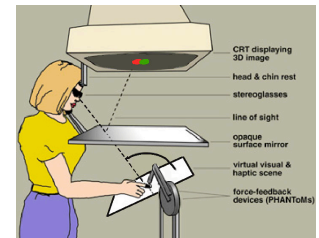


What if we change your motor variability?

Experiment: Increase “Motor Noise”

Changes in movement endpoints in response to visually-imposed changes in movement variability

Experimental set-up:



Trommershäuser, Gepshtein, Maloney, Landy, Banks (2005).
J. Neurosci., 25, 7169.

Experiment: Increase “Motor Noise”

Idea:

- finger visually represented by red point
- on each trial: unpredictable perturbation of the visual feedback of the finger tip by

$$\begin{bmatrix} \Delta x \\ \Delta y \end{bmatrix} \sim \text{Gaussian} \left(0, \begin{bmatrix} \sigma_{pert}^2 & 0 \\ 0 & \sigma_{pert}^2 \end{bmatrix} \right)$$

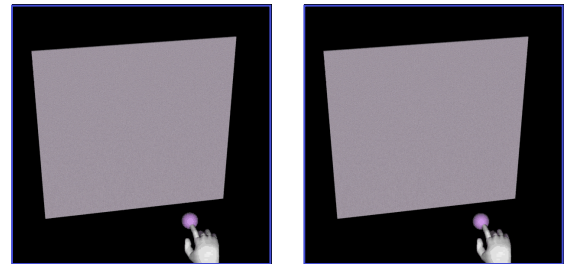
- Points are scored based on the perturbed finger position

⇒ Increase of subject's **effective movement variability** σ_{eff}^2

Trommershäuser, Gepshtein, Maloney, Landy, Banks (2005).
J. Neurosci., 25, 7169.

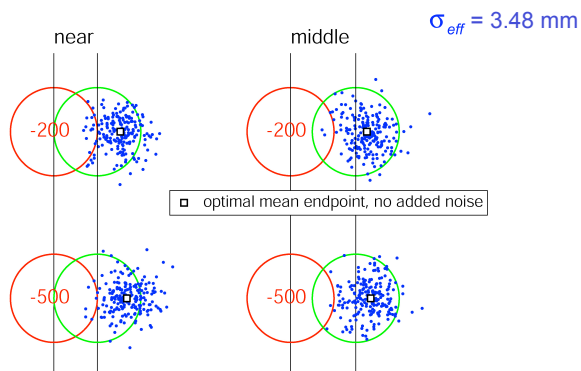
Experiment: Increase “Motor Noise”

Experimental manipulation of effective movement variability:



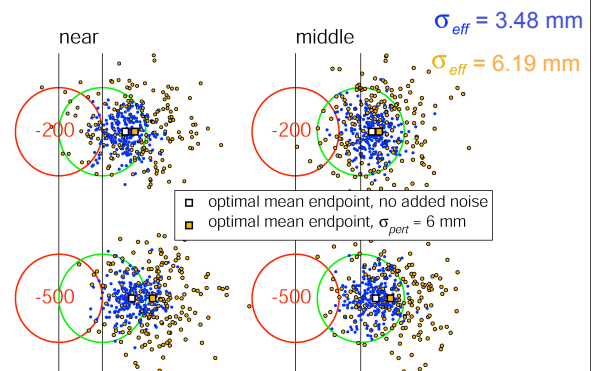
Trommershäuser, Gepshtein, Maloney, Landy, Banks (2005).
J. Neurosci., 25, 7169.

Prediction



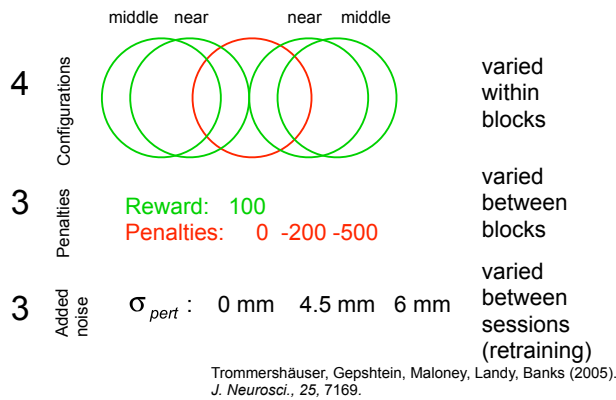
Trommershäuser, Gepshtein, Maloney, Landy, Banks (2005).
J. Neurosci., 25, 7169.

Prediction



Trommershäuser, Gepshtein, Maloney, Landy, Banks (2005).
J. Neurosci., 25, 7169.

Design



Design

Six subjects

1 practice session: 300 trials,
decreasing time limit

per noise condition:

1 learning session: 300 trials

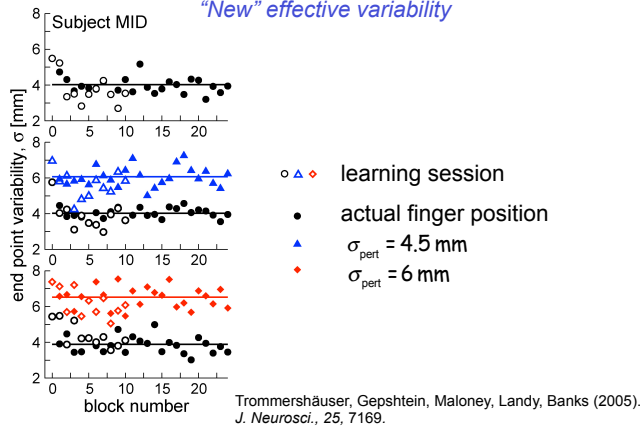
2 sessions of data collection: 360 trials each

40 repetitions per condition

Trommershäuser, Gepshtein, Maloney, Landy, Banks (2005).
J. Neurosci., 25, 7169.

