

# Representation of high bandwidth sensory information by small circuits in the whisker system

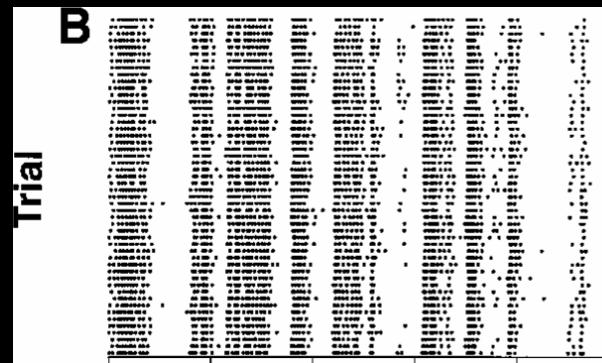
Rasmus Petersen

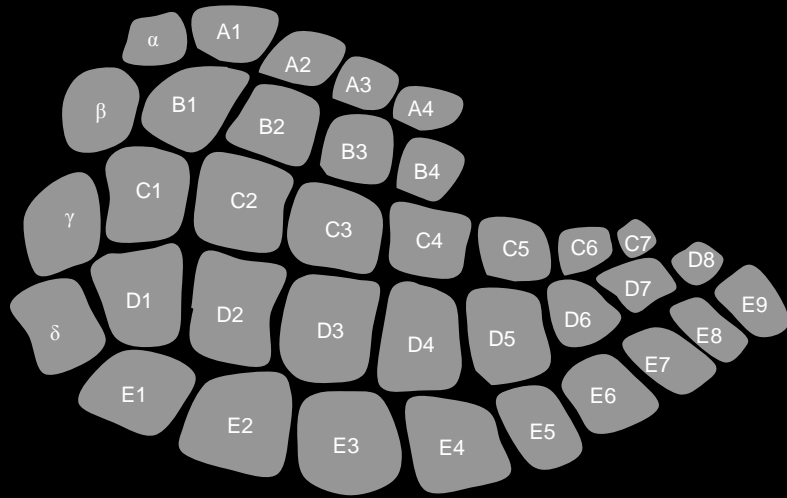
# Two views of neural coding

- Low bandwidth  
e.g., visual cortex

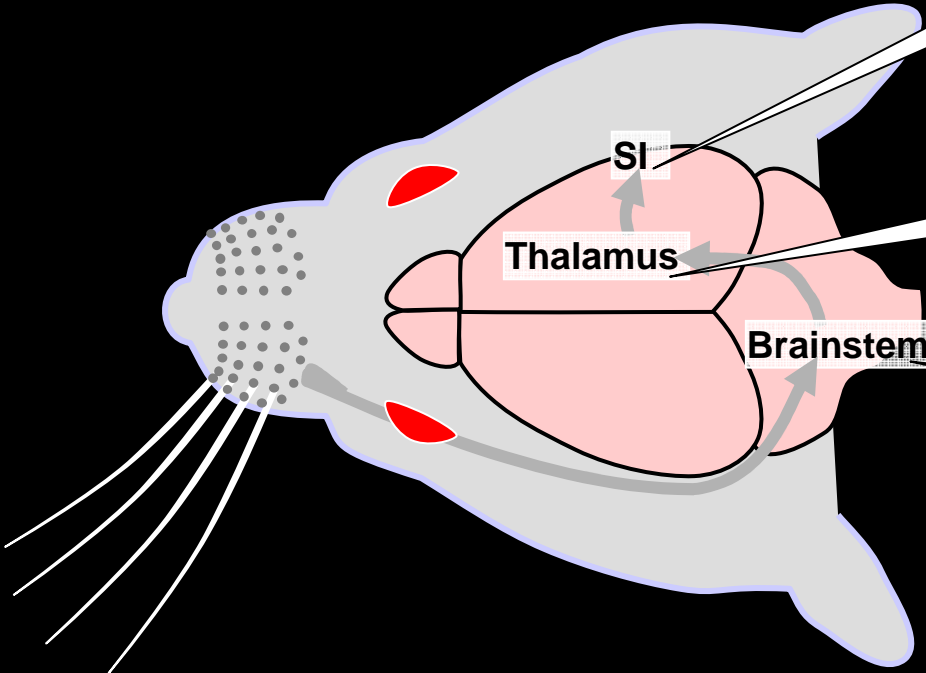


- High bandwidth  
e.g., fly H1



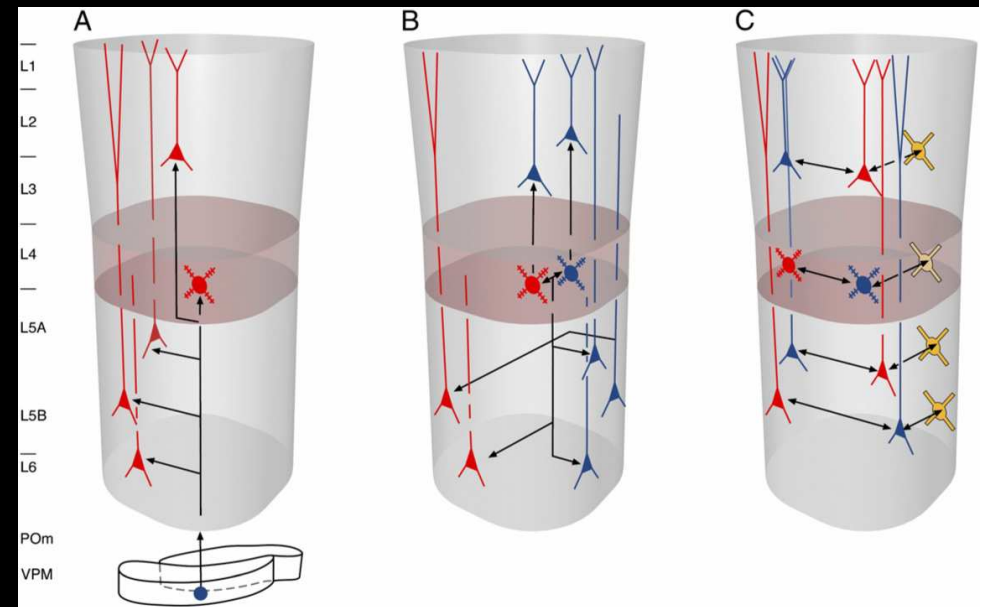
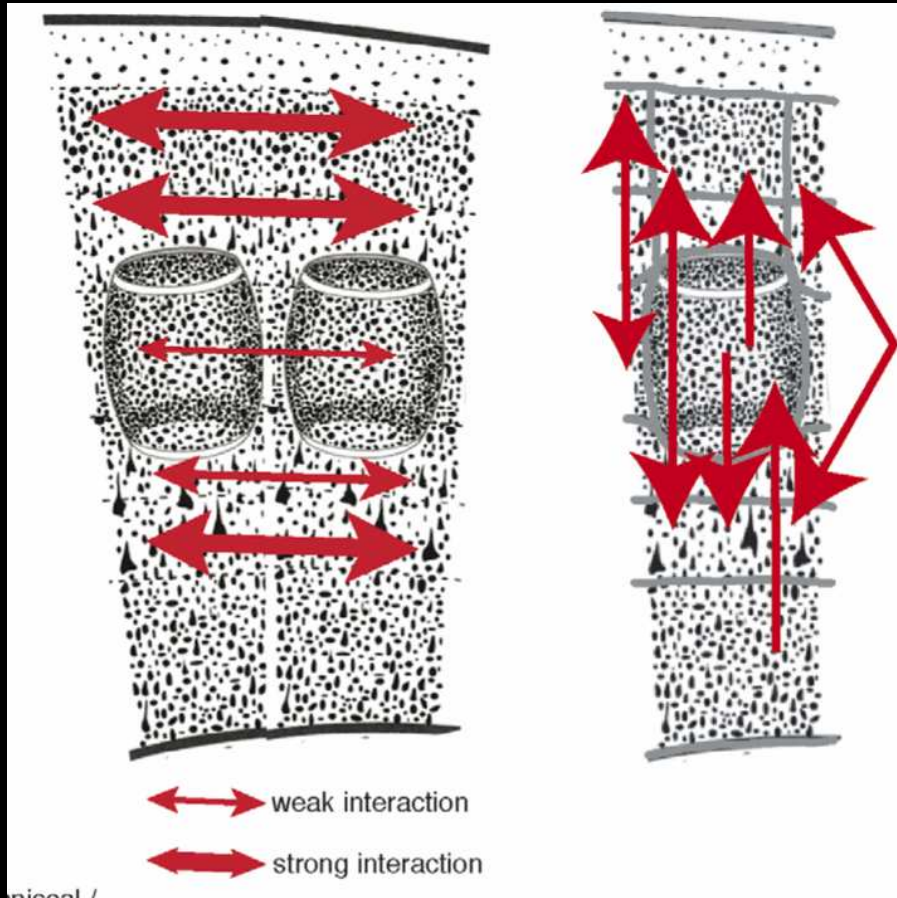


