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Dissociable Effects of Uncertainty in Perceptual and Cognitive Control

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control;

Abstract

7 The ability to manage uncertainty is a hallmark of 45 flexible control. But uncertainty is not a unitary 46 8 9 construct: it arises from various sources that 47 10 challenge diverse processing domains - from 48 11 ambiguous perceptual inputs to conflicting cues 49 about task-relevant features. To disentangle 50 12 13 uncertainty sources in perceptual and cognitive 51 control, we independently manipulated perceptual 52 14 15 uncertainty (relative choice evidence) and task 53 16 uncertainty (uncertainty about relevant feature 54 17 sets) in a dynamic perceptual decision task. Across 55 18 three experiments, we observed a double 56 19 dissociation in behavioral effects: perceptual 57 20 uncertainty reduced accuracy, while task 58 21 uncertainty primarily slowed response times. 59 22 Conceptually replicating and extending prior work, 60 functional MRI revealed robust engagement of a 23 24 fronto-thalamic network in response to task, but 25 not perceptual, uncertainty. We propose that 26 thalamocortical circuits track uncertainty in a 27 differentiated fashion to exert domain-specific 28 control. By establishing robust and dissociable 29 effects, this work provides a foundation for 30 understanding how the human brain manages 31 diverse uncertainties. 61

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33 Keywords: uncertainty; cognitive34 perceptual control

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36 How limited or contrasting information (i.e., 37 uncertainty) is managed by perceptual, choice, and executive neural systems is fundamental to 38 understanding flexible control in the brain (Halassa 39 & Kastner, 2017; Okazawa & Kiani, 2022). Here, we 40 41 manipulate two uncertainty sources to establish 42 their consequences on behavior, computations, and 43 neural engagement.

Perceptual and task uncertainty

To induce distinct challenges to cognitive and perceptual processing, we adapted a task (Kosciessa et al., 2021; 2024) that requires participants to sample up to four visual features in a dynamic compound stimulus (Kéri et al., 2004; Mante et al., 2013) to indicate either the predominant color (red/green), movement direction (left, right), size (small, large), or saturation (low, high) (Fig. 1). Upon stimulus offset, a probe gueried the prevalence of one target feature via 2-AFC. A cue informed participants about the feature set from which a probe would be selected. Task uncertainty was manipulated by cueing either one or features. Perceptual four uncertainty was manipulated via relative choice evidence, individually titrated for each feature to 65% and 90% accuracy.



Figure 1: Schematic uncertainty manipulations.

Experiments & Methods

64 Three experiments investigated behavioral
65 uncertainty effects. Experiments were approved by
66 local ethics boards (ECSS or METC). Participants
67 were healthy adults between 16 and 36 years of age.
68 **EXP1: EEG (N=15).** Two 144-trials task runs were
69 acquired during 64-channel EEG acquisition.

70 **EXP2: MRI & EEG (N=23).** Three separate runs 71 were acquired during one 3T fMRI and two 40-72 channel EEG sessions (N=20) prior to non-invasive

73 brain stimulation.

74 EXP3: Behavioral Control (N=26). High task 75 uncertainty prevents response preparation prior to 76 probe onset while the target feature remains 77 unknown. RT slowing could therefore reflect added 78 response mapping demands. To test this possibility, 79 we performed a control experiment in which 80 left/right response mappings were either fixed or 81 variable within blocks. Task uncertainty was varied 82 at low perceptual uncertainty.

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Figure 2: Behavioral uncertainty sensitivity.

Results: Behavior

87 We observed robust behavioral effects (Fig. 2) that were marked by a double dissociation: perceptual 88 89 uncertainty predominantly decreased accuracy 90 (expected based on evidence titration), whereas 91 task uncertainty predominantly slowed responses. 92 Response slowing due to task uncertainty did not 93 differ between fixed and variable response 94 mappings, suggesting that it does not primarily 95 reflect increased motor preparation demands 96 following unpredictable probes.

Results: fMRI

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98 Replicating prior work using parametric manipulation 99 (Kosciessa et al., 2021; 2024), task uncertainty 100 increased activation of a multi-demand network 101 during stimulus presentation (Fig. 3), including in the 102 mediodorsal thalamus. Decreased activation was 103 observed in medial prefrontal cortex (Muller et al., 104 2019). Task uncertainty modulation was more 105 pronounced when perceptual uncertainty was low. 106 Notably, the network was not sensitive to perceptual 107 uncertainty levels. Results indicate that fronto-108 thalamic network engagement modulation by task 109 uncertainty is robust, replicable, and source-specific. 110



Figure 3: fMRI: First latent variable (*permuted* p <
0.001) from task PLS analysis of condition-specific
stimulus regressors. BSR = bootstrap ratio. N = 23.

Conclusion

116 Decisions are subject to various uncertainties (Bach 117 & Dolan, 2012). By disentangling uncertainty about 118 relevant task sets from that regarding perceptual 119 choice information, we show that decision behavior 120 and neural engagement are differentially sensitive to 121 either uncertainty source. This provides a starting 122 point for clarifying specific computational and neural 123 adjustment to either uncertainty type. We currently 124 probe such via multimodal fMRI and EEG 125 neuroimaging recorded during task performance, and 126 test effects of deep brain ultrasound stimulation on 127 behavior, computations, and neural dynamics.

Acknowledgements. JQK was supported by theRadboud Excellence Initiative.

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