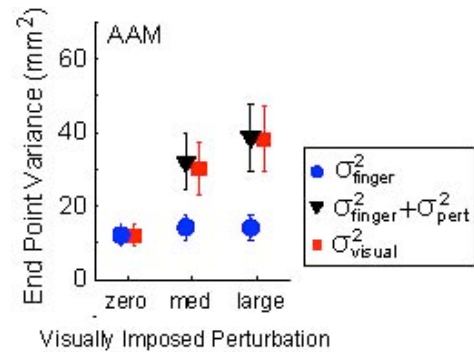
Results

"New" effective variability



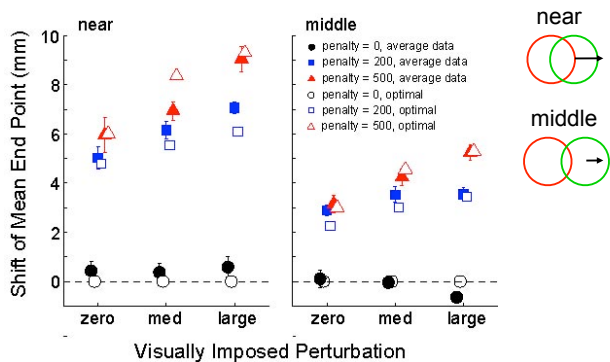
Results

Additivity of Variances



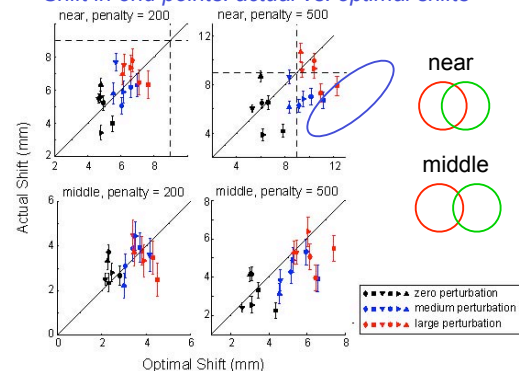
Results

Shift in end points: average subject data



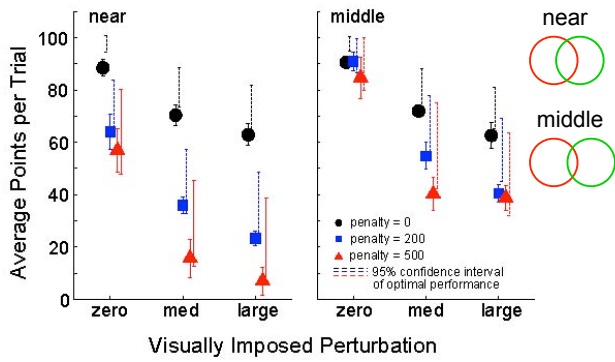
Results

Shift in end points: actual vs. optimal shifts



Results

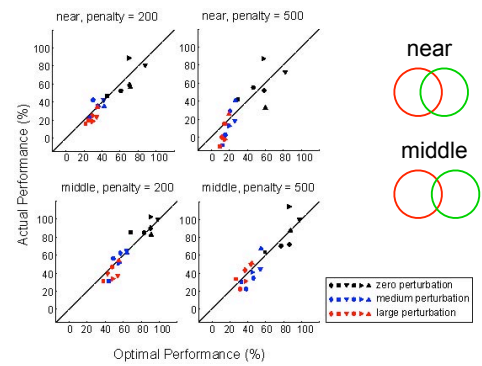
Scores: average subject data



Trommershäuser, Gepshtein, Maloney, Landy, Banks (2005).
J. Neurosci., 25, 7169.

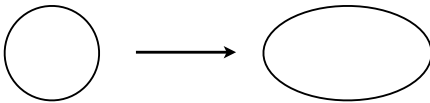
Results

Scores: actual vs. optimal performance

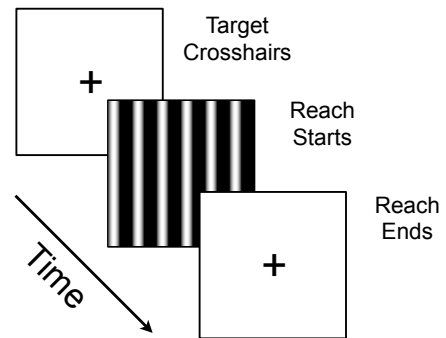


Trommershäuser, Gepshtein, Maloney, Landy, Banks (2005).
J. Neurosci., 25, 7169.

What if we change your motor variability, making it anisotropic?

