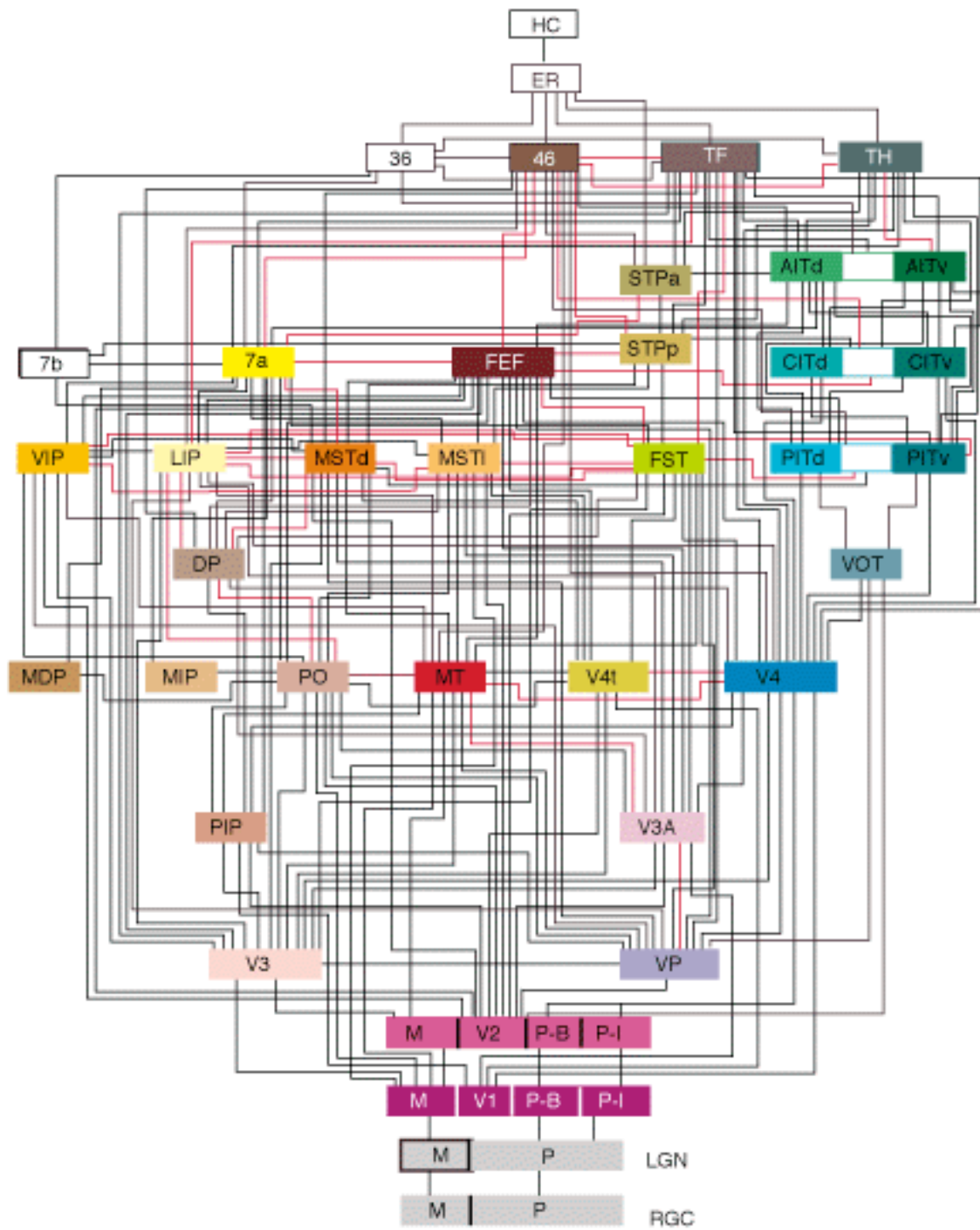


# Top-down signals in cortex & sampling

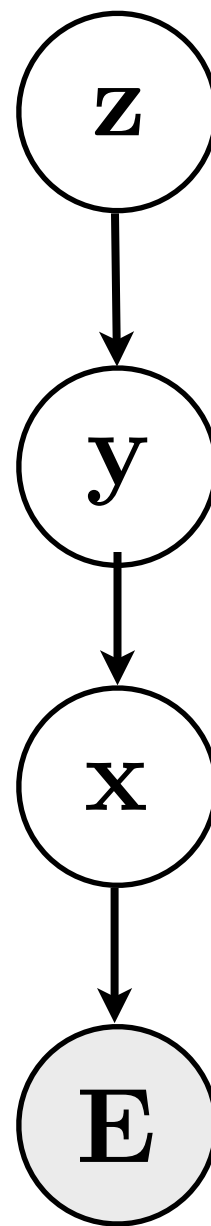
Ralf M. Haefner

Brain & Cognitive Sciences, University of Rochester (NY)

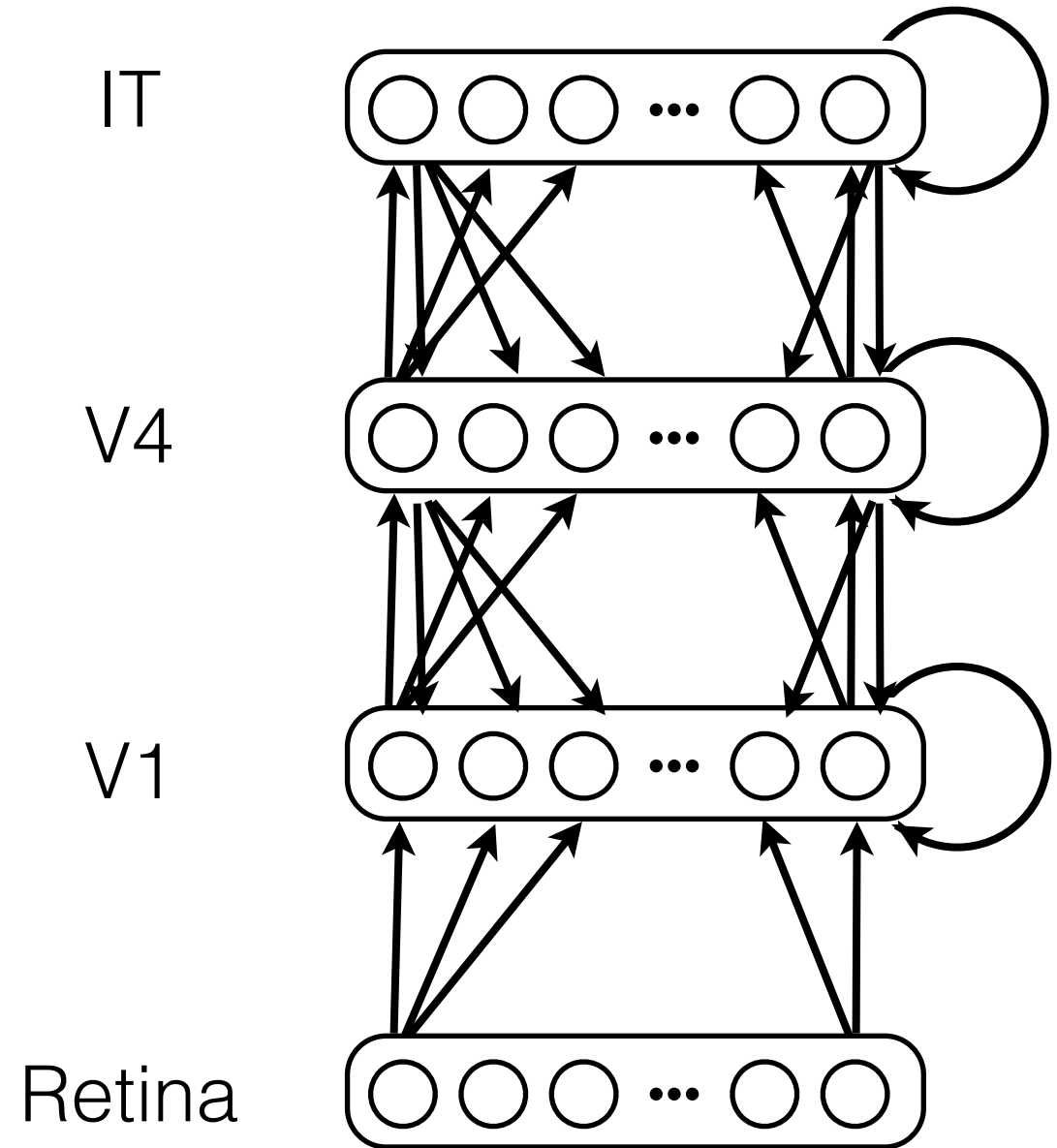
# Full internal model for natural vision unknown



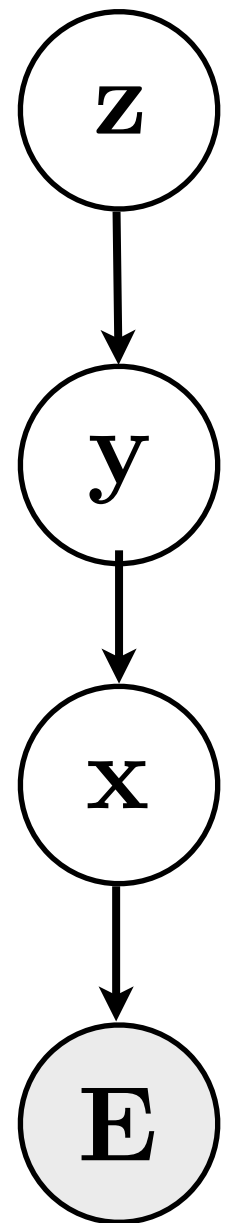
Felleman & van Essen 1991



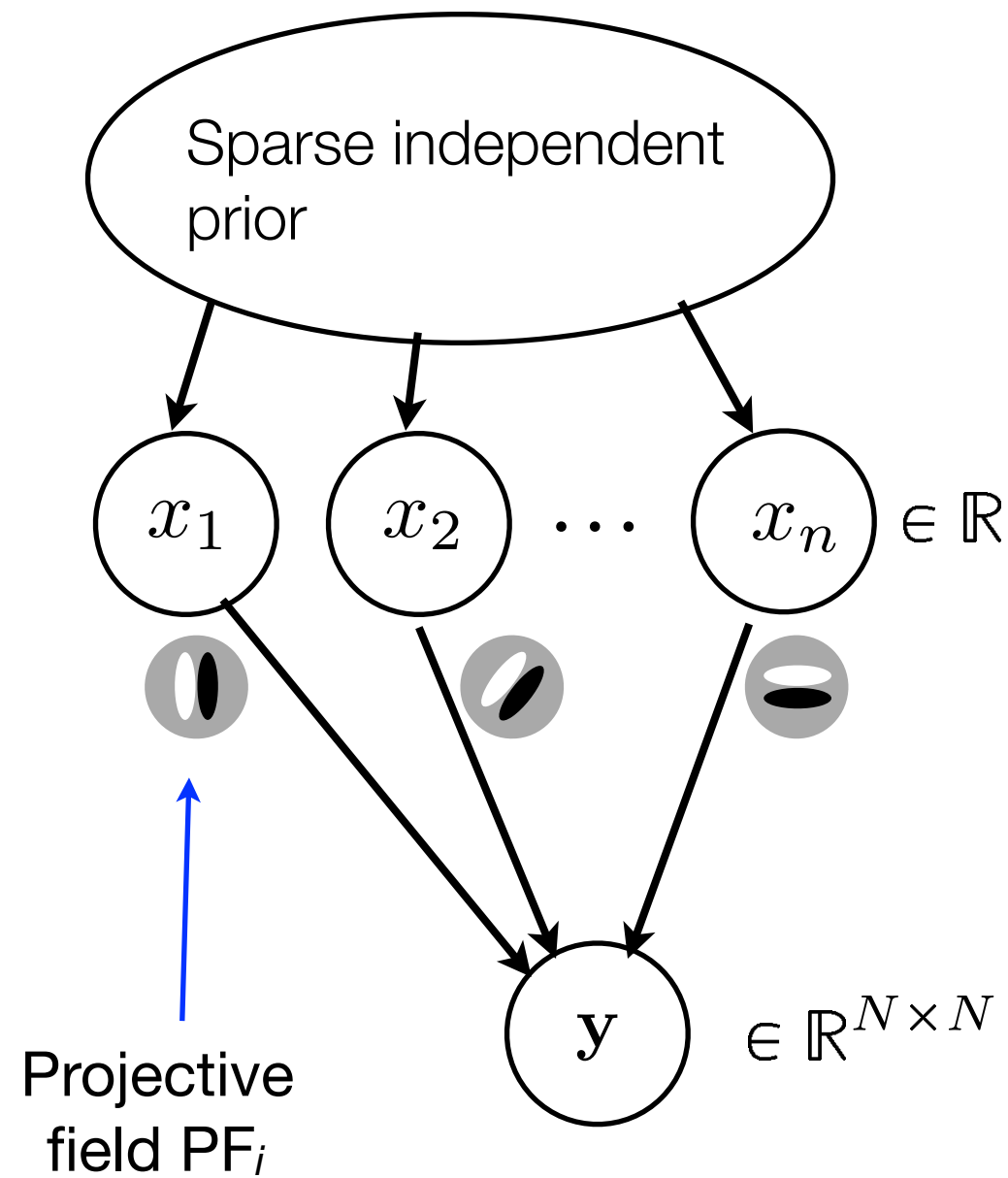
## Feed-forward & feed-back



# Full internal model for natural vision unknown



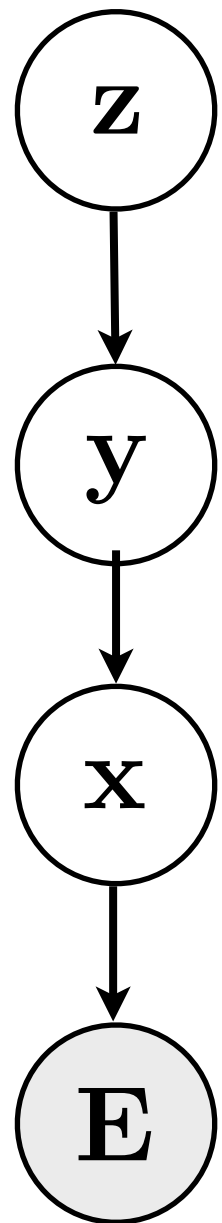
V1  
Retina



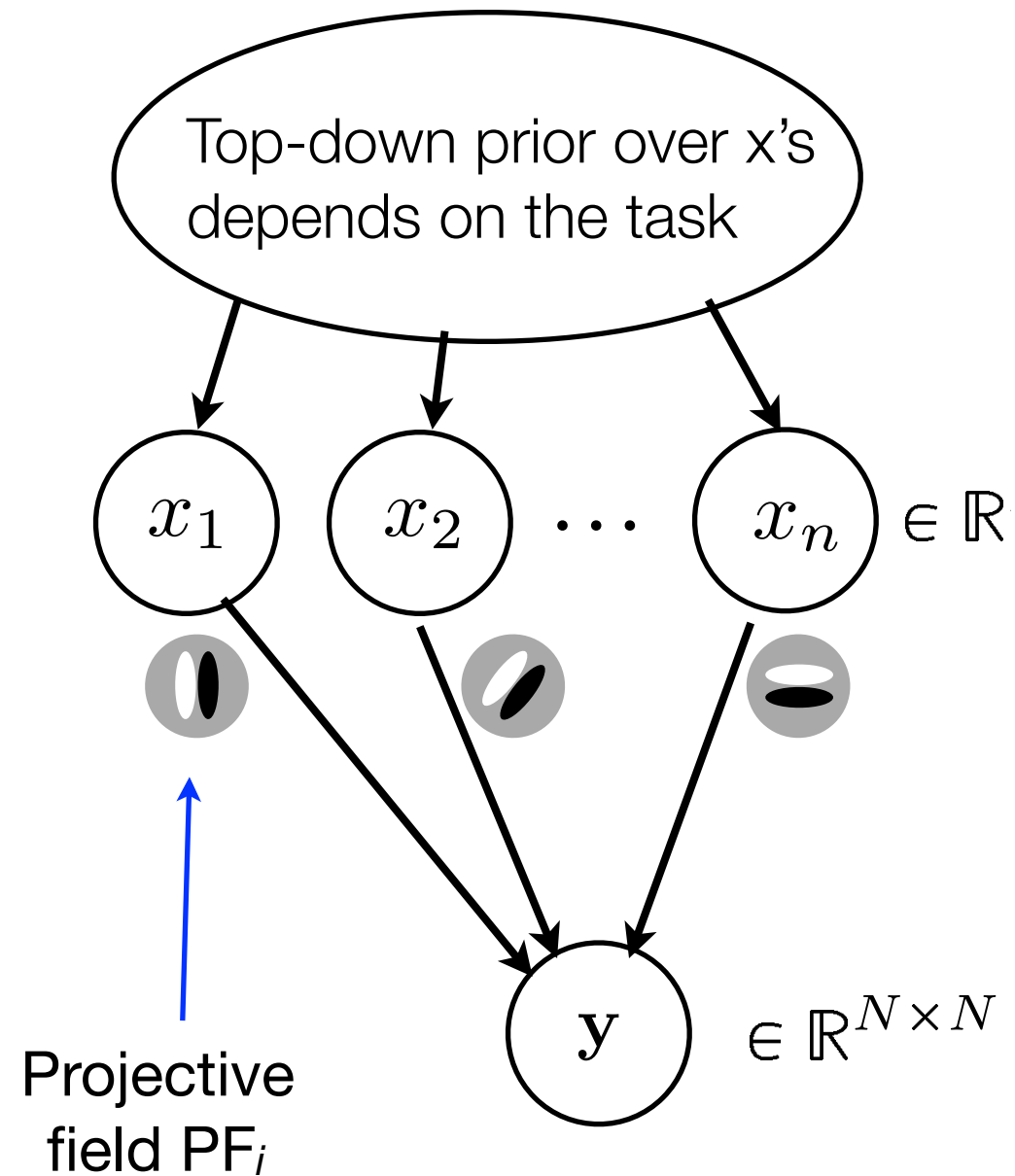
Olshausen &  
Field, 1996,  
1997

$$p(\mathbf{y}|\mathbf{x}) = \mathcal{N} \left( \mathbf{y} : \sum_i PF_i x_i, \sigma^2 \right)$$

# Perturbation approach



V1  
Retina

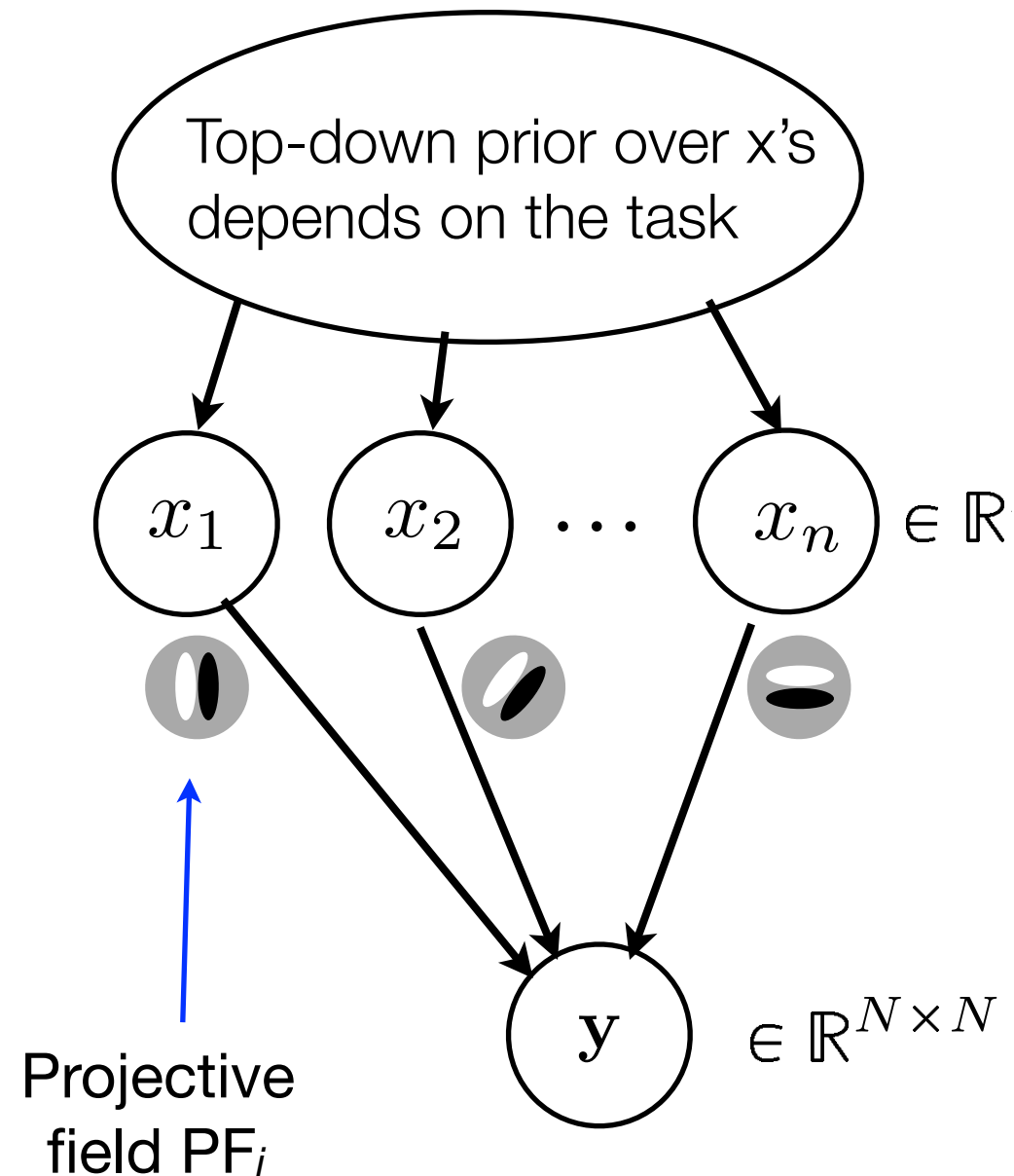


$$p(\mathbf{y}|\mathbf{x}) = \mathcal{N} \left( \mathbf{y} : \sum_i \text{PF}_i x_i, \sigma^2 \right)$$

# Perturbation approach

## Predictions for:

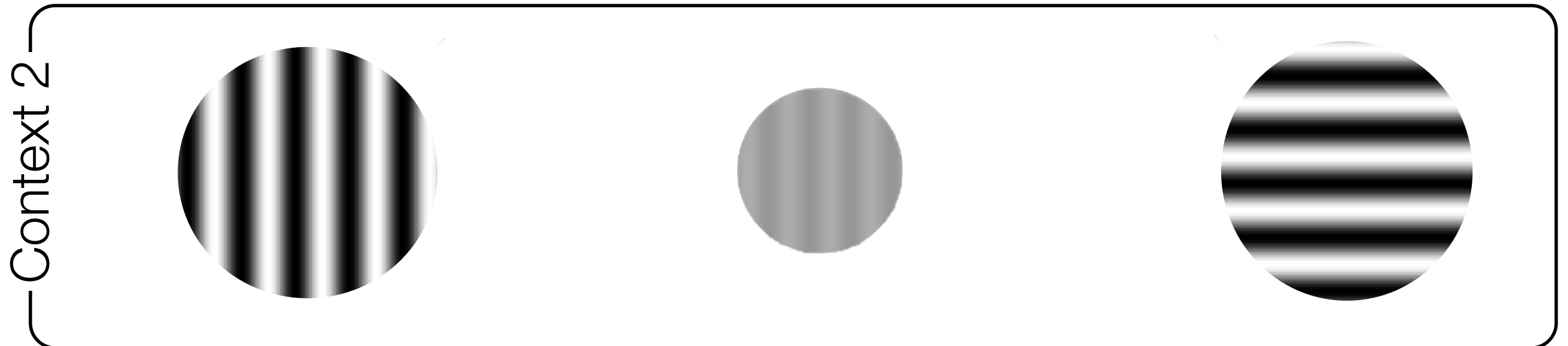
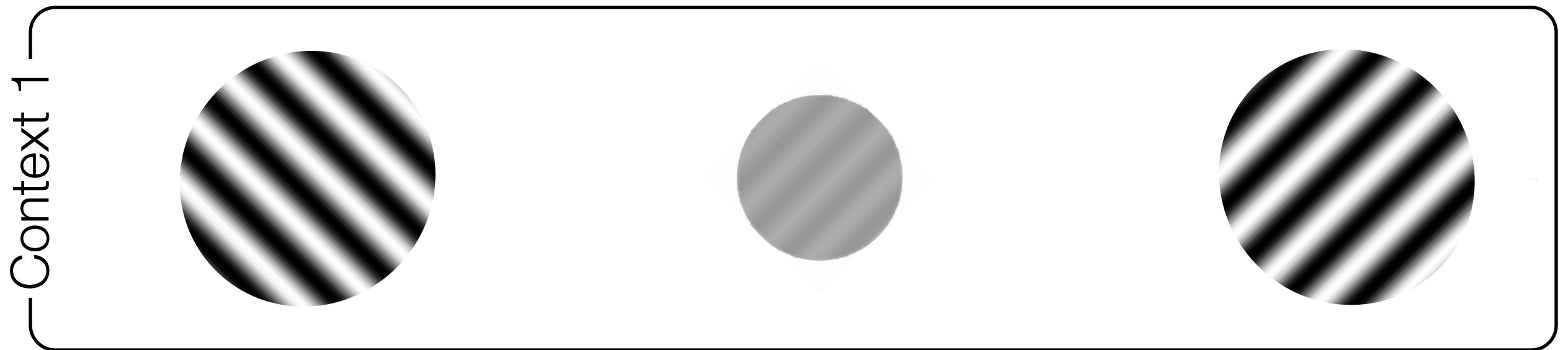
- Difference between before and after learning
- Difference between two different tasks



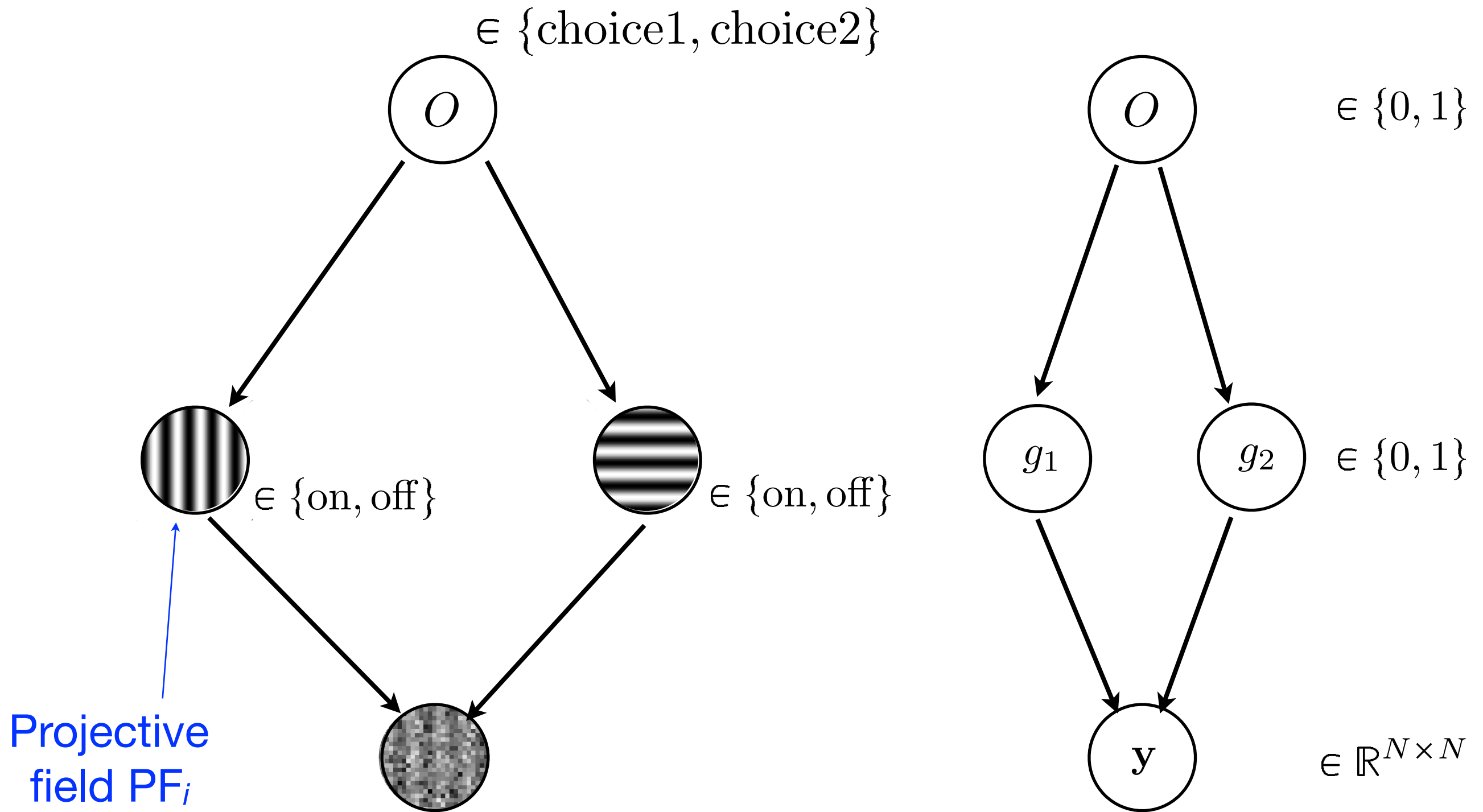
$$p(\mathbf{y}|\mathbf{x}) = \mathcal{N} \left( \mathbf{y} : \sum_i \text{PF}_i x_i, \sigma^2 \right)$$

# Orientation discrimination task

Which of two perpendicular gratings caused the noisy image on the screen?



# 2AFC task: generative model

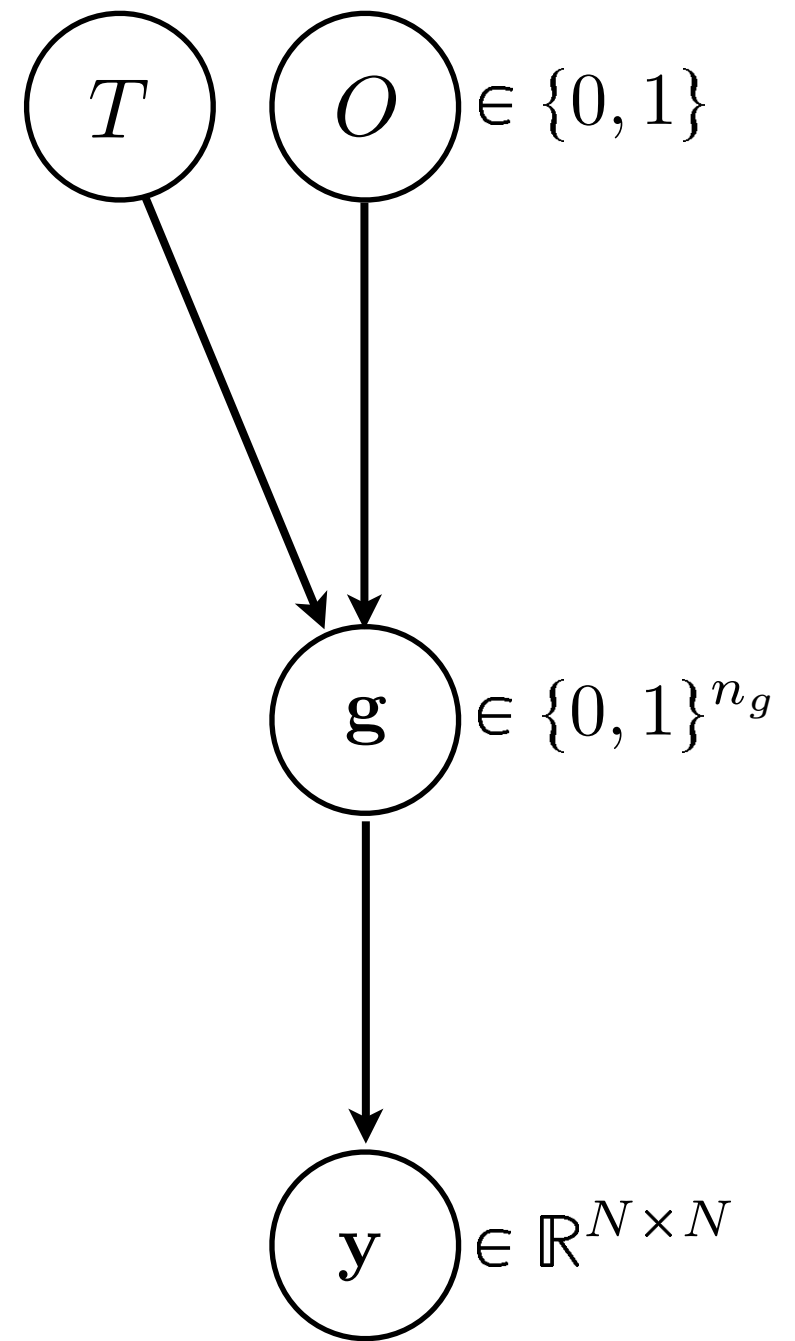
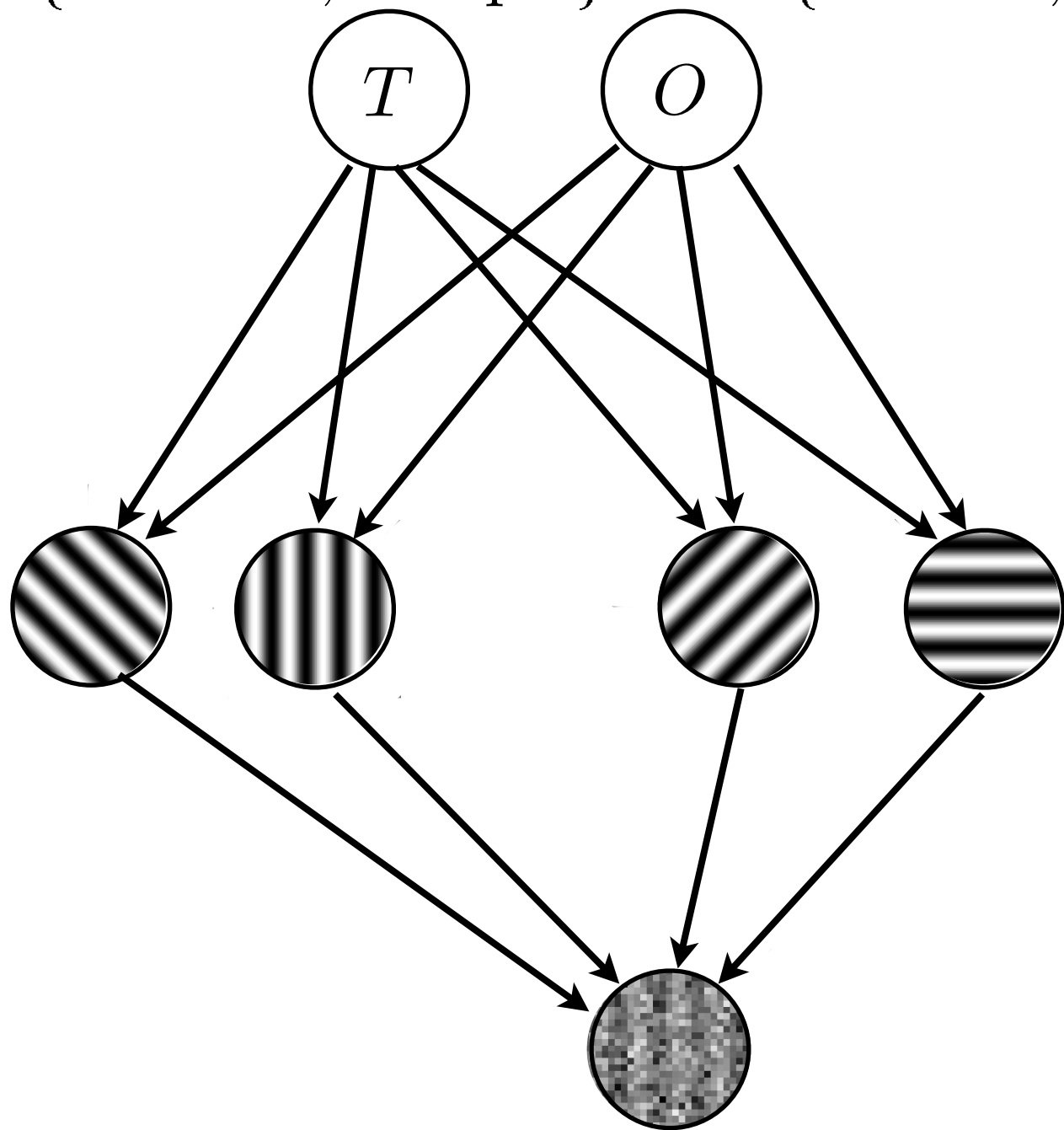


$$p(\mathbf{y}|\mathbf{g}) = \mathcal{N} \left( \mathbf{y} : \sum_i \text{PF}_i g_i, \sigma_y^2 \right)$$

# 2AFC task: generative model

$\in \{\text{cardinal, oblique}\}$

$\in \{\text{choice1, choice2}\}$



$$p(\mathbf{y}|\mathbf{g}) = \mathcal{N} \left( \mathbf{y} : \sum_i \text{PF}_i g_i, \sigma_y^2 \right)$$

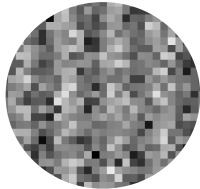


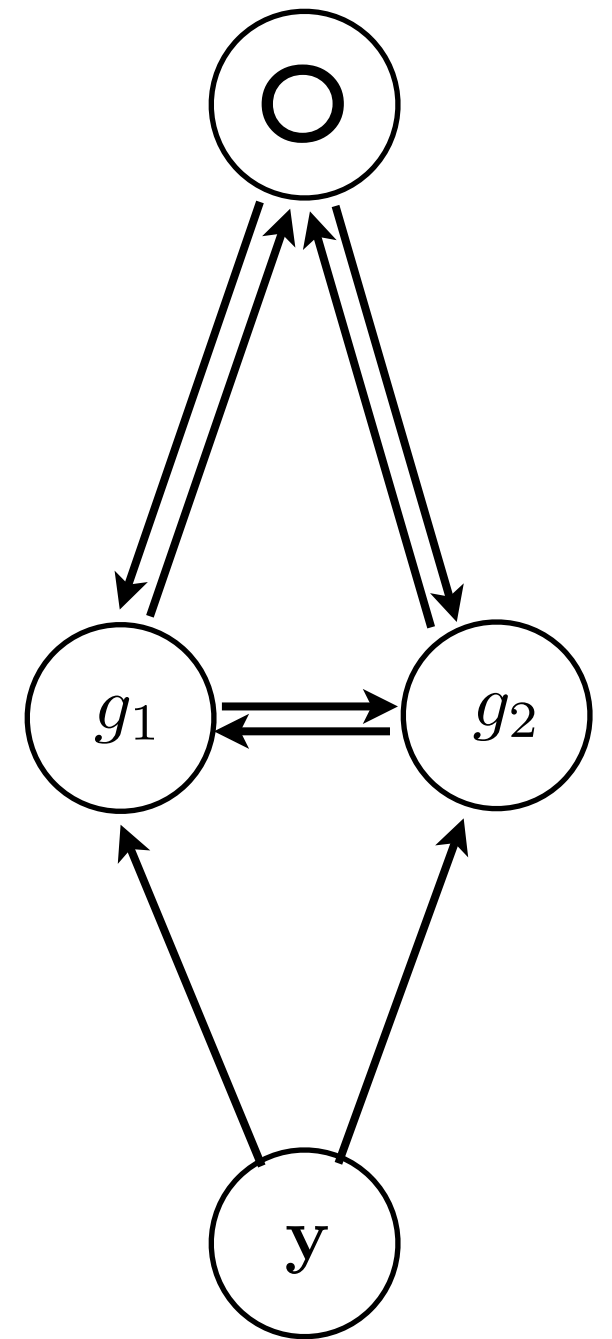
# Inference by neural sampling

$O \sim p(O|g_1, g_2)$  

$g_1 \sim p(g_2|O, g_1, \mathbf{y})$  

$g_2 \sim p(g_1|O, g_1, \mathbf{y})$  

$\mathbf{y} =$  



# Predictions

$$O \sim p(O|g_1, g_2)$$



$$p(O|\mathbf{y})$$

$$g_1 \sim p(g_1|O, g_2, \mathbf{y})$$



$$p(g_1|\mathbf{y})$$

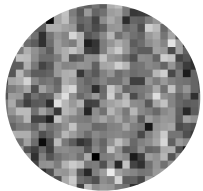
$$g_2 \sim p(g_2|O, g_1, \mathbf{y})$$



$$p(g_2|\mathbf{y})$$

$$p(g_1, g_2|\mathbf{y})$$

$\mathbf{y} =$



Noise correlations...

