Characterizing diverse neuronal dynamics in sensory circuits



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The sensory decoding problem

- Same distal object, many proximal representations
- Example: phonetic perception



- Misleading or absent segmentation
- Coarticulation: context effects from anticipatory configuration of vocal tract
- Variations in pitch and duration (prosody, different speakers)
- Noise and interference (cocktail party problem)

The sensory decoding problem

 Different distal objects, similar proximal representations



Liberman et al (1961)

Towards biophysical models of sensory decoding

- 1. Songbirds as a model for speech decoding
- 2. Diverse dynamics in auditory circuits
- 3. Combining sensory response models with dynamical neuron models to understand the function of neuronal diversity



Songbirds as models for auditory and vocal learning

- Songbirds imitate an adult tutor heard in a critical period
- Juveniles refine their imitation of the tutor through sensorimotor learning
- Production and perception are tightly linked









The songbird auditory pathway



• Auditory stimuli are processed hierarchically



The songbird auditory pathway





 Auditory stimuli are processed hierarchically



Woolley (2012)

Circuits for sensory decoding

Convergent and recurrent networks



Models for neural computation

A very simple neuron model



gives the linear-nonlinear-Poisson (LNP) RF model



adapted from Desbordes et al. (2010)

Intracellular electrophysiology in CM









McCormick et al (1985)

Zebra finch CM neurons are phasic







Juveniles



Chen & Meliza (in revision)



Do dynamics matter?

• Phasic neurons are tuned to faster modulations







Dynamical receptive field models



Stimulus enters the dynamics as a driving current:

$$C_{m} \frac{dV}{dt} = g_{L}(E_{L} - V)$$

$$+ g_{Na}(t)(E_{Na} - V)$$

$$+ g_{K}(t)(E_{K} - V)$$

$$+ \cdots$$

$$+ I_{stim}(t)$$

$$I_{stim}(t) = \sum_{f} \sum_{\tau} h(f, \tau) \cdot s(f, t - \tau)$$

Phenomenological neuron models

Multi-timescale Adaptive Threshold (MAT) model

$$\tau_m \frac{dV}{dt} = -V + RI(t)$$

$$\theta(t) = \sum_k H(t - t_k) + \omega$$

$$H(t) = \sum_{j=1}^L \alpha_j \exp(-t/\tau_j)$$

$$\mathbf{x}(t) = (V(t), \theta(t))$$

$$\mathbf{p} = (\tau_m, R, \omega, \alpha_1, \tau_1, \dots, \alpha_L, \tau_L)$$



MAT model firing patterns and parameters



ß

Kobayahsi et al (Front Comp Neurosci 2009) Yamauchi et al (Front Comp Neurosci 2011)

Intrinsic dynamics matter!

5 stimuli @ 2 s

ms

120 140

I_noise \rightarrow 60% sync

True Filter





Characterizing neural diversity with DA

• Method: extracellular recordings from living birds







MAT model assimilated to spike times







dSTRF twin experiments



Tyler Robbins









dSTRF twin experiments





Phasic



2 4500 ms r = 0.79



Tonic

dSTRF of zebra finch neuron



Diverse dynamics from extracellular data



Conclusions

- Data assimilation of dynamical models is a tool for characterizing sensory neural circuits
- Sensory-evoked extracellular responses *alone* convey information about dynamical properties of neurons

and Future directions

- Circuit models
 - . inferring network connectivity
 - predicting/interpreting optogenetic manipulations
- Biophysical models
 - predicting/interpreting pharmacological effects

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