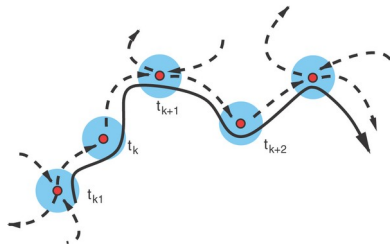
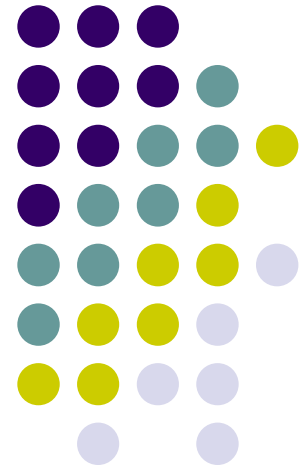


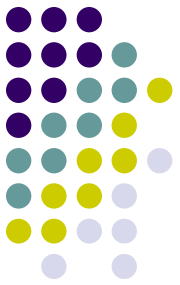
Winnerless Competition in Neuroscience



M. Ivanchenko, UoL
M. Rabinovich, UCSD



The Team



NONLINEAR DYNAMICS

Mikhail Rabinovich

Valentin Afraimovich



OLFACTORY SYSTEM

Ramon Huerta

Alex Volkovskii

Valentin Zhigulin

Henry Abarbanel

Gilles Laurent

SONGBIRD

Ramon Huerta

Marta Garcia-Sanches



CLIONE

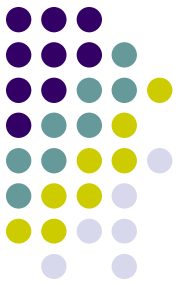
Rafael Levi

Pablo Varona

Yuri Arshavsky

Allen Selverston





Reading List

- **Vision:**

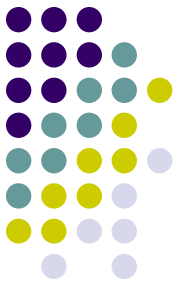
Rabinovich MI, Huerta R, Laurent G. Transient Dynamics for Neural Processing, Science 321: 48, 2008.

- **Maths:**

Afraimovich VS, Rabinovich MI, Varona P. Heteroclinic Contours in Neural Ensembles and the Winnerless Competition Principle. International Journal of Bifurcation and Chaos, Vol. 14(4): 1195-1208, 2004.

- **Neuroscience:**

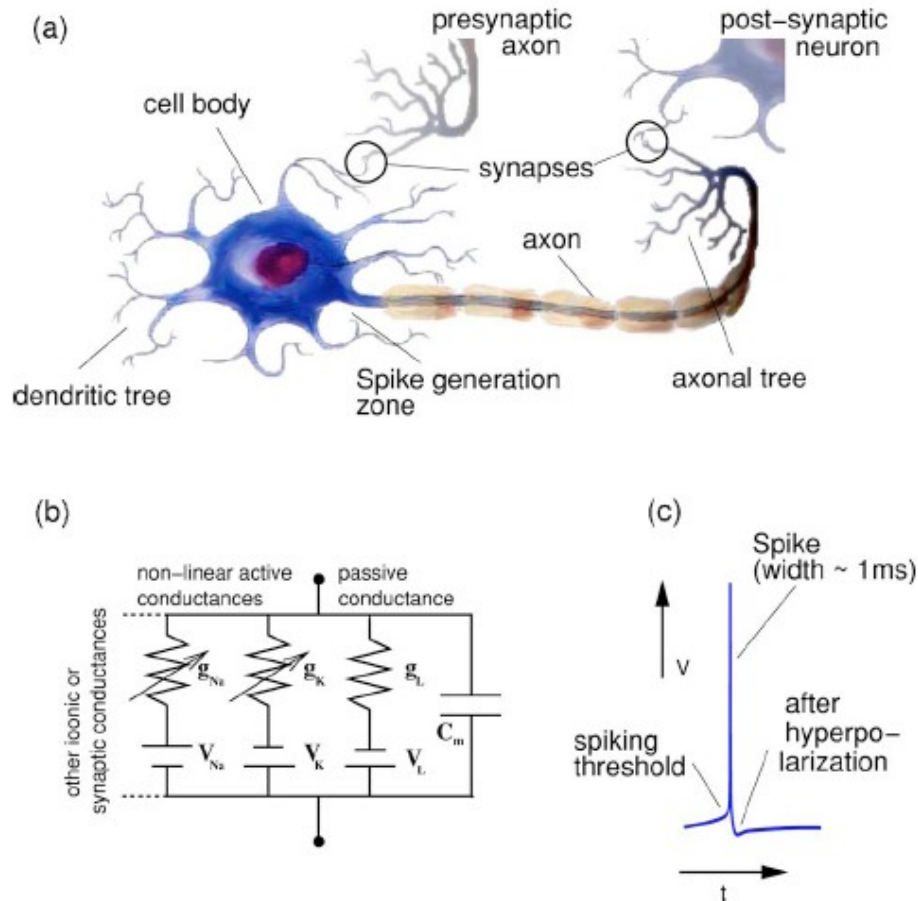
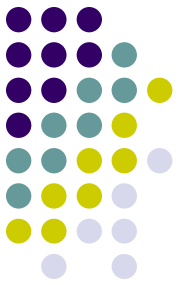
Rabinovich MI, Varona P, Selverston AI, Abarbanel HDI. Dynamical Principles in Neuroscience. Reviews of Modern Physics 78(4): 1213, 2006.



