep groups are more likely to correctly accept old words as previously seen, and accurately reject new words as unseen. If so, then this would provide evidence in support of a positive role of sleep on memory consolidation and improving accuracy of memory. In contrast, the effects of sleep on false recognition are still under review, and so we might expect to see a larger discriminability and response bias score for the sleep groups if sleep increases false recognition (Monaghan et al., 2017), larger scores for the wake groups if sleep reduces false recognition (Fenn et al., 2009), or no difference in discriminability and response bias if sleep does not influence false recognition (Diekelmann et al., 2008).

Method

To collect the relevant data, we conducted searches in both Scopus and Web of Science [23-06-2017], using the keywords “sleep OR nap AND false memories”. Scopus produced 113 results, and Web of Science produced 139 results. Our next step was then to check for duplicates, yielding a total of 169 unique entries. An additional 2 papers from our own research lab were also included in the final analysis, although these were not produced during the main searches due to being submitted for review or in preparation at the time of the searches. These entries were then screened using the following inclusion criteria: (1) Behavioural studies conducted with adult participants, who were (2) exposed to DRM word lists and (3) asked to take part in a recall or recognition task (4) after a period of sleep (which could be overnight or a nap), with (5) a wake group comparison condition. This screening led to the inclusion
of nine papers in total, with some papers containing multiple experiments (13 individual experiments with a total of 596 participants overall; see Table 1 for summary data and moderators for each experiment).

Meta-Analysis

The effect sizes reported are the standardized mean difference in proportion of responses to each word type given as old (in the recognition tests) and proportion recalled (in the recall tasks) between the sleep and wake group, with positive values meaning increased proportion of responses in the sleep group than the wake group. Effect sizes for previously seen (old) words, unseen, related (lure) words and unseen, unrelated (new) words were calculated and analysed separately (see Table 1 for means and effect sizes). True recognition and false recognition d’ and C sensitivity measures were also computed for the studies testing recognition memory. When not enough data was provided in the paper to calculate effect size and sensitivity measures, authors were contacted for means and standard deviations.

We computed Hedge’s $g$ based on the means and variance reported in each study for the wake and sleep groups. Hedge’s $g$ is a variation of Cohen’s $d$ that corrects for biases due to small sample sizes. We then fitted a random effects model using the R package metafor (Viechtbauer, 2010). A random effects analysis was chosen because this method, in contrast to a fixed effects meta-analysis, allows for inconsistencies between the studies analysed, calculation of possible sampling error, and assessment of the effects of moderators on the size of the effect (Borenstein, Hedges, Higgins, & Rothstein, 2010). We introduced the six moderator variables, (1) recall or recognition testing, (2) number of words in each list, (3) number of lists learned, (4) whether words were presented auditorially or visually, (5) emotionality of
lists, and (6) overnight or nap study to the model, to examine any possible influence of these moderators on the effect size of sleep.
<table>
<thead>
<tr>
<th>Authors</th>
<th>New Words</th>
<th>Lure Words</th>
<th>Old Words</th>
</tr>
</thead>
<tbody>
<tr>
<td>Year</td>
<td>Mean (SD)</td>
<td>Mean (SD)</td>
<td>Mean (SD)</td>
</tr>
<tr>
<td>2000</td>
<td>4.43 (1.86)</td>
<td>4.43 (1.86)</td>
<td>4.43 (1.86)</td>
</tr>
<tr>
<td>2000</td>
<td>4.43 (1.86)</td>
<td>4.43 (1.86)</td>
<td>4.43 (1.86)</td>
</tr>
<tr>
<td>2000</td>
<td>4.43 (1.86)</td>
<td>4.43 (1.86)</td>
<td>4.43 (1.86)</td>
</tr>
</tbody>
</table>

*Table 1: Descriptions of means, standard deviations, Hedge's g and standard error for lure, old and new words for each experiment.*

