## Stochastic facilitation in the brain?

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# Unsolved Problems of Neural Noise

- To what extent may the oscillatory behavior of individual neurons, or linked populations of neurons, best be characterized as noise-driven quasi-cycles or as noisy limit cycles (or both)?
- What role(s) does neural noise play in the synchronization of, and information transmission between, neural populations located far from one another in the brain?
- To what extent do noise-driven quasi-patterns arise in the brain and affect its overall functioning?

## Neural oscillations

Measure electrical activity of brain in any of several ways to get fluctuating brioadband signal with roughly 1/f poser spectrum with bumps at some particular frequencies



Greenwood, McDonnell, Ward, 2015, Neural Computation

## Noisy limit cycles or quasi-cycles?

- A common model of interaction of Excitatory and Inhibitory neurons in cortex of brain
- $S_{EE}$ ,  $S_{II}$ ,  $S_{EI}$ ,  $S_{IE}$ : Synaptic efficacies between neurons





# Noisy limit cycles or quasicycles?

A general rate/local field potential (LFP) model of E-I neural 0 interaction (orig. Wilson & Cowan, Biophysical Journal, 1972):  $\tau_E dV_E(t) = \left[ -V_E(t) + g \left[ a_E (S_{EE} V_E(t) - S_{EI} V_I(t) - \theta_E + P_E(t)) \right] \right] dt + \sigma_E dW_E(t)$  $\tau_I dV_I(t) = \left[ -V_I(t) + g \left[ a_I (S_{II} V_I(t) - S_{IE} V_E(t) - \theta_I) \right] \right] dt + \sigma_I dW_I(t)$  $V_F$ : firing rate/LFP of excitatory neurons;  $P_F(t)$ : input current  $V_{I}$ : firing rate/LFP of inhibitory neurons;  $\tau_{F}$ ,  $\tau_{I}$ : time constants  $S_{EE}$ ,  $S_{II}$ ,  $S_{EI}$ ,  $S_{IE}$ : synaptic efficacies between neurons g: threshold function, usually  $g(x) = 1/(1+e^x)$ , i.e. logistic  $W_F(t)$ ,  $W_I(t)$ : standard Brownian motions (NOISE!)  $\Theta_{\rm F}, \Theta_{\rm I}, \sigma_{\rm F}, \sigma_{\rm I}, a_{\rm F}, a_{\rm I}$ : constants

 $\tau_{E} dV_{E}(t) = \left[-V_{E}(t) + g\left[a_{E}(S_{EE}V_{E}(t) - S_{EI}V_{I}(t) - \theta_{E} + P_{E}(t))\right]\right]dt + \sigma_{E}dW_{E}(t)$  $\tau_{I}dV_{I}(t) = \left[-V_{I}(t) + g\left[a_{I}(S_{II}V_{I}(t) - S_{IE}V_{E}(t) - \theta_{I})\right]\right]dt + \sigma_{I}dW_{I}(t)$ 

•  $g(x) = 1/(1+e^{-x})$ 

SIF

- $\sigma_E = \sigma_I = 0$ : deterministic limit cycles around unstable fixed point
- $\sigma_E = \sigma_I > 0: \text{ noisy limit}$ cycles



Wallace et al., 2011

## Quasi-cycles

 $\tau_E dV_E(t) = \left[ -V_E(t) + g \left[ a_E (S_{EE} V_E(t) - S_{EI} V_I(t) - \theta_E + P_E(t)) \right] \right] dt + \sigma_E dW_E(t)$  $\tau_I dV_I(t) = \left[ -V_I(t) + g \left[ a_I (S_{II} V_I(t) - S_{IE} V_E(t) - \theta_I) \right] \right] dt + \sigma_I dW_I(t)$ 

•  $g(x) = x; a_E = a_I = 1;$ 

SIF

 $S_{FI}$ 

- $\mathbf{O} \quad P_E(t) = 0; \ \boldsymbol{\Theta}_E = \boldsymbol{\Theta}_I = 0$
- $\sigma_E = \sigma_I = 0$ : damps to stable fixed point
- $\sigma_E$ ,  $\sigma_I$  > 0: quasi-cycles i.e. noisedriven oscillations at frequency determined by  $S_{EE}$ ,  $S_{II}$ ,  $S_{EI}$ ,  $S_{IE}$  and  $\tau_E$ ,  $\tau_I$
- Inputs to neurons are typically Poissonlike, i.e. noise-driven



Wallace et al., 2011

# Noisy limit cycles or quasi-cycles?

- Both models can generate narrow band oscillations
- Both models generate bursting, but with different mechanisms
- Both models generate a physiological range of local field potential oscillation frequencies
- We need new methods to distinguish the mechanism in data from brain recordings



Greenwood, McDonnell, Ward, 2015, Neural Computation

### Noisy synchronization?

- Noise can enhance synchronization between oscillators
- Both noisy limit cycles and quasi-cycles display synchronization between distant populations
- Synchronization is supposed to benefit neural information transfer
- Does noise benefit information transfer via synchronization?



#### Greenwood, McDonnell, Ward, 2015, unpublished



Daffertshoffer & vanWijk, 2011, Frontiers

### Noisy multiplexing?

- Population neural codes are noisy – sparse neural firing
- Noisy oscillations can be modulated by a signal and decoded at another area to recover the signal
- Does the noise play a computational role? Via quasi-cycles?



Akam & Kullman, 2014, Nat Rev Neurosci

## Spatial quasi-patterns?

- Noise-driven systems can exhibit both quasi-cycles and quasi-patterns
- Quasi-patterns are analogous to quasicycles but are spatial and characterized by a wave number instead of a frequency
- Demonstrated in predator-prey and epidemiological models.
- Quasi-patterns of quasi-cycles in brains? Do they affect information processing and information transmission?



Butler & Goldenfeld, 2011, PRE



McKane et al, 2014, Bull Math Biol

### Cortical waves

- Binocular-rivalryinduced cortical traveling waves: waves of dominance move across the visual cortex as perception of dominant stimulus changes
- EEG standing alpha (8-12 Hz) waves: areas with similar colour are roughly in phase; dark and light are 180 deg out of phase.



Wilson, Blake, Lee, Nature, 2001



Nunez & Srinivassan, Brain Research, 2014

### Mexican Hat coupling on a 100 x 100 lattice of stochastic E-I processes

- Quasi-cycles at ~70 Hz at each lattice location
- Local Mexican hat coupling only; run for 10,000 iterations
- Quasi-Turing patterns at ≈ 5-6 cycles in amplitudes (100/19=5.26)
- Little mathematical understanding







#### Greenwood & Ward, 2015, unpublished

### Summary

• Noisy limit cycles or quasi-cycles?

- Noise-driven synchronization and information transfer?
- Spatial quasi-patterns mixed with noisy or noise-driven oscillations?