The Plan

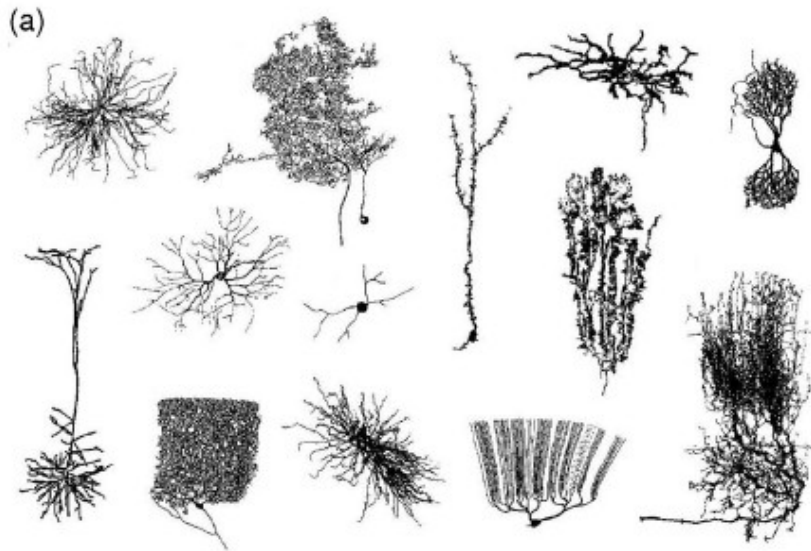
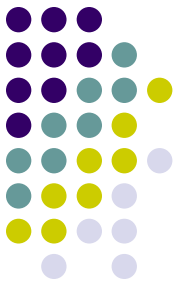
1. Transient dynamics and winnerless competition in neuroscience
2. Heteroclinic sequences in asymmetric networks
3. Beyond the rate model