...and their task-dependence

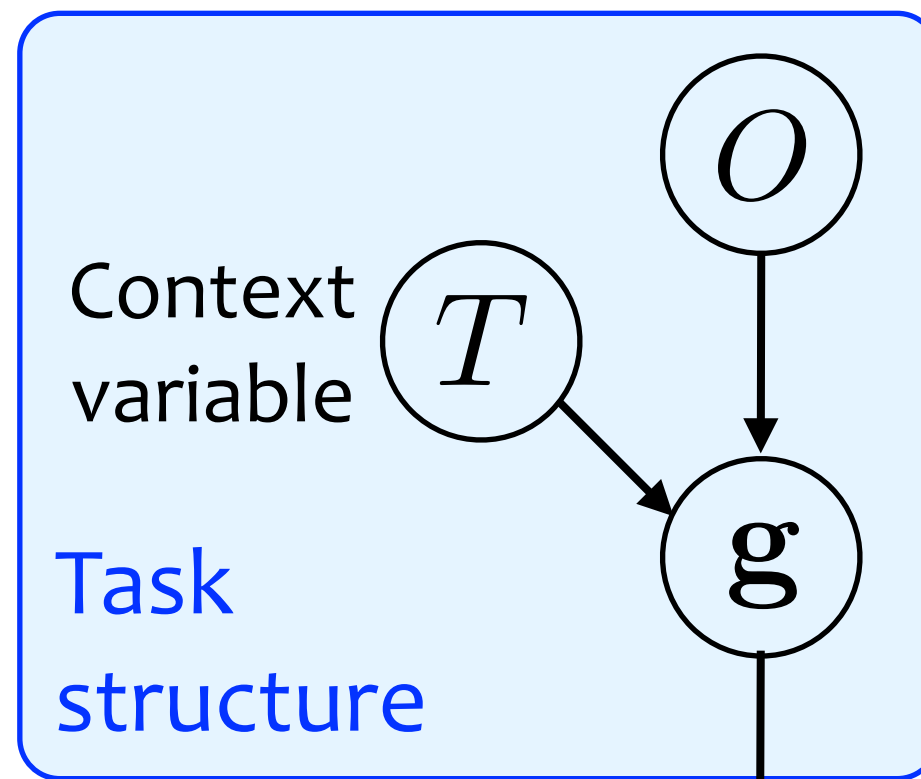
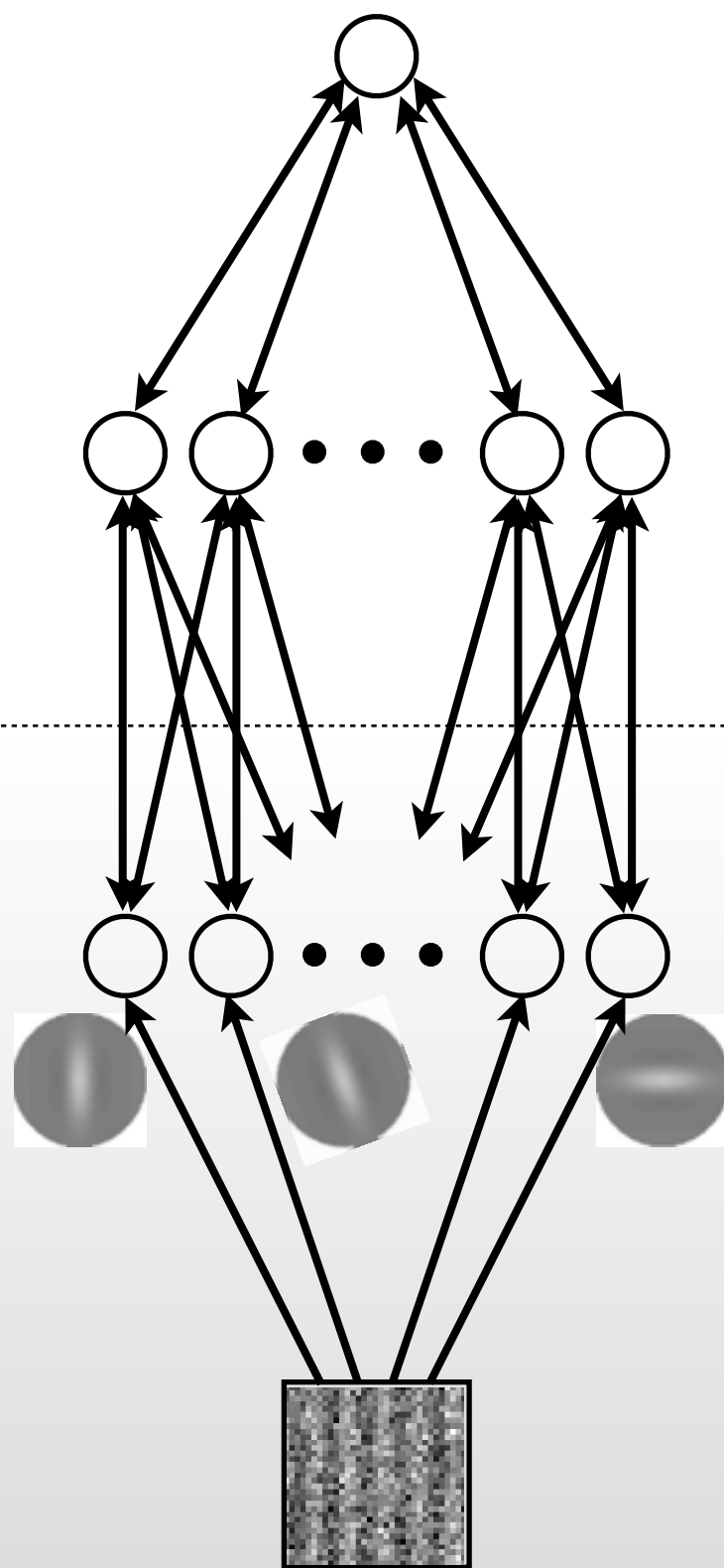
Choice probabilities and psychophysical kernels

# Correspondence to the visual system

LIP/PPC/  
PFC...?

V2/V4/IT?

V1



Decision  
variable

Grating  
variables

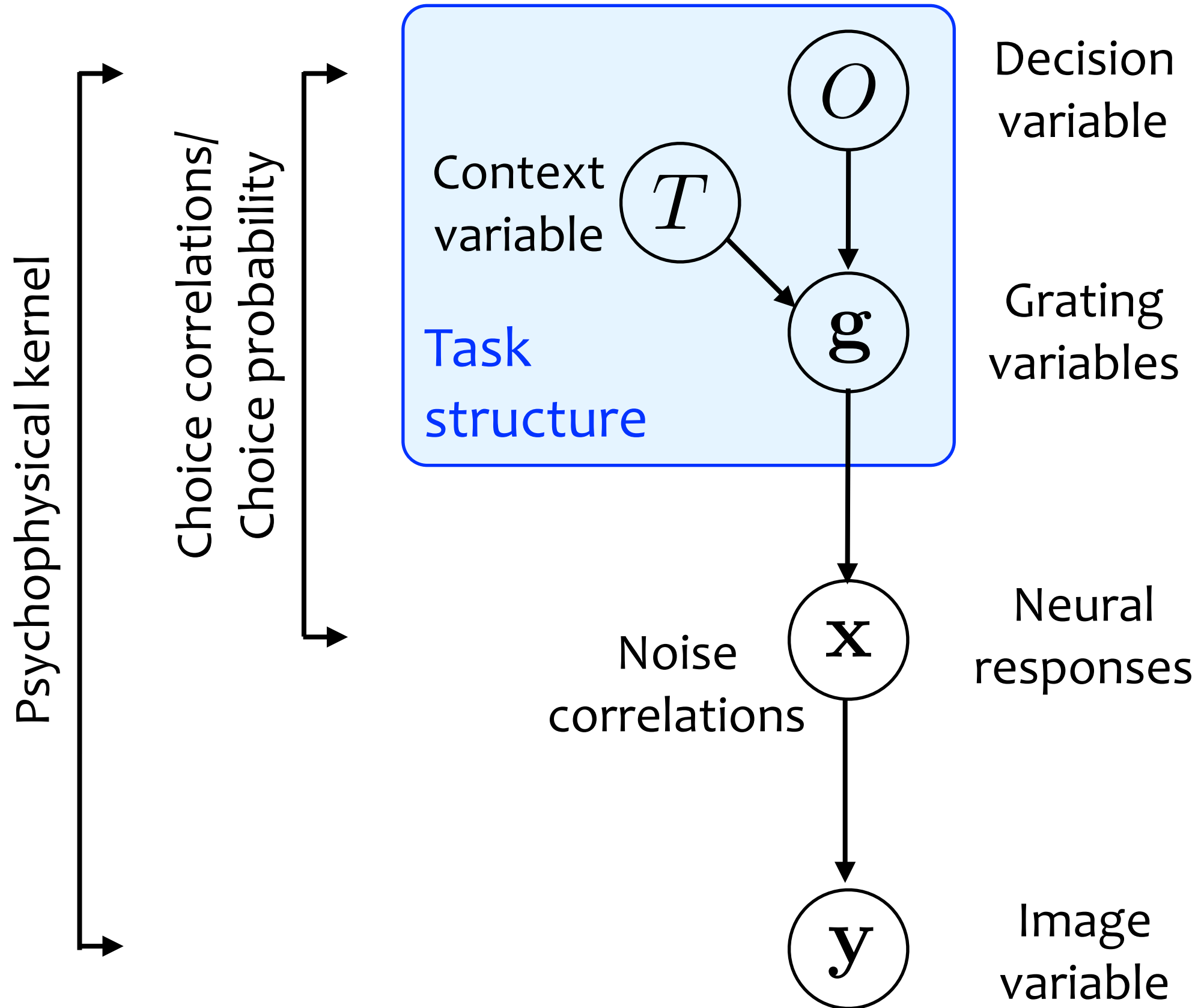
Neural  
responses

Image  
variable

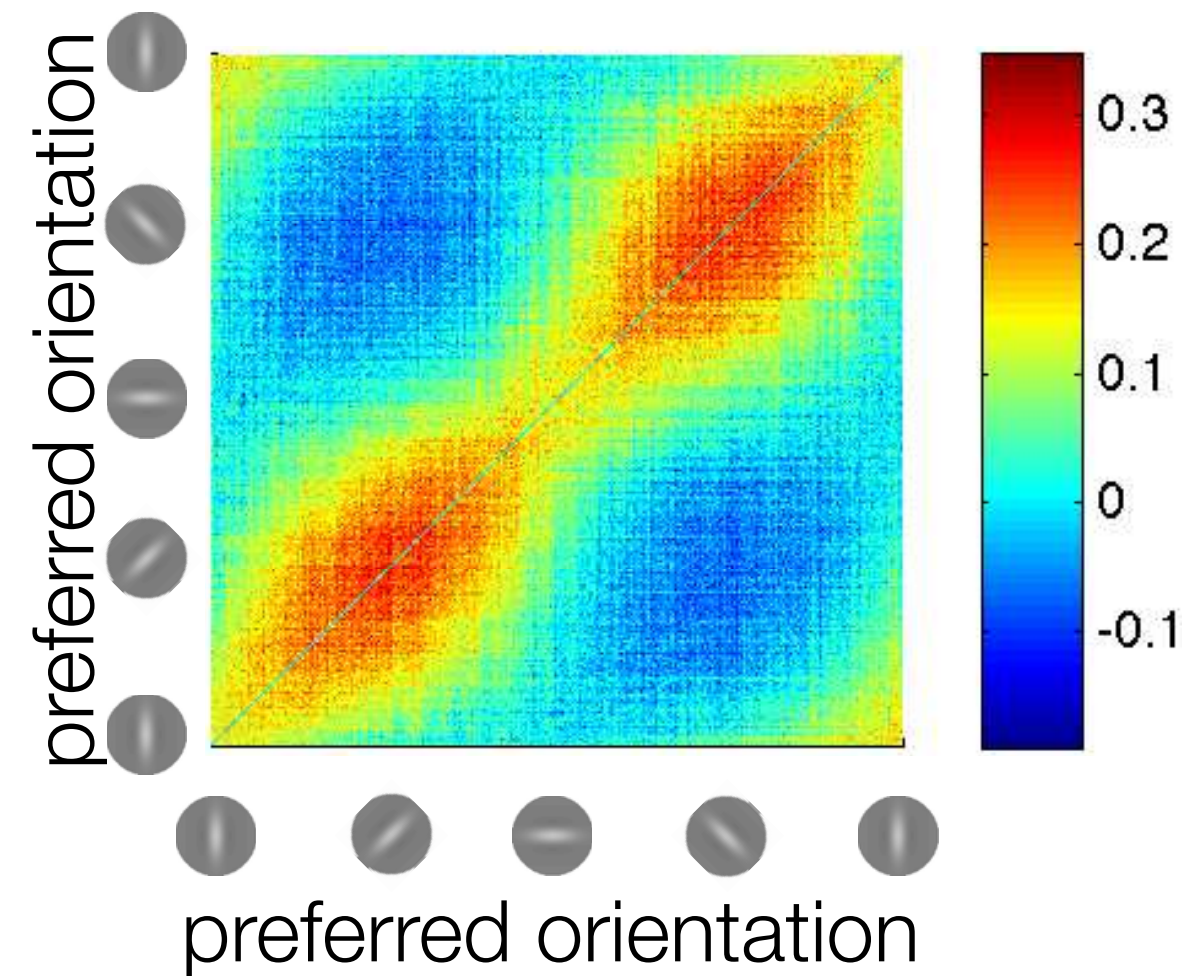
Probabilistic sparse-coding model: Hoyer & Hyvarinen 2003  
Gaussian scale mixture: Schwartz & Simoncelli, 2001

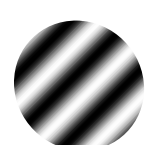

V1/early vision

# Dependencies and observables

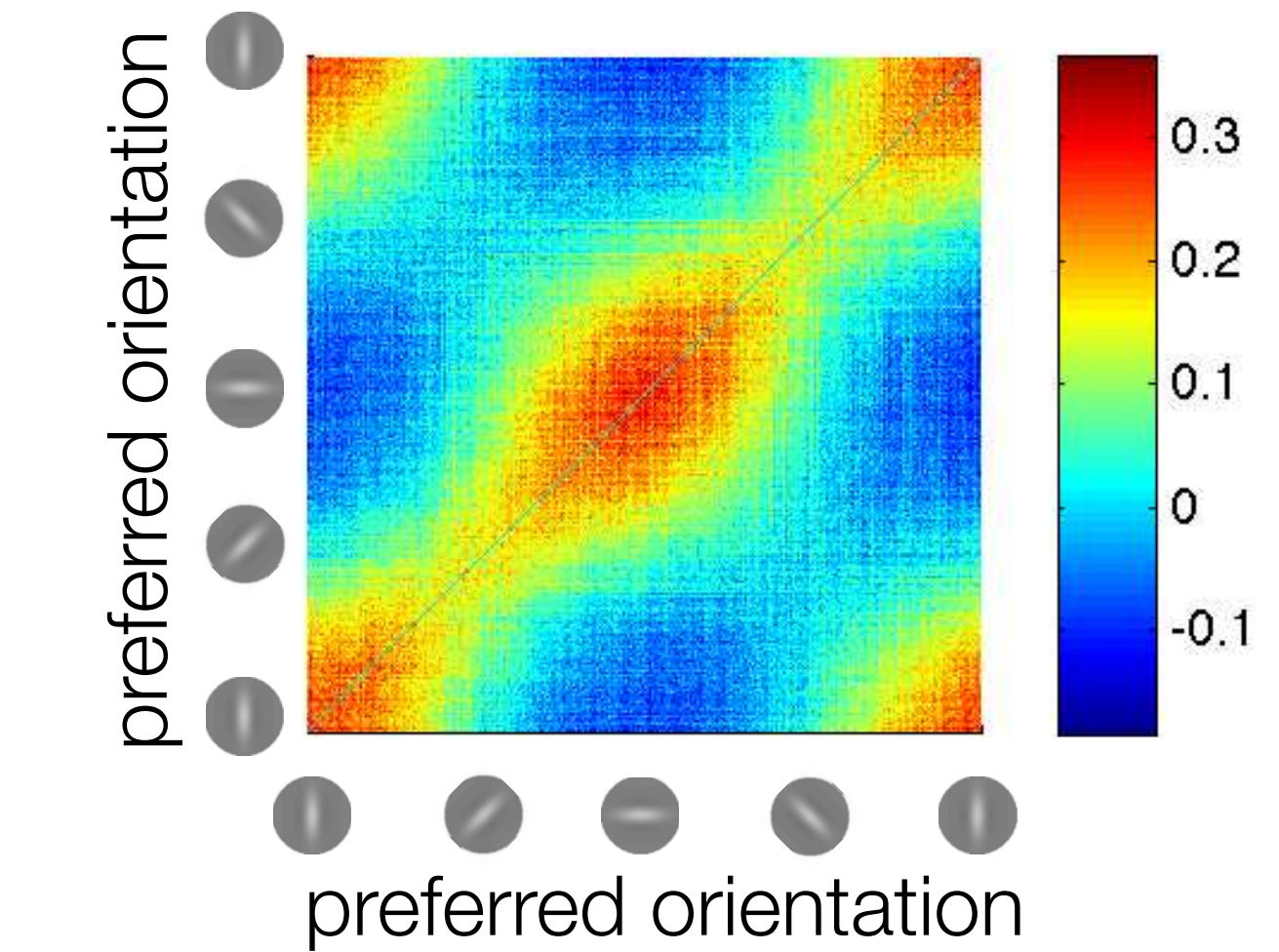


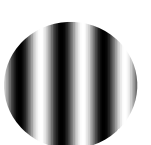
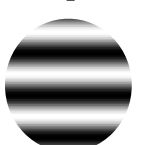
# Model: noise correlations in $\mathbf{x}$



Task:  vs 

**Context 1**

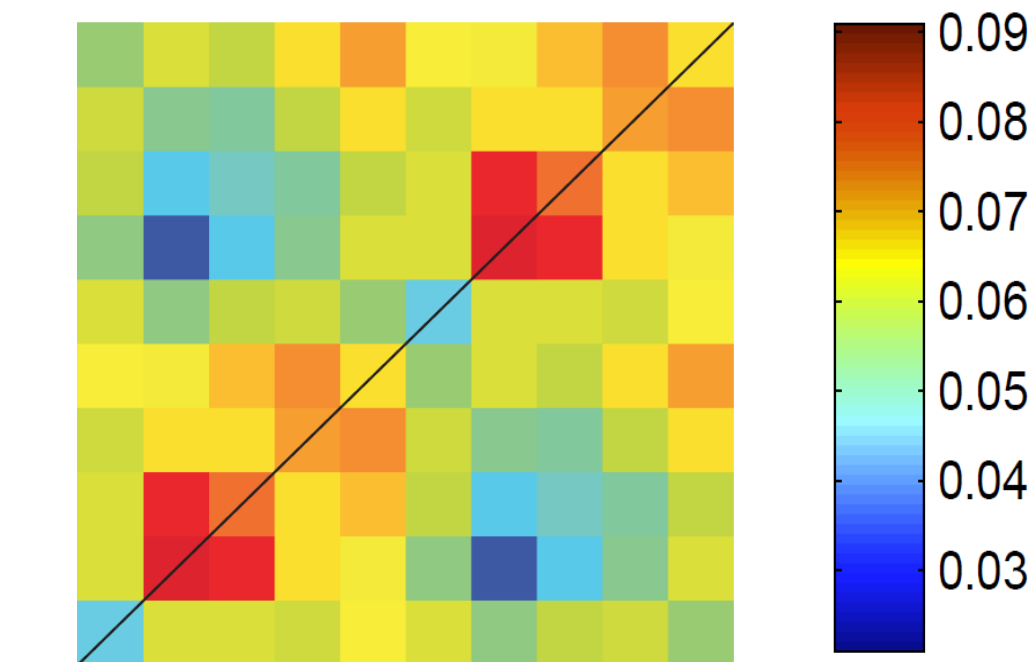
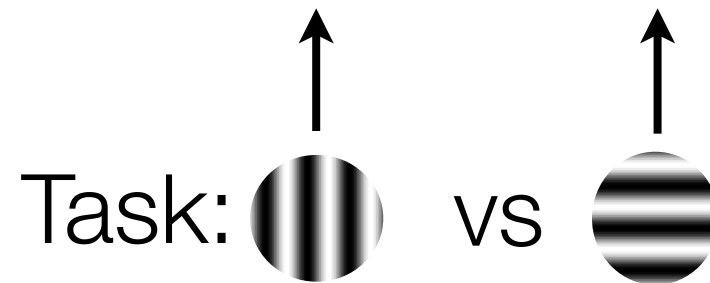
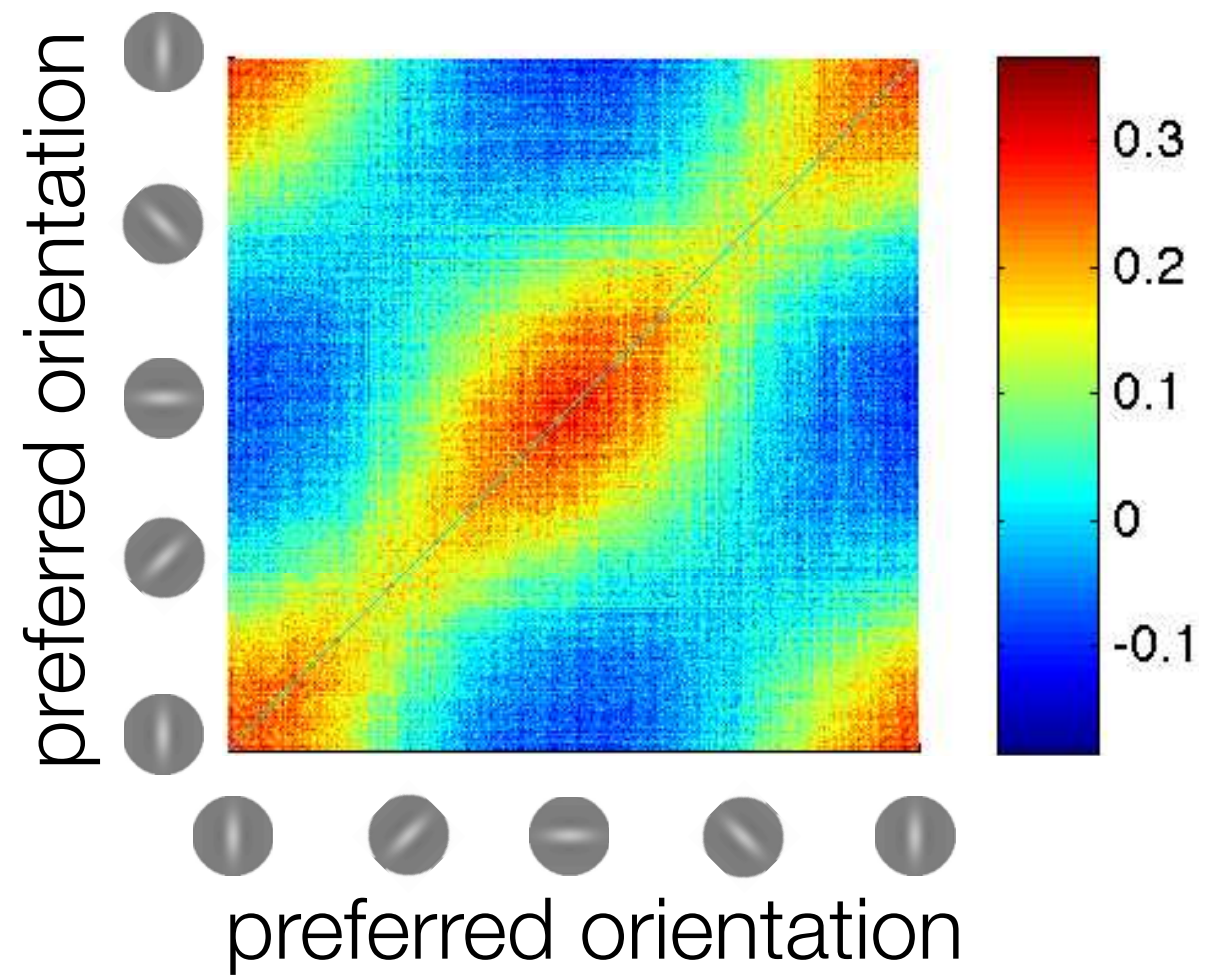
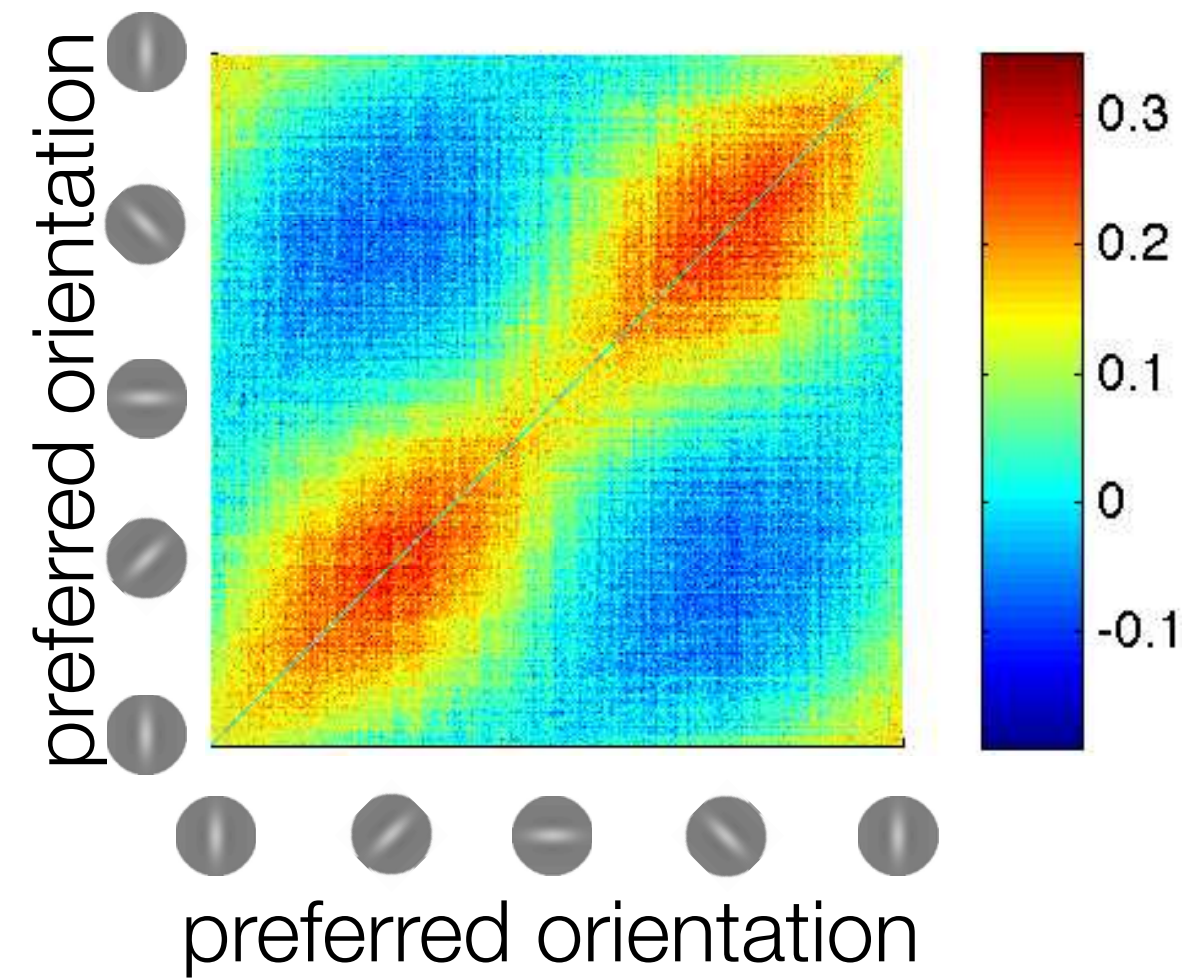


Task:  vs 

**Context 2**



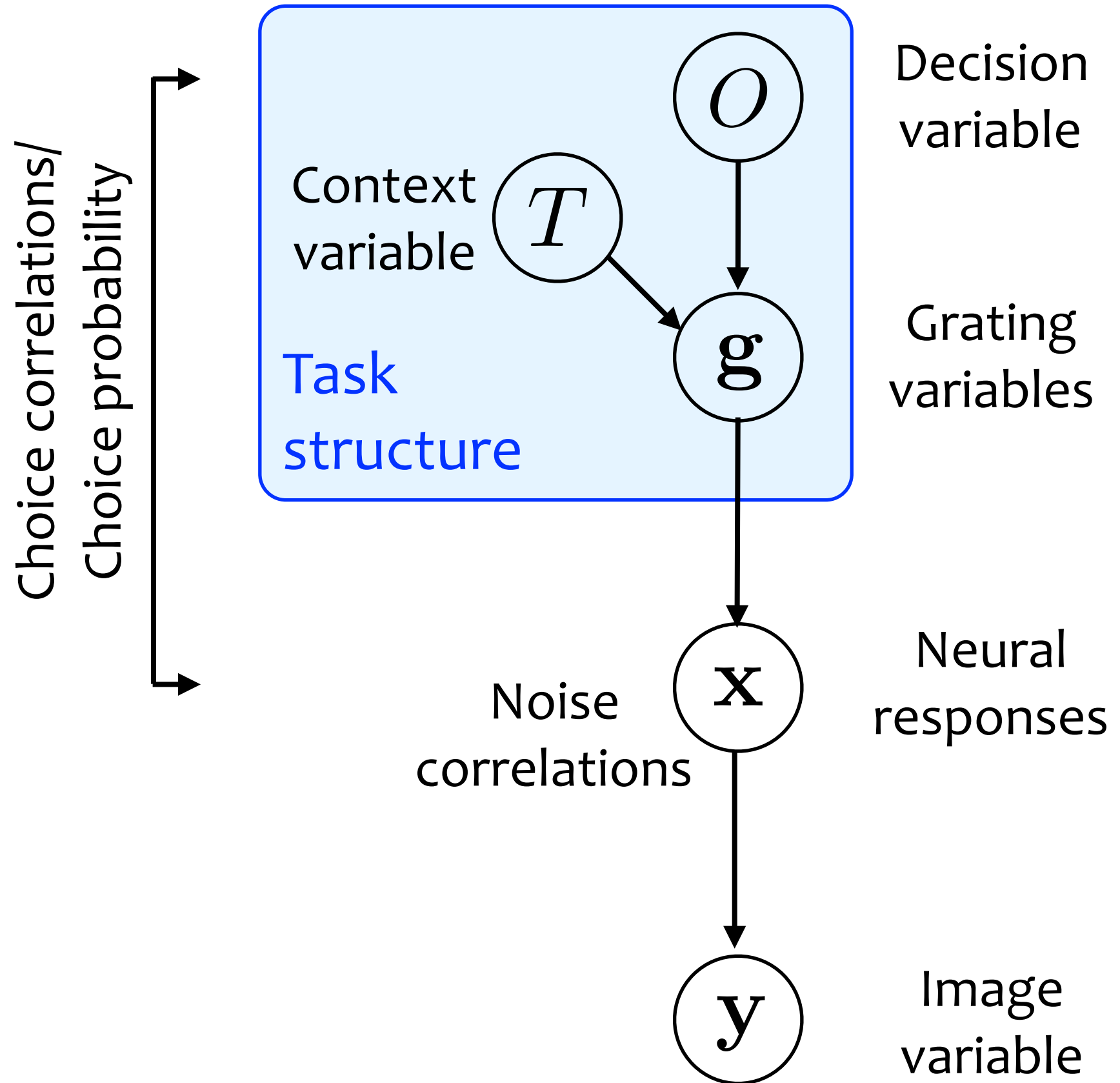
# Model: noise correlations in $\mathbf{x}$