barrels



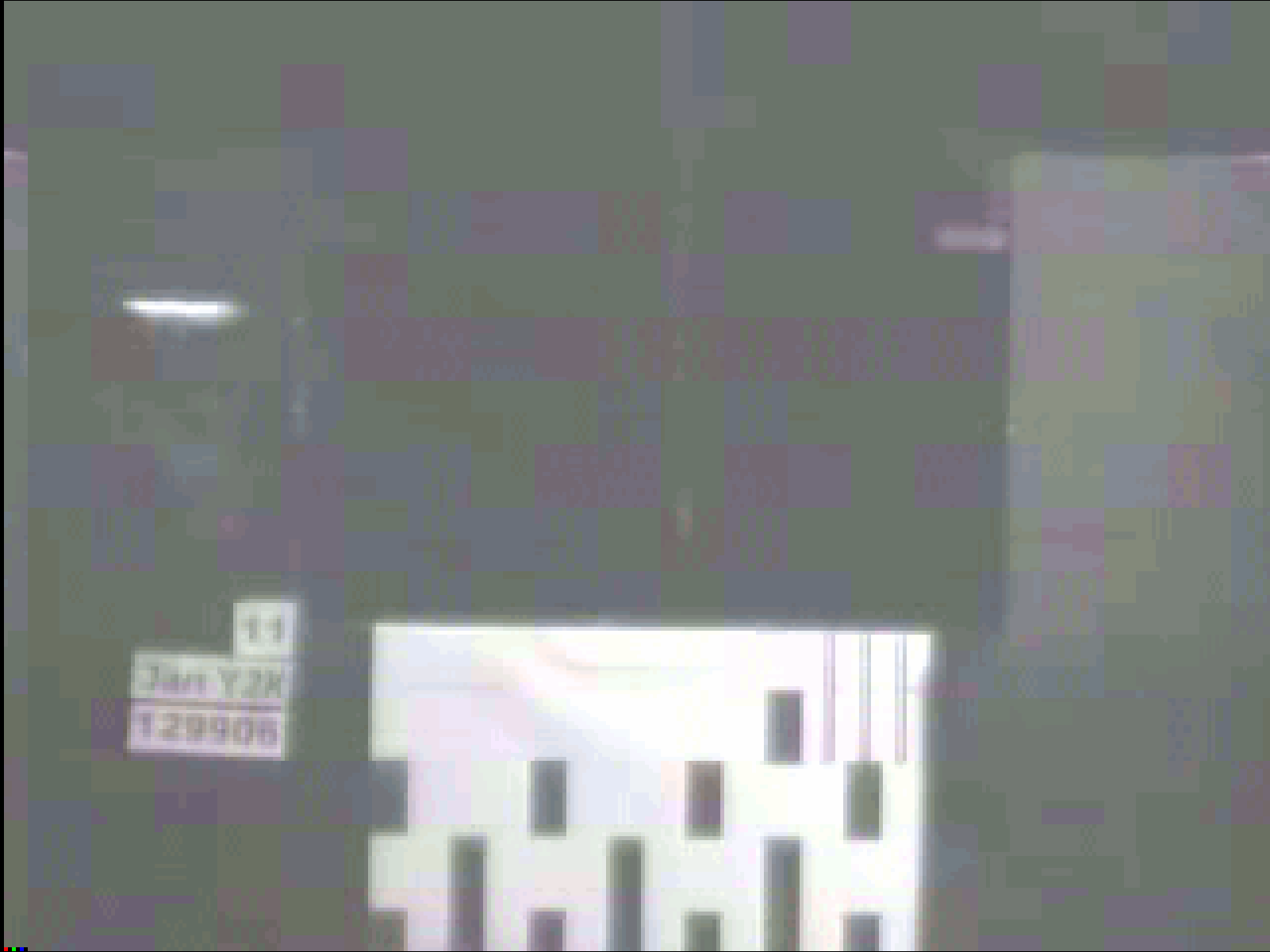
barreloids

barrelettes

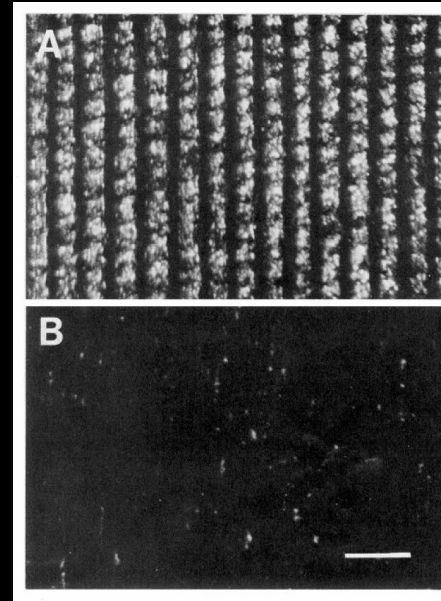
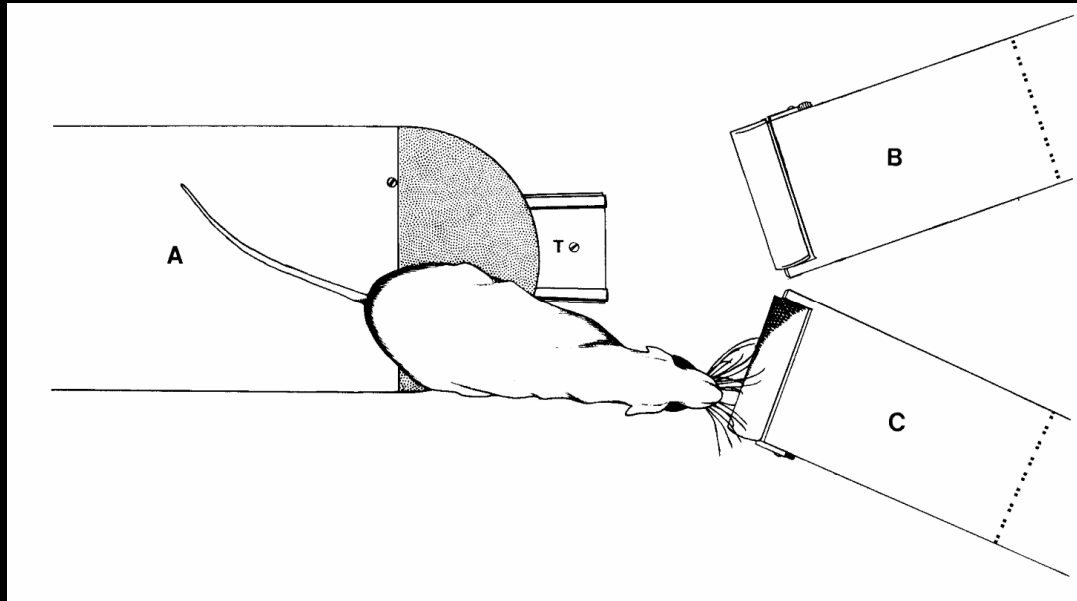


