

Topology, dynamics and coding in neuronal networks

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Outline

- 1 Introduction
- 2 Random topologies
- 3 Effects of topology in dynamics
 - Broader degree distributions
 - Randomizing the weights
 - Local connectivity
 - Networks with hubs
- 4 Implications for neural coding
 - Stimulus induced variability reduction
 - Topology in feature space

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2 Random topologies

3 Effects of topology in dynamics

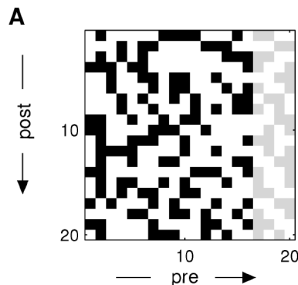
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What is topology?

- Graph: nodes and vertices, directed and undirected.
- Adjacency matrix.
- In-degree and out-degree distributions.



Kriener et al. 2008

What do we care about?

- Distribution of mean output firing rates.
- Cellular variability:
 - ▶ within a trial: CV of the ISI distribution.
 - ▶ between trials: Fano factor of the spike count distribution
- Co-variability between activity of different cells:
 - ▶ Fano factor of the PSTH.
 - ▶ Distribution of pairwise correlations coefficients.
 - ▶ Higher order correlations.
 - ▶ Power spectrum.

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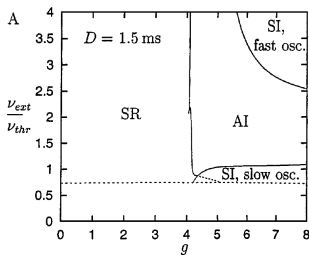
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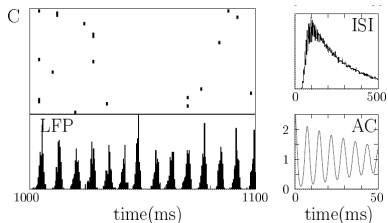
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Is random good enough?

- Sparse randomly connected balanced EI networks can produce activity:
 - ▶ Skewed and long-tailed firing rate distributions.
 - ▶ High single cell variability (irregularity).
 - ▶ Very low co-variability between pairs of cells (asynchrony).
- While also offering a variety of other dynamical states.
- Fast and slow oscillations where individual neurons fire irregularly skipping cycles are also possible.
- Mean-field techniques can be used.



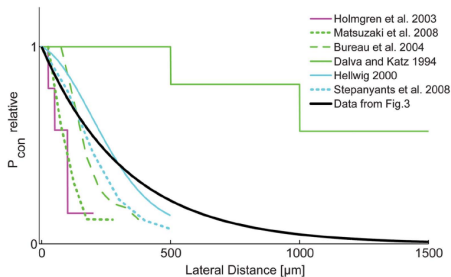
Brunel 2000



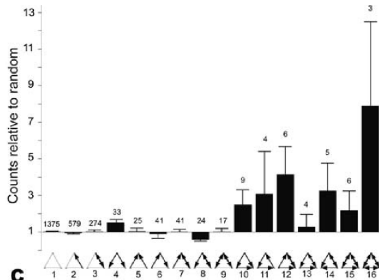
Brunel and Hakim 1999

Random is not enough

- Random connectivities can't explain different phenomena that are observed in the brain.
- Recent experimental studies have shown that the wiring in the brain is far from random.



Boucsein et al. 2011



Song et al. 2005

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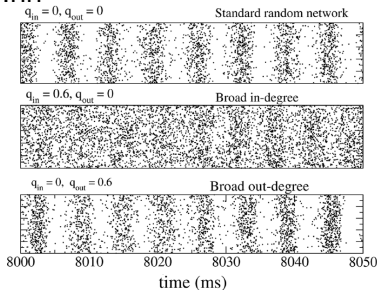
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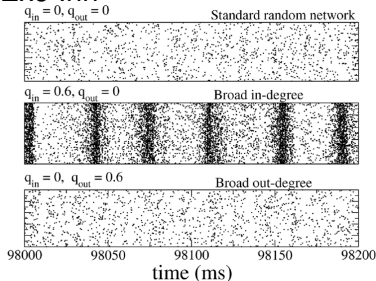
The effect of degree distributions

- Increasing the width of the in-degree distribution affects the global state of the network by driving transitions between asynchronous behavior and oscillations.

