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**Correlated individual differences suggest a common mechanism underlying metacognition**

**in visual perception and visual short-term memory**

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**Abstract**

Adaptive behavior depends on the ability to accurately introspect about one's own performance.

Whether this metacognitive ability is supported by the same mechanisms across different tasks

has thus far been investigated with a focus on correlating metacognitive accuracy between

perception and long-term memory paradigms. Here, we investigated the relationship between

metacognition of visual perception and metacognition of visual short-term memory (VSTM), a

cognitive function thought to be more intimately related to visual processing. Experiments 1 and

2 required subjects to estimate the perceived or remembered orientation of a grating stimulus and

rate their confidence. We observed strong positive correlations between individual differences in

metacognitive accuracy between the two tasks. This relationship was not accounted for by

individual differences in task performance or average confidence, and was present across two

different metrics of metacognition and in both experiments. A model-based analysis of data from

a third experiment showed that a cross-domain correlation only emerged when both tasks shared

the same task-relevant stimulus feature. That is, metacognition for perception and VSTM were

correlated when both tasks required orientation judgments, but not when the perceptual task was

switched to require contrast judgments. In contrast to previous results comparing perception and

long-term memory, which have largely provided evidence for domain-specific metacognitive

processes, the current findings suggest that metacognition of visual perception and VSTM is

supported by a domain-general metacognitive architecture, but only when both domains share

the same task-relevant stimulus feature.

## Introduction

When humans make decisions they are capable of estimating the likelihood that their decision was accurate. This introspective ability falls under a class of cognitive processes known as metacognition because it entails cognizing about the quality of a decision-making process (1). Intuitively, an individual has high metacognitive accuracy if their estimate of the accuracy of their decision (e.g., as expressed by a confidence rating) corresponds well with the actual accuracy of their decision (2). Because decisions can be made on the basis of information from a plethora of sources—for example, deciding on the basis of current sensory input versus deciding on the basis of information culled from long-term memory—an outstanding question is whether metacognitive processes are domain-general or domain-specific (3). A domain-general metacognitive monitoring process would be expected to evaluate the accuracy of decisions made from both perceptual inputs as well as those based on memory. In contrast, a domain-specific metacognitive system would use independent neural resources or computations to estimate the quality of memory- versus perception-based judgments, for example.

Recent work on this topic has focused on correlating individual differences in metacognition during perception and long-term memory and has resulted in mixed findings. Several studies have reported non-significant relationships between individual's metacognitive ability in a perceptual task and their metacognitive ability in a long-term memory task (4–6), suggesting domain-specific metacognition. However, an experiment using similar tasks did find a reliable positive correlation between metacognitive abilities in both domains (7), and other work has shown correlated metacognitive performance across different perceptual tasks (8), suggesting some shared underlying resources. A number of the above-mentioned studies, however, have

also reported that structural and function brain imaging data from distinct regions correlated with

metacognitive abilities for the distinct tasks, reinforcing domain-specificity at the neural level

(4,5,7). Additional evidence for domain-specificity between perception and long-term memory

has come from a recent study of patients with lesions to anterior portions of prefrontal cortex.

These patients showed a selective deficit in visual perceptual metacognition, but not memory

metacognition for a recently studied word list (9).

A lack of cross-task correlation in metacognition may sometimes be difficult to interpret because

this could result from procedural differences between tasks not necessarily related to the

cognitive construct under investigation (e.g., the use of different stimuli in the perception versus

memory task). Furthermore, perception and long-term memory are themselves quite distinct

cognitive functions (although they can certainly interact in some situations, e.g., (10)), and an

underexplored question is whether perceptual metacognition relates to metacognition for other

cognitive functions more closely related to perception. Across three experiments, we examined

whether metacognition in visual perceptual judgments is related to metacognition for visual

short-term memory (VSTM) judgments using tasks with the same stimuli that differ only in the

requirement for memory storage over a short delay (Experiments 1 and 2), or tasks that differ

also in the relevant stimulus feature (Experiment 3). Because perception of and VSTM for a

given stimulus feature are hypothesized to rely on shared neural representations (11–14), we

might anticipate that metacognition in these domains is also based on some shared resource,

leading to positively correlated individual differences in metacognition across tasks, but only

when the task-relevant stimulus feature is shared.

**Materials and Methods**

**Data availability.** In accordance with the practices of open science and reproducibility, all raw

data and code used in the present analyses are freely available through the Open Science

Framework (https://osf.io/py38c/).

**Experiments 1 and 2.** Because of their similarities, the methods pertaining to Experiment 1 and

2 are described together in this section, followed by the methods for Experiment 3.

**Participants**. Forty subjects (twenty in Experiment 1: mean age = 21 years, $SD$ = 1.67, 10

female, and twenty in Experiment 2: mean age = 20.6 years, $SD$ = 2.01, 14 female) from the

University of Wisconsin-Madison community participated in these experiments and received

monetary compensation. All subjects provided written consent, reported normal or corrected-to-

normal visual acuity and color vision, and were naive to the hypothesis of the experiment. The

University of Wisconsin-Madison Institutional Review Board approved all experiments reported

here.

