



Working memory and attentional control abilities predict individual differences in visual long-term memory tasks[☆]

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ARTICLE INFO

Keywords:

Working memory
Recognition memory
Source memory
Individual differences
Attention control

ABSTRACT

Working memory predicts cognitive abilities like fluid intelligence (gF) and source memory. This suggests these abilities depend on working memory and attentional control. When attentional resources were occupied by a secondary task, previous research shows that source memory performance is more impaired than recognition memory, implying that working memory abilities exert less influence on recognition memory performance than source memory performance. Here, we directly tested if working memory and attentional control differences predict visual recognition memory performance across four experiments ($n = 841$ in total). Surprisingly, we found that working memory and attentional control nearly always predicted recognition memory performance as robustly as source memory (Studies 1, 3 and 4), with the exception of when rapid presentation rates exceeded the temporal limits of attention during encoding (Study 2). Additionally, source memory and recognition memory, regardless of encoding presentation rates across experiments, remained highly correlated across individuals. Together, our findings suggest that working memory and attention control resources play a role in performance of both recognition and source memory tests of visual long-term memory.

Individual differences in working memory capacity are known to predict various cognitive abilities, including fluid intelligence (gF, [Unsworth et al., 2014](#)), reading comprehension ([Turner & Engle, 1989](#)), and complex real-world tasks ([Draheim et al., 2022](#)). One proposal for this relationship has been that working memory capacity measures appear to largely reflect variations in the efficacy of exerting attention control ([Meier & Kane, 2017](#); [Shipstead et al., 2014](#); [Unsworth et al., 2014](#)). While there is still ongoing debate about whether measures of working memory capacity and measures of attention control (WMAC) reflect a single or separable factors, measurements of both are strong predictors of ability across many cognitive domains. In particular, multiple WMAC measures have been shown to be predictive of source memory performance, paired associated recall and delayed free recall performance ([Unsworth et al., 2014](#)). One follow-up study replicated the findings that multiple working memory task measures collectively predicted variance in delayed free recall performance ([Miller et al., 2019](#)). Moreover, pupil dilation during encoding, a measure of attention control, explained significant variance in delayed free recall performance, suggesting that successful recall of verbal materials relies on attentional control resources ([Van der Wel & Van Steenbergen, 2018](#)). These studies

collectively suggest that attentional control differences have been shown to impact performance across a wide range of long-term memory tasks.

The studies above measured long-term memory abilities using either recall of verbal materials ([Miller et al., 2019](#)), or source memory of visual images ([Unsworth et al., 2014](#)). One common feature between these two paradigms is that the memory tests appear to be highly attention demanding, such that they exhibit higher costs under divided attention conditions as compared to recognition memory for single items. For instance, in a signature study, [Craik et al. \(1996\)](#) showed that when participants were asked to perform a secondary task while studying verbal materials, both recall and recognition memory were slower and less accurate compared to when participants were fully concentrating on the memory task alone. However, free recall performance showed more than twice as much accuracy cost (46 %) compared to recognition memory (22 %) when the divided attention task was administered during encoding. Furthermore, the response time cost was nearly five times larger for free recall (146 %) than recognition task (32 %) when the divided attention task was performed during retrieval. Therefore, free recall tasks appear to require more attentional control resources than recognition memory tasks during both encoding and retrieval.

[☆] This article is part of a special issue entitled: 'Individual differences in memory' published in Journal of Memory and Language.

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Similarly, [Troyer et al. \(1999\)](#) showed that source memory, particularly spatial location retrieval, showed a higher memory cost with divided attention than item memory. They concluded that source processing required more attentional control resources than forming item memory.

The divided attention literature raises the possibility that the robust correlations between attentional control and long-term memory may be specific to source memory and recall tasks, due to the higher level of attentional control these paradigms require. Therefore, simple recognition memory performance for items, because of its reduced reliance on attention compared to either recall or source retrieval, may not be predicted by attentional control differences. Echoing this view, individual differences studies of item memory, source memory and recall had been shown to be better explained by a dual-factor or three-factor model than a single long-term memory factor model ([Unsworth, 2019; Unsworth & Brewer, 2009](#)). In their analyses, the recollection factor, shared by item recognition, source memory and recall, was explained by the working memory factor. By contrast, the familiarity factor, on which only item recognition loaded, did not predict the working memory factor formed with span tasks. This evidence converges with the divided-attention literature in suggesting that recognition memory for items has fewer demands for attentional or working memory resources than more effortful retrieval tasks, such as free recall or source memory tasks.

A compelling alternative hypothesis is that recognition memory, despite being less effortful to perform at a high level, still requires substantial attentional control resources for successful performance. Supporting this view, more recent literature using divided attention paradigms has suggested that performing a secondary task during encoding induces similar costs for both item and source memory. In a series of experiments, [Naveh-Benjamin et al. \(2003\)](#) showed that the detrimental effects of divided attention were equal for item memory and source memory between the item and its source information, such as fonts, a paired word, or a paired non-word. Moreover, they tested the participants with free recall, cued recall, and recognition memory under divided attention conditions, and found that performance under divided attention condition was equally hurt with all three test conditions. The findings suggest that the encoding of item memory may rely on similar resources to those during the encoding of source information in memory. A parallel line of research on individual differences discovered that span task performance loaded positively on both recognition and recall factors, suggesting that attentional control might be involved in both recognition and recall ([Unsworth, 2010](#)). One potential caveat to the study, however, is that the authors included the source recognition task into the recognition factor, on which item recognition memory also loaded. As a result, the structural equation model in the paper did not separate the contribution of attentional resources to source memory from those to item recognition memory, leaving the question open for future studies.

Our review above was based on divided attention and attentional control side of the question. Although it's currently debatable on whether attentional control and working memory reflect the same construct, our primary goal here is to test whether working memory capacity predicts simple recognition memory. For instance, prior research has suggested that individual differences in working memory capacity stem from variations across multiple cognitive processes, including primary memory capacity, attentional control, and secondary memory abilities ([Unsworth, 2016](#)). More recent studies have further indicated that both attentional control and secondary memory processes, such as recognition and source memory, mediate the relationship between working memory capacity and higher-order cognitive functions, including reading comprehension and fluid intelligence ([Robison et al., 2024](#)). While these studies have examined how different facets of working memory support complex cognitive tasks, our focus in the current study is on whether working memory capacity also plays a role in simple recognition memory tasks, and if so, what factors may affect the correlation between working memory capacity and recognition memory performance. On one hand, working memory and recognition

memory differ substantially in terms of capacity and activation levels. That is, because recognition memory is often supported by long-term memory, a correlation with working memory might not necessarily be expected. On the other hand, because working memory and recognition memory may share similar decision-making processes under uncertainty, it's possible that this shared component leads to a measurable correlation between working memory capacity and simple recognition memory performance.

The work reviewed above raises an important secondary question about whether individual differences in recognition and source memory form a coherent long-term memory factor. Prior studies have suggested that item recognition memory and source memory rely on distinct neural signatures. For example, when participants were instructed to encode the source of a visual item, the event-related potential (ERP) old/new effect during the test phase was observed to be larger and earlier compared to when they were instructed to memorize the item itself without source information ([Guo et al., 2006](#)). Additionally, item memory elicited an earlier onset of visual ERP than source memory, indicating that source memory retrieval required more extensive conscious processing ([Thakral & Slotnick, 2015](#)). However, other evidence supports a more unified view, such that recognition and source memory depend upon on similar neural processes used by long-term memory. For instance, a neuroimaging study suggested that item and source memory rely on the medial temporal lobe to a similar degree ([Gold et al., 2006](#)). Furthermore, they found that hippocampal lesion participants were similarly impaired on item and source memory tasks. Together, the question of whether item and source memory task rely on different levels of attentional resources, or instead form a single coherent long-term memory factor, remains unresolved.

In our current study, we directly tested whether working memory differences predict recognition memory performance. In each of our Studies, subjects completed a visual working memory task, a source memory task, and a recognition memory task within a single session. In Study 1, we tested whether working memory abilities, measured with a change detection paradigm with visual arrays ([Luck & Vogel, 1997](#)), predicted simple item recognition memory differences for real-world objects ([Brady et al., 2008](#)). In Study 2, we examined whether working memory differences continued to predict item recognition memory performance when items were presented during study at rapid rates (2 items per second) that are known to tax the temporal limits of attentional and working memory encoding ([Potter, 1976](#)). In Study 3, we tested whether the lack of correlation between working memory and rapid serial visual presentation (RSVP) recognition performance was due to the brief exposure duration of each image rather than the rapid rate of incoming images during the study phase. Lastly, to generalize our findings to the working memory and attentional control (WMAC) psychological factor, we administered a wide range of newly developed attentional control tasks to measure the construct in Study 4 ([Burgoyne et al., 2023; Martin et al., 2021; Zhao et al., 2022](#)), along with a visual working memory capacity measure. This design also allowed us to test a secondary hypothesis within each of our experiments regarding whether source memory and recognition memory were highly correlated with each other, despite the differences in retrieval demands.

Study 1

Humans have an exceptional ability to recognize visual objects. For example, after only a single brief presentation, we can accurately (~90 %) recognize thousands of visual scenes and individually presented objects ([Standing, 1973; Brady et al., 2008](#)). This exceptional level of recognition memory ability stands in sharp contrast with other measures of long-term memory, such as free recall or source memory judgments in which accurate mnemonic performance is much more effortful and limited even for much smaller sets of stimuli. One potential explanation for the difference is that tasks such as free recall and source memory may depend more on attention-demanding processes at encoding and

retrieval, as compared to less effortful recognition judgements which may rely more heavily on a general sense of familiarity. In line with this proposal, previous research has demonstrated that individual differences in WMAC measures predict source memory performance (Unsworth et al., 2014). However, whether the easier, familiarity-driven visual recognition process also demands WMAC resources remains unresolved. To test on this question, in Study 1, we measured working memory, recognition memory and source memory abilities of participants. We aimed to test if 1) individual differences in working memory predicted recognition memory, 2) individual differences in familiarity-driven recognition memory predicted recollection-driven source memory, and 3) individual differences in working memory uniquely predicted familiarity-driven recognition memory even after the contribution of source memory was regressed out.

