

Unveiling Criticality in Noisy Nonequilibrium Systems

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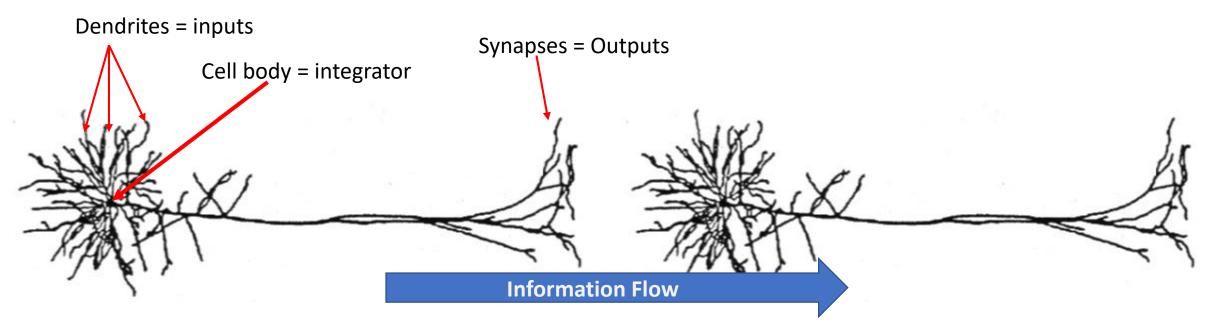
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Outline

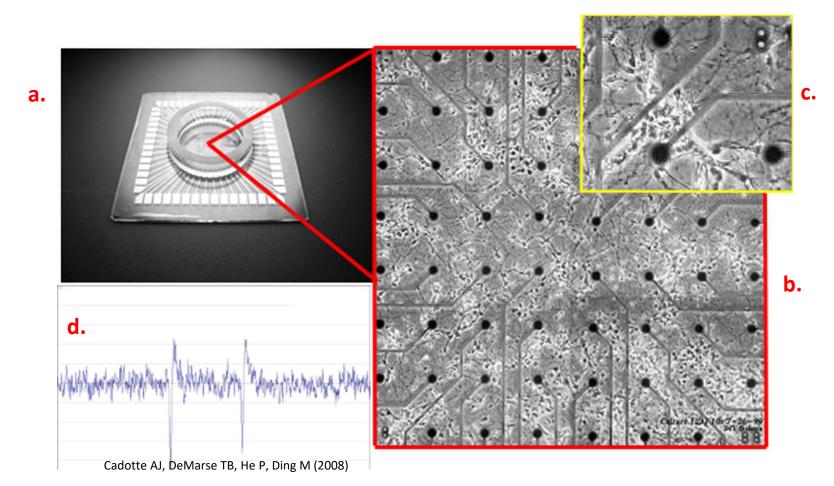
- Neural avalanches as a critical branching process
- Branching process with noise
- Measures of criticality

What are neurons?

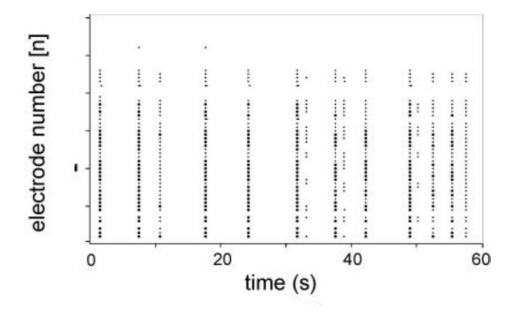
- Are directed computational elements
- All or nothing electrical response
- Sum inputs from multiple parents