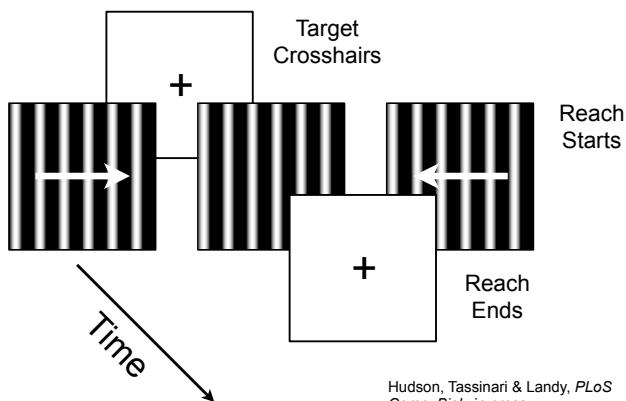


Experiment: Training: No-Drift Trials



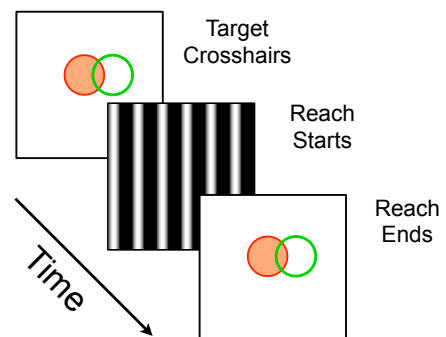
Hudson, Tassinari & Landy, *PLoS Comp. Biol.*, in press.

Experiment: Training: Drift Trials



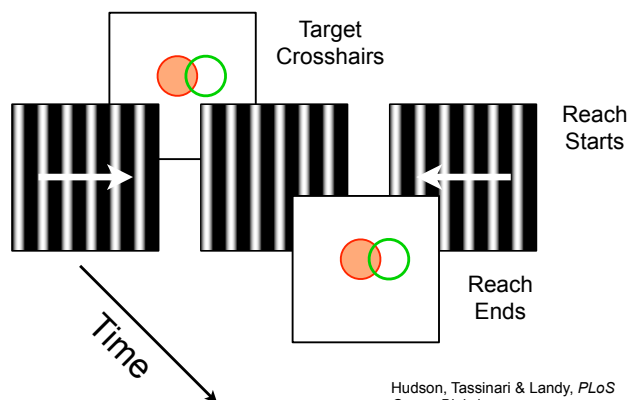
Hudson, Tassinari & Landy, *PLoS Comp. Biol.*, in press.

Experiment: Test: No-Drift Trials



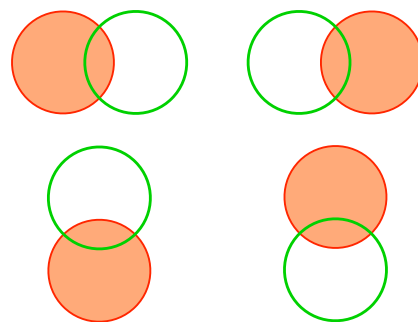
Hudson, Tassinari & Landy, *PLoS Comp. Biol.*, in press.

Experiment: Test: Drift Trials



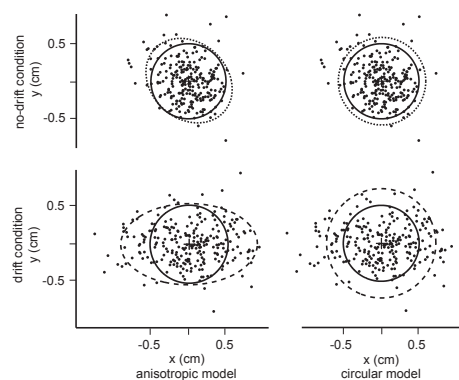
Hudson, Tassinari & Landy, *PLoS Comp. Biol.*, in press.

Payoff-Penalty Configurations



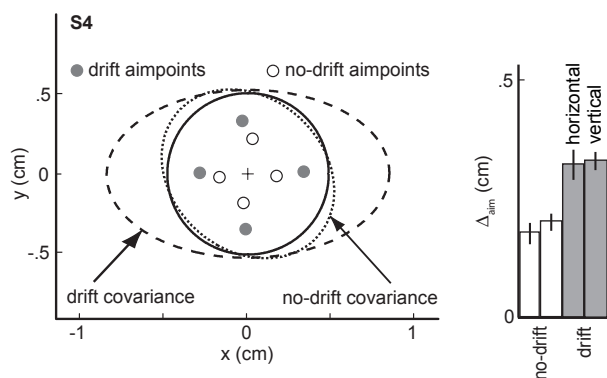
Hudson, Tassinari & Landy, *PLoS Comp. Biol.*, in press.

Results



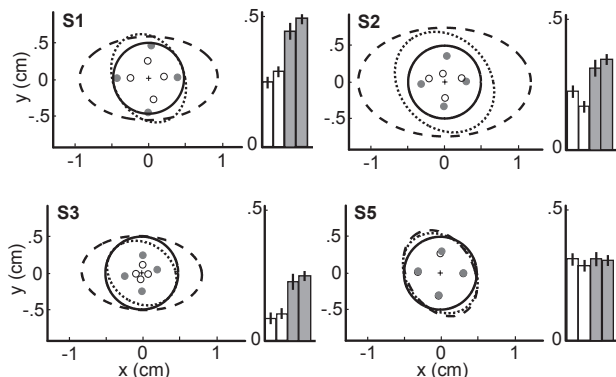
Hudson, Tassinari & Landy, *PLoS Comp. Biol.*, in press.