Method

Participants

For Study 1, 385 young residents of the United States (18–35 years old) were recruited through Prolific and received monetary compensation (\$10.00/hour). All participants reported normal or corrected-to-normal vision, no color blindness, fluency in English, no history of mental illness/condition, and no cognitive impairment. All participants had successfully completed 90 % or more of the studies that they had participated in previously on Prolific (filtered by approval rate ≥ 90

%).

Materials and procedure

For Study 1, all participants signed an informed consent, and completed three tasks. Each participant started with an attentional control task (Fig. 1A), followed by a visuo-spatial source memory task (Fig. 1B) and a visual recognition memory task (Fig. 1C).

Working memory Task: Change detection paradigm

During each trial, six colored squares were displayed on the screen simultaneously for 250 ms. This was followed by a blank retention interval lasting 1,000 ms. Subsequently, a single-colored square appeared, and participants were required to determine whether the color of this square had changed compared to the one presented earlier in that position. Changes happened with a probability of 0.5. Participants completed a total of 140 trials for the change detection task. The dependent variable of interest was the K capacity measure ($N \times (\text{hit-false alarm})$) of the working memory task.

Source memory Task: Visuo-spatial source memory

In the study phase, we presented 30 images of real-world objects. The average picture size was approximately 4.6° by 4.6° of visual angle, assuming the subject was seated 80 cm from the screen. Each image was centered on the screen during both the study and test phases.

During the study phase, each trial started with a 500-ms fixation

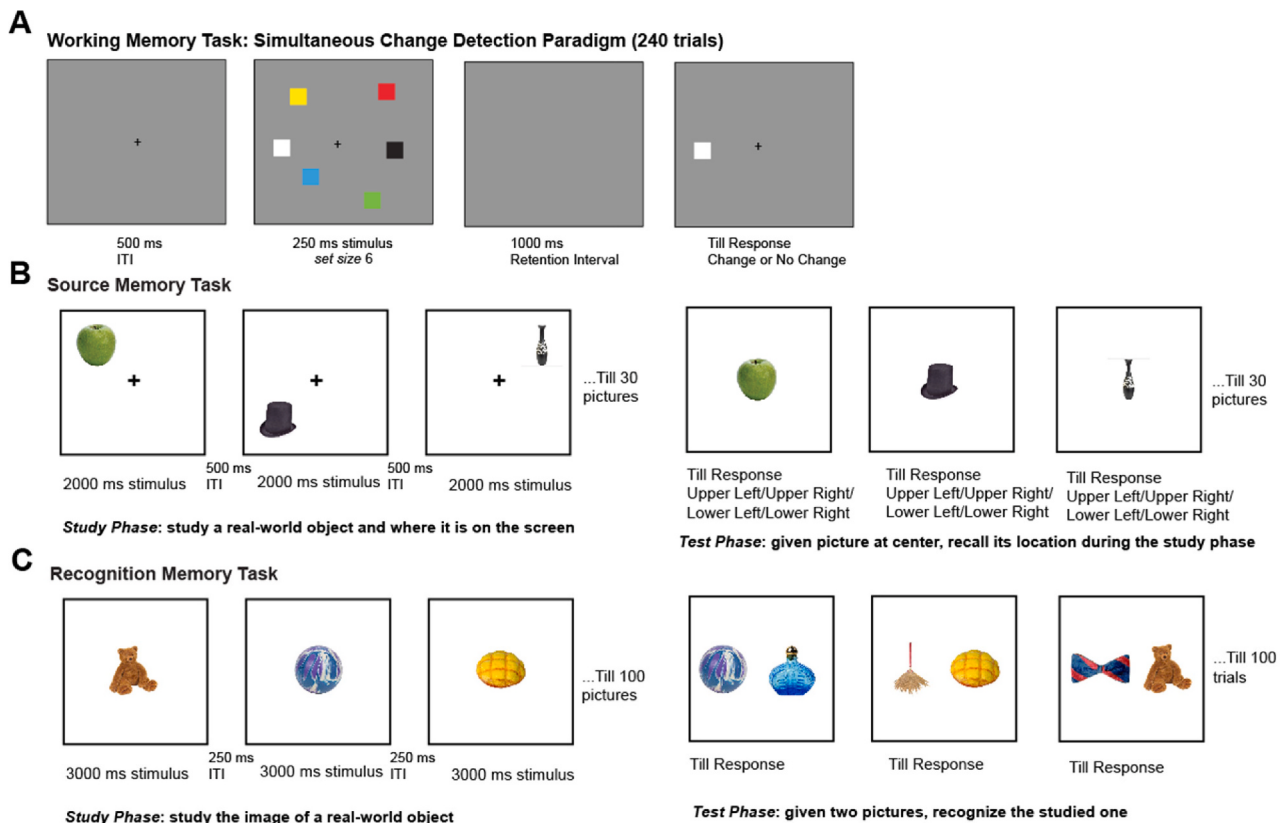


Fig. 1. Experimental Procedures for Study 1. (A) Attentional Control task: simultaneous change detection paradigm. Six colored squares appeared on the screen simultaneously. At the end of the trial, participants were cued with a square at one of the six original locations, with 50% exhibited a color change and 50% without change. (B) Source Memory task. During each trial of the study phase, a real-world object appeared at one of the four quadrants on the screen for 2 s. The participants need to encode the content of the picture as well as where it was placed on the screen. We had thirty pictures in total for the study phase. During each trial of the test phase, we cued the participants with the studied object at the center of the screen, and they were asked to recall which quadrant the object was in during the study phase. (C) A sample of the recognition memory paradigm used in Study 1. During each trial of the study phase, a real-world object appeared at the center of the screen for 3 s. The participants need to encode the content of the picture. During each trial of the test phase, participants were shown two real-world images on the screen, one on the left and the other on the right. They were asked to recognize which of the image was studied. In each trial, one image was always a studied picture, and the other image was a new image that the participants had never studied in the experiment.

cross. Following that, a picture was displayed in one of the four quadrants of the screen for 2000 ms. The participants were instructed to memorize the visual appearance as well as the spatial location of the image. In the test phase, we presented each participant with the same 30 pictures one at a time, and the participants were asked to press W/E/S/D to indicate which quadrant the image was originally at (W for upper left, E for upper right, S for lower left and D for lower right). Each trial in the test phase began with a 500-ms fixation cross, followed by the presentation of the test picture at the center of the screen until the participant made a button press to record their source memory response. The stimuli were drawn from a previously published set of real-world objects (Brady et al., 2008). Both parts of the experiment were programmed using the jsPsych package (De Leeuw, 2015). The dependent variable of interest was the accuracy of the source memory task.

Recognition memory task

In the study phase, we presented 100 images of real-world objects. The average picture size was approximately 4.6° by 4.6° of visual angle, assuming the subject was seated 80 cm from the screen. Each image was centered on the screen during both the study and test phases (Fig. 2A).

During the study phase, each trial started with a 250-ms fixation

cross. Following that, a picture was displayed at the center of the screen for 3000 ms. In the test phase, we presented each participant with 100 pairs of pictures, two at a time, one on the left side of the screen and the other one on the right side of the screen. Among the two images on screen, one image had already been shown in the study phase and the other one was novel to the participant. The participants were asked to perform a two-alternative forced choice task to report which image was the one they had seen during the study phase. Each trial in the test phase began with a 250-ms fixation cross, followed by the presentation of the test picture at the center of the screen until the participant made a button press to record their old versus new response, along with their confidence level.

Participants used the number keys on a keyboard to indicate their confidence and which image was the one they studied earlier. Specifically, the number keys 1 and 2 indicated that the item on the left side of the screen was the studied one, with high and low confidence levels, respectively. On the other hand, the number keys 9 and 8 indicated that the item on the right side of the screen was the studied one, with high and low confidence levels, respectively. The stimuli were drawn from a previously published set of real-world objects (Brady et al., 2008). Both parts of the experiment were programmed using the jsPsych package (De

