

# Anisotropic Ocular Drift

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With lots of help from Janis Intoy

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# Overview

1. Brownian Motion as a model of Ocular Drift

2. Evidence of Optimal Drifts in Certain Tasks

3. Isotropic vs. Anisotropic Optimal Drifts

4. Results from tilted-grating detection and discrimination, Letter task



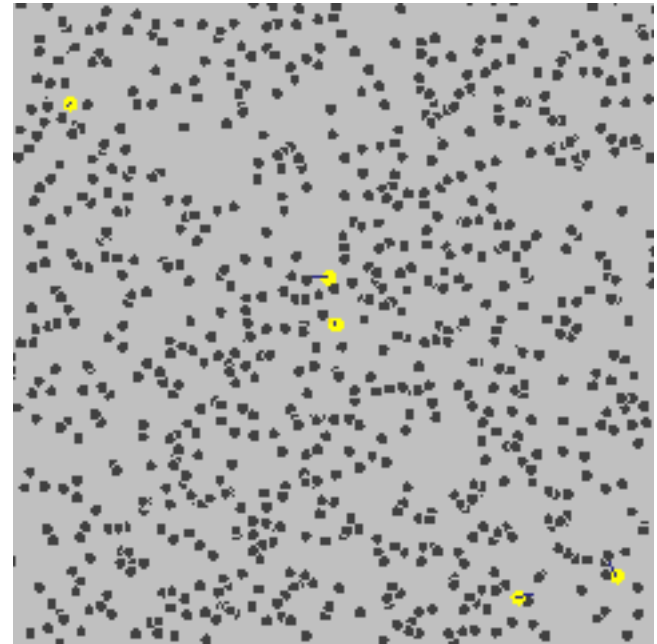
A method for  
quantifying  
and comparing  
the shape of drift

5. Next steps for investigating anisotropic drift

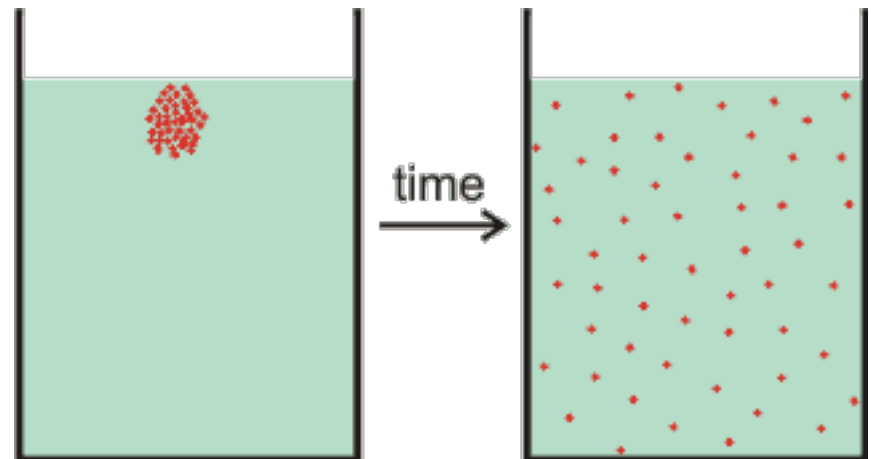
# Brownian Motion

- Brownian motion is the apparently random motion of small particles suspended in fluid.
- Brownian motion drives diffusion such that over time particles will move from regions of high concentration to low concentration.
- When diffusion is the same in every direction, i.e. *isotropic*, movement is characterized by a single diffusion coefficient ( $D$ ).

**Brownian Motion**



**Diffusion**



# Brownian Motion: Applied to Drift

Probability of gaze displacement over time



smaller



larger

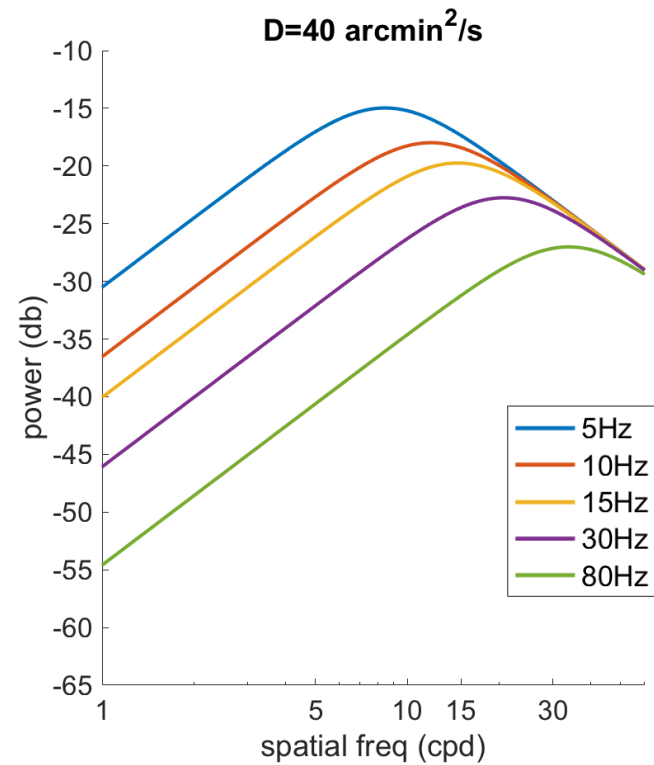
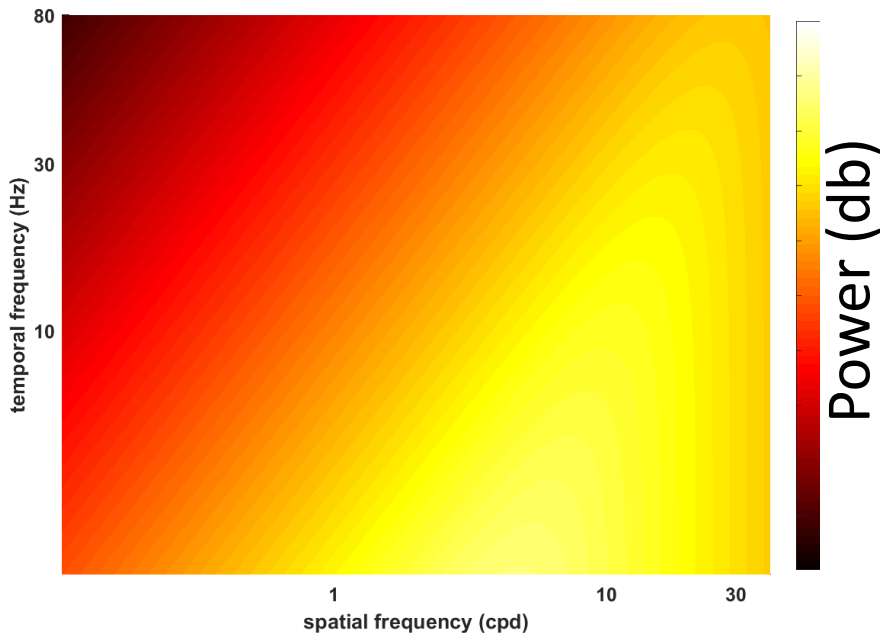
Diffusion Constant

- Parameterized by the diffusion constant,  $D$
- Gaze displacement at any given time follows the normal distribution
  - Variance increases **linearly** with time
  - $\langle r^2 \rangle = 4Dt$

# Brownian Motion: Applied to Drift

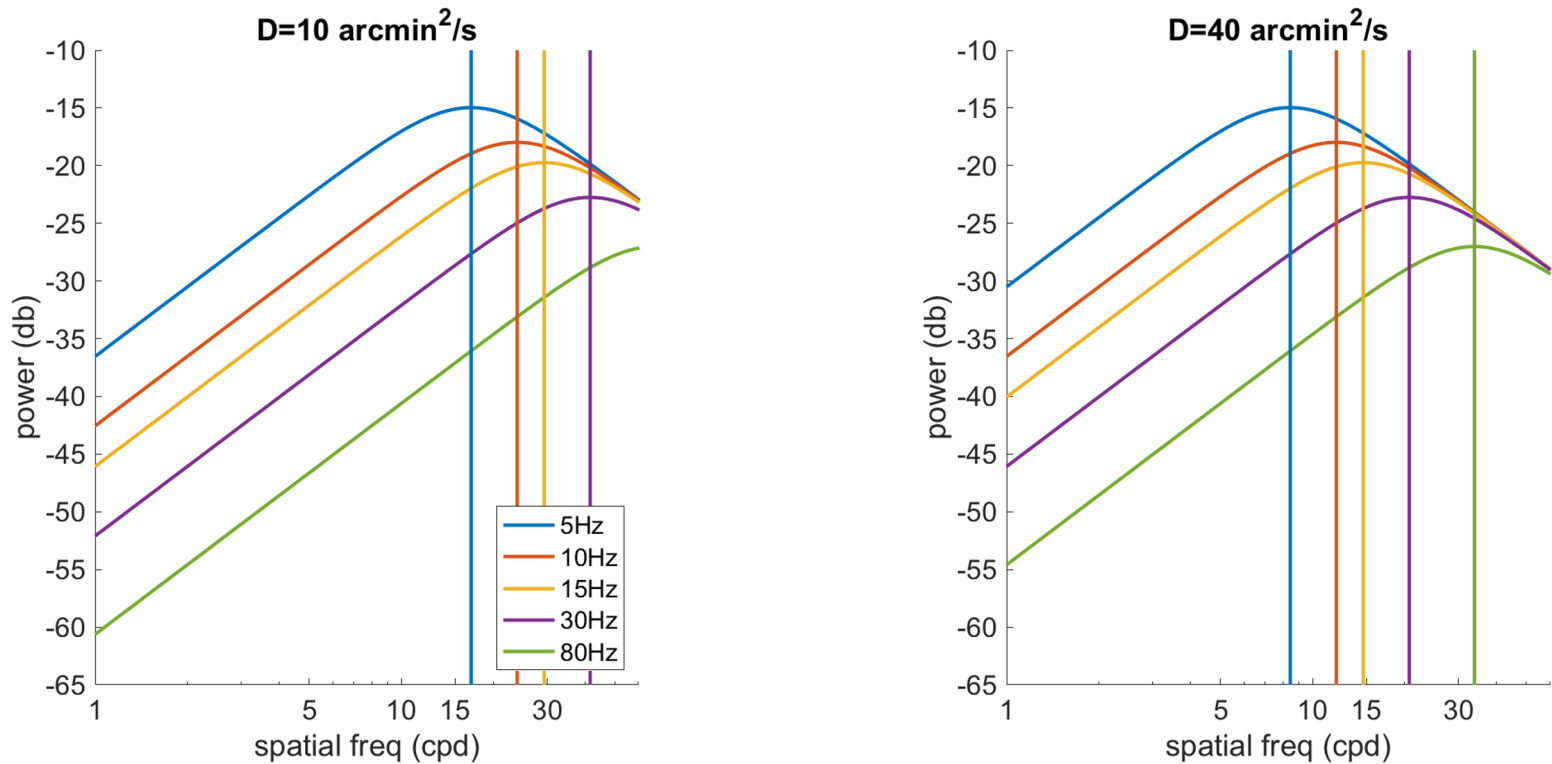
Power provided by drift

$$Q(k, f; D = 40)$$



# Brownian Motion: Applied to Drift

## Power provided by drift

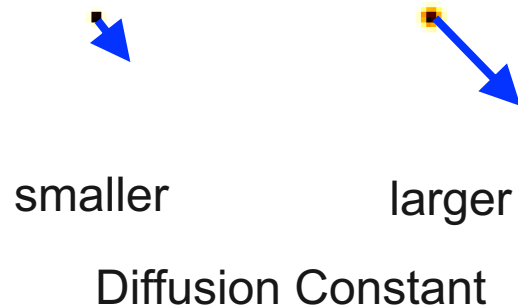


Smaller D --> Increases Critical Frequency

# “Optimal” Drifts

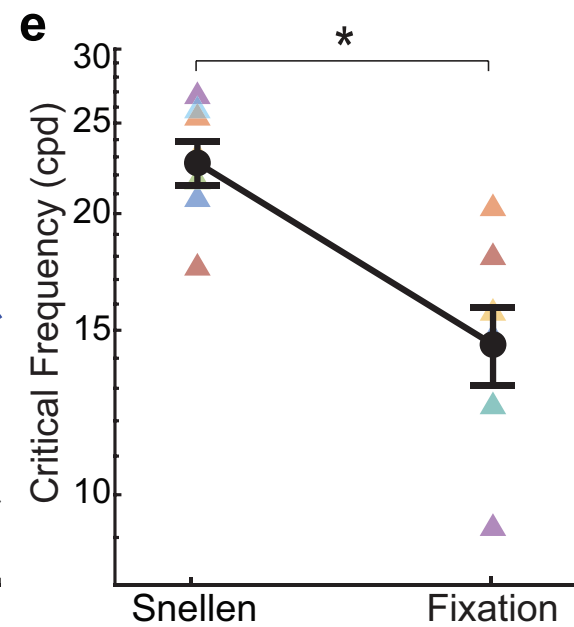
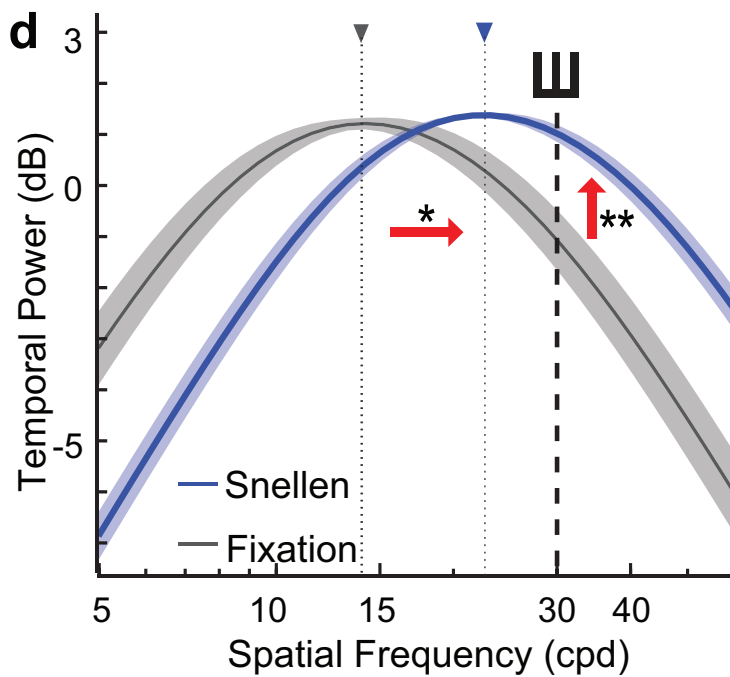
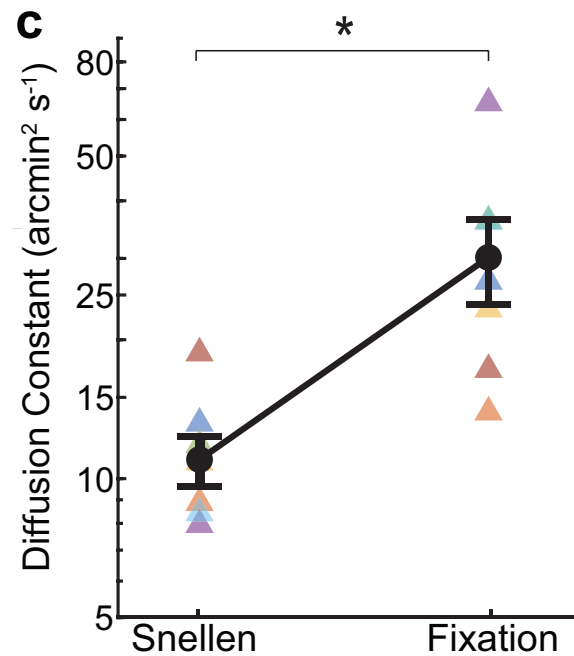
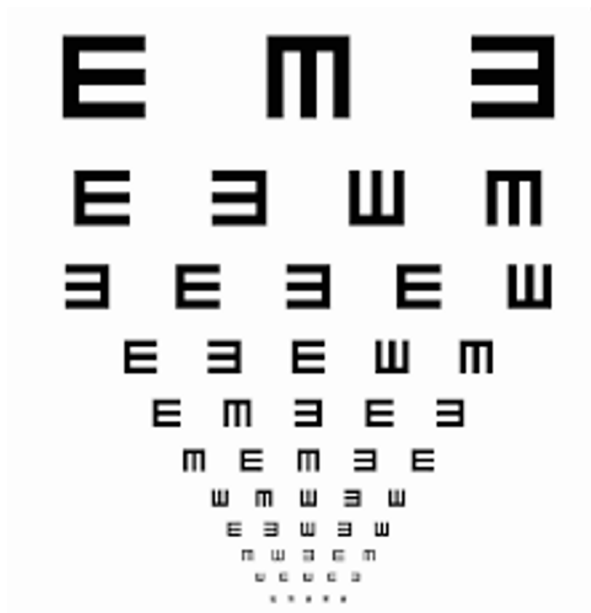
Smaller D --> Increases Critical Frequency