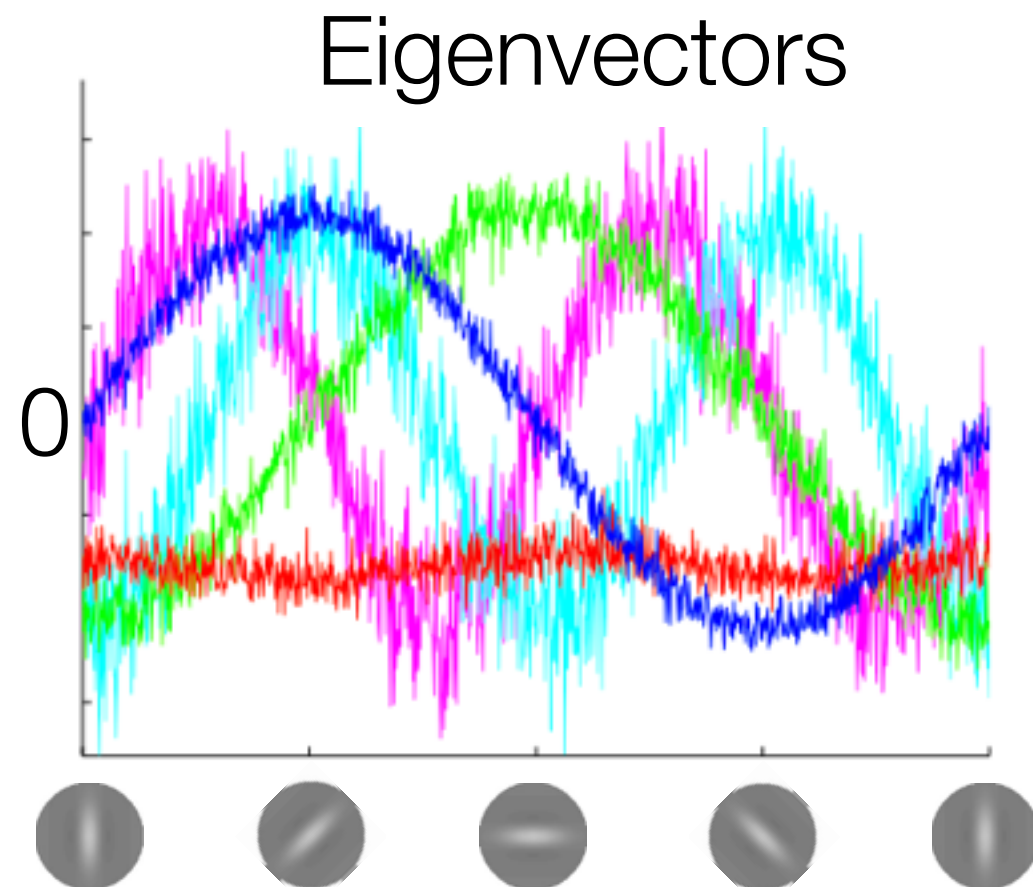
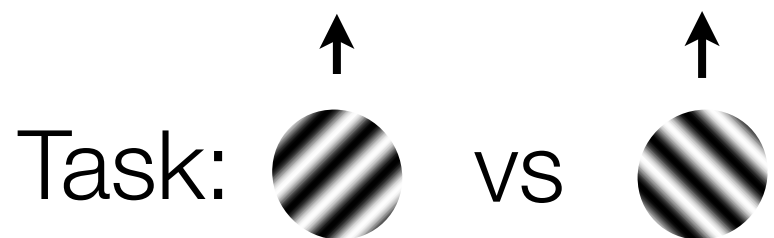
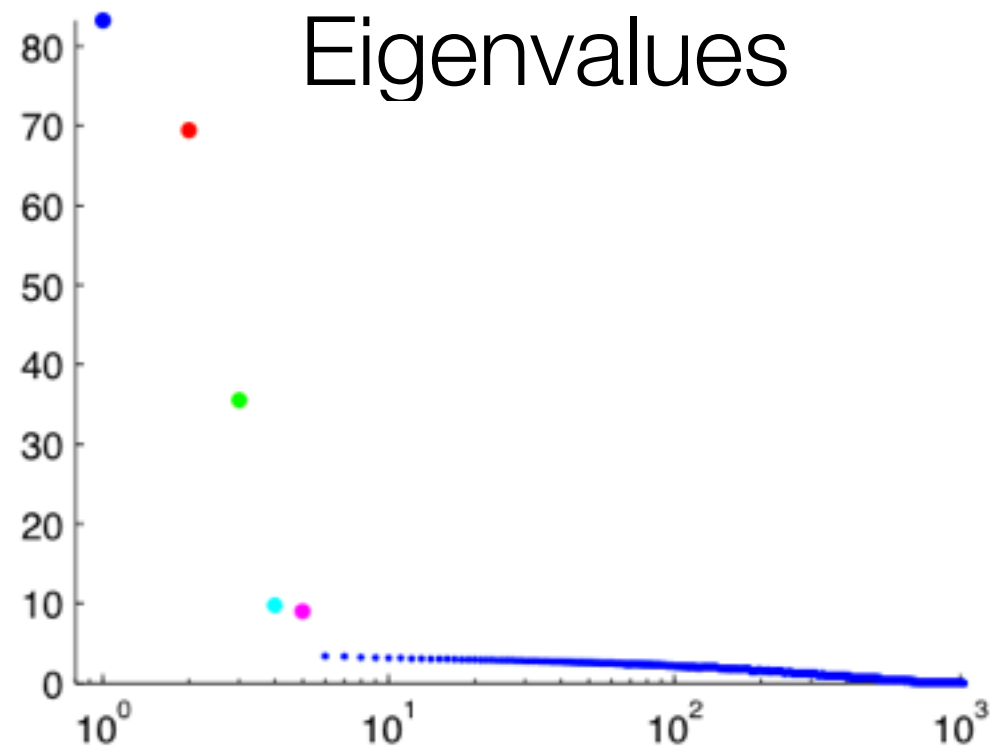
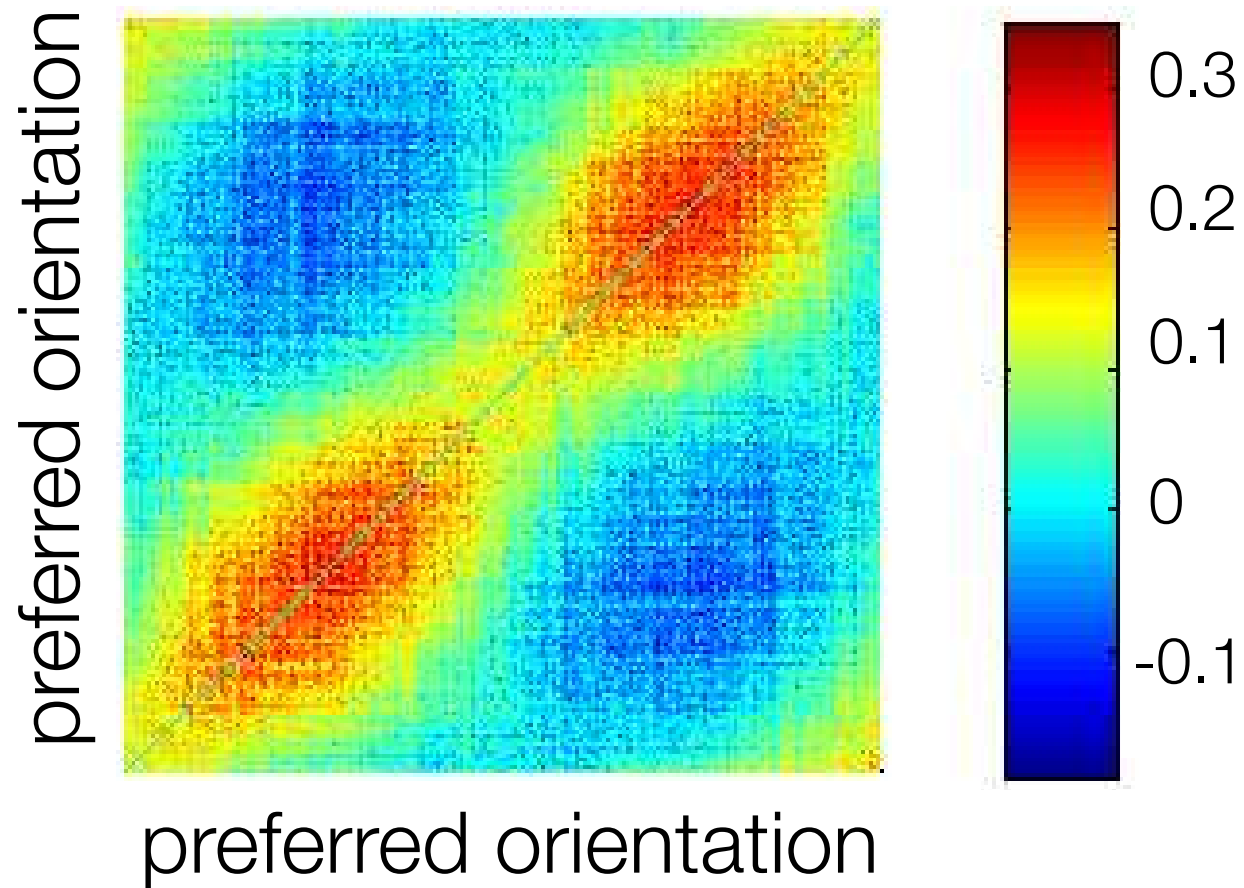
Bondy & Cumming, bioRxiv 2016

Haefner, Berkes & Fiser, Neuron 2016

# Dependencies and observables

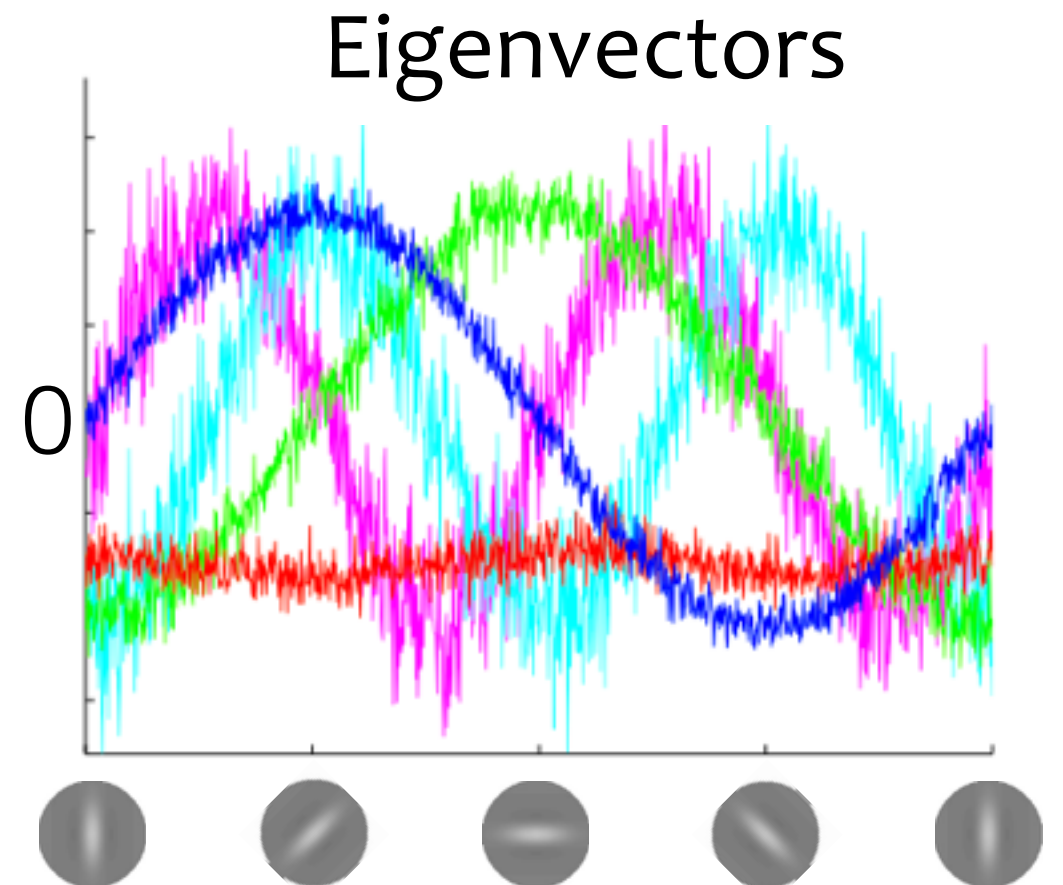
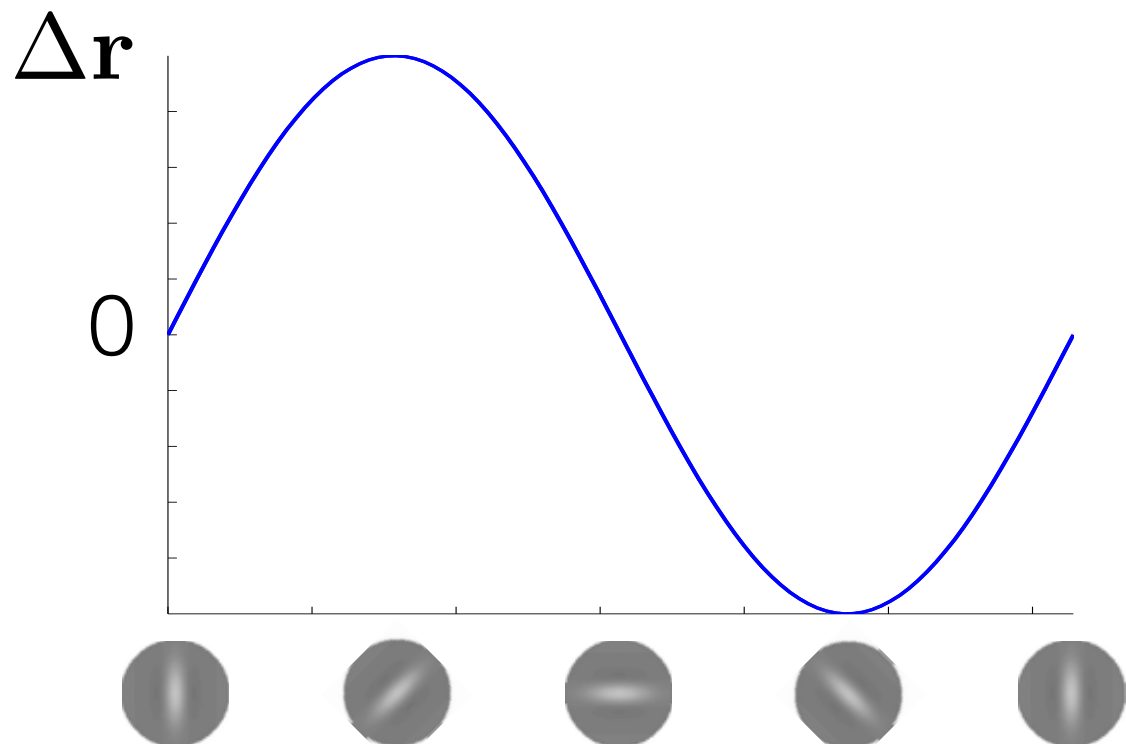
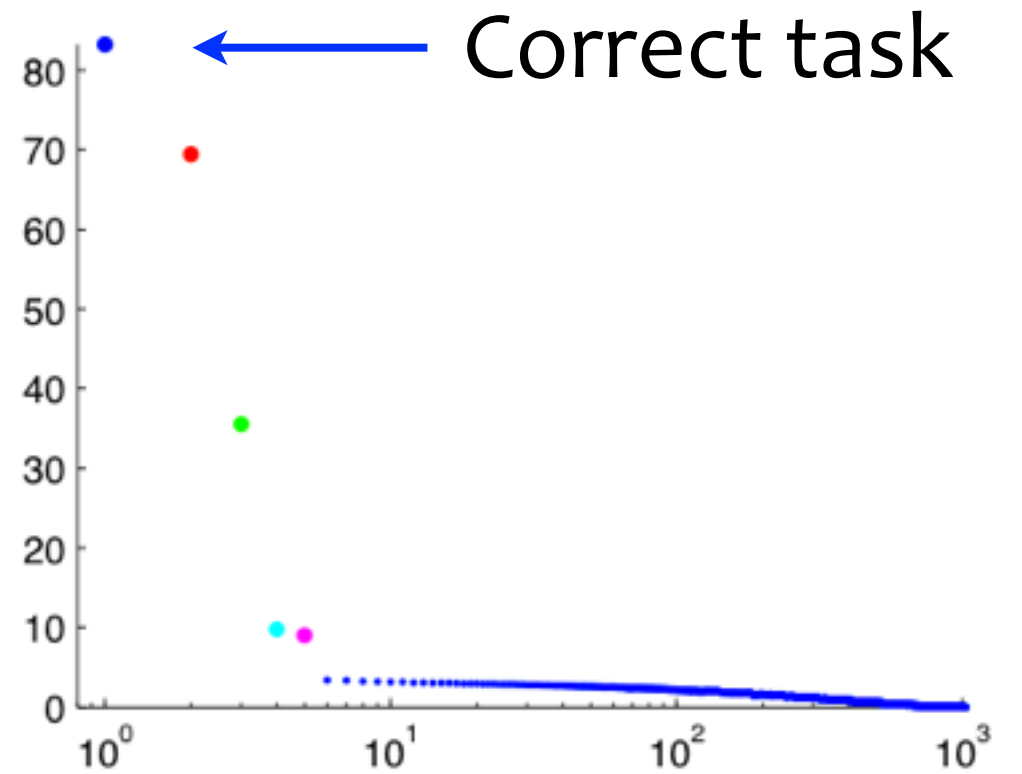
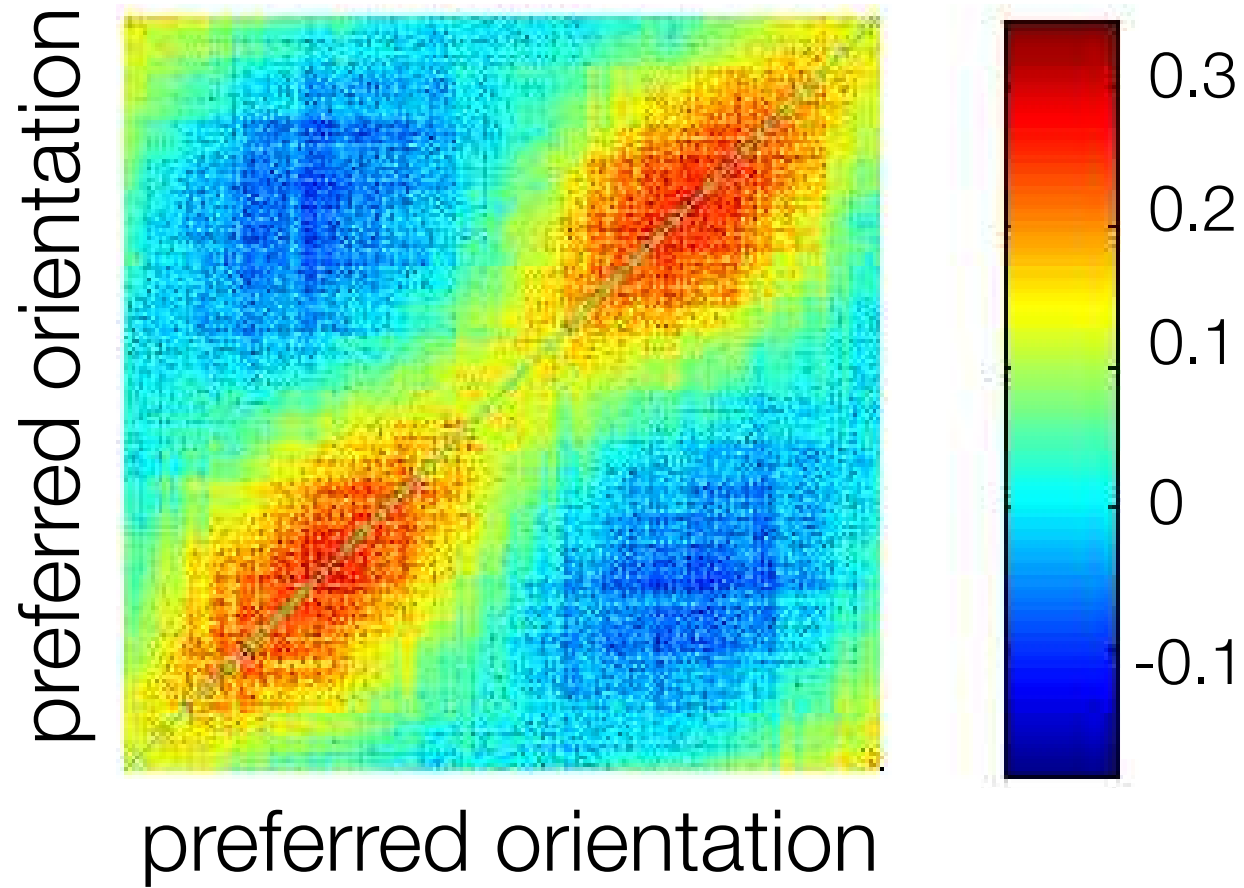


# Model: noise correlations in V1

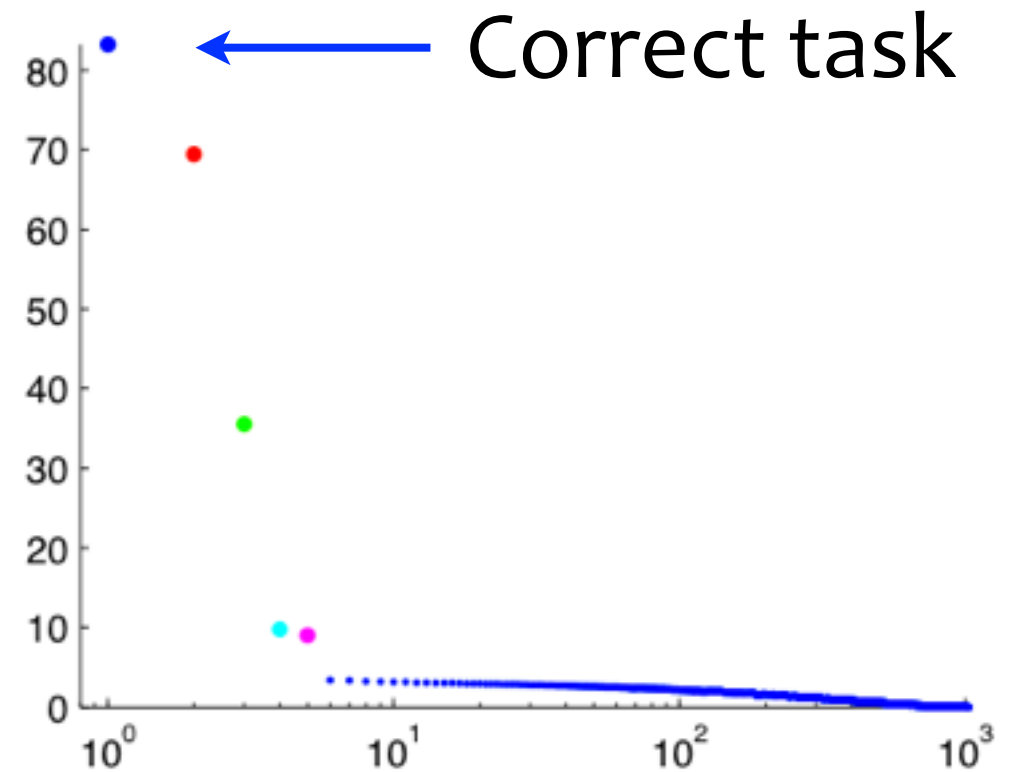
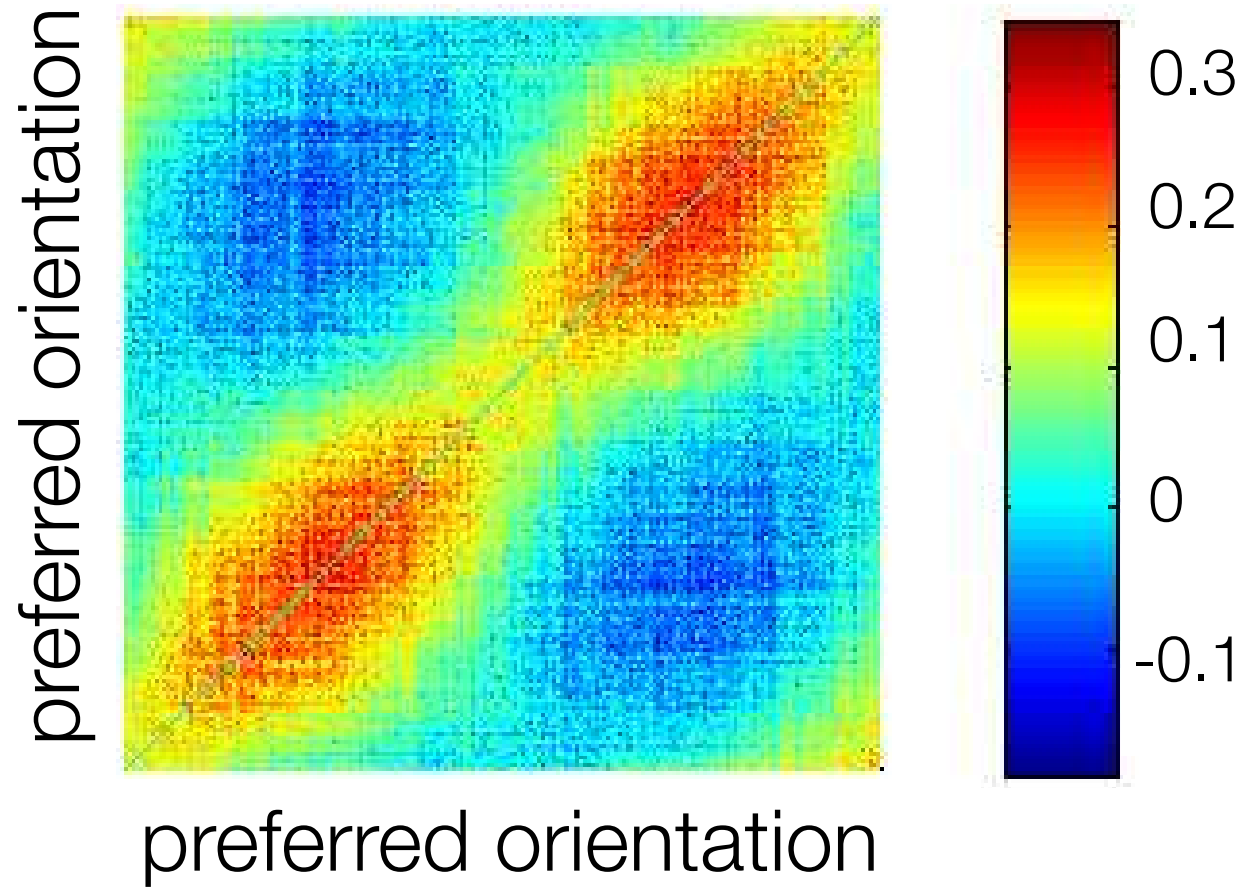




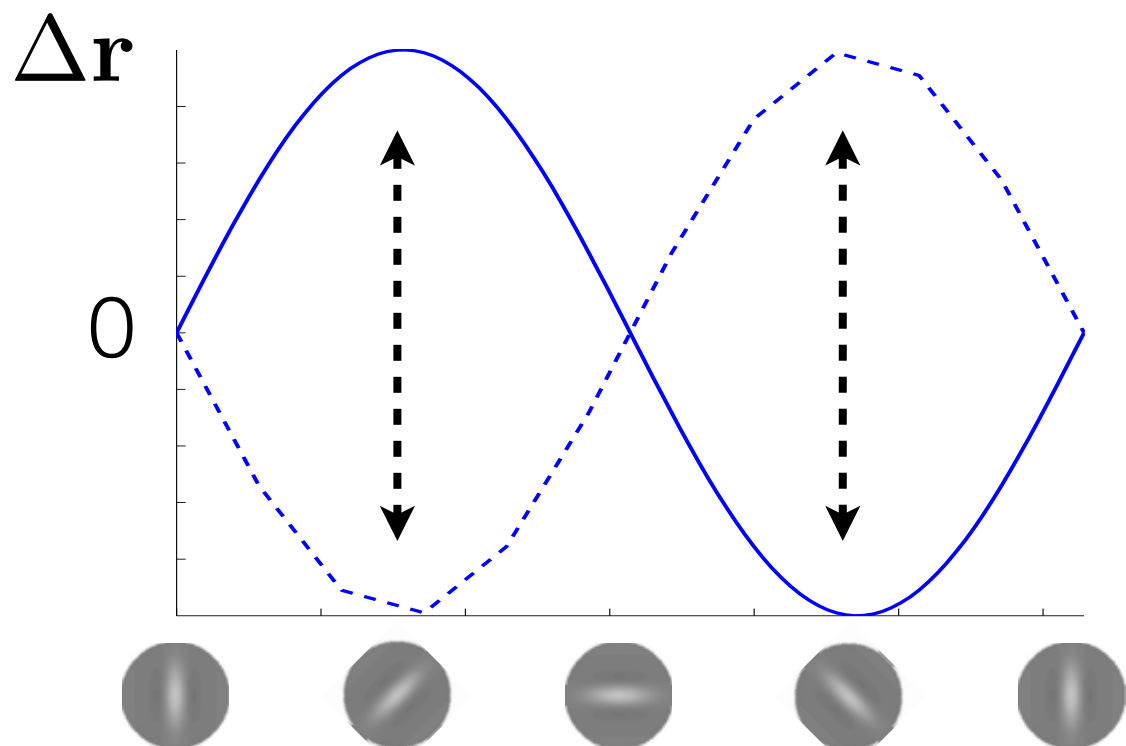
# Model: noise correlations in V1



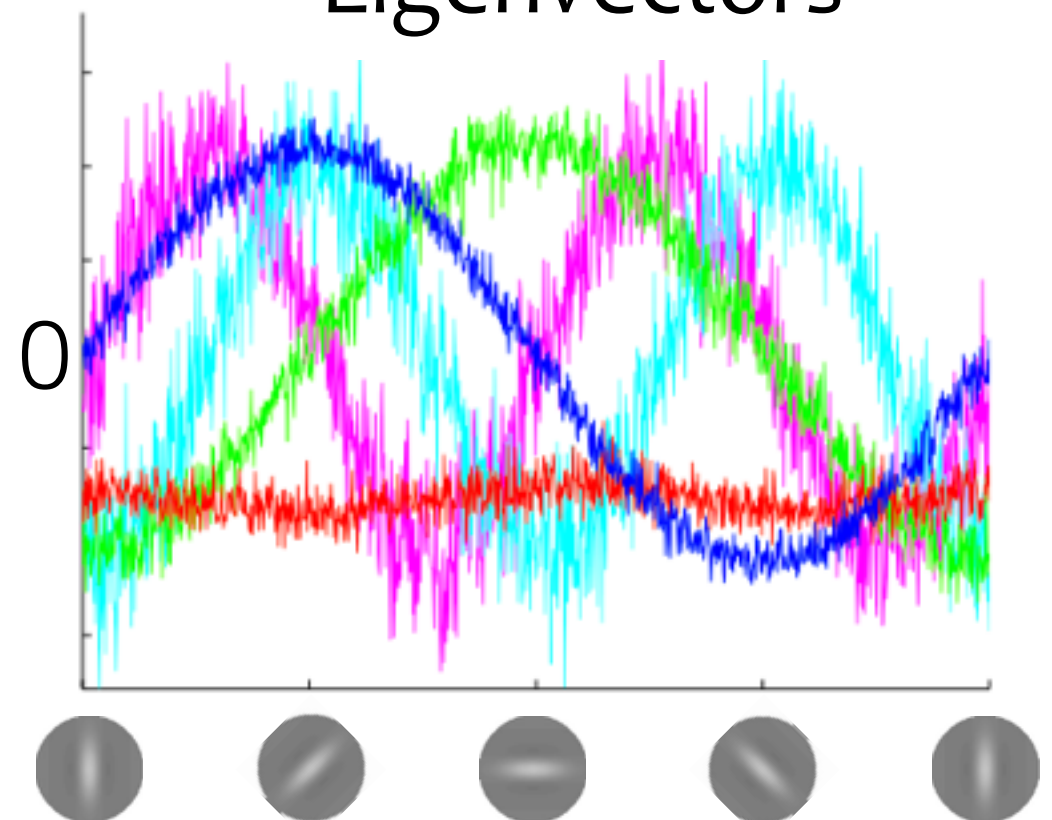
# Model: noise correlations in V1



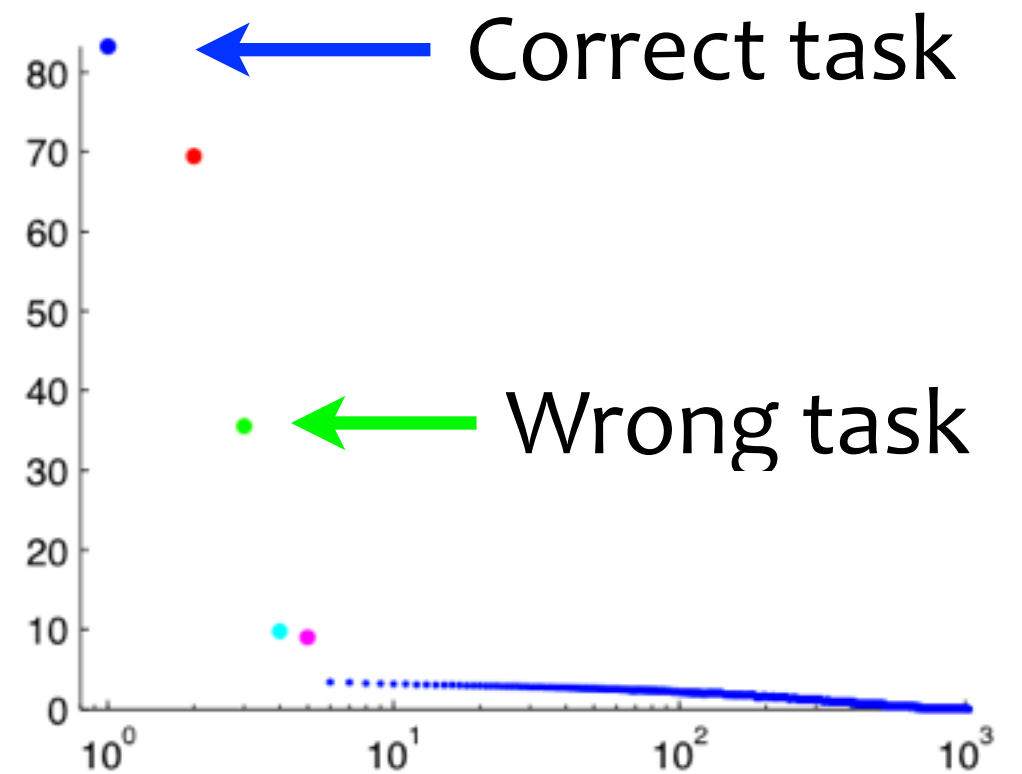
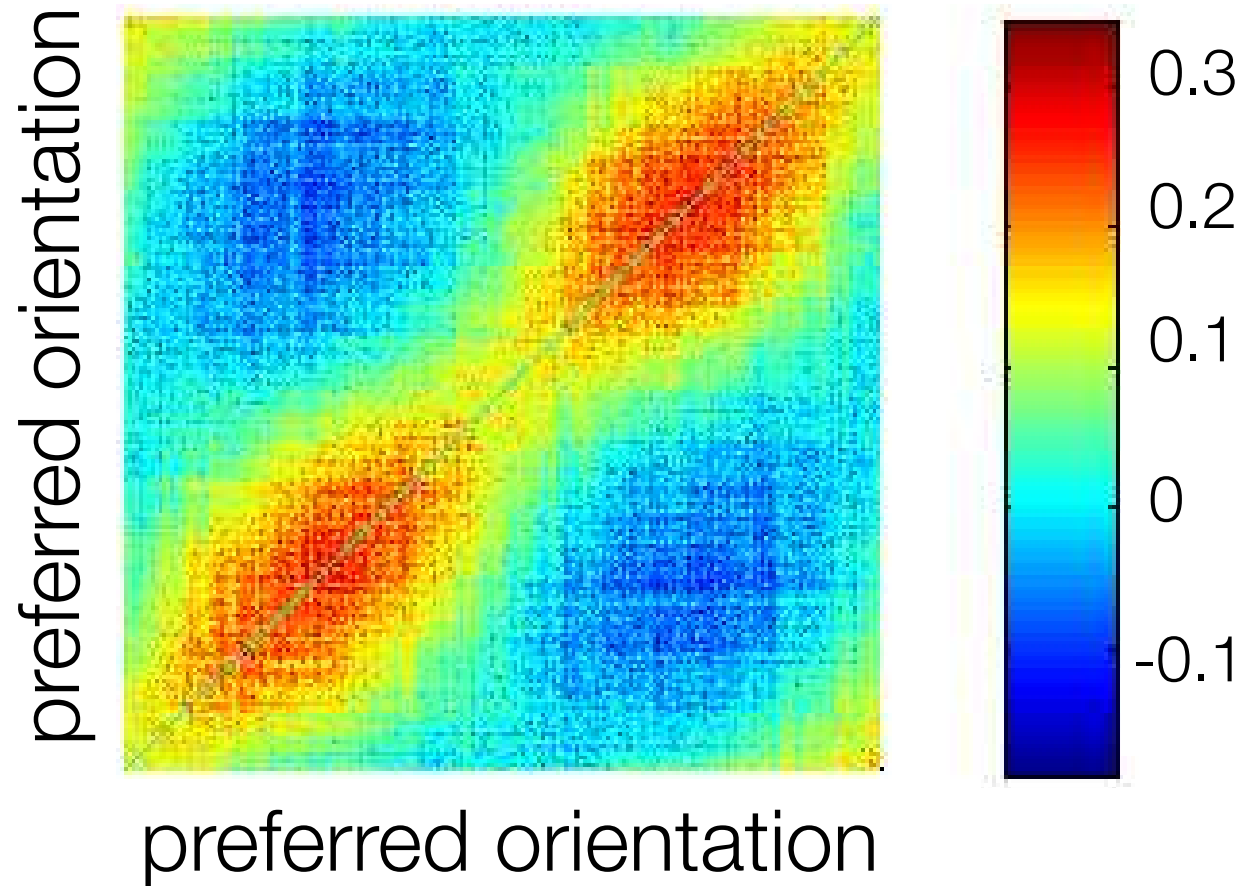
$\mathbf{f}'\mathbf{f}'^\top$  – covariance