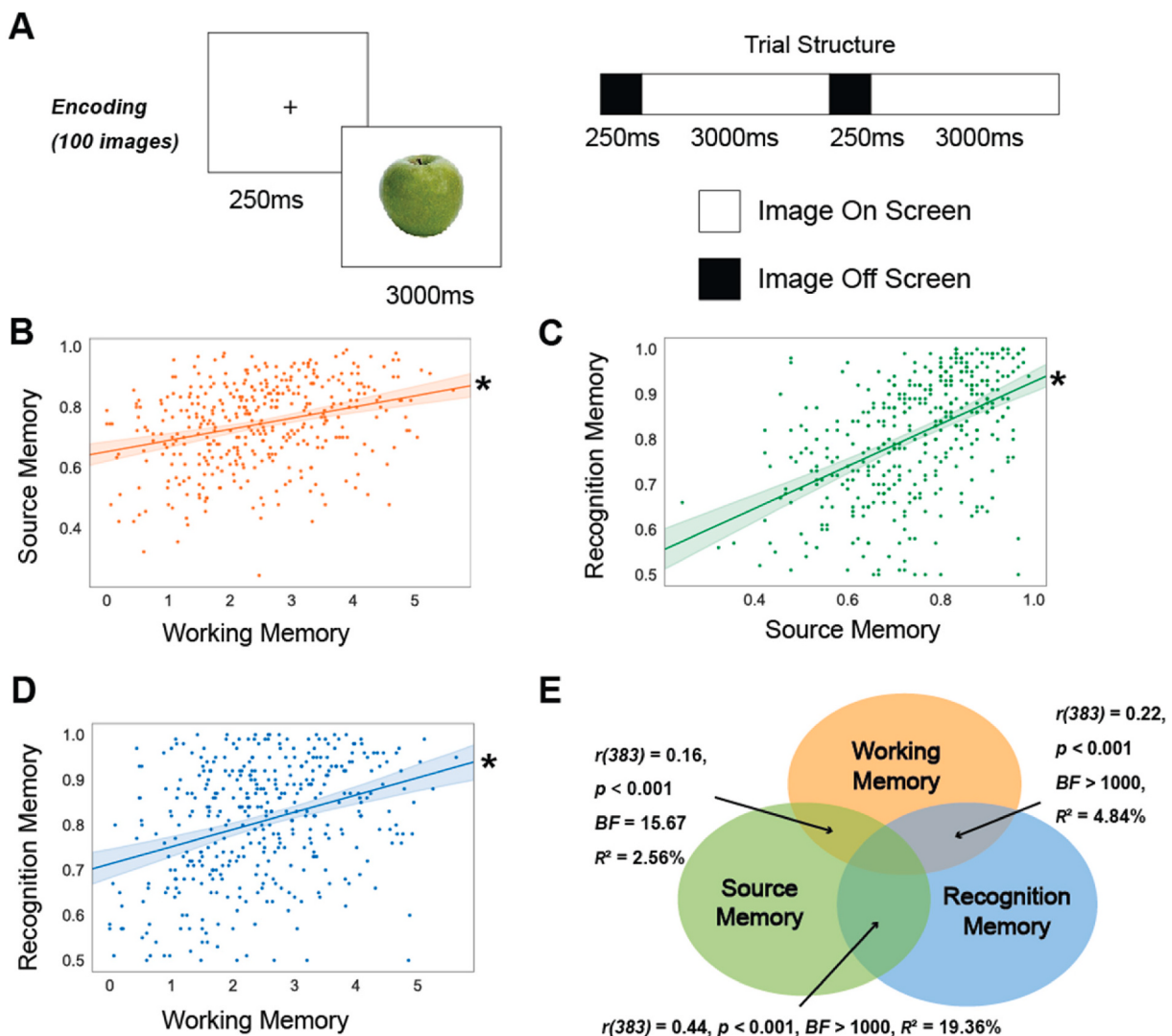


Fig. 2. Results of Study 1. (A) Schema of Recognition Memory Task in Study 1: the images were presented on screen for 3250 ms, and 250 ms of inter-stimulus interval. (B) Working memory differences significantly predicted source memory performances ($r(383) = 0.30$, $p < 0.001$). (C) Source memory performances significantly predicted recognition memory accuracies ($r(383) = 0.49$, $p < 0.001$). (D) Working memory differences significantly predicted recognition memory accuracies ($r(383) = 0.33$, $p < 0.001$). (E) A sample Venn diagram on variance explained by working memory, source memory and recognition memory.

Leeuw, 2015). The dependent variable of interest was the accuracy of the recognition memory task.

Sample size estimation and exclusion criterion

In Study 1, a stopping criterion was established when the Bayes Factor (BF) exceeded 3, indicating support for the alternative hypothesis (i.e., working memory predicted simple recognition memory), or fell below 0.33, indicating support for the null hypothesis (i.e., working memory did not predict simple recognition memory). Here, without prior data on the correlation strength between working memory and recognition memory measures, we estimated the required sample size based on a correlation of $r = 0.15$ and a beta of 0.2 to capture meaningful correlations between the two variables. This estimation indicated a need for 347 subjects. A multivariate exclusion criterion was applied to the behavioral data, excluding participants whose task performance fell below chance and more than 2.5 standard deviations below the mean.

Results

Table 1 lists descriptive statistics and split half reliability for working memory, recognition memory and source memory tasks used in Study 1. Table 2 lists correlations between working memory, recognition memory and source memory tasks.

In Study 1, we first tested whether individual differences in working memory predicted source memory accuracy. We found out that working memory differences positively correlated to source memory performance differences ($r(383) = 0.30, p < 0.001$, Fig. 2B), as shown in previous studies (Unsworth et al., 2014). More importantly, we tested whether working memory also predicted visual recognition memory, a paradigm that was presumably relying less on visual attentional resources than source memory. If the hypothesis held, then we would expect a weaker or nonsignificant correlation than that with source memory. Surprisingly, our correlation analysis revealed that working memory positively predicted recognition memory accuracies ($r(383) = 0.33, p < 0.001$, Fig. 2D), and the correlation strength was not significantly different from the correlation between working memory and source memory ($z(383) = 0.62, p = 0.27$). A further correlational analysis revealed that recognition and source memory strongly correlated with each other ($r(383) = 0.49, p < 0.001$, Fig. 2C), suggesting the existence of a more generalized visual long-term memory construct. These results suggest that both source and item recognition versions of visual long-term memory tasks rely on working memory resources.

To better quantify the level of working memory resources involved in each visual long-term memory task, we performed a three-way partial regression analysis to determine the pairwise shared variance among working memory, source memory, and recognition memory tasks. Echoing our correlational results, we first found that working memory differences continued to share variance with source memory performance after recognition memory was regressed out ($r(383) = 0.16, p < 0.001$, BF = 15.68, Fig. 2E). Likewise, working memory differences also continued to explain a significant amount of variance with recognition memory even after source memory performance was regressed out ($r(383) = 0.22, p < 0.001$, BF > 1000, Fig. 2E). Additionally, recognition memory and source memory continued to share significant portions of variance even after working memory was regressed out ($r(383) = 0.44, p < 0.001$, BF > 1000, Fig. 2E). Collectively, our findings suggest that simple visual recognition memory tasks appear to rely on working

Table 2

Correlations among all measures in Study 1 (** indicates $p < .01$ level).

Correlation	Working Memory	Recognition Memory
Working Memory	—	—
Recognition Memory	0.33**	—
Source Memory	0.30**	0.49**

memory resources just as much as source memory tasks. Moreover, visual recognition memory and source memory differences remained highly correlated with or without attention control differences in the regression model, suggesting that both may form a coherent visual long-term memory factor in explaining individual differences in human cognition.

Study 2

Contrary to our predictions, in Study 1 we found that individual differences in working memory capacity predicted performance in a relatively easy visual recognition memory task; a relationship that was comparable to what we observed for the more effortful source memory task. One possibility for the correlation between recognition memory and working memory capacity is that the relatively slow encoding rate (3 s) during study provided sufficient time to rely on attentional and working memory mechanisms to bolster performance. Prior work using rapid serial visual presentation (RSVP) during the study phase of visual recognition tasks has shown that these rapid rates produce overall lower recognition performance in part because they exceed the temporal limits of an attention-demanding short-term consolidation process (e.g., Potter, 1975). However, despite the reduced recognition rate, the RSVP items appear to still receive relatively deep amounts of perceptual and semantic processing, reflected by the intact accuracy in picture and name search tasks (Potter, 1976). Considering the depth of perceptual and semantic processing, these rapidly presented items may still leave a detectable mnemonic trace. Furthermore, rapid streams of items are known to exceed the limits of working memory encoding. Participants are often unable to encode the second target into memory if it is presented too close in time following the first target (i.e., attentional blink, Shapiro et al., 1997). Similarly participants often miss the repetition of targets in rapid presentation streams (i.e., repetition blindness, Kanwisher & Potter, 1990). In Study 2, we directly tested whether WM differences will continue to predict recognition memory for rapid streams of images. That is, do attention control and working memory still play a role in recognition performance for items that were presented at rates that exceed the temporal limits of attention and working memory encoding?

Method

Participants

For Study 2, 199 young residents of the United States (18–35 years old) were recruited through Prolific and received monetary compensation (\$10.00/hour). All participants reported normal or corrected-to-normal vision, no color blindness, fluency in English, no history of mental illness/condition, and no cognitive impairment. All participants had successfully completed 90 % or more of the studies that they had participated in previously on Prolific (filtered by approval rate ≥ 90 %).

Materials and procedure

For Study 2, all participants provided informed consent, and completed three tasks. Each participant started with a working memory task, followed by a visuo-spatial source memory task and a visual recognition memory task, as in Study 1.

Table 1

Descriptive statistics and reliability estimates for all measures in Study 1.

Measure	Mean	SD	Skewness	Kurtosis	Reliability
Working Memory	2.46	1.15	0.17	−0.51	0.84
Recognition Memory	0.81	0.13	−0.51	−0.64	0.92
Source Memory	0.74	0.14	−0.56	−0.14	0.89

Working memory Task: Change detection paradigm

The working memory task was the same as in Study 1.

Source memory Task: Visuo-spatial source memory

The source memory task was the same as in Study 1.

Recognition memory Task: Rapid presentation

The recognition memory task used the same materials and study-test procedures as in Study 1. Different from Study 1, during the study phase, each trial started with a 250-ms fixation cross. Following that, a picture was displayed at the center of the screen for 250 ms. Therefore, each participant was presented images at a 250-ms on and 250-ms off pace (see Fig. 3A). Note that this presentation rate is more than six times faster than the rate used in Study 1.

Sample size estimation and exclusion criterion

In Study 2, a stopping criterion was established for when the Bayes Factor (BF) exceeded 3, indicating support for the alternative hypothesis, or fell below 0.33, indicating support for the null hypothesis. Here, with our observations from Study 1, we estimated the required sample

size based on a correlation of $r = 0.3$ and a beta of 0.2 to capture meaningful correlations between the two variables. This estimation indicated a need for 85 subjects. A multivariate exclusion criterion was applied to the behavioral data, excluding participants whose task performance fell below chance and more than 2.5 standard deviations below the mean.