Probability of gaze displacement over time

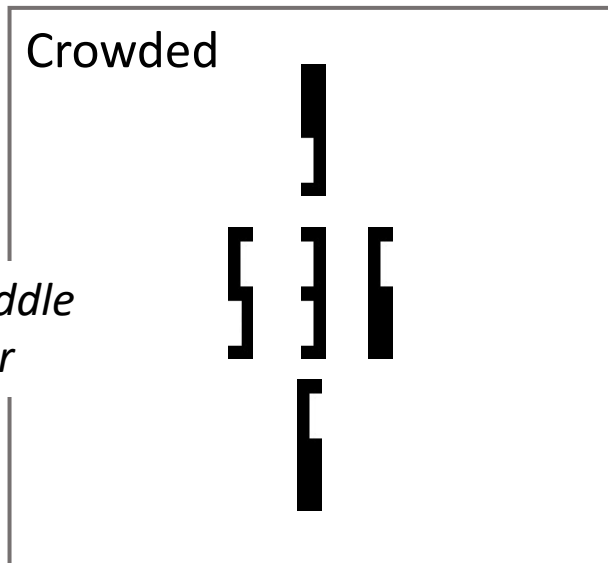
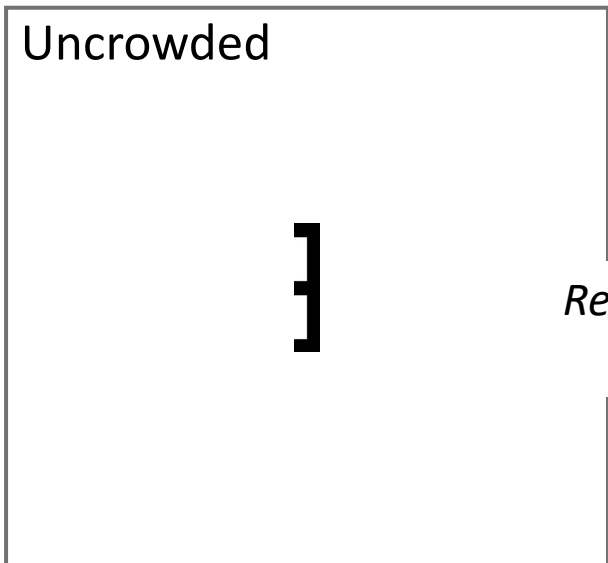


Three Pieces of Evidence:

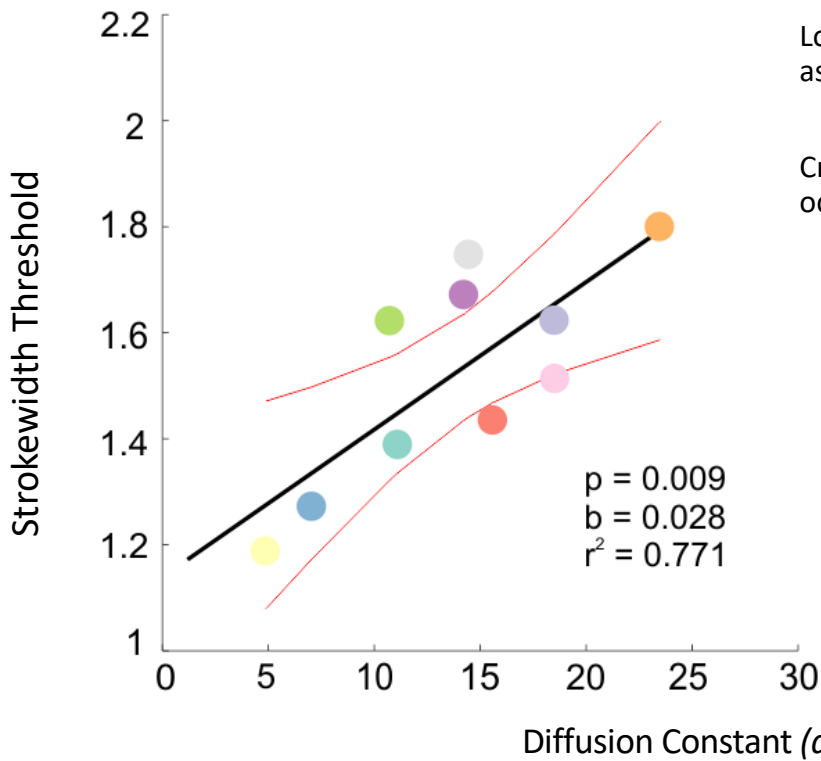
1. Snellen (Intoy & Rucci, 2020).
2. Crowding (Clark et al, VSS).
3. Drift Meta Analysis (Intoy et al., VSS)





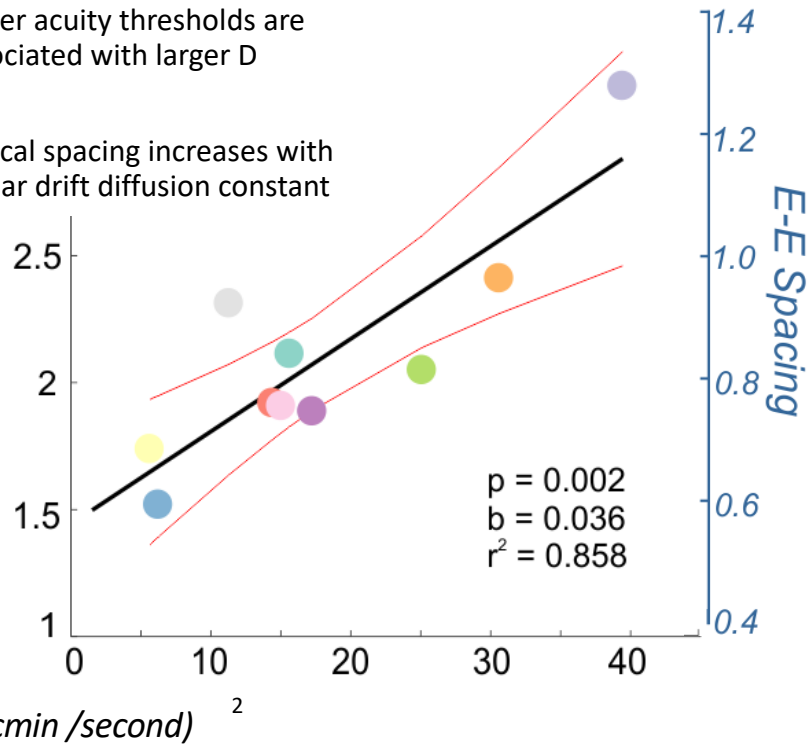


Report middle number

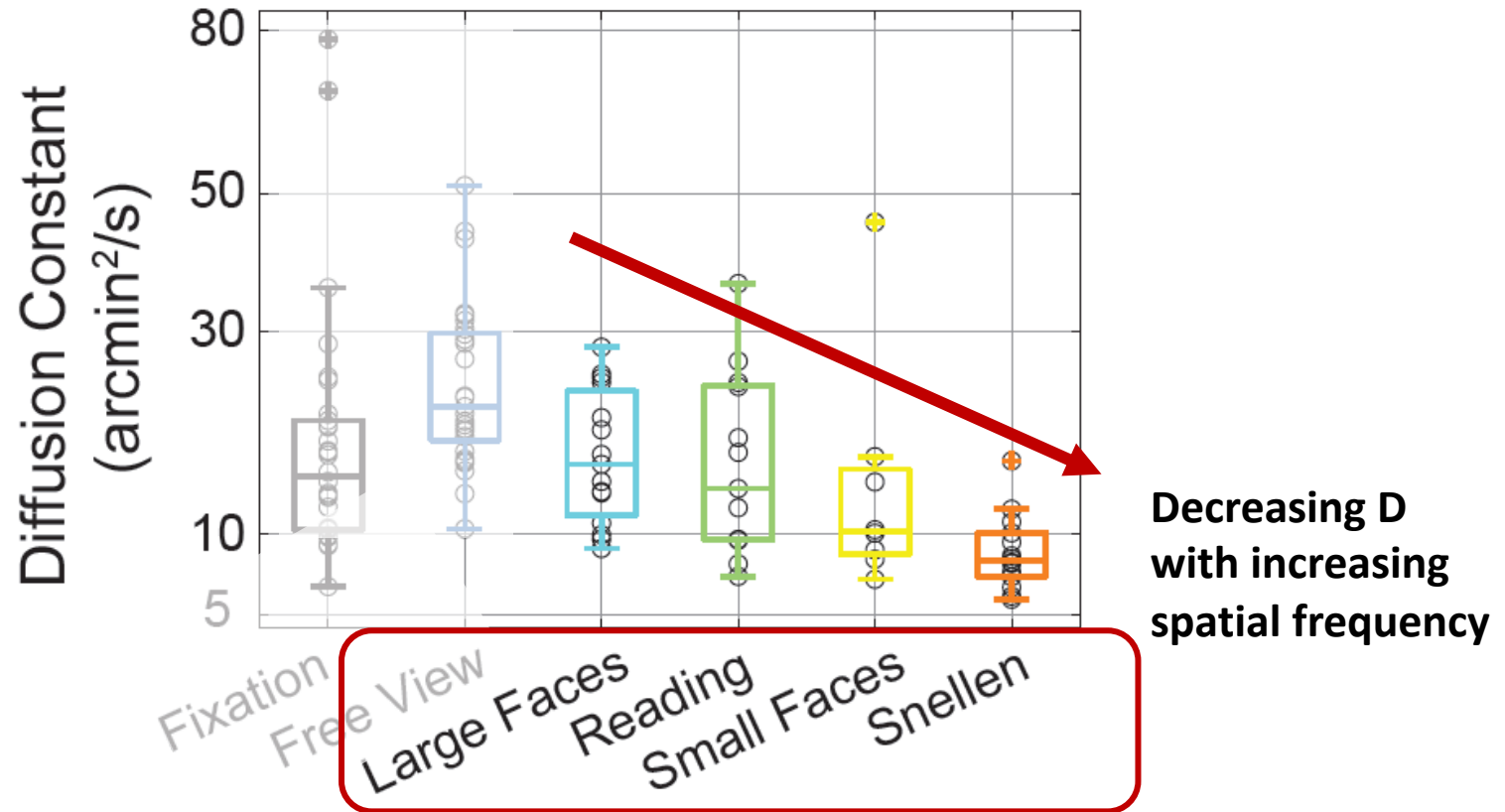


Lower acuity thresholds are associated with larger D

Critical spacing increases with ocular drift diffusion constant



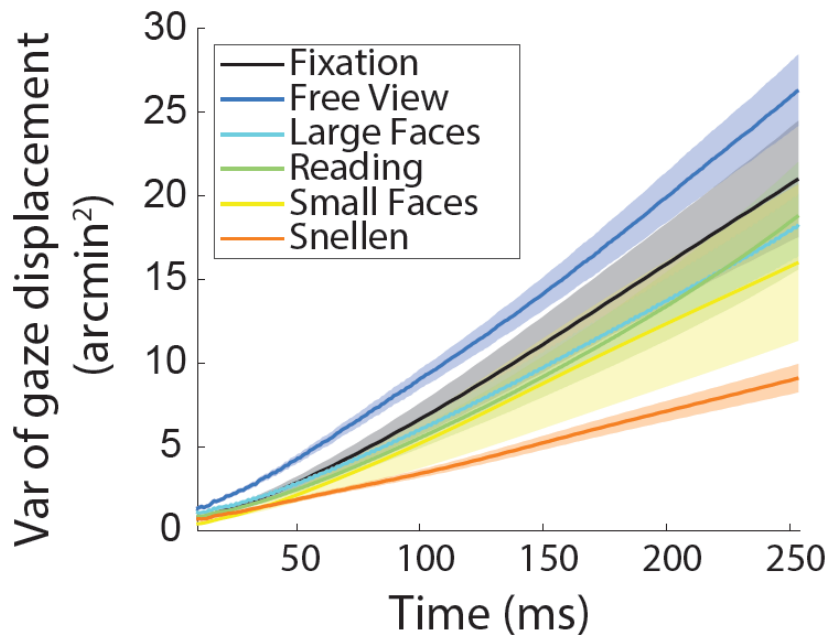
# Task-dependent Changes in Drift



The diffusion constant (which captures changes in speed and curvature) is one parameter by which drift may be controlled.

# Is Brownian motion a good model of drift?

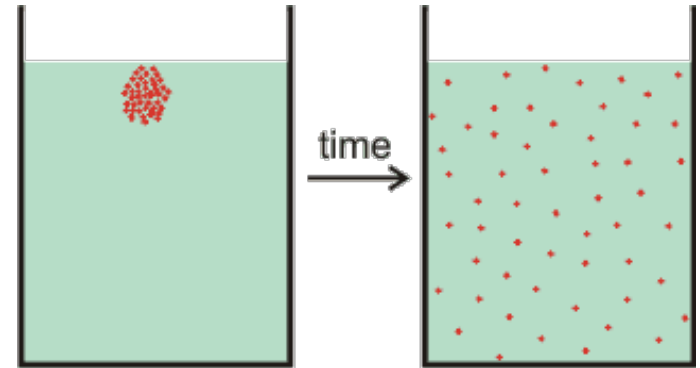
## Ocular Drift is Brownian



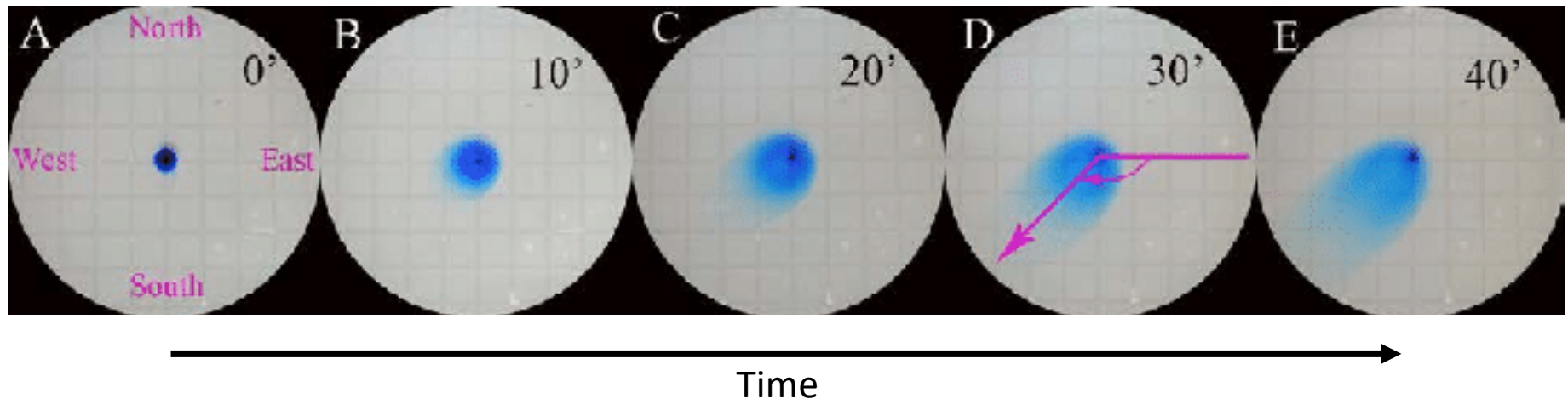
**In all tasks, the overall characteristics of ocular drift are compatible with a Brownian motion model.**

The variance of gaze displacement increases linearly with time, a signature of Brownian motion ( $R^2 > 0.95$ ). However, the diffusion constant, the slope of the increase, varies with task.