As well as information on modulators.
Results

Lure words

The overall effect size for the mean difference in the proportion of responses to lure words given as old between the sleep and wake group, measured by Hedge’s $g$, was 0.129 (SE = .210), which indicated no significant difference from zero (95% CI [-0.282, 0.539], $p = .540$). See Figure 1 for a forest plot of effect sizes. Since at the time of data analysis one study within the meta-analysis was unpublished (Newbury & Monaghan, submitted), we conducted a second analysis without this dataset. The overall effect size did not significantly change, (Hedge’s $g = 0.165$, SE = 0.227, 95% CI [-0.281, 0.610], $p = .469$), and so we continued our analysis of the full dataset. There are possible moderator variables that may lead to differing directions of effects, highlighted by the significant heterogeneity ($Q(12) = 63.227, p < .001$), indicating that there is variance in the data that cannot be explained by the random measurement error. We therefore analysed the effects of each of the moderators (see Table 2 for significance of each moderator).
Figure 1. A forest plot containing effect sizes and 95% confidence intervals for the difference in proportion of old responses between sleep and wake groups for lure words. Studies are split by two moderators: number of words in DRM lists, and recall versus recognition studies. Effect sizes further to the right indicate more lure words falsely recalled or recognised as old in the sleep group than the wake group, and therefore increased false memories after sleep.

Table 2

The effect of each moderator on the overall effect size difference between sleep and wake groups for lure words.

<table>
<thead>
<tr>
<th>Moderator</th>
<th>df</th>
<th>Heterogeneity (Q)</th>
<th>p</th>
</tr>
</thead>
<tbody>
<tr>
<td>Recall versus Recognition</td>
<td>1</td>
<td>3.685</td>
<td>.055+</td>
</tr>
<tr>
<td>Number of lists</td>
<td>1</td>
<td>0.291</td>
<td>.590</td>
</tr>
<tr>
<td>Number of words in each list</td>
<td>2</td>
<td>18.368</td>
<td>&lt;.001***</td>
</tr>
<tr>
<td>Auditory versus Visual</td>
<td>1</td>
<td>0.387</td>
<td>.534</td>
</tr>
<tr>
<td>Emotional versus Neutral</td>
<td>1</td>
<td>0.264</td>
<td>.608</td>
</tr>
<tr>
<td>Nap versus Overnight sleep</td>
<td>1</td>
<td>1.818</td>
<td>.178</td>
</tr>
</tbody>
</table>

Note: +p < .1, ***p < .001
**Moderator Analysis: Recall versus Recognition**

We found no significant effect of test type (Q(1) = 3.86, \( p = .055 \)). However, since the moderator test was close to significance, we ran effect size analyses of the recall and recognition studies separately. The recall studies showed a medium effect of sleep, with sleep increasing the number of lure words that were falsely recalled as old words, Hedge’s \( g = 0.606 \) (SE = 0.299), which was significantly different from zero (95% CI [0.020, 1.192], \( p = .043 \)). The recognition studies showed a very small non-significant effect in the opposite direction, with sleep reducing the proportion of old responses to lure words, Hedge’s \( g = -0.150 \) (SE = 0.243), indicating no significant difference from zero (95% CI [-0.626, 0.327], \( p = .538 \)).

**Moderator Analysis: Number of words**

Studies varied in use of either 10, 12, or 15 words within each list. The moderator test indicated a significant effect of number of words (Q(2) = 18.368, \( p < .001 \)). Studies that used 10 words in each list showed a significant increase in the proportion of lure words falsely recalled or recognised as old after sleep than wake (Hedge’s \( g = 0.920 \), SE = 0.193, 95% CI [0.541, 1.300], \( p < .001 \)). No significant effect for 12 words was found (Hedge’s \( g = 0.315 \), SE = 0.302, 95% CI [-0.277, 0.908], \( p = .297 \)). The effect for 15 words was however found to be significantly different from zero (Hedge’s \( g = -0.495 \), SE = 0.165, 95% CI [-0.818, -0.172], \( p = .003 \)), with an increase in false memories in the wake group compared to the sleep group.

