# Comparing confidence biases in decision about perception and general knowledge



### Introduction

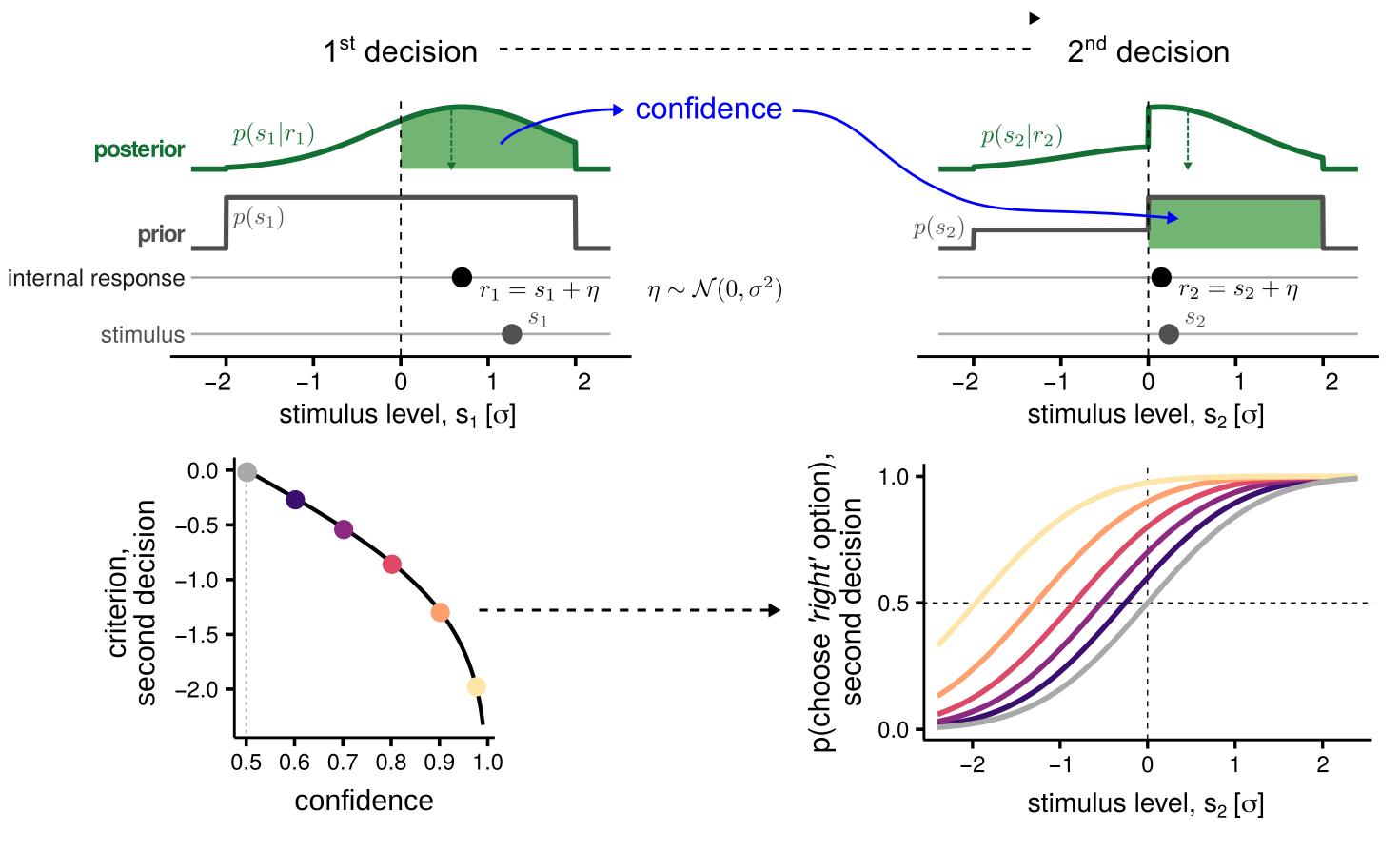
Confidence biases — over/under estimation of uncertainty — are a pervasive source of errors in decision-making.