Results



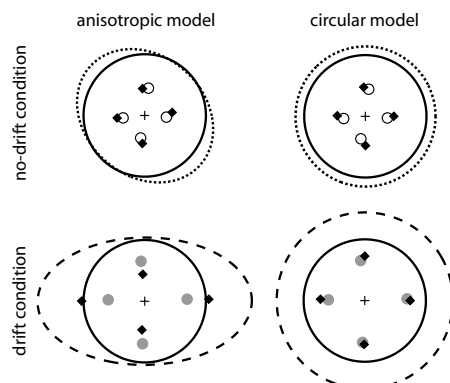
Hudson, Tassinari & Landy, *PLoS Comp. Biol.*, in press.

Results



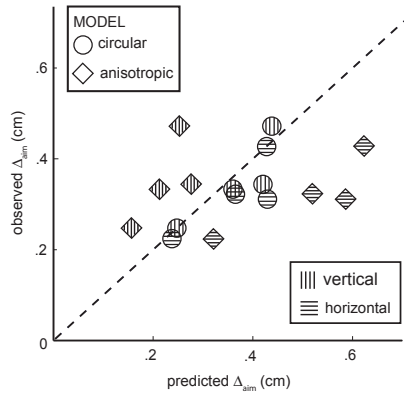
Hudson, Tassinari & Landy, *PLoS Comp. Biol.*, in press.

Results



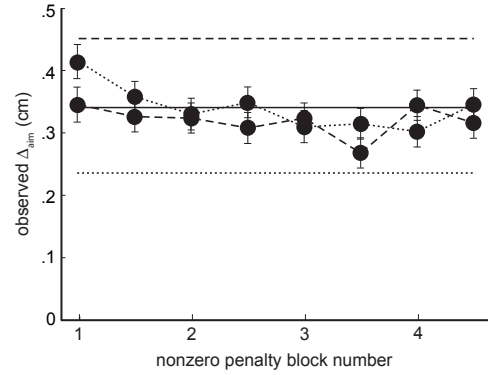
Hudson, Tassinari & Landy, *PLoS Comp. Biol.*, in press.

Results



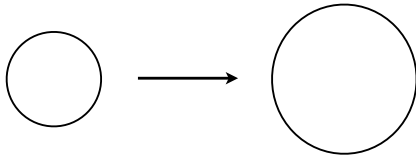
Hudson, Tassinari & Landy, *PLoS Comp. Biol.*, in press.

Results

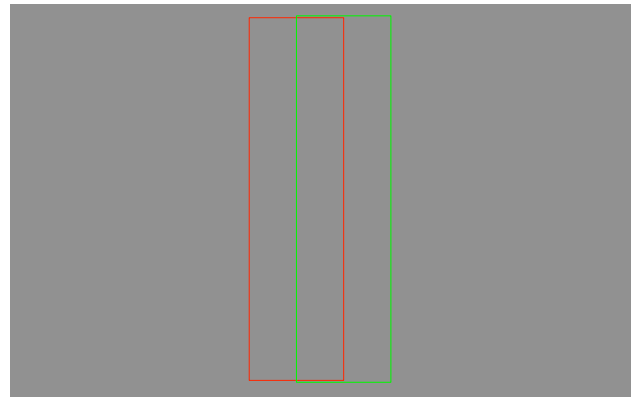


Hudson, Tassinari & Landy, *PLoS Comp. Biol.*, in press.

What if we change your outcome variability, trial-by-trial during a task? Will you learn the new variability? How?

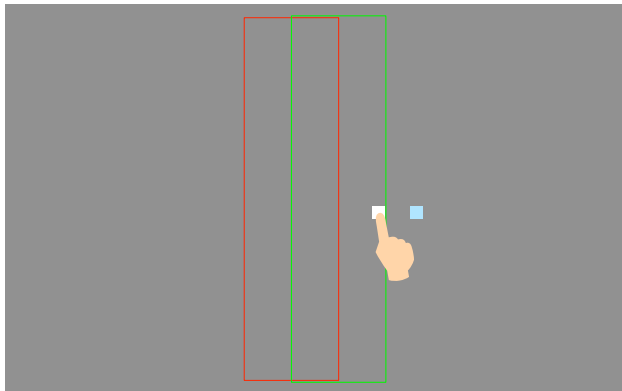


Experiment: Task



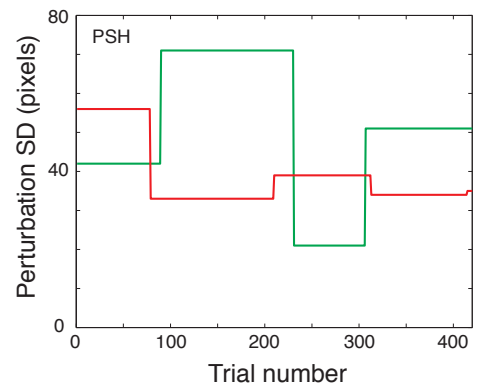
Landy, Daw & Trommershäuser, *J Neurosci*, in press

Experiment: Task



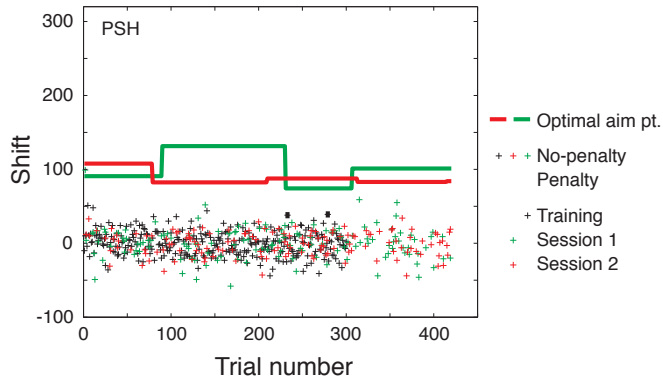
Landy, Daw & Trommershäuser, *J Neurosci*, in press