**Stimuli.** Target stimuli were identical for both experiments and consisted of sinusoidal

luminance gratings embedded in white noise presented within a central circular aperture (see

Figure 1A). Gratings subtended 2 degrees of visual angle (DVA), had a spatial frequency of 1.5

cycles/DVA and a phase of zero. Fixation (a light gray point, 0.08 DVA) was centered on the

screen and was dimmed slightly to indicate trial start (see Figure 1A). Noise consisted of white

noise luminance values generated randomly on each trial for each pixel in the noise patch. The

contrast of the grating was determined for each subject by an adaptive staircase procedure prior

to the main tasks. On a random half of the trials the contrast of both the signal and the noise was

halved. This was not expected to impact orientation estimation performance because the signal-

to-noise ratio of the stimulus was unchanged (15), however it led to a relatively small but reliable

performance difference in Experiment 1 (difference in error = 1.7$^{o}$, p<0.001), but not in

Experiment 2 (difference = 0.15$^{o}$, p = 0.79). This manipulation was not further explored here.

Stimuli were presented on an iMac computer screen (52 cm wide $\times$ 32.5 cm tall; 1920 $\times$ 1200

resolution; 60 Hz refresh rate). Subjects viewed the screen from a chin rest at a distance of 62

cm. Stimuli were generated and presented using the MGL toolbox (http://gru.stanford.edu)

running in MATLAB 2015b (MathWorks, Natick, MA, USA).

**Perceptual task**. To probe each individual's perceptual metacognitive abilities, we employed an

orientation estimation task with confidence ratings (16). On each trial, a target grating was

presented centrally for 33 ms with a randomly determined orientation between 1-180$^{o}$, followed

shortly (600 ms) by a highly visible probe grating without noise, whose orientation could be

rotated via mouse movement. This short interval between the target and probe was necessary to

ensure that the probe had no visual masking effect. Subjects were instructed to match the

perceived orientation as closely as possible. Subjects pressed the spacebar to input their

orientation response and then used number keys 1-4 to provide a confidence rating. Because

performance in this task varies continuously (as opposed to a binary correct/incorrect outcome)

we instructed subjects to use the confidence scale to indicate how close they think they came to

the true orientation using the scale labels 1 = "complete guess" and 4 = "very close". These

perceptual task parameters were the same for both experiments. See Figure 1A for complete trial

timings.

**VSTM task**. To probe metacognitive abilities for VSTM, we introduced a delay period between

the target and the response probe. In Experiment 1, the delay period was fixed at 7 seconds and

in Experiment 2 it was randomly sampled from the set: 3.45, 6.30, 9.15, or 12.00 seconds. The

stimuli and all other task events were identical to the perceptual task in order to minimize any

differences between tasks that are unrelated to the cognitive manipulation of interest

**Staircase**. To minimize performance differences across subjects, both experiments began with

100 trials of a 1-up, 3-down adaptive staircase procedure, which classified responses as correct

or incorrect depending on whether they were within $25^o$ of a trial's true orientation. This

procedure aimed to produce ~80% of trials with less than $25^o$ error. The staircase began with the

grating component of the stimulus having a Michelson contrast of 50%, which was then

averaged with a 100% contrast white noise patch. The step size in grating contrast was adapted

according to the PEST algorithm (17), with an initial starting step size of 20% contrast.

Procedurally, the staircase task was identical to the perceptual task (described above). The

resulting mean (SEM) contrast of the grating (prior to averaging with noise) was 7.8% (0.47) for

Experiment 1 and 8.5% (0.61) for Experiment 2, and was held constant throughout the rest of

both experiments.

**Procedure**. For Experiment 1, perceptual and working memory tasks were performed in separate

blocks. Following the staircase, each subject completed one block of 120 trials of the perceptual

task, followed by three blocks of 60 trials each of the VSTM task, followed by another block of

the perceptual task. This resulted in a total of 240 perceptual trials and 180 VSTM trials per

subject, completed in a single 1.5 hour session. Experiment 2 differed in that perceptual and

VSTM trials were intermixed within blocks and randomly determined with equal probability to

be either a perceptual trial or one of the four delay periods (between 3.45 - 12 seconds) of the

VSTM task. Intermixing perception and VSTM trials further minimized procedural differences

between tasks by eliminating any task-related expecations (since subjects did not know which

type of trial would come next) and by removing temporal delays between task performance.

Each subject completed 300 trials, seperated into 5 blocks, resulting in an average (± SD) of 55.5

(6.4) perceptual trials and 59 (8.0) trials of each delay period of the VSTM task, after removal of

trials based on response times (see below). Total task time was ~1.5 hours.