Since the moderator results also indicated a marginally significant difference in performance between recall and recognition studies, we analysed whether the significant effect of number of words in each list was evident in only those studies
using recognition testing. The same effect was confirmed ($Q(2) = 22.043, p < .001$), with lists of 10 words leading to increased false recognition after sleep than wake (Hedge’s $g = 0.853, SE = 0.223, 95\% CI [0.417, 1.290], p < .001$). Lists with 12 words had no effect (Hedge’s $g = -0.250, SE = 0.265, 95\% CI [-0.769, 0.269], p = .345$), where as word lists with 15 words led to an increase in false recognition in the wake group compared to the sleep group (Hedge’s $g = -0.495, SE = 0.165, 95\% CI [-0.818, -0.172], p = .003$). There was insufficient variation in the list length for the recall test studies to be able to analyse these separately.

**Publication Bias**

Funnel plots show the distribution of effect sizes around the mean based on sample sizes, with confidence intervals indicating where studies are likely to be positioned, if there is no publication bias. If many studies fall outside the confidence intervals then this indicates that there may be a publication bias (i.e., only studies with larger effect sizes published). Figure 2 shows a funnel plot of effect sizes for proportion of lure words given as old in the sleep versus wake group. An Egger’s regression test for funnel plot asymmetry, used for smaller meta-analyses (<25 studies), was run to test for possible publication bias (Egger, Smith, Schneider, & Minder, 1997). There are a number of effect sizes that are outside of the expected distribution; however Egger’s regression test indicated no significant funnel plot asymmetry ($z = 0.910, p = .365$) so no evidence for publication bias.
Figure 2. Funnel plot showing standard error of the effect size between sleep and wake groups for lure words, with 95% and 99% confidence intervals.

Old Words

The overall Hedge’s $g$ effect size for old words was 0.159 (SE = 0.126), which again indicated no significant difference from zero (95% CI [-0.088, 0.406], $p = .206$), see Figure 3 for a forest plot of effect sizes. Again we ran the analysis without the unpublished data, and found no significant change in the effect size, (Hedge’s $g = 0.203$, SE = 0.130, 95% CI [-0.052, 0.458], $p = .458$). We therefore continued with the full dataset. There was significant heterogeneity, indicating variance in the data that cannot be explained by the random measurement error, (Q(12) = 28.159, $p = .005$). We therefore again analysed the effects of each of the moderators (see Table 3 for significance of each moderator).
Figure 3. A forest plot containing effect sizes and 95% confidence intervals for the difference in the proportion of old words recalled or recognised between sleep and wake groups for old words. Effect sizes further to the right indicate an increase in the proportion of old words accurately recalled or recognised in the sleep group than the wake group.

Table 3

The effect of each moderator on the overall effect size difference between sleep and wake groups for old words.

<table>
<thead>
<tr>
<th>Moderator</th>
<th>df</th>
<th>Heterogeneity (Q)</th>
<th>p</th>
</tr>
</thead>
<tbody>
<tr>
<td>Recall versus Recognition</td>
<td>1</td>
<td>3.933</td>
<td>.047*</td>
</tr>
<tr>
<td>Number of lists</td>
<td>1</td>
<td>1.376</td>
<td>.241</td>
</tr>
<tr>
<td>Number of words in each list</td>
<td>2</td>
<td>7.151</td>
<td>.028*</td>
</tr>
<tr>
<td>Auditory versus Visual</td>
<td>1</td>
<td>0.088</td>
<td>.767</td>
</tr>
<tr>
<td>Emotional versus Neutral</td>
<td>1</td>
<td>0.321</td>
<td>.571</td>
</tr>
<tr>
<td>Nap versus Overnight sleep</td>
<td>1</td>
<td>0.259</td>
<td>.611</td>
</tr>
</tbody>
</table>