• We can record and observe neurons in bulk using electrode arrays

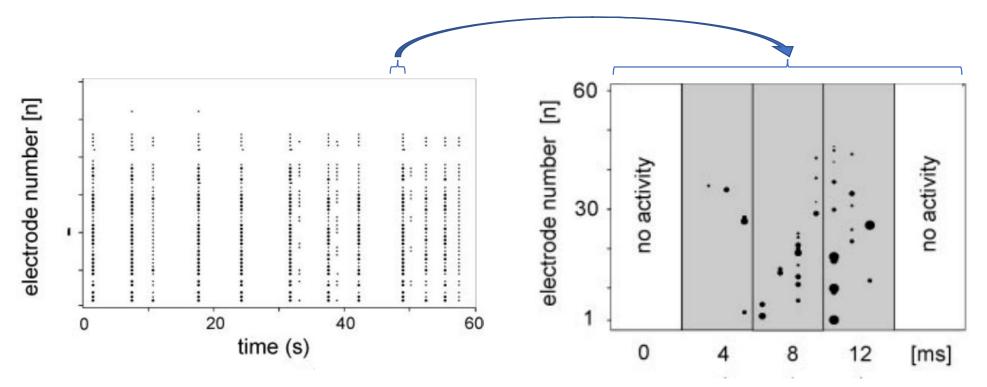


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- You will see periods of activity that we can call "avalanches."



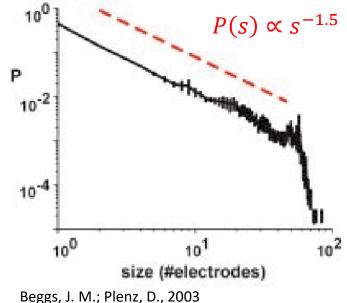
Beggs, J. M.; Plenz, D., 2003

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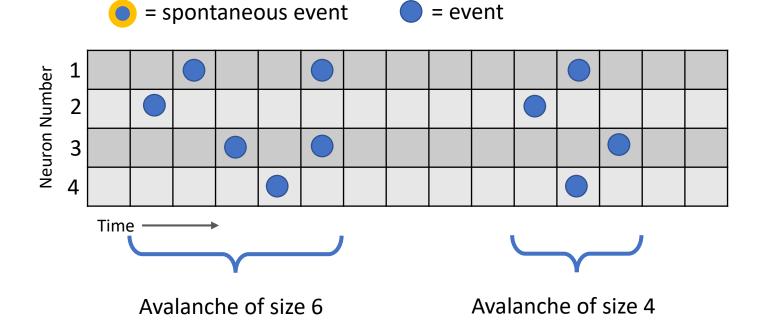


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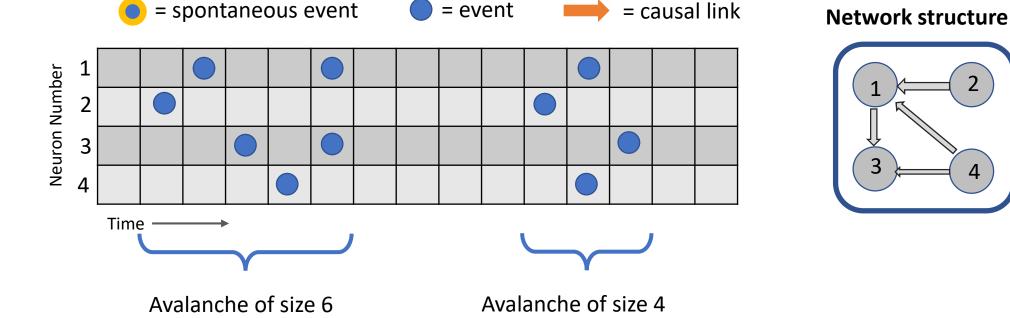
- We can record and observe neurons in bulk using electrode arrays
- You will see periods of activity that we can call "avalanches."
- The neurons per avalanche is power-law distributed $P(s) \propto s^{-1.5}$
- Modelled as a branching process