Previous studies suggest correlations in confidence ratings across decision domains, hinting at a *domain-general* mechanism.

However, self-reported confidence may mix genuine confidence biases with response biases.

To address this, we used a *dual-decision* method [1] to explore confidence biases across domains.

## Computational modelling

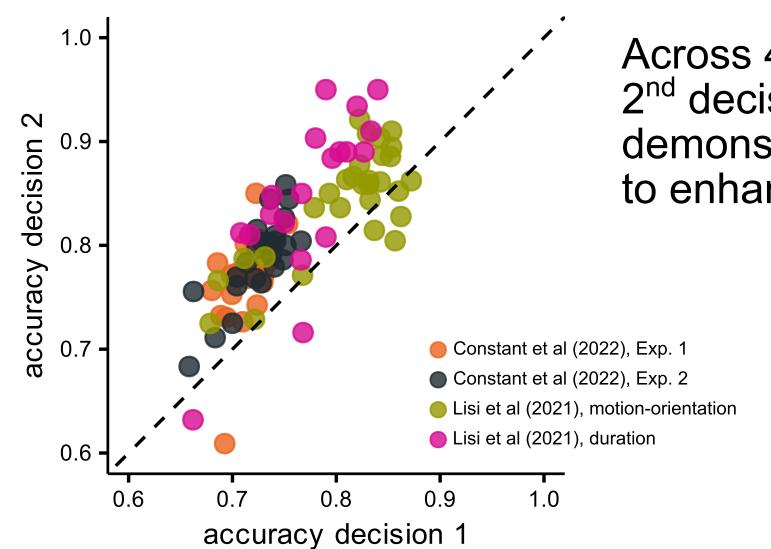


In the 'standard' Bayesian observer model, the only free parameter represents uncertainty, specifically the standard deviation of internal noise ( $\sigma$ ).

The Bayesian observer can be modified to incorporate a confidence bias by assuming it uses an inaccurate estimate  $(\hat{\sigma})$ of its own internal noise: biased

Confidence bias =  $\log -$ 

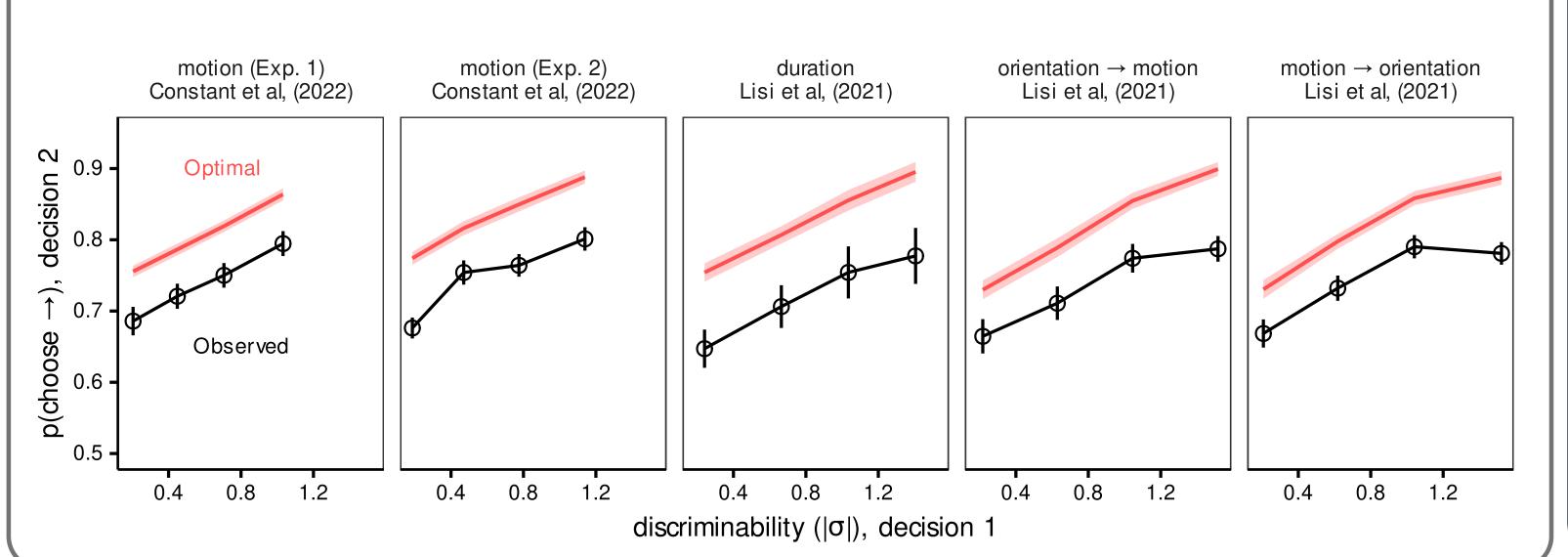
# Previous studies — underconfidence in perception