# Anisotropic Diffusion

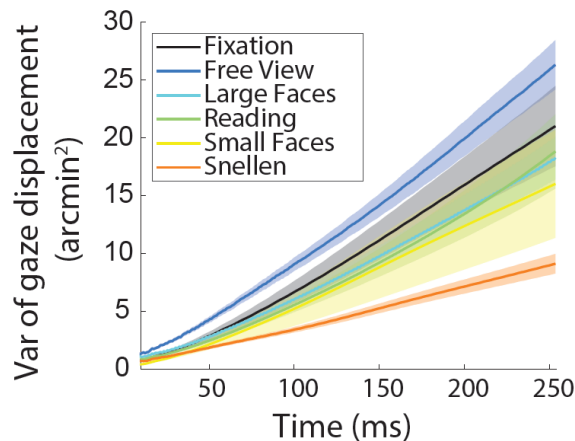


Long-term anisotropic (asymmetric) diffusion of toluidine blue solution in water.  
Angles indicate the maximum diffusion trend (MDT).



# Is Brownian motion a good model of drift?

- Investigate long-term correlations present in drifts.
- BM is uncorrelated in time.
- Fractional Brownian motion (fBM) has a parameter for temporal correlations (Hurst index,  $H$ )



Brownian Motion(BM):

$$\langle r^2 \rangle = 4Dt$$

Fractional Brownian Motion (fBM):

$$\langle r^2 \rangle = 4Dt^H$$

$H = 1 \rightarrow$  Brownian Motion

# Fractional Brownian Motion

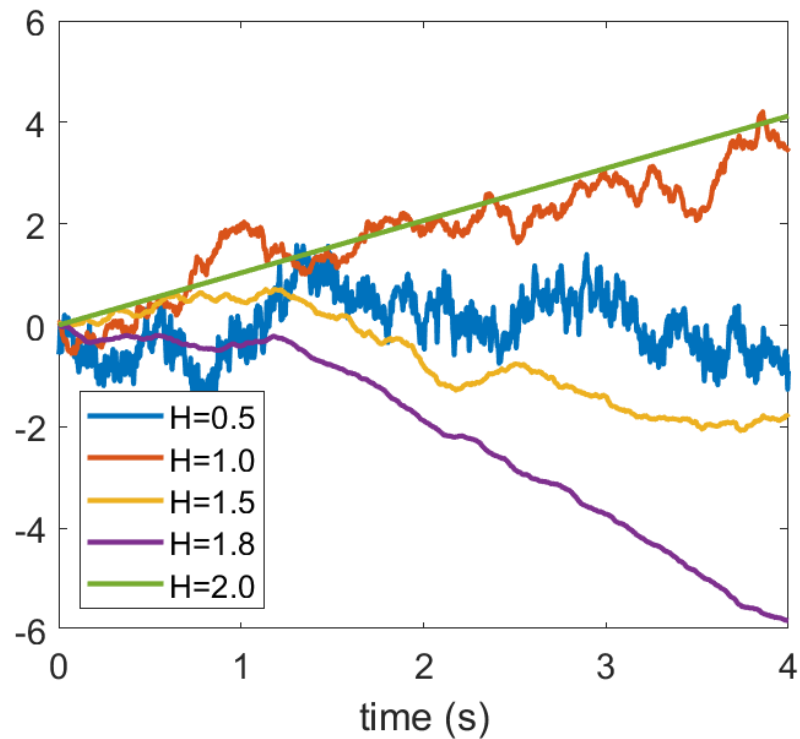
## Hurst Index (H)

$H > 1$  : correlated in time

$H = 1$ : uncorrelated in time (BM)

$H < 1$ : anticorrelated in time

## *Simulated fBM Drifts*



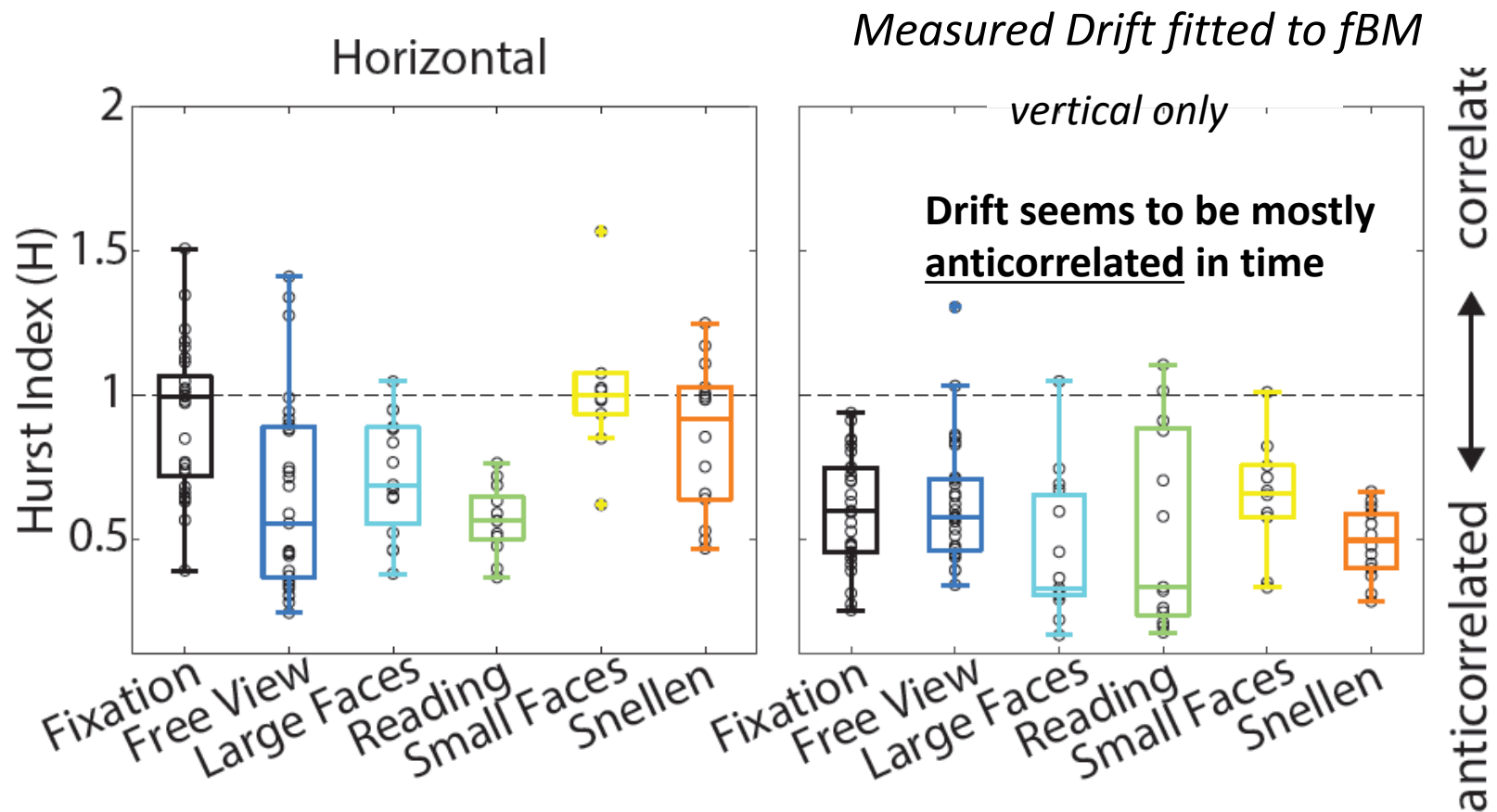
# Fractional Brownian Motion

## Hurst Index (H)

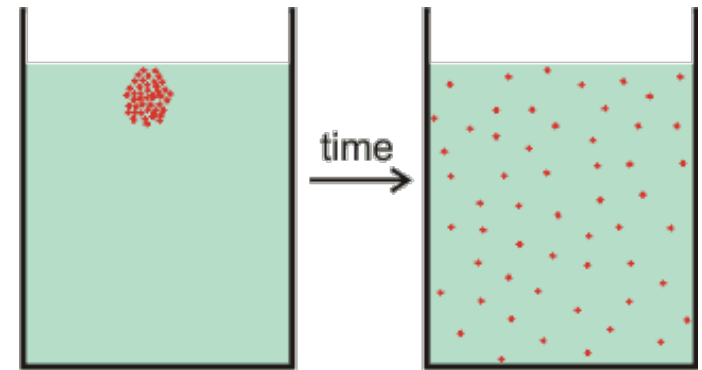
$H > 1$  : correlated in time

$H = 1$  : uncorrelated in time (BM)

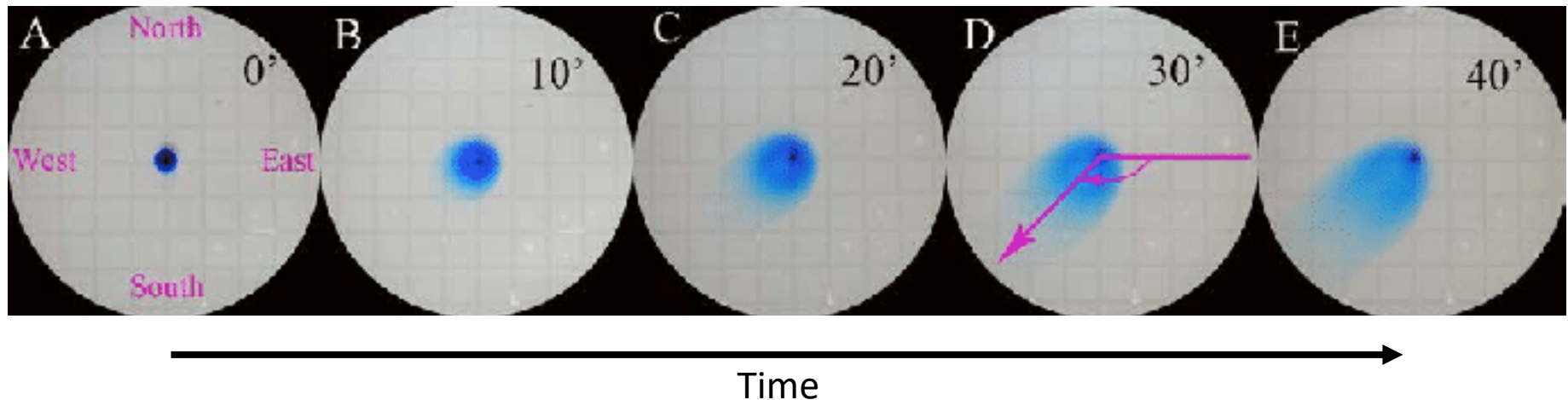
$H < 1$  : anticorrelated in time



# Anisotropic Diffusion



Long-term anisotropic (asymmetric) diffusion of toluidine blue solution in water.  
Angles indicate the maximum diffusion trend (MDT).





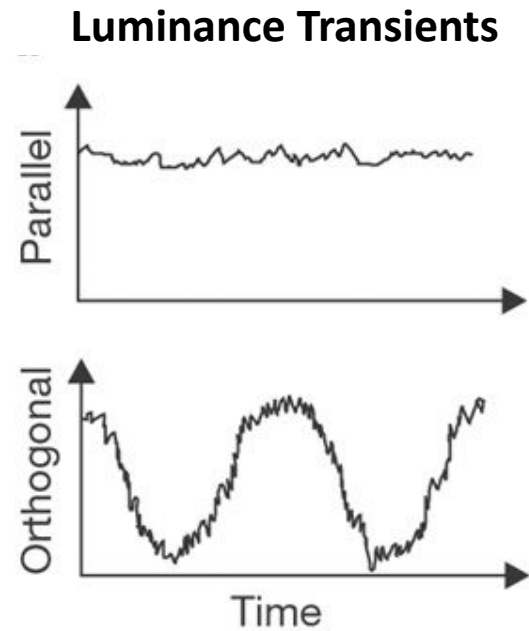
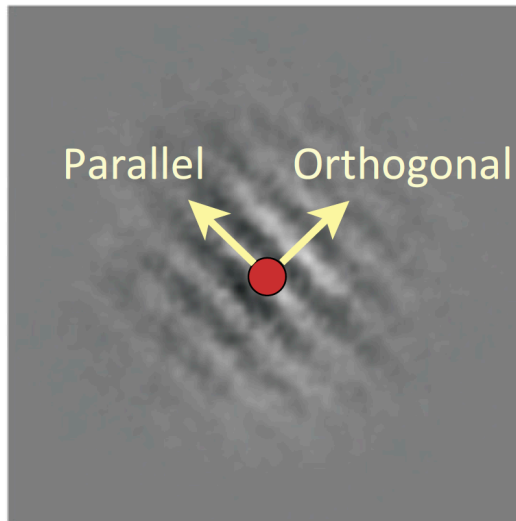
# Past and Current Empirical Investigations

*Switch Speaker*

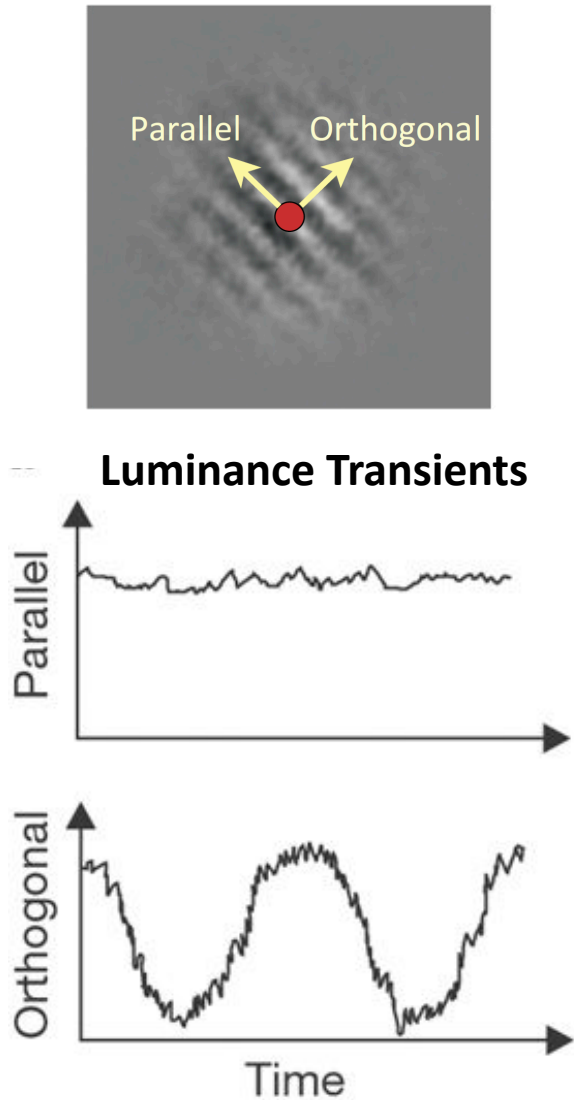
# Tasks to Test Stimulus-Dependent Anisotropic Drift

1. Grating detection task
2. Grating discrimination task
3. Letter discrimination task

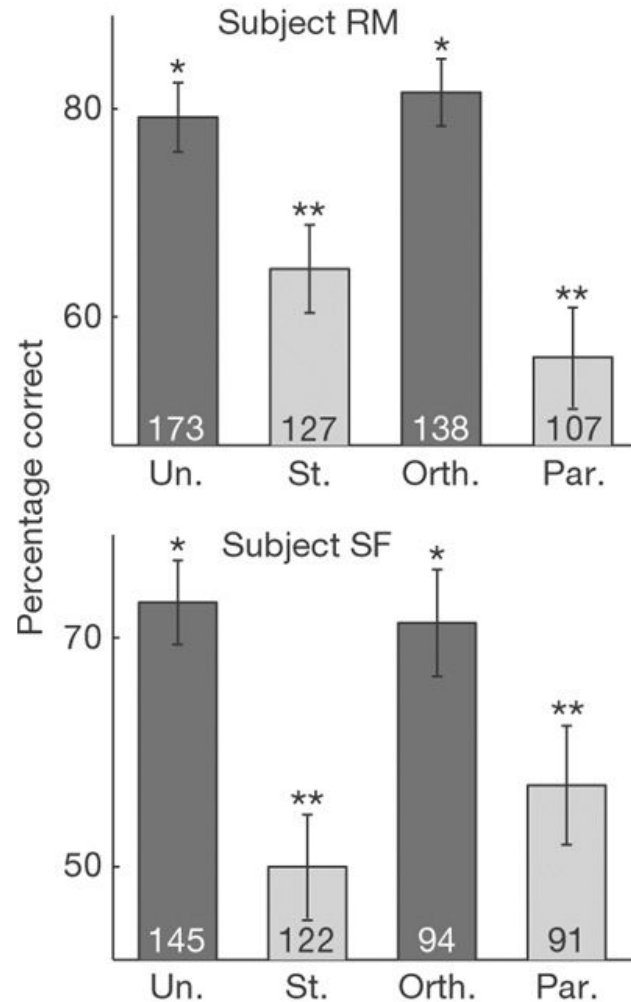
# Optimal Anisotropic Drift?