Single neurons: modelling

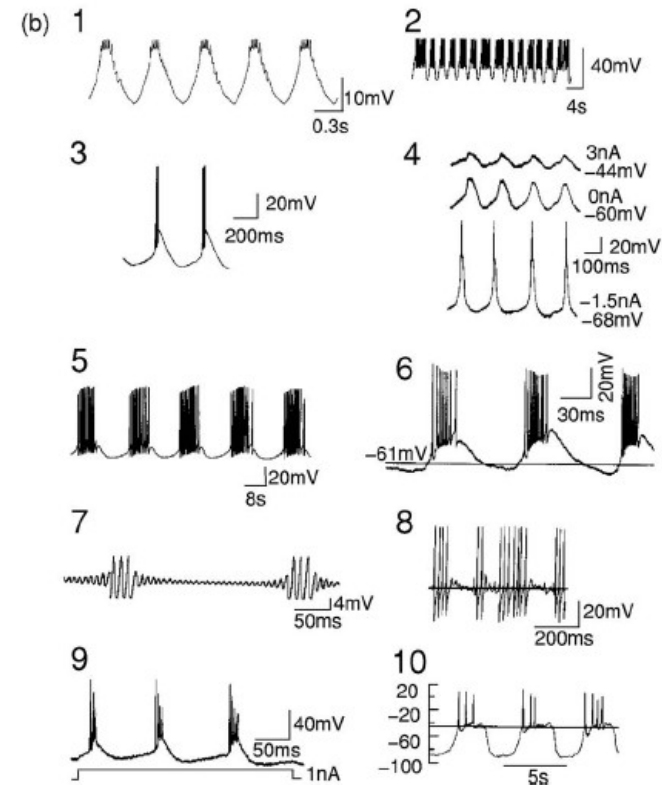


Rabinovich, Varona, Selverston, Abarbanel,
Reviews of Modern Physics **78**, 1213, 2006

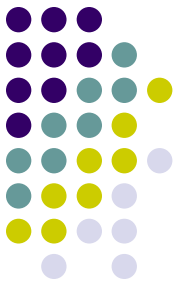
Single neurons: diversity



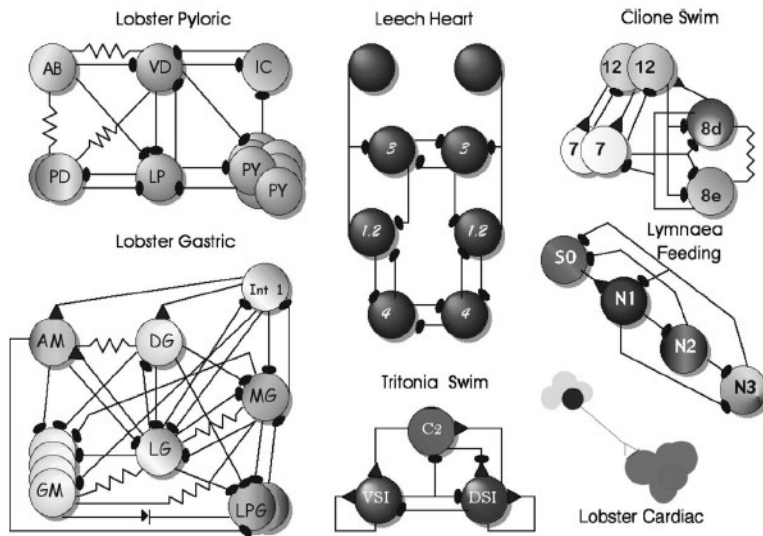
Rabinovich, Varona, Selverston, Abarbanel,
Reviews of Modern Physics **78**, 1213, 2006