Brecht (2007) COIN

Helmstaedter et al. (2007) Br Res Rev

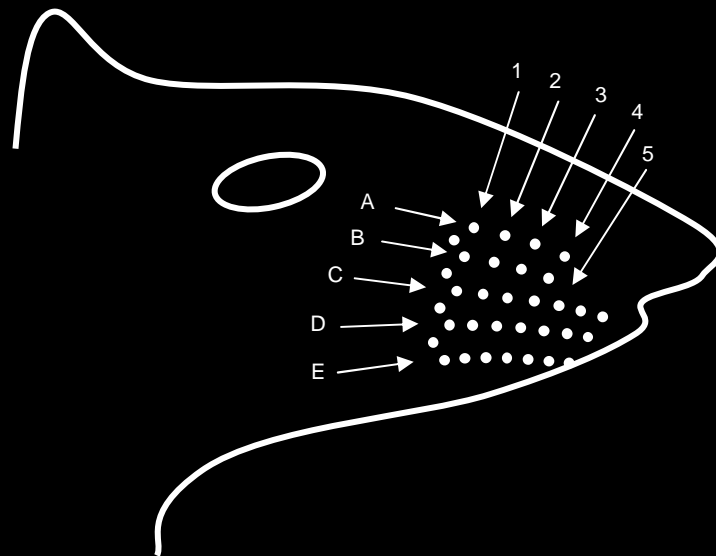


Tansu Celikel

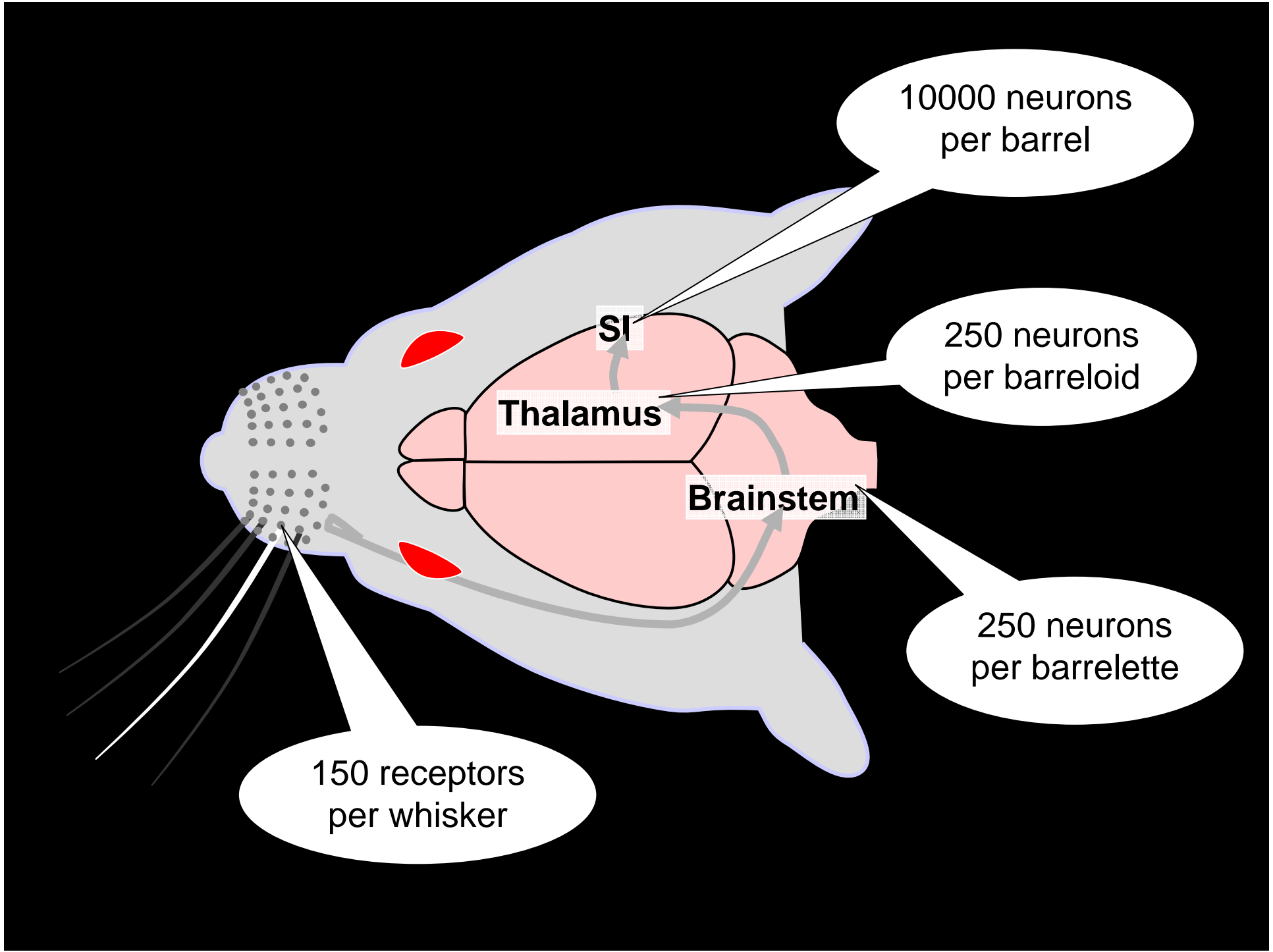


30 micron  
grooves

smooth



Single whisker discrimination



10000 neurons  
per barrel

SI

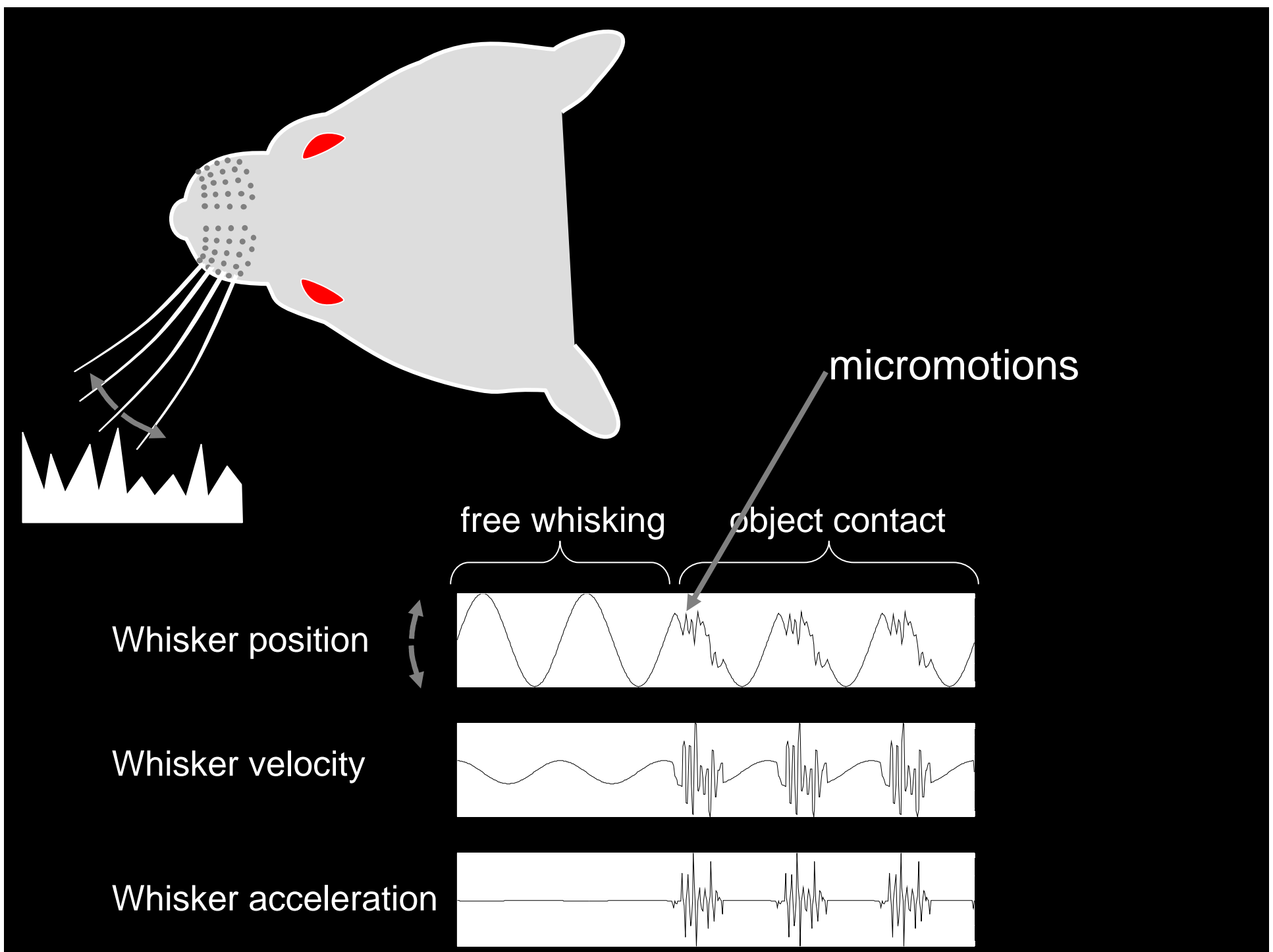
250 neurons  
per barreloid

Thalamus

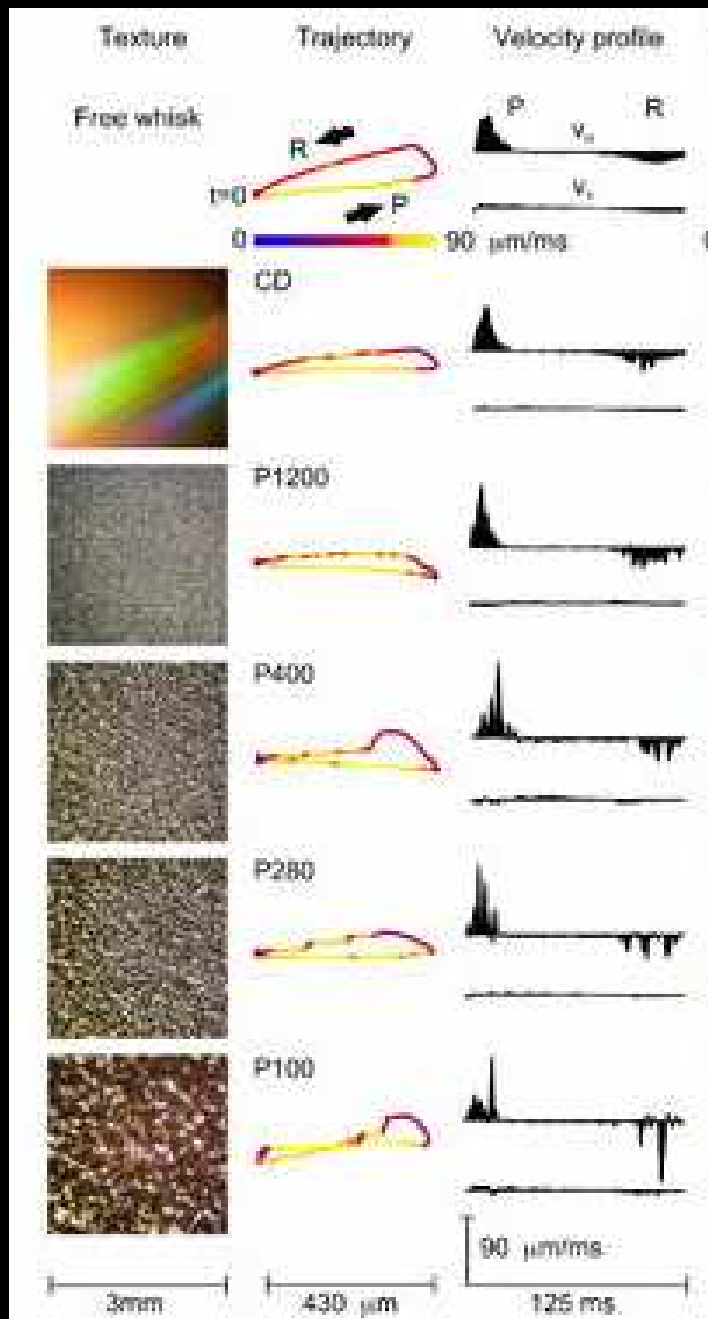
Brainstem

250 neurons  
per barrelette

150 receptors  