# Optimal Anisotropic Drift?



## Performance Under Partial Stabilization



# Expected Results

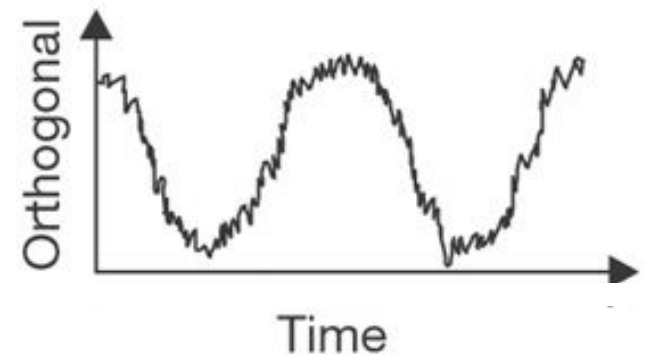
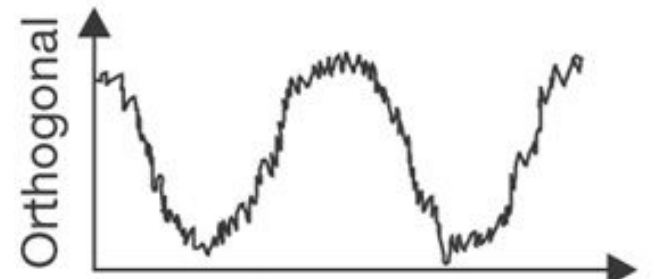
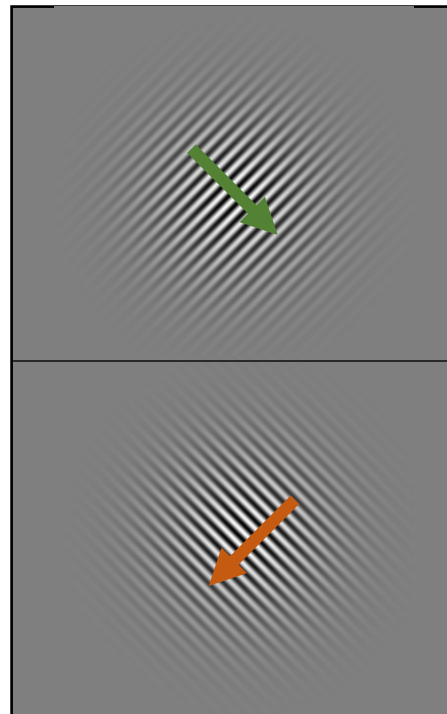
Optimal bias ocular drifts strategy  
– from the luminance change point of view

Active, controlled anisotropic FEM

FEM would change for each stimulus

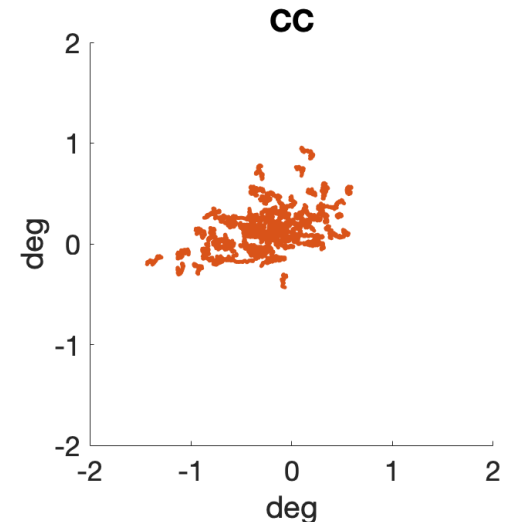
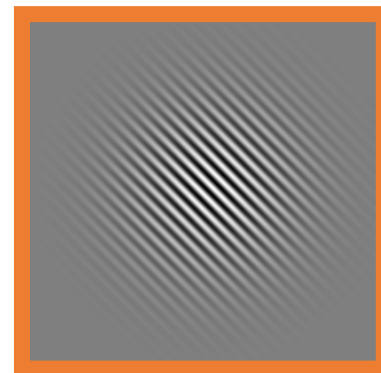
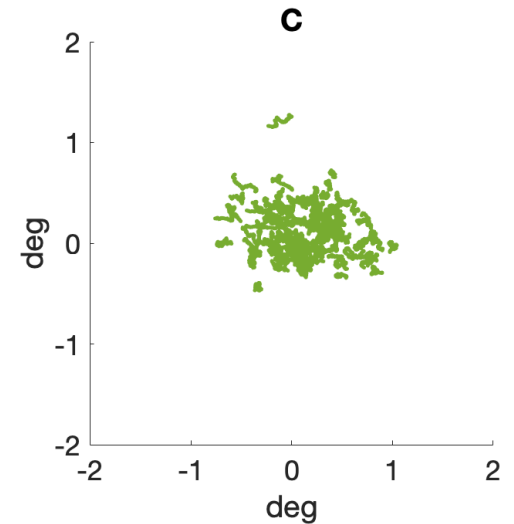
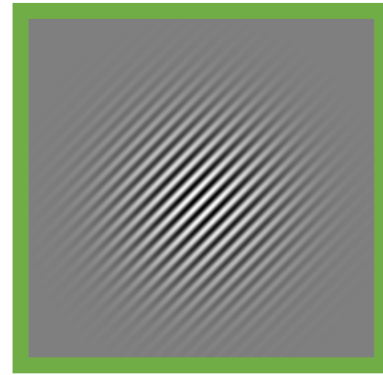
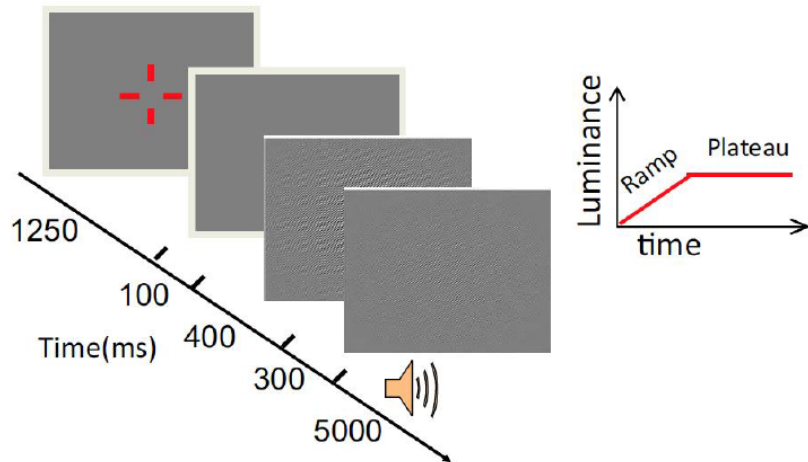
**STIMULUS-DRIVEN**

increase luminance  
power for all stimuli



# Stimulus-Dependent Anisotropic Drift?

## Tilted grating detection task



- 8 cpd, 11 deg in size
- Trials are block design; subjects knows the grating orientation
- **Do we see any evidence that the DIRECTION of drift changes with task?**

# Tilted Grating Detection: Previous and Repeated Analyses

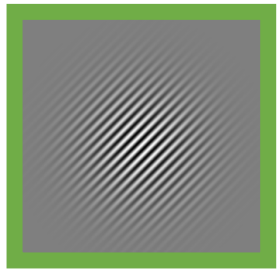
- Drift Metrics
  - Position
  - Velocity
  - Curvature
  - Speed
- Data Segmentation
  - Plateau vs. Drift
  - Fast vs. Slow Drifts
  - Performance
    - Overall detection (d-prime)
    - Hits, Miss, Correct Rejection, False Alarms

In all cases, no consistent differences in metrics between right-tilted and left-tilted sessions

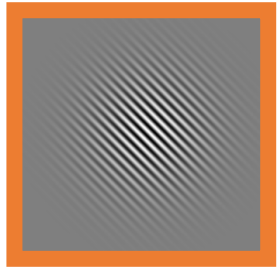
# No difference in Velocity Angles between Two Grating Viewing