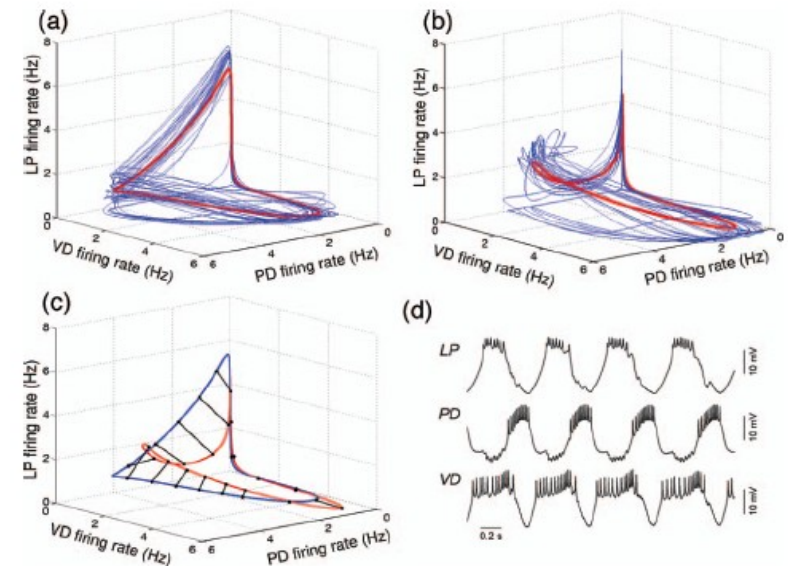
Small scale: mini-circuits



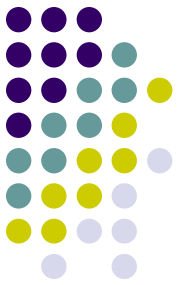
Invertebrate CPGs



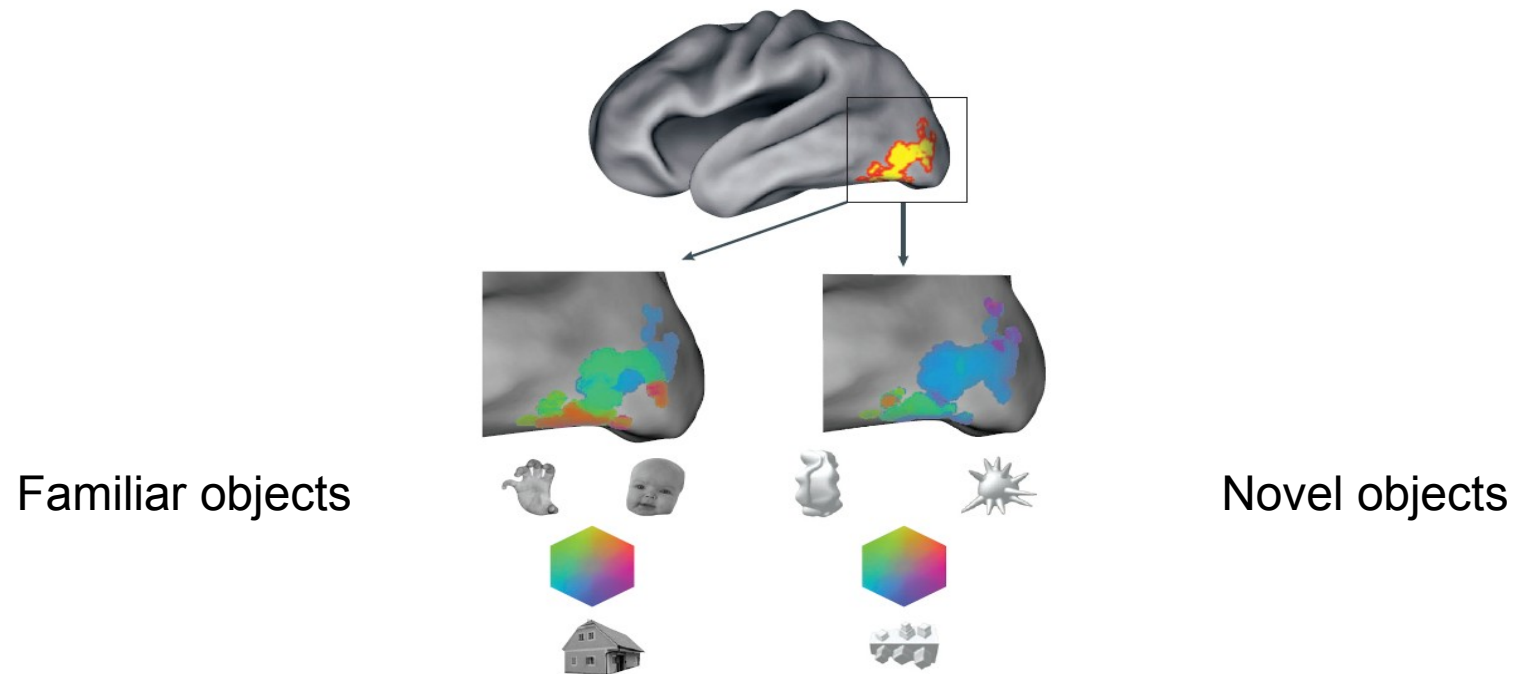
Rev. Mod. Phys., Vol. 78, No. 4, October–December 2006