per whisker

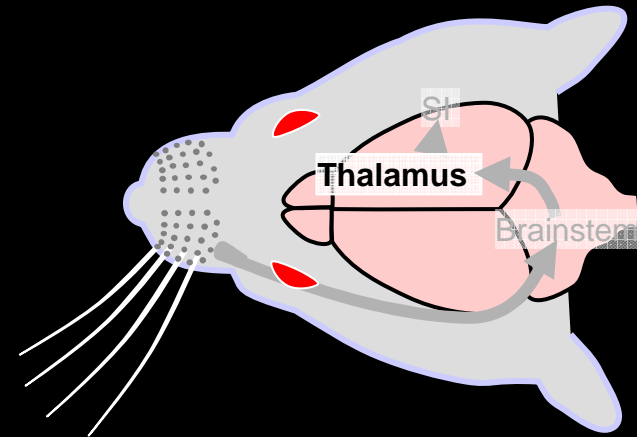
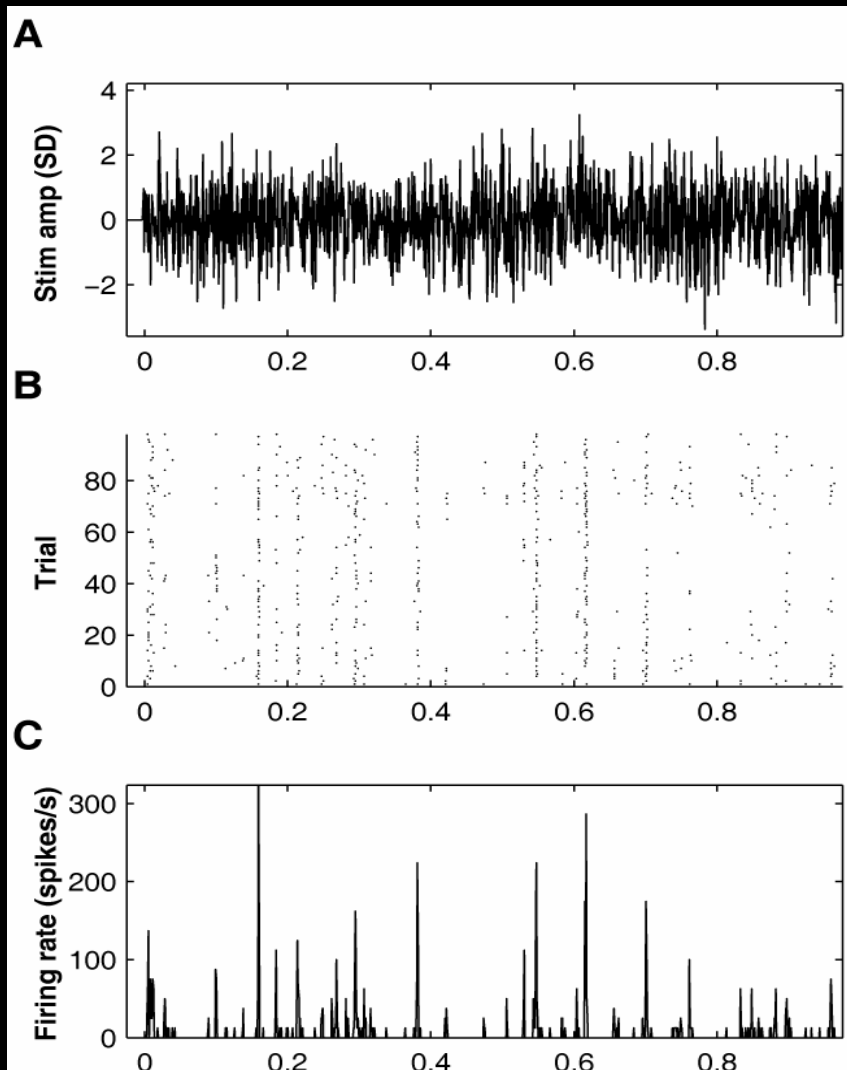






- Experimentally measured micromotions
- Importance of temporal domain

# Coding temporal signals



## 1. Response question:

What is the 'information-bearing element' of the neuronal response?

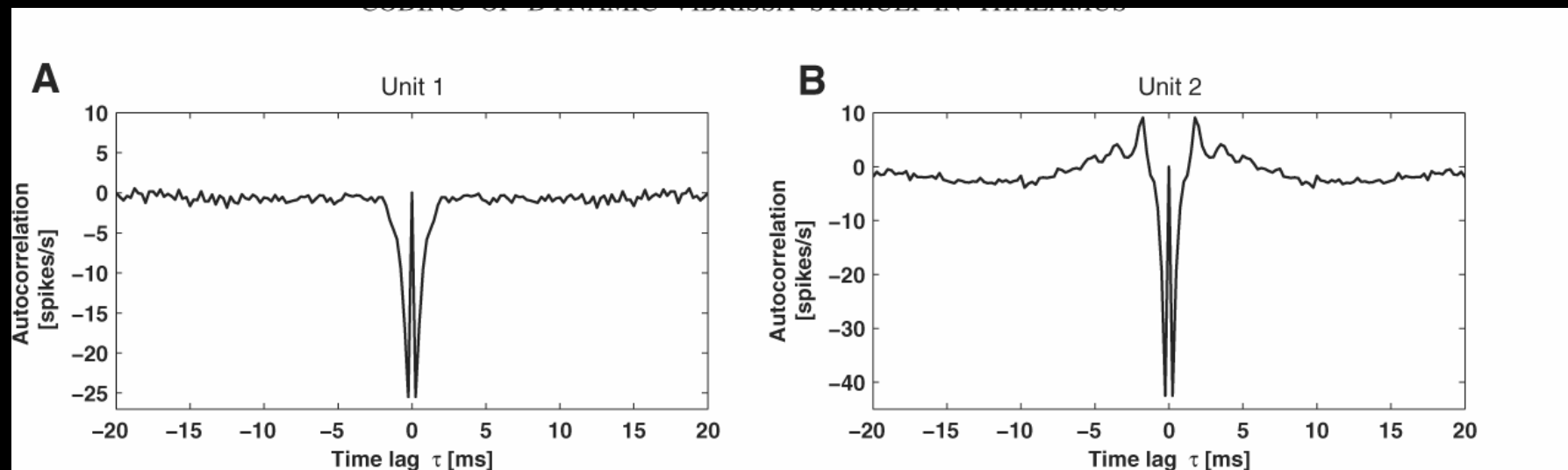
## 2. Stimulus question:

What stimulus features do the information-bearing elements signify?