## Tilted grating detection task

### Histogram of Velocity Angles

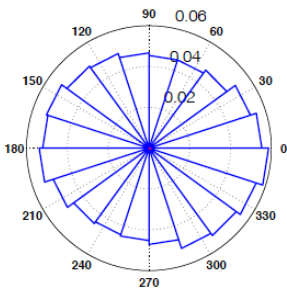


Left

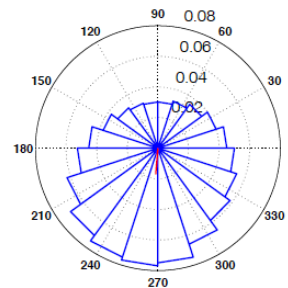


Right

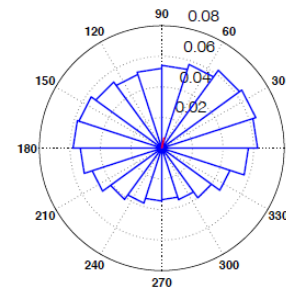
Andy



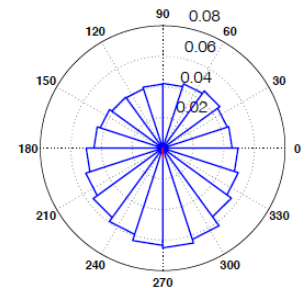
Ninjin



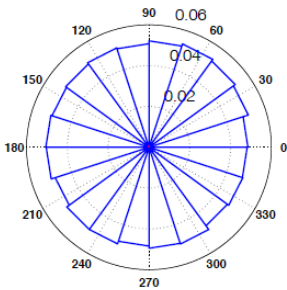
Sara



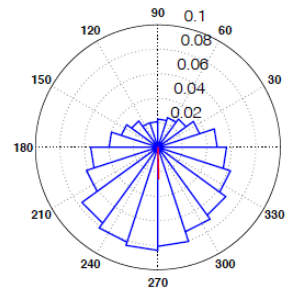
Tracy



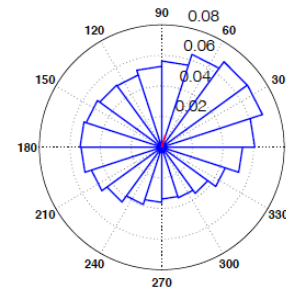
Andy



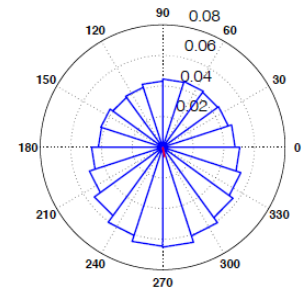
Ninjin



Sara

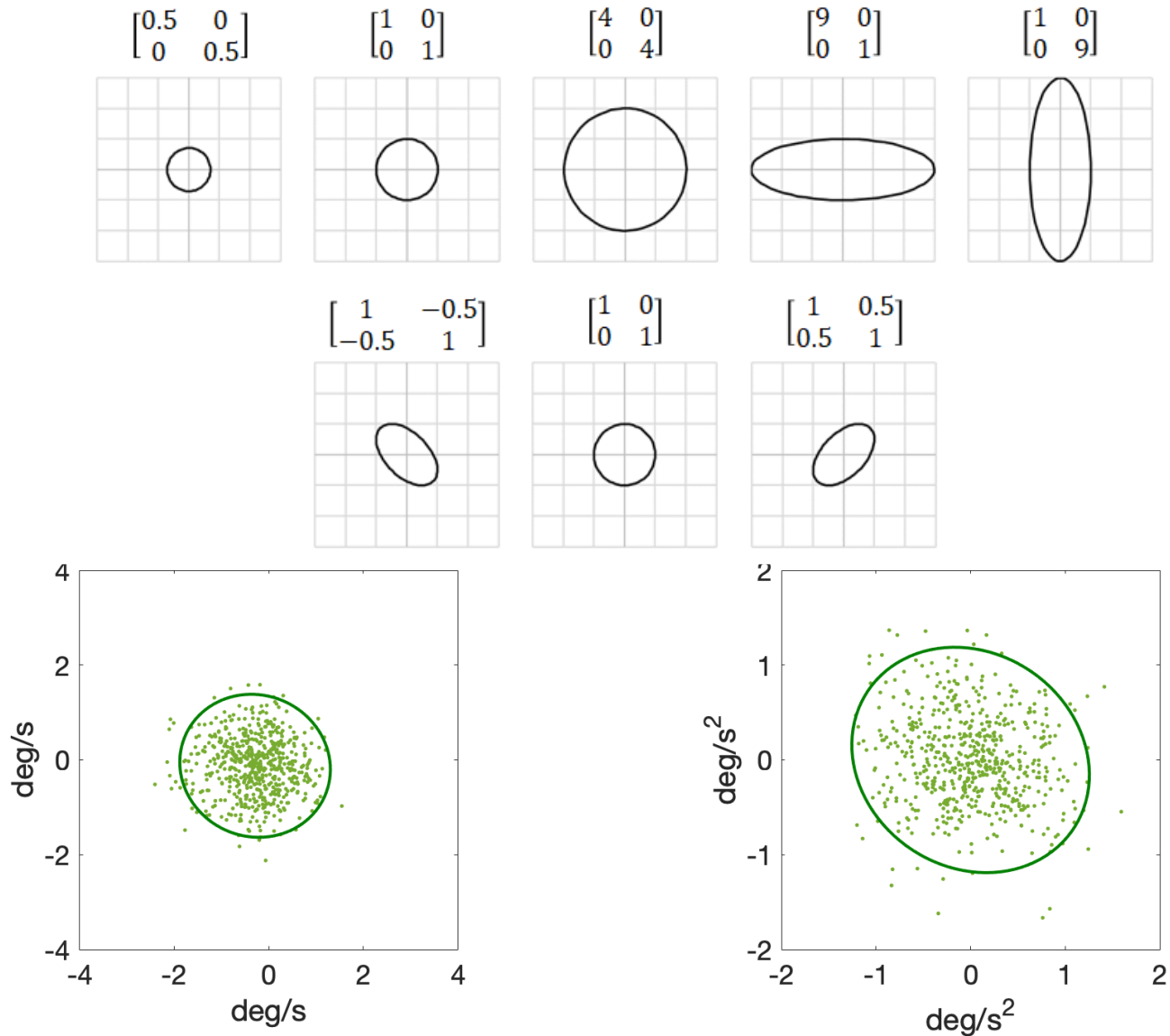


Tracy

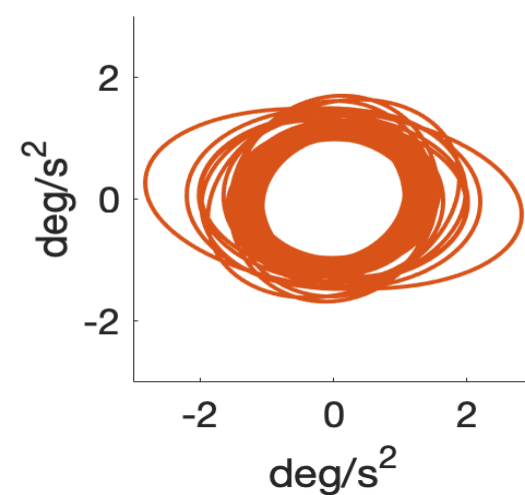
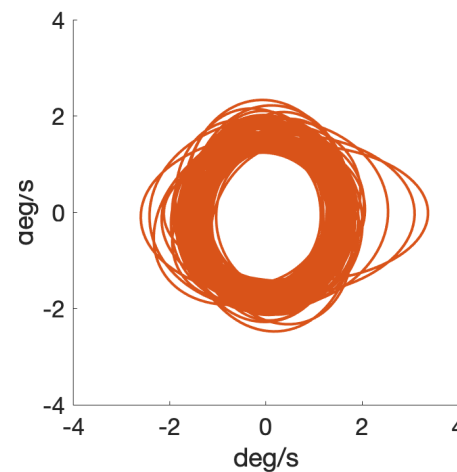
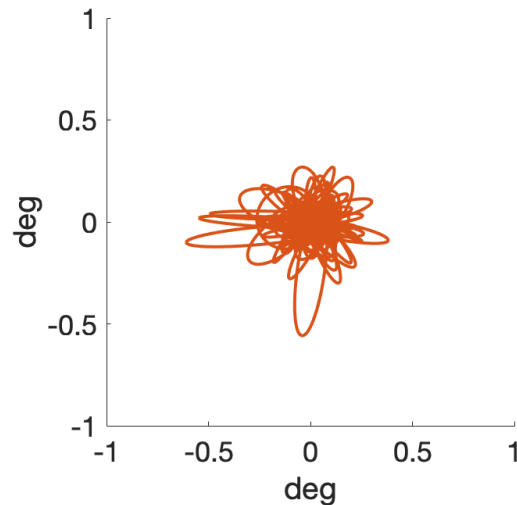
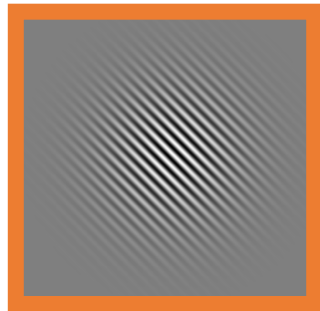
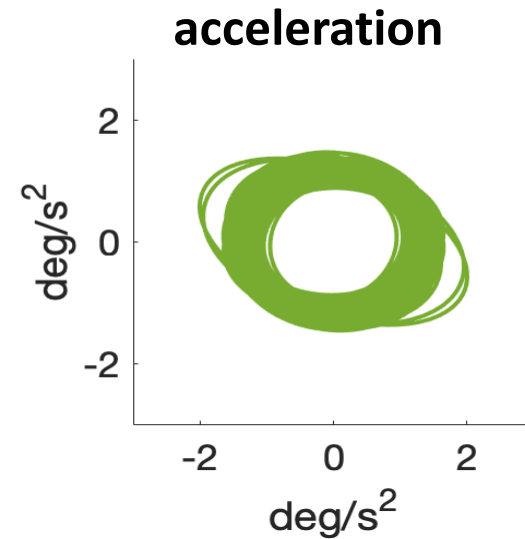
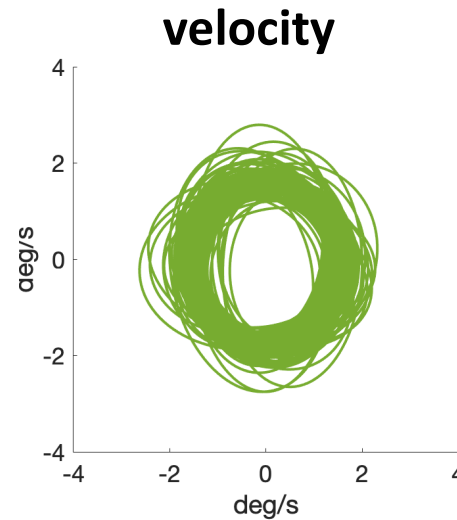
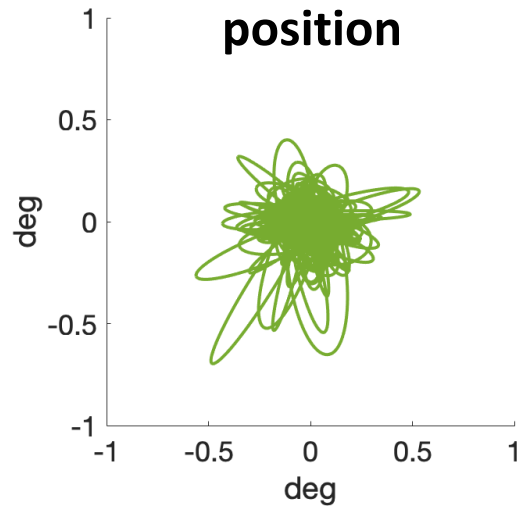
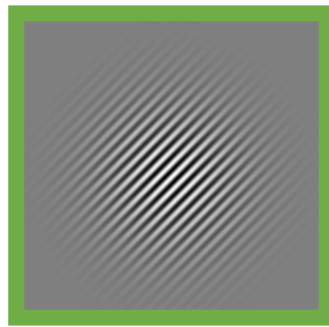




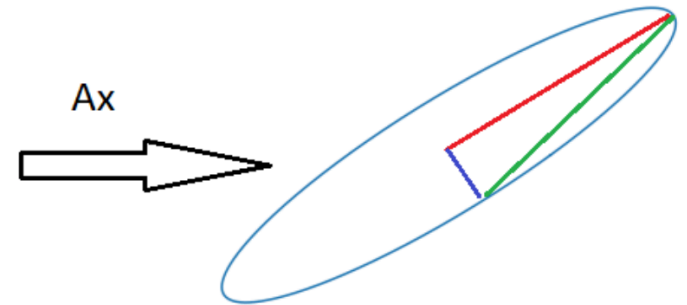
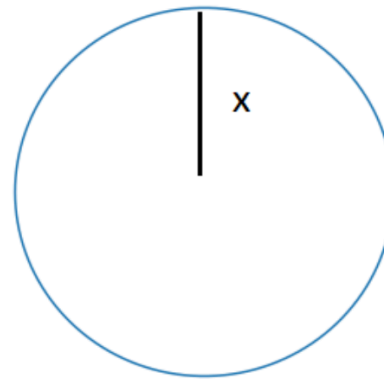
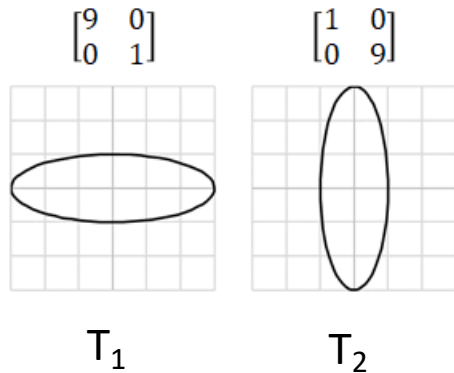
# Capturing the 2D Features of Drifts



# Capture the Features of Drifts on A Trial by Trial Basis



# Distance between Two Covariance Matrices

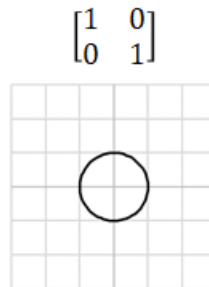


Eigenvalues:  $s_1$   
 $s_2$

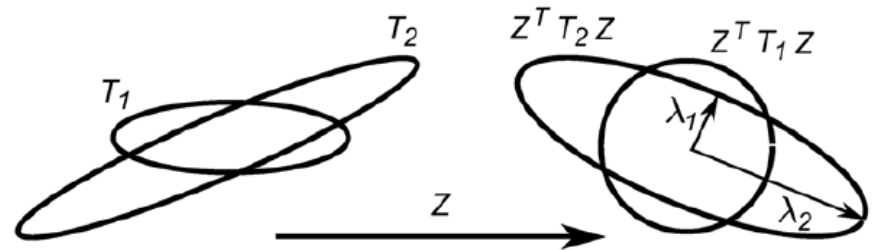
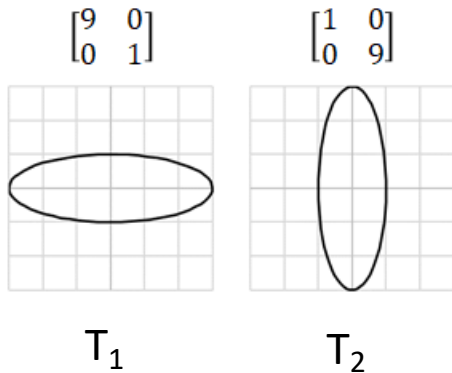
Frobenius norm:  $\sqrt{s_1^2 + s_2^2}$

$$Q = T_1^{-1} T_2$$

We want to know how close  $Q$  is to identity



# Distance between Two Covariance Matrices



**Figure 1. Geometric interpretation of the distance between two tensors.** The two structure tensors  $T_1$  and  $T_2$  are represented by the elliptic blobs shown in the lefthand side of the figure. After the change of coordinates defined by the matrix  $Z$ ,  $T_1$  is represented by the unit disk and the principal axes of  $T_2$  are equal to the eigenvalues  $\lambda_1$  and  $\lambda_2$  that appear in (6), see text.  
doi:10.1371/journal.pcbi.1000625.g001

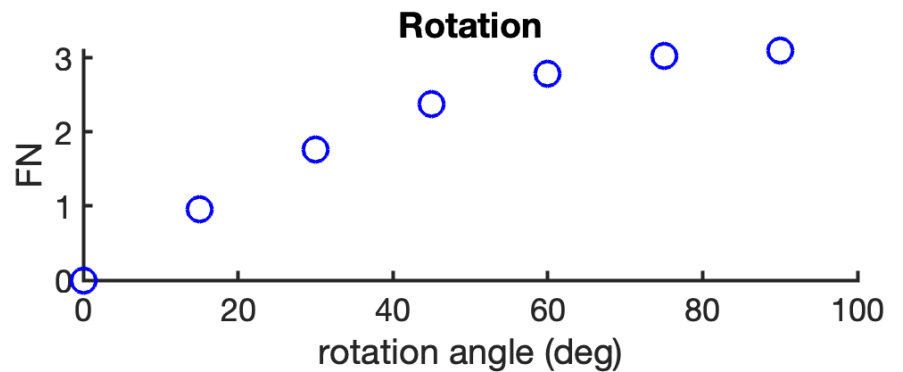
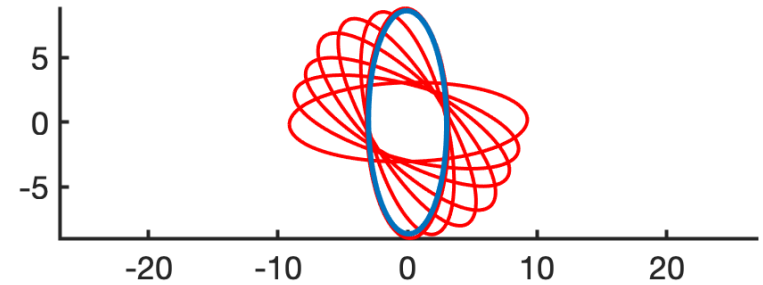
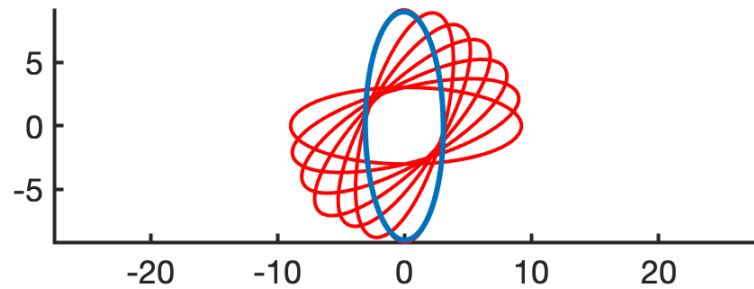
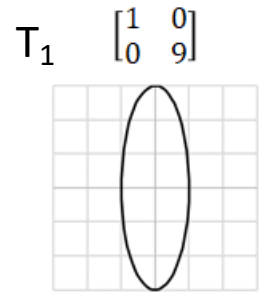
$$Q = T_1^{-1} T_2$$

We want to know how close  $Q$  is to identity

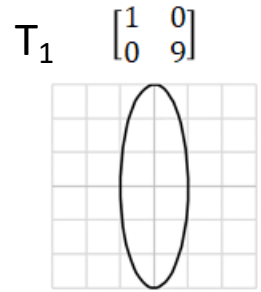
**the Frobenius norm of the matrix-log of  $Q$**

$$d_0(T_1, T_2) = \left\| \log T_1^{-1} T_2 \right\|_F = \sqrt{\log^2(\lambda_1) + \log^2(\lambda_2)}$$

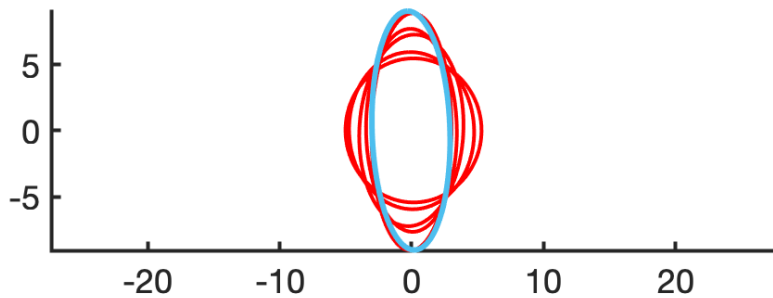
# Rotation Changes the Standard Distance between $T_1$ and $T_2$



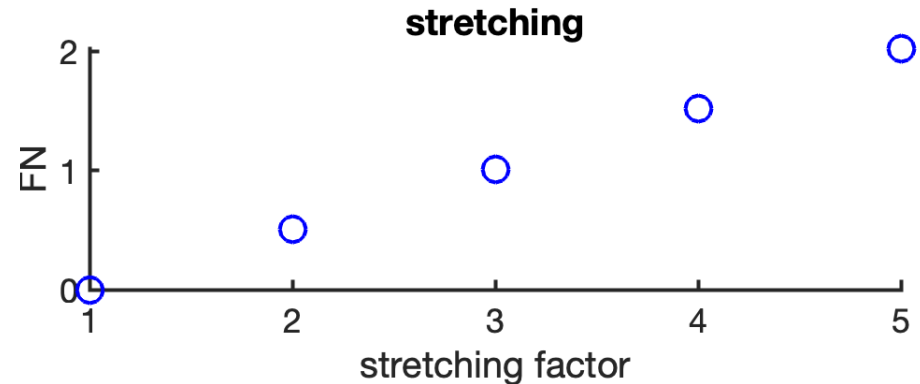
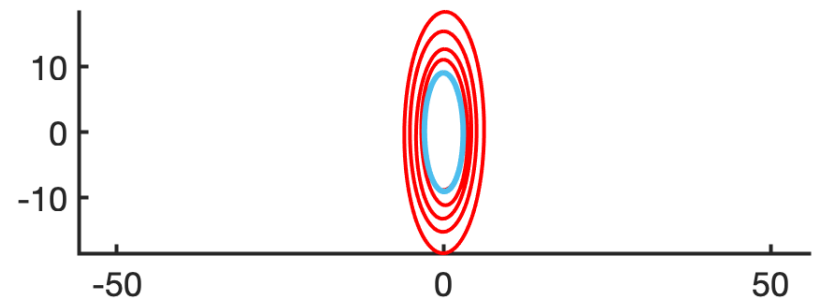
# Stretching Change the Standard Distance between $T_1$ and $T_2$



Without changing the size

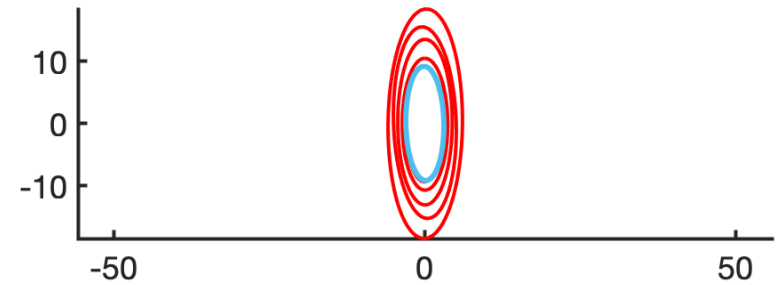
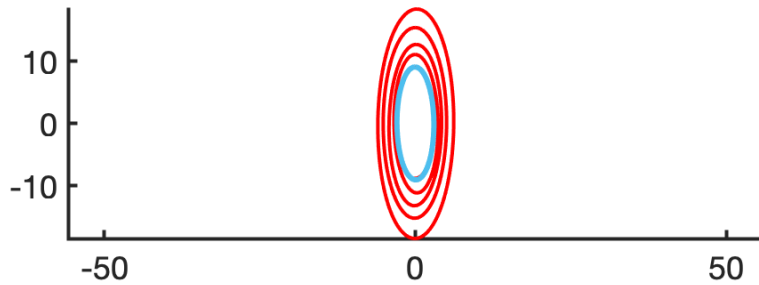