Large scale: the brain

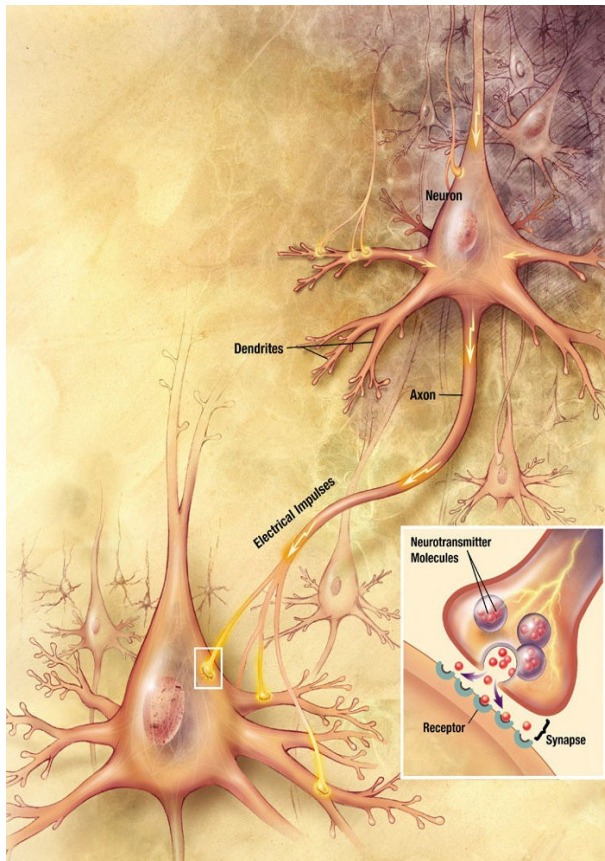
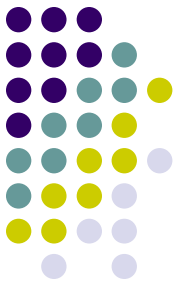


fMRI: activation patterns

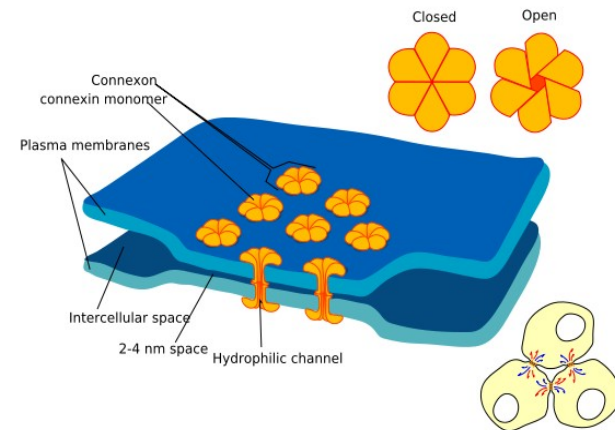


De Beeck, Haushofer, Kanwisher,
Nature Reviews Neuroscience, **9**, 123, 2008

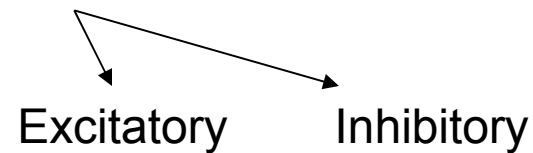
Synapses



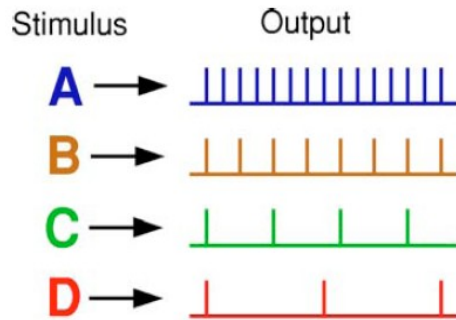
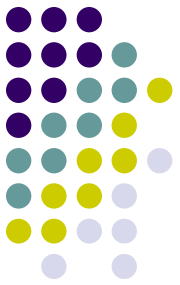
Electrical