*Figure 1. Orientation estimation tasks and confidence-error relationships for Experiment 1. (A)*
*On each trial of the perception and VSTM task, subjects moved a computer mouse to match the*

*perceived or remembered orientation and then provided a confidence rating on a 1-4 scale to indicate how close they thought they came to the true orientation where 1 = "complete guess" and 4 = "very close". The tasks differed only by the addition of a 7 second delay period for the VSTM task. (B) Distributions of responses relative to the true orientation (i.e., error) show a clear scaling with confidence ratings, suggesting that subjective ratings track objective performance at the group level. (C) Median absolute error scales with confidence and VSTM trials produced overall greater error, indicating that representations became noisier when held in VSTM. ITI: inter-trial interval. Shaded bands and error bars denote ± 1 SEM.*

**Quantifying metacognition**. Task performance is measured as error (in degrees) between the subject's response and the true orientation on each trial (see Figure 1B). To relate this continuously varying performance metric to subjective confidence ratings, we computed rank correlations between each trials' absolute error and confidence rating to capture how well confidence tracks performance. Error should decrease with increasing confidence so a subject with good metacognition would have a stronger negative correlation between confidence and error than a subject with poor metacognition. Although intuitive, and used elsewhere (18,19), this metric is potentially influenced by factors not necessarily related to metacognitive accuracy per se, such as task difficulty and biases in confidence scale use (e.g., under or overconfidence; (2)). Although we used a staircase procedure to match difficulty, there was still considerable variability across subjects in median absolute error in both Experiment 1 (range: 8 - 16.5$^{\circ}$) and Experiment 2 (range: 6.9 – 23.3$^{\circ}$). A recently introduced measure called meta-d'/d' can correct for these influences (20), however, meta-d'/d' has been developed only for tasks with discreet outcomes amenable to signal detection theory analysis (e.g., hits, misses) and cannot be applied to the continuous estimations tasks we employed in Experiments 1 and 2 (but see Experiment 3). In order to control for these influences when testing our primary hypothesis about the relationship between perceptual and VSTM metacognition, we ran two additional multiple regression models that included covariates for average and task-specific error and confidence

(see *Statistics* below). In the case of models with multiple predictors, the relationship between

perceptual and memory metacognition was visualized (Figure 2 and 4) using added variable plots

(MATLAB function *plotAdded.m*), which use the Frisch–Waugh–Lovell theorem to partial out

the effects of other predictors in the model, revealing the effect of a single predictor while all

other predictors are held constant. Predictor $R^2$ for these models was computed as the sum of

squares for the perceptual metacognition predictor divided by the total sum of squares for all

other predictors and error.

Additionally, we verified that the results of this analysis were robust to our particular metric of

metacognition by repeating all analyses using the non-parametric area under the type 2 receiver

operating characteristics curve ($A_{ROC}$; (21–23) as our measure of metacognitive accuracy. This

measure is obtained by taking the area under the curve formed by plotting the type 2 false alarm

rate by the type 2 hit rate at different type 2 criteria. A type 2 false alarm is an incorrect but high

confidence trial and type 2 hit is a correct and high confidence trial and the number of

confidence criteria is the number of ratings on the scale minus 1. At values of 0.5, this metric

indicates that confidence ratings do not discriminate between correct and incorrect trials and

values of 1 indicate perfect discriminability. $A_{ROC}$ was computed using the method outlined in

(21). Because this metric requires binarizing the data into correct and incorrect responses, we

defined thresholds for each subject based on the 75th percentile of their response error

distributions such that a trial with error larger than this threshold was considered incorrect. This

analytically set performance at 75% for each subject, equating accuracy for this analysis. Using a

common threshold of 25° for each subject did not change the statistical significance of any

analyses reported with this metric. Prior to any analysis, trials with response times below 200

milliseconds or above the 95$^{th}$ percentile of the distribution of response times across all subjects

were excluded. The same trial exclusion procedure was applied to both experiments.

**Statistics**. We used linear regression to predict individual differences in VSTM metacognition

from variation in perceptual metacognition scores (Figures 2 and 4). In a first, "basic model", we

considered only these two variables. Then, to control for individual differences in task

performance and confidence ratings, we ran two additional regression models. One included each

subject's mean error and mean confidence as covariates (3 predictors in total) and the other

included task-specific confidence and error as covariates (i.e., mean perceptual error and

confidence and mean VSTM error and confidence; 5 predictors in total). These three models

were run for each metric of metacognition (r values and $A_{ROC}$; see above) and for both

experiments. To test for linear effects of confidence on error (Figures 1C and 3D) we regressed

single-trial confidence ratings on absolute error for each subject and task and tested the resulting

slopes against zero at the group level using a t-test. To test for performance differences between

tasks we compared median absolute error between the perception and VSTM task with a paired

t-test. We additionally tested for a linear effect of delay period duration in Experiment 2 (Figure

3B) by fitting slopes to each subject's single-trial absolute error by delay period data and testing

these slopes against zero at the group level with a t-test. All tests were two-tailed.

**Experiment 3**. This experiment was conducted to test whether the correlation between

perceptual and VSTM metacognition depended on both tasks sharing the same task-relevant

stimulus feature (e.g. orientation). To this end, we compared metacognition in an orientation

perception task and a contrast perception task (Figure 5A) to metacognition in an orientation

VSTM task. If the perception of and short-term memory for orientation depend on overlapping

representations (13,14), then individual differences in metacognition may be correlated between

orientation perception and orientation VSTM, but not between contrast perception and

orientation VSTM. Furthermore, we used 2-choice discrimination tasks in Experiment 3 which

are amenable to a recently developed Bayesian hierarchical model-based analysis of

metacognition that controls for individual differences in task performance and confidence biases

while appropriately accounting for variability in individual-subject parameter estimates at the

group level (24).