With changing the size

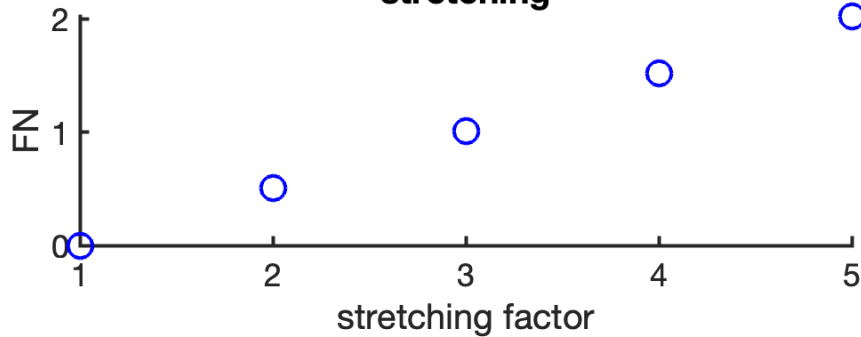


# Change in Size Can Be Normalized by A Single Factor

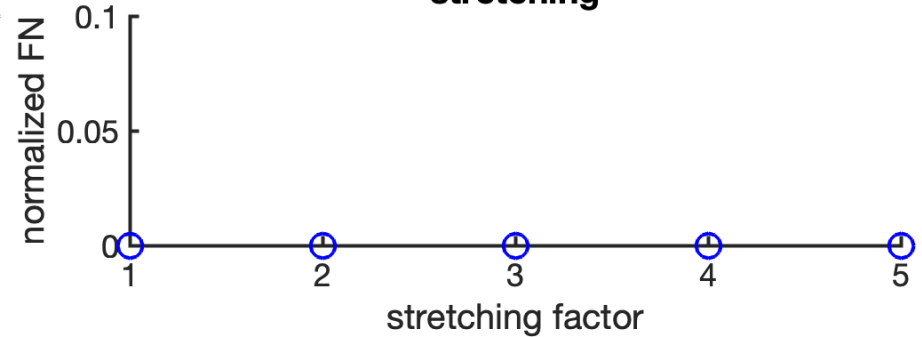
Normalized by the size



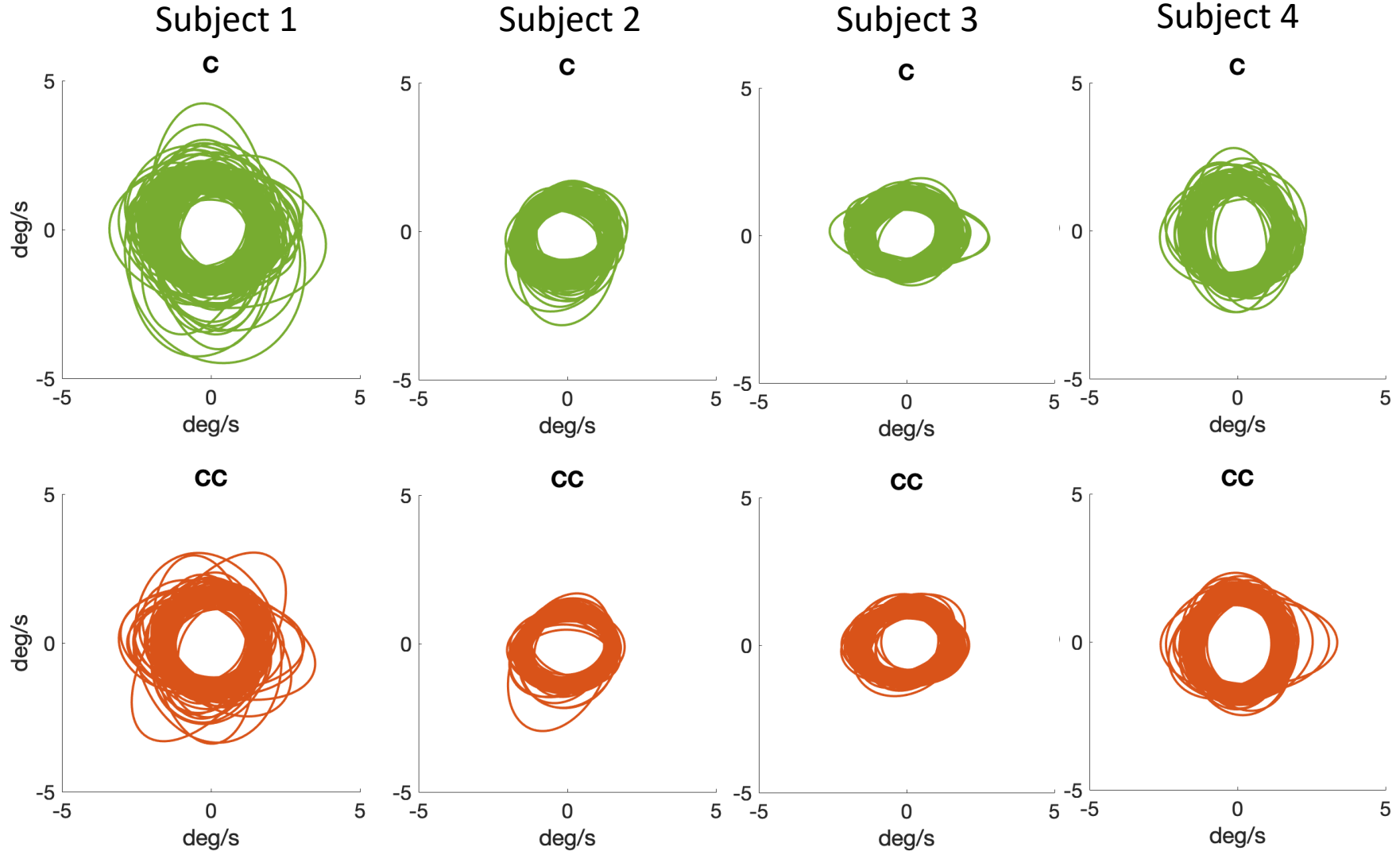
stretching



stretching



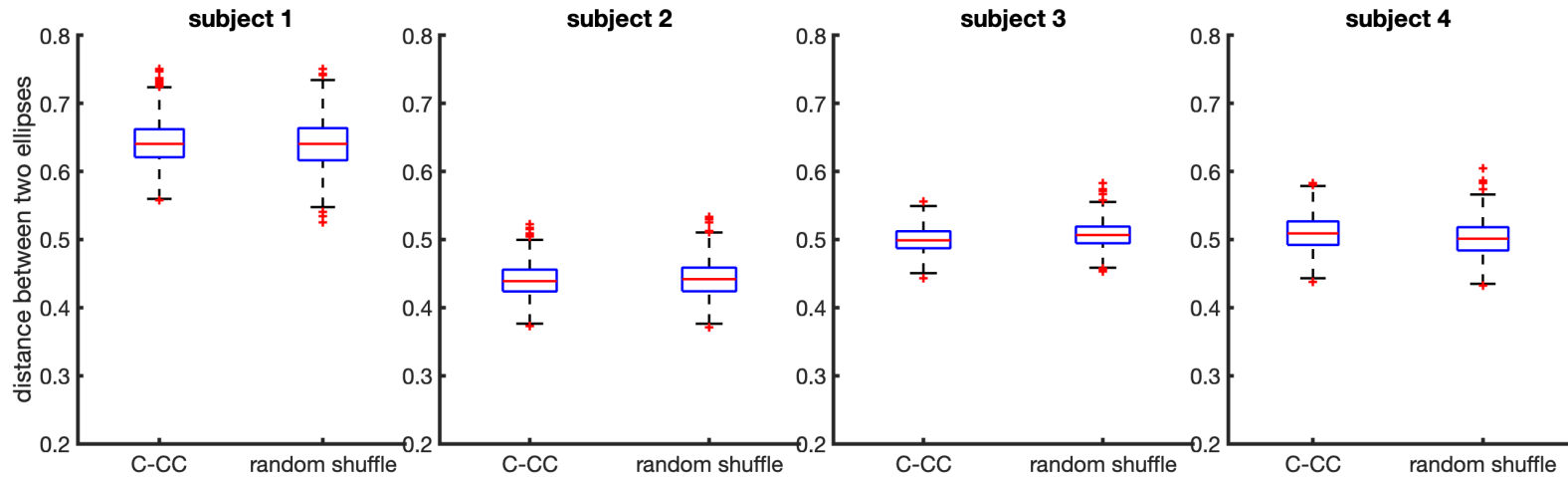
# Difference in Directional Bias between Two Grating Viewing?



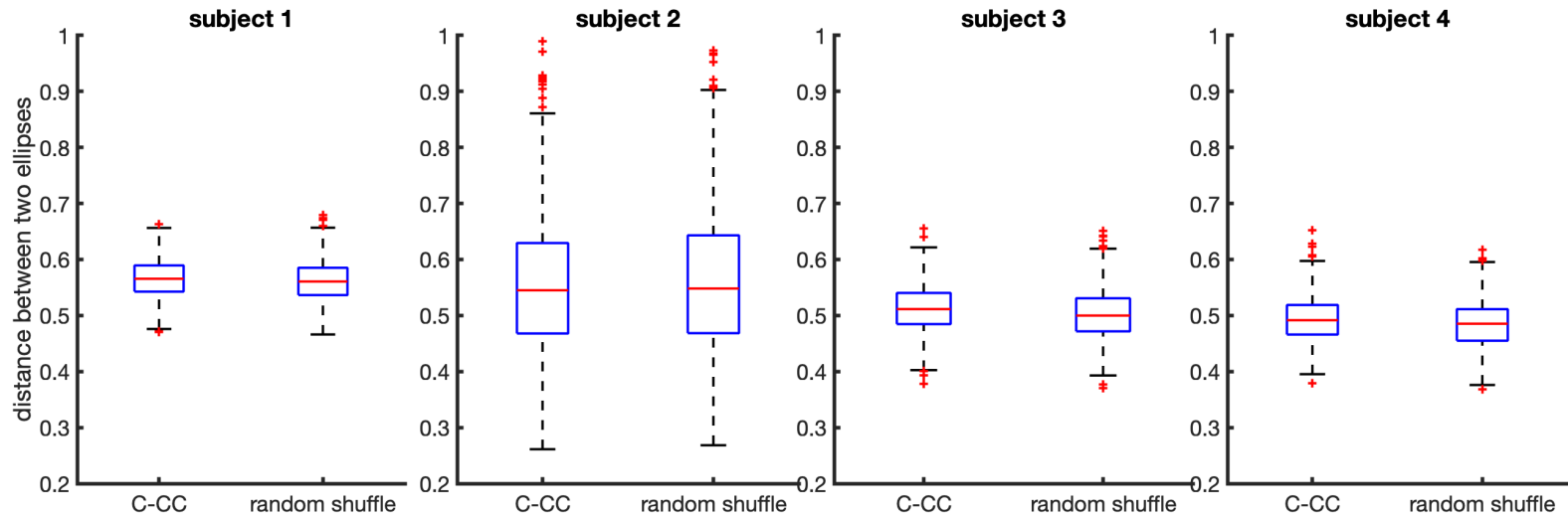


# No Difference in Directional Bias between Two Grating Viewing

Hit

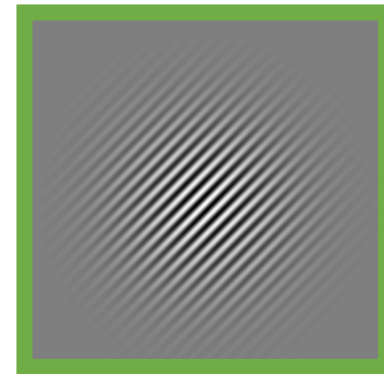
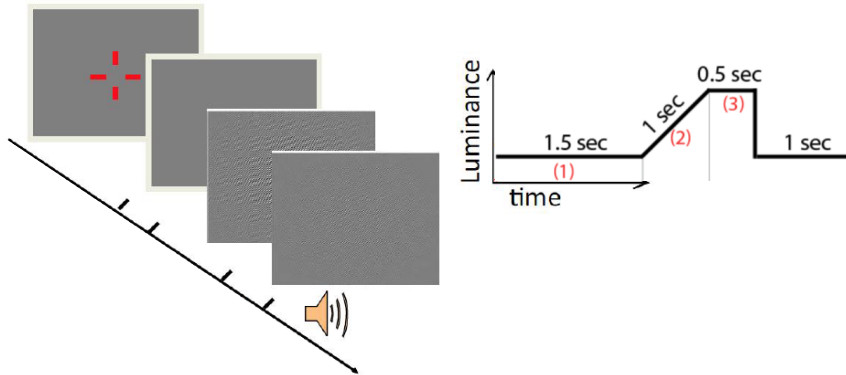


Miss



# Tasks to Test Stimulus-Dependent Anisotropic Drift

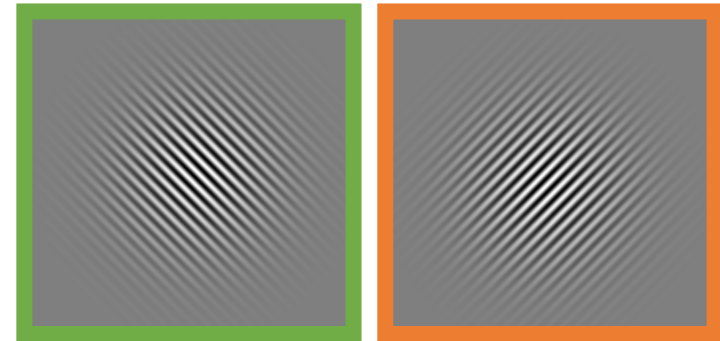
## Tilted grating discrimination task



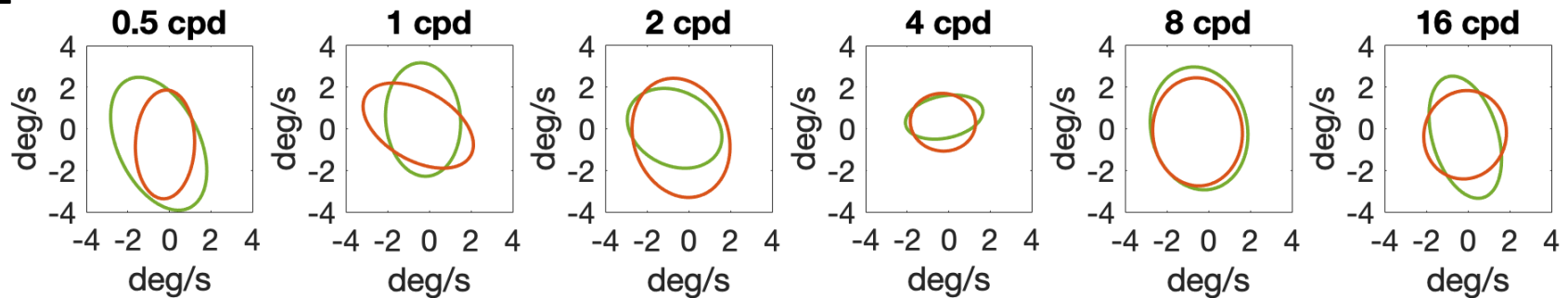
- 0.5 - 16 cpd, 4 deg in size
- Trials are block design; subjects know the spatial frequency of the grating
- **Do we see any evidence that the DIRECTION of drift changes with task?**