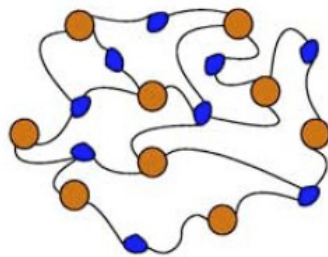
Chemical



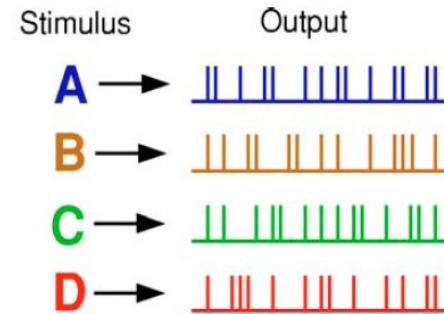
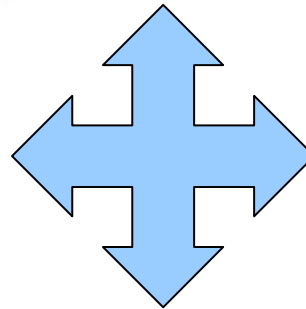
Coding



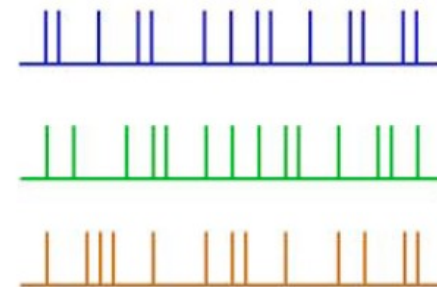
Rate coding



Spatial

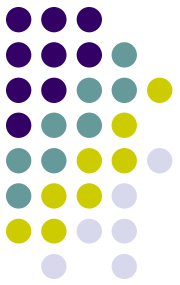


Timing coding



Spatio-temporal

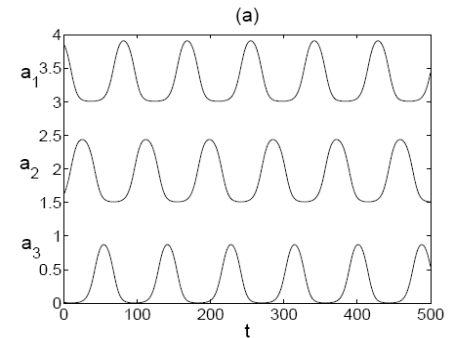
Models: neurons



Rate models

$$\dot{a}_i(t) = F_i(a_i(t)) [G_i(a_i(t)) - \sum_j \rho_{ij} Q_j(a_j(t))]$$

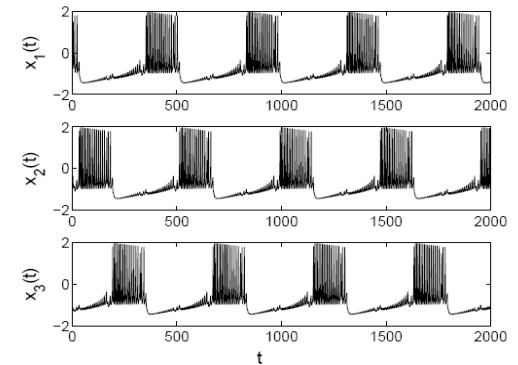
$a_i(t) > 0$ is the spiking rate of the i th neuron or cluster; ρ_{ij} is the connection matrix; and F, G, Q are polynomial functions.