**Participants.** Twenty subjects (mean age = 21.8 years, *SD* = 3.18, 13 female) from the

University of Wisconsin-Madison community participated in these experiments in exchange for

monetary compensation.

**Stimuli.** Sinusoidal luminance gratings subtending 2 DVA were centered 1.5 DVA to the right

and/or left of fixation (Figure 5A). As in Experiments 1 and 2, the gratings were averaged with

white noise. The contrast of the grating and the noise components were adjusted for each subject

using a staircase procedure (see below). Stimuli were presented on an iMac computer screen (52

cm wide × 32.5 cm tall; 1920 × 1200 resolution; 60 Hz refresh rate) and viewed by subjects from

a chin rest 62 cm away. Stimuli were generated and presented using the MGL toolbox

(http://gru.stanford.edu) running in MATLAB 2015b (MathWorks, Natick, MA, USA).

**Contrast perception task.** Subjects were instructed to indicate whether the left or right stimulus

contained a higher contrast grating using the left and right arrow keys, respectively. Subjects

then indicated their confidence using number keys 1-4, where 1 denotes a "complete guess" and

4 denotes "very confident". Subjects were encouraged to use the full range of the scale and were

instructed to understand a number 4 rating as indicating the highest confidence they would feel

in this task, given the difficult nature of the task. This confidence rating procedure was the same

for all three tasks in this experiment. A target and a standard stimulus were presented

simultaneously to the left and right of fixation for 50 ms. The location containing the target was

randomly determined on each trial. Each stimulus also had a randomly and independently

determined orientation that was task irrelevant. Responses could be made as soon as the stimuli

were presented and there was no time limit for responding. The standard stimulus was created by

averaging a 10% Michelson contrast grating with an 80% contrast noise patch and the contrast of

the target grating was adapted for each subject with a staircase procedure (see below).

**Orientation perception task.** This task required subjects to indicate whether the two gratings

had the same or different orientation. Both stimuli appeared simultaneously for 50 ms and then

subjects indicated their choice followed by their confidence. On "same" trials, both stimuli had

an identical orientation, which was randomly determined on each trial (between 1 and 180$^{\circ}$),

whereas on "different" trials one stimulus was offset by 25$^{\circ}$ clockwise or counter-clockwise

(randomly determined). Whether a trial was same or different was randomly determined.

Difficulty was controlled by a staircase procedure that adapted the contrast of both stimuli.

**Orientation VSTM task.** This task also required that subjects indicate whether two gratings

were the same or different, but here there was a temporal delay of 3 seconds between the first

and second stimulus. Thus, the orientation of the first stimulus must be maintained over the delay

in order to perform the task. Each grating was presented for 50 ms and subjects provided their

choice and confidence, with no time pressure, following the second "probe" stimulus. As in the

orientation perception task, both stimuli had an identical randomly determined orientation on

same trials, and a difference in orientation of $25^{o}$ (clockwise or counter-clockwise) on different

trials. Trial type was randomly determined as was the location (left or right of fixation) of both

stimuli, although the location of both stimuli was always the same for any given trial.

**Staircase.** Each task began with 60 trials of a 1-up, 3-down staircase procedure, aimed to

converge on ~80% accuracy. During these 60 trials, stimulus contrast was adjusted according to

the PEST algorithm (17), with a starting step size of 8% contrast for all task. For the contrast

perception task, the staircase adapted the contrast of the grating component of the target and

modulated the contrast of the noise component of the target in the opposite direction so that the

overall stimulus contrast always matched the standard (see Figure 5A *left*). For example, if the

contrast of the grating component of the target was +12% relative to the grating component of

the standard then the noise component of the target was reduced by 12%, thereby matching total

stimulus contrast between the target and standard, but producing higher contrast in just the

grating component of the target. Starting contrast of the target grating was +20% relative to the

standard. For the orientation perception and VSTM tasks the starting contrast of the grating

component of each stimulus was 30%, which was averaged with 80% contrast noise. After these

initial PEST trials, a threshold was computed as the mean contrast from the last 4 staircase

reversals. The staircase continued throughout the duration of each task but with a fixed step size

of 0.5% for the contrast task and 0.25% for the orientation tasks, with the starting threshold

determined from the initial PEST staircase in the case of the first block of each task, or from the

mean of the last 4 reversals from the most recent block.

**Procedure**. Each subject completed 3 blocks of 100 trials each for each of the three tasks,

resulting in 300 trials per task (with the exception of one subject who only completed 100 trials

of the contrast perception task). Blocks of the same task were completed sequentially and task

order was randomized. Prior to the start of each new task, subjects completed 60 trials of the

initial PEST staircase corresponding to that task. These 60 trials were not included in any

analysis. Total task time was ~1.5 hours.