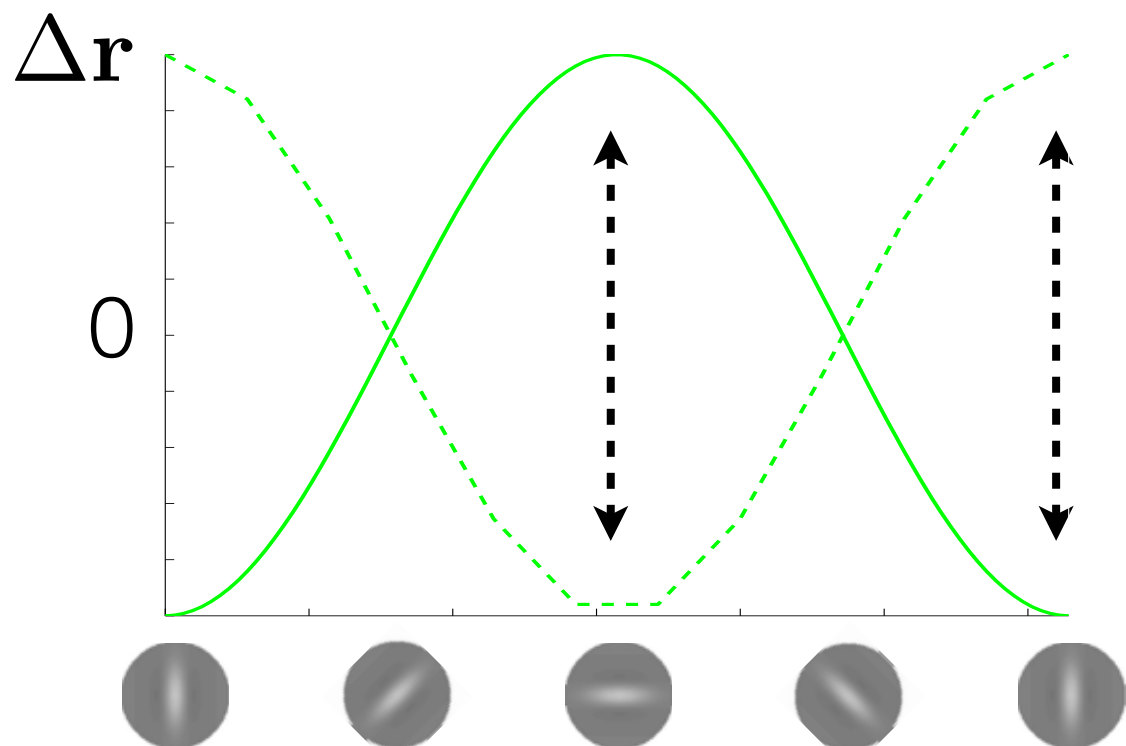
Eigenvectors



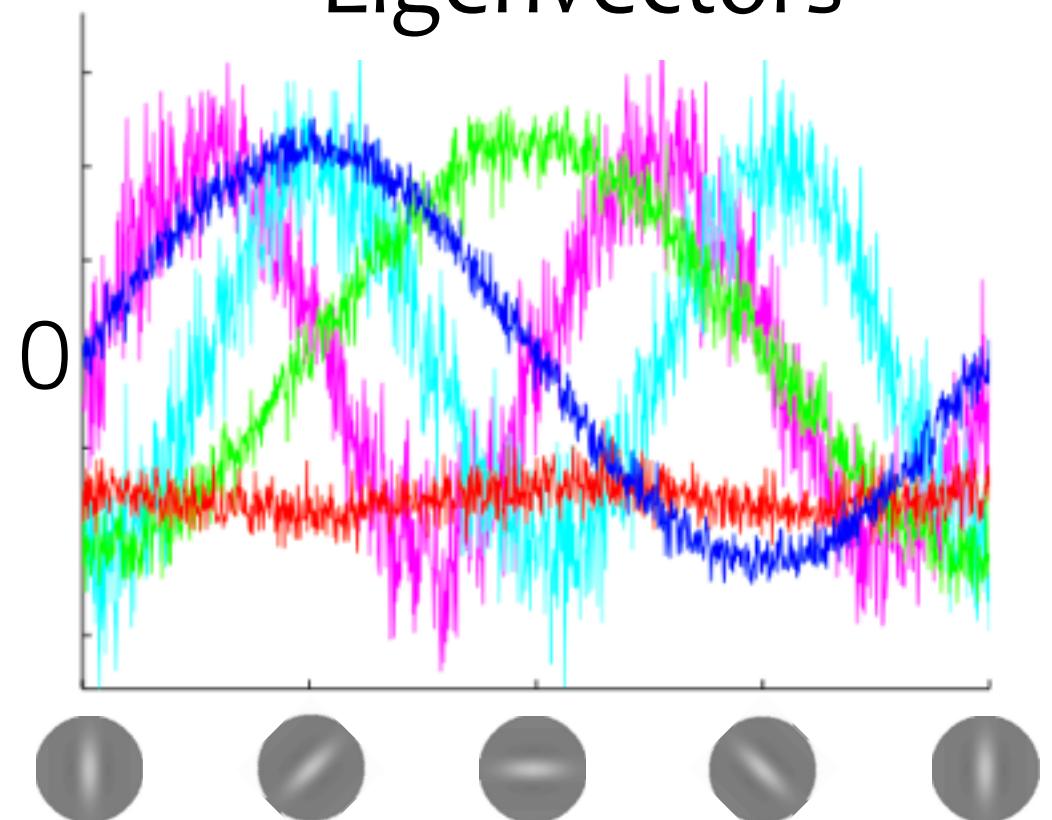
# Model: noise correlations in V1



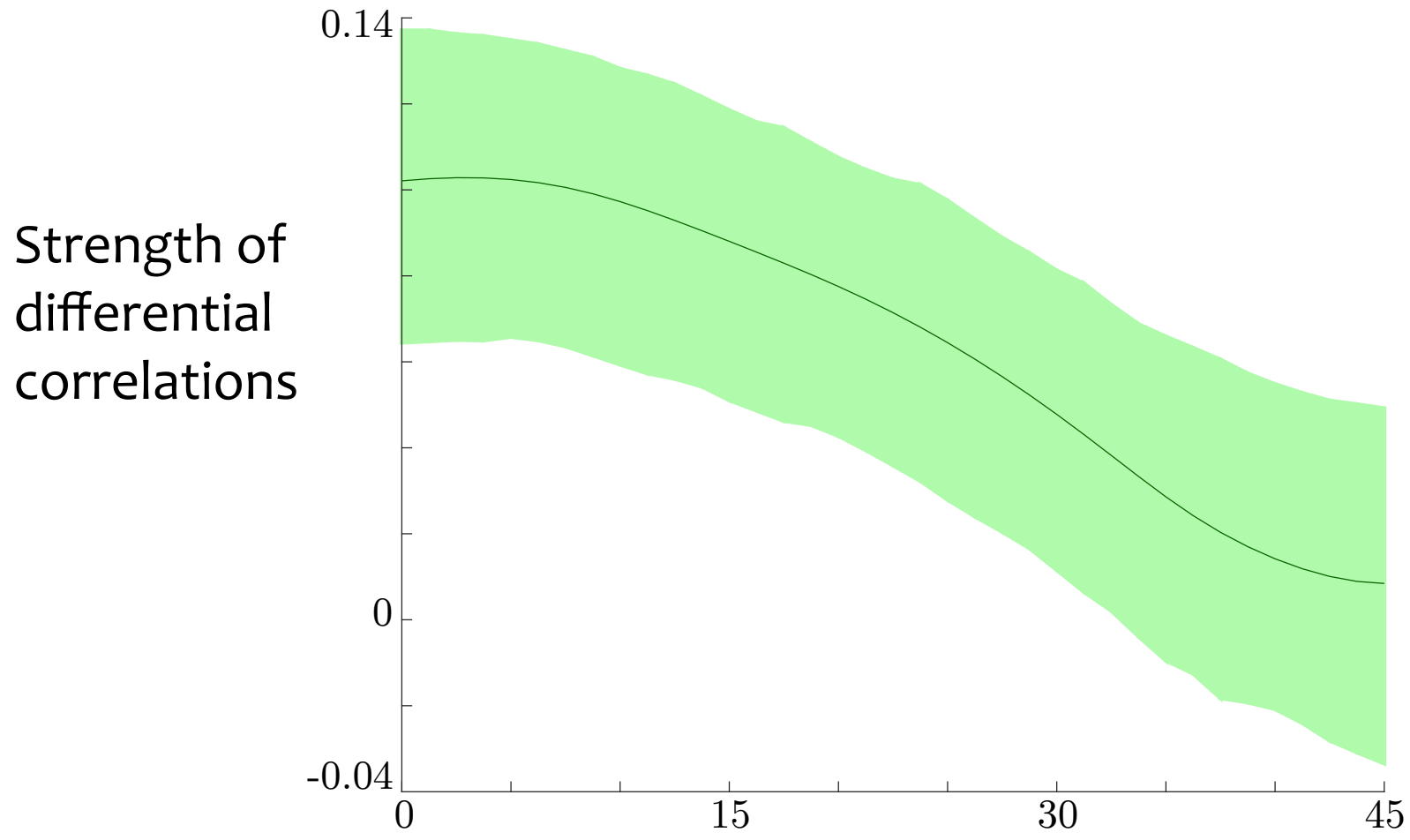
$\mathbf{f}'\mathbf{f}'^T$  – covariance



Eigenvectors

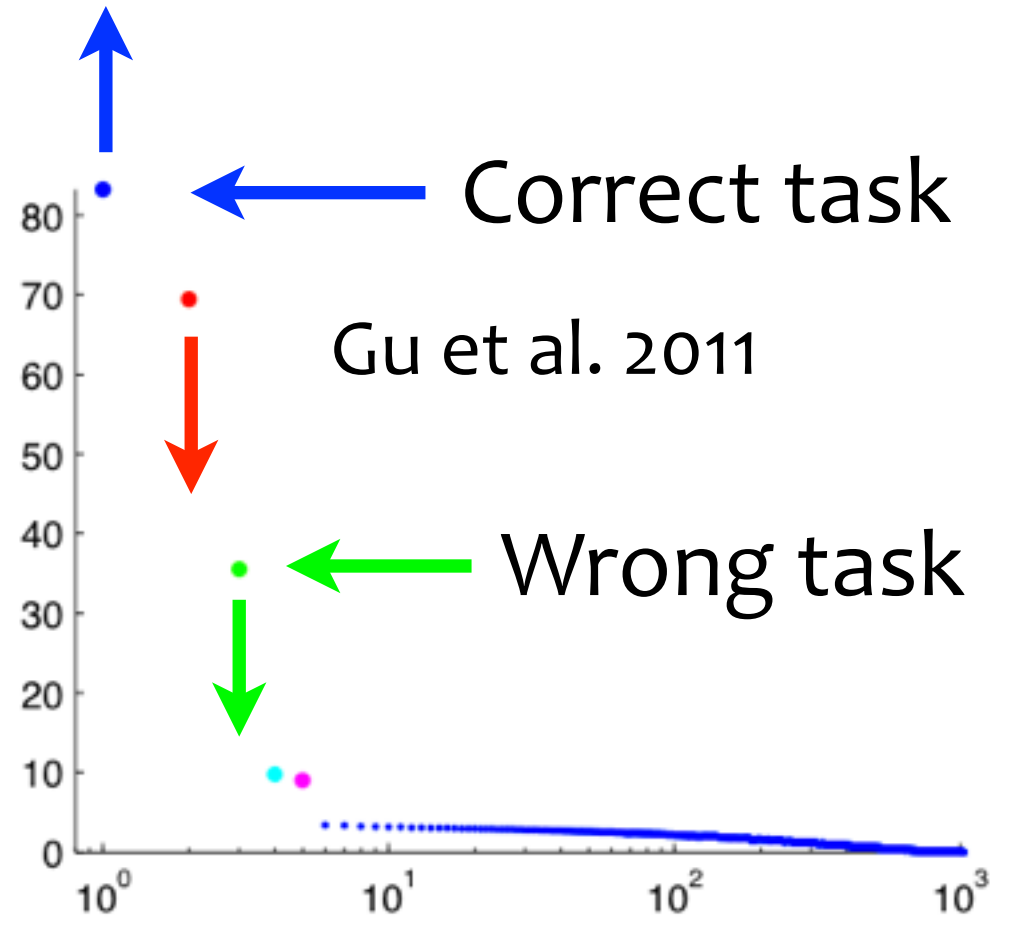
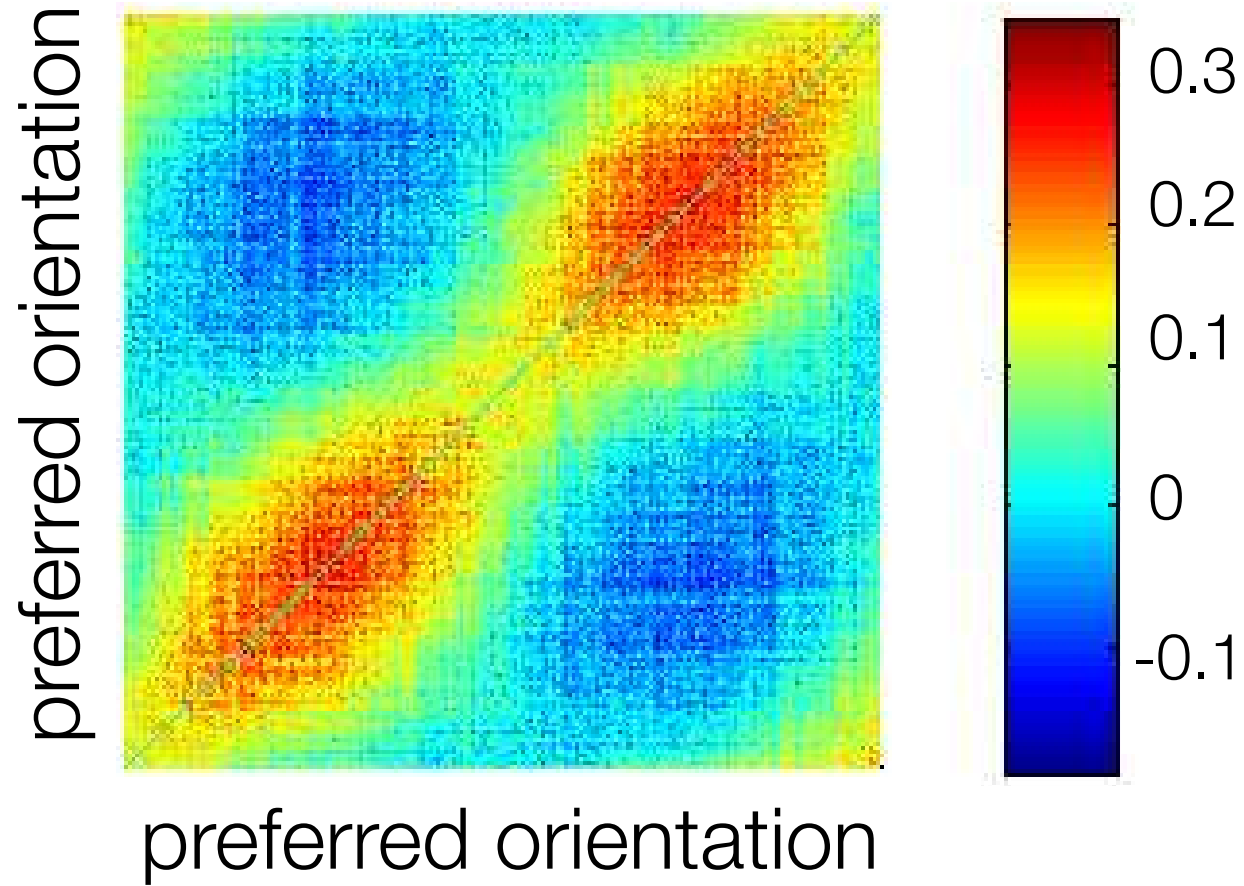


# Differential correlations as a function of task



Difference between actual task performed by monkey and task for which prediction was made

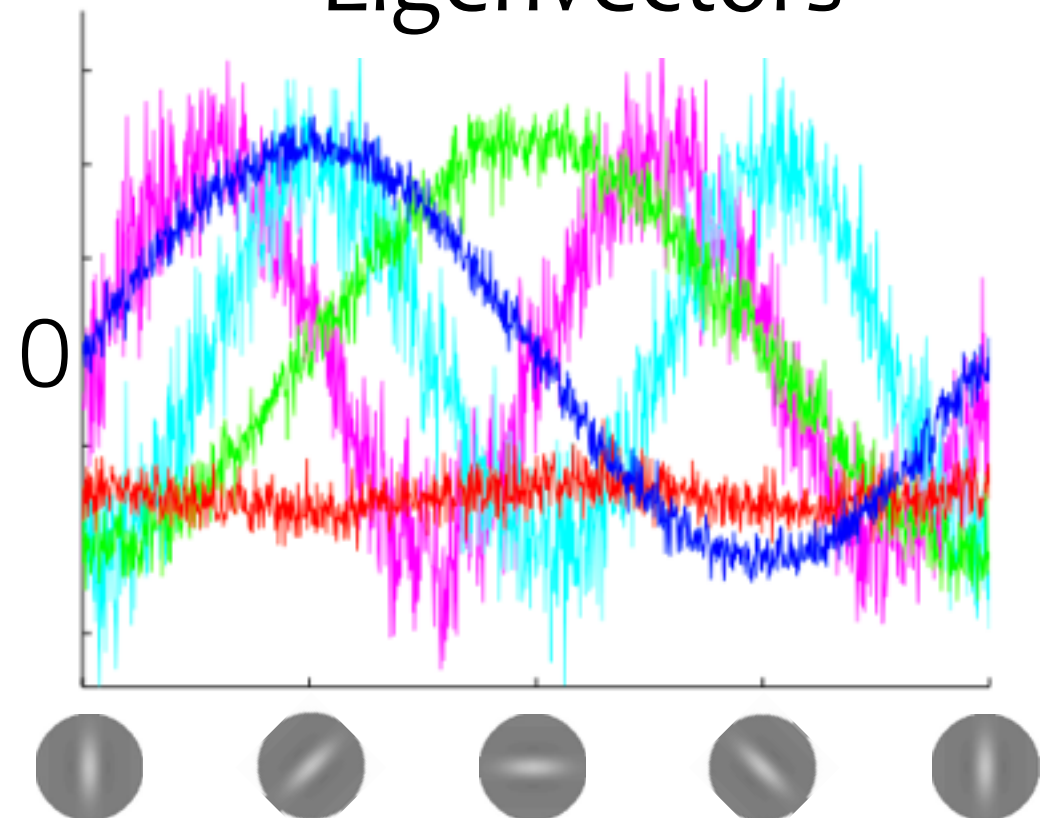
# Learning



## Learning:

- Increasing EV for correct task
- All other EVs decreasing

## Eigenvectors







Camille Gomez-  
Faberge  
(Harvard)



Rick Born  
(Harvard)

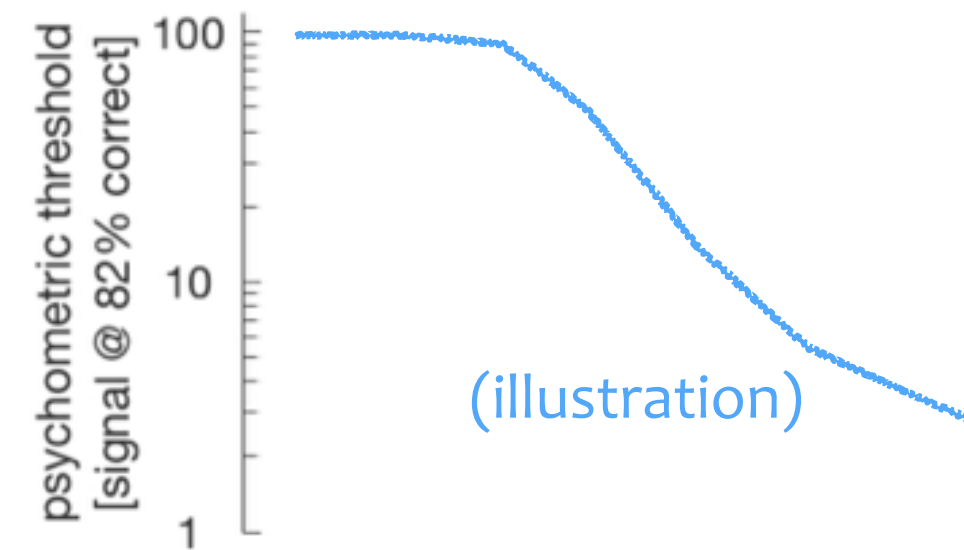


Richard Lange

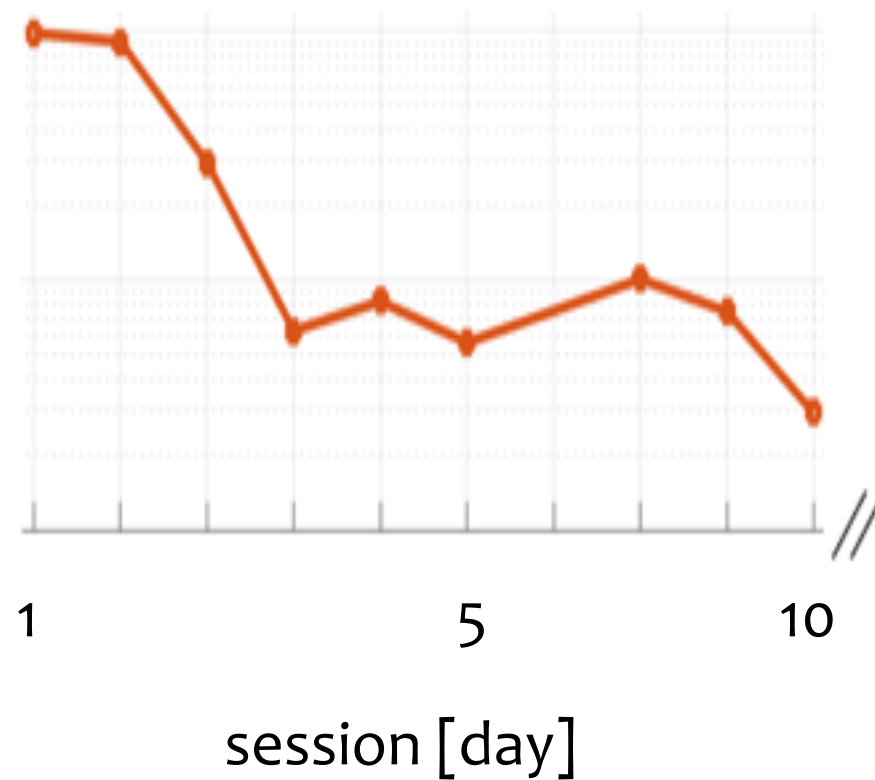
- Learning
- Inference over the task/multi-tasking
- Reversibly inactivate top-down connections to V1 (by cooling V2)

# Preliminary results (1 monkey)

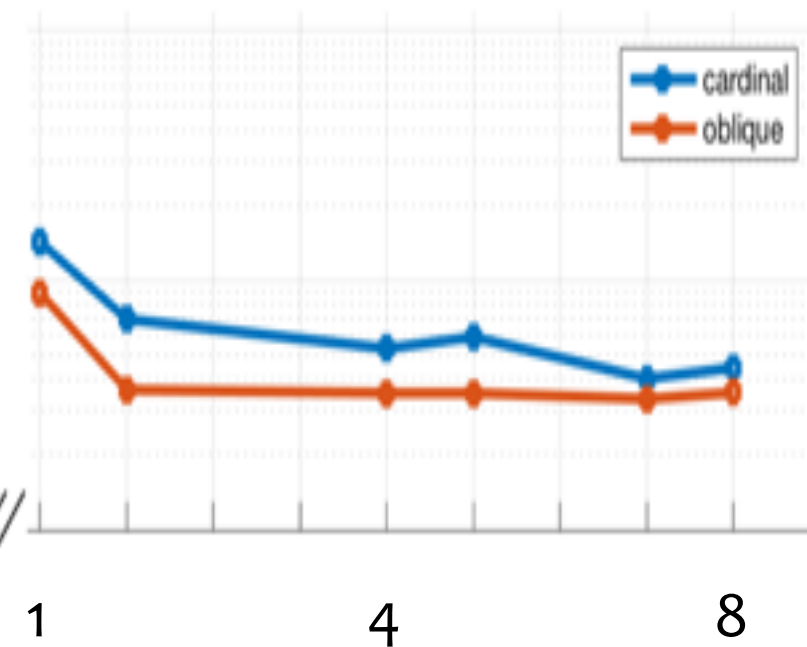
## Cardinal task



## Oblique task

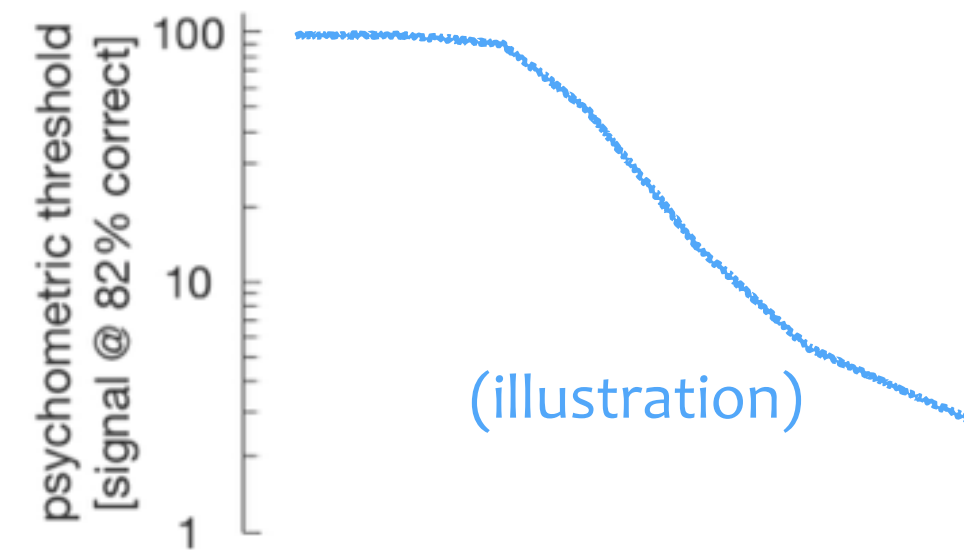


## Interleaved

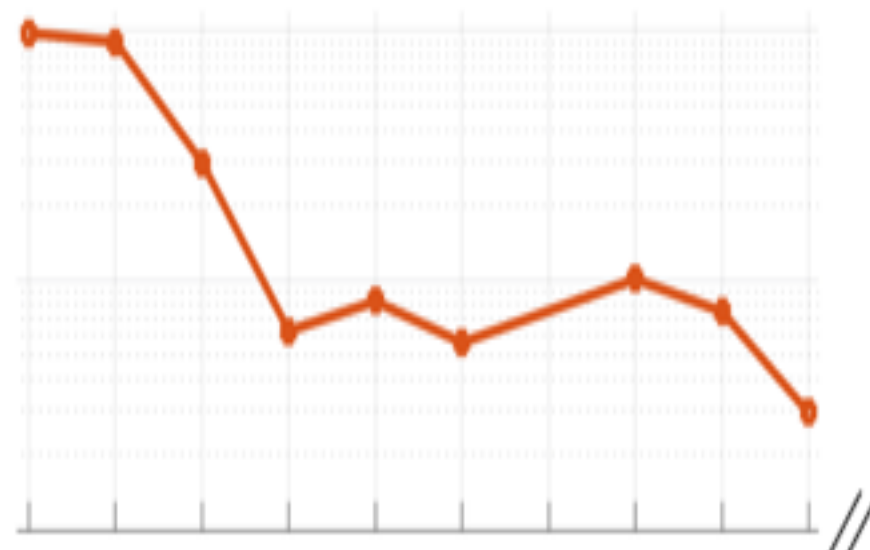


# Preliminary results! (1 monkey)

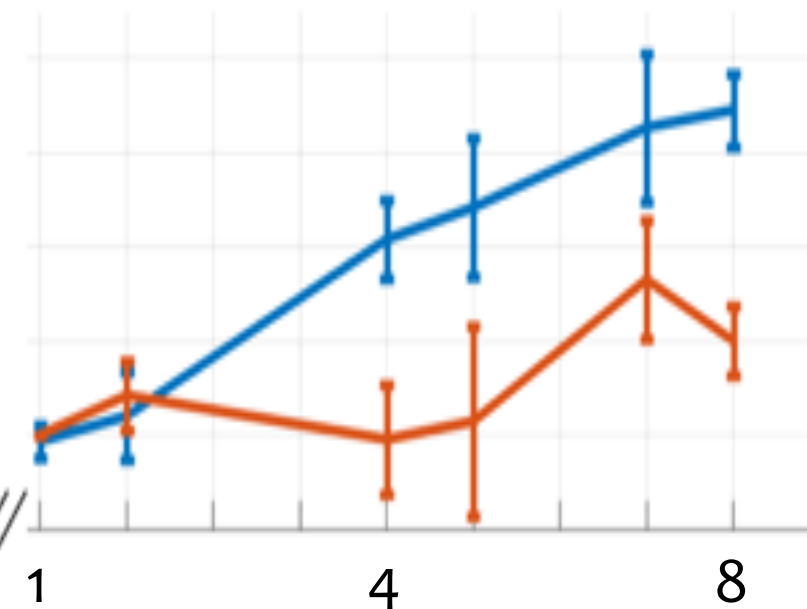
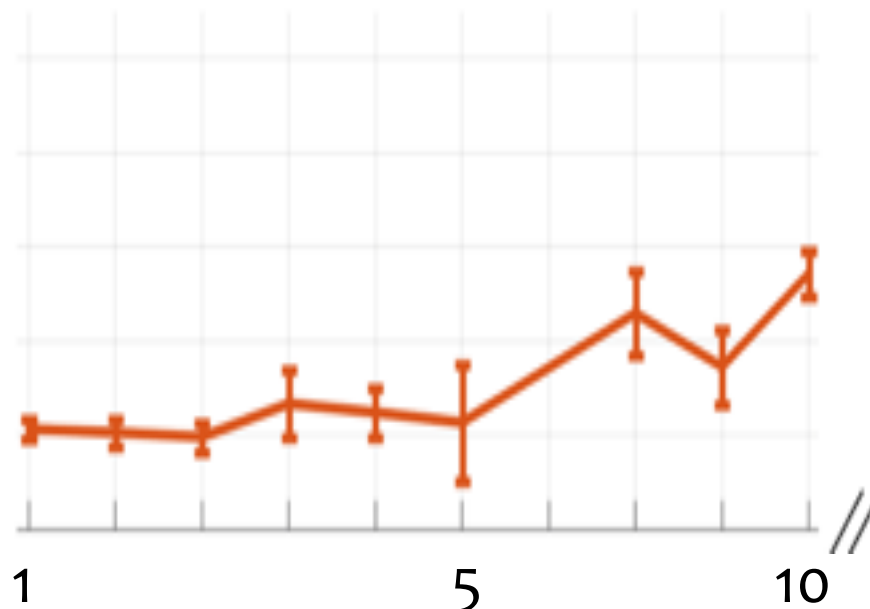
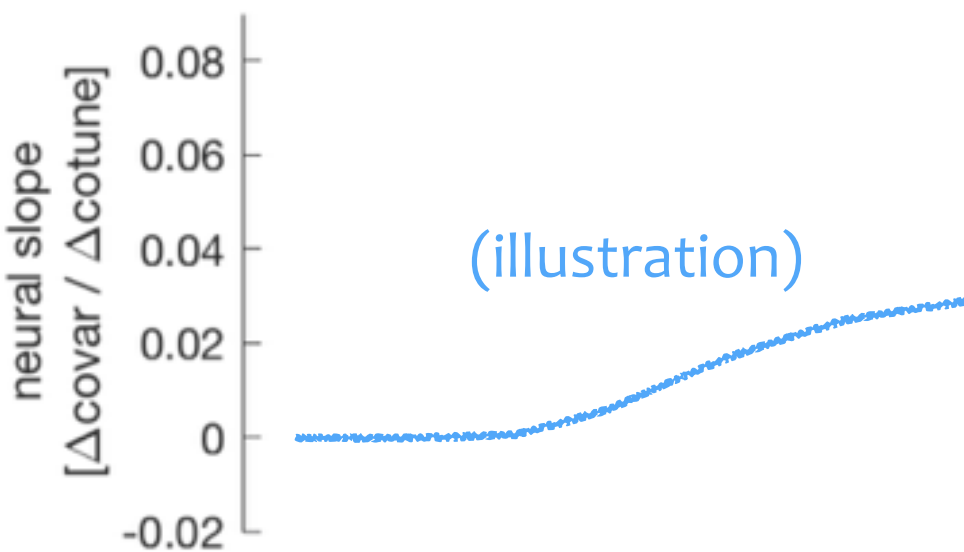
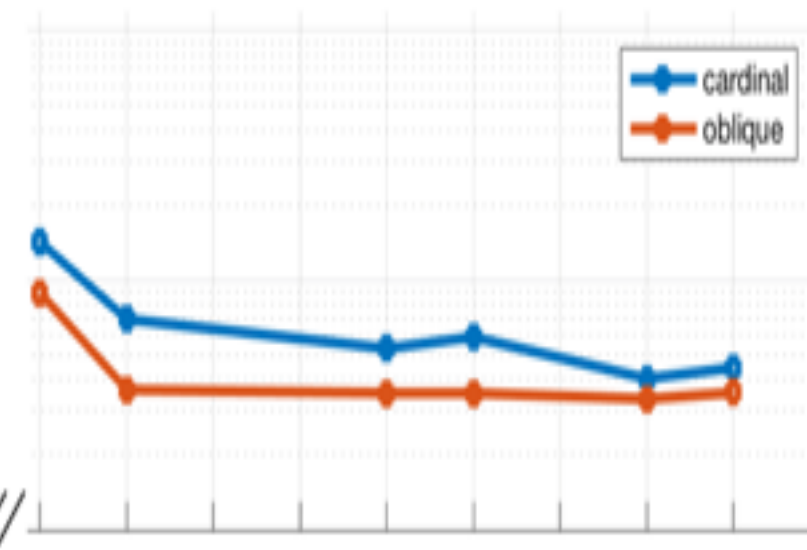
## Cardinal task



## Oblique task



## Interleaved



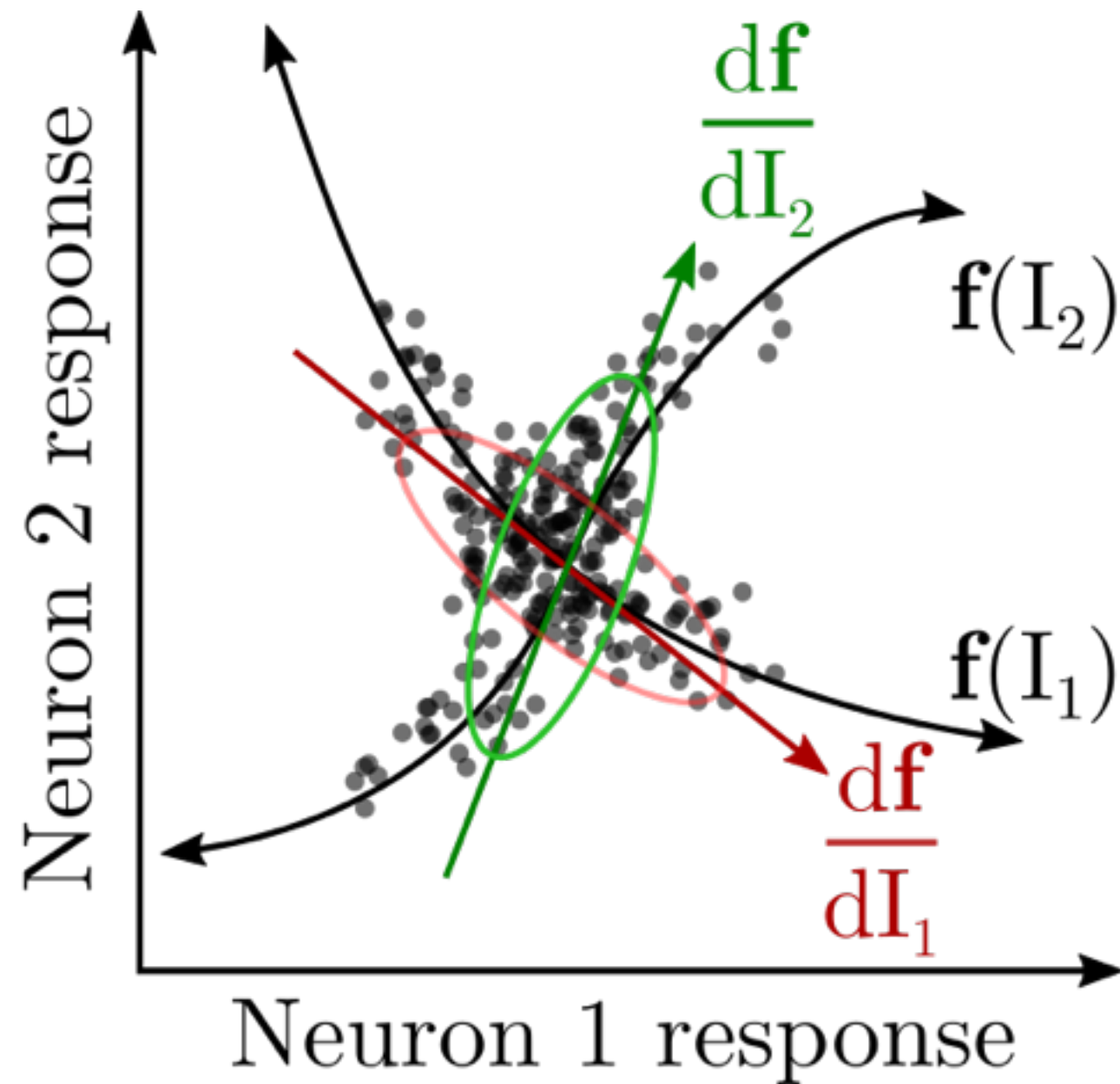
session [day]

