

Response to Changes in Variability During Movement Under Risk

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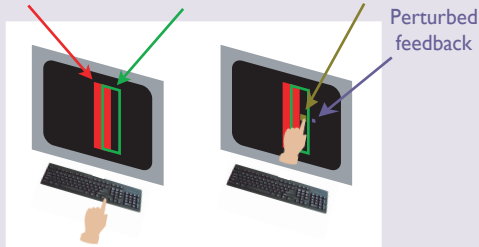
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Introduction

- Movements under risk are nearly optimal¹⁻². The optimal strategy, maximizing expected gain, requires the subject to *shift* the aimpoint from the center of the target away from the penalty region;
- In these tasks, humans take into account changes in task-relevant variability³⁻⁴.
- How do they estimate task-relevant variability?
- Researchers typically study reward-based learning of a mean, but not of a variance.

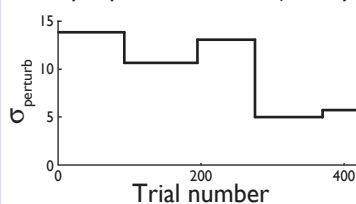
Task: Speeded reach under risk

Penalty: -5 points Target: 1 point True endpoint



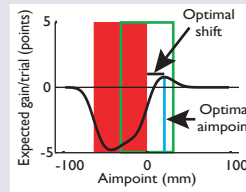
- Horizontal random Gaussian perturbation
- Points awarded based on perturbed endpoint
- $\sigma_{\text{effective}} = (\sigma_{\text{motor}}^2 + \sigma_{\text{perturb}}^2)^{1/2}$
- Random “sample-and-hold” perturbation variability over trials:

Example perturbation trajectory

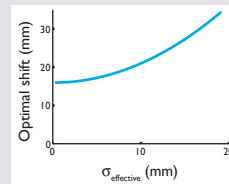


Gain Landscape

The optimal strategy, maximizing expected gain, requires the subject to *shift* the aimpoint from the center of the target away from the penalty region;

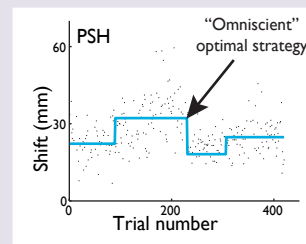


Optimal shift depends on $\sigma_{\text{effective}}$



But, subjects have to estimate σ_{perturb} dynamically.

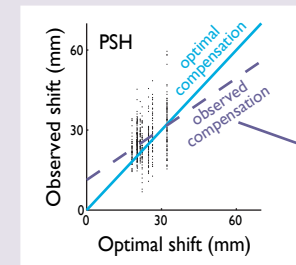
Sample data



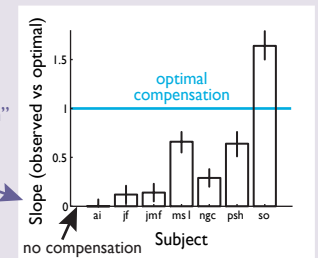
“Omniscient” optimal strategy based on current trial’s value of σ_{perturb}

Observed vs optimal shift

One subject



Summary of 7 subjects

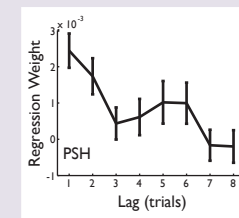


This subject, and all but one of the seven to varying extent, compensates for increased task-relevant variability by shifting the aimpoint further from the penalty.

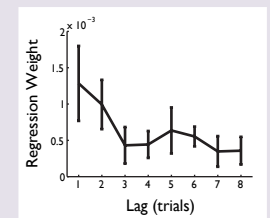
Predicting the current shift

We computed a linear regression of the shift as a function of the squared perturbations of the past eight trials.

One subject



Summary of 7 subjects



Humans dynamically estimate task-relevant variability by computing an exponentially-weighted average of recent squared perturbations.

References: 1. Trommershäuser et al., *Spat Vis* 2003. 2. Trommershäuser et al., *JOSA A* 2003. 3. Trommershäuser et al., *J Neurosci* 2005. 4. Tassinari et al., *SfN Annual Mtg* 2007. Support: NIH EY08266 Contact: landy@nyu.edu