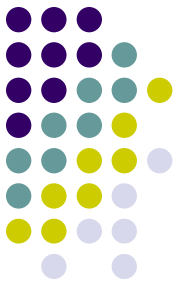
Hodgkin-Huxley

$$\begin{aligned} C\dot{v}(t) &= g_L[v_L - v(t)] \\ &\quad + g_{Na}m(t)^3h(t)[v_{Na} - v(t)] \\ &\quad + g_{K}n(t)^4[v_K - v(t)] + I, \\ \dot{m}(t) &= \frac{m_\infty(v(t)) - m(t)}{\tau_m(v(t))} \\ \dot{h}(t) &= \frac{h_\infty(v(t)) - h(t)}{\tau_h(v(t))} \\ \dot{n}(t) &= \frac{n_\infty(v(t)) - n(t)}{\tau_n(v(t))} \end{aligned}$$

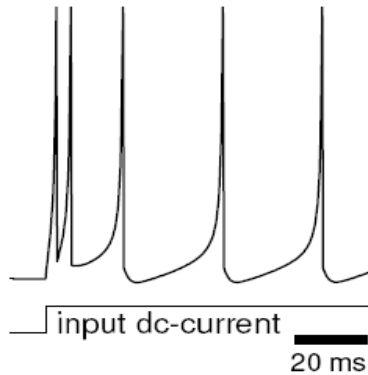
$v(t)$ is the membrane potential, $m(t)$, and $h(t)$, and $n(t)$ represent empirical variables describing the activation and inactivation of the ionic conductances; I is an external current. The steady-state values of the conductance variables $m_\infty, h_\infty, n_\infty$ have a nonlinear voltage dependence, typically through sigmoidal or exponential functions.