Note: *p < .05
**Moderator Analysis: Recall versus Recognition**

Recall versus recognition as a moderator was a significant effect (Q(1) = 3.933, \( p = .047 \)). We therefore ran effect size analyses of recall and recognition studies separately. For studies using a test of recall, we found no significant effect of sleep versus wake (Hedge’s \( g = 0.407 \), SE = 0.256, 95% CI [-0.094, 0.909], \( p = .112 \)), nor was there a significant effect for recognition studies (Hedge’s \( g = 0.005 \), SE = 0.100, 95% CI [-0.190, 0.200], \( p = .958 \)). Therefore, although recall studies significantly differed from recognition studies, with recall studies showing increased performance accuracy after sleep than recognition studies, there was no significant difference in performance accuracy between sleep and wake groups for tests of recall or recognition analysed separately.

**Moderator Analysis: Number of words**

We found the number of words in each list (10, 12, 15) to be a significant moderating variable (Q(2) = 7.151, \( p = .028 \)). We found a medium effect based on 10 words in the lists (Hedge’s \( g = 0.683 \), SE = 0.230, 95% CI [0.231, 1.134], \( p = .003 \)), with an increase in performance accuracy after sleep compared to wake. We found no significant effect based on 12 words (Hedge’s \( g = 0.116 \), SE = 0.505, 95% CI [-0.334, 0.565], \( p = .614 \)), or 15 words (Hedge’s \( g = -0.094 \), SE = 0.124, 95% CI [-0.338, 0.149], \( p = .448 \)).

Again we tested the effect of number of words for recognition studies only. The same significant effect was found (Q(2) = 6.841, \( p = .033 \)), with lists of 10 words leading to a significant increase in performance accuracy after sleep than wake (Hedge’s \( g = 0.484 \), SE = 0.213, 95% CI [0.066, 0.902], \( p = .023 \)), and no effect based on 12 words.
(Hedge’s $g = -0.280$, SE = 0.265, 95% CI [-0.797, 0.239], $p = .290$), or 15 words
(Hedge’s $g = -0.094$, SE = 0.124, 95% CI [-0.338, 0.149], $p = .448$).

**Publication Bias**

Figure 4 shows a funnel plot of effect sizes for accurate recall or recognition of previously seen (old) words in the sleep versus wake group. There are a number of effect sizes that are outside of the expected distribution, however an Egger’s regression test indicated no significant funnel plot asymmetry ($z = -0.272$, $p = .786$), so no evidence of publication bias.

![Figure 4. Funnel plot showing standard error of the effect size between sleep and wake groups for old words, with 95% and 99% confidence intervals.](image)

**New Words**

For new words, the overall Hedge’s $g$ effect size was -0.277 (SE = 0.079), which significantly differs from zero (95% CI [-0.333, -0.022], $p = .026$), suggesting
that new words were falsely recalled or recognised as old significantly more in the wake group than the sleep group, see Figure 5 for a forest plot of effect sizes per experiment. Removing the unpublished data did not significantly change the results, (Hedge’s $g = 0.204$, SE = 0.083, 95% CI [-0.367, -0.041], $p = .014$) and so we continued with the full dataset. Unlike for lure and old words, heterogeneity was not significant, suggesting that the moderators were not influencing the effect, and any variance in the data can be explained by random measurement error ($Q(12) = 7.440$, $p = .827$).

**Figure 5.** A forest plot containing effect sizes and 95% confidence intervals for the difference in the proportion of new words recalled or recognised as old between sleep and wake groups. Effect sizes further to the right indicate an increase in the proportion of new words falsely recalled or recognised as old in the sleep group than the wake group.
Publication Bias

Figure 6 shows a funnel plot of effect sizes for accurate rejection of new words not previously seen for the sleep versus wake groups. There are two effect sizes that are outside of the expected distribution, and an Egger’s regression test indicated no significant funnel plot asymmetry ($z = -0.179, p = .858$), so no evidence of publication bias.