# The information-bearing element of a spike train

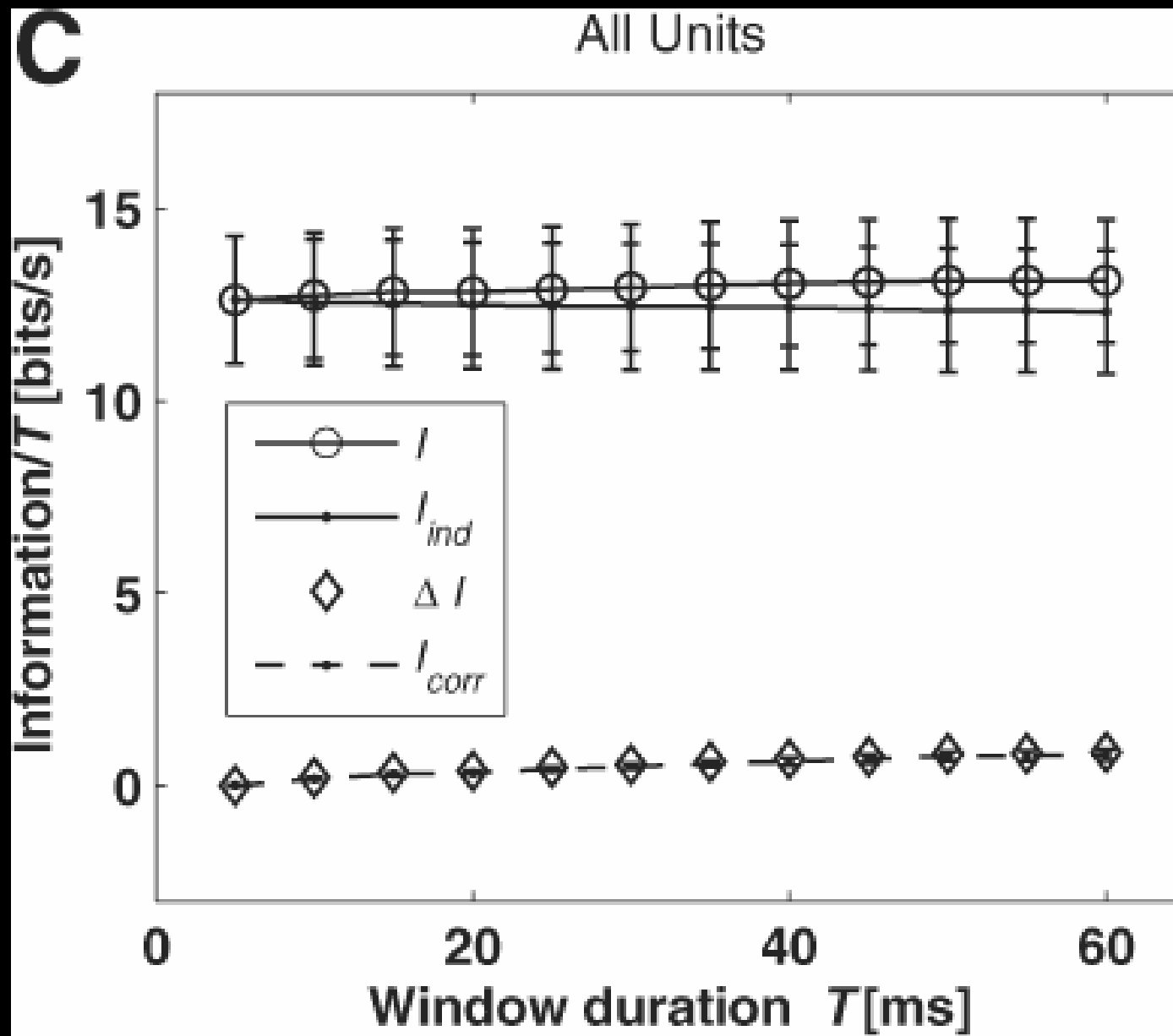
Hypothesis 1. Firing rate

Hypothesis 2. Correlations between spikes



# Firing rate vs temporal correlations

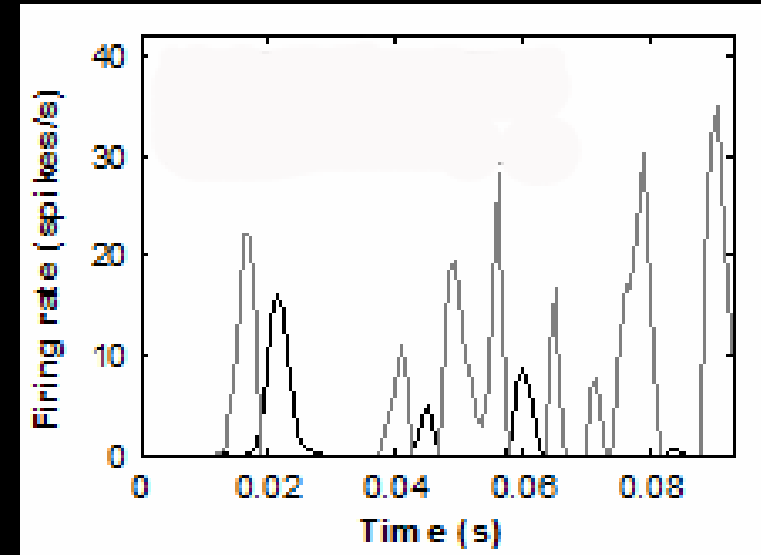
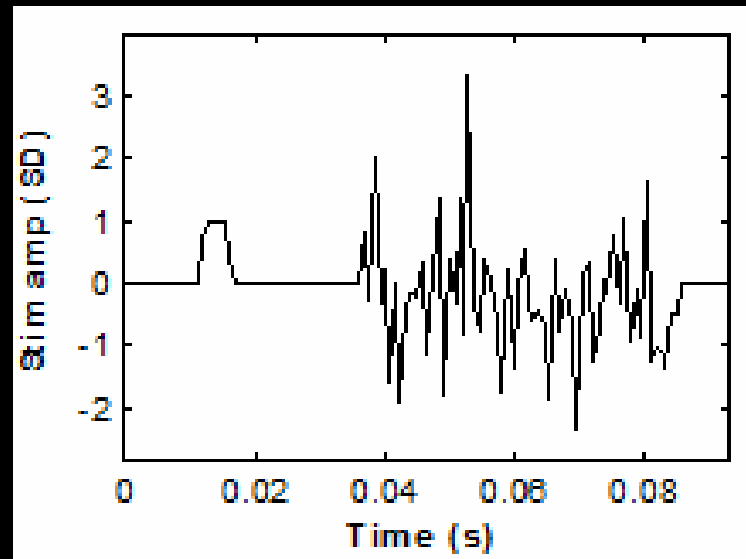
- **Effect of correlations on encoded information ( $I_{\text{corr}}$ )**  
How much does the presence of noise correlations change the information compared to a hypothetical neural system where the neuron has identical mean firing rate but zero noise correlations? (eg Schneidman et al., 2003)
- **Effect of correlations on decodable information ( $\Delta I$ )**  
If a hypothetical downstream decoding circuit attempts to decode using a model that includes mean firing rates but ignores noise correlations, how much information will be lost? (Nirenberg et al., 2001)

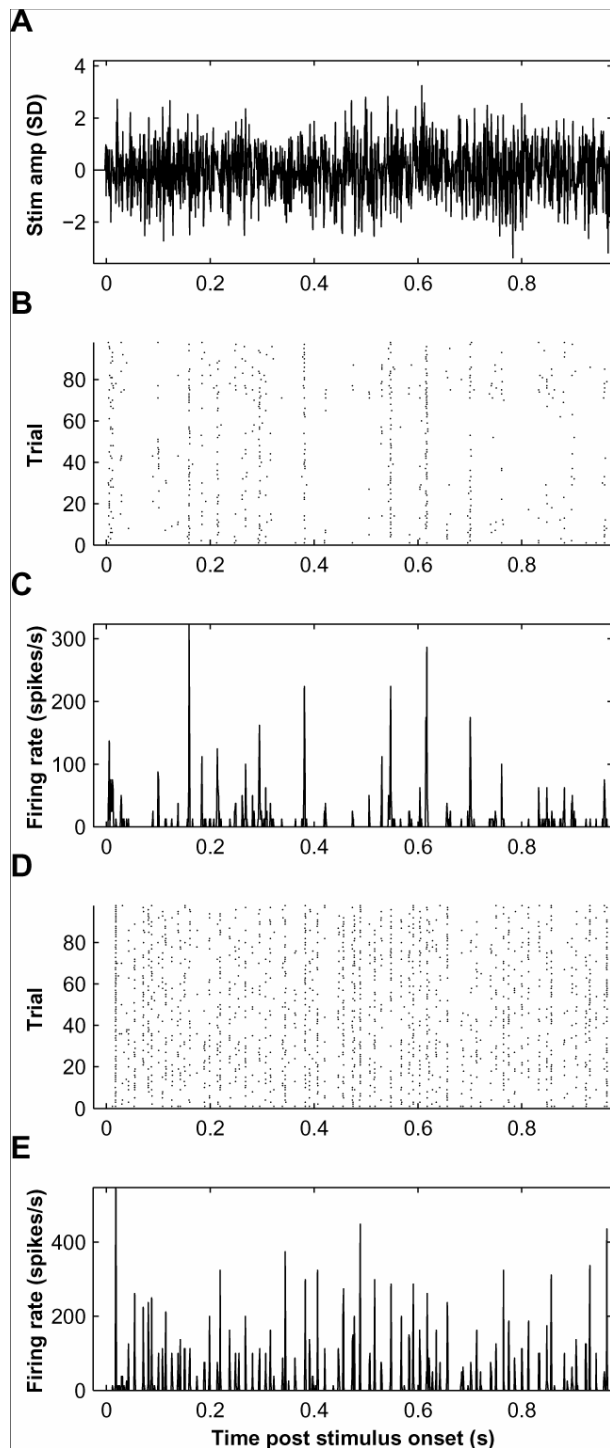


# Summary

- For VPM neurons stimulated by white noise, key “information-bearing element” is firing rate
- For most VPM neurons, higher order temporal correlations have minor impact