Results

Table 3 lists descriptive statistics and split half reliability for working memory, recognition memory and source memory tasks used in Study 2. Table 4 lists correlations between working memory, recognition

Table 3
Descriptive statistics and reliability estimates for all measures in Study 2.

Measure	Mean	SD	Skewness	Kurtosis	Reliability
Working Memory	2.57	1.23	0.14	-0.38	0.73
Recognition Memory	0.69	0.12	-0.11	-0.31	0.85
Source Memory	0.73	0.15	-0.73	0.17	0.91

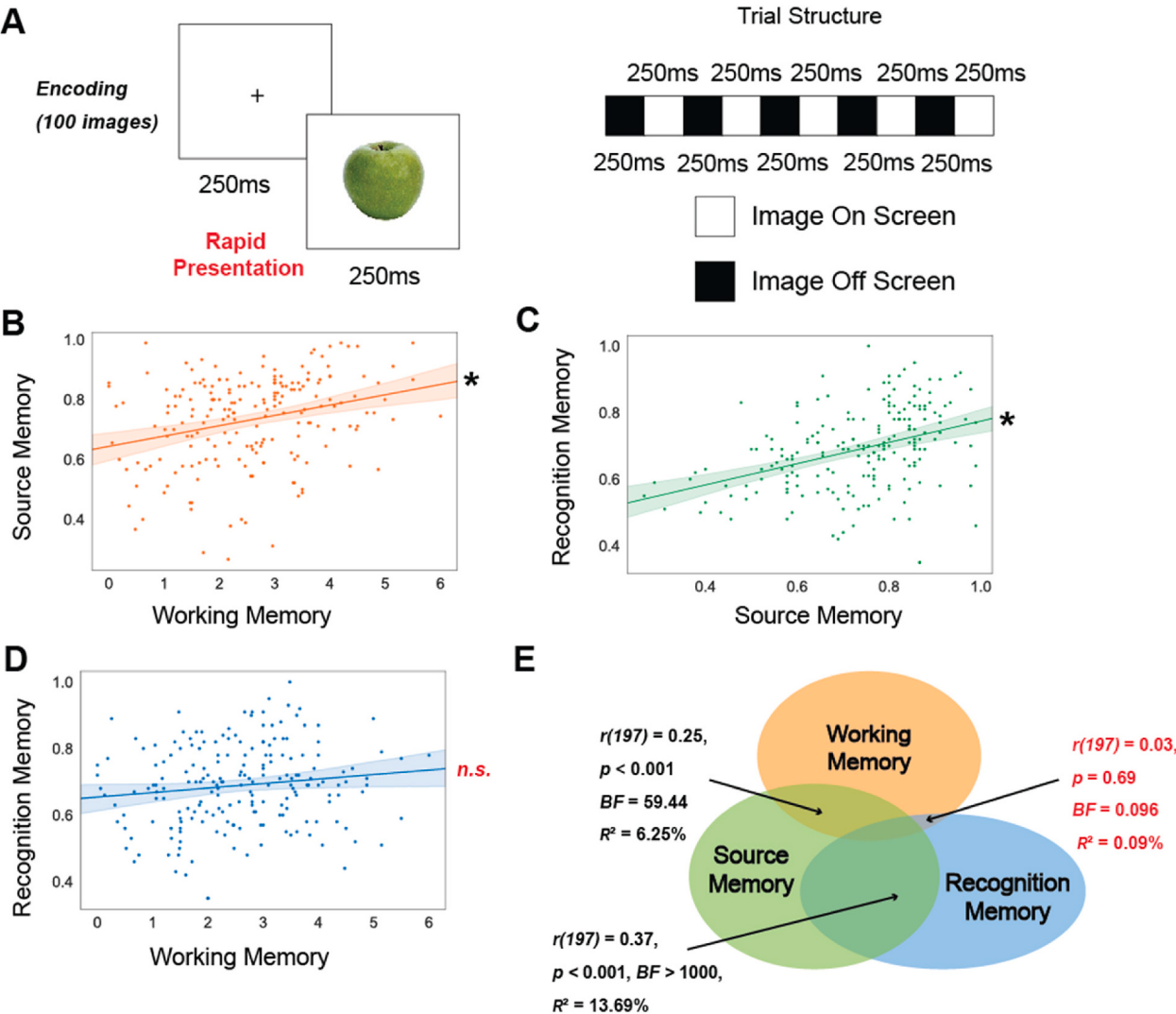


Fig. 3. Results of Study 2. (A) Schema of Recognition Memory Task in Study 2: the images were presented on screen for 250 ms, and 250 ms of inter-stimulus interval, at a presentation rate six times faster than used in Study 1. (B) Working memory differences significantly predicted source memory performances ($r(197) = 0.28$, $p < 0.001$). (C) Source memory performances significantly predicted recognition memory accuracies at a fast presentation rate ($r(197) = 0.39$, $p < 0.001$). (D) Working memory differences did not predict recognition memory accuracies at a fast presentation rate ($r(197) = 0.14$, $p = 0.057$). (E) A sample Venn diagram on variance explained by working memory, source memory and recognition memory.

Table 4
Correlations among all measures in Study 2 (** indicates $p < .01$ level).

Correlation	Working Memory	Recognition Memory
Working Memory	—	—
Recognition Memory	0.14	—
Source Memory	0.28**	0.39**

memory and source memory tasks.

In Study 2, we first tested whether individual differences in working memory predicted source memory accuracy. We replicated our findings in Study 1 that working memory differences positively correlated to source memory performance differences ($r(197) = 0.28, p < 0.001$, Fig. 3B). Furthermore, we tested whether working memory predicted visual recognition memory when the images were presented rapidly. If the ability to encode a rapid stream of images similarly relies on working memory resources, then we would expect a strong positive correlation between working memory and recognition memory with fast presentation rates. However, in contrast with Study 1 when images were presented slowly (> 3000 ms per image), our correlation analysis revealed that working memory did not predict recognition memory accuracies with rapid presentation ($r(197) = 0.14, p = 0.057$, Fig. 3D). Moreover, we found that this nonsignificant correlation with working memory was significantly weaker than the correlation between working memory and source memory ($z(197) = 1.83, p = 0.03$). One possibility for the lack of correlation is that recognition memory with RSVP presentation no longer relies on the common visual long-term memory factor we observed in Study 1. If this is the case, we would expect RSVP recognition to no longer predict source memory performance. However, instead we found that rapid recognition and source memory were still strongly correlated with each other ($r(197) = 0.39, p < 0.001$, Fig. 3C). The robust correlation between RSVP recognition memory and the slowly-presented source memory task indicates that both forms of memory continue to form a general visual long-term memory factor, suggesting that RSVP recognition performance still utilizes the same common long term memory mechanisms as we observed in Study 1. Instead, the null correlation between working memory and RSVP recognition performance suggests that the reduced attentional and working memory processing for these rapidly presented items is what disconnected recognition performance from working memory capacity.

To better quantify the level of working memory resources involved in each visual long-term memory task, we performed a three-way partial regression analysis to determine the pairwise shared variances among working memory, source memory, and RSVP recognition tasks. Similar to our correlational findings, we first replicated our findings that working memory differences shared variance with source memory performance when recognition memory was regressed out ($r(197) = 0.25, p < 0.001, BF = 59.44$, Fig. 3E). Different from Study 1, working memory differences stop explaining shared variances with recognition memory when images were presented rapidly ($r(197) = 0.03, p = 0.69, BF = 0.096$, Fig. 3E). However, RSVP recognition memory and source memory shared significant portions of variance after working memory was regressed out ($r(197) = 0.37, p < 0.001, BF > 1000$, Fig. 3E). Collectively, our findings reveal that visual recognition memory tasks do not rely on working memory resources if images were presented rapidly during study. Moreover, despite the severe difference between presentation rates at encoding, recognition memory and source memory individual differences remained highly correlated, suggesting that both tasks form a coherent visual long-term memory factor.

Study 3

In Study 2, we found that working memory differences did not predict RSVP recognition memory performance. However, it is unclear whether the lack of correlation was due to the rapid rate of item presentation (2 per second) or brief stimulus exposure (250 ms) of each item

presented at study because these two factors were confounded. Here, we tested whether working memory differences predicted recognition memory for items with brief (250 ms) exposure durations but with a slow presentation rate between items by increasing the inter-stimulus interval to 3000 ms. If the null correlation between working memory and RSVP recognition was due to the rapid rate of item presentation, rather than the brief exposure duration, we would expect that the correlation will reemerge for briefly presented items that were presented at a rate that is slow enough between items to allow for full attention and working memory processing for each item.

Method

Participants

For Study 3, 112 young residents of the United States (18–35 years old) were recruited through Prolific and received monetary compensation (\$10.00/hour). All participants reported normal or corrected-to-normal vision, no color blindness, fluency in English, no history of mental illness/condition, and no cognitive impairment. All participants had successfully completed 90 % or more of the studies that they had participated in previously on Prolific (filtered by approval rate ≥ 90 %).

Materials and procedure

For Study 3, all participants provided informed consent, and completed three tasks. Each participant started with a working memory task, followed by a visuo-spatial source memory task and a visual recognition memory task, as in Exp 1.

Working memory Task: Change detection paradigm. The working memory task was the same as in Study 1.

Source memory Task: Visuo-spatial source memory. The source memory task was the same as in Study 1.

Recognition memory Task: Long inter-stimulus interval. The recognition memory task used the same materials and study-test procedures as in Exp 1. Unlike Study 1, each study item was displayed at the center of the screen for 250 ms. Following that, a 3250-ms fixation cross was displayed during the blank inter-stimulus interval. Therefore, each participant saw the images at a 250-ms on and 3250-ms off pace (Fig. 4A), such that the exposure duration was the same as Study 2, but the presentation rate was the same as used in Study 1.