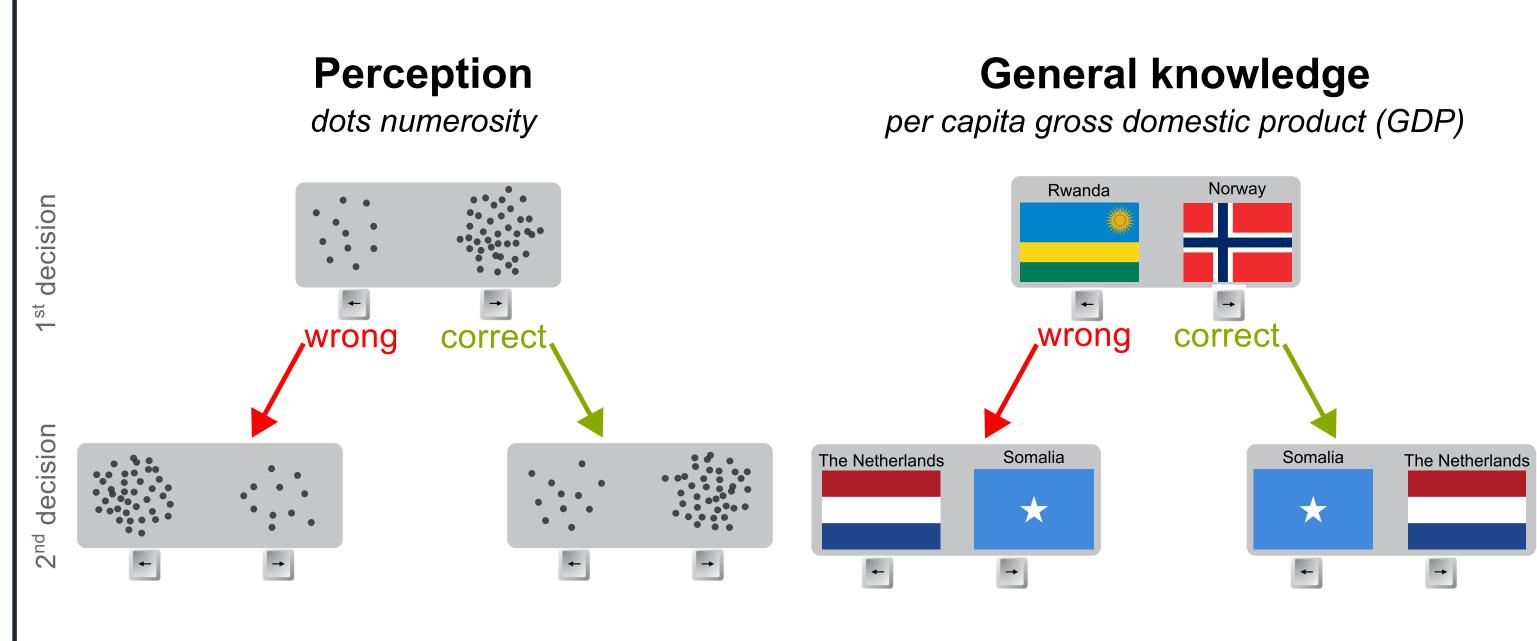
Across 4 experiments from previous studies,[1,2] 2<sup>nd</sup> decisions are more often correct than 1<sup>st</sup>, demonstrating effective use of confidence to enhance performance

> Participants choose → less frequently than optimal across tasks, suggesting they underestimate the reliability of their perception (indicative of an underconfidence bias)

confidence



## 'Dual-decision' tasks



**Two decisions per trial**: the correct response in the  $2^{nd}$  decision ( $\leftarrow$  vs  $\rightarrow$ ) is determined by accuracy in the 1<sup>st</sup> decision.