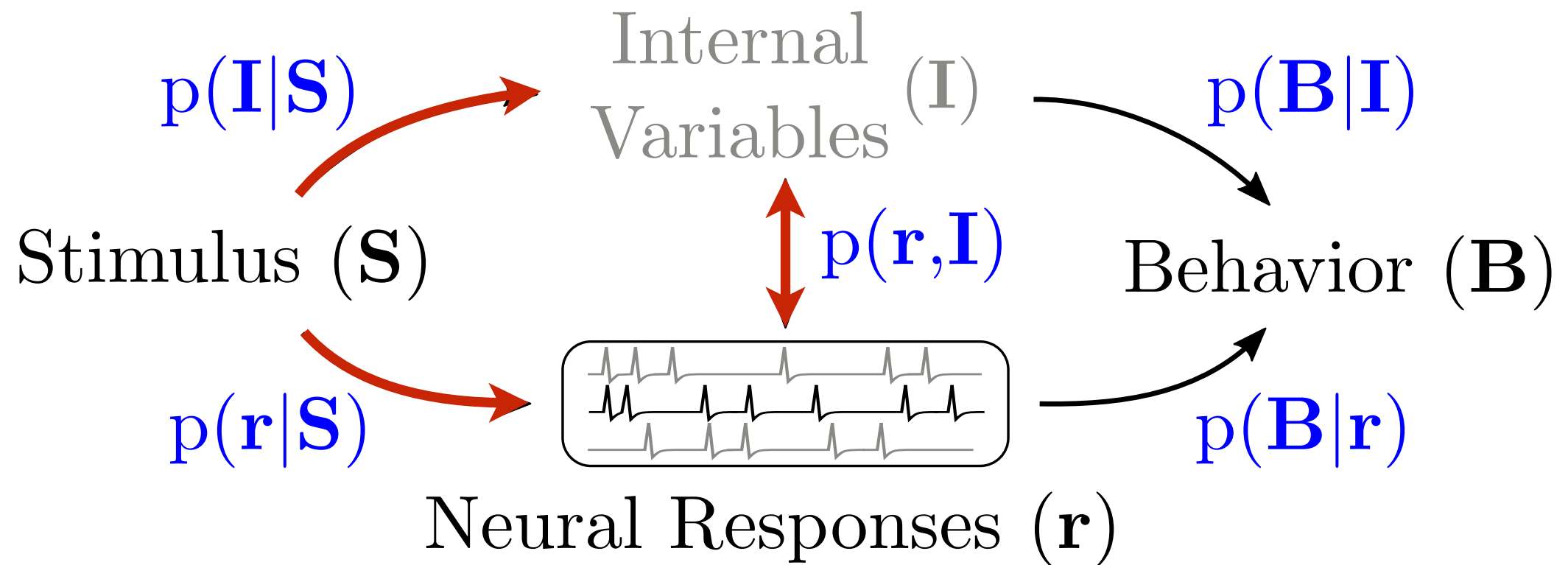


Next steps: compare correlations & cooling data



# Variability in internal states induces statistical signature in neural population responses





- knowledge of relationships  
stimulus–responses & responses–internal variables

► infer relationships: stimulus–internal variables

# Conclusions so far

- Probabilistic inference by neural sampling makes strong predictions about the task-dependence of neural correlations
- Data so far confirm those predictions
- Population recordings let us reverse-engineer internal beliefs
  - e.g. track them over learning
- We can interpret them in stimulus space
- Sampling? -> time scales

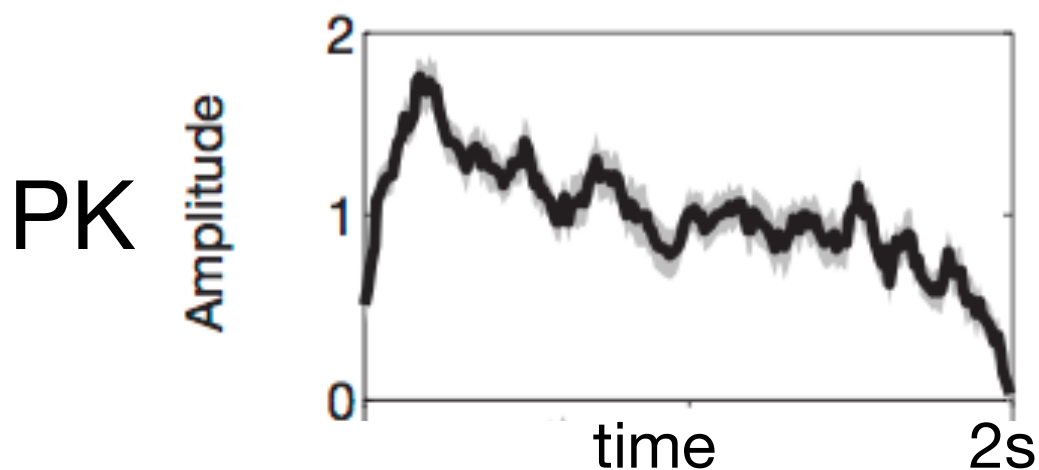
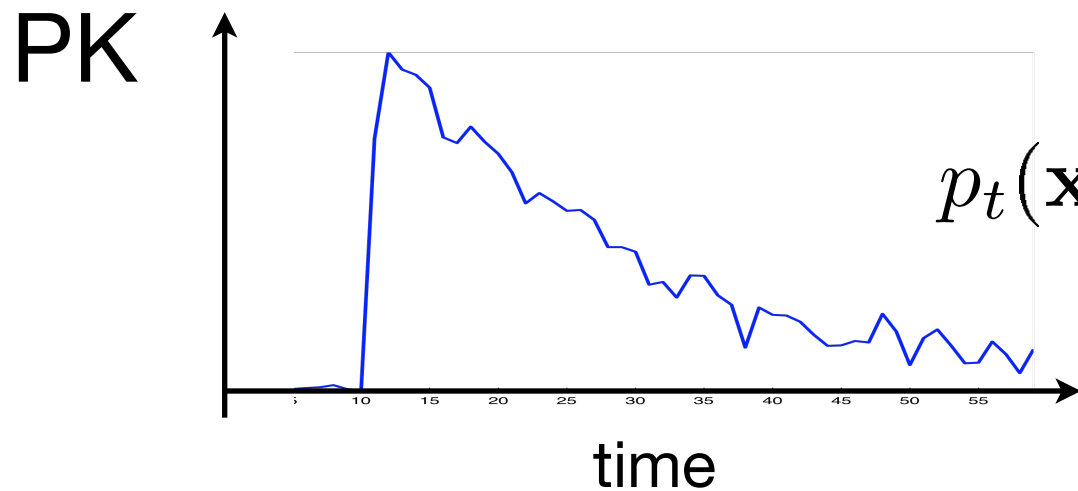
# Decreasing psycho-physical kernel (PK)

## Confirmation bias:

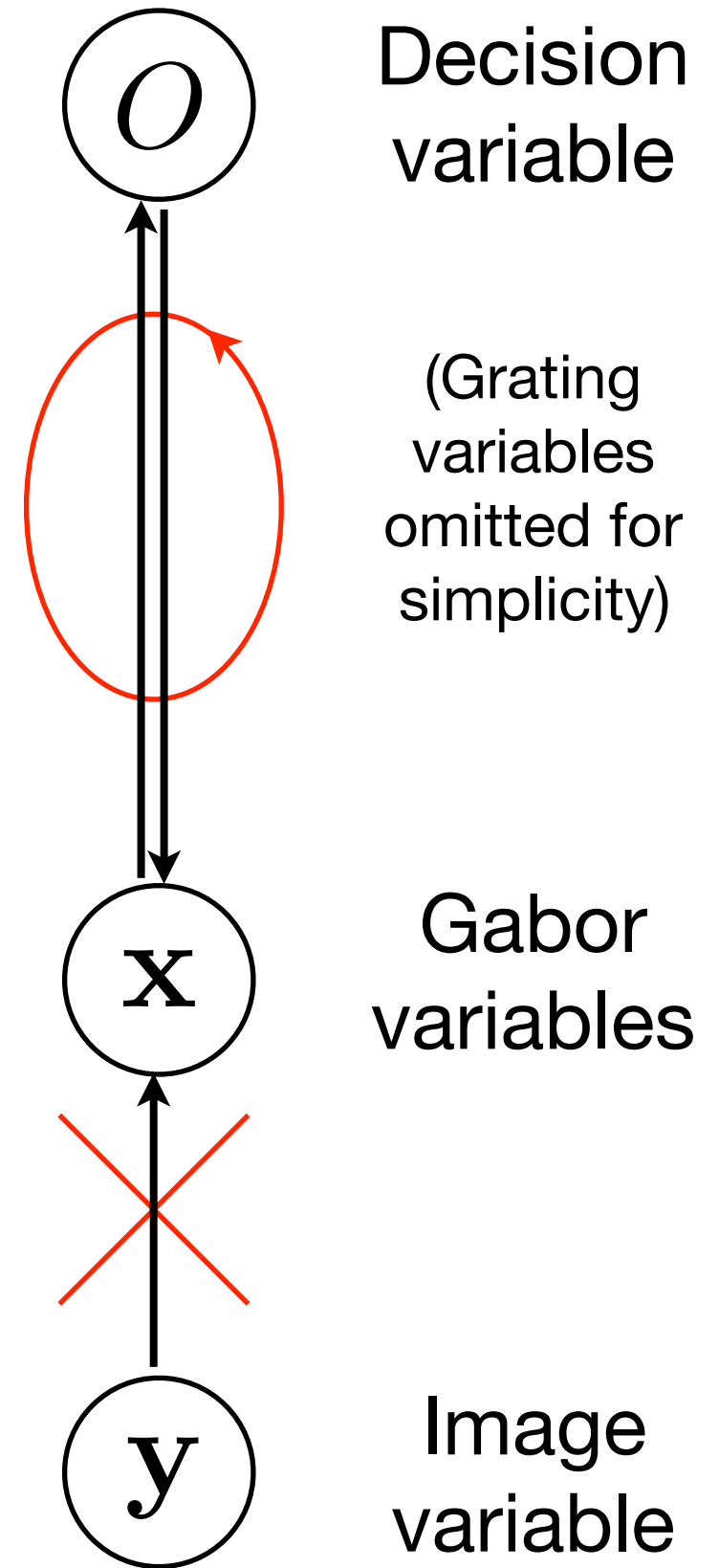
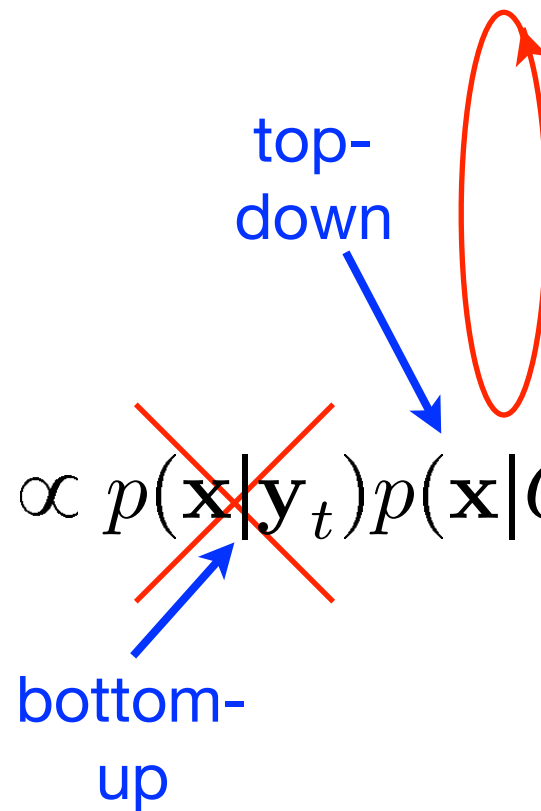
- One sample at a time
- Decision based on inferred, not directly observed variables

$$p_t(O) \propto p_{t-1}(O)p(O|\mathbf{x}_t)$$

$$p_t(\mathbf{x}) \propto p(\mathbf{x}|\mathbf{y}_t)p(\mathbf{x}|O_t)$$



Nienborg & Cumming, Nature 2009

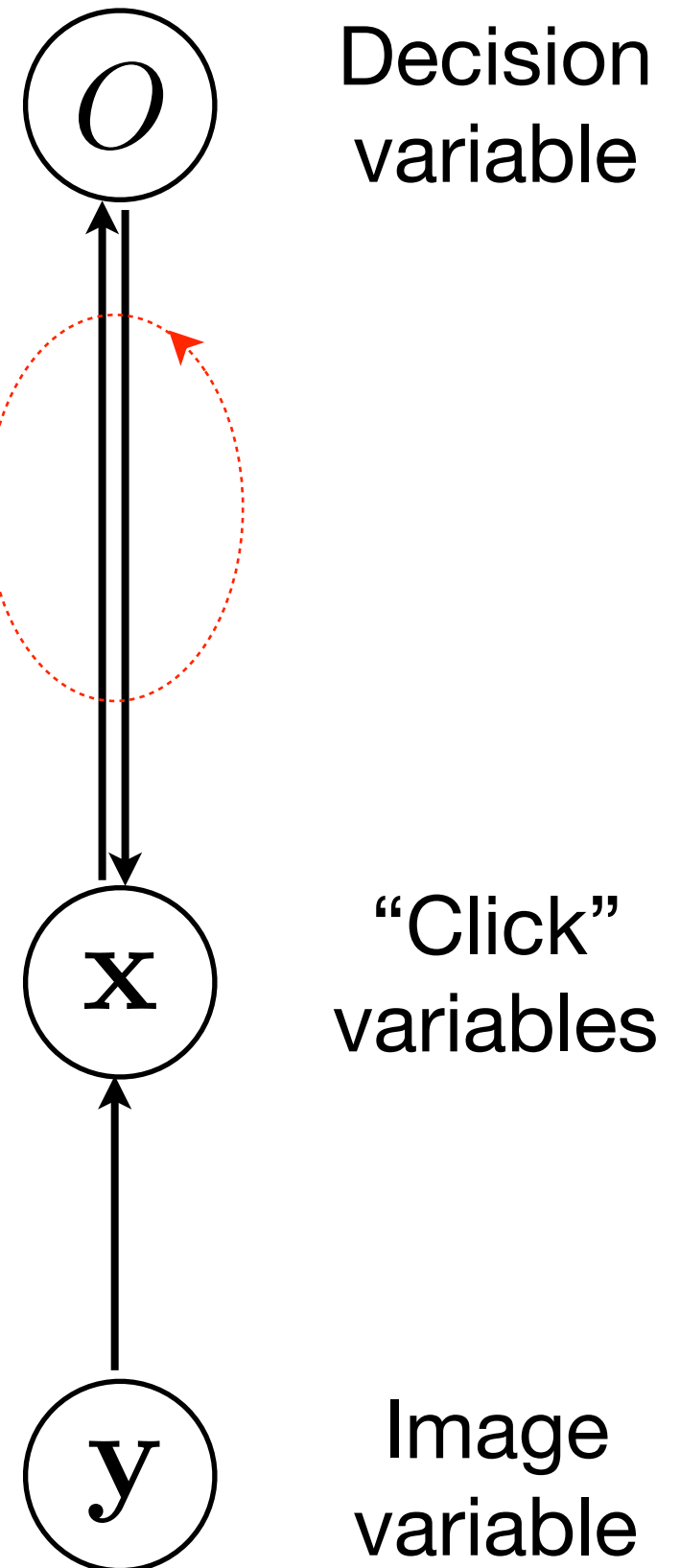


# Why constant psycho-physical kernel?

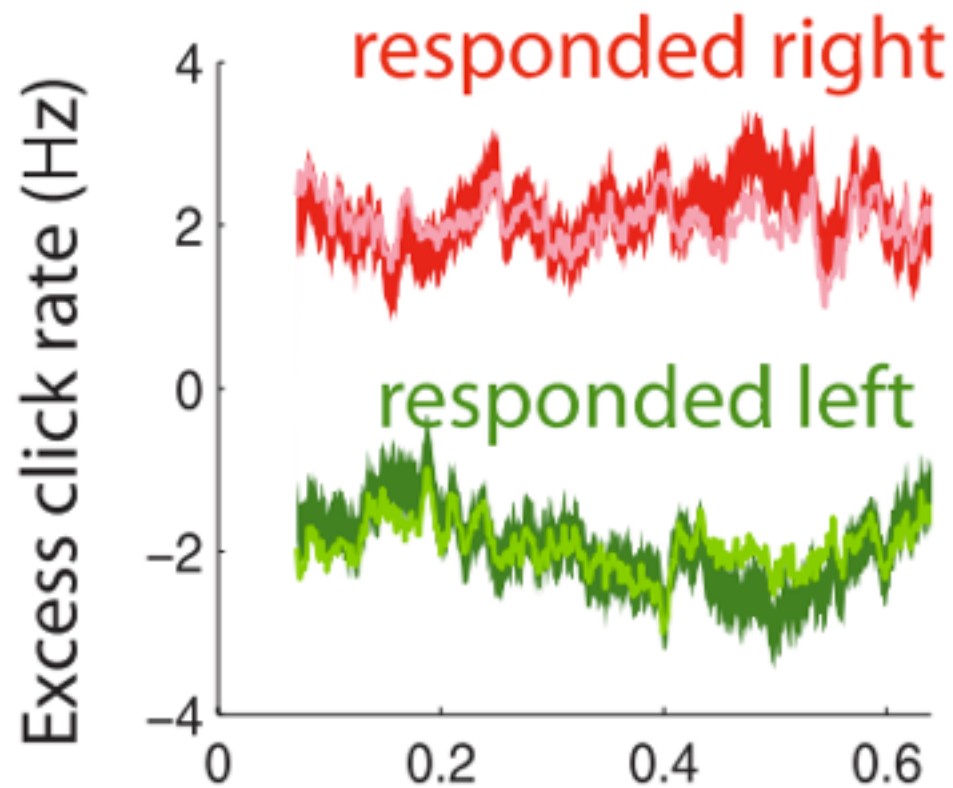
$$p_t(O) \propto p_{t-1}(O)p(O|\mathbf{x}_t)$$

$$p_t(\mathbf{x}) \propto p(\mathbf{x}|\mathbf{y}_t)p(\mathbf{x}|O_t)$$

Each click far above threshold, i.e. likelihood dominates prior.



Poisson click task



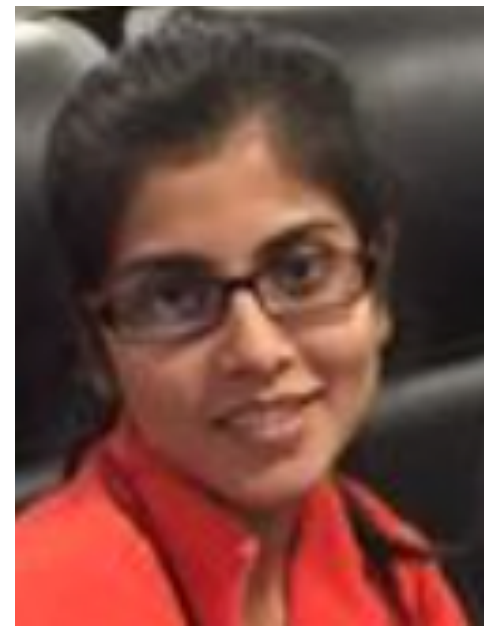
Brunton et al., Science 2013

*Prediction:* Soft clicks  $\rightarrow$  decreasing PK

# Confirmation bias project

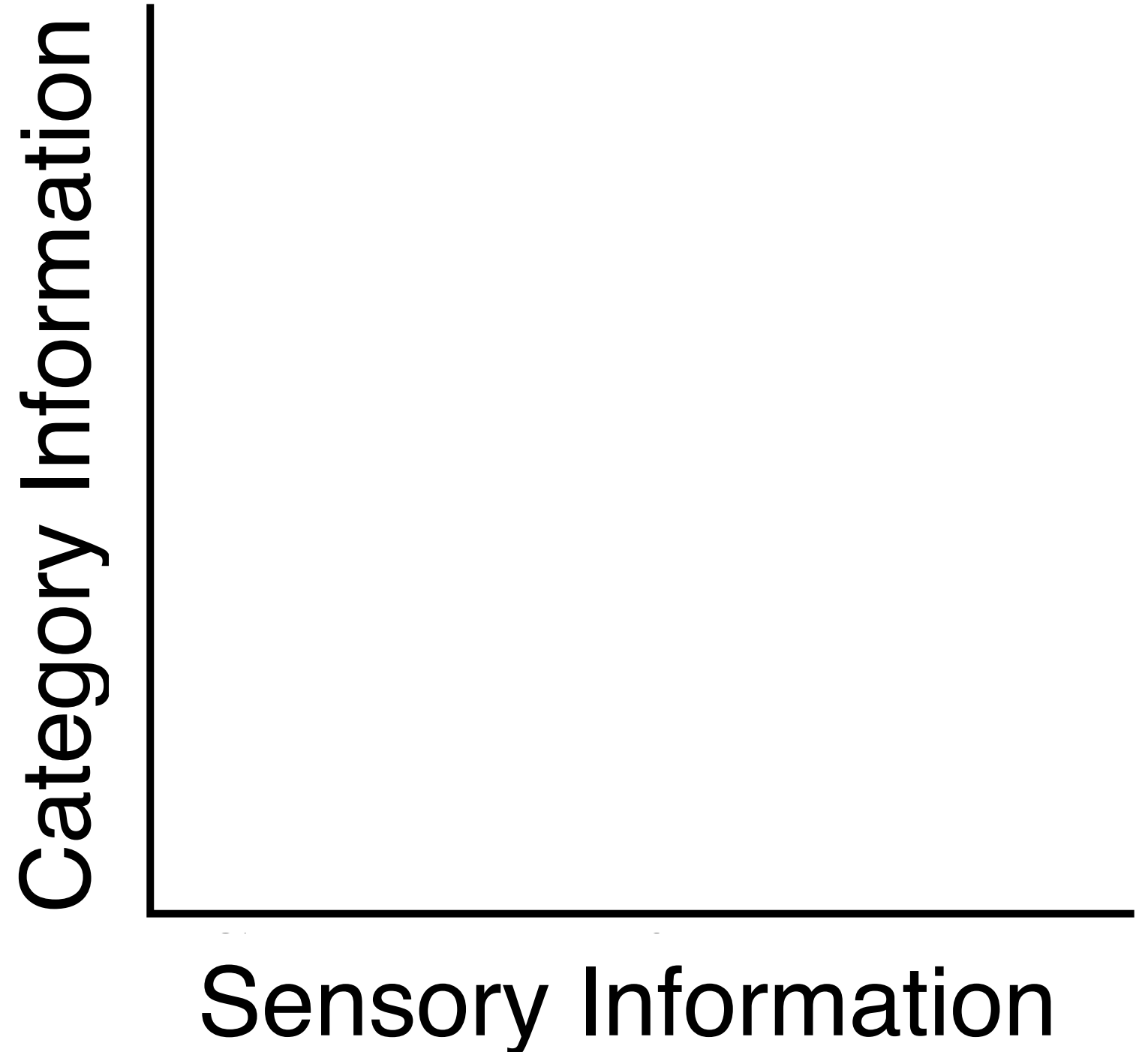
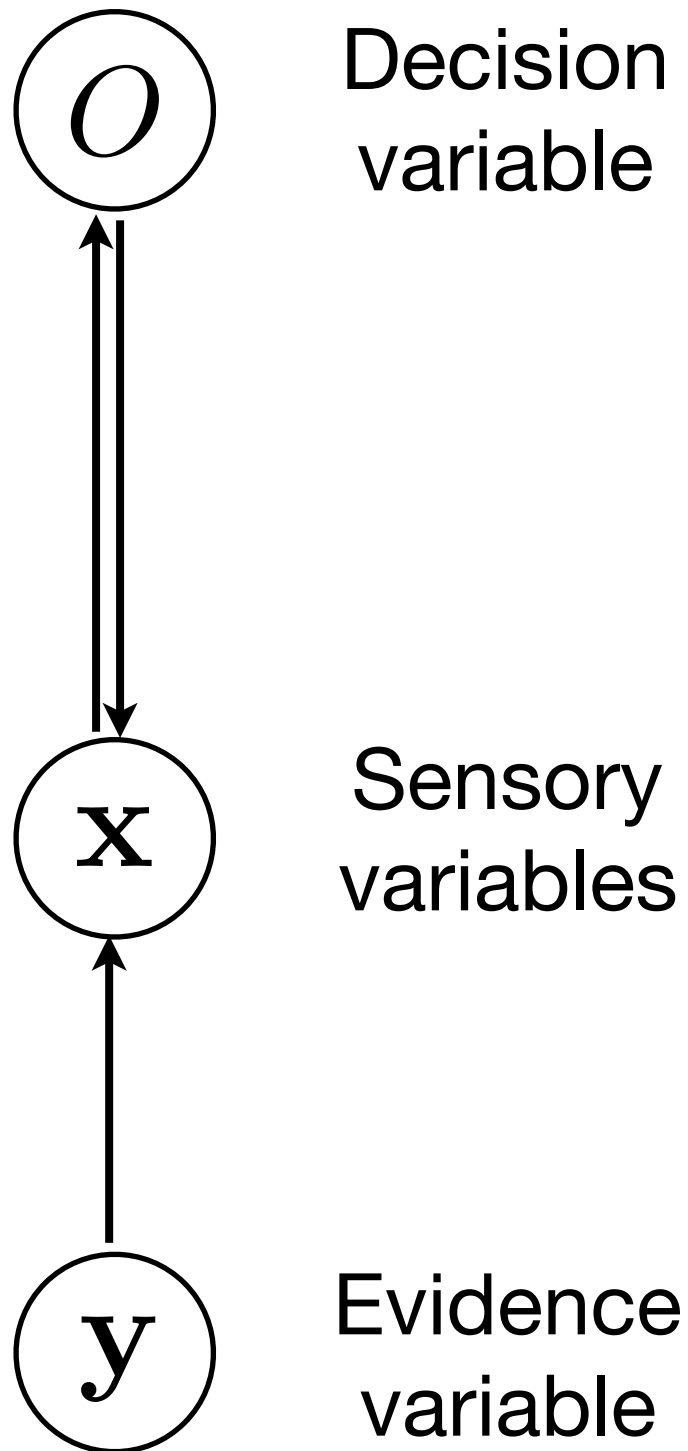


Richard Lange

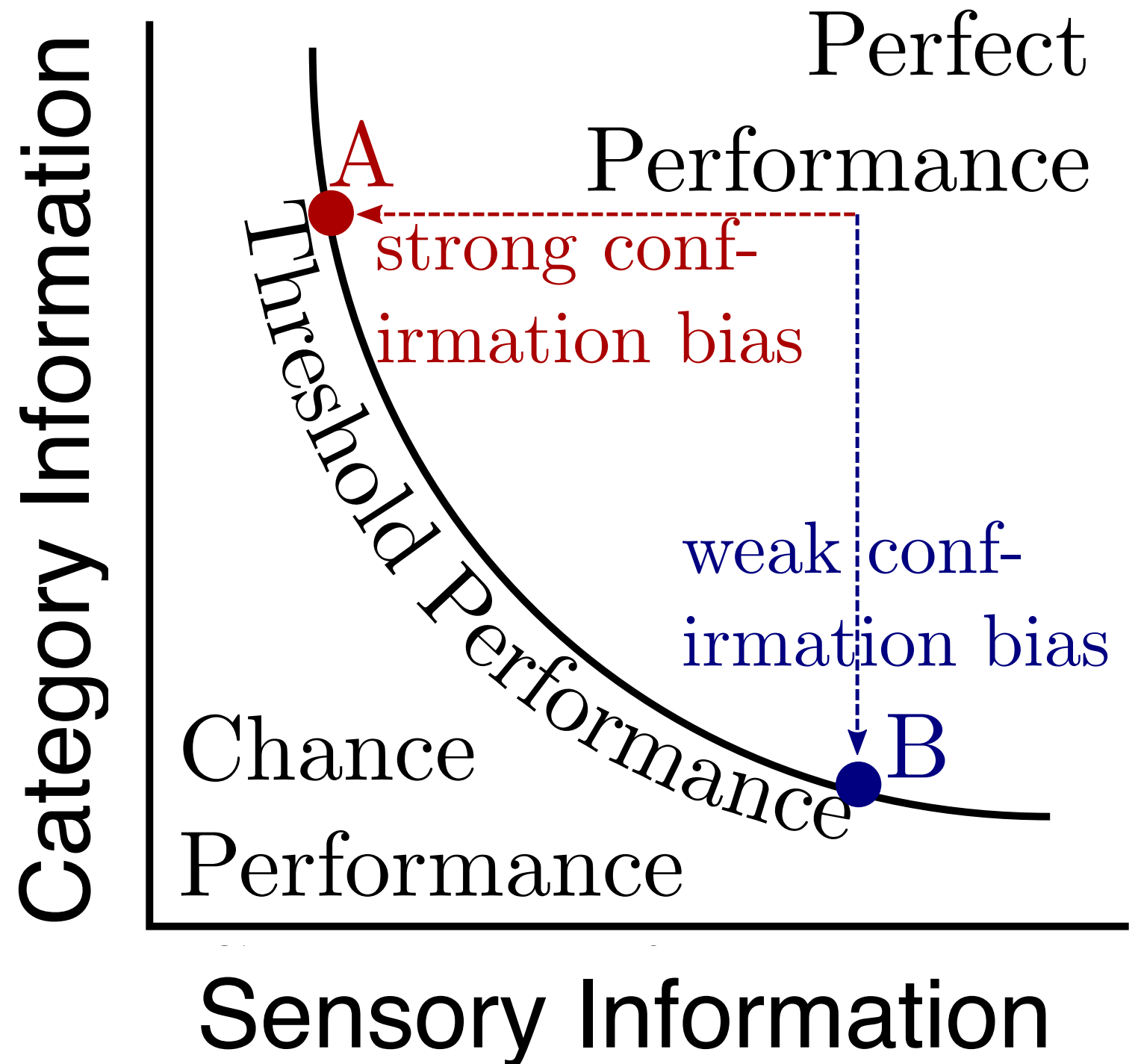
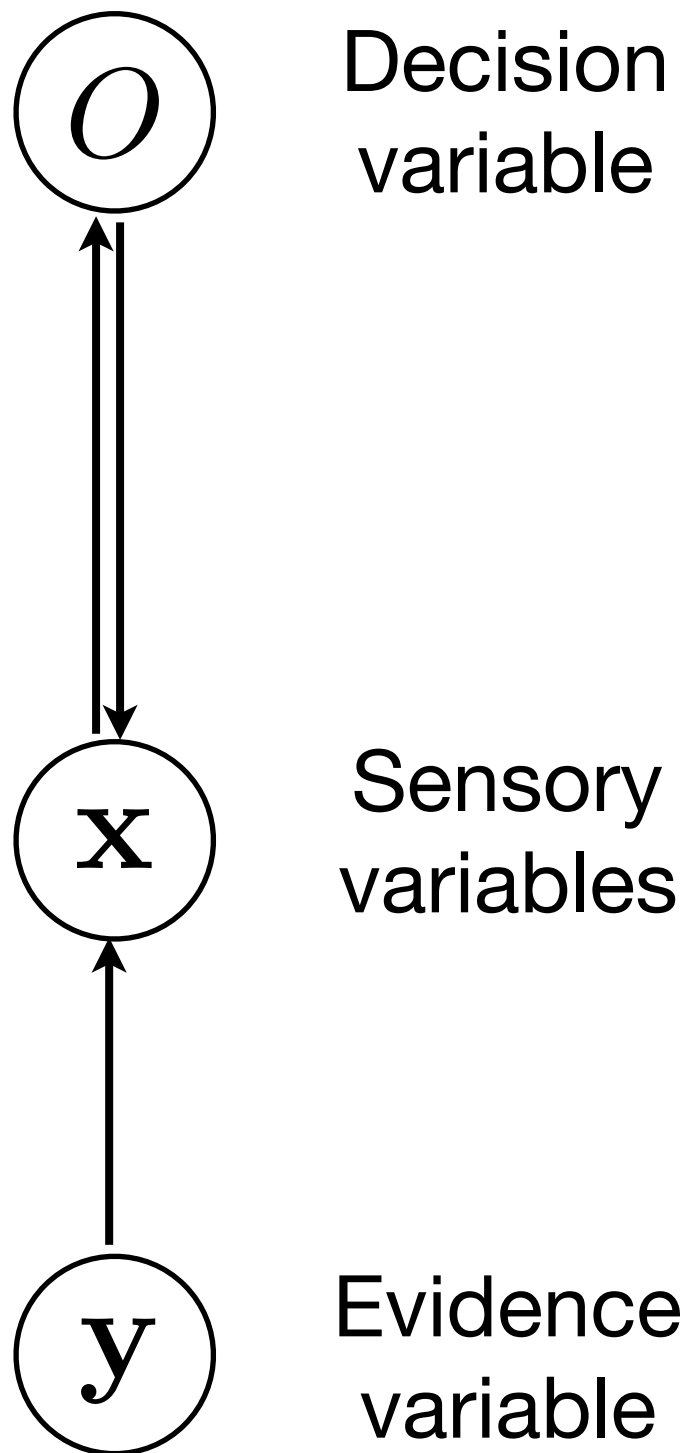