![Funnel plot showing standard error of the effect size between sleep and wake groups for new words, with 95% and 99% confidence intervals.](image)

*Figure 6.* Funnel plot showing standard error of the effect size between sleep and wake groups for new words, with 95% and 99% confidence intervals.

Signal Detection Analyses

For those studies in which participants were given a recognition task, we calculated the mean difference between sleep and wake groups in their overall discriminability ($d'$) and response bias ($C$) for old versus new words (true recognition) and lure versus new words (false recognition). See Table 4 for $d'$ and $C$ scores per experiment.
Table 4

*Descriptions of Discriminability (d’) and Response Bias (C) for false recognition (lure versus new words) and true recognition (old versus new words) for sleep and wake groups.*

<table>
<thead>
<tr>
<th>Authors</th>
<th>Discriminability (d’)</th>
<th>Response Bias (C)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Sleep Groups</td>
<td>Wake Groups</td>
</tr>
<tr>
<td>Fenn et al. (Experiment 1)</td>
<td>0.819</td>
<td>0.267</td>
</tr>
<tr>
<td>Fenn et al. (Experiment 2)</td>
<td>0.986</td>
<td>0.455</td>
</tr>
<tr>
<td>Fenn et al. (Experiment 3)</td>
<td>0.927</td>
<td>0.382</td>
</tr>
<tr>
<td>Diekelmann et al. (Experiment 1)</td>
<td>0.713</td>
<td>0.188</td>
</tr>
<tr>
<td>Lo et al.</td>
<td>0.498</td>
<td>-0.014</td>
</tr>
<tr>
<td>Monaghan et al.</td>
<td>0.536</td>
<td>0.174</td>
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<td>Newbury &amp; Monaghan</td>
<td>0.389</td>
<td>0.038</td>
</tr>
<tr>
<td>Shaw &amp; Monaghan</td>
<td>0.636</td>
<td>0.251</td>
</tr>
</tbody>
</table>

**Discriminability (d’)**

**False Recognition.** D-prime (d’) for false recognition was analysed by calculating the z-inverse hit rate (lure words falsely accepted as old/total number of lure words), minus the z-inverse false alarm rate (new words falsely accepted as old/total number of new words) for each experiment.

The overall Hedge’s *g* effect size did not significantly differ from zero (Hedge’s *g* = 0.039, SE = 0.098, 95% CI [-0.153, 0.230], *p* = .692) (see Figure 7 for effect sizes). The test of heterogeneity was not significant, suggesting that there were no potential moderators influencing the result, (Q(7) = 0.736, *p* = .998).
Figure 7. A forest plot containing effect sizes and 95% confidence intervals for false recognition discriminability (d’) scores. Effect sizes further to the right indicate an increase in discriminability for the sleep group than wake group.

**True Recognition.** D-prime (d’) for true recognition was analysed by calculating the z-inverse hit rate (number of hits/total number of old words), minus the z-inverse false alarm rate (new words falsely accepted as old/total number of new words) for each experiment.

The overall Hedge’s g = -0.044 (SE = 0.098), which did not significantly differ from zero (95% CI [-0.236, 0.147], p = .650) (see Figure 8). The test of heterogeneity was not significant, suggesting that there were no potential moderators influencing the result, (Q(7) = 4.082, p = .770).
Figure 8. A forest plot containing effect sizes and 95% confidence intervals for true recognition discriminability (d’) scores. Effect sizes further to the right indicate an increase in discriminability for the sleep group than wake group.

Response Bias (C)

False Recognition. Response bias (C) for false recognition was calculated by the z-inverse transformation of hit rate (lure words) + false alarm rate/2. We found no significant effect of sleep on response bias (Hedge’s $g = 0.037$, SE = 0.098, 95% CI [-0.155, 0.229], $p = .706$), see Figure 9 for effect sizes. There was no significant heterogeneity, indicating that there were no potential moderators influencing the effect ($Q(7) = 0.287$, $p = 1.000$).
Figure 9. A forest plot containing effect sizes and 95% confidence intervals for false recognition response bias (C) scores. Effect sizes further to the right indicate more conservative responses in the sleep group than the wake group.