Sample size estimation and exclusion criterion

In Study 3, a stopping criterion was established when the Bayes Factor (BF) exceeded 3, indicating support for the alternative hypothesis, or fell below 0.33, indicating support for the null hypothesis. Here, with our observations from Study 1, we estimated the required sample size based on a correlation of $r = 0.3$ and a beta of 0.2 to capture meaningful correlations between the two variables. This estimation indicated a need for 85 subjects. A multivariate exclusion criterion was applied to the behavioral data, excluding participants whose task performance fell below chance and more than 2.5 standard deviations below the mean.

Results

Table 5 lists descriptive statistics and split half reliability for working memory, recognition memory and source memory tasks used in Study 3. Table 6 lists correlations between working memory, recognition memory and source memory tasks.

In Study 3, we first tested whether individual differences in working memory predicted source memory accuracy. We replicated the findings

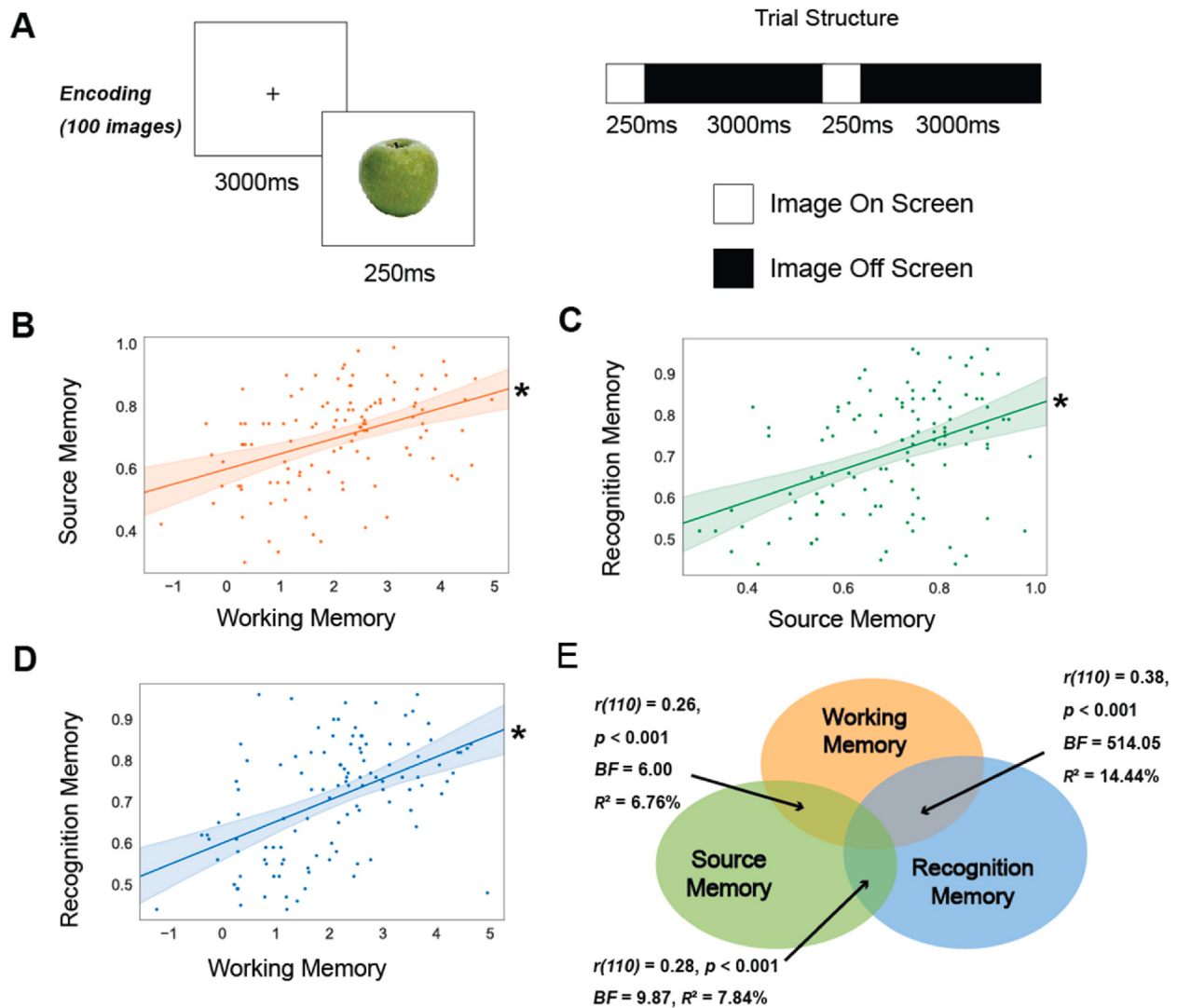


Fig. 4. Results of Study 3. (A) Schema of Recognition Memory Task in Study 3: the images were presented on screen for 250 ms, and 3250 ms of inter-stimulus interval. (B) Attentional control differences significantly predicted source memory performances ($r(110) = 0.42, p < 0.001$). (C) Source memory performances significantly predicted recognition memory accuracies ($r(110) = 0.49, p < 0.001$). (D) Attentional control differences significantly predicted recognition memory accuracies ($r(110) = 0.43, p < 0.001$). (E) A sample Venn diagram on variance explained by attentional control, source memory and recognition memory.

Table 5

Descriptive statistics and reliability estimates for all measures in Experiment 3.

Measure	Mean	SD	Skewness	Kurtosis	Reliability
Working Memory	2.06	1.29	-0.01	-0.56	0.76
Recognition Memory	0.71	0.14	-0.21	-1.00	0.93
Source Memory	0.70	0.15	-0.49	-0.31	0.89

Table 6

Correlations among all measures in Experiment 3 (** indicates $p < .01$ level).

Correlation	Working Memory	Recognition Memory
Working Memory	—	—
Recognition Memory	0.43**	—
Source Memory	0.42**	0.49**

from Experiment 1 and 2 that working memory differences positively correlated to source memory performance differences ($r(110) = 0.42, p < 0.001$, Fig. 4B). Furthermore, we tested whether working memory predicted visual recognition memory when the images were presented briefly, but with a slow presentation rate to allow for attention and

working memory processing. Similar to Study 1, we found that that working memory significantly predicted recognition memory accuracy with brief exposure durations ($r(110) = 0.43, p < 0.001$, Fig. 4D), and this correlation strength was no different from the correlation we observed between working memory and source memory ($z(110) = 0.12, p = 0.45$). Our findings suggest that the lack of correlation between working memory abilities and RSVP recognition in Study 2 was not due to the brief exposure duration of the images themselves but was instead due to the high number of images presented per second. Thus, when items are presented at a rate that is sufficient for attention and working memory processing, the recognition of even briefly presented items (250 ms) is well predicted by working memory ability. Moreover, we performed a correlational analysis between recognition memory with brief presentations and source memory. Similar to Study 1 and 2, we found that recognition and source memory remained strongly correlated with each other ($r(110) = 0.49, p < 0.001$, Fig. 4C). The robust correlation between recognition memory and source memory are consistent with our hypothesis that both forms of memory collectively form a generalized long-term memory construct.

To better quantify the level of working memory resources involved in each visual long-term memory task, we performed a three-way partial

regression analysis to determine the pairwise shared variances among working memory, source memory, and brief-exposure recognition memory tasks. Similar to our correlational findings, we first replicated our findings that working memory differences shared variance with source memory performance after recognition memory was regressed out ($r(110) = 0.26, p < 0.001, BF = 6.00$, Fig. 4E). Similar to Study 1, working memory differences explained significant portions of shared variances with brief-exposure recognition memory ($r(110) = 0.38, p < 0.001, BF = 514.05$, Fig. 4E). Moreover, recognition memory and source memory shared significant portions of variance after working memory was regressed out ($r(110) = 0.28, p < 0.001, BF = 9.87$, Fig. 4E). Collectively, our findings demonstrate that visual recognition memory tasks rely upon working memory resources if sufficient time is provided for attention and working memory processing. Additionally, recognition memory and source memory differences remained highly correlated after working memory contributions are removed, suggesting that both tasks may form a coherent visual long-term memory factor.

Study 4

In Studies 1–3, we found that working memory capacity was predictive of recognition memory so long as the presentation rate of items during study did not exceed attention and working memory limits. However, it is unclear whether broader measures of attentional control abilities (e.g., Kane & Engle, 2002) would also show a similar relationship with visual long-term memory performance. In Study 4, we administered a battery of four WMAC tasks and tested whether they revealed a relationship with recognition and source memory that was comparable to what we have observed with measures of working memory capacity.

Method

Participants

For Study 4, 145 young residents of the United States (18–35 years old) were recruited through Prolific and received monetary compensation (\$10.00/hour). All participants reported normal or corrected-to-normal vision, no color blindness, fluency in English, no history of mental illness/condition, and no cognitive impairment. All participants had successfully completed 90 % or more of the studies that they had participated in previously on Prolific (filtered by approval rate ≥ 90 %).

Materials and procedure

For Study 4, all participants provided informed consent and completed six tasks. Each participant started with four attentional control tasks (change localization paradigm, filtering change localization paradigm, Flanker square task and Simon square task, see Fig. 5), followed by a visuo-spatial source memory task and a visual recognition memory task, as in Study 1.

Working Memory and Attentional Control (WMAC) Tasks: Change localization, filtering change localization, Flanker Square, Simon square

Change Localization

The Change Localization task was adopted from the color Change Localization task used in prior research (Zhao et al., 2022). During every trial, six colored squares were simultaneously displayed for a duration of 250 ms, followed by a blank retention interval lasting 1,000 ms. Subsequently, the same six squares reappeared in their original positions, but with one of the colors was changed to a hue not previously shown in that trial. Each square was assigned a digit ranging from 1 to 6, and participants were required to press the corresponding key to identify the square that changed color. The spatial arrangement of the six numbers was randomized across trials. The Change Localization measure was the capacity of the working memory, calculated from the accuracy of the

task (Zhao, Vogel & Awh, 2023).