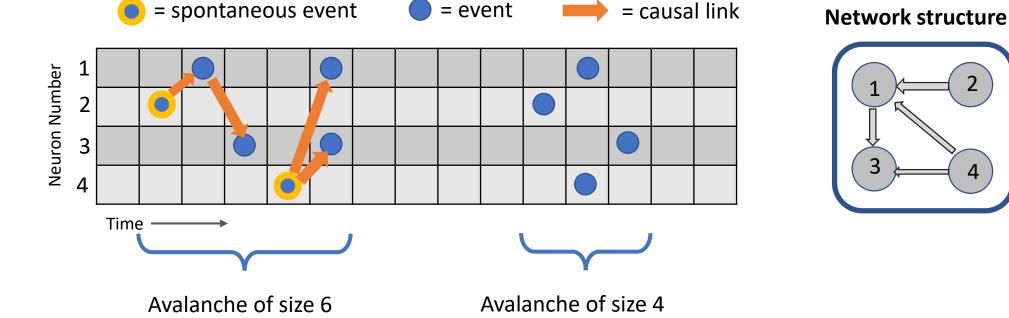
- Neurons are noisy
 - 'Minis', unobserved couplings
- Avalanches should be causal



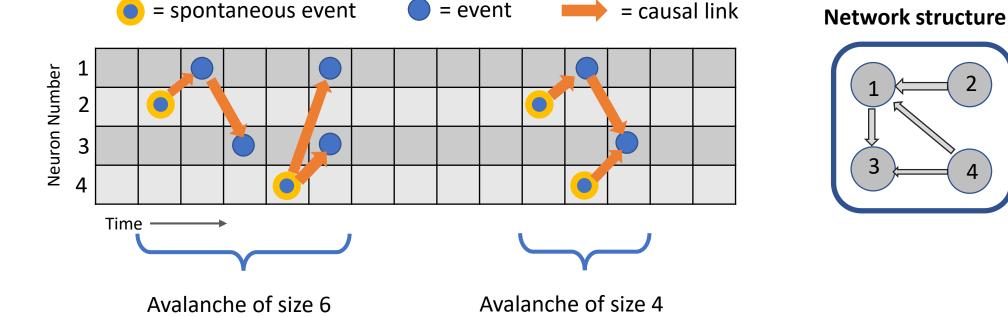
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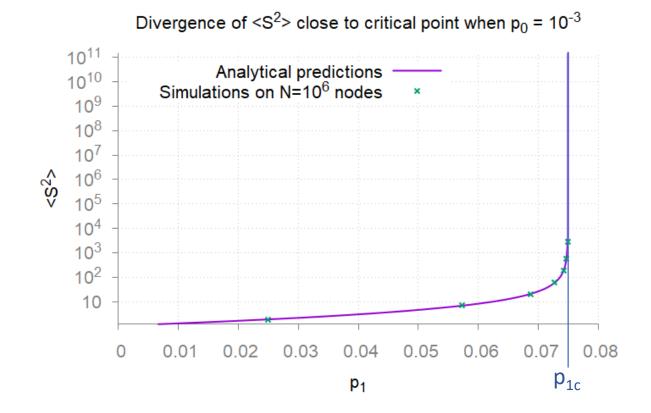
Branching process with noise

• Each time step each node fires with probability:

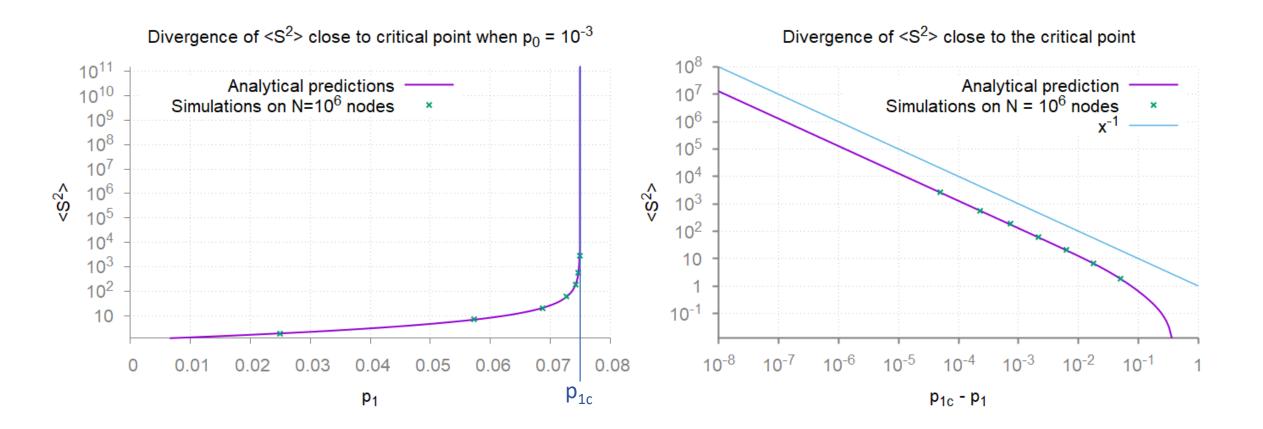
$$P(m \text{ parents}) = 1 - (1 - p_0)(1 - p_1)^m$$