**Model-based analysis of metacognition**. Because Experiment 3 employed 2-choice

discrimination tasks we quantified metacognition in a bias-free signal detection theory model

(20,24). We used an estimate of metacognitive efficiency (M-ratio) that quantifies the extent to

which confidence ratings discriminate between correct and incorrect decisions (i.e., type 2

performance), given the underlying difficulty of the discrimination itself (i.e., type 1

performance), thereby optimally controlling for task difficulty and confidence biases (2). M-ratio

is the ratio between the d' estimated from the confidence data according to a metacognitively

ideal observer and the actual d' computed from task performance. Because both metrics are in

the same units, M-ratio will approach 1 if all the information used for the type 1 decision was

also available to the type 2 decision, indicating optimal metacognition. Values below 1 reflect

suboptimal metacognition.

We used a recently introduced hierarchical modeling approach to estimate the cross-task

correlation between individual differences in M-ratio, as is implemented in the freely available

toolbox HMeta-d (24, https://github.com/smfleming/HMM) for MATLAB. This toolbox is a

hierarchical Bayesian extension of Maniscalco and Lau's (20) meta-d' model. The advantage of

a Bayesian model in this context is that the estimation of group-level parameters of interest (i.e.,

M-ratio correlation coefficient across tasks) takes into account parameter uncertainty at the

single subject level. This means that a subject whose M-ratio is estimated with high uncertainty

will contribute less to the group-level parameter estimate than a subject whose M-ratio is

estimated more precisely. In typical maximum likelihood or sum of squares fitting (20), this

knowledge of parameter uncertainly is discarded. Simulations suggest this approach produces

more accurate parameter recovery and lower false positive rates than non-Bayesian alternatives

(24). Cross-task M-ratio correlations were estimated using the HMeta-d function

*fit_meta_d_mcmc_groupCorr.m.*

Posterior distributions of parameters were sampled using Markov-Chain Monte-Carlo methods

(MCMC) implemented in JAGS (http://mcmc-jags.sourceforge.net). We ran 3 chains of 20,000

samples each with 5,000 burn-in samples. Each subject's log(M-ratio) for each domain are

specified as draws from a bivariate Gaussian. We used a weakly informative normal prior on

log(M-Ratio) which encompasses estimates from 167 previous subjects (24), and a uniform prior

between -1 and 1 for the correlation coefficient. To assess convergence we ensured that all

MCMC chains were well mixed and that the Gelman and Rubins $R^{\wedge}$ convergence statistics were

between 1 and 1.1. Statistical significance for each correlation was assessed by computing the

proportion of MCMC samples that fell below zero, multiplied by 2 (akin to a two-tailed non-

parametric frequentist test) and by computing 95% high-density intervals (HDI) on the

correlation posterior distributions.

**Results**

**Experiment 1.** Distributions of response error as a function of confidence are shown in Figure

1B. Absolute error significantly decreased with increasing confidence for both the perceptual

task ($t(19) = -13.48$, p < 0.0001) and the VSTM task ($t(19) = -14.88$, p < 0.0001), indicating that

subject's confidence reasonably reflected their task performance at the group level (Figure 1C).

Error was also significantly greater in the VSTM task as compared to the perceptual task ($t(19) =$

-2.10, p = 0.049), reflecting an expected degradation of orientation information when the task

required short-term memory maintenance. Confidence ratings were distributed similarly for

perception and VSTM tasks (Supplementary Figure 1), and average confidence ratings did not

significantly differ between tasks (p = 0.15). See Supplementary Figure 2 for a breakdown of

accuracy and response time by block.

Central to our hypothesis, we found a positive relationship across individuals between perceptual

metacognition and VSTM metacognition (Figure 2). This relationship was observed when using

confidence-error correlations as the measure of metacognition (slope = 0.51, t = 2.63, predictor

$R^2 = 0.28$, p = 0.016; Figure 2A) and, importantly, was still present after controlling for average

confidence and error (slope = 0.54, t = 2.55, predictor $R^2 = 0.28$, p = 0.021) and in the model

controlling for task-specific confidence and error (slope = 0.62, t = 2.66, predictor $R^2 = 0.29$, p =

0.018). All covariate predictors in both control models were not statistically significant (ps >

0.30). These results indicate that, although the confidence-error correlation may be influenced by

task performance and confidence biases, these factors did not account for the across-subjects

correlation between perceptual and VSTM metacognition.

**Figure 2.** *Positive relationship between perceptual and VSTM metacognition in Experiment 1.*
403
404    *(A) Cross-task regression using confidence-error correlations as the metric of metacognition.*
405    *Increasingly complex regression models controlling for task performance and confidence shown*
406    *from left to right (see Methods). (B) Same models as in A, but using the area under the type 2*
407    *ROC curve ($A_{ROC}$) as a measure of metacognitive performance. Dashed lines denoted 95%*
408    *confidence intervals on the linear fit. Black points are individual subjects.*
409

The same relationship was observed when using $A_{ROC}$ as the metric of metacognition (Figure

2B). With the basic model, perceptual metacognition significantly predicted VSTM

metacognition (slope = 0.77, t = 3.96, predictor $R^2$ = 0.46, p = 0.0009). This relationship held

when controlling for average confidence and error (slope = 0.82, t = 3.78, predictor $R^2$ = 0.46, p