Filtering Change Localization

The Filtering Change Localization task was modified from the Filtering Change Detection task (Luck & Vogel, 1997; Martin et al., 2021). On the start of each trial, a word, either RED or BLUE, denoting the color of the items to be remembered (the selection instruction), was presented for 200 ms, followed by a 100-ms interval. Subsequently, 10 bars were displayed for 250 ms, with half of them being drawn in red and the other half in blue, which is effectively a set size 5 condition. After a 900-ms delay, only the bars corresponding to the attended color reappeared. During test phase, only one of the bars changed its orientation compared to the encoding phase. The participants were asked to determine which one of the five bars had changed its orientation compared to the initial presentation. This Filtering Localization phase had 60 trials in total. The Filtering Localization measure was the accuracy of the localization task.

Flanker Square

The Flanker Square task is a modified Flanker task (Burgoyne et al., 2023). In each trial of the Flanker Square task, participants were presented with a target stimulus alongside two possible responses. Both the target stimulus and response options consisted of sets of five arrows arranged horizontally (i.e., $\ll \gg$). Participants were instructed to choose the response option where the middle arrow aligned in direction with the outer arrows in the target stimulus. For instance, if the target stimulus displayed arrows pointing left and right (i.e., $\ll \gg$), participants would select the response option with a central arrow pointing left (i.e., $\gg \ll$). Therefore, the task required participants to focus on the outer arrows of the target stimulus and the central arrow of the response options while disregarding the central arrow of the target stimulus and the outer arrows of the response options. Each participant completed 30 s of practice, and 90 s of test phase. The Flanker Square score was calculated as the difference between number of correct and incorrect responses.

Simon Square

The Simon Square task was a modified Simon task (Burgoyne et al., 2023). In each trial of the Simon Square task, participants were presented with a target stimulus and two response options. The target stimulus was represented by an arrow, and the response options were the words “RIGHT” and “LEFT.” Participants were instructed to choose the response option that corresponded to the direction indicated by the arrow in the target stimulus. For instance, if the arrow in the target stimulus pointed to the left, the participant should select the response option with the word “LEFT.” Both the target stimulus arrow and the response options could appear on either side of the computer screen with equal probability. Consequently, participants had to attend to the direction indicated by the target stimulus arrow while understanding the meaning of the response options. Simultaneously, they must disregard the side of the screen where the target stimulus arrow and response options were presented. Each participant completed 30 s of practice, and 90 s of test phase. The Simon Square score was calculated as the difference between number of correct and incorrect responses.

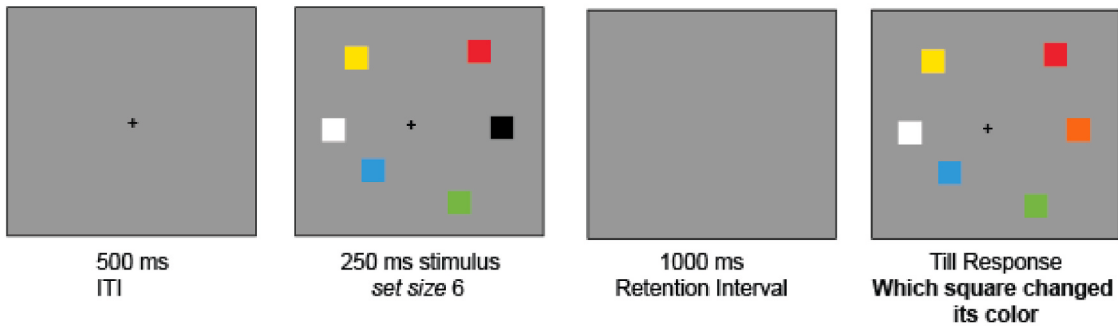
Source Memory Task: Visuo-spatial source memory. The source memory task was the same as in Study 1.

Recognition Memory Task: long inter-stimulus interval. The recognition memory task was the same as in Study 1.

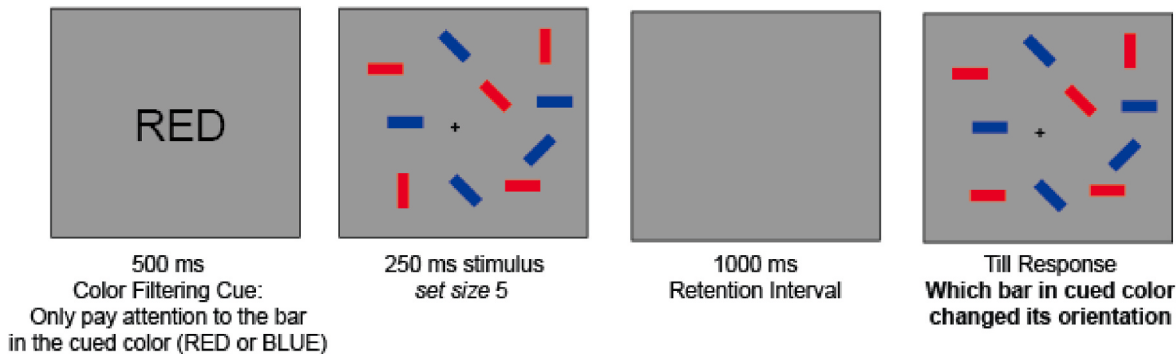
Sample size estimation and exclusion criterion

In Study 4, a stopping criterion was established when the Bayes Factor (BF) exceeded 3, indicating support for the alternative

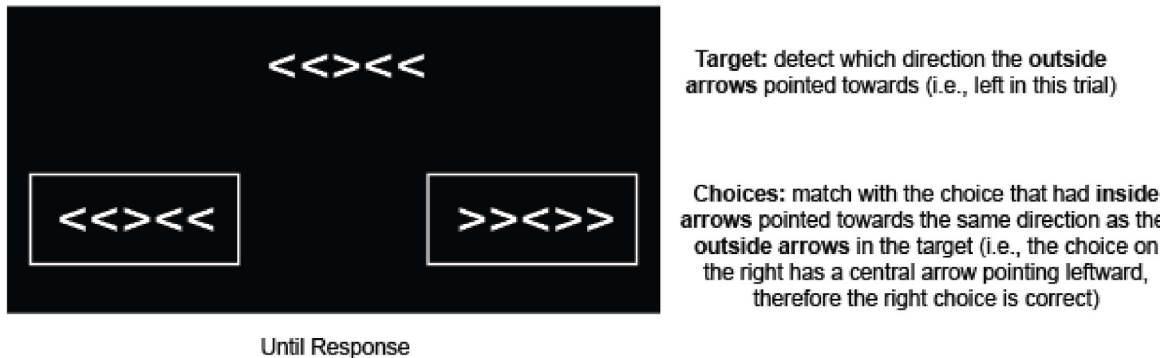
A Change Localization Paradigm



B Filtering Change Localization Paradigm



C Flanker Square Paradigm



D Simon Square Paradigm

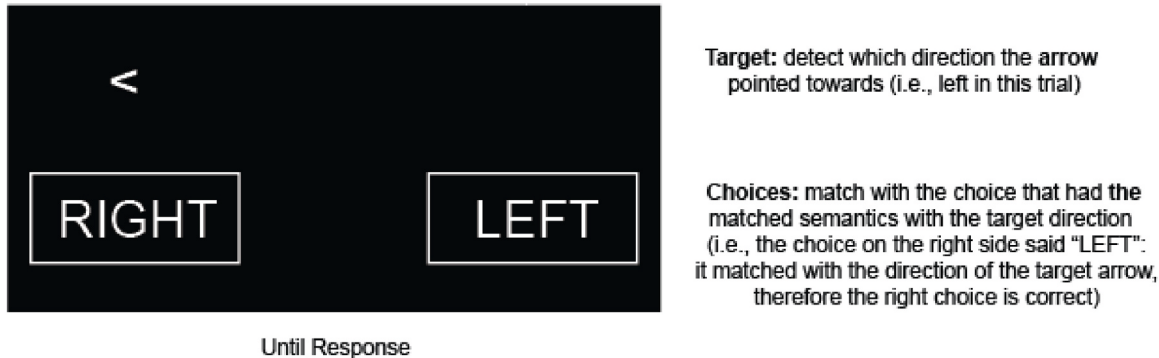


Fig. 5. Experimental Procedures for Study 4. (A) Schema of Change Localization Task. (B) Schema of the Filtering Change Localization Task. (C) Schema of the Flanker Square Task. (D) Schema of the Simon Square Task.

hypothesis, or fell below 0.33, indicating support for the null hypothesis. Here, with our observations from Study 1, we estimated the required sample size based on a correlation of $r = 0.3$ and a beta of 0.2 to capture meaningful correlations between the two variables. This estimation indicated a need for 85 subjects. A multivariate exclusion criterion was applied to the behavioral data, excluding participants whose task performance fell below chance and more than 2.5 standard deviations below the mean.

Results

Table 7 lists descriptive statistics and split half reliability for working memory & attentional control, recognition memory and source memory tasks used in Study 4. Table 8 lists correlations between working memory & attentional control, recognition memory and source memory tasks.