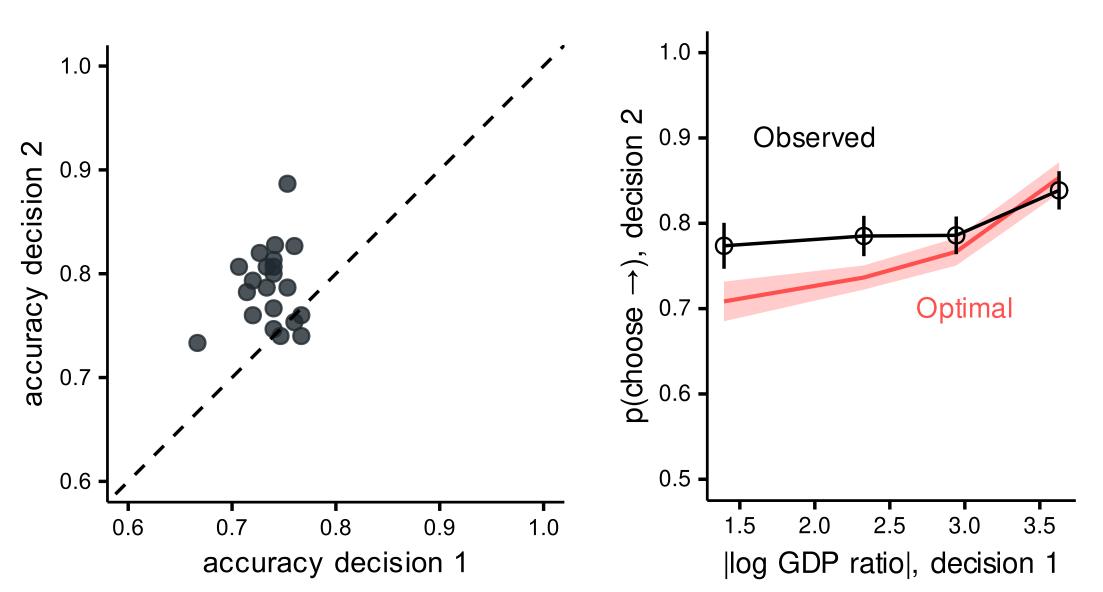
**Confidence as a prior**: confidence in the 1<sup>st</sup> decision serves as a prior to guide the 2<sup>nd</sup> decision.

**Assessing confidence biases**: the frequency of  $\rightarrow$  responses in the 2<sup>nd</sup> decision, compared to the optimal frequency, reveals participants' confidence biases.

### Results

#### **Experiment 1**

Participants (N=21) completed 150 trials of the GDP task.



Differently from previous results, in the general knowledge (GDP) task participants  $responded \rightarrow as$ frequently as the optimal Bayesian model.