**True Recognition.** Response bias (C) for true recognition was calculated by the $z$-inverse transformation of hit rate (old words) + false alarm rate/2. We found no significant effect of sleep on response bias for true recognition, ($\text{Hedge’s } g = 0.032, \text{SE} = 0.098, 95\% \text{ CI } [-0.159, 0.224], p = .741$), see Figure 10. There was no significant heterogeneity, indicating that there were no potential moderators influencing the effect ($Q(7) = 0.148, p = 1.000$).
Figure 10. A forest plot containing effect sizes and 95% confidence intervals for true recognition response bias (C) scores. Effect sizes further to the right indicate more conservative responses in the sleep group than the wake group.

Discussion

The present study examined the effect of sleep on consolidation of seen words, as well as susceptibility to false memories using the DRM procedure. Although we found no overall significant effect of sleep on false memories, the current meta-analysis helps to clarify mixed findings within the literature, by demonstrating that recall versus recognition testing, and shorter list lengths, enhance sleep-based increases in DRM false memories.

Based on the conclusions of a previous meta-analysis conducted by Chatburn et al. (2014), we hypothesised that this lack of an overall effect may have been due to differences between tests of recall and recognition. Based on the previous meta-analysis, as well as the studies presented in the current analysis, we predicted a strong enhancement effect of sleep compared to wake on false recall (Payne et al., 2009). In
contrast, for tests of false recognition, sleep has been found to reduce, have no effect, or enhance false memories (Diekelmann et al., 2008; Fenn et al., 2009; Monaghan et al., 2017). A moderator test examining the effect of sleep on false recall and recognition separately found a significant effect of recall, with greater false memories after sleep than wake, whereas recognition tests did not have this same effect. The lack of an effect of sleep on false recognition was further supported by the signal detection analysis, which revealed no significant difference in discriminability or response bias between sleep and wake groups. Thus, this meta-analysis supports Chatburn et al.’s (2014) smaller meta-analysis indicating a significant effect of sleep on false recall, but no effect for false recognition.

However, the larger set of studies investigated in the current meta-analysis enabled us to go further to determine the role of additional task constraints on the effect of sleep on memory. In particular, the results also indicated that list length moderated the effect of sleep on false memories. The studies examined in this analysis used lists consisting of 10, 12 or 15 words. Based on previous research indicating an increase in false memories when more list items were presented, due to increasing associations (Robinson & Roediger, 1997), we predicted two possible hypotheses. If sleep aids in spreading activation of memories equally regardless of the density of the word lists, then we would expect to see no effect of list length on the overall effect size. However, word lists of shorter list length create fewer semantic associations at encoding, thus priming fewer similar, unseen words. If sleep aids memory by increasing the spreading activation in long-term semantic associative memory, then this is more likely to result in activation of lure words for shorter lists where the activation within a network containing fewer semantically related items is sparse compared to a more densely activated network resulting from a longer list (Cai
et al., 2009; Sio et al., 2013). The results of the analysis supported this, with an increase in false memories after sleep when studies used lists of 10 words, whereas studies containing lists of 15 words led to a reduction in the proportion of old responses to lure words in the sleep group than the wake group.

For old words, we also found no overall significant difference between the sleep and wake groups on memory performance. This contrasts with previous literature examining the positive effect of sleep on veridical memory consolidation (Dumay & Gaskell, 2007; Plihal & Born, 1997; Wilson et al, 2012). Despite this, two moderators were found to influence the effect size. Tests of recall led to increased performance accuracy after sleep than did tests of recognition for veridical memory of old words. We also found an effect of list length; shorter word lists of ten words led to an increase in accurate memory performance after sleep compared to wake. Therefore, sleep appears to be more beneficial when participants were required to encode fewer words per list. Importantly, this enhancement of sleep effects from short lists for both false and veridical memory was not due to an increase in response bias associated with sleep, as confirmed by the signal detection analyses. The effects were rather specific: only for sparse sets of thematically-related words did sleep improve recognition of old words, and increase acceptance of related but unseen lure words.