- p_0 is the noise parameter
- p_1 is the branching parameter
- Can be studied analytically in the limit of an infinite random graph, or numerically in simulations on a finite random graph

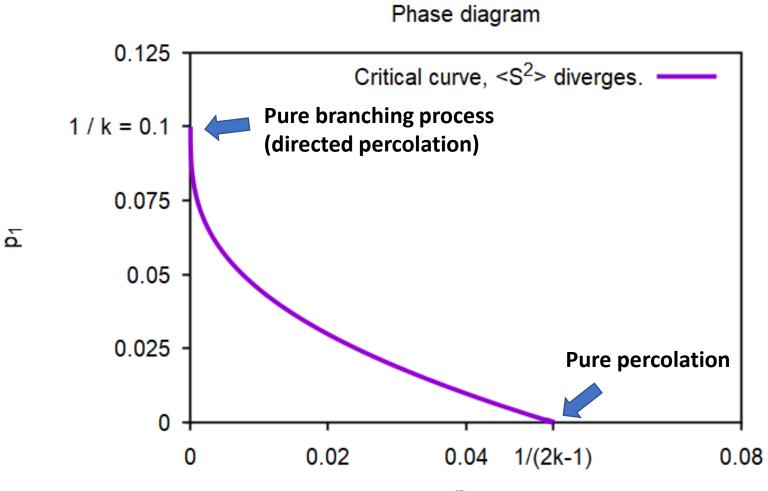
Divergence of causal web size $\langle S^2 \rangle$



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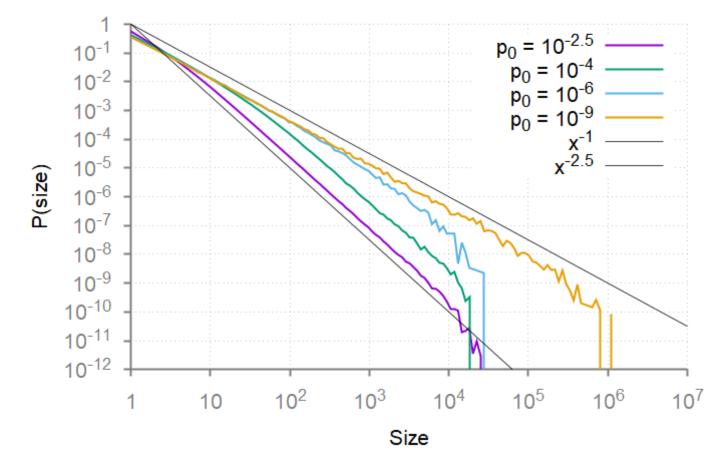
Phase diagram



 \mathbf{p}_0

Causal web distributions

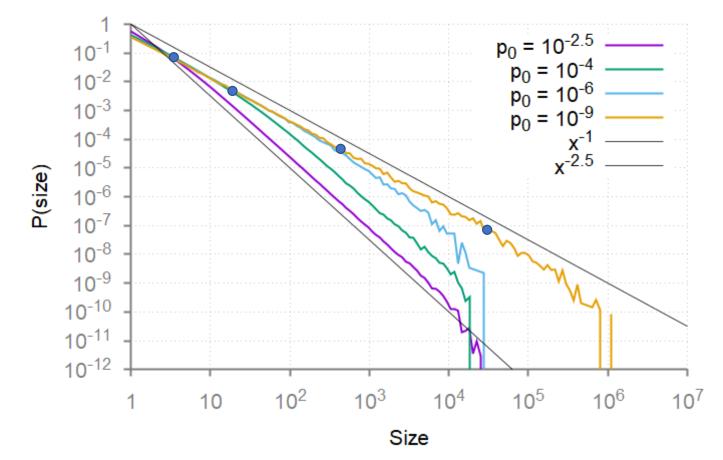
Critical causal webs with varying noise on N=10⁶ nodes