Ankani Chatteraj

# Two kinds of information

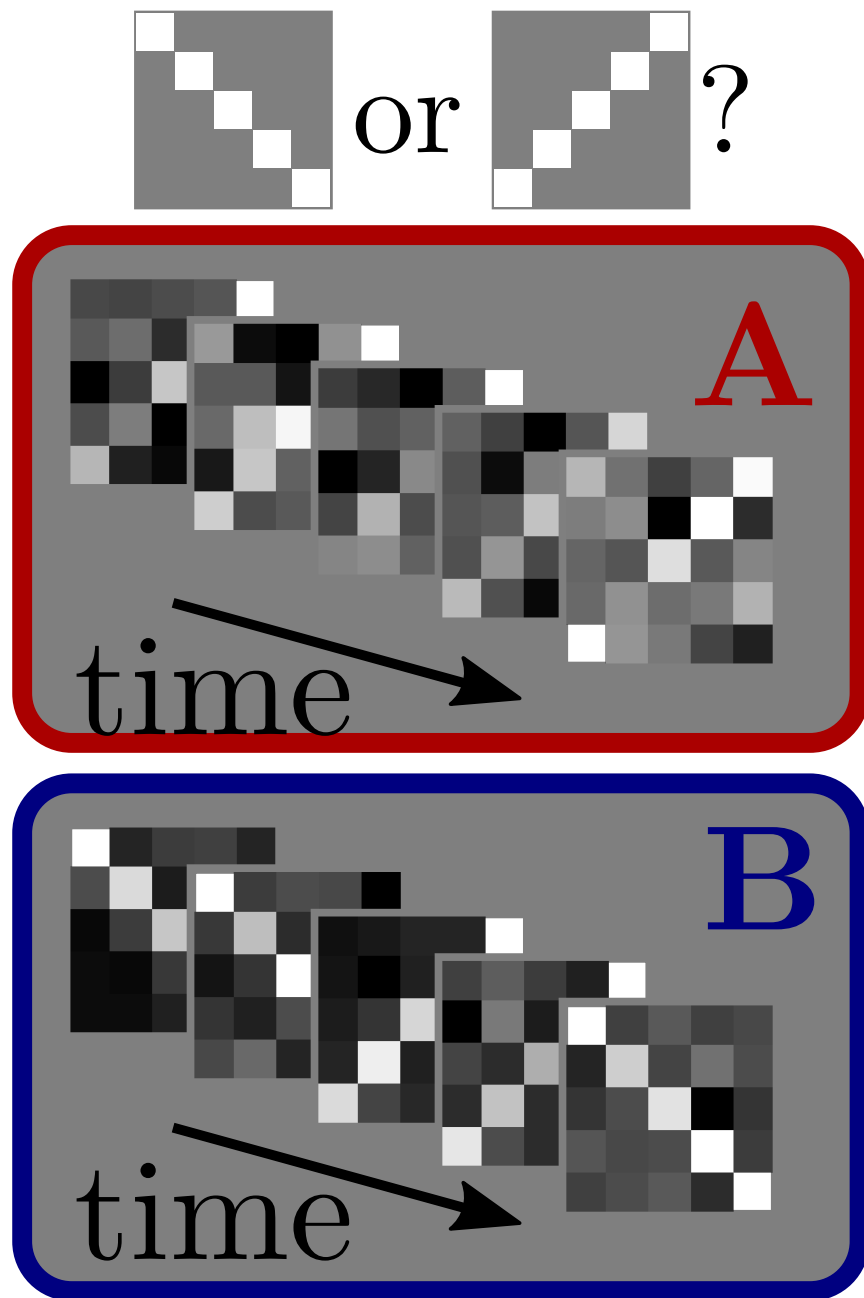


# Two kinds of information

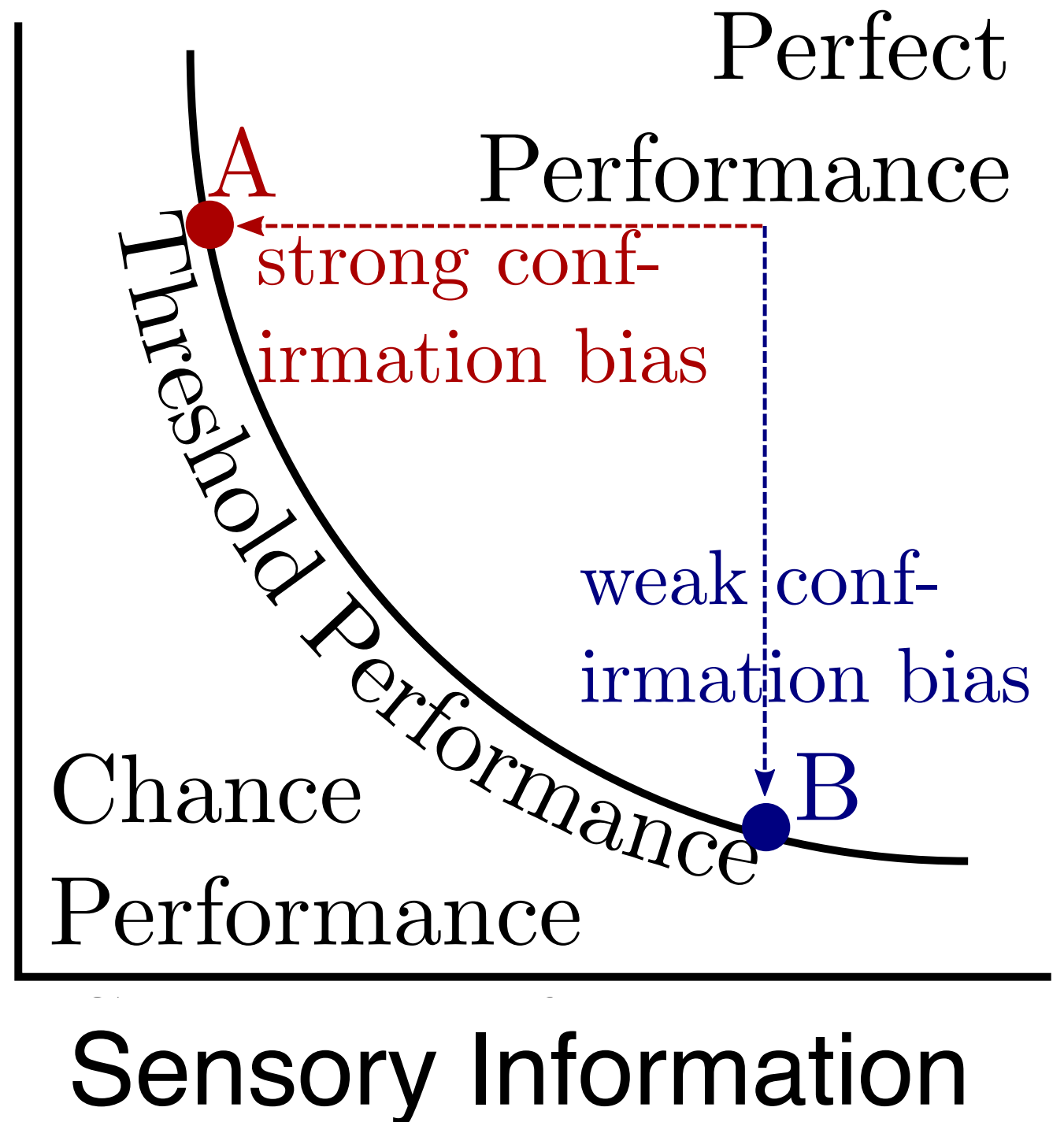




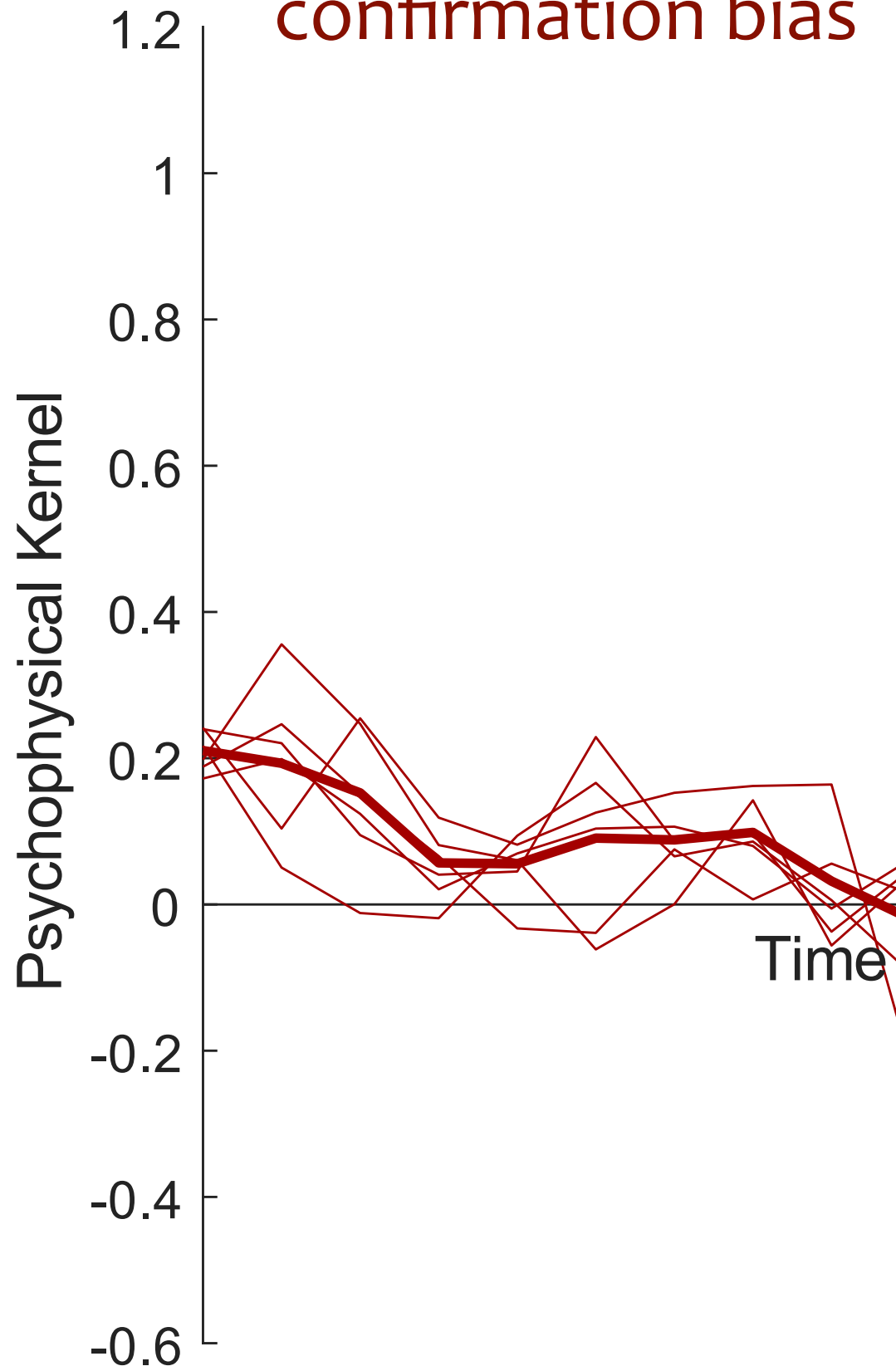
# Two different tasks



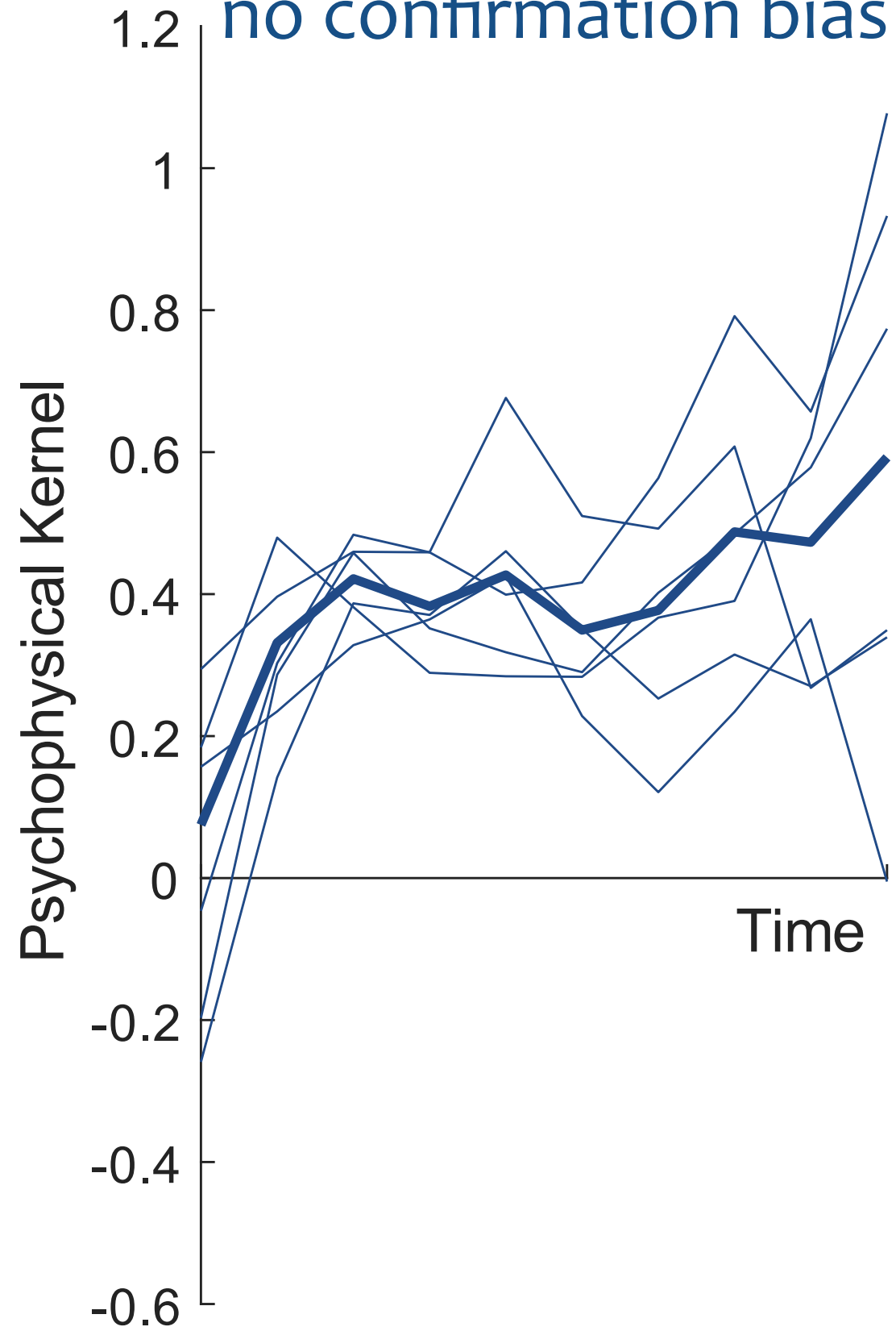
Category Information



**Task A:**  
**confirmation bias**



**Task B:**  
**no confirmation bias**



# Conclusion

- Approximate nature of inference process:
  - Infer task-model learnt by the brain from population responses
  - Explain confirmation bias -> poster!

# Acknowledgements

## **Lab**

Richard Lange  
Ankani Chatteraj  
Sabya Shivkumar

## **Theoretical collaborators**

Pietro Berkes  
Jozsef Fiser (CEU)

## **Experimental collaborators**

Adrian Bondy (Princeton)  
Bruce Cumming (NIH)

Camille Gomez-Faberge (Harvard)  
Rick Born (Harvard)



# University of Rochester (NY)



Looking for postdocs!



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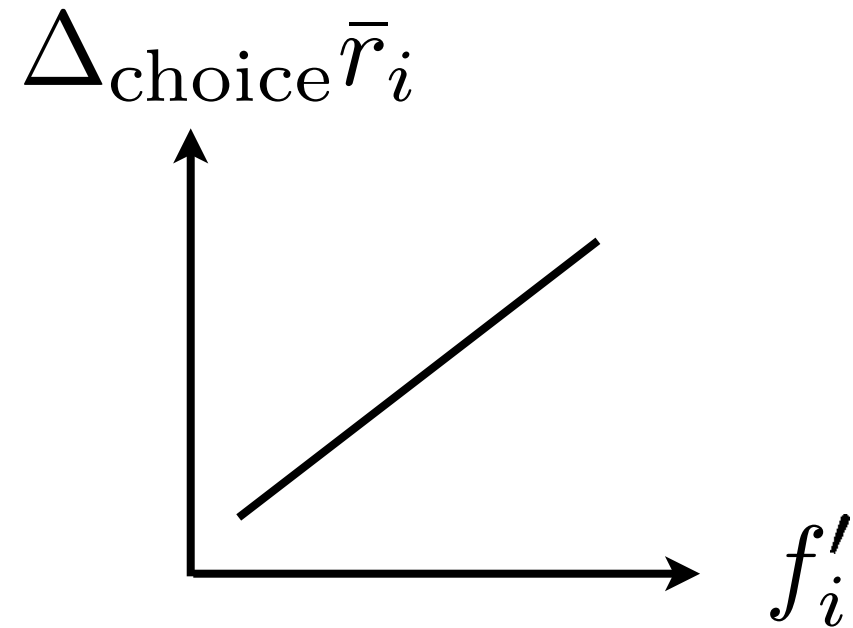
**BACKUP**

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# Predictions

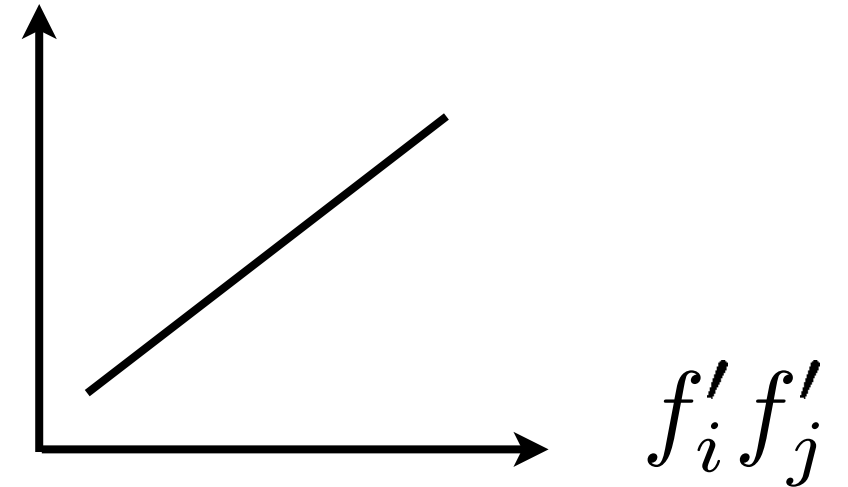
1<sup>st</sup> order

Haefner et al, NatNeuro 2013



2<sup>nd</sup> order

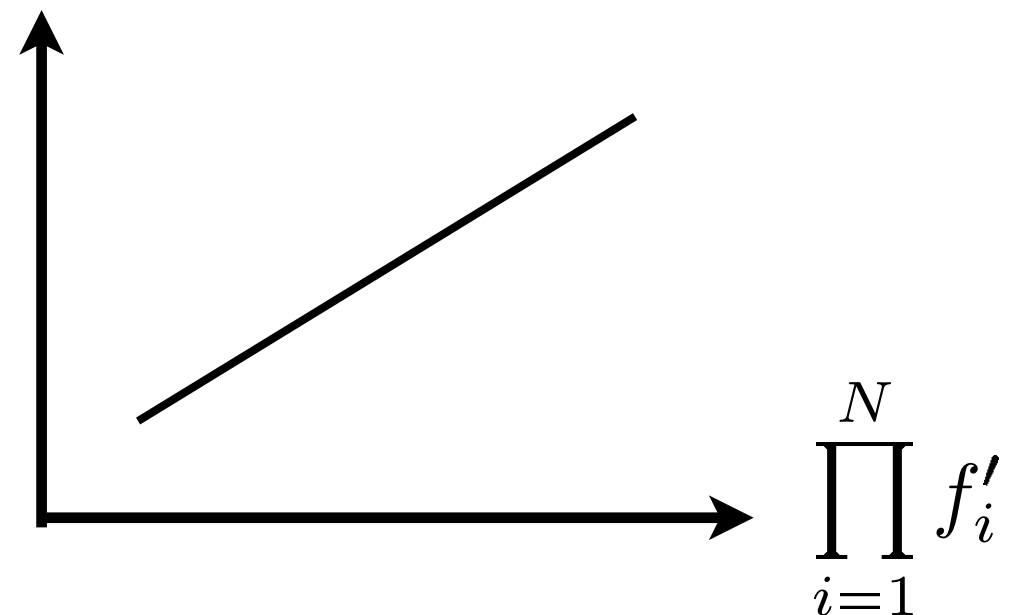
$\text{COV}(r_i, r_j)$



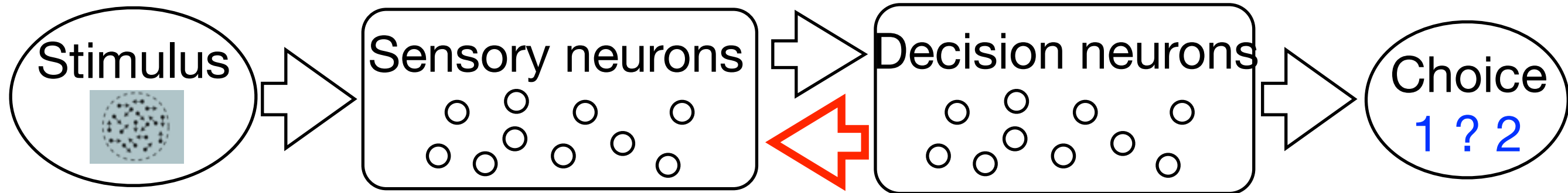
N<sup>th</sup> order

Odd  $N$ : choice-triggered  
(like 1st order)  
Even  $N$ : across all trials  
(like 2nd order)

(...)

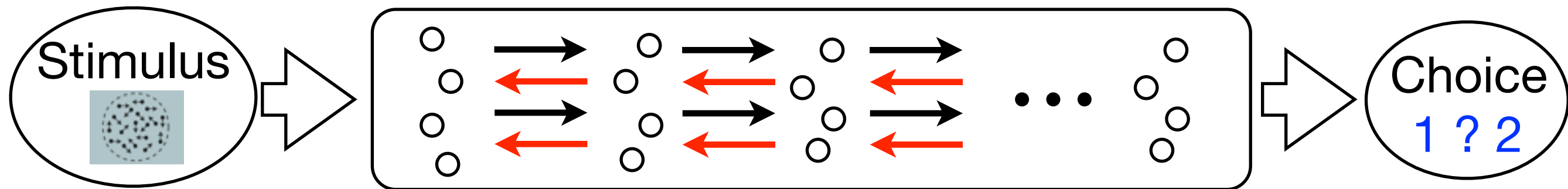


# 1) Brain's computation

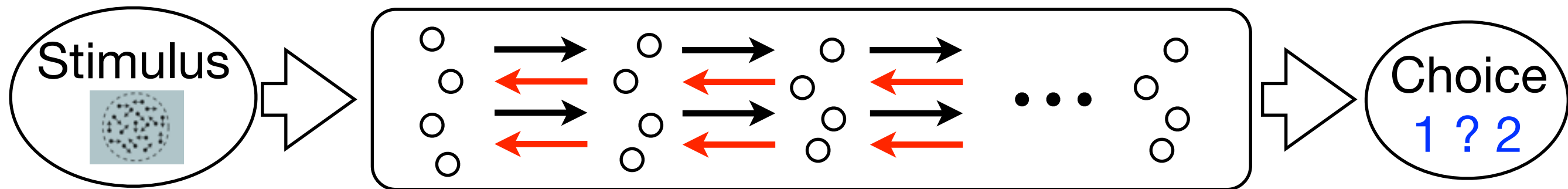




# 1) Brain's computation

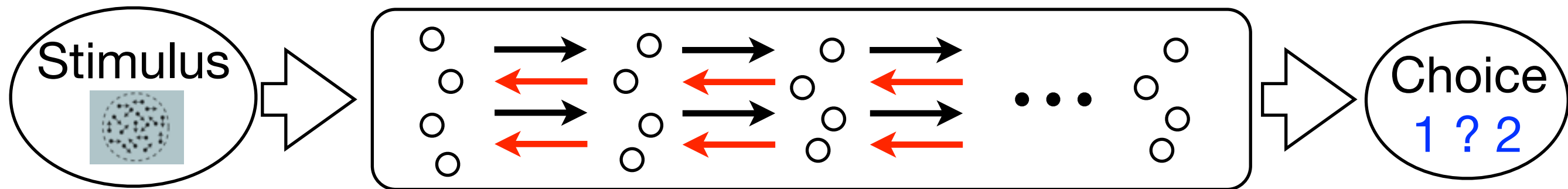


# 1) Brain's computation



Confirmation bias: Poster 1

# 1) Brain's computation



# 2) Neural data analysis

# Computational Neuroscience Lab

PI: Ralf Haefner, Brain & Cognitive Sciences



Richard Lange



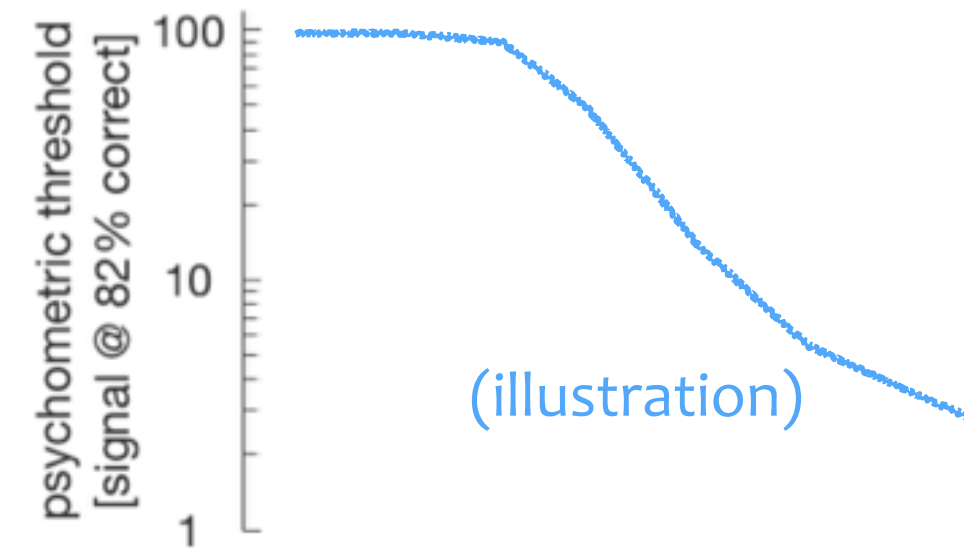
Ankani Chattoraj



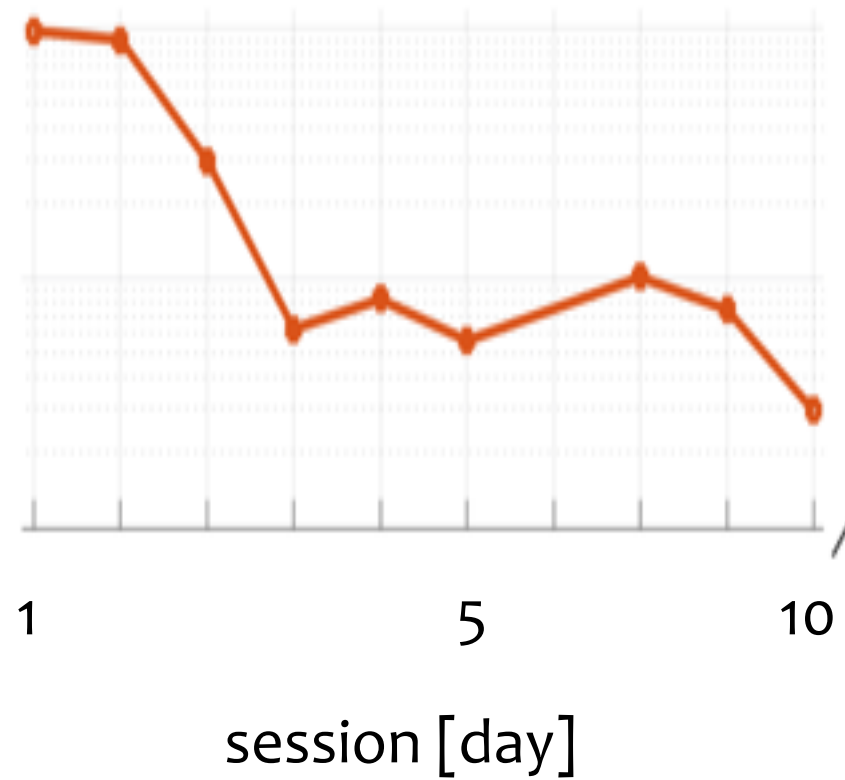
Sabya Shivkumar

# Preliminary results! (1 monkey)

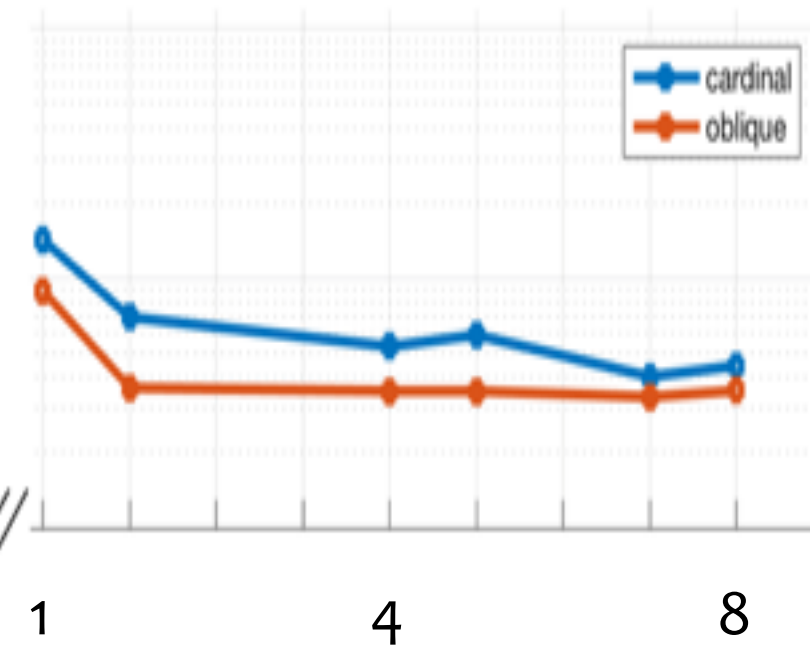
## Cardinal task



## Oblique task

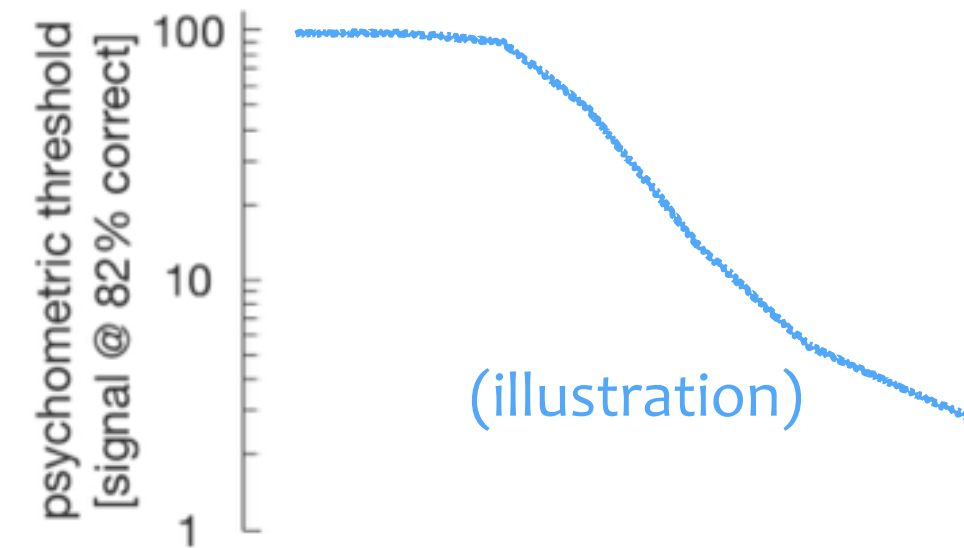


## Interleaved

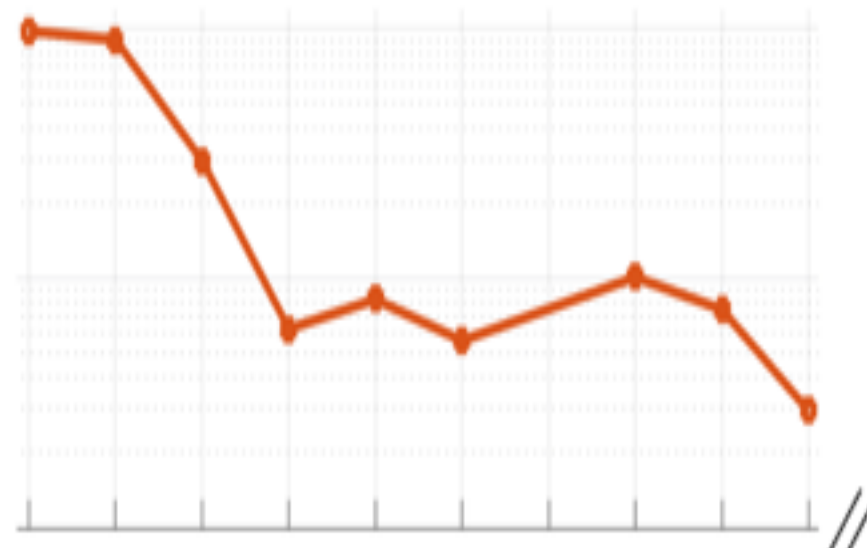


# Preliminary results! (1 monkey)

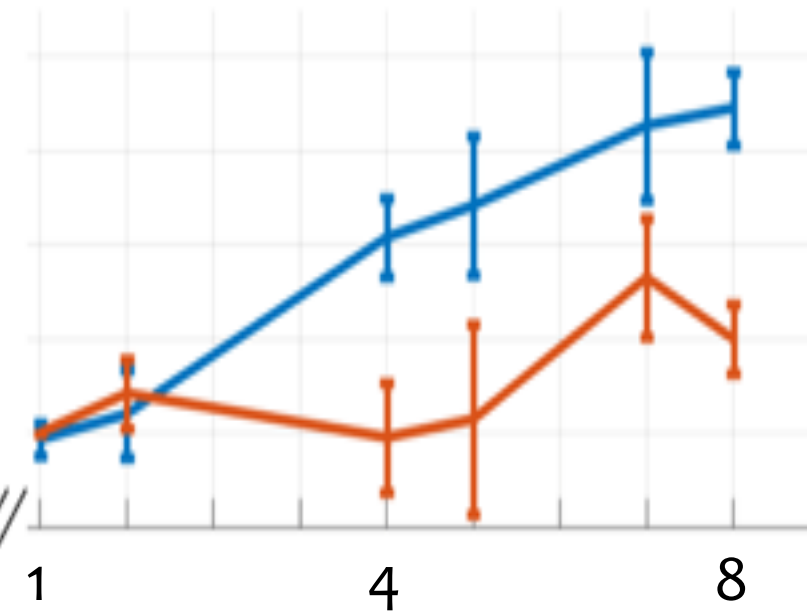
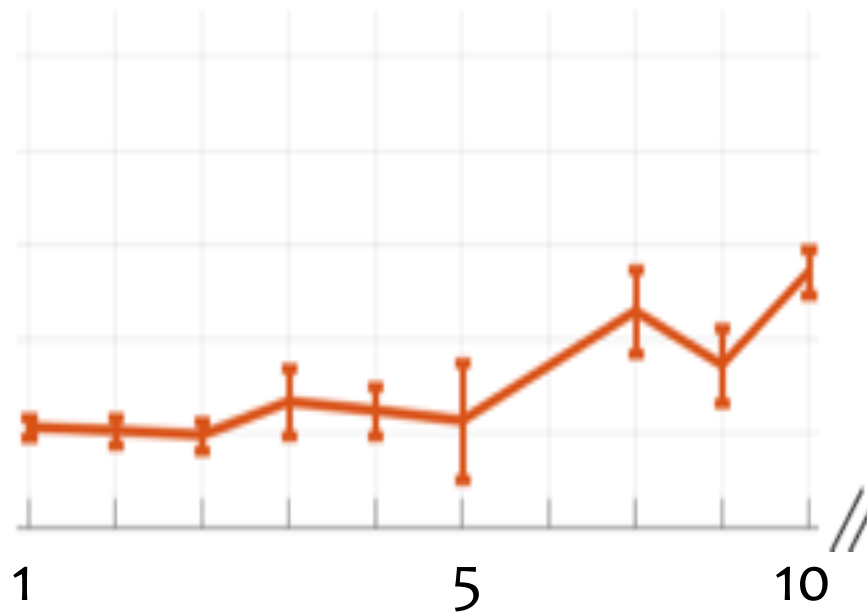
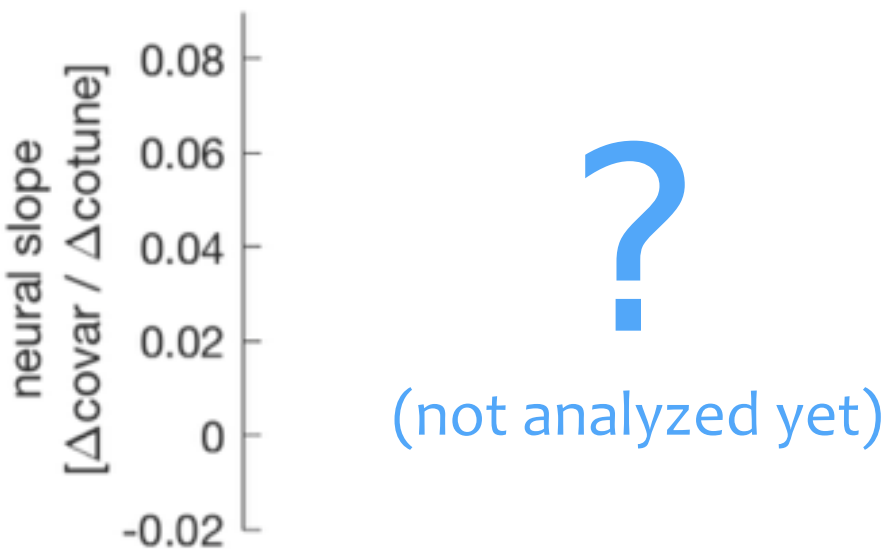
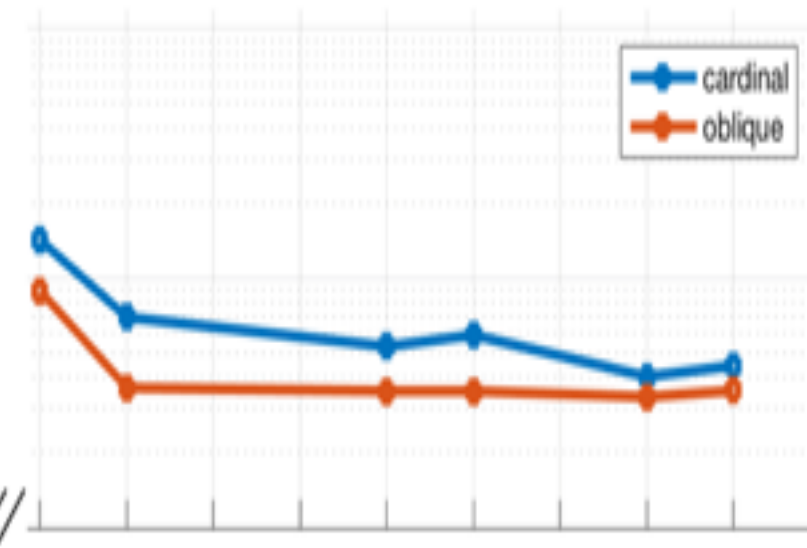
## Cardinal task



## Oblique task



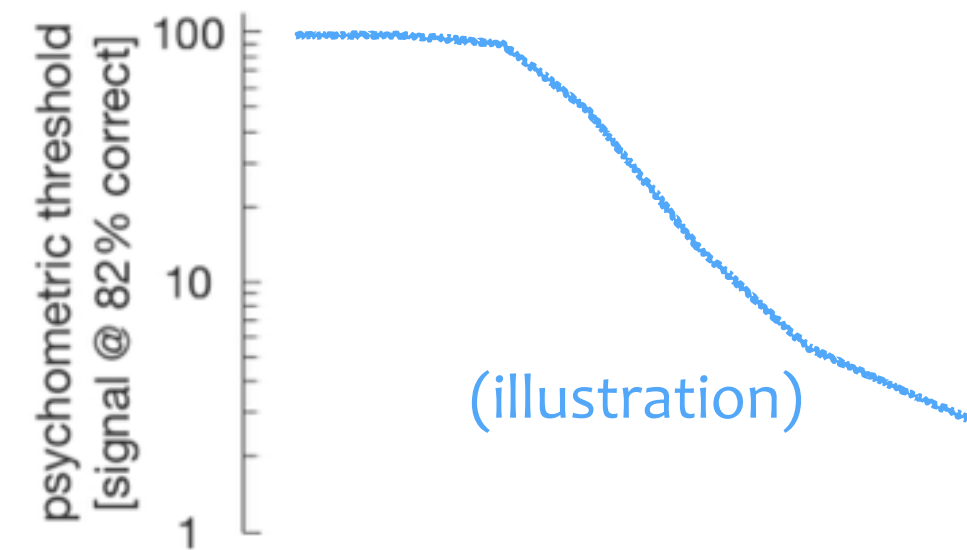
## Interleaved



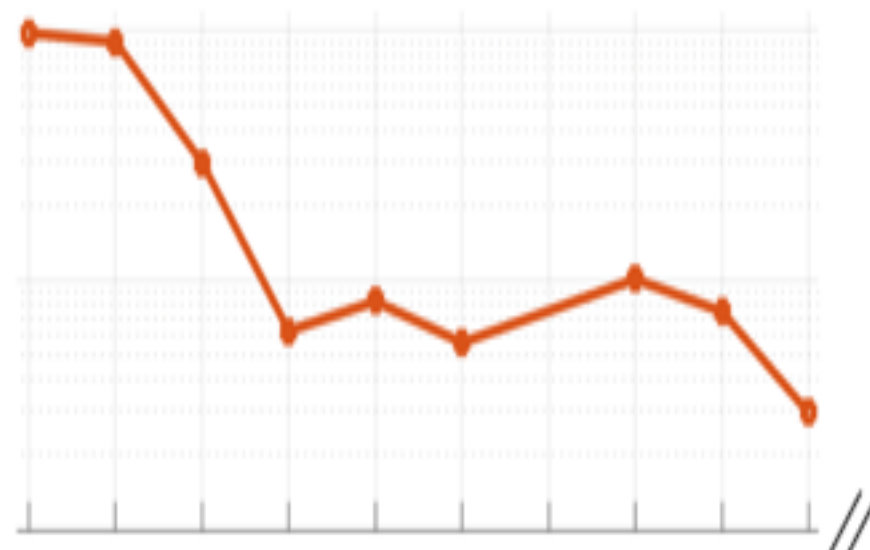
session [day]