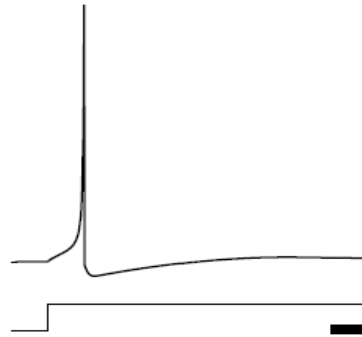
Types



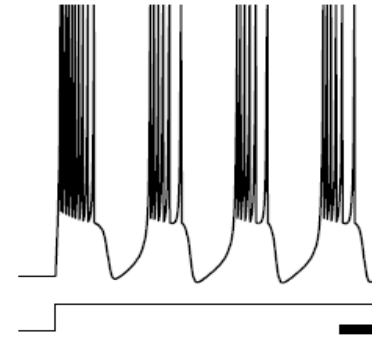
(A) tonic spiking



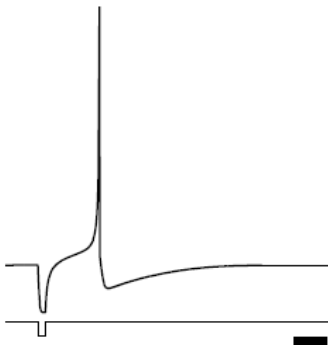
(B) phasic spiking



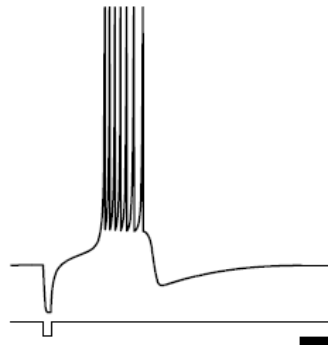
(C) tonic bursting



(M) rebound spike

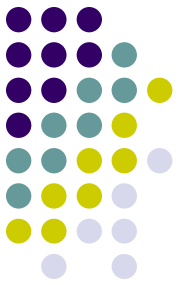


(N) rebound burst



Izhikevich, IEEE Trans. Neural Networks, 2004

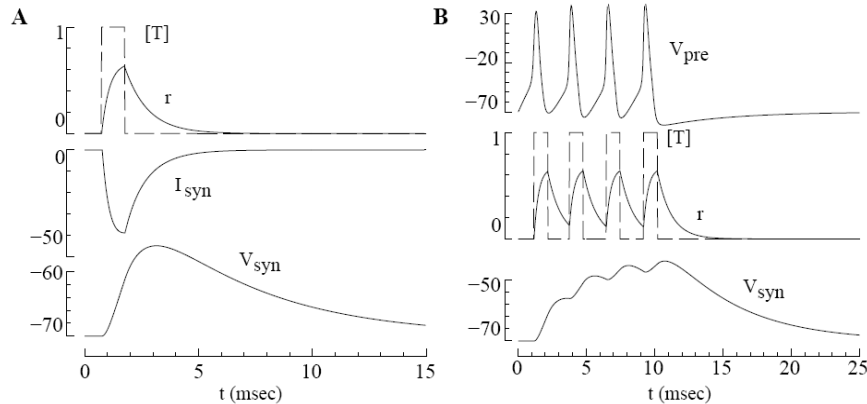
Models: synapses



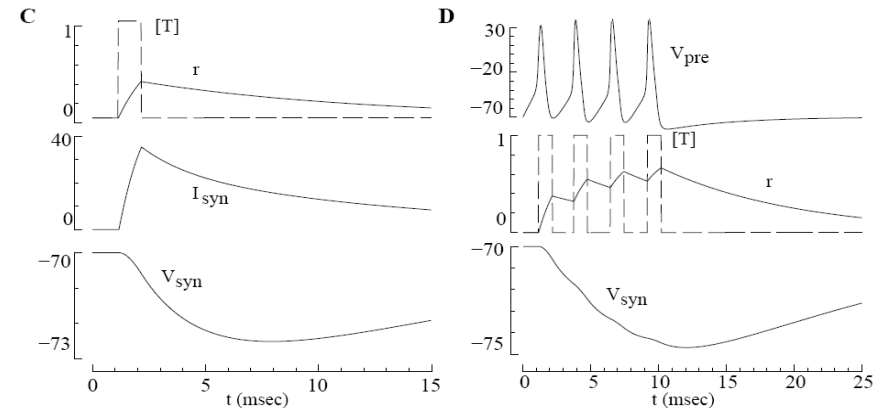
Electrical: diffusive $I_{syn}(t) = g_{syn}(V_{post} - V_{pre})$

Chemical: integrators $I_{syn}(t) = g_{syn}r(t)(V_{post} - E_{syn}); \quad dr/dt = a[T](1-r) - br$

Excitatory

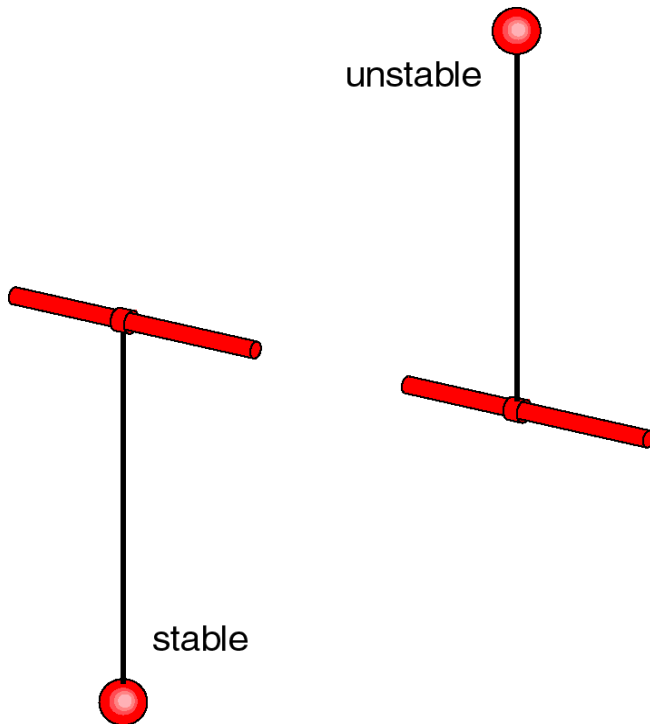