Causal web distributions on random graph, $N=10^6$ and (in/out)degree = 10

Causal web distributions

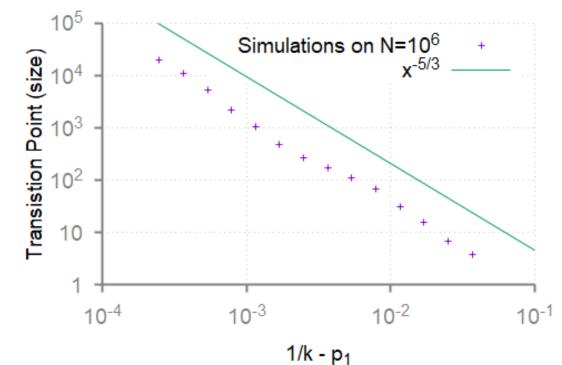
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Causal web distributions

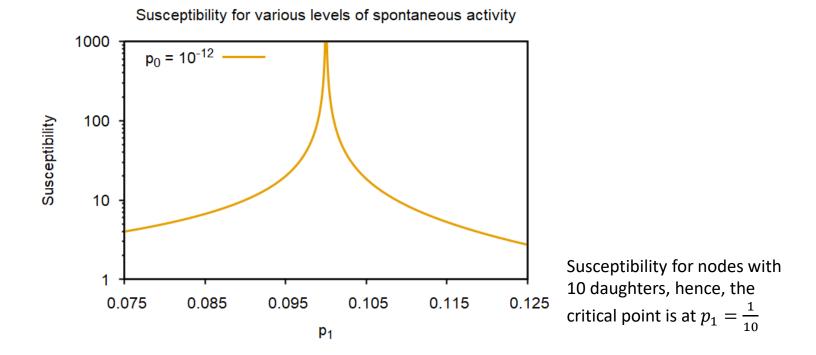
• The transition point from a pure branching process to percolation diverges as a power law



Divergence of the transistion point

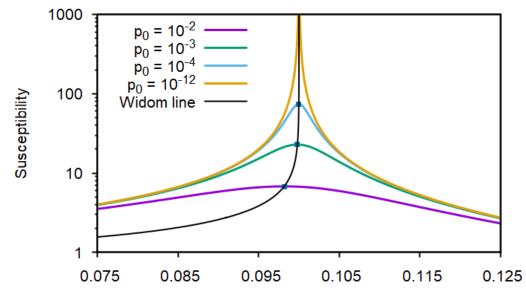
Susceptibility

- The active fraction Φ is the proportion of cells firing per unit time
- In the zero noise limit ($p_0 \to 0$), the susceptibility $\frac{\partial \Phi}{\partial p_0}$ diverges at the critical point



Susceptibility

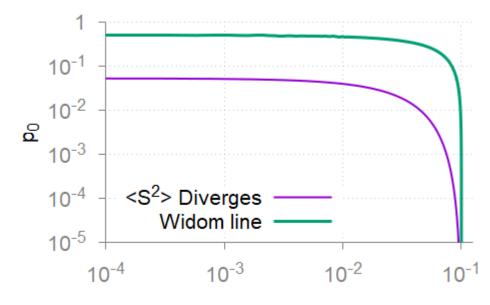
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Susceptibility for various levels of spontaneous activity

Susceptibility

- The active fraction Φ is the proportion of cells firing per unit time
- In the zero noise limit $(p_0 \rightarrow 0)$, the susceptibility $\frac{\partial \Phi}{\partial p_0}$ diverges at the critical point, but with noise the susceptibility exhibits a finite maximum Phase curve