# Preliminary results! (1 monkey)

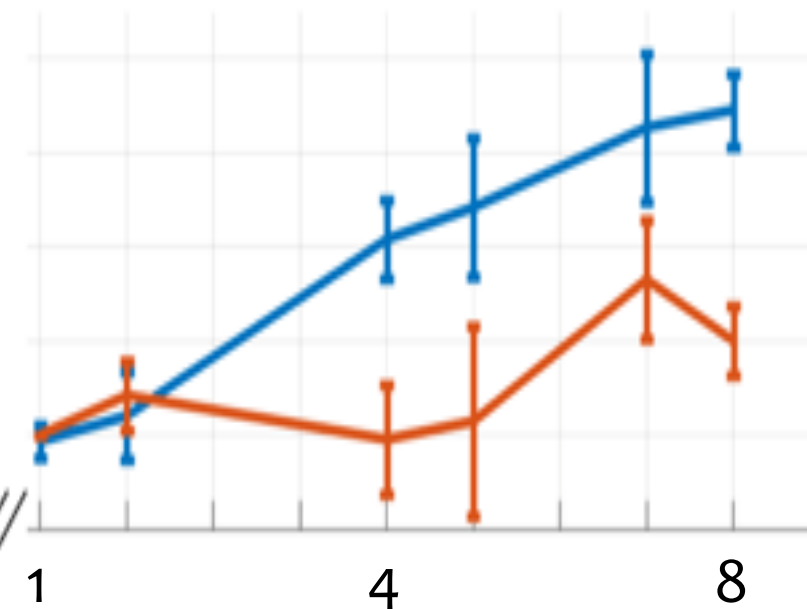
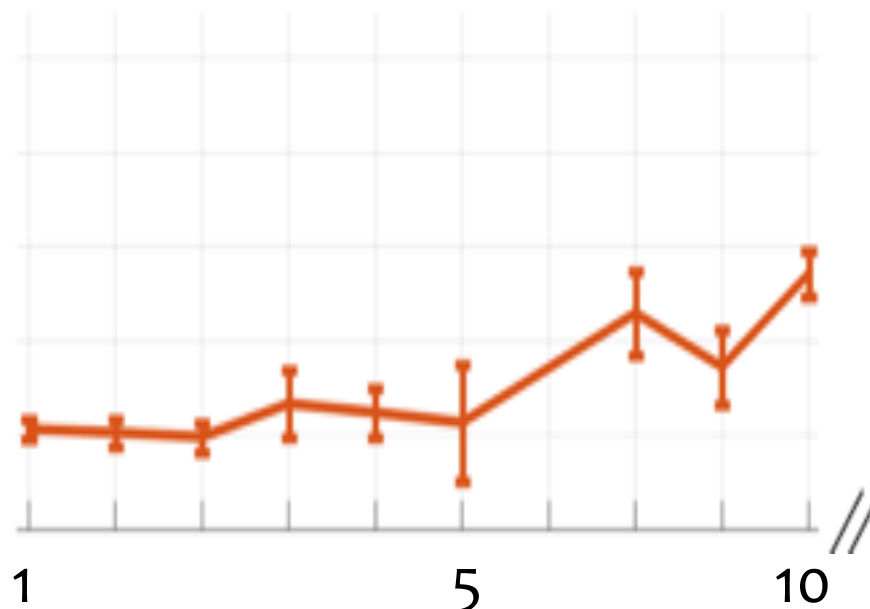
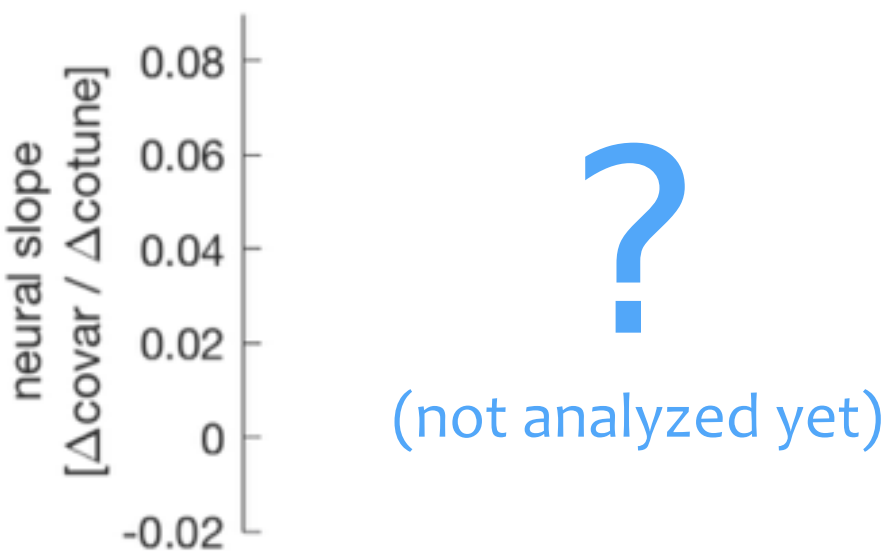
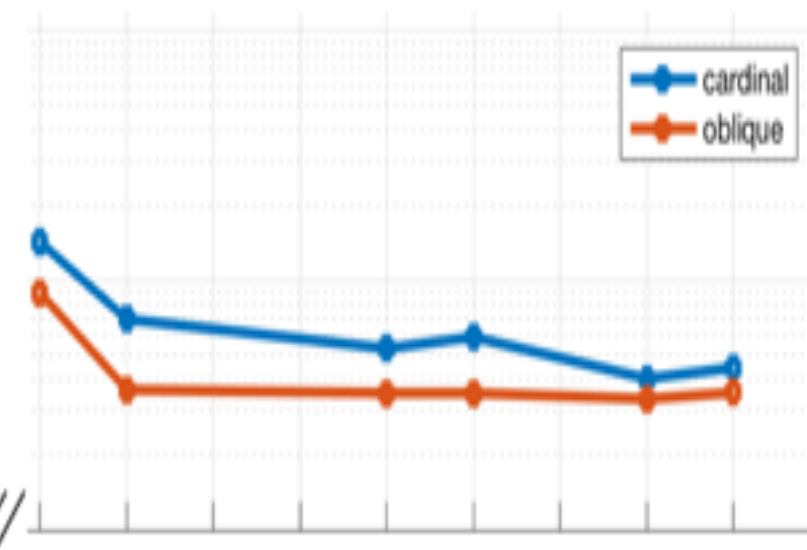
## Cardinal task



## Oblique task

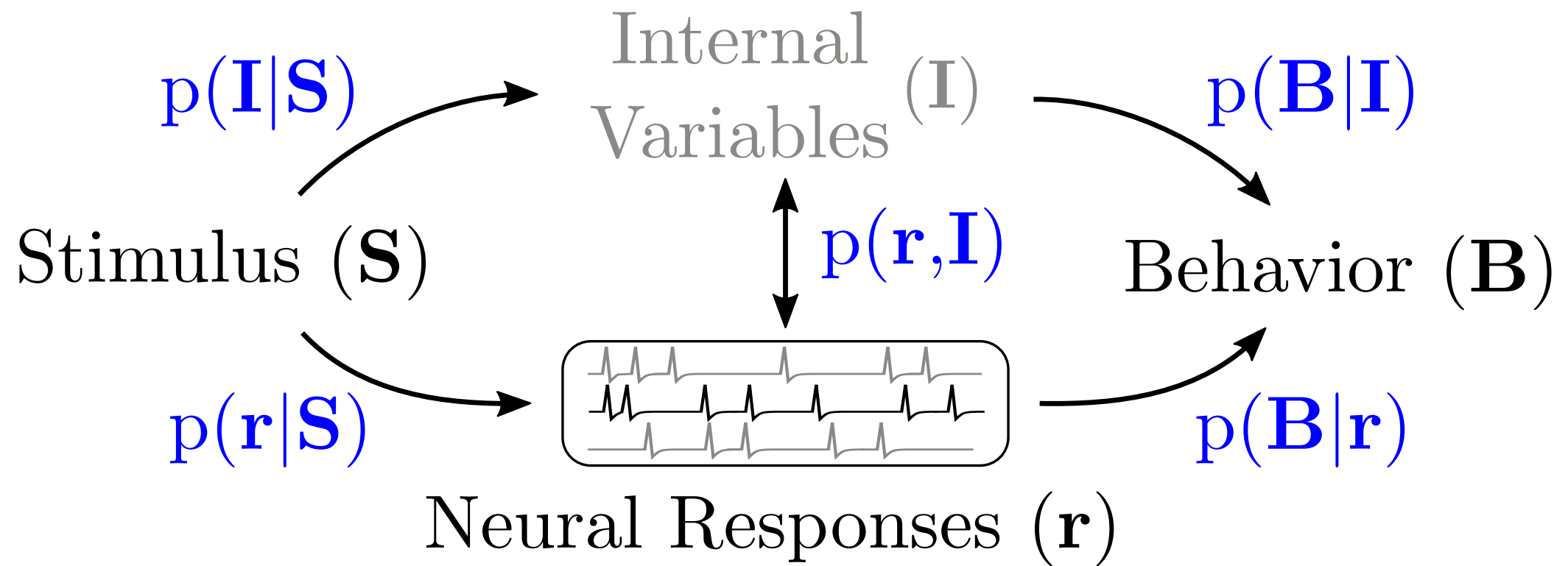


## Interleaved



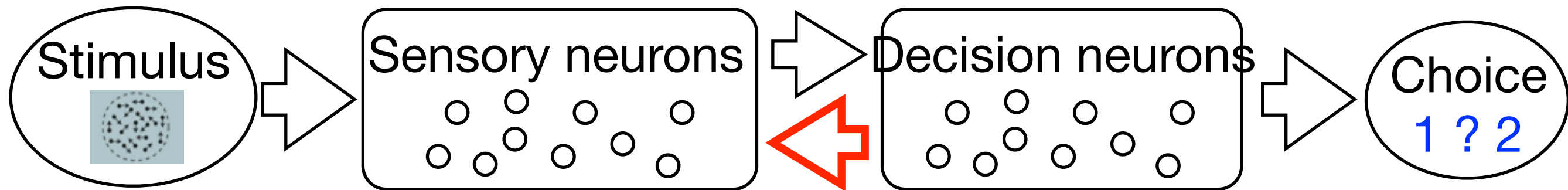
session [day]

▶ Next steps: analyze opposite task, compare cooling data



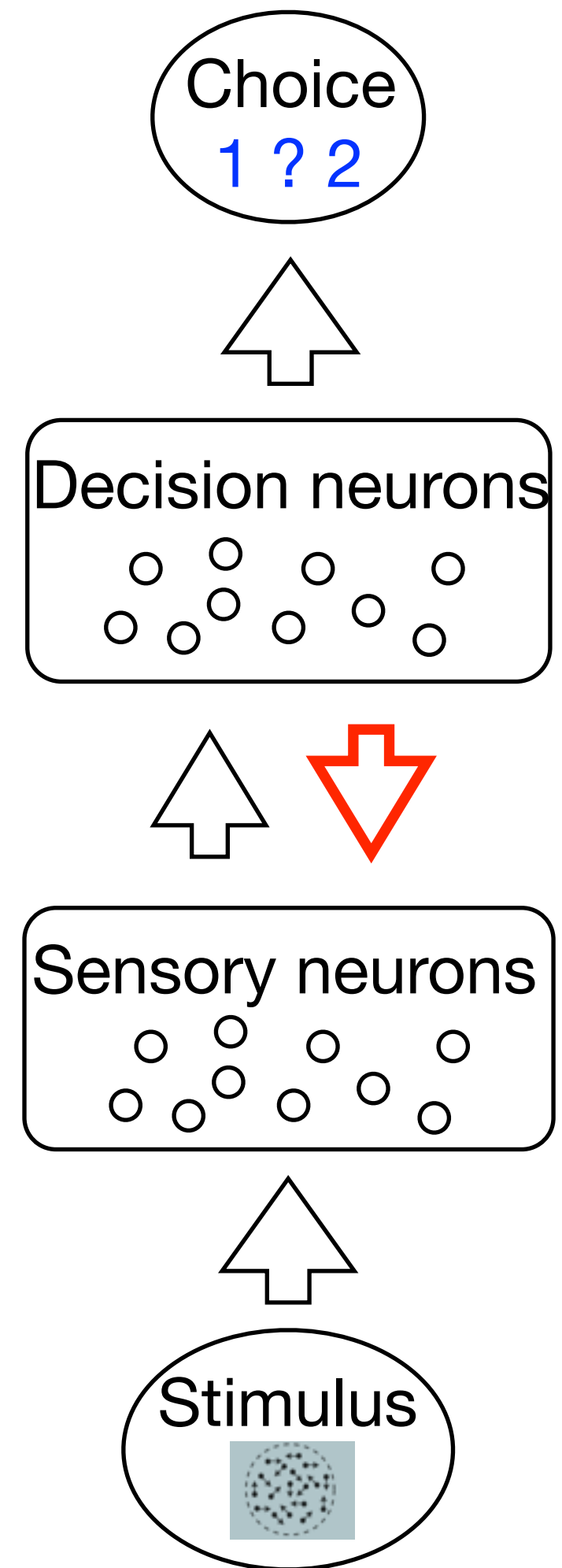


# 1) Brain's computation



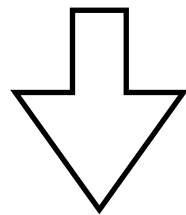
# 2) Neural data analysis

# 1) Brain's computation

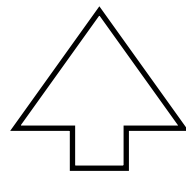


# 1) Brain's computation

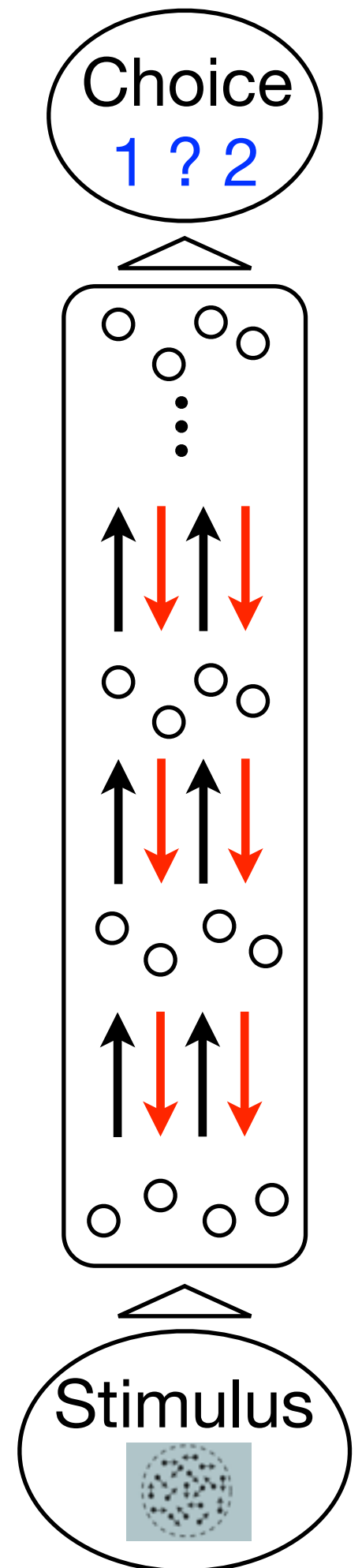
Prior knowledge



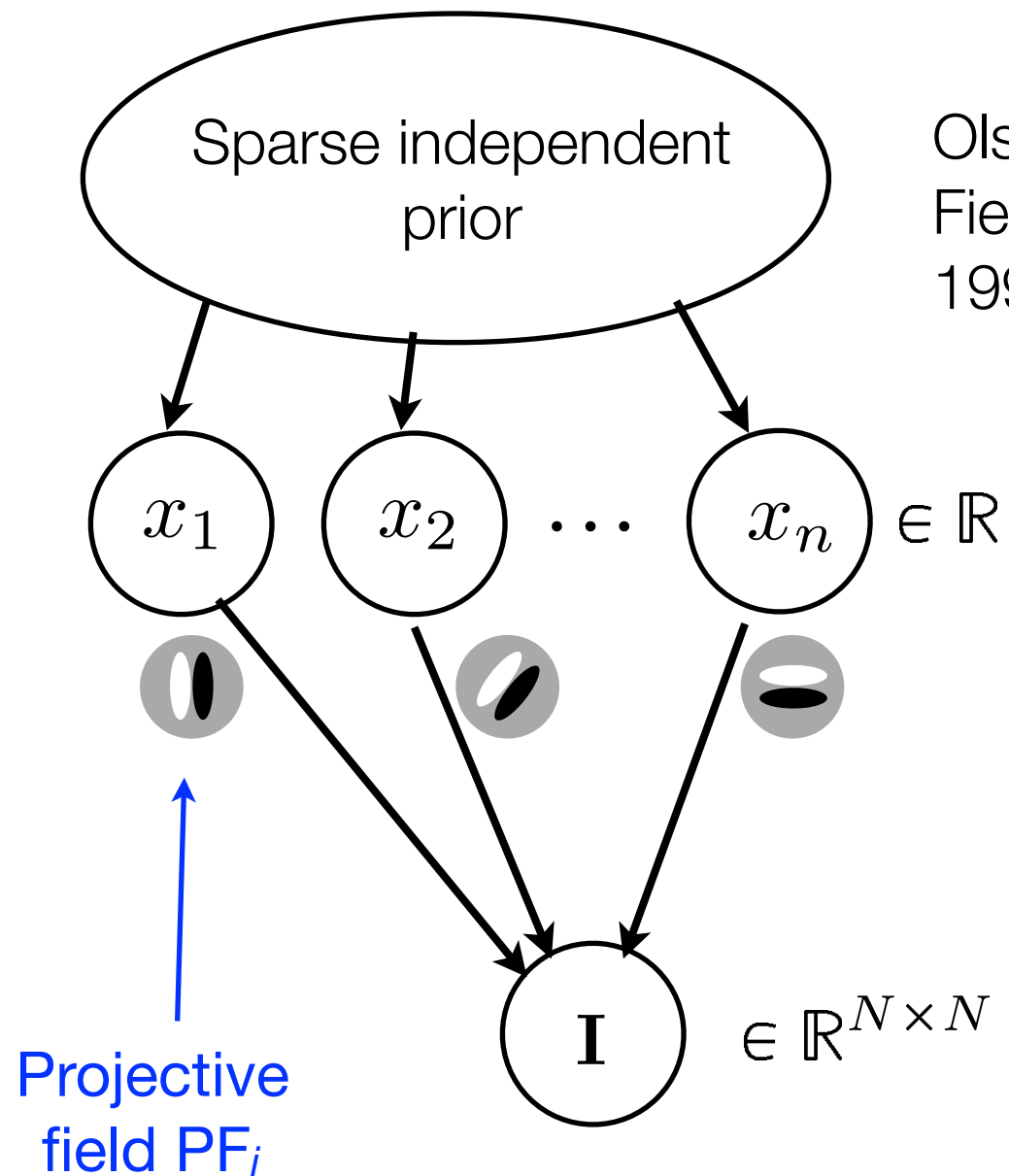
Posterior beliefs



External inputs



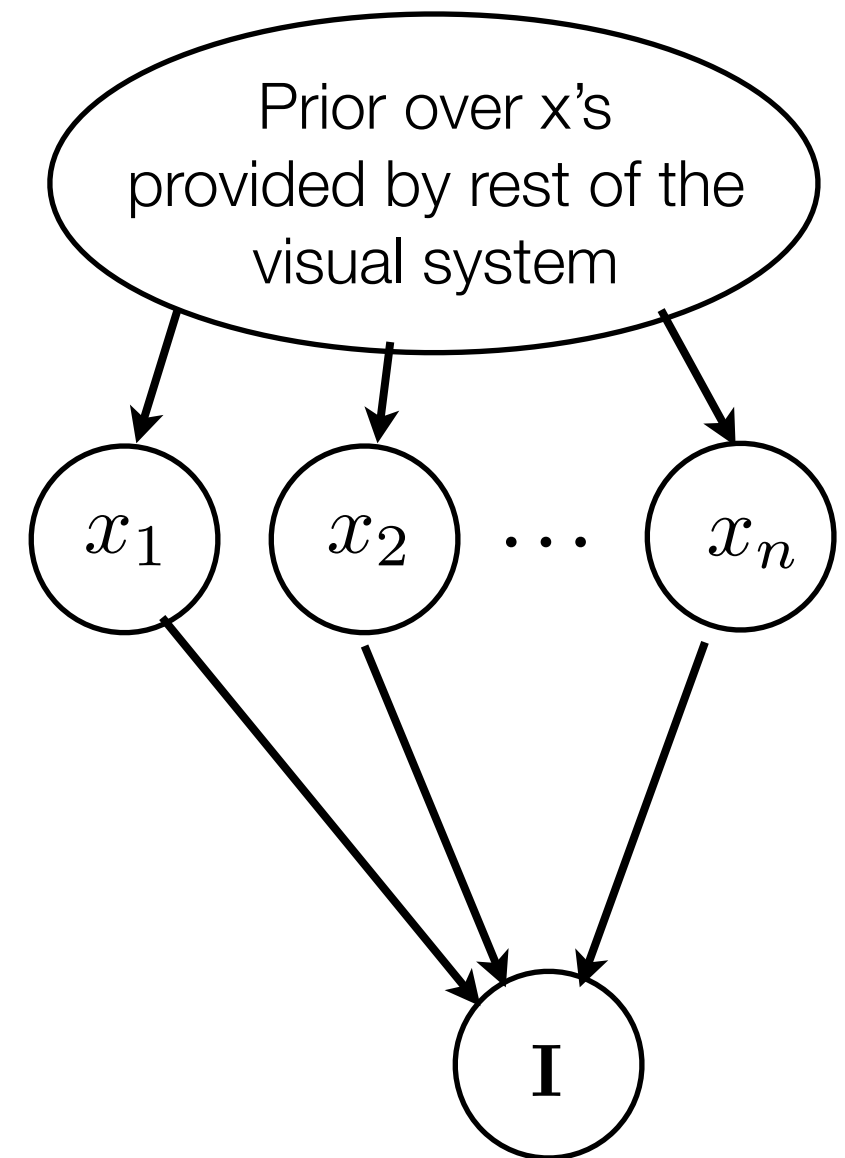
# Sampling in a probabilistic model of V1



$$p(\mathbf{I}|\mathbf{x}) = \mathcal{N} \left( \mathbf{I} : \sum_i PF_i x_i, \sigma^2 \right)$$

Sampling: Hoyer & Hyvarinnen, NIPS 2003

Neural network implementation: Berkes et al, NIPS 2009

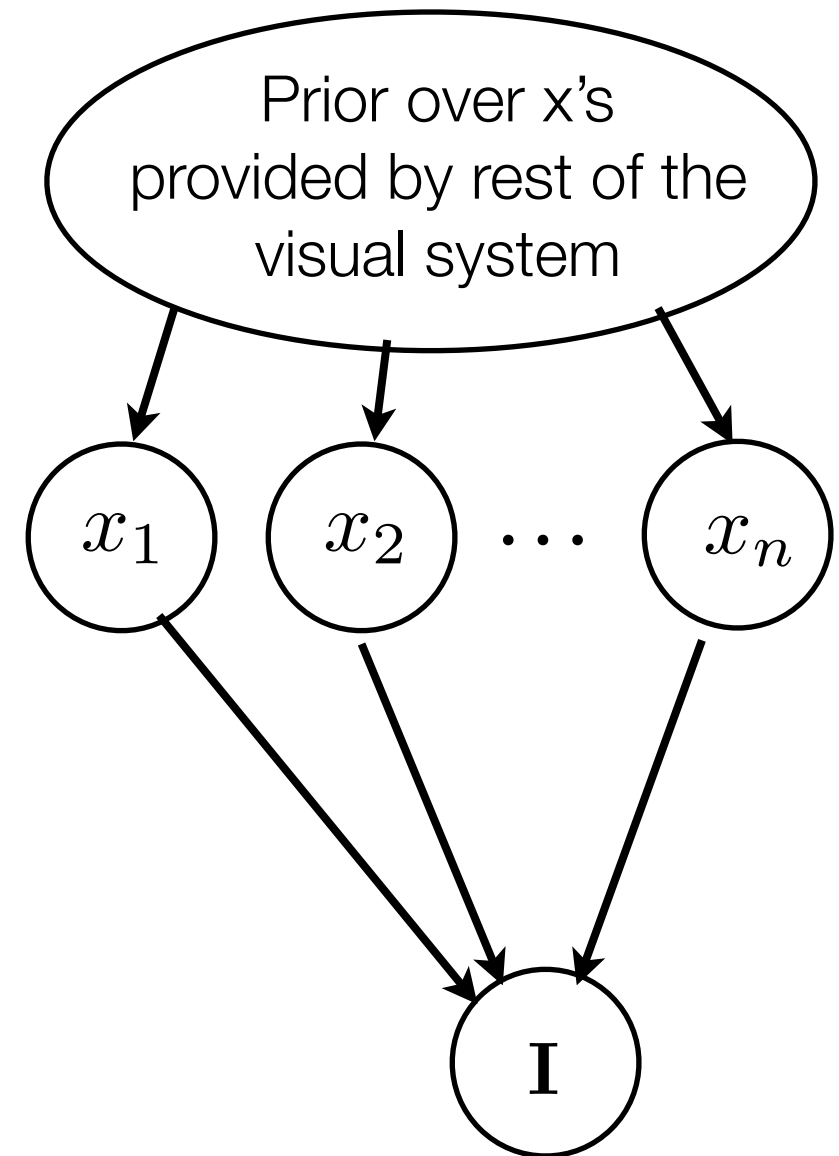


Perturbation approach:  
 Training on a psychophysical task changes internal model in predictable way  
 Allows predictions for neuronal behavior (e.g. correlations)

# Prediction for *changes* in neural activity

## Predictions for:

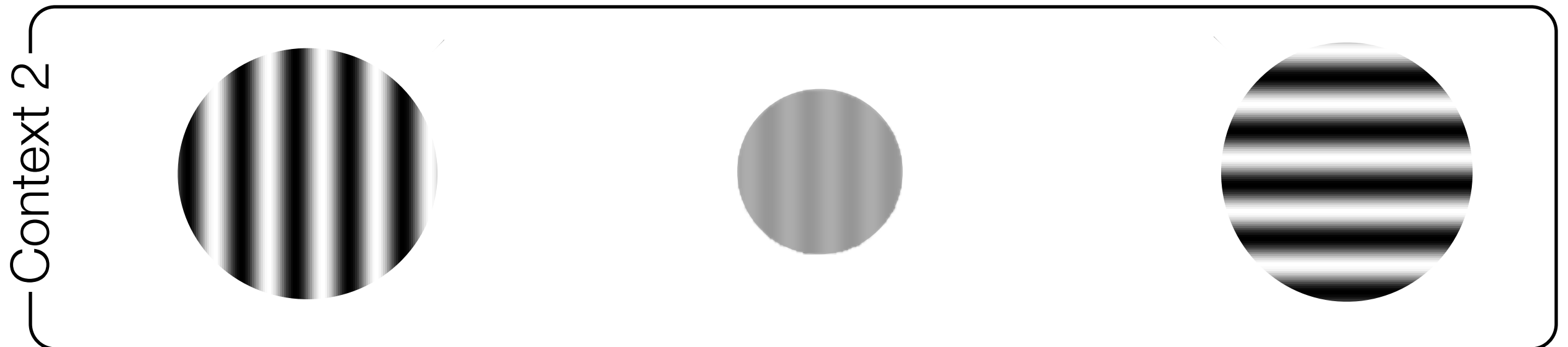
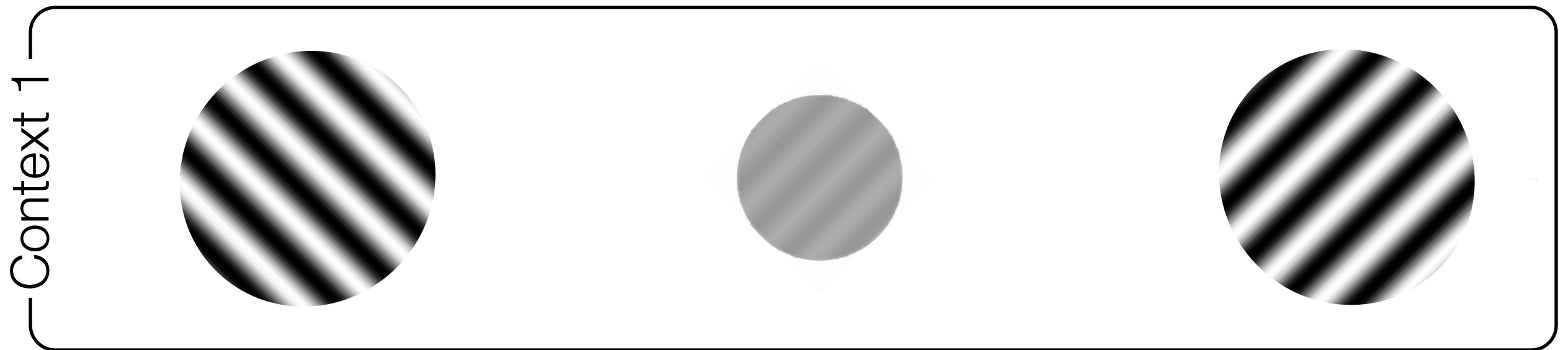
- Difference between before and after learning
- Difference between two different tasks



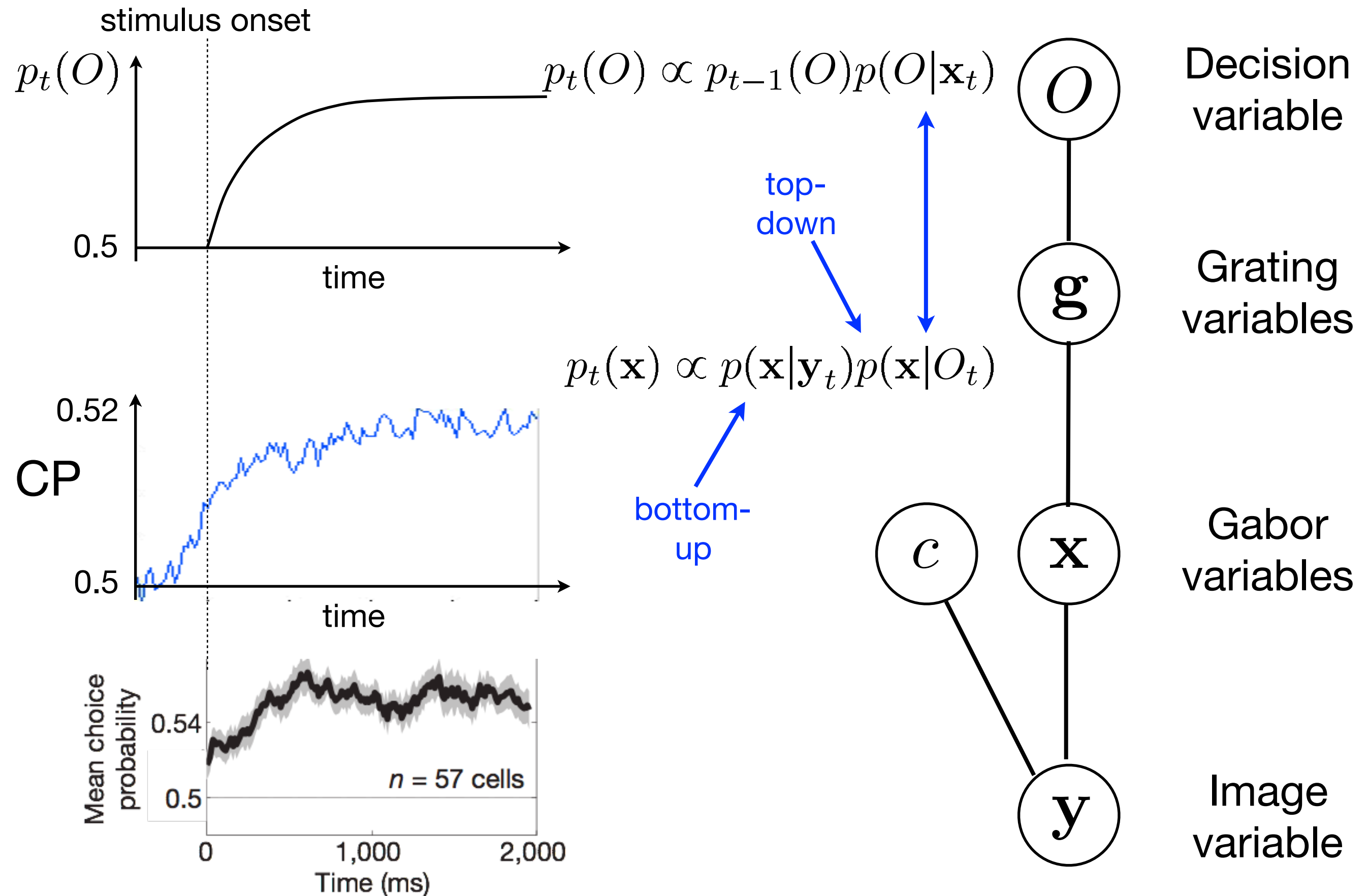
Perturbation approach:  
Training on a psychophysical task changes internal model in predictable way  
Allows predictions for neuronal behavior (e.g. correlations)

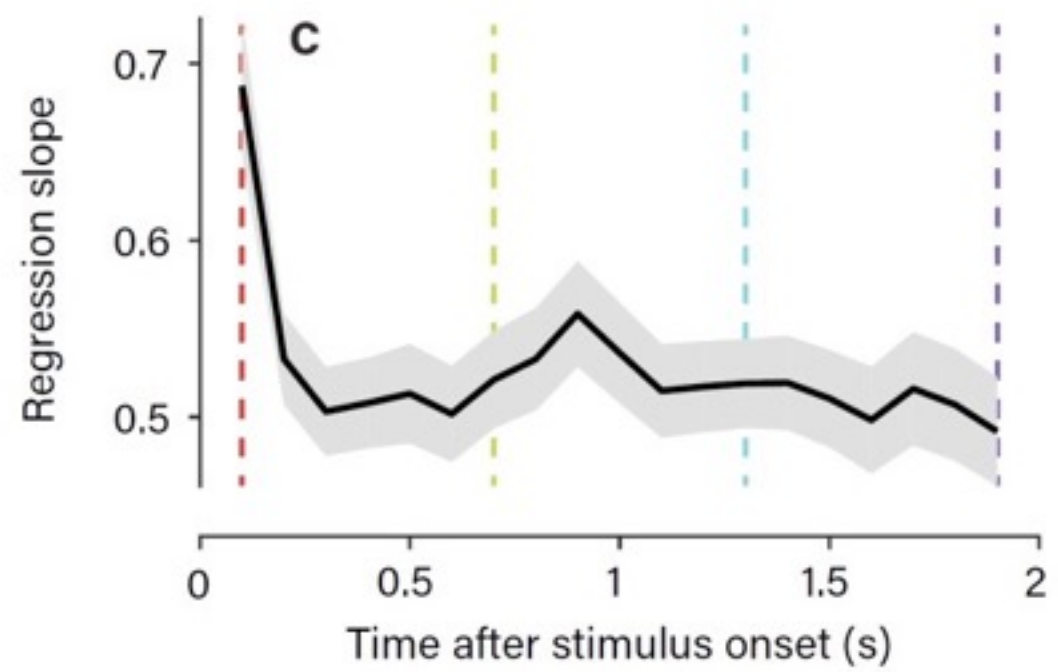
# Orientation discrimination task

Which of two perpendicular gratings caused the noisy image on the screen?

