

Trial-by-trial Fluctuations in Decision Criterion Shape Decision Confidence

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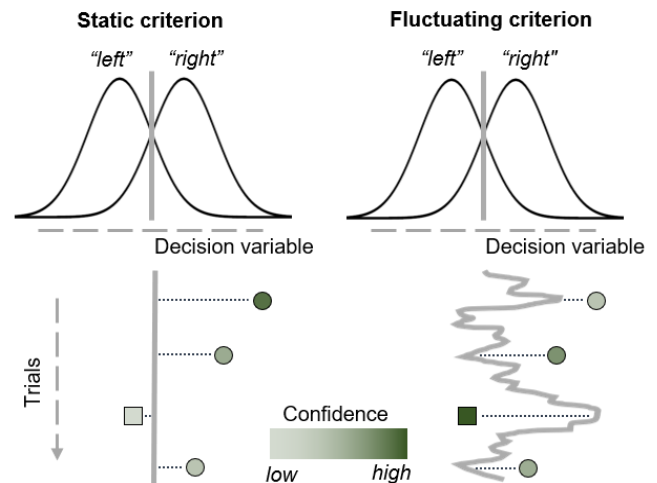
Abstract

1 Many choices we make are accompanied by a sense of
2 confidence. Within Signal Detection Theory (SDT),
3 confidence is traditionally conceptualized as the
4 absolute distance between a decision variable and a
5 decision criterion. Whereas this criterion is typically
6 modeled as being stable over time, increasing evidence
7 suggests that it undergoes trial-to-trial fluctuations.
8 Based on theory and model simulations, we predict that
9 fluctuations in the decision criterion shape confidence.
10 Using the Hierarchical Model for Fluctuations in
11 Criterion (hMFC) to obtain single-trial criterion
12 estimates, we found robust evidence for this hypothesis
13 across 15 datasets. When analyzing each dataset
14 individually the effect was found in 13 out of 15 datasets,
15 indicating a stable pattern across a variety of paradigms
16 and confidence scales. Our results demonstrate that
17 what has been previously interpreted as noise in
18 confidence ratings, instead reflects variability driven by
19 fluctuations in the decision criterion.

20 **Keywords:** decision making; decision confidence; signal
21 detection theory; criterion fluctuations

Introduction

22 To understand the computational mechanisms
23 underlying decision confidence researchers often rely
24 on computational models. Signal detection theory
25 (SDT) is one of the most influential computational
26 frameworks of decision making (Green & Swets,
27 1966), and easily allows to model decision confidence.
28 SDT assumes that observers generate an internal
29 representation of the relevant stimulus information,
30 typically referred to as the decision variable (DV). To
31 make a binary decision, this DV is compared against
32 an internal decision criterion (Figure 1; static criterion,
33 gray line). Within SDT, decision confidence can be
34 computed as the absolute distance between the
35 decision variable and the decision criterion, with a
36 larger distance indicating higher confidence (Hebart et
37 al., 2016; Kepecs et al., 2008; Shekhar & Rahnev,
38 2021).



39 **Figure 1.** For identical decision variables, predicted
40 confidence differs depending on whether the criterion is static
41 or fluctuating.

42 By shifting the criterion, SDT naturally
43 accounts for biases (i.e., when one response option is
44 chosen more often than the other). For example,
45 shifting the criterion to the right will result in overall
46 more 'left' responses, while leaving the overall
47 sensitivity unaffected. To estimate this decision
48 criterion it is assumed that the criterion remains
49 constant over all trials. However, increasing evidence
50 suggests that computational variables, like the
51 decision criterion, are not constant, but instead
52 fluctuate over trials (Ashwood et al., 2022; Cowley et
53 al., 2020; Gupta & Brody, 2022; Roy et al., 2021;
54 Vloeberghs et al., 2024). Importantly, trial-by-trial
55 fluctuations in the decision criterion should in theory
56 also affect the computation of confidence, given that
57 confidence is thought to reflect the distance between
58 the criterion and the decision variable (Figure 1;
59 fluctuating criterion). Therefore, the current study aims
60 to examine whether confidence ratings are shaped by
61 trial-to-trial fluctuations in the decision criterion.

Methods

To investigate how criterion fluctuations shape confidence we first simulated trials from a SDT observer with a fluctuating criterion. On each trial, a decision variable was sampled from a normal distribution with mean = [-3, -2, -1, 1, 2, 3] and standard deviation = 1. To obtain a binary response, the decision variable was compared to the decision criterion, which fluctuated trial-to-trial according to a first-order autoregressive process with an autoregressive coefficient = .9995 and noise standard deviation = .05. Confidence was quantified as the absolute distance between the decision variable and criterion.