# Overall Differences between Two Tilted Grating Viewing

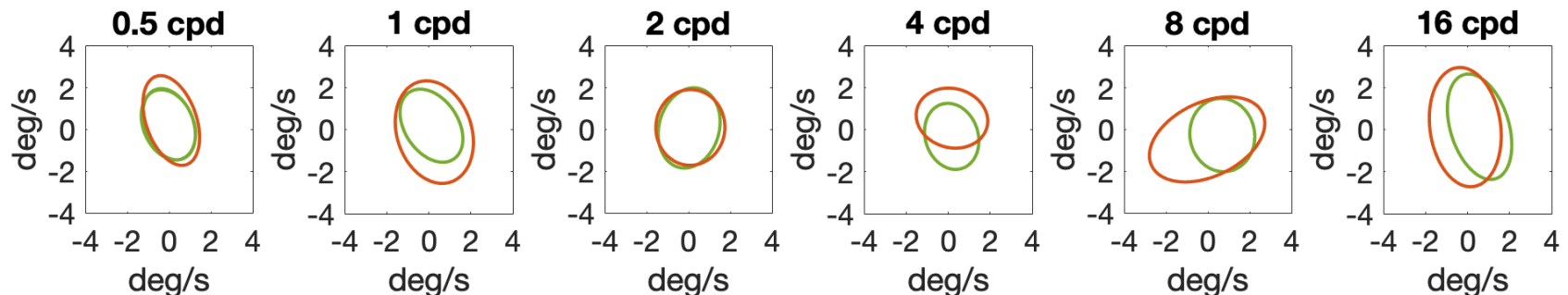
ALL DATA FITTED WITH 1 ELLIPSE  
not trial-by-trial



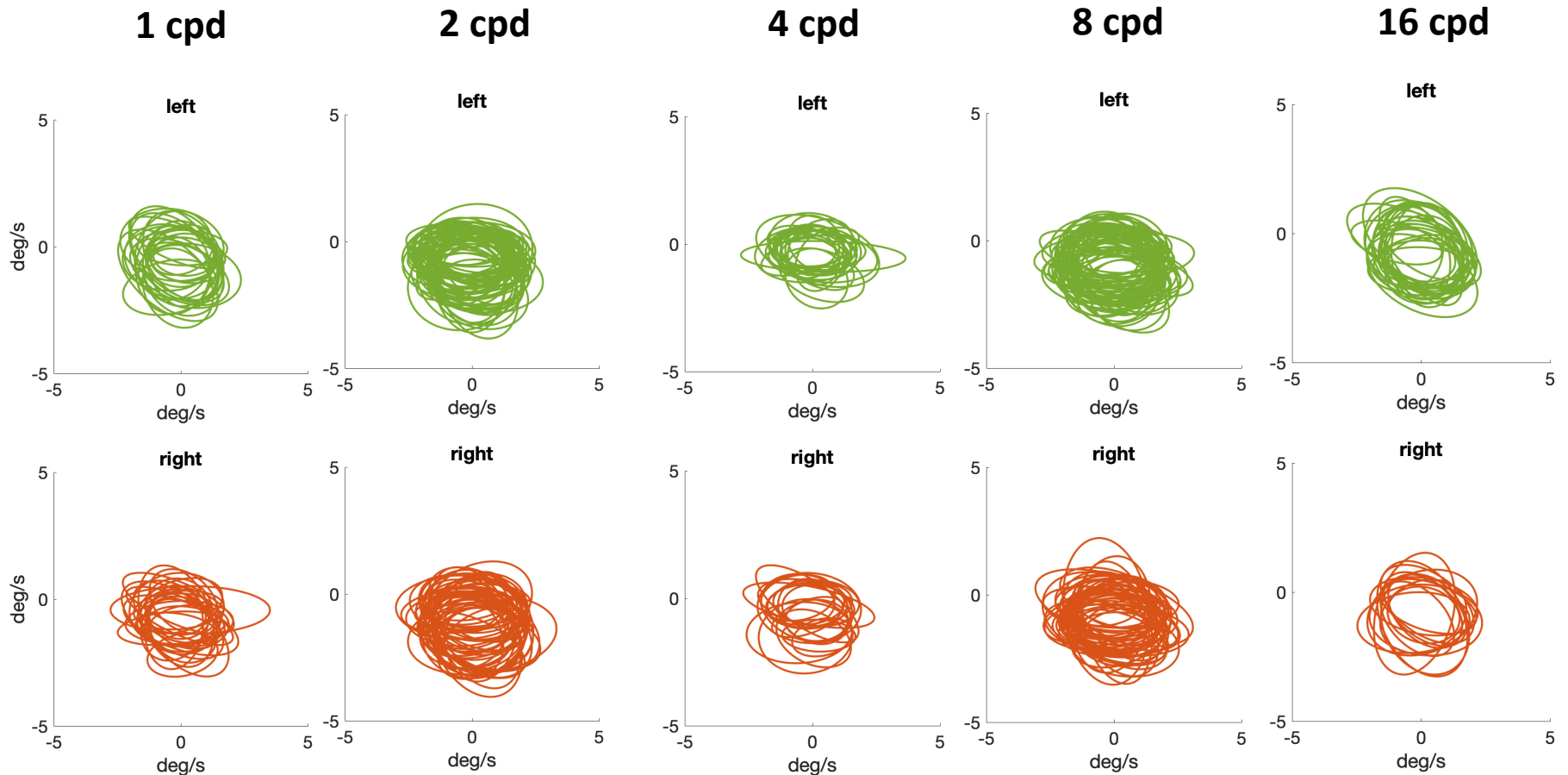
Subj 1



Subj 2



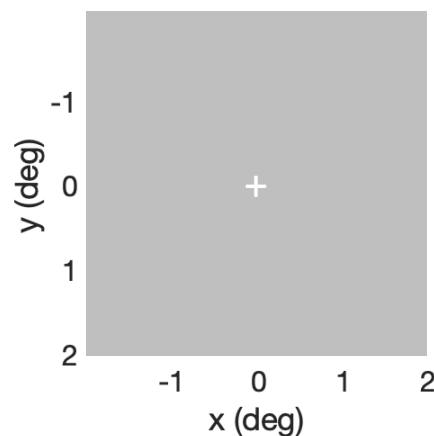
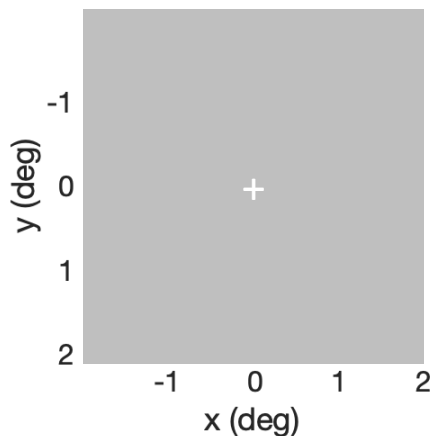
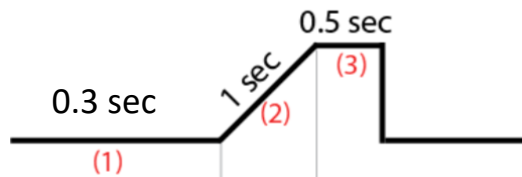
# No Trial-by-Trial Differences between 2 Tilted Grating Viewing



# Tasks to Test Stimulus-Dependent Anisotropic Drift

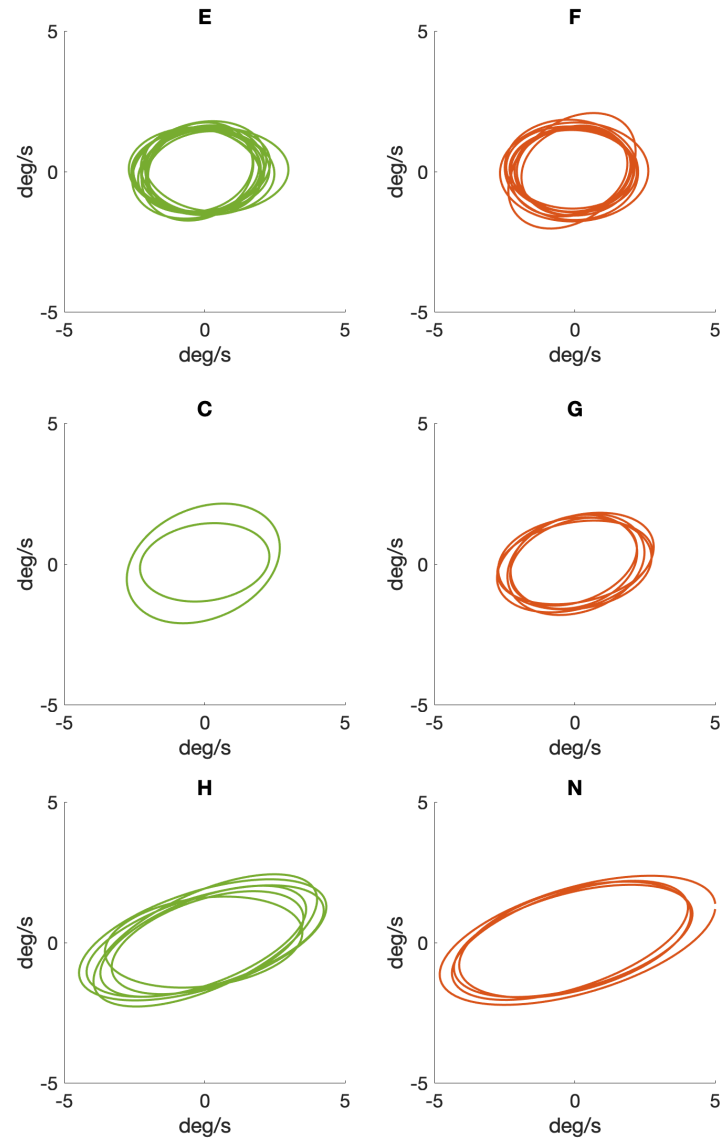
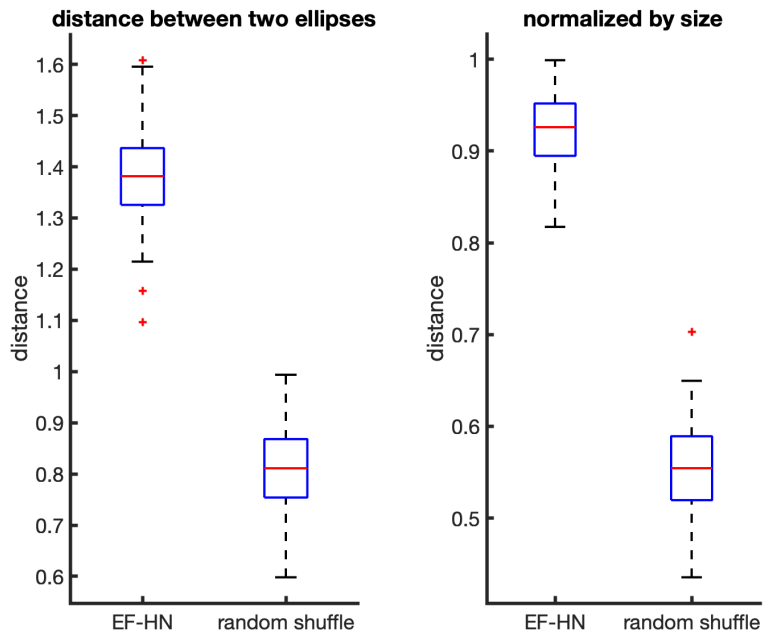
## Letter discrimination task

- Letter pairs,  $\sim 1.5$  deg in size
- Trials are block design; subjects know which letter pairs within trials
- **Do we see any evidence that the DIRECTION of drift changes with task?**



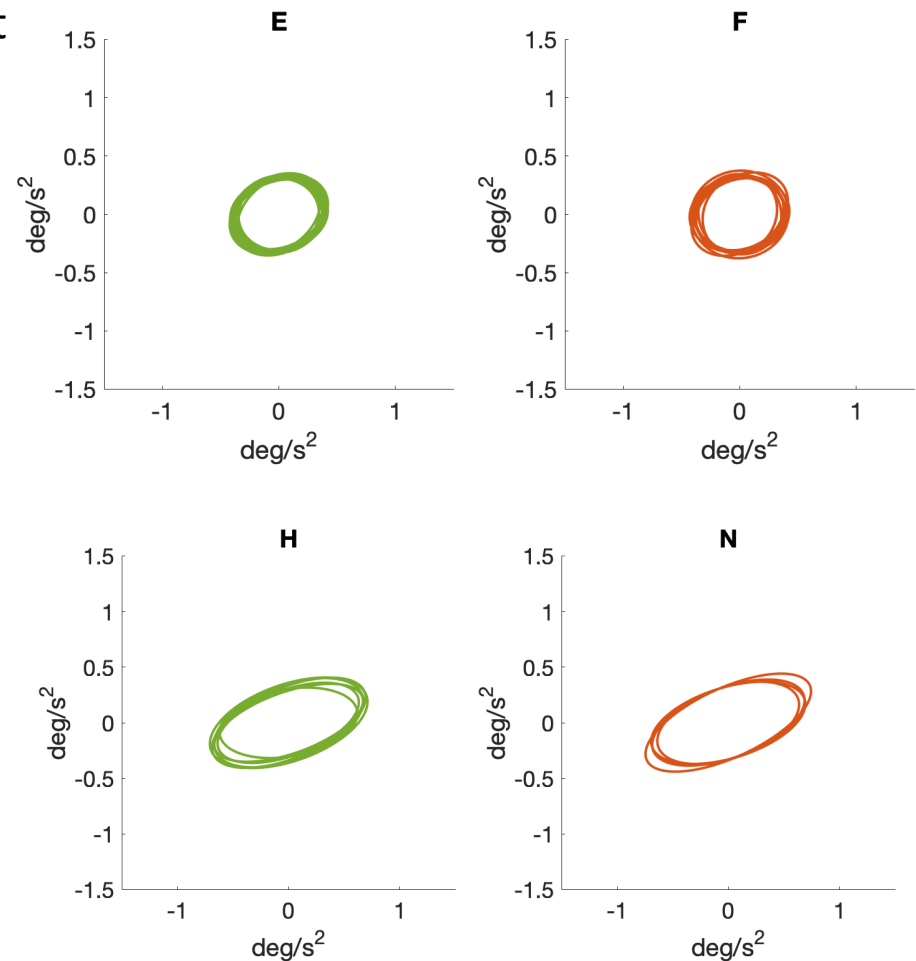
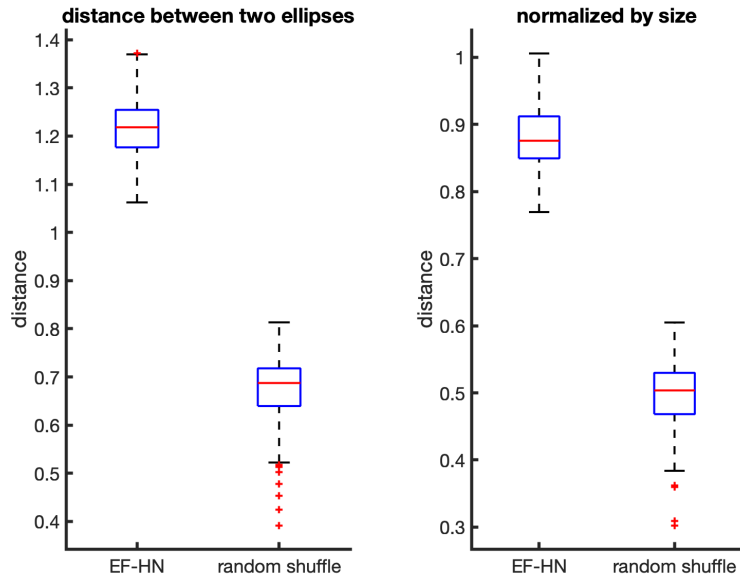
# Difference in velocities between pairs

Preliminary data suggests no directional bias between different letters within blocks, but difference between different pairs



# Difference in accelerations between pairs

Preliminary data suggests no directional bias between different letters within blocks, but difference between different pairs



# Overview of Results from All 3 Tasks

	Grating detection	Grating discrimination	Letter discrimination
Between Stimuli	No difference	Maybe?	No difference
Between Stimuli – trial-by-trial	No difference	No difference	No difference
Between Tasks – trial-by-trial	N.A.	N.A.	YES (need more data)