Inh



Exc-Inh



Roxin 2011

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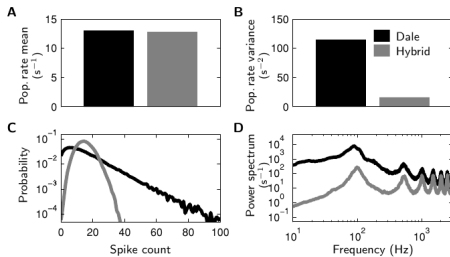
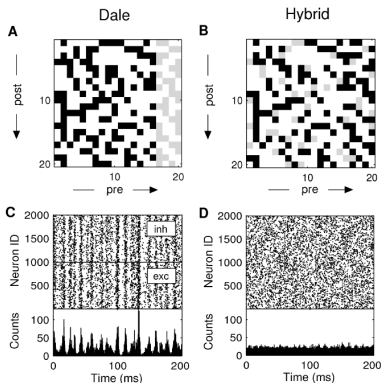
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Hybrid topologies

- Mean inputs are identical in hybrid and Dale spiking networks.
- Correlations vanishes when randomizing weights even for densely connected networks.



Kriener et al. 2008

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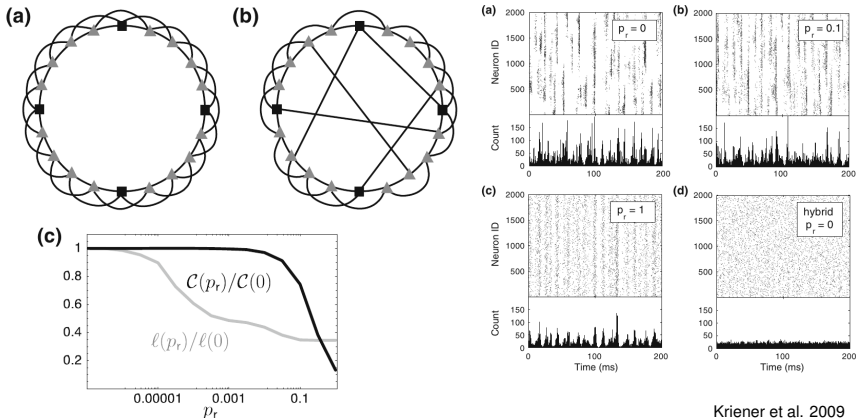
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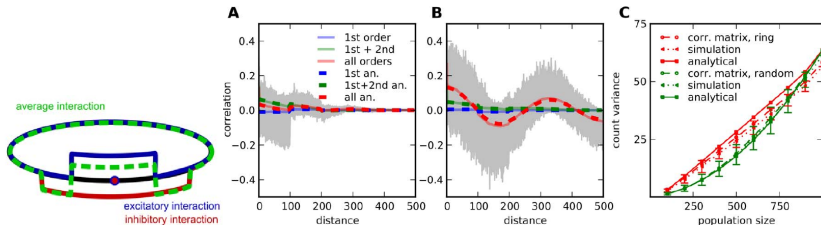
Distance dependent connectivity

- The Fano Factor of the PSTH is higher in networks with higher clustering coefficient but only locally.
- Dale beats topology!



Distance dependent connectivity

- In networks with local connectivity, the distribution of correlation coefficients (cc) is broader, while average correlations remain similar compared to the random case.
- In EI networks broader inhibitory connectivity broadens the cc distribution.
- Locally fluctuations increase faster in the ring network than in the random case.



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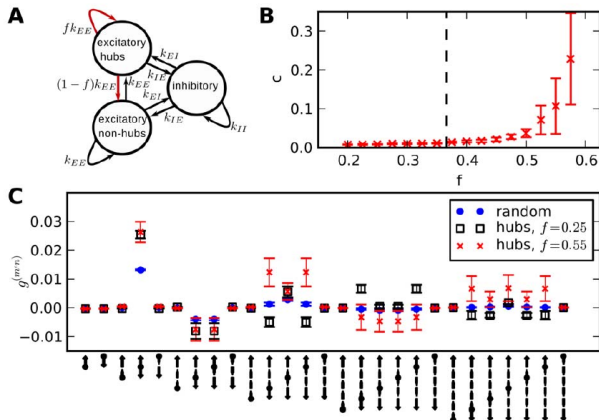
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Networks with hubs

- Making the out-degree distribution broader and favoring connections between neurons within similar out-degree increases average correlations.



