Week 6 – part 5 : Rate Codes versus Temporal Codes



Neuronal Dynamics: Computational Neuroscience of Single Neurons

Week 6 – Noise models:

Escape noise

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6.1 Escape noise

- stochastic intensity and point process
- **√ 6.2 Interspike interval distribution**
 - Time-dependend renewal process
 - Firing probability in discrete time
- √ 6.3 Likelihood of a spike train
 - generative model
- **√** 6.4 Comparison of noise models
 - escape noise vs. diffusive noise
 - 6.5. Rate code vs. Temporal Code
 - timing codes
 - stochastic resonance

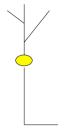
Week 6 – part 5 : Rate Codes versus Temporal Codes

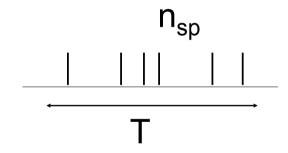


- 6.1 Escape noise
 - stochastic intensity and point process
- **4** 6.2 Interspike interval distribution
 - Time-dependend renewal process
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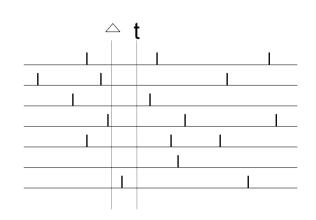
Neuronal Dynamics – 6.5 Rate codes versus temporal codes

3 rate codes



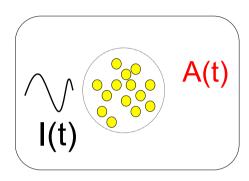


Temporal averaging



$$PSTH(t) = \frac{n(t; t + \Delta t)}{K \Delta t}$$

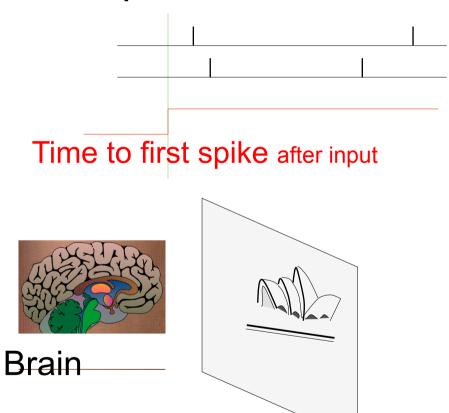
Trial averaging



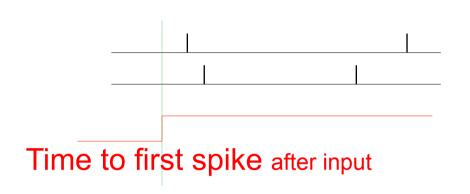
$$A(t) = \frac{n(t; t + \Delta t)}{N\Delta t}$$
population
averaging

Neuronal Dynamics – 6.5. Temporal codes

The problem of neural coding: temporal codes

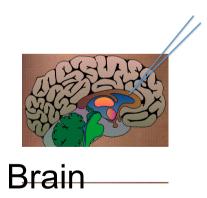


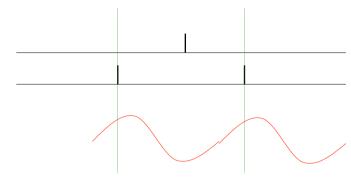
Neuronal Dynamics – 6.5. Temporal codes



Spike timing codes:

- -time-to-first spike
- -phase code





Phase with respect to oscillation

Neuronal Dynamics – 6.5. Stochastic Resonance

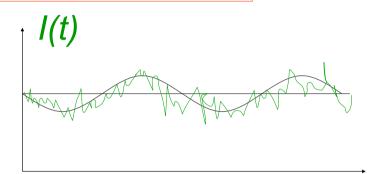
Stochastic Resonance: changing the noise level

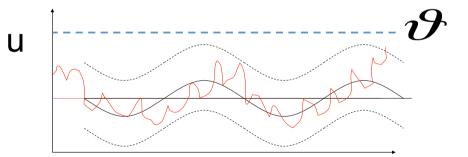
$$I(t) = I_0 \cos(\omega t)$$

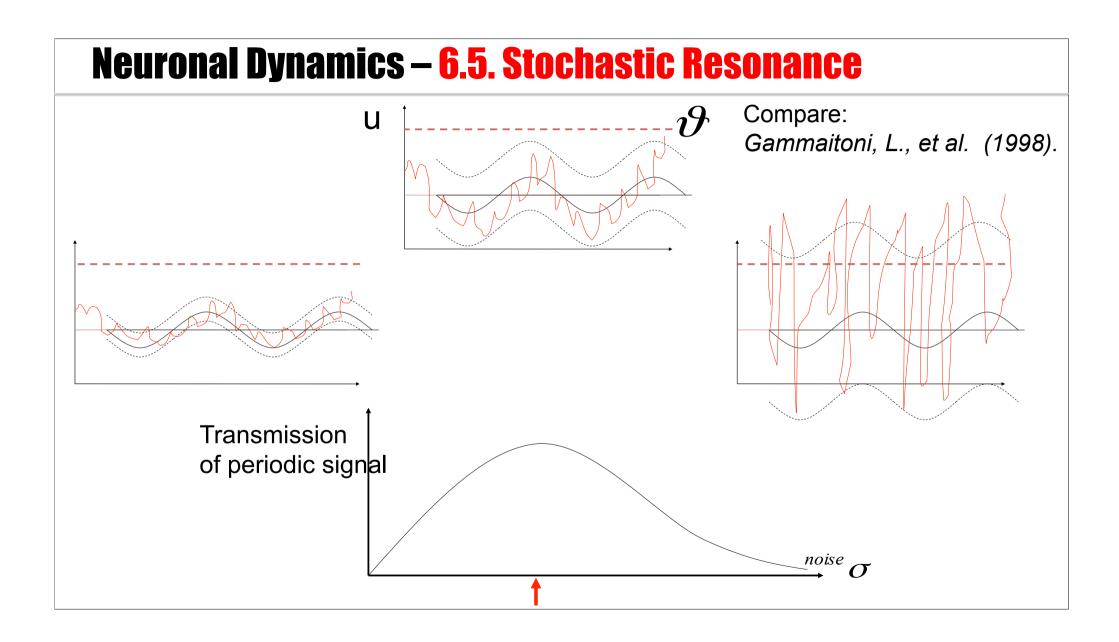
$$I^{noise}(t) = \sigma \xi(t)$$

Sinusoidal input

- + noise
- + threshold







Neuronal Dynamics – 6.5 Rate codes versus temporal codes

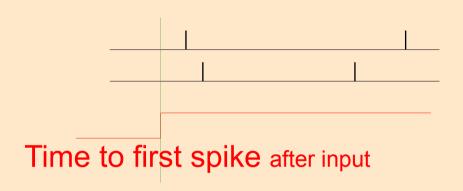
-Rate codes

- population rate

-Temporal Codes

- time-to-first spike
- phase of spike
- stochastic resonance

Neuronal Dynamics – Homework assignment 6.2



With deterministic model
With Poisson model
With noisy IF (escape noise)

Neuronal Dynamics weeks 5+6- References and Suggested Reading

Reading: W. Gerstner, W.M. Kistler, R. Naud and L. Paninski, Neuronal Dynamics: from single neurons to networks and models of cognition. Ch. 7-9: Cambridge, 2014 **OR** W. Gerstner and W. M. Kistler, Spiking Neuron Models, Chapter 5, Cambridge, 2002

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