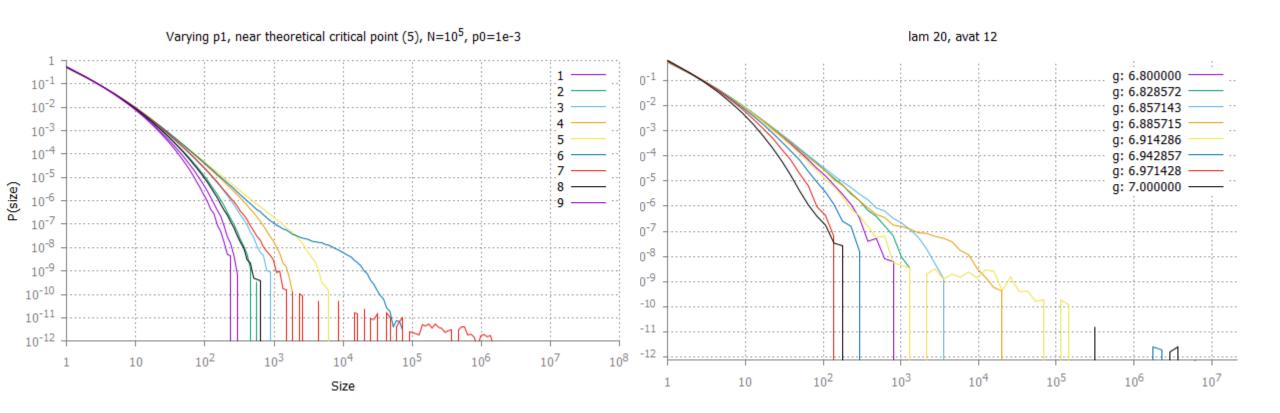


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Conclusions

- Susceptibility of the active fraction is a limited indicator of criticality
- Studying causal activity allows for independent activity to be resolved
- The branching process with noise exhibits a critical line, that varies between the branching process and percolation
- Qualitative agreement with biological neuron model

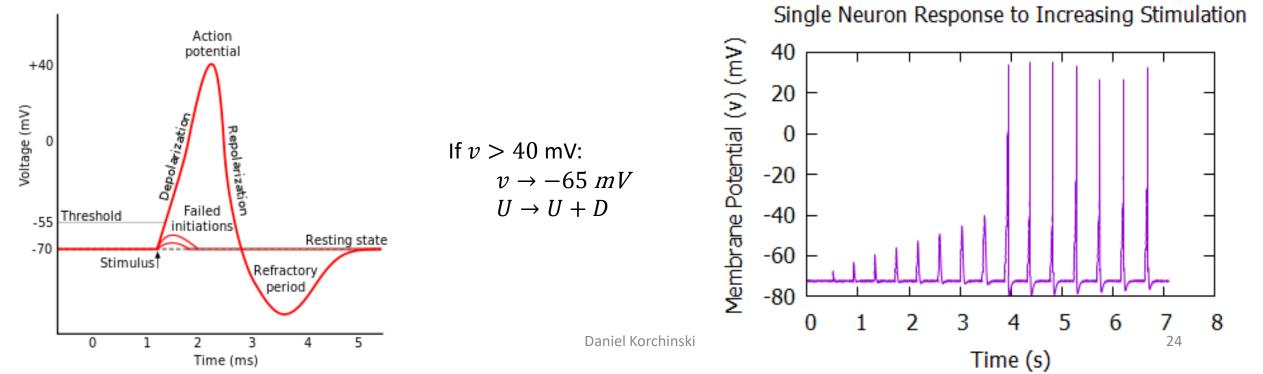
Appendix: Comparison to 'biological' neurons



Methods: Dynamical Model

• Izhikevich Model (2007)

$$\frac{dv}{dt} = k(v - v_r)(v - v_t) - U + \eta + i_c$$
$$a\frac{dU}{dt} = b(v - v_r) - U$$



С

Dynamical Model : Parameter Constraints

- Many parameters, but most have physical interpretation
- <u>http://neuroelectro.org/</u> : Pyramidal Neurons