# What stimulus features does the firing rate signify?





## Approaches:

- Pseudolinear model:
  - Spike-triggered averaging
- Non-linear models:
  - Spike feedback (Paninski) model
  - Spike-triggered covariance analysis



# Linear Nonlinear Poisson (LNP) Model

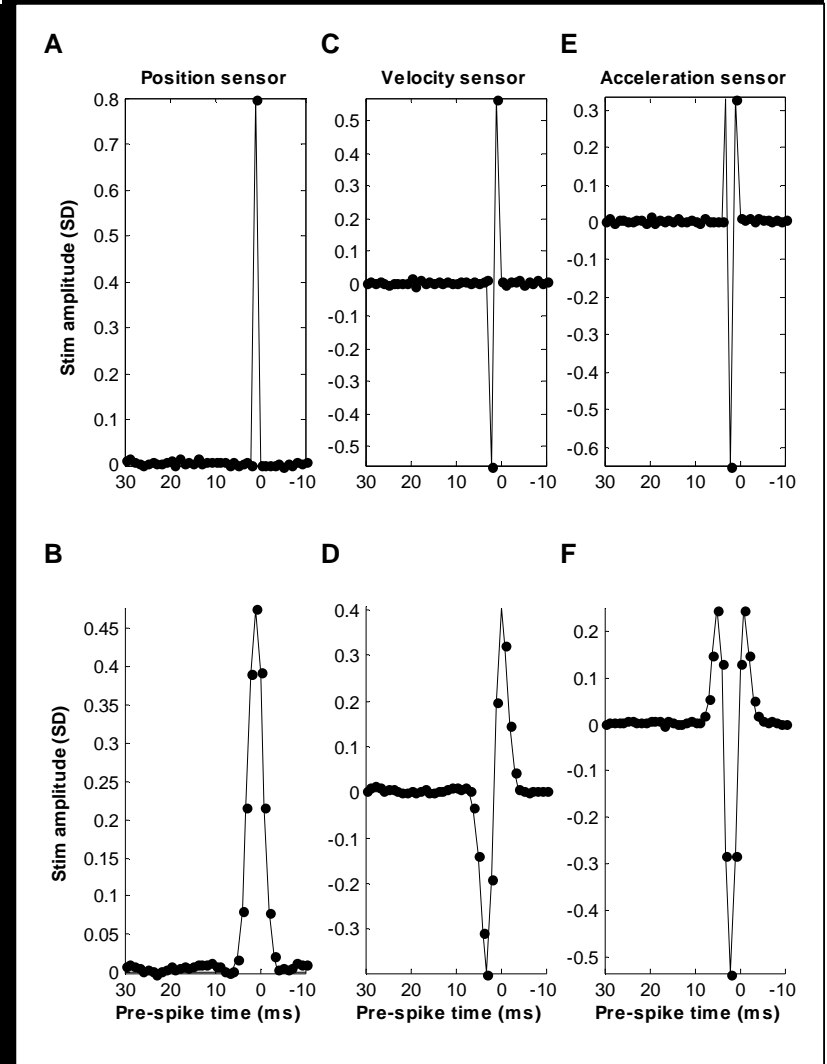
$$k(t) = \sum_{i=0}^{T/\delta t} s(t - i\delta t) f(i\delta t)$$