Experiment: Variance Variation



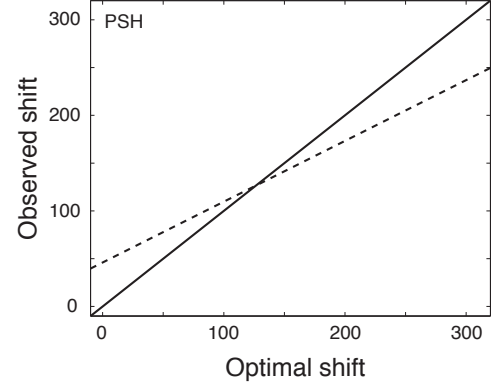
Landy, Daw & Trommershäuser, *J Neurosci*, in press

Experiment: Trial-by-Trial Data



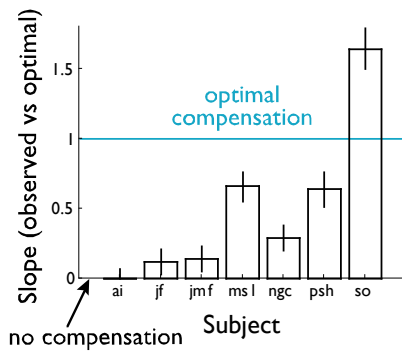
Landy, Daw & Trommershäuser, J Neurosci, in press

Experiment: Raw Data vs. Optimal



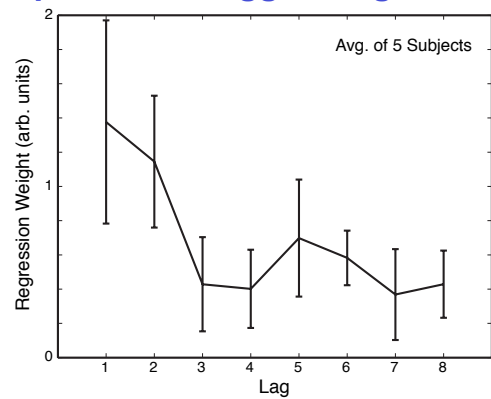
Landy, Daw & Trommershäuser, J Neurosci, in press

Experiment: Raw Data vs. Optimal



Landy, Daw & Trommershäuser, J Neurosci, in press

Experiment: Lagged Regression

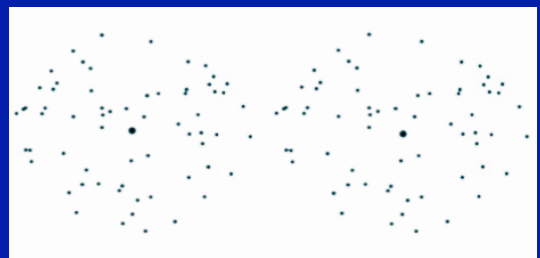


Landy, Daw & Trommershäuser, J Neurosci, in press

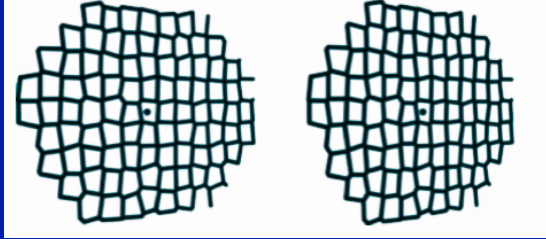
Do cue weights depend on local viewing conditions?

Hillis et al. (2004)

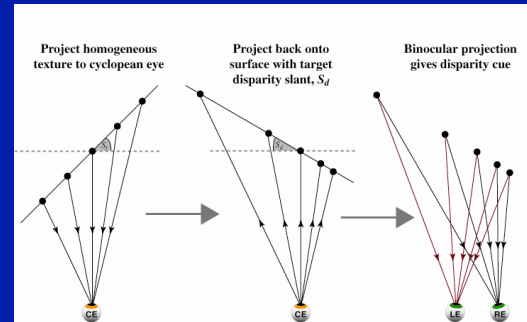
Stimuli – Disparity-only



Stimuli – Voronoi textures



Cue Conflict Stimuli



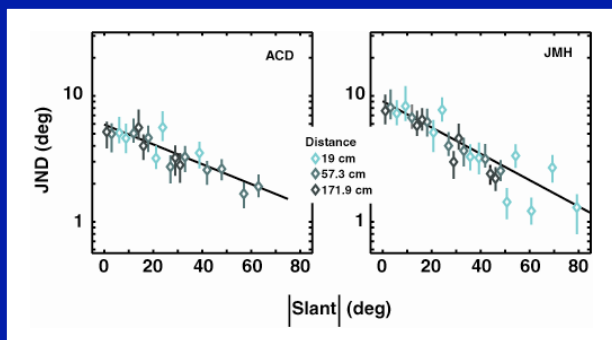
Methods

- Task: 2IFC slant discrimination
- Single-cue and two-cue blocks
- Opposite-sign slants mixed across trials in a block to avoid slant adaptation
- One stimulus fixed, other varied by staircase; several interleaved staircases
- Analysis: fit psychometric function to estimate PSE and JND