= 0.0016) and when controlling for task-specific confidence and error (slope = 0.88, t = 3.85,

predictor $R^2$ = 0.45, p = 0.0017). As before, all other covariate predictors across both control

models were non-significant (ps > 0.26). We examined the correlation between all predictor

variables in all of our models (Table 1) and found that there was collinearity between several,

quite expected, covariate predictors (e.g., average confidence predicted average error, perceptual

confidence predicted VSTM confidence). Importantly, however, there were no significant

correlations between our predictors of interest (both perceptual metacognitive scores, $A_{ROC}$ or

rho) and any other covariate predictors, indicating that task performance and confidence are

unlikely to be driving the cross-task correlation in metacognition. These results indicate that the

relationship observed between perceptual and VSTM metacognition was independent of the

particular metric used and was not accounted for by correlated individual differences in task

performance or average confidence.

| Predictor – Exp. 1, *Exp. 2* | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 |
|---|---|---|---|---|---|---|---|---|
| 1. Perceptual metacognition (rho) | - | -0.13 | -0.03 | -0.16 | -0.04 | -0.10 | -0.01 | X |
| 2. Average confidence | 0.33 | - | -0.41 | X | X | X | X | 0.15 |
| 3. Average error | 0.01 | **-0.54\*** | - | X | X | X | X | 0.01 |
| 4. VSTM confidence (average) | 0.33 | X | X | - | -0.42 | **0.95\*** | -0.38 | 0.17 |
| 5. VSTM error (average) | 0.08 | X | X | **-0.54\*** | - | -0.42 | **0.91\*** | -0.01 |
| 6. Perceptual confidence (average) | 0.32 | X | X | **0.94\*** | **-0.44\*** | - | -0.38 | 0.13 |
| 7. Perceptual error (average) | -0.05 | X | X | **-0.53\*** | **0.83\*** | **-0.53\*** | - | 0.03 |
| 8. Perceptual metacognition ($A_{ROC}$) | X | -0.29 | 0.07 | -0.28 | 0.01 | -0.14 | -0.29 | - |

427 ***Table 1.*** *Correlation matrix for every predictor in each model in Experiment 1 and Experiment 2*
428 *(gray regions). X's denote predictor combinations that were not used in any model. Significant*
429 *correlations (p<0.05) are noted in bold and with asterisks.*
430

**Experiment 2.** This experiment served to replicate the cross-task correlation observed in

Experiment 1 while further minimizing procedural differences between tasks by intermixing

perceptual and VSTM trials of differing delays (Figure 3A). Error increased monotonically with

delay duration ($t(19) = 2.85$, p $= 0.010$. Figure 3B), and perception trials had lower error than

VSTM trials, collapsing across delays ($t(19) = 3.33$, p $= 0.003$), indicating the expected loss of

information in VSTM relative to perception. As in Experiment 1, error decreased with increasing

confidence during both perception ($t(19) = -7.56$, p $< 0.0001$) and VSTM trials ($t(19) = -8.99$, p

< 0.0001), indicating that confidence reliably tracked performance at the group level (Figure 3C

& 3D). Average confidence was lower on VSTM (mean = 2.59, SEM = 0.133) than on

perception trials (mean = 2.71, SEM = 0.132; $t(19)$ = 2.97, p = 0.0077; Supplementary Figure 1).

*Figure 3.* Task and behavior for Experiment 2. (A) Perceptual trials (delay 0.6 seconds) and VSTM trials (delay between 3.45 and 12 seconds) were intermixed within blocks. (B) Error increased with increasing delay length, indicating a loss of information when the orientation needed to be maintained in VSTM. (C) Response error distributions show a clear scaling with confidence. (D) Error decreased as confidence increased in both perceptual and VSTM trials. Shaded bands and error bars indicate ± 1 SEM.

Importantly, we replicated the positive relationship between perceptual and VSTM

metacognition with quantitatively better model fits in a new set of subjects. Using confidence-

error correlations (Figure 4A) perceptual metacognition robustly predicted VSTM metacognition

in the one-predictor basic model (slope = 0.60, t = 5.21 predictor $R^2$ = 0.60, p < 0.0001), the

three-predictor model controlling for average confidence and error (slope = 0.60, t = 4.91,

predictor $R^2 = 0.59$, p = 0.0001), and in the five-predictor model controlling for task-specific

confidence and error (slope = 0.59, t = 4.44, predictor $R^2 = 0.58$, p = 0.0005. All covariate

predictors in both control models were non-significant (ps > 0.64). As in Experiment 1, some

covariate predictors were significantly correlated (Table 1; gray region), but no covariates were

significantly correlated with perceptual metacognition, the predictor of interest. This effect was

also observed when using $A_{ROC}$ as the metric of metacognition for the basic model (slope = 0.53,

t = 3.88, predictor $R^2 = 0.45$, p = 0.0011), the three-predictor model (slope = 0.52, t = 3.58,

predictor $R^2 = 0.44$, p = 0.002), and the five-predictor model (slope = 0.54, t = 3.43, predictor $R^2$

= 0.44, p = 0.004). All covariates in both control models were non-significant (ps > 0.52).