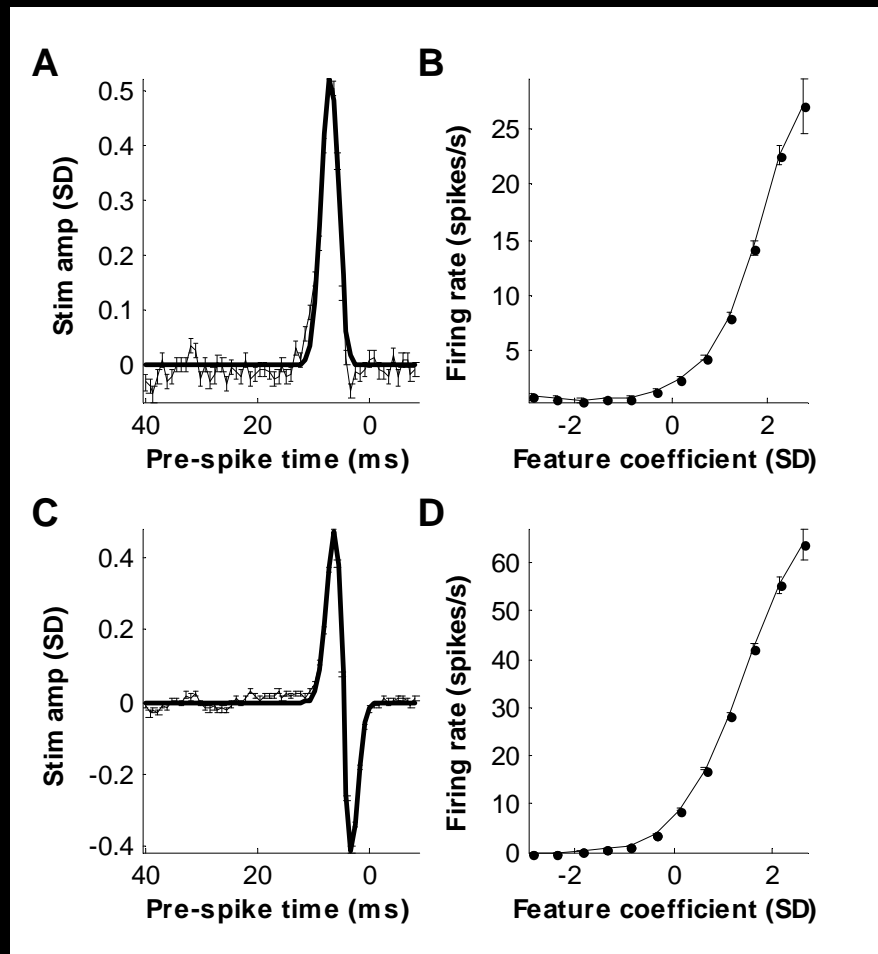
$$r(t) = g[k(t)]$$

Position  $r(t) = g[s(t)]$

Velocity  $r(t) = g[s(t) - s(t - \delta t)]$

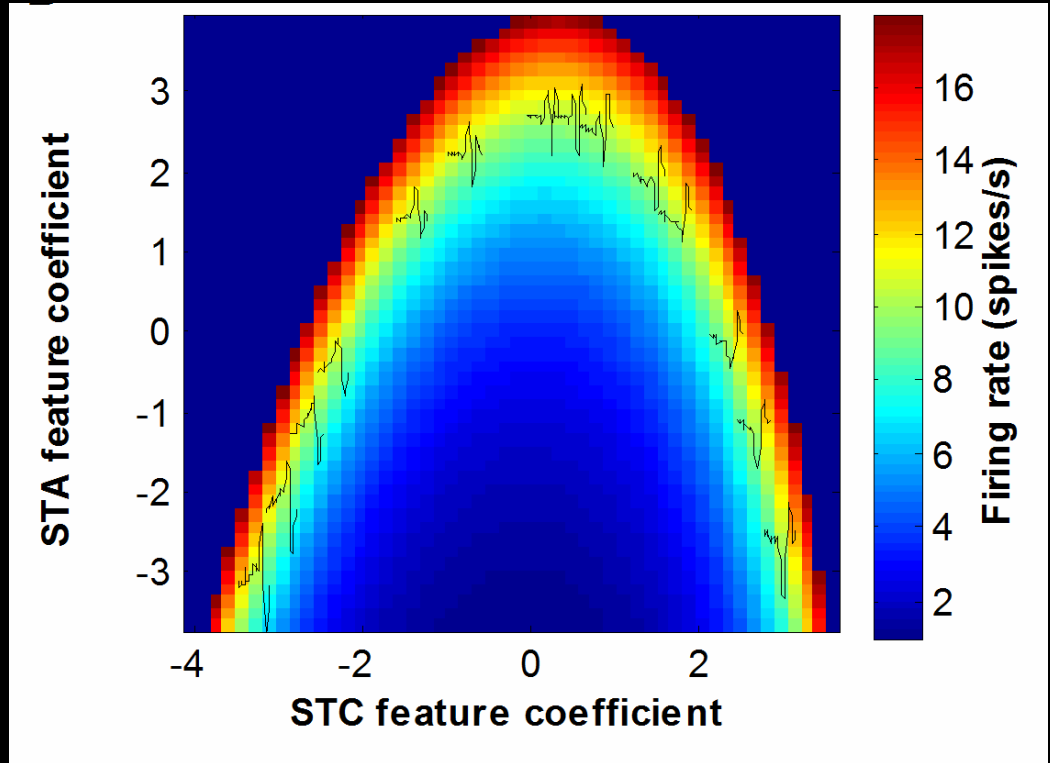
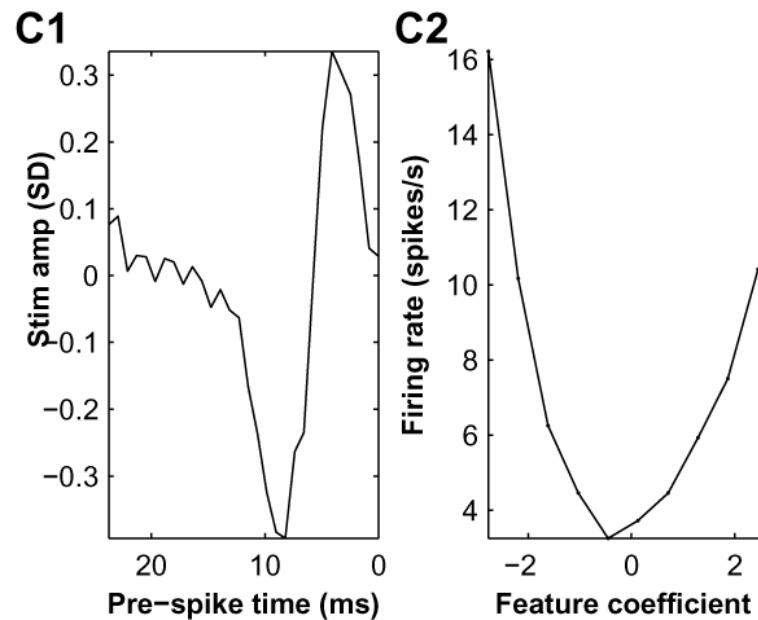
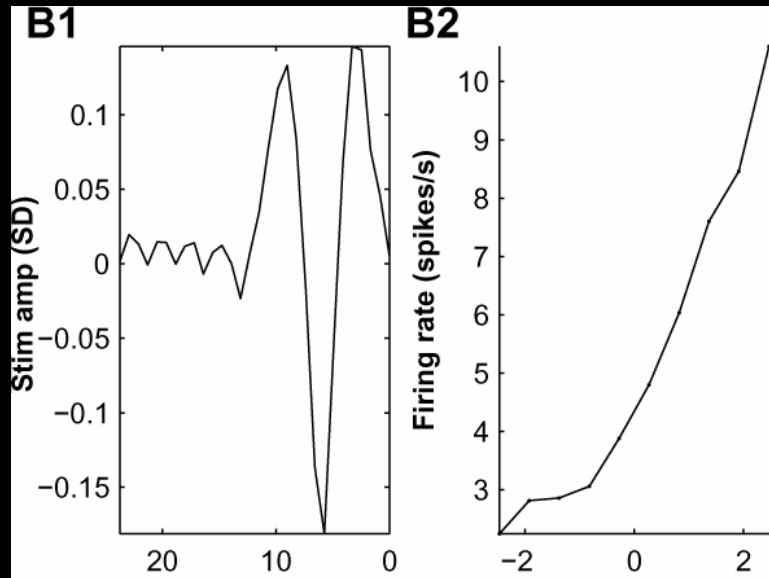
Acceler.  $r(t) = g\left[\frac{1}{2}s(t) - s(t - \delta t) + \frac{1}{2}s(t - 2\delta t)\right]$

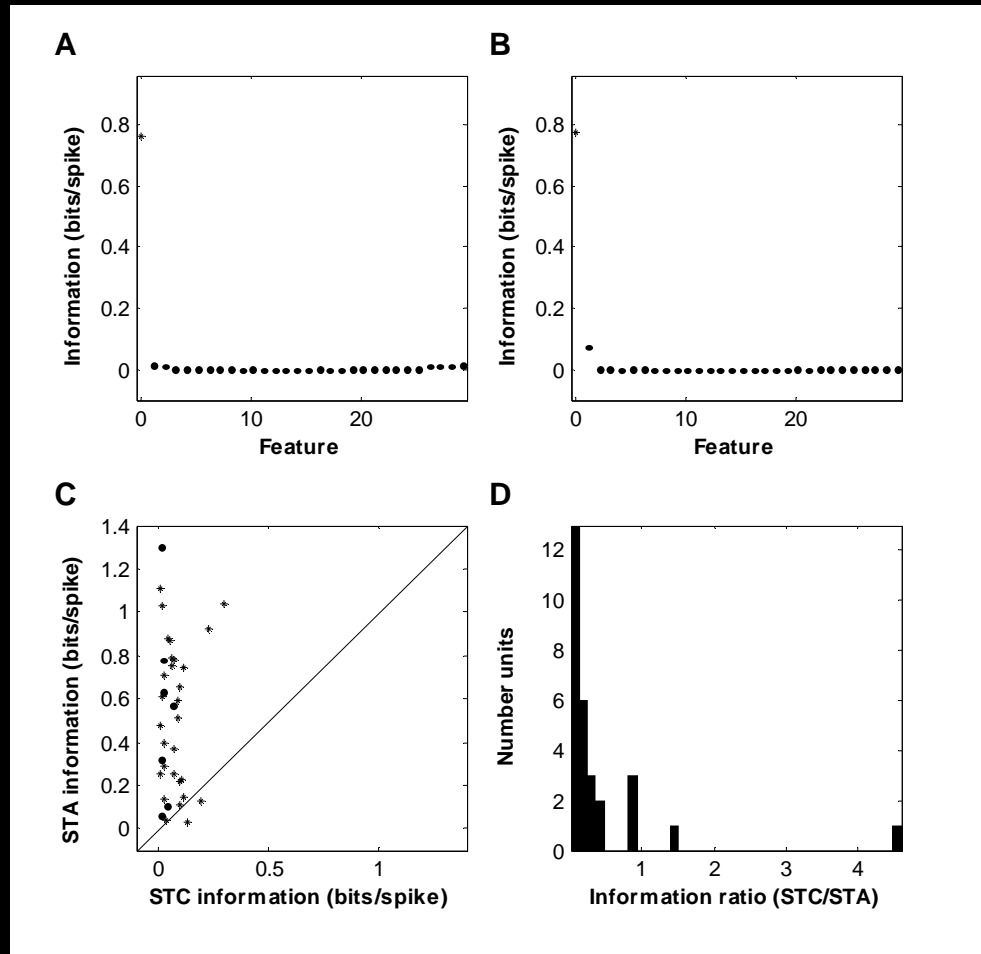




# How accurate is the LNP model?

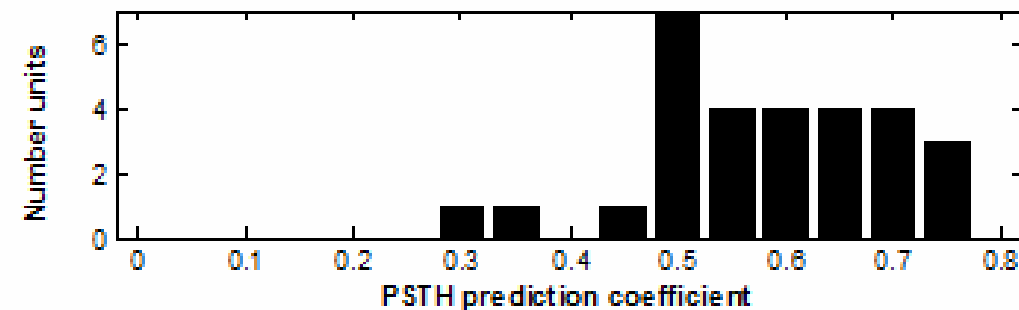
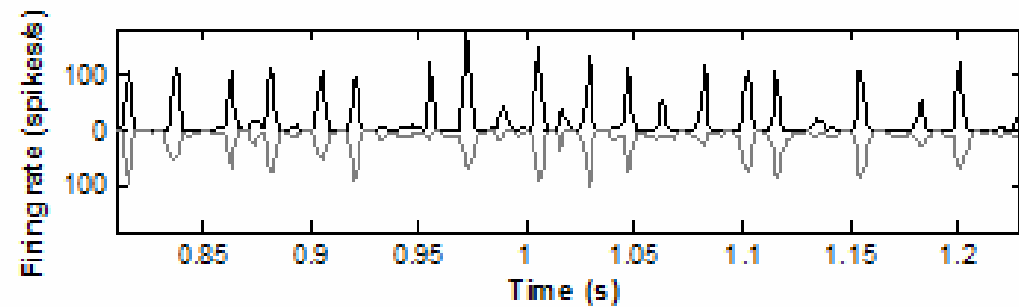
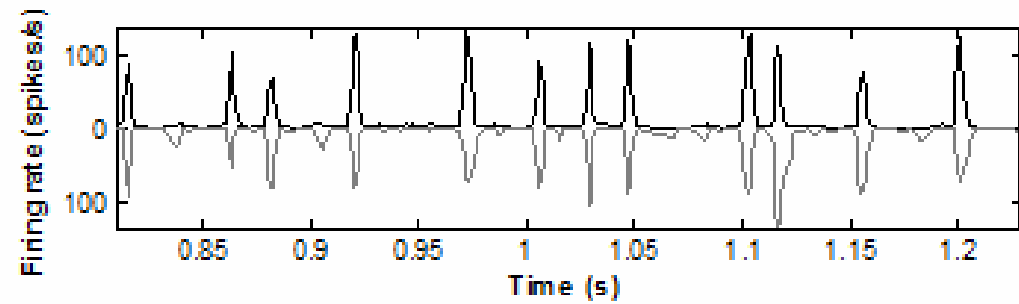
- Non-linear approaches:
  - Maximum likelihood fitting of spike feedback (Paninski) model
  - Spike-triggered covariance analysis