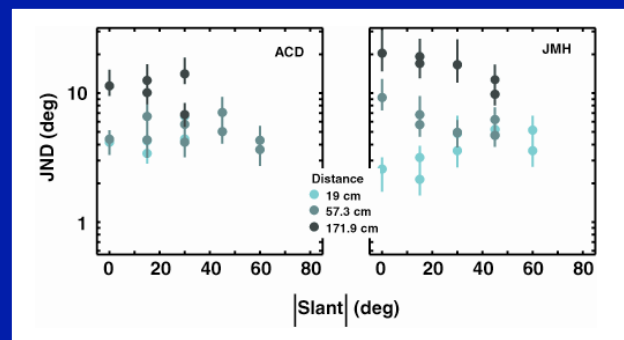
Outline

- Background: Optimal cue combination
- Methods: slant discrimination
- Single-cue results
- Two-cue results: perceived slant
- Two-cue results: JNDs
- Conclusions

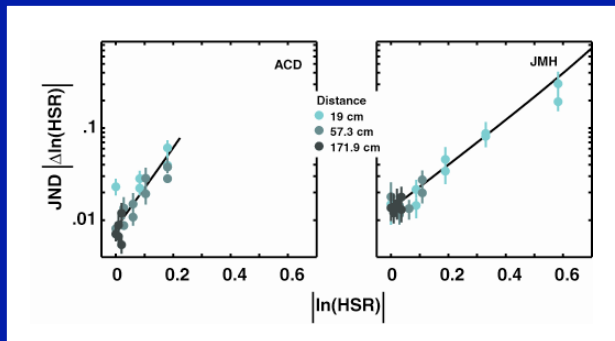
Single-cue JNDs: Texture



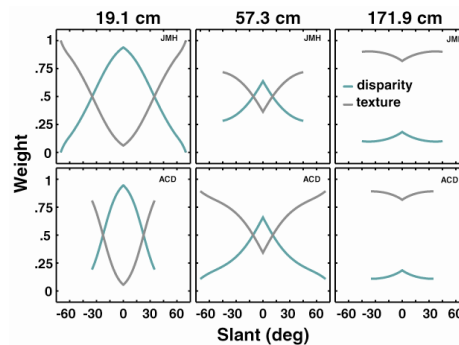
Single-cue JNDs: Disparity



Single-cue JNDs: Disparity



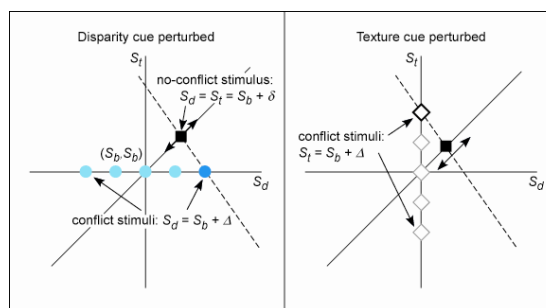
Predicted Cue Weights



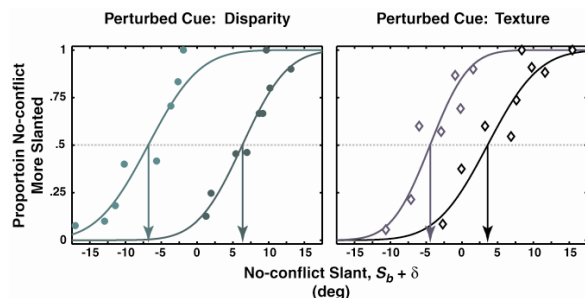
Outline

- Background: Optimal cue combination
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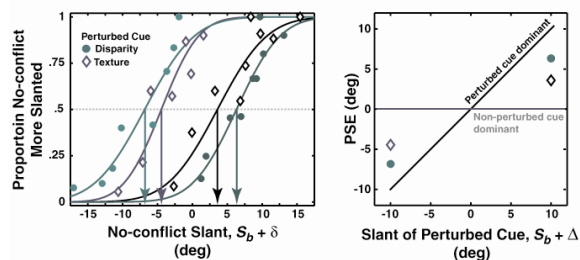
Cue Conflict Paradigm