We tested whether a broad set of working memory attentional control (WMAC) measures predict visual long-term memory performance similar to what we observed with working memory capacity in our previous experiments. First, we tested if WMAC abilities, measured with four paradigms, predicted source memory performance and found that performance in all four tasks was positively correlated to source memory performance (change localization: $r(143) = 0.35, p < 0.001$; filtering change localization: $r(143) = 0.26, p = 0.002$; Flanker square: $r(143) = 0.29, p < 0.001$; Simon square: $r(143) = 0.26, p = 0.002$, see Fig. 6 A-D), which generalizes our findings from Studies 1–3. Moreover, we also found that the four WMAC tasks were all predictive of simple recognition memory accuracies (change localization: $r(143) = 0.39, p < 0.001$; filtering change localization: $r(143) = 0.34, p < 0.001$; Flanker square: $r(143) = 0.32, p < 0.001$; Simon square: $r(143) = 0.23, p = 0.006$, see Fig. 6 E-H). These results indicate that the relationship we observed between visual long-term memory performance and working memory generalizes to other attention control and working memory measures. Furthermore, when we divide the WMAC measures into working memory and attentional control measures, visual long-term memory differences shared unique variance with both the WM and AC factors, which replicates and extends our findings with WM measures (Fig. 7C). Therefore, recognition memory appears to rely on similar levels of WMAC resources compared to source memory, despite prior evidence suggesting that source memory performance is impaired more severely under split-attention tasks than recognition memory.

Lastly, we replicated our findings in Study 1 that source memory and recognition memory were highly correlated ($r(143) = 0.60, p < 0.001$, see Fig. 7A). To further examine if these two tasks form a coherent visual long-term memory factor, we performed an exploratory factor analysis with our four WMAC tasks and two visual long-term memory tasks. We first performed a Bartlett's Sphericity test and got a significant chi square of 237.79 ($p < 0.001$), fulfilling a prerequisite to perform factor analysis. We then conducted a Kaiser-Meyer-Olkin criterion test with KMO of 0.79, indicating that our observed variables were sufficiently correlated with each other. To determine the optimal number of factors, we calculated the eigenvalues for our data. Our first two eigenvalues returned as 2.96 and 0.92, higher than the third eigenvalue 0.74 and were both close to or larger than 1.0. Therefore, we performed a two-factor factor analysis on our data (see Fig. 7B). We found that Flanker

Table 8

Correlations among all measures in Study 4 (** indicates $p < .01$ level).

Correlation	1	2	3	4	5
1. Change Localization	—	—	—	—	—
2. Filtering Localization	0.58**	—	—	—	—
3. Flanker Square	0.40**	0.40**	—	—	—
4. Simon Square	0.35**	0.32**	0.43**	—	—
5. Recognition Memory	0.39**	0.34**	0.32**	0.23**	—
6. Source Memory	0.45**	0.37**	0.36**	0.31**	0.60**

Square (0.52), Simon Square (0.58), Change Localization (0.67) and Filtering Localization (0.66) all significantly (> 0.3) loaded onto Factor 1, and Source Memory (0.6) and Recognition Memory (0.86) both significantly loaded onto Factor 2. The exploratory factor analysis confirmed that a common WMAC factor underlay our four tasks. More importantly, the two forms of long-term memory formed a shared visual long-term memory factor, converging with our findings in Studies 1–3. Lastly, all six of our tasks loaded positively to both WMAC and long-term memory factors, suggesting that the predictive power between the two abilities may stem from the fact that visual long-term memory tasks generally require attentional control resources. To further validate if working memory and attentional control abilities would separately predict long-term memory abilities (Factor 2), we performed another partial correlational analysis on working memory (change localization and filtering localization), attentional control (Flanker square and Simon square), and long-term memory abilities (recognition and source memory, see Fig. 7C). We confirmed that attentional control and working memory abilities were closely correlated ($r(143) = 0.37, p < 0.001$), and both working memory ($r(143) = 0.36, p < 0.001$) and attentional control ($r(143) = 0.21, p < 0.001$) predicted long-term memory.

General Discussion

In our study, we examined the relationship between individual differences in working memory, attentional control, source memory and recognition memory. First, we replicated previous findings that working memory and attentional control abilities predict source memory performance. Surprisingly, recognition memory performance, a paradigm that has been presumed to be less attention demanding than source memory, was also robustly predicted by working memory and attentional control abilities (Studies 1, 3 and 4). The only exception that we observed to this relationship was when the study items were presented rapidly (2 per second), at which point, working memory no longer predicted recognition memory (Study 2). In addition, we found that irrespective of differences in presentation rate and exposure duration of items during study, recognition memory performance remained highly correlated with source memory performance in all four Studies. An exploratory factor analysis confirmed our correlational results and suggests that both tasks load onto a common latent factor (Study 4), and that each of the WMAC tasks coherently loaded onto a separate latent factor.

Our study suggests that under most circumstances, an individual's overall recognition memory performance has significant contributions from his or her working memory and attentional control abilities. All five metrics of working memory and attentional control tasks, change detection in Studies 1 and 3, change localization, filtering localization, Flanker square and Simon square in Study 4 significantly predicted recognition memory performance. A potential challenge to this robust relationship would be that although certain levels of working memory and attentional control are necessary for recognition memory, source memory may require much higher contributions of working memory and attentional control resources than simple recognition. If this hypothesis was correct, then the remaining shared variance between WMAC measures and recognition would be expected to be close to zero

Table 7

Descriptive statistics and reliability estimates for all measures in Study 4.

Measure	Mean	SD	Skewness	Kurtosis	Reliability
Change Localization	2.43	1.00	−0.11	−0.01	0.81
Filtering Localization	0.47	0.15	0.30	−0.22	0.83
Flanker Square	31.41	13.93	−0.36	−0.40	0.94
Simon Square	38.34	13.21	−0.39	0.83	0.93
Recognition Memory	0.77	0.15	−0.52	−0.60	0.93
Source Memory	0.79	0.14	−0.98	0.53	0.93

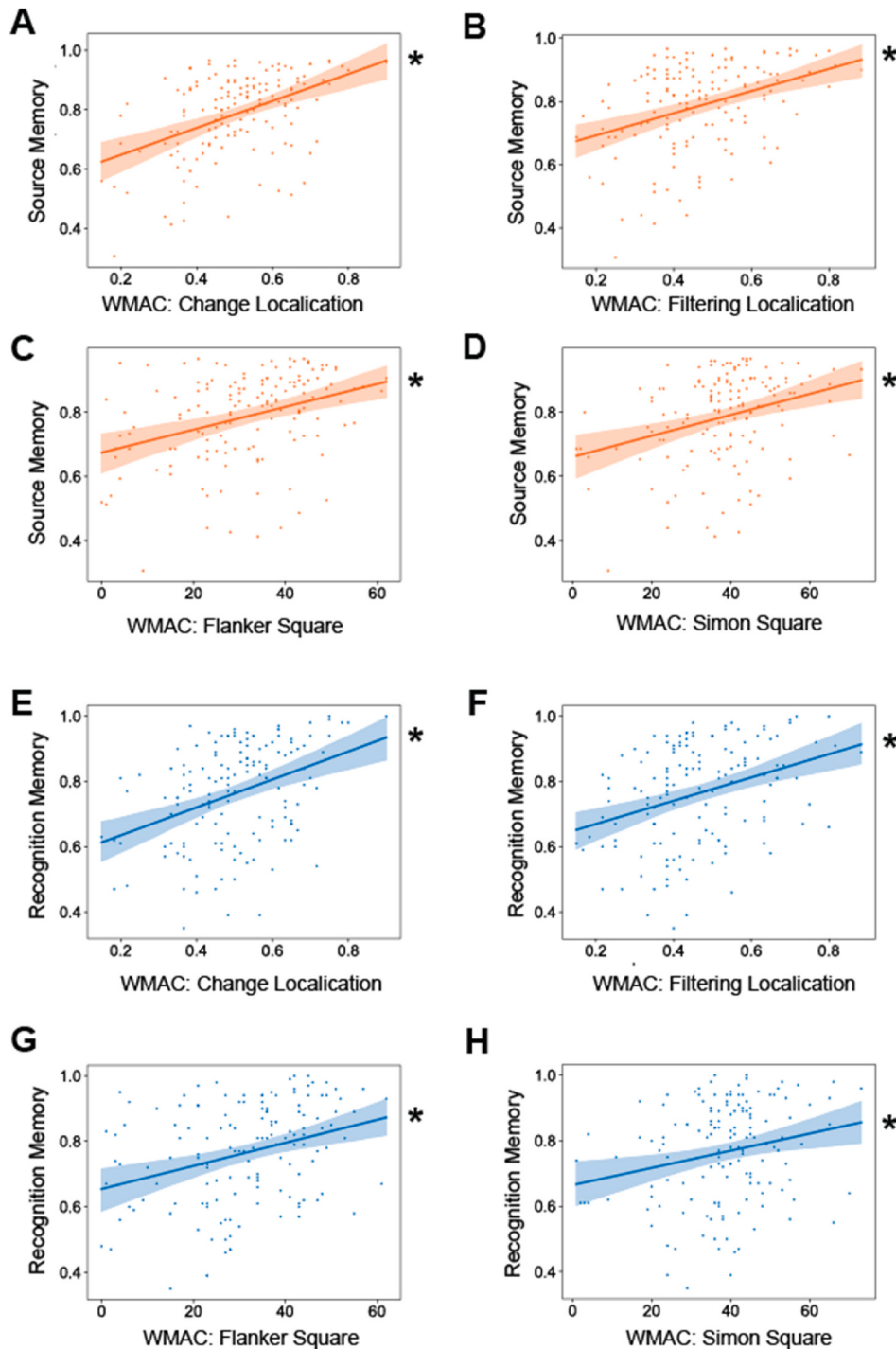


Fig. 6. Results of Study 4. (A-D) Working Memory Attentional Control (WMAC) differences significantly predicted source memory performances ($p < 0.001$). (E-H) Working Memory Attentional Control (WMAC) differences significantly predicted recognition memory performances ($p < 0.001$).

once contributions from source memory were regressed out. Contrary to such prediction, we found that the shared variance between recognition memory and WMAC measures remained significant even after regressing source memory performance in Studies 1, 3 and 4. Therefore, simple item recognition memory appears to require extensive attentional control resources, and this reliance is partially independent from the attentional control resources that source memory requires.