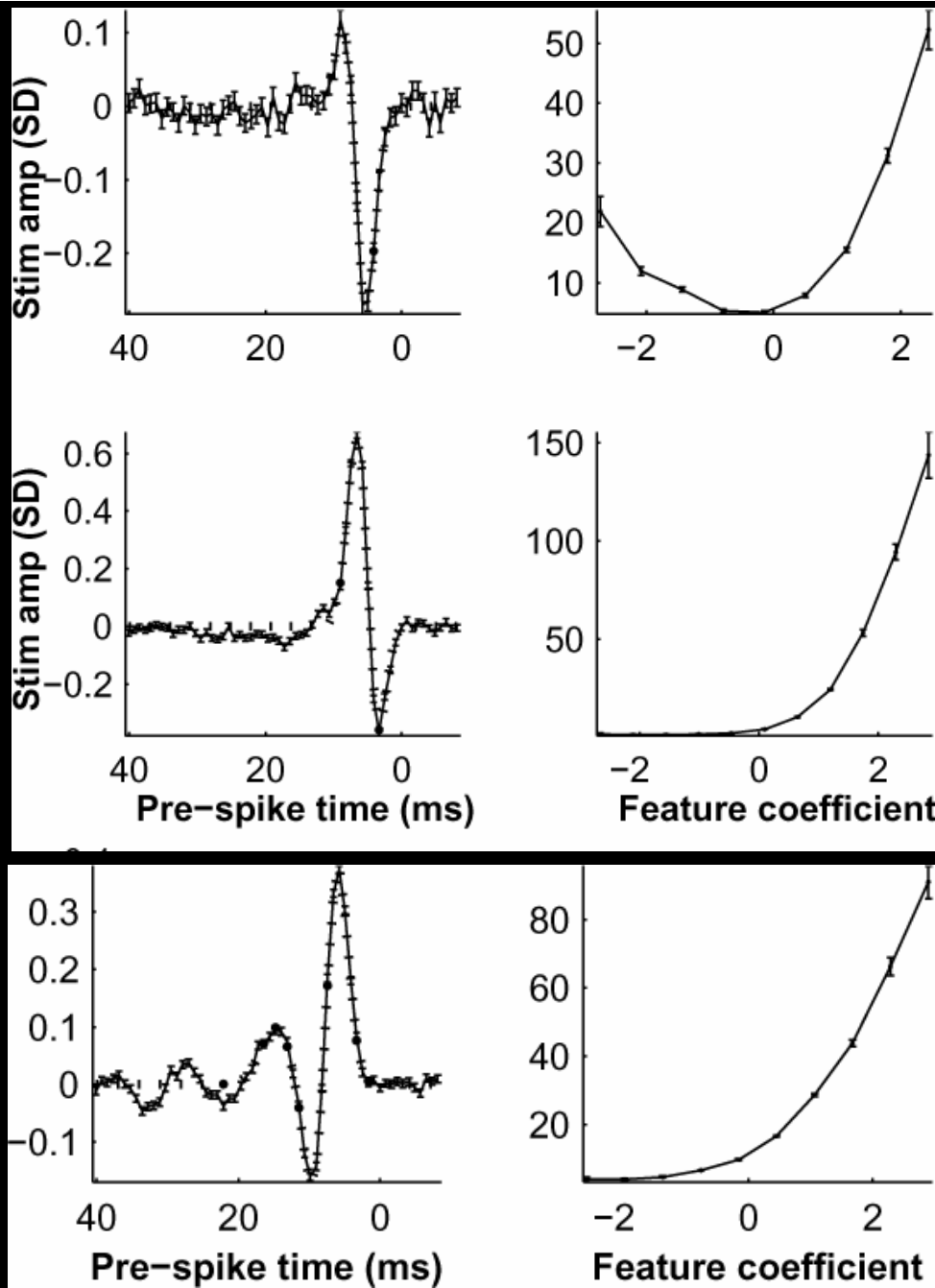


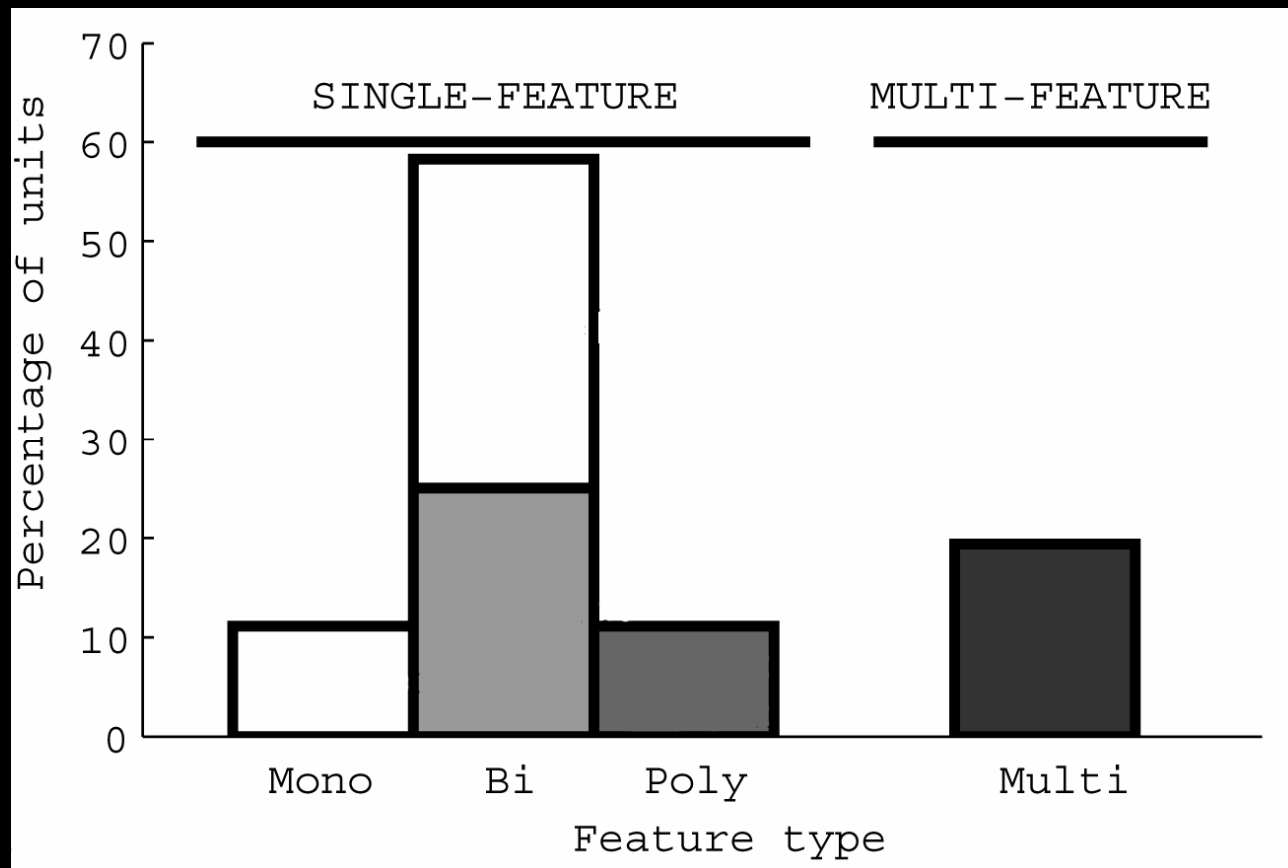


### LNP model

$$k(t) = \sum_{i=0}^{i=T/\delta t} s(t - i\delta t) f(i\delta t)$$
$$r(t) = g[k(t)]$$









# Conclusions

- VPM neurons reliably encode complex whisker motion
- Key information-bearing element of response is firing rate on millisecond time-scale
- High reliability permits small subcortical circuits to support demanding sensory processing
- Most VPM neurons can be described by LNP models
- VPM exhibits a distributed code for whisker motion

# Acknowledgements

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EXTRA