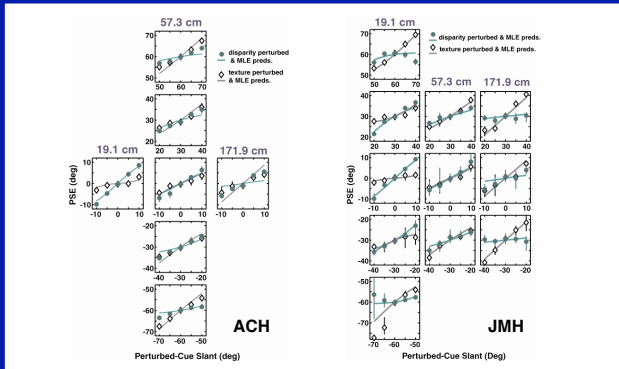
Determination of PSEs



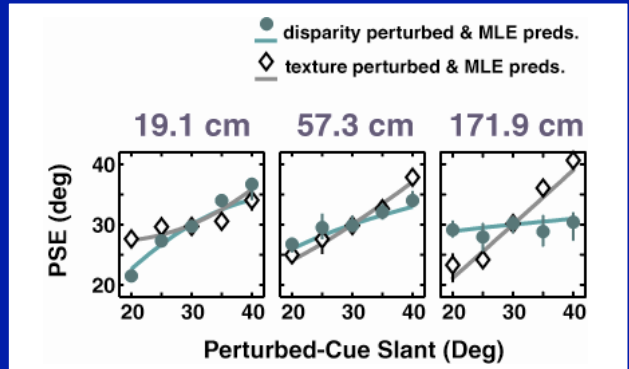
Determination of Weights



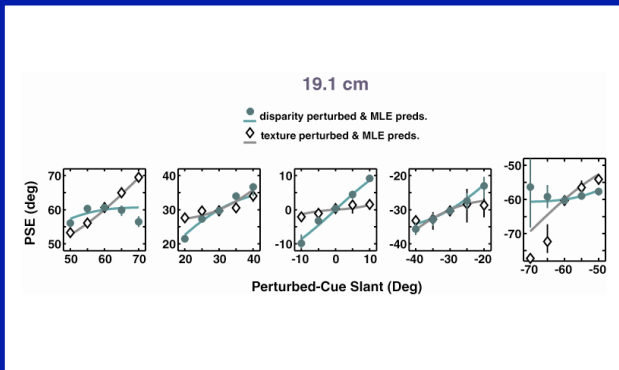
Full Two-Cue Dataset



Effect of Viewing Distance



Effect of Base Slant



Outline

- Background: Optimal cue combination
- Methods: slant discrimination
- Single-cue results
- Two-cue results: perceived slant
- Two-cue results: JNDs
- Conclusions

Improvement in Reliability with Cue Combination

If the optimal weights are used:

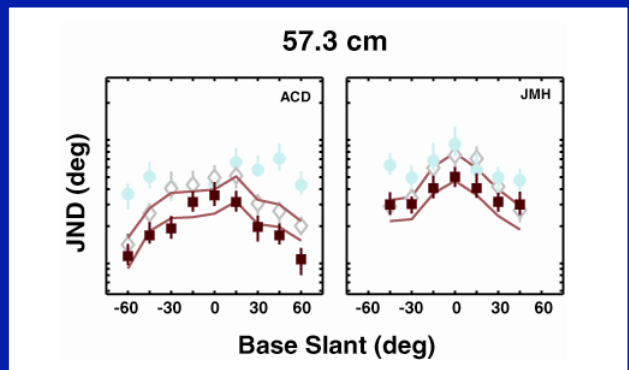
$$w_i = \frac{1/\sigma_i^2}{1/\sigma_i^2 + 1/\sigma_d^2} \quad \text{and} \quad w_d = \frac{1/\sigma_d^2}{1/\sigma_i^2 + 1/\sigma_d^2}$$

then the resulting variance

$$\frac{\sigma_i^2 \sigma_d^2}{\sigma_i^2 + \sigma_d^2}$$

is lower than that achieved by either cue alone.

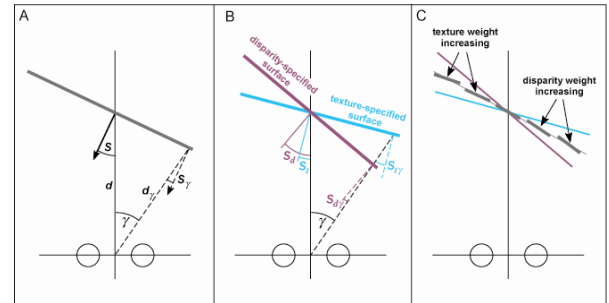
Improvement in JND with 2 Cues



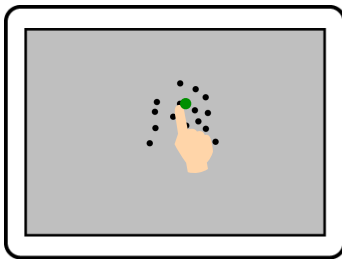
Conclusion

- The data are consistent with optimal cue combination
- Texture weight is increased with increasing distance and increasing base slant, as predicted
- Two cue JNDs are generally lower than the constituent single-cue JNDs
- Thus, weights are determined trial-by-trial, based on the current stimulus information and, in particular, the two single-cue slant estimates

Are Cue Weights Chosen Locally?



Experiment: Priors and Likelihoods



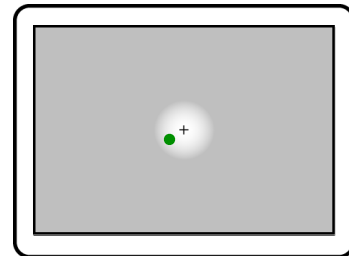
Training:

- Subject shown 'splash'
- Task is touch where coin landed
- Feedback provided
- 150 Trials

Tassinari, Hudson & Landy (2006). *J. Neurosci.*, 26, 10154-10163.

See also: Körding & Wolpert (2004). *Nature*, 427, 244-247.

Teaching Subjects the Prior



Tassinari, Hudson & Landy (2006). *J. Neurosci.*, 26, 10154-10163.

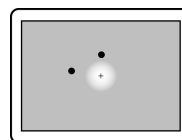
Experimental Trials



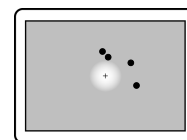
- 900 Trials
- Task is touch where coin landed
- Only hit/miss feedback is given
- Hits worth \$0.05

Tassinari, Hudson & Landy (2006). *J. Neurosci.*, 26, 10154-10163.