For unseen, unrelated (new) words, we expected to see either no difference in performance between sleep and wake groups, due to higher performance accuracy evident in both groups (Monaghan et al., 2017; Newbury & Monaghan., submitted), or an increase in accurate rejection of new words after sleep compared to wake due to an overall increase in performance accuracy after sleep (Rasch & Born, 2007). The meta-analysis revealed a small increase in the proportion of new words falsely recalled or recognised as old in the wake group compared to the sleep group.
Therefore, the sleep groups were significantly more accurate at rejecting new words as previously seen, supporting previous research indicating a benefit of sleep on accurate memory performance (Davis et al., 2009; Dumay & Gaskell, 2007; Plihal & Born, 1997; Wilson et al., 2012).

Although the current results cannot be extended to apply to general verbal memory consolidation, as the DRM paradigm is designed primarily to examine susceptibility to DRM false memories, and not to investigate veridical memory performance, it should be noted that veridical and false memory within DRM tests are often correlated (e.g., Payne et al., 2009; Shaw & Monaghan, 2017). For those studies that used recall testing and shorter word lists, we saw both an increase in veridical memory, and greater susceptibility to false memories after sleep than wake. This, along with the finding that unseen, unrelated new words were rejected more easily by the sleep group, provides support for spreading activation theories of sleep and memory. The current results indicate a role of sleep in associative activation theory (Howe et al., 2009; Gagnon & Plumpton, 2009; Roediger et al., 2001), suggesting that shorter word lists with fewer semantic associations benefit from sleep-dependent spreading activation, leading to false acceptance of critical lures to a greater extent than wake, as well as accurate rejection of words with no semantic association.

Conclusions

The current meta-analysis of the effects of sleep on veridical and false memory consolidation in DRM tests indicated no overall significant effects. Despite this, it is clear that there are several moderating variables that influence offline memory consolidation. Furthermore, the studies presented in this meta-analysis contain further differences in methodology that may explain the lack of an effect of
sleep on both veridical and false memories. For instance, Newbury and Monaghan (submitted) found that sleep improved consolidation of old words to a greater extent than during wake, but only for word lists of negative valence. Monaghan et al., (2017) and Shaw and Monaghan (2017) found evidence for sleep aiding veridical consolidation specifically for those word lists presented to the left hemisphere. Furthermore, Lo, Sim, and Chee (2014) found a reduction in false recognition specifically in older adults, who have previously been found to show different levels of susceptibility to false memories than young adults (Dennis, Kim & Cabeza, 2007; Kensinger & Corkin, 2004) whereas Diekelmann et al. (2010) found an increase in false recall after sleep only for those participants who had an overall low general memory performance. The DRM paradigm does however provide us with evidence for only one type of false memory illusion. Thus, we cannot make firm conclusions regarding the effects of sleep on other forms of veridical and false memories, for example during eye-witness testimony or autobiographical memory for past events.

In conclusion, sleep may therefore improve performance accuracy differentially depending on a number of factors, but the current results do indicate that observations of sleep-enhancement of veridical and false memory effects are task-dependent – potentially sensitive to source monitoring constraints in memory tasks – as well as subject to constraints emergent from the structure of semantic associative memory, as measured by list length, which reflects the density of interconnections within networks of associated words (Monaghan et al., 2017; Robinson & Roediger, 1997). Further investigation as to the effects that these different factors may have on the integration and consolidation of specific information from the short-term to the long-term memory stores will allow for a greater understanding as to the complexities of memory consolidation under different conditions.
References