Next, in a pre-registered analysis we selected 15 datasets from the Confidence Database (Rahnev et al., 2021), resulting in 382 subjects with 463,088 trials in total. Trial-to-trial fluctuations in decision criterion were estimated using Hierarchical Model for Fluctuations in Criterion (hMFC; Vloeberghs et al., 2024). Using (nested) linear mixed models, confidence was predicted based on stimulus strength, stimulus direction, and single-trial criterion estimates across all datasets and for each dataset individually.

Results

SDT simulations confirm that confidence is affected by criterion fluctuations (Figure 2A). More specifically, when the criterion fluctuates towards the right (yellow), on average the distance between a decision variable favoring the “left” stimulus and the criterion will increase, leading to higher confidence. On the contrary, when the criterion fluctuates towards the left (blue) the distance between a decision variable favoring the “left” stimulus and the criterion will decrease, leading to lower confidence. For “right” stimuli, this pattern flips.

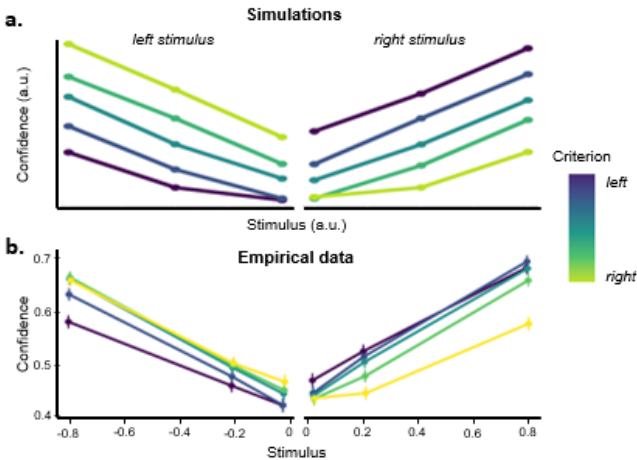


Figure 2: A. Simulations from the SDT framework reveal a clear relation between trial-to-trial criterion fluctuations and decision confidence. **B.** Analysis of 15 datasets show a similar pattern as the simulations, confirming the hypothesis that criterion fluctuations shape decision confidence.

In line with the model simulations, we find across 15 datasets that confidence is significantly predicted by an interaction between stimulus direction and criterion ($\beta = -0.053$, $SE = 0.002$, $F(1,458132) = 529.376$, $p < .001$), showing that confidence is shaped by criterion fluctuations (Figure 2B). Moreover, when analyzing the datasets individually this crucial two-way interaction between stimulus direction and criterion was significant in 13 out of 15 datasets, demonstrating a robust effect independent of the specific paradigm or confidence scale (Figure 3).

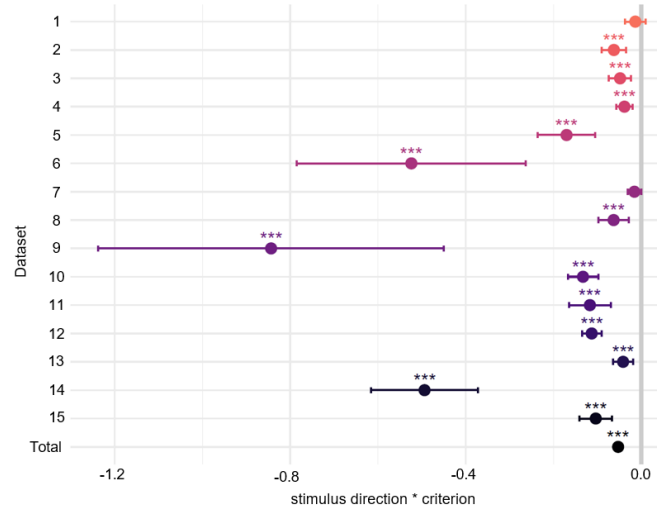


Figure 3. Parameter estimates for the crucial interaction between stimulus direction and decision criterion and their 95% confidence intervals, separately for each dataset. The estimate plotted at the bottom, labelled “Total”, shows the parameter estimate from the model fitted across all datasets (taking into account the nesting of subjects within datasets).

Conclusion

Our study provides robust evidence that confidence is shaped by trial-to-trial fluctuations in the decision criterion. This finding shows that variability in confidence judgements, which is typically assumed to reflect noise, actually reflects genuine computation of confidence as distance-to-criterion. One important consequence of quantifying confidence as the distance between the decision variable and a fluctuating criterion, is that confidence will naturally become autocorrelated across trials, a phenomenon referred to as the confidence leak (Rahnev et al., 2015). Whereas in the literature these two observations (variability and autocorrelation) are typically explained via different mechanisms, fluctuations in decision criterion could provide a parsimonious account for both explaining variability and autocorrelation in confidence ratings. With this work new insights are provided into the computational underpinnings of decision confidence.

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