Critical states of slow pattern in neuronal networks

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Outline

- Motivation
 - The Basic Problem That We Studied
 - Previous Work
- Our Results/Contribution
 - Main Results
 - Basic Ideas for Proofs/Implementation

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Network Dynamics

- Network response correspond to specific neuronal parameter, including fire rate, degree of irregularity, spatiotemporal patterns in neuronal spike trains and neuronal critical dynamics.
- Explor the influence of simulation size of neuronal network as well as the community stuctural network. 2000, 5000, 10000, ..., 100 million.
- Synapase density and input heterogeneity.(to be confirmed)

Theoretical explanation

Mainly three aspects...

- Explain the mechanism underly the trainsition dynamics.
 - input current variablity analysis.
 - Mean-filed equation and hopf bifuraction
 - The real part of fixed point is decreasing.
- fit the network response with a simple f unction.

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Biological plausible neuronal network

Consider a balanced E-I network with *N* neurons, in which 80% are excitatory neurons and the others inhibitory ones. Each neuron is equipped with a biological plausible neuronal model, leaky integrate and fire model,

$$C\frac{dV}{dt} = -g_l(V - V_l) + I_{syn} + I_{ext},$$

and the conductance-based synaptical filter

$$egin{aligned} I_{\mathit{syn}}(t) &= \sum_{u} g_{u} \mathcal{S}_{u}(t) (V_{\mathit{rev},u} - V) \ \mathcal{S}_{i,u}(t) &= rac{1}{ au_{u}} e^{-t/ au_{u}} * \sum_{n,j \in \partial^{i}} w_{j} \delta\left(t - t_{j}^{n}
ight) \end{aligned}$$

 $u = \{AMPA, NMDA, GABA_A, GABA_B\}.$

Balance condition

Under the given network conditions and default neuron parameters, we need to seek for a group of appropriate parameters (g_u) to make the network self-sustaining in a stable firing state.

we can roughly estimate a gruop of parameter by adopting the first order diffusion approximation.

$$\frac{\textit{V}_{\textit{th}} - \textit{V}_{\textit{reset}}}{\sum_{\textit{u}} \langle \textit{g}_{\textit{u}} \textit{S}_{\textit{u}} (\textit{V}_{\textit{u}} - \textit{V}) \rangle - \langle \textit{g}_{\textit{l}} (\textit{V} - \textit{V}_{\textit{l}}) \rangle} = \frac{1}{\textit{r}_{\textit{equillibrium}}},$$

and then we can dervie that

$$egin{cases} 34g_{AMPA} + 250g_{NMDA} = 1 \ 2g_{gaba_A} + 36g_{GABA_B} = 1 \end{cases}$$

For convenience, we denote the first term in the two equations as *AMPA* Contribution and *GABA_A* Contribution respectively.

Default Paramters

Symbol	Description	Value
N	Total number of neurons	2000
N_E	Total number of excitatory neurons	600
N_I	Total number of inhibitory neurons	400
K_{in}	Mean of in-degrees	400
$ au_{ref}$	Refractory period	5 <i>ms</i>
tau _u	decay time of receptors	(8, 40, 10, 50)
$V_{rev,u}$	reverse voltage	(0,0,-75,-100))

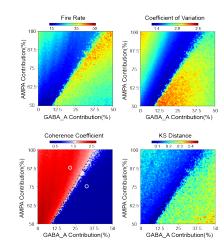
Table: Default values of model parameters used in numerical simulations



Rich dynamics in parameter submainfold

We find rich dynamics in this small balanced E-I network.

- Obvious transition in the fire rate.
- CV is larger than 1, and multi stratification.
- Coherence coefficient reflects the spike coherence.
- Critical dynamics occurs in the stratification line paramete space.



avalanches pehenomena

Mainly due to E-I delay feedback...

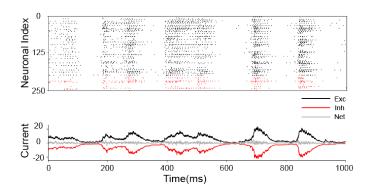


Figure: raster plot of slow dynamics



avalanches pehenomena

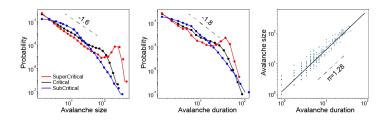


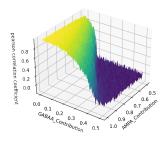
Figure: avalanches pehenomena

The slope of best fit powerlaw distribution for avalanche size is -1.6 and avalanche duration is -1.8

Criticality in large-scale network

Fit to function $s \cdot tanh(ax + by + c) + t$

- in the small blcok, slop is 1.6
- in the large-scale block, slop is 1.6



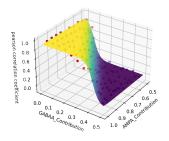


Figure: size=2k

Figure: size=100m

Criticality in large-scale network

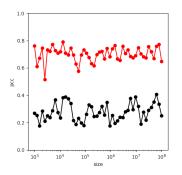


Figure: pcc with respect to different sizes

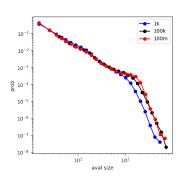


Figure: avalanches distribution



Criticality in large-scale network

Some simulation case from a fixed parameter of critical space.

 The avalanche size increases with the size of the simulation.

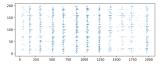


Figure: size=10k

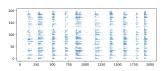
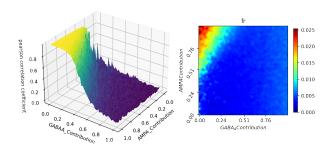


Figure: size=100m

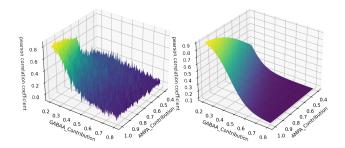
Dependence on connection density

grid search on network of 2000 neurons with 300 in-connections. x and y range is [0, 1], each 50 points.



Dependence on connection density

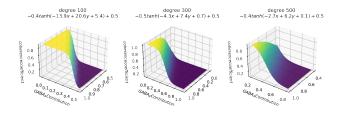
grid search on network of 2000 neurons with 500 in-connections.



As we can see, the plane is much more smooth than its in degree=100 and the one-dimensional linear submanifold disappears.

Dependence on connection density

If all case fit to function $s \cdot tanh(ax + by + c) + t$, we can find that its slope is smaller with the degree increasing.



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Summary

- The first main message of your talk in one or two lines.
- The second main message of your talk in one or two lines.
- Perhaps a third message, but not more than that.
- Outlook
 - Something you haven't solved.
 - Something else you haven't solved.

For Further Reading I



A. Author.

Handbook of Everything.

Some Press, 1990.



S. Someone.

On this and that.

Journal of This and That, 2(1):50–100, 2000.