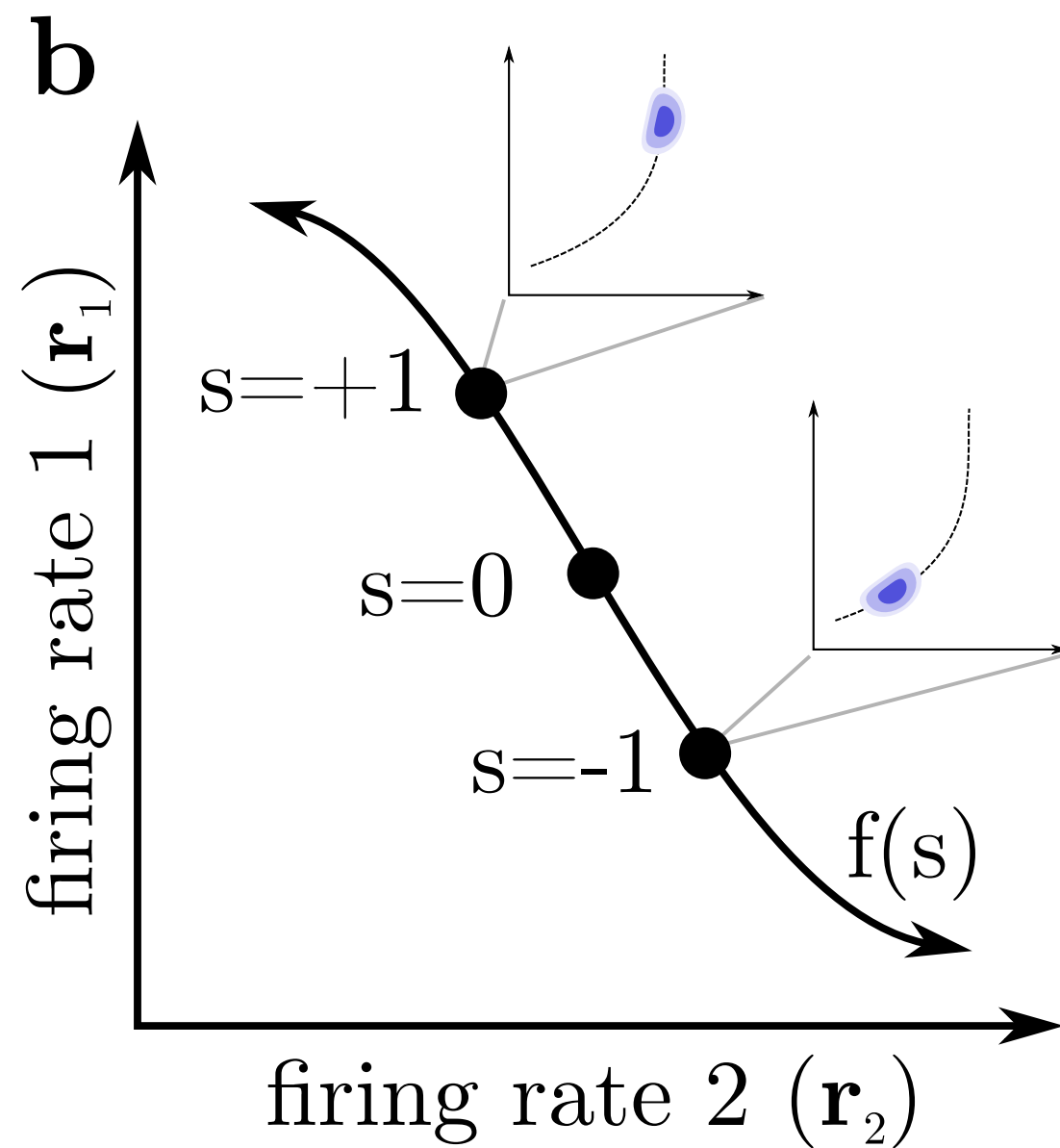
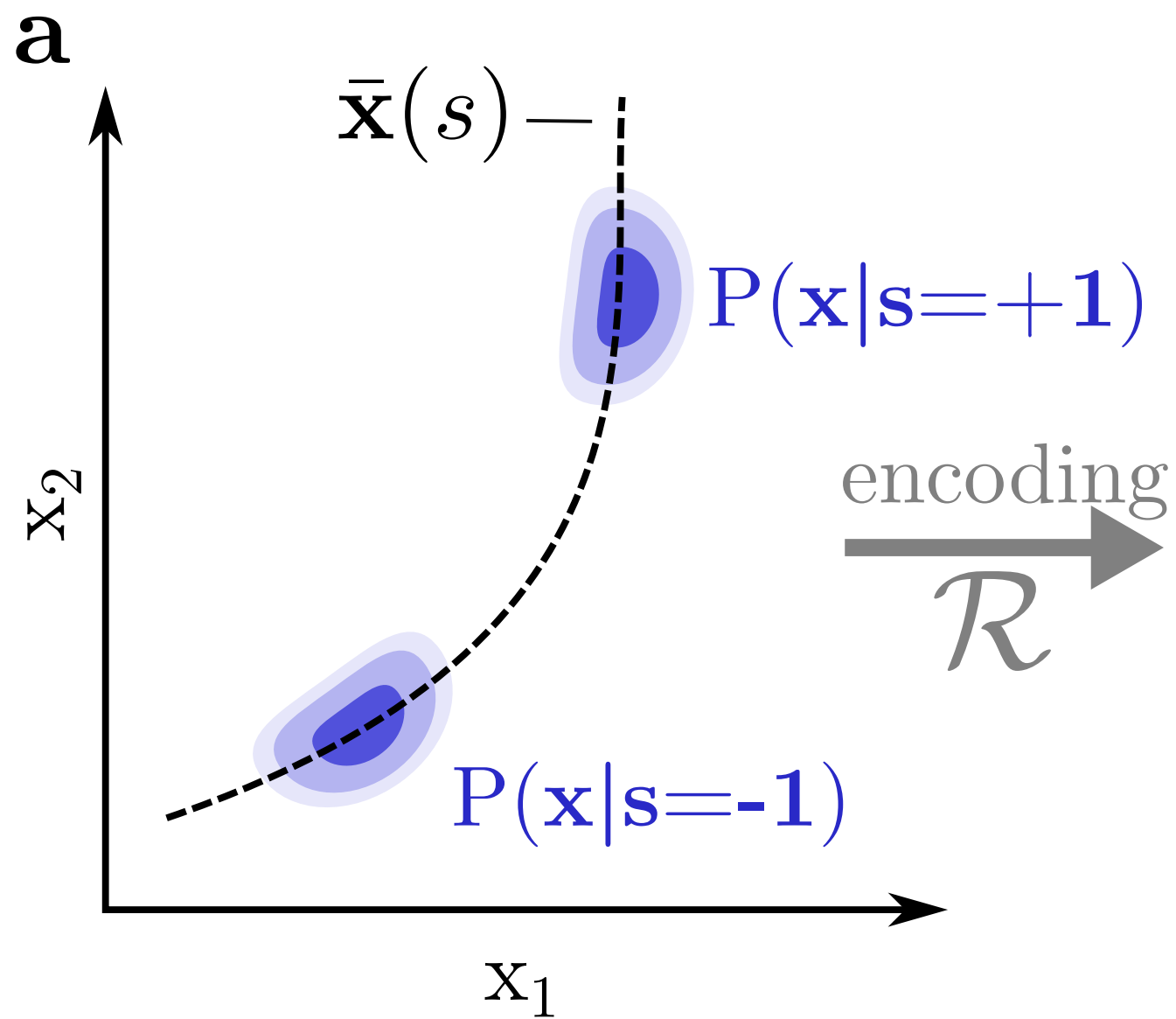


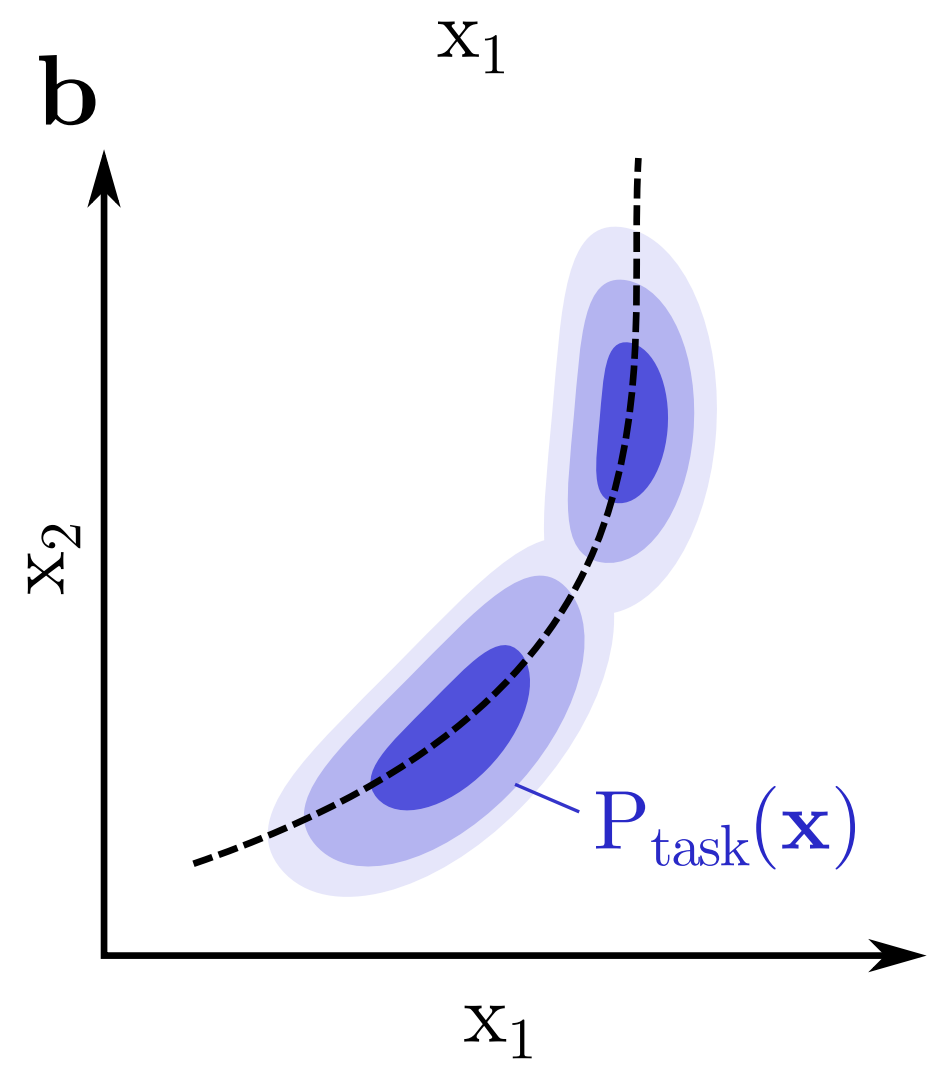
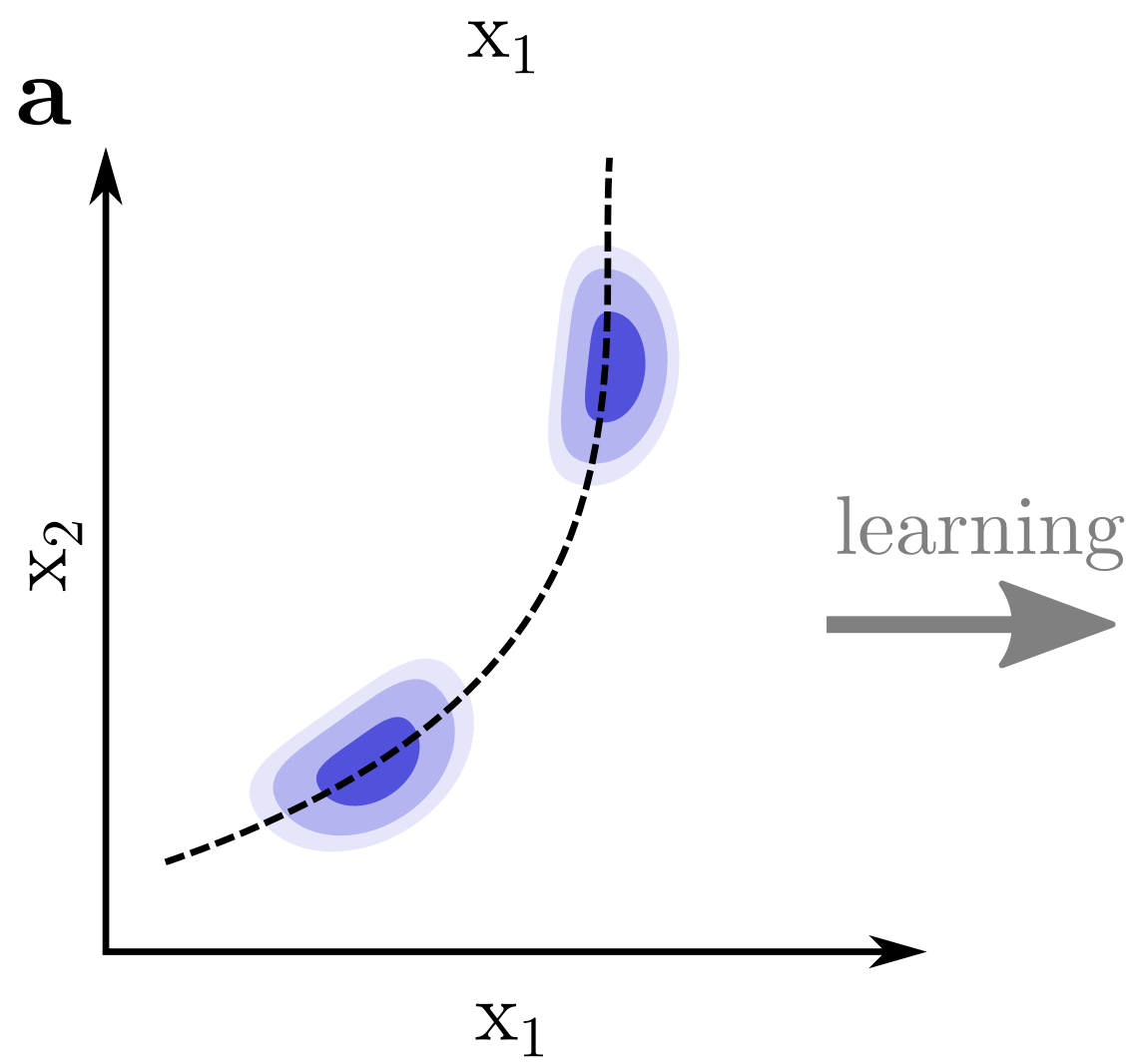
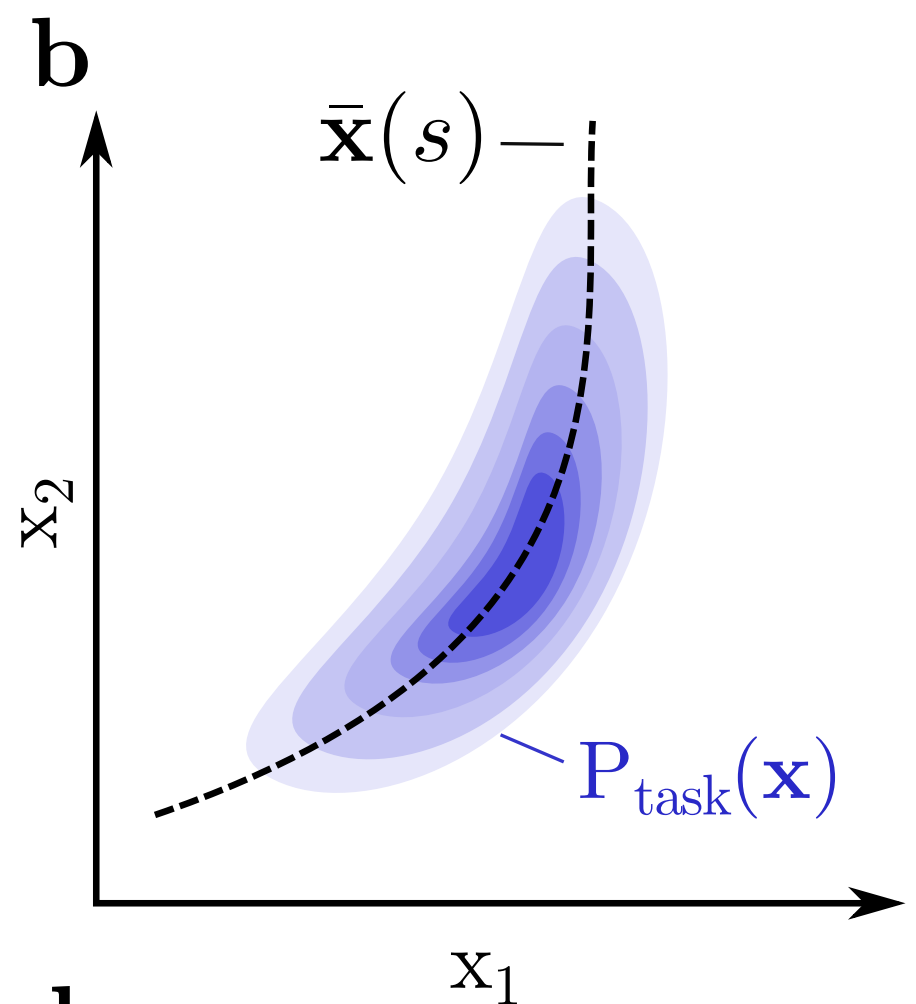
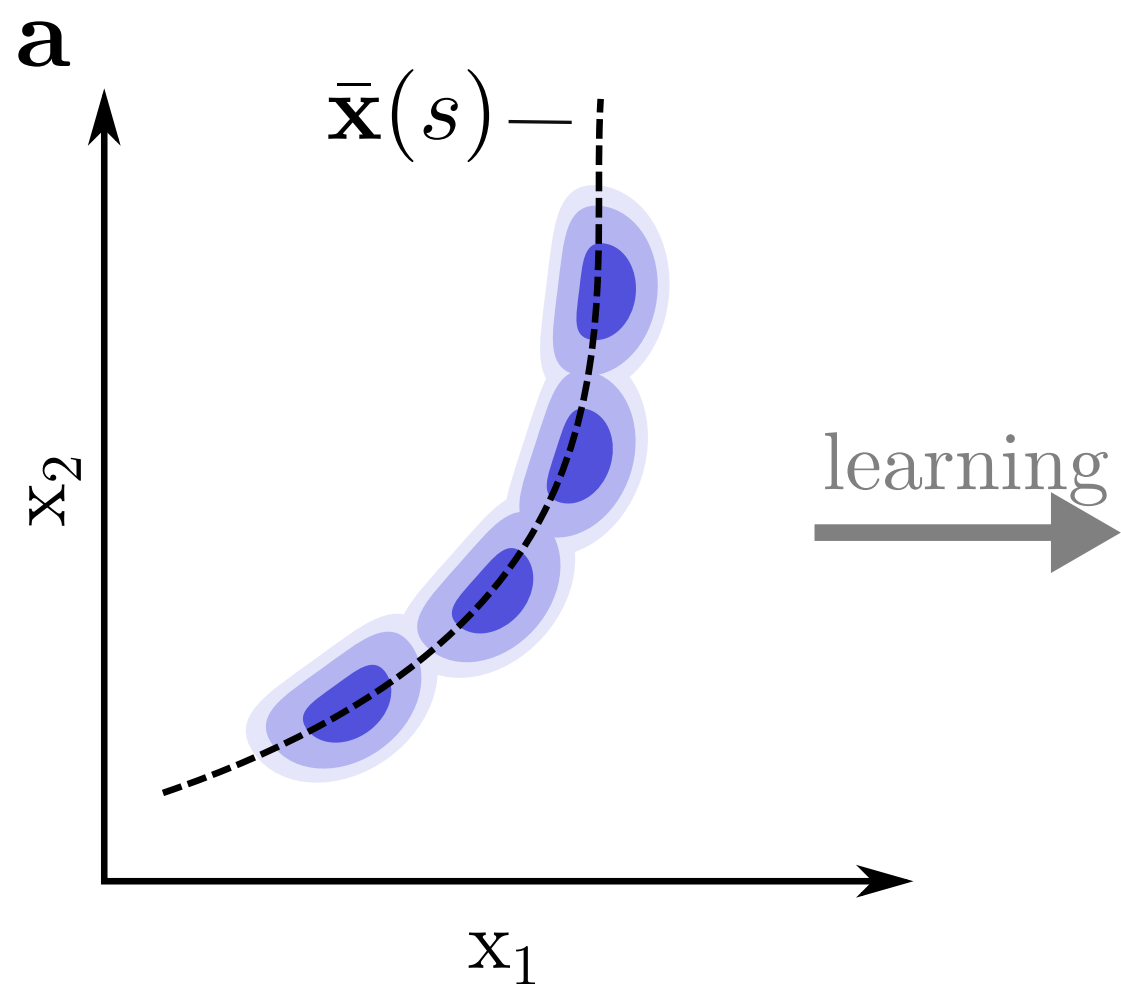
# Choice probabilities in $\mathbf{x}$



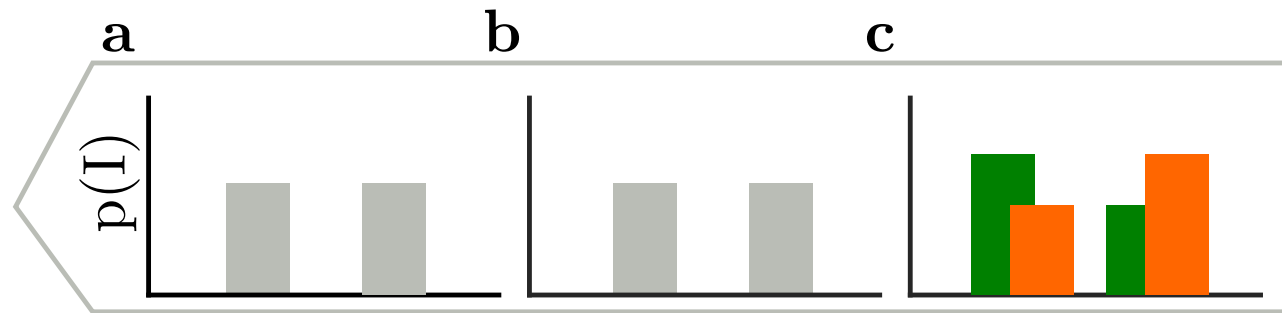




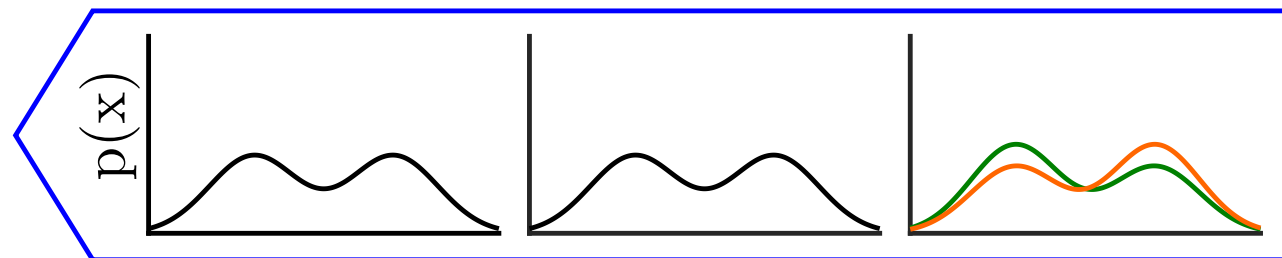




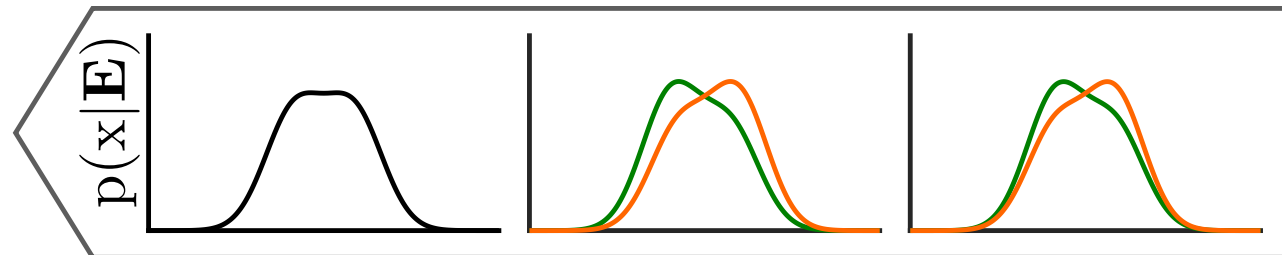
Discrimination  
Task  
categorical  
prior



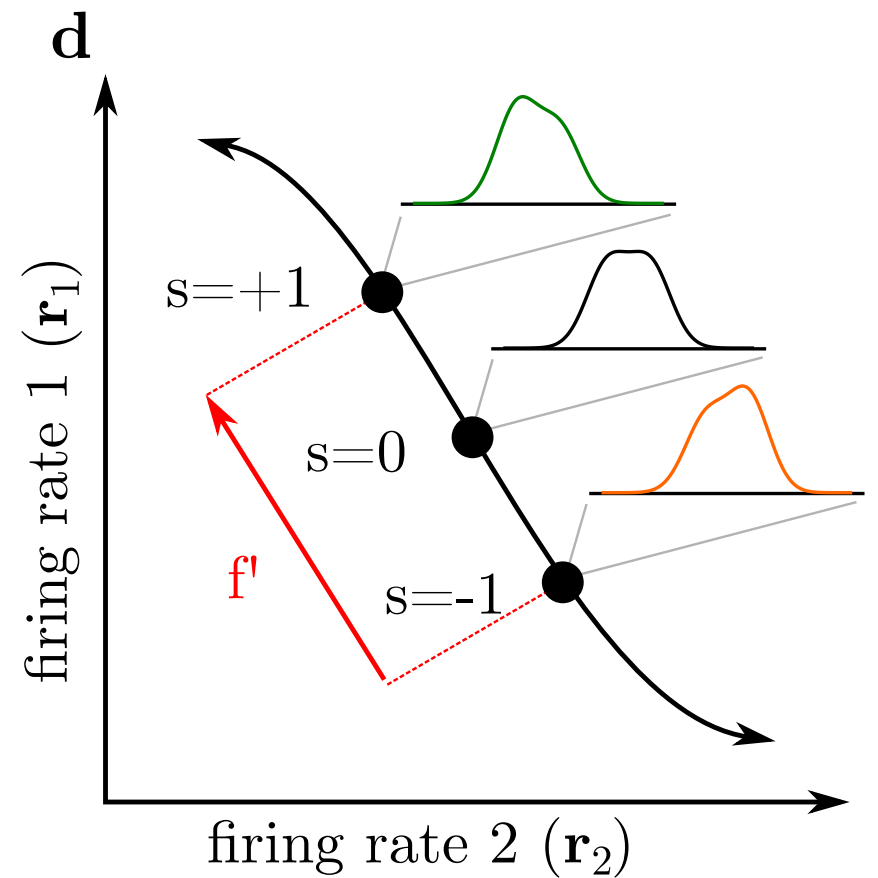
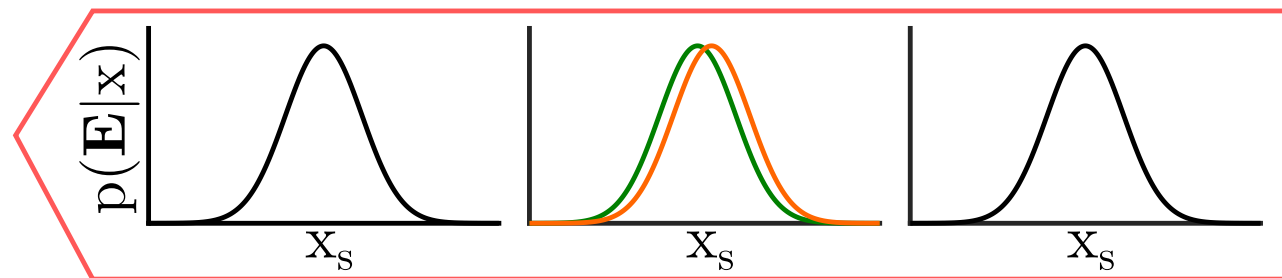
sensory  
prior



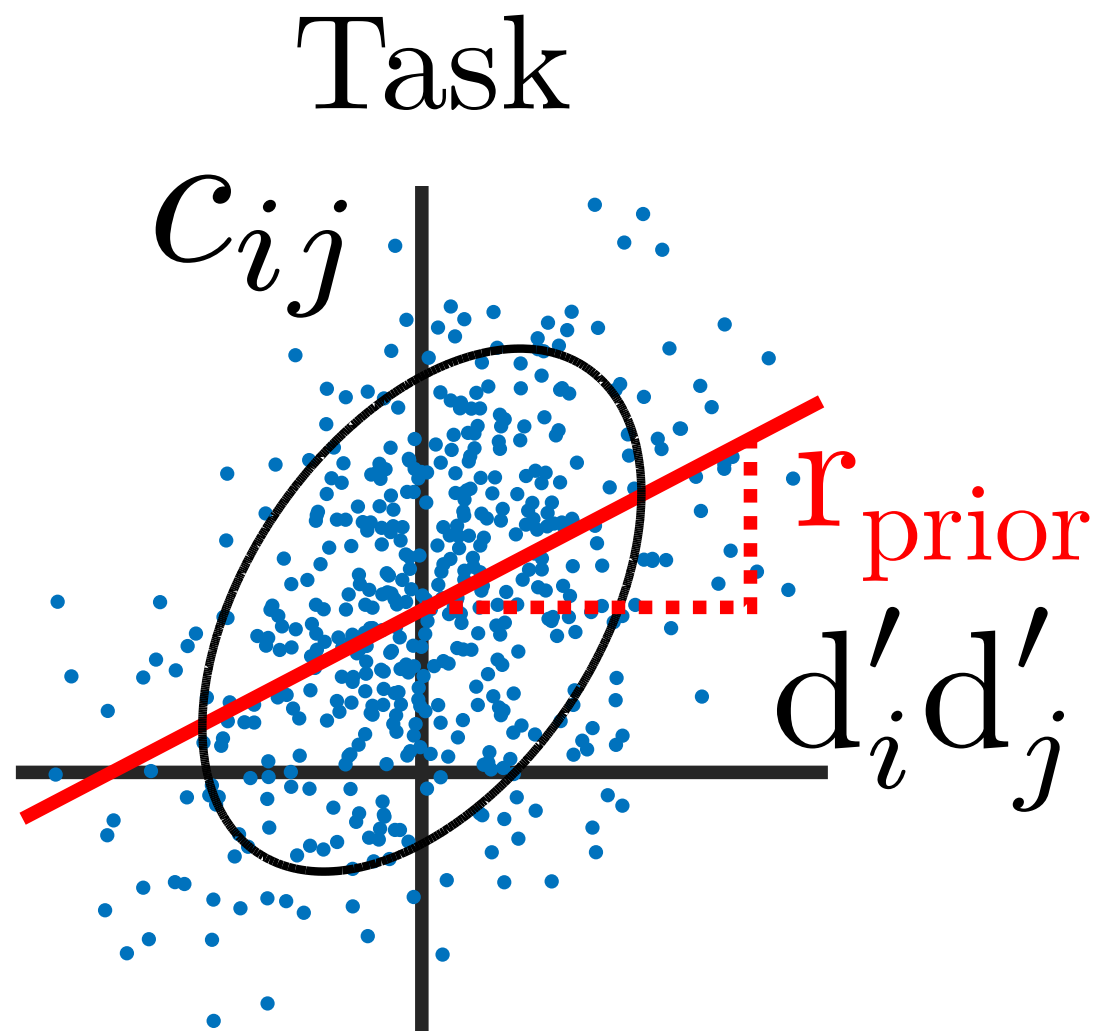
posterior



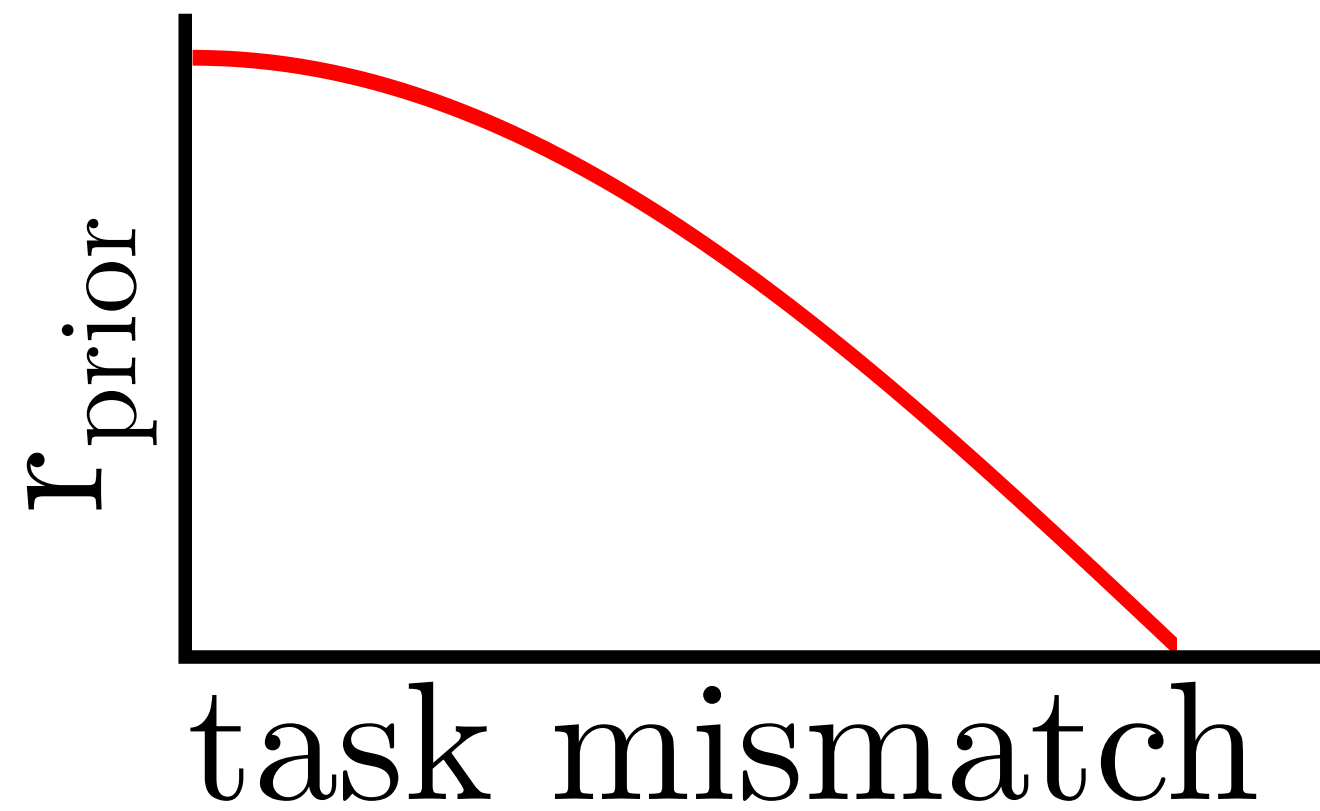
likelihood



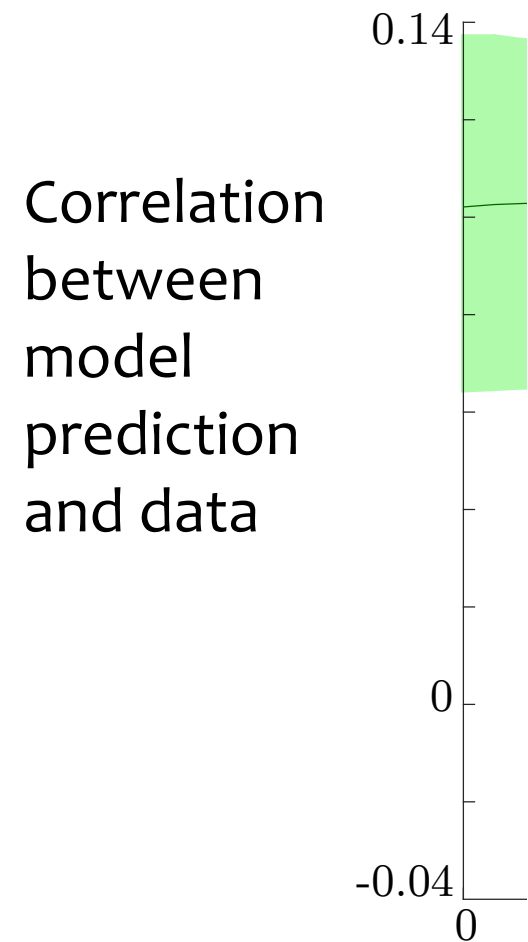
Differential correlations (proportional to  $f'f'$ )



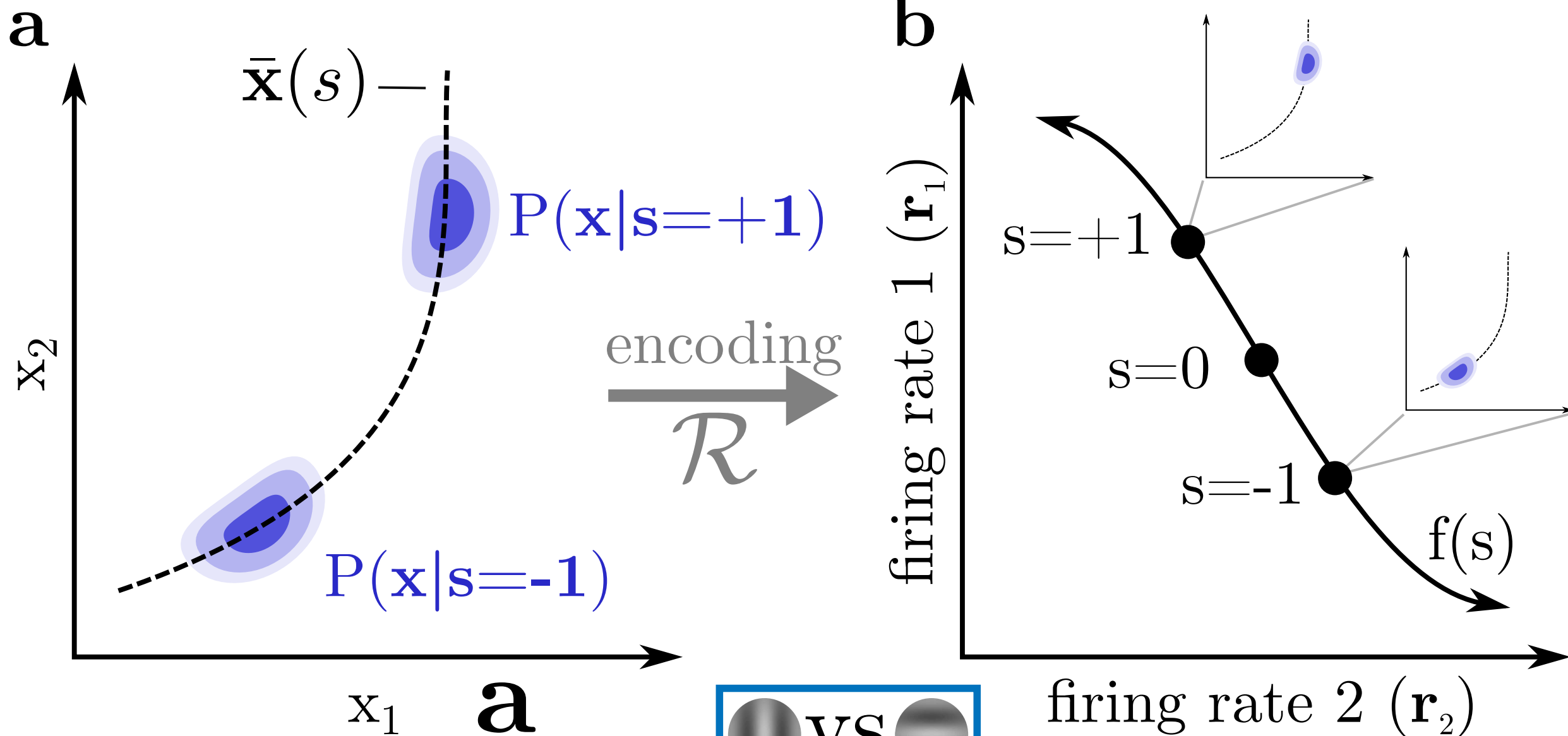
**e**



# Empirical test of 2nd order prediction



Difference between actual task performed by monkey and task for which prediction was made



$x_1$  **a**