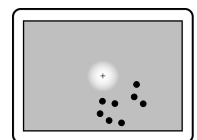
Experimental Stimuli



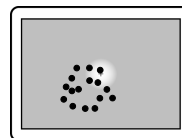
2 dots



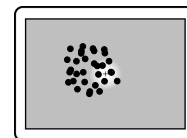
4 dots



8 dots



16 dots



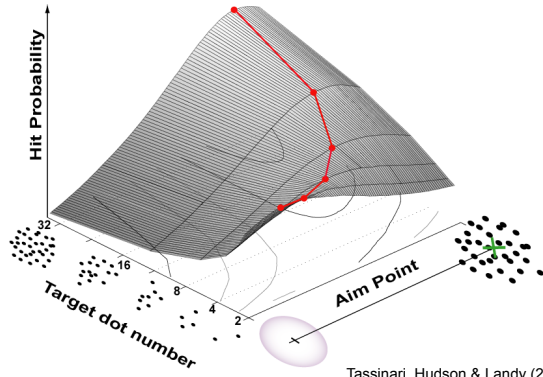
32 dots

Splash $\sigma = 4$ cm
Prior $\sigma = 2$ cm

Where should one aim to maximize number of hits?

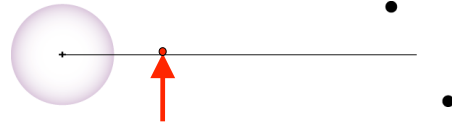
Tassinari, Hudson & Landy (2006). *J. Neurosci.*, 26, 10154-10163.

Gain Landscape



Tassinari, Hudson & Landy (2006).
J. Neurosci., 26, 10154-10163.

2 Dot condition:

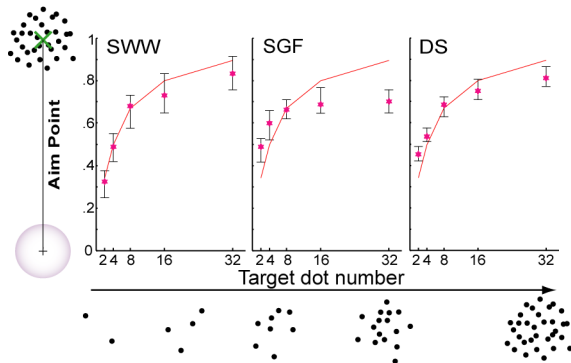


32 Dots:



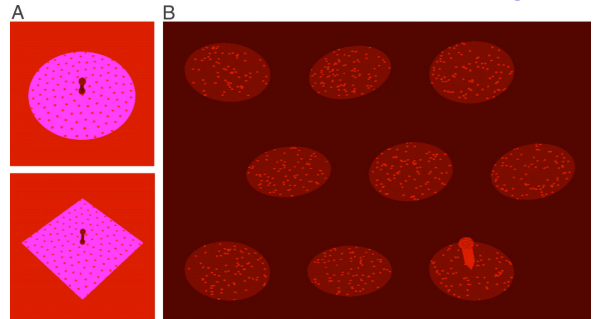
Tassinari, Hudson & Landy (2006).
J. Neurosci., 26, 10154-10163.

Results

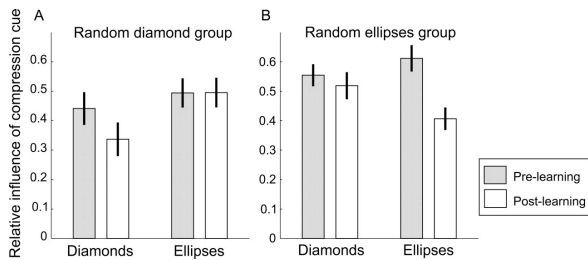


Tassinari, Hudson & Landy (2006).
J. Neurosci., 26, 10154-10163.

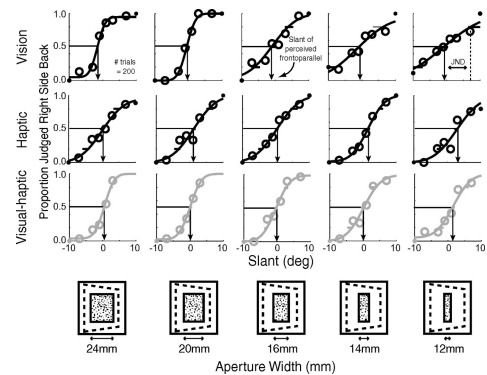
Learning a prior on stimulus distribution, hence reliability



Seydell et al. J Vis 2010



Cue-combination data from one representative subject.



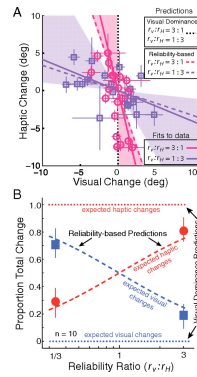
Burge J et al. J. Neurosci. 2010

A

age (deg)

Visual Dominance
 $f_V/f_R = 3:1$...
 $f_V/f_R = 1:3$

Reliability-based
 $f_V/f_R = 3:1$ - - -
 $f_V/f_R = 1:3$ = = =



Relevant Visual Error Irrelevant Visual Error