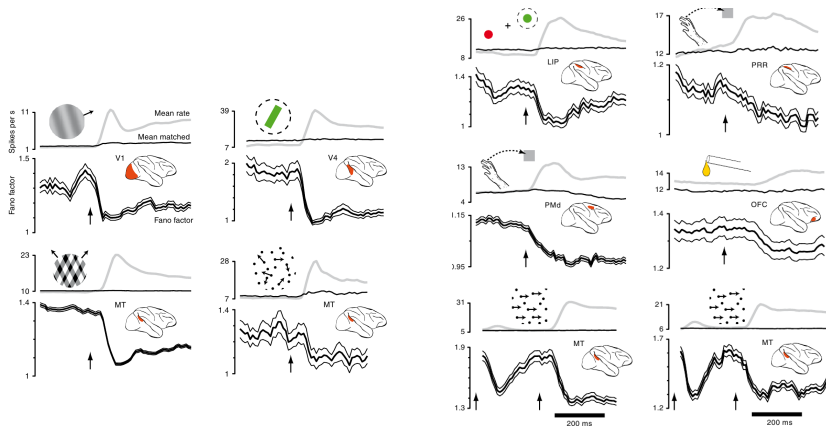
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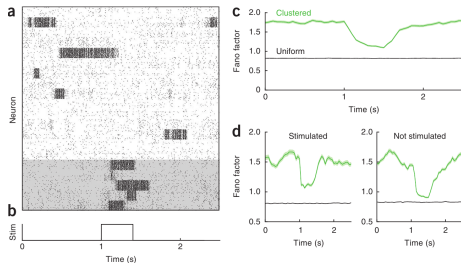
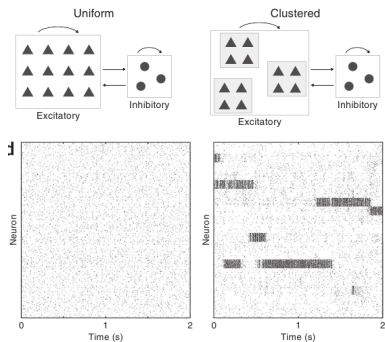
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Stimulus induced variability reduction



Churchland et al. 2010

Stimulus induced variability reduction



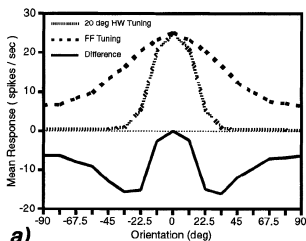
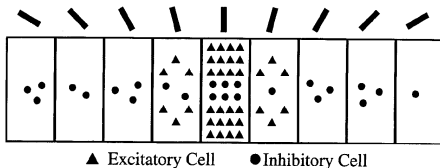
Litwin-Kumar et al. 2012

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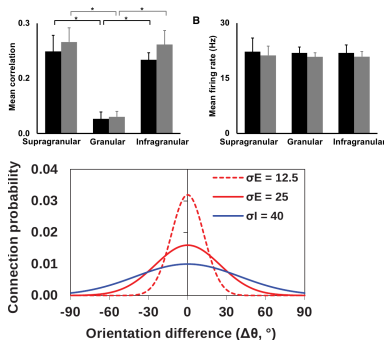
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Topology in *feature space*

- Useful to get sharpening of tuning curves.
- Also to get contrast invariance.
- To reduce noise correlations?



Somers et al. 1995



Hansen et al. 2012

Summary

- Random connectivity is a good model but it is not enough.
- Degree distributions have a strong impact in the activity of the network.
- The presence of hub neurons can increase the amount of correlated activity in the network.
- Local connectivity can explain the reduction of inter-trial variability, as well as sharpening of tuning curves and contrast invariance.