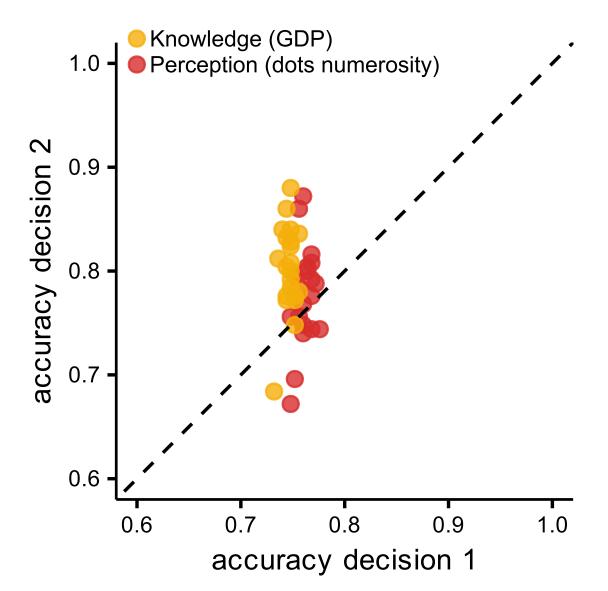
#### **Experiment 2**

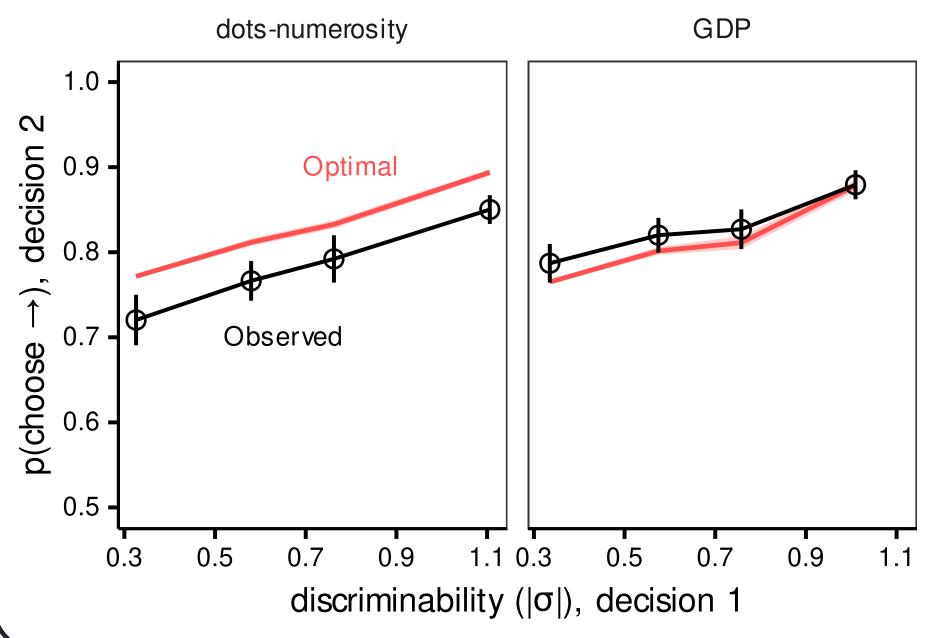
Within-subject comparison:

Participants (N=23) completed 250 trials each of the GDP and dots numerosity tasks in a randomized order.

Difficulty adjusted with a 3-up-1-down staircase, based on performance in Decision 1.

Accuracy in decision 2 exceeded decision 1 in both tasks, with a larger effect size in knowledge tasks (odds ratios: 1.1 for perception, 1.4 for knowledge).





Participants responded → less frequently than the optimal Bayesian model in the perceptual task, suggesting under-confidence, but showed well-calibrated confidence in the knowledge task

## Fat-tailed internal noise distribution?

Simulations show that excess kurtosis does not influence the estimated confidence bias.

The optimal strategy using internal response from decision 1 as criterion shift for decision 2 holds for any symmetric,

unimodal distribution (scan QR code for a proof).

excess kurtosis Data simulated using Pearson Type VII distributions with varying excess kurtosis, refitted with a biased Bayesian model assuming Gaussian noise

### Discussion

Across all studies, participants effectively used the dual-decision task structure to improve decision 2 performance.

Modeling indicated an under-confidence bias in perceptual tasks, but no bias in the general knowledge task.

Within-participant comparison in Exp. 2 showed a significant difference in confidence bias. A meta-analysis found no bias in the GDP task and a greater bias in perceptual tasks — 95% CI [0.30, 1.59] In units increase.

These results may reflect over-confidence in knowledge tasks [3], combined with a cautious strategy to mitigate errors due to confidence "noise."

A limitation is that only a GDP-based knowledge task was tested, leaving open the question of whether this pattern extends to other knowledge-based tasks.