# Current Hypotheses & Next Steps

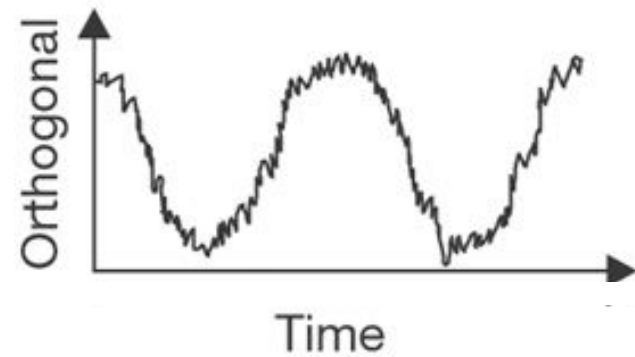
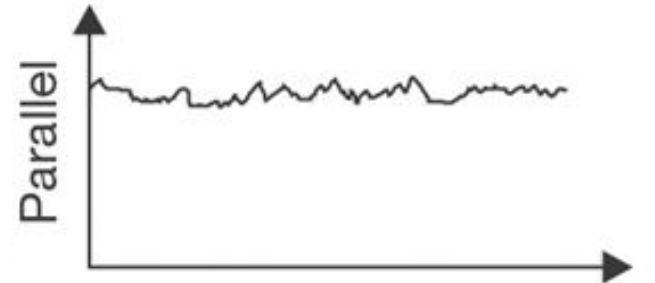
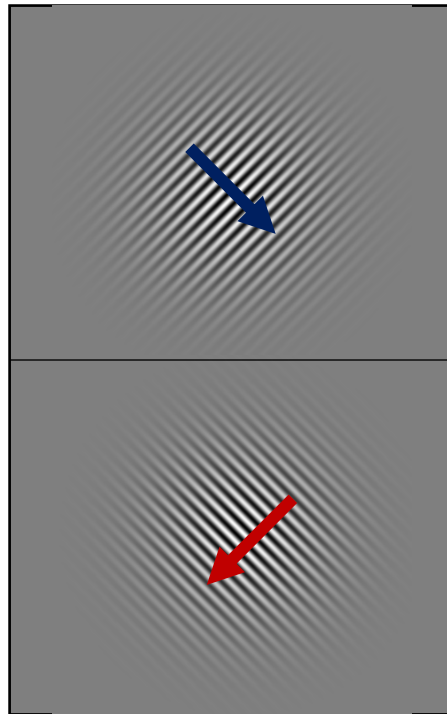
*Switch Speaker*

Optimal bias ocular drifts strategy  
– from the luminance change point of view

**STIMULUS-DRIVEN**  
increase luminance power  
for all stimuli

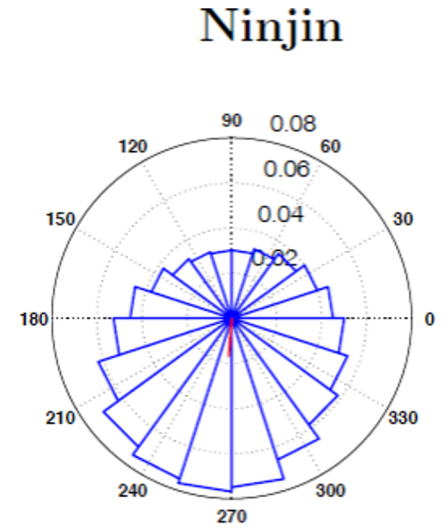
Active, controlled anisotropic FEM

FEM would change  
for each stimulus



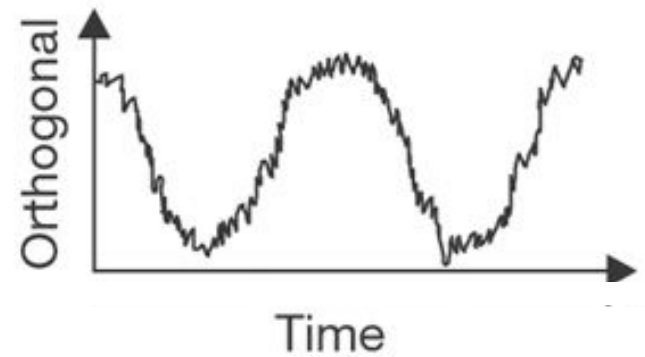
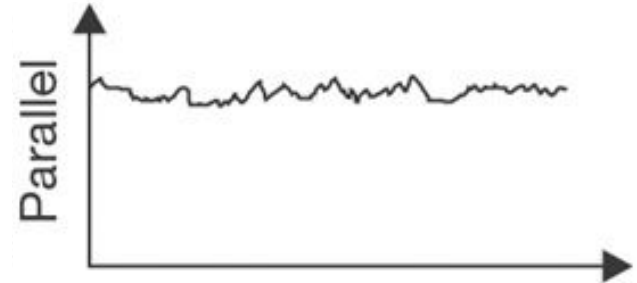
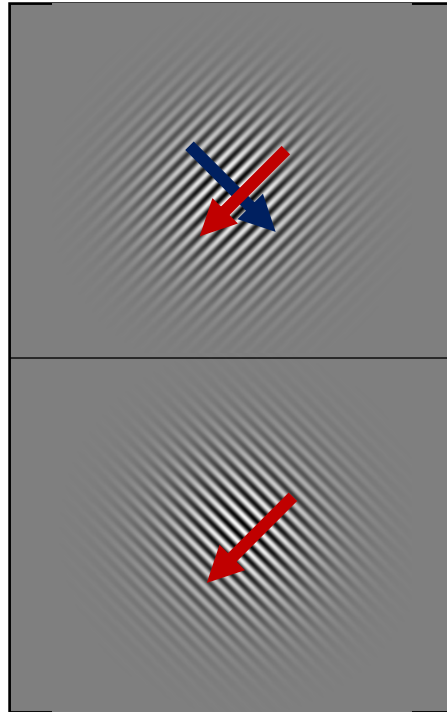
Optimal bias ocular drifts strategy  
– from the luminance change point of view

**STIMULUS-DRIVEN**  
increase luminance power  
for all stimuli



Active, controlled anisotropic FEM

~~FEM change for each stim~~

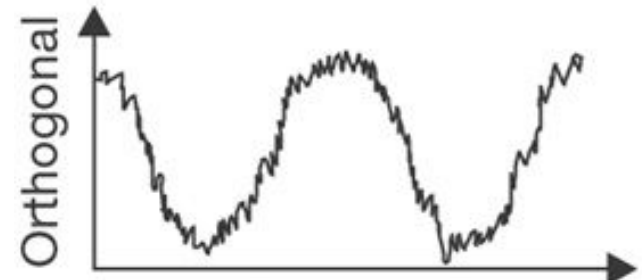
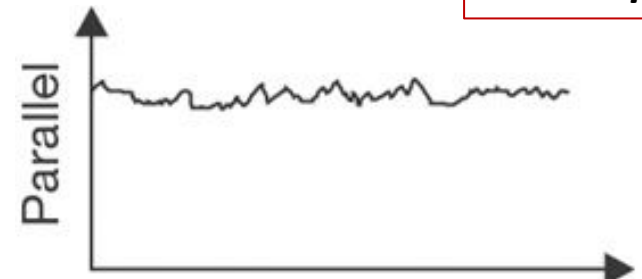
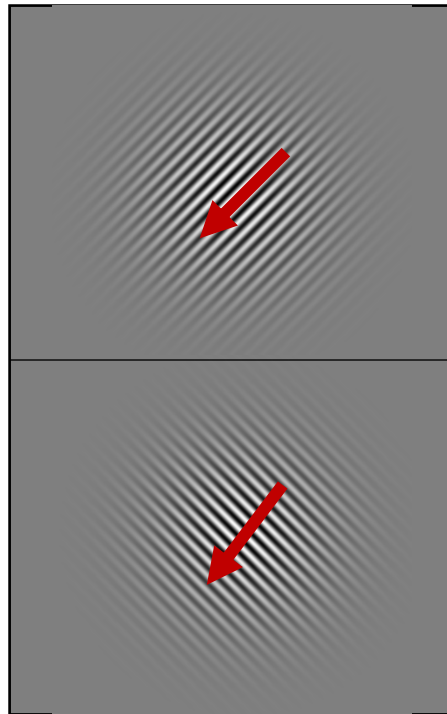


Optimal bias ocular drifts strategy  
– from the luminance change point of view

**STIMULUS-DRIVEN**  
increase luminance power  
for all stimuli

**TASK-DRIVEN**  
increase luminance power difference  
between stimuli

*discriminate  
based on  
overall power*



Time

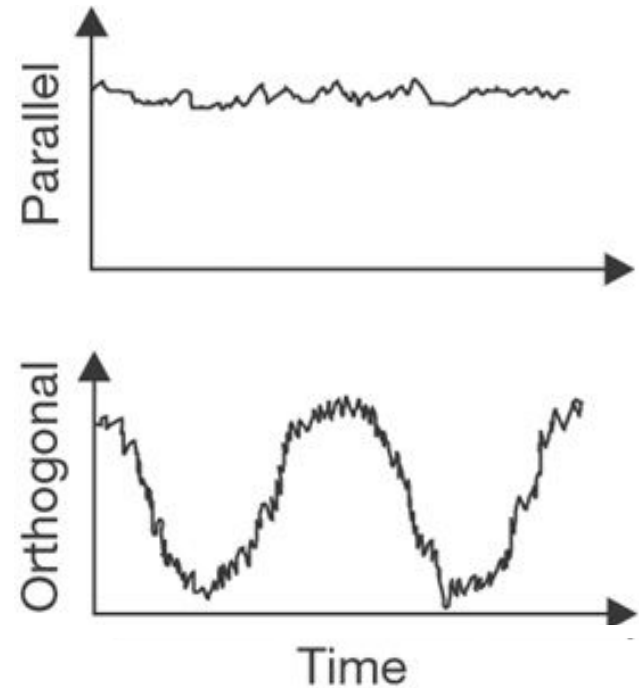
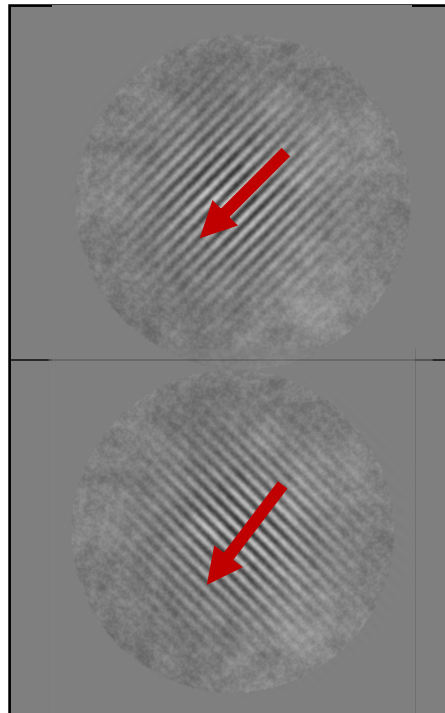
Optimal bias ocular drifts strategy  
– from the luminance change point of view

**STIMULUS-DRIVEN**  
increase luminance power  
for all stimuli

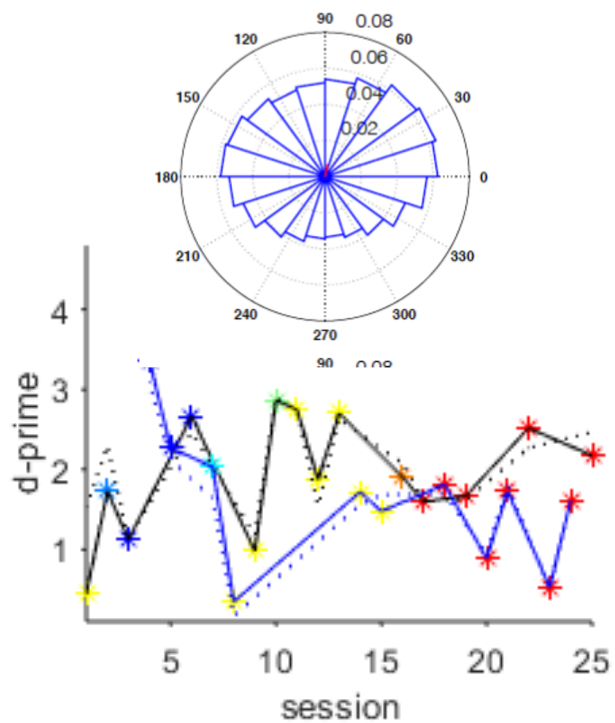
**TASK-DRIVEN**  
increase luminance power difference  
between stimuli

### Detection in noise?

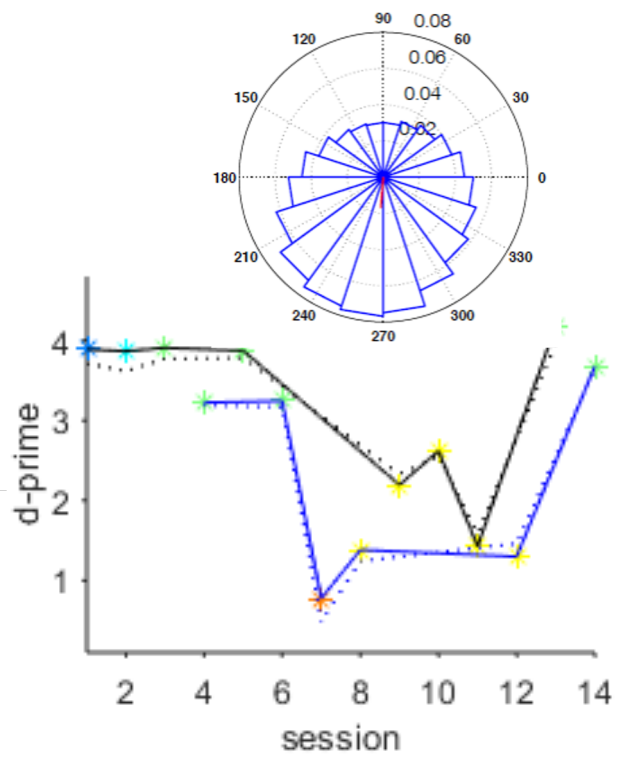
Most benefit  
when eye  
movement is  
orthogonal



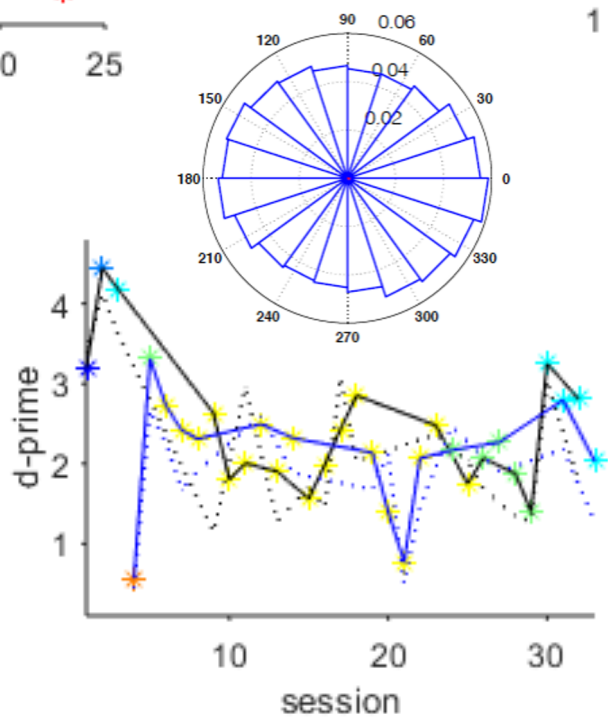
Sara



Ninjin



Andy



Optimal bias ocular drifts strategy  
– from the luminance change point of view

**STIMULUS-DRIVEN**  
increase luminance power  
for all stimuli

**TASK-DRIVEN**  
increase luminance power difference  
between stimuli

Active, controlled anisotropic FEM

Intrinsic, fixed anisotropic FEM

FEM would change  
for each stimulus

FEM would not change, but  
performance would be  
better when there's more  
luminance power in general

FEM would change  
for each task

FEM would not change, but  
performance would be better  
when luminance difference  
between stimuli is higher

??????????

Tilted grating, systematic  
change in performance for  
subjects with more  
anisotropy in drift

Letter task,  
differences  
between letter  
but movement is  
unclear from a  
luminance  
increase  
perspective

??????????

# Next Steps

- Develop and test measures of drift anisotropy
- Analyze multiple datasets for evidence of change in drift anisotropy across conditions
- *a priori* definition of an optimal drift anisotropy for any given task and stimulus that is widely adaptable