***Figure 4.*** *Replication of the positive relationship between perceptual and VSTM metacognition*
*in Experiment 2. (A) Same regression models as in Figure 2, indicating the cross-task*
*relationship using confidence-error correlations as the metric of metacognition. (B) Same as in*
*A, but with $A_{ROC}$ as the metric of metacognition. Dashed lines denoted 95% confidence intervals*
*on the linear fit.*

**Experiment 3.** This experiment was conducted to test whether correlated individual differences

in metacognition depended on the perception and VSTM tasks sharing the same task-relevant

stimulus feature (i.e., orientation). Task accuracy (% correct) and metacognitive efficiency (M-

ratio) for each task and subject are shown in Figure 5B. Accuracy was comparable between the

contrast perception task (mean = 78.6%, SEM = 0.012) an the orientation perception task (mean

= 80.4 %, SEM = 0.009; p = 0.33), as well as between the contrast perception task and the

orientation VSTM task (mean = 76.5%, SEM = 0.018%, p = 0.39), but differed significantly

between the orientation perception and the orientation VSTM tasks ($t(19) = 2.68$, p = 0.015).

This is party because four subjects reached the maximum contrast allowable by the staircase

(25%), so the VSTM task never got easier for them. In general, higher target contrasts were

needed in the VSTM task (mean = 15.5%, SEM = 0.012) compared to both the orientation

perception task (mean = 8.5%, SEM = 0.005; $t(19) = 6.50$, p < 0.0001), and the contrast

perception task (mean = 6.2%, SEM = 0.005; $t(19) = 6.49$, p < 0.0001), indicating that more

signal was needed in the VSTM task to achieve threshold performance. Metacognitive

efficiency, on the other hand, did not significantly differ between tasks (ps > 0.16; see

Supplementary Figure 3 for full posterior distributions of M-ratio for each task, and comparisons

between tasks), and was well below the optimal M-ratio of 1 for all tasks (contrast perception:

mean M-ratio = 0.4, HDI = [0.27, 0.54], orientation perception: mean = 0.44, HDI = [0.30, 0.59],

orientation VSTM: mean = 0.53, HDI = [0.38, 0.68]). Mean confidence (see Supplementary

Figure 1) also did not differ between either of the perceptual tasks and the VSTM task (ps >

0.08), but was significantly lower in the contrast perception task as compared to the orientation

perception task ($t(19) = 3.06$, p = 0.006).

To our primary question of correlated individual differences in metacognition, we observed a

large positive correlation between metacognitive efficiency in the orientation perception task and

orientation VSTM task (Figure 5C; rho = 0.90, HDI = [0.58 0.99], p < 0.0001), consistent with

the results of Experiments 1 and 2. However, we did not observe a significant correlation

between metacognition in the contrast perception task and the orientation VSTM task (rho =

0.32, HDI = [-0.32 0.85], p = 0.32), suggesting that the correlation between perception and

VSTM metacognition depends on both tasks sharing the same task-relevant stimulus feature.

Interestingly, metacognition was also not correlated between the two perceptual tasks (rho =

0.24, HDI = [-0.35 0.76], p = 042), again underscoring the importance of the similarity of

stimulus feature used. We also computed the difference between these correlation distributions

and found that the correlation between orientation perception and orientation VSTM memory

was significantly larger than the correlation between contrast perception and orientation

perception (HDI = [0.07 1.27], p = 0.027) and trending larger than the correlation between

contrast perception and orientation VSTM (HDI = [-0.03 1.24], p = 0.063).

**Figure 5.** *Tasks, behavior, and metacognitive correlations from Experiment 3. (A) To compare metacognition across discrimination tasks while varying task and the task-relevant stimulus feature, subjects performed 1) a contrast perception task, judging which stimulus contained a higher contrast grating, 2) an orientation perception task, judging whether the two stimuli had the same orientation, and 3) an orientation VSTM task, judging whether a memorized target grating had the same orientation as a probe grating that appeared 3 seconds later. (B) Individual subject accuracies (proportion correct) and estimates of metacognitive efficiency (M-ratio) for each task. Note that because M-ratio for each subject is estimated in the same model, estimates are not fully independent (24). (C) Posterior distributions of cross-task correlations in metacognitive efficiency, which reveal a strong positive correlation between orientation perception and orientation VSTM metacognition, but not between other tasks.*

## Discussion

Metacognition is an important aspect of decision-making (25,26), learning (27), development (28), and perhaps certain aspects of conscious experience (29,30), and can be compromised in psychiatric disorders (31–33). It is currently unclear whether an individual with good

metacognitive ability in one domain also has good metacognition in other domains. In

Experiment 1, we found that individuals with more accurate metacognition in perceptual

judgments also showed more accurate metacognition in a VSTM task requiring stimulus

maintenance over a 7 second delay period. This relationship was present when using two

different measures of metacognitive performance and regression models controlling for task

performance and mean confidence revealed that this effect was not driven by correlated

individual differences in task performance or confidence biases. We then replicated these

findings in Experiment 2 with a new set of subjects using a task that intermixed perceptual and