The one exception to this robust correlation that we observed was that when the sequence of 100 images was presented at a rapid presentation rate during study, working memory abilities no longer predicted recognition memory performance. Storage of items in working

memory is known to depend upon a somewhat slow, attention-demanding consolidation process (Potter, 1976; Vogel et al., 2006) that can take several hundred milliseconds per item (Jolicœur & Dell'Acqua, 1998). Our RSVP presentation rate of 2 items per second likely exceeded the temporal limits of this consolidation process, which would explain why RSVP recognition performance no longer depended on working memory ability. This implies that recognition performance for rapidly presented items reflects learning that occurred in the absence of significant contributions from attention and working memory resources.

Our findings may be consistent with prior reports that individual differences in working memory capacity (as measured by OSPAN)

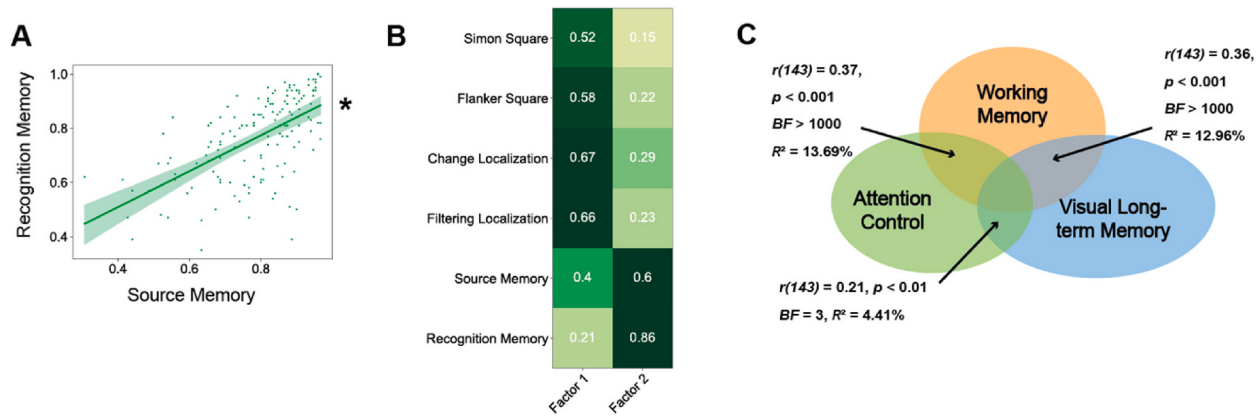


Fig. 7. Results of Study 4. (A) Source Memory differences significantly predicted recognition memory performances ($r(143) = 0.60$, $ps < 0.001$). (B) Exploratory factor analysis results for Study 4. A darker green suggested a higher loading of the task on the latent factor. We observed that Flanker Square, Simon Square, Change Localization and Filtering Change Localization loaded significantly onto Factor 1. Furthermore, Recognition and Source Memory significantly loaded onto Factor 2, suggesting the existence of a long-term memory factor. (C) A sample Venn diagram on variance explained by attentional control, working memory and visual long-term memory. (For interpretation of the references to color in this figure legend, the reader is referred to the web version of this article.)

predict the magnitude of the attentional blink (Dale et al., 2013; Willems & Martens, 2016). While Attentional Blink paradigms use similar rapid stimulus presentation procedures to what we used in our RSVP recognition task, they differ substantially in the working memory storage demands they require. In Attentional blink tasks, subjects must selectively consolidate and store just two predetermined targets from the stream into working memory. By contrast, our RSVP recognition task requires the consolidation and storage of ALL of the items in the stream. It is plausible that when attempting to store just a few items from a rapid stream (as in Attentional Blink tasks), individuals with high WMAC ability have an advantage over those with low ability because the task may rely heavily on successfully selecting just the targets for storage and filtering out the remaining distractors: an ability that is well known to be reduced in individuals with low working memory capacity (Adam et al., 2015; Fukuda & Vogel, 2011; Vogel et al., 2005). By contrast, the RSVP recognition task we used in Study 2 had no distractor filtering demands during encoding because all items were to be remembered. Here, we presume RSVP recognition performance does not correlate with working memory ability because the heavy storage demands exceed even the highest capacity individual's ability to attend and store each item in the rapid sequence into working memory.

Our findings suggest that the process of consolidating items into working memory may play a key role in explaining the correlation between WMAC and long-term memory. This process is thought to reflect the binding of item information to its spatial and temporal context (Karlsen et al., 2010). Computational models have long proposed similar binding mechanisms as a key component of working memory (Oberauer, 2019), plausibly binding an item to both the location and time it was encountered (Thyer et al., 2022). Supporting this hypothesis, individual differences in working memory, as typically measured using simultaneous change detection (i.e., entire set of to-be-remembered objects presented together during encoding), are highly predictive of change detection measured with sequential presentations (i.e., one item presented at a time, Zhao & Vogel, 2023). Therefore, visual array tasks, or WMAC tasks in general, may engage the participants' ability to bind multiple items to different positions in displays with items presented together in time, as well as the ability to temporally bind multiple items to the same location when presented at different times within the sequence. In our visuo-spatial source memory task, working memory uniquely predicted spatial binding ability, which was not utilized in the recognition memory task. Conversely, in the recognition memory task, temporal binding appeared to play a more significant role in recognition memory formation, given that a temporal structure is often present for recognition memory even when order is not being tested (Schwartz

et al., 2005). Supporting this hypothesis, we argue that the rapid stream of images in Study 2 prevented the formation of sufficient temporal binding in recognition memory. Consequently, working memory performance no longer predicted recognition memory performance when items were presented rapidly.

We consistently found that recognition memory and source memory for real-world images remained correlated regardless of the presentation rate and exposure duration of the images in all four of our Studies. Moreover, the shared variance between source and recognition memory remained significant even after attentional control abilities were regressed out. Therefore, we believe that individual differences in visual long-term memory form a construct that is somewhat correlated with, yet distinct from, the known attentional control factor established in previous research (Kane et al., 2004; Kane & Engle, 2002). Evidence from Study 4 shows that both source memory and recognition memory exhibit significantly positive loadings for the same underlying factor, and that this factor is not the same as the attentional control factor from the exploratory factor analysis. Therefore, our results suggest that there is a generalized visual long-term memory factor, as suggested by a recent theoretical review (Unsworth, 2019). Future studies on this long-term memory factor are necessary to better explore the correlational and mediational relationship between this factor and other established factors such as attentional control (Engle et al., 1999), fluid intelligence (Unsworth et al., 2014), and false memory (Roediger, III & Mcdermott, 1995), to name a few.

The current study examined the relationships amongst performance measures from several attention, working memory and visual long term memory tasks. However, because there are other general ability measures such as fluid intelligence and processing speed that are known to correlate with WMAC measures (Kail & Salthouse, 1994; Unsworth et al., 2014), there are always lingering concerns that these untested measures may be contributing to the correlational structure we are observing between these constructs. For example, one possibility is that long-term memory tasks require certain aspects of overall general ability such as fluid intelligence (gF) instead of attention control per se. Previous research has suggested that measures of working memory capacity, attentional control, and secondary memory all mediate the relationship between working memory storage and gF (Unsworth et al., 2014). Given the positive correlations among gF, WMAC, and long-term memory, it remains unclear whether gF and WMAC explain distinct variance in long-term memory. Likewise, declines in processing speed have been shown to impact working memory declines associated with aging (Salthouse, 1991, 1996, 2000). Thus, it is also plausible that the common resources that we observed between visual long-term memory

tasks and WMAC measures may be impacted by the individual's processing speed. That said, the results of Study 2 cast some doubt on whether these alternative general ability constructs are sufficient to explain our current findings. For example, our RSVP recognition study was considerably more difficult than our slow recognition memory procedure as evidenced by the large difference in average recognition accuracy (i.e., 69 % vs 81 %), yet it is only the more challenging RSVP recognition task that is not correlated with working memory capacity. Individuals with high fluid intelligence are generally better at just about all novel task environments (Kyllonen & Kell, 2017; Tschentscher et al., 2017; Zook et al., 2004) and it's not obvious why this general ability wouldn't have also given these individuals an advantage in the challenging RSVP recognition task. Relatedly, if processing speed differences were what was producing the relationship between WMAC measures and visual long-term memory one would expect that such a relationship would continue (if not become stronger) when we increased the encoding speed demands of the stimulus presentation in Study 2. Instead, we found that this seemingly more processing speed-dependent task condition no longer correlated with working memory. These arguments notwithstanding, without measuring these other constructs there will always be remaining uncertainty about the role of these general factor constructs in the relationship between WMAC measures and visual long-term memory. Future work should measure these constructs directly to better reveal the relationships between working memory, attention control and visual long term memory. Furthermore, our current study assesses recognition and source memory constructs using single tasks for each, which may introduce task-specific variance into our results. To achieve a more robust construct-level analysis that is less susceptible to task-specific influences, future research should incorporate multiple tasks representing both recognition and source memory constructs.

CRedit authorship contribution statement

Chong Zhao: Writing – review & editing, Writing – original draft, Visualization, Software, Methodology, Formal analysis, Data curation, Conceptualization. **Edward K. Vogel:** Writing – review & editing, Supervision, Project administration, Investigation, Funding acquisition, Conceptualization.

Declaration of competing interest

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

Acknowledgements

We acknowledge the funding from National Institute of Mental Health (grant ROIMH087214); Office of Naval Research (grant N00014-12-1-0972).

Data availability

The data, stimuli, and experimental codes will be posted at the Open Science Framework (<https://osf.io/vt8wu/>). This study was not preregistered.

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