

Correlated dynamics in spiking neural networks

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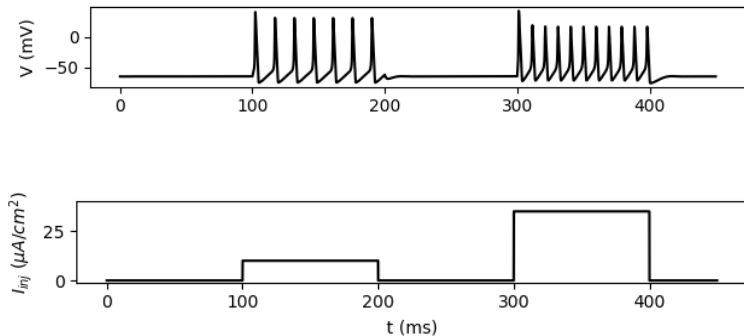
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Outline

Synaptic connectivity as an internal model

References

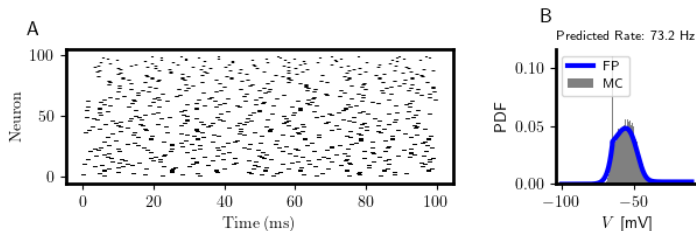
Integrate and fire (IF) neuron models



$$\tau \dot{V}(t) = g_{\ell}(E - V) + g_{\ell} \cdot \psi(V) + I(t)$$

Monte-Carlo simulation of uncoupled IF neurons

When $\psi(V) = g_{\ell} \Delta_{\tau} \exp\left(\frac{V - V_L}{\Delta_{\tau}}\right)$ we have the exponential integrate and fire model

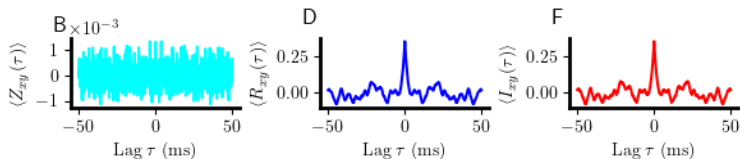


Langevin equations have a corresponding Fokker-Planck equation

$$\frac{\partial P}{\partial t} = \frac{\sigma^2}{\tau} \frac{\partial^2 P}{\partial V^2} + \frac{\partial}{\partial V} \left(\frac{V - E + \psi}{\tau} P \right)$$

Synaptic coupling can induce correlations in spiking activity

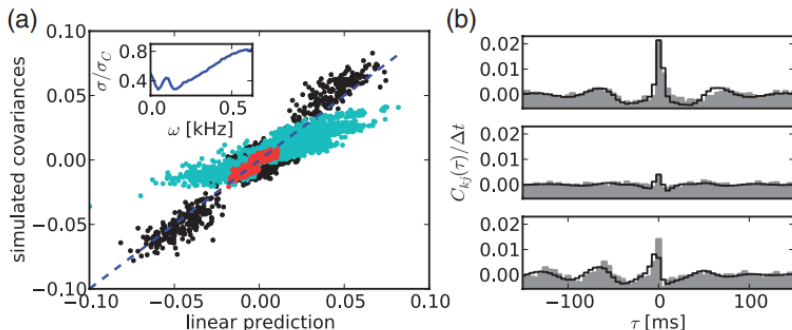
For special synaptic connectivity regimes dynamical variables can remain uncorrelated between neurons



Uncorrelated neural activity captures irregular spiking seen *in-vivo*

Predicting neuron correlations

The linear response of $r(t)$ allows us to also estimate the matrix of cross-correlations $C_{kj}(\tau)$ from the synaptic connectivity \mathcal{C}



This has important implications for brain-inspired machine learning

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