VSTM trial types within blocks. Intermixing trial types in Experiment 2 more than doubled the

proportion of variance in VSTM metacognition explained by perceptual metacognition in the

models using error-confidence correlations relative to Experiment 1 when trial types were

blocked (mean increase in $R^2 = 0.30$, a factor of ~2.2), highlighting the importance of

minimizing procedural differences between tasks. A comparable increase across experiments

was not seen, however, when using the AUC metric, which already showed a very large effect

size in both experiments and across all models (mean $R^2 = 0.45$, Cohen's $d = 1.81$). In

Experiment 3 we compared an orientation VSTM task to an orientation perceptual task and a

contrast perception task. We again found a large positive correlation ($R^2 = 0.81$) between

metacognition in the orientation VSTM and orientation perception task, but not between the

orientation VSTM and contrast perception task, nor between contrast perception and orientation

perception tasks, highlighting the importance of both tasks sharing the same relevant stimulus

feature. Importantly, given known biases in VSTM metacognitive judgments (34), metacognition

in Experiment 3 was quantified within a signal detection theory model (20,24) that controls for

confidence biases and task performance. Taken together, these results provide the first evidence

in humans for a medium-to-high positive correlation between an individual's metacognitive

abilities in perception and VSTM, when both domains share a common stimulus feature.

The present results contrast with recent experiments examining the relationship between

metacognition of visual perception and long-term memory, which have typically observed no

correlation (4–6; but see 7). We reason that, in contrast to long-term memory, VSTM for a given

stimulus feature is thought to rely on the same neural representations that support perception of

that stimulus feature (11–14), and this may underlie the cross-task correlation in metacognitive

performance. This explanation follows naturally from "first-order" models of metacognition

according to which confidence and task performance are driven by the same internal

representation of stimulus evidence (35–38). For example, in signal detection theoretic models,

the absolute distance of the decision variable from the decision criterion is a proxy for

confidence (39,40). Thus, if perception and VSTM were supported by the same internal

representation of the stimulus, then the computation of confidence across the two tasks would

also be based on the same representations, leading to correlated behavior. "Second-order"

models of metacognition, in contrast, posit an architecture with a secondary confidence read-out

process, which may be influenced by additional sources of noise (41) or other signals not directly

related to the stimulus, such as action-related states (42,43), cortical excitability (44), or arousal

(45,46).

The results of Experiments 1 and 2 are also compatible with second-order models of

metacognition, although several possible relationships between first- and second-order processes

could explain our findings. Shared first-order (sensory) representations across tasks might be

enough to produce a behavioral correlation despite separate second-order readout mechanisms.

Alternatively, both first- and second-order processes may be shared across tasks, or only the

second-order process shared, though this latter possibility is unlikely given existing neural

evidence for shared representations in visual regions across perception and VSTM (14,47,48).

The results of Experiment 3, however, provide support for 1st-order models because they suggest

that shared sensory representations underlie the cross-task correlation in metacognition. Because

metacognition was not reliably correlated when tasks differed in their relevant stimulus feature,

even when both tasks were perceptual, this points towards a first-order model of metacognition.

Yet another alternative is that the correlation was dependent on the task structure, for example,

because both orientation tasks involved same/different judgments. This account may also explain

why a previous report comparing metacognition for contrast and orientation judgments in the

context of a visual search paradigm did find correlated individual differences (8), but recent

work comparing a variety of perceptual paradigms with different task structures and stimuli did

not find a correlation (49).

Although the present findings are consistent with a domain-general model of metacognition for

perception and VSTM, correlations at the behavioral level raise further questions about what

specific aspects of metacognitive processing are shared. For example if one's ability to learn

stable confidence criteria over time improves metacognitive accuracy (38), then metacognitive

abilities may be high across domains for an individual with superior learning abilities, perhaps

related to recent work implicating hippocampal myelination in perceptual metacognition (50).

However, this need not imply that the underlying neural substrate responsible for computing the

appropriate levels of confidence is itself domain-general. Similarly, recent work has highlighted

specific factors beyond stimulus evidence that modulate confidence, leading to dissociations of

confidence and task performance within an individual (15,51,52). For example, spontaneous

trial-to-trial fluctuations in oscillatory neural activity in the alpha-band (8-13 Hz), which are

thought to reflect visual cortical excitability (53,54), have been shown to bias confidence ratings,

but not objective performance in a visual discrimination task (44). Perhaps a subject who is less

susceptible to such influences from sources not directly related to the difficulty of stimulus

discrimination would show better metacognition across different domains. Future work

examining neural correlates of metacognitive performance across different domains may

contribute in a substantive way to this issue. As an example, McCurdy and colleagues (7)

observed a positive correlation between metacognition of perception and recollection memory at

the behavioral level, but found distinct (as well as overlapping) neural structures whose gray

matter volume related to metacognitive performance in the different tasks. This suggests that

only a portion of the processing stages or computations involved in generating confidence need

be shared across tasks in order to produce a behavioral correlation. Nevertheless, the experiments

reported here provide an important first step for future work by demonstrating a clear correlation

between metacognitive behavior in perception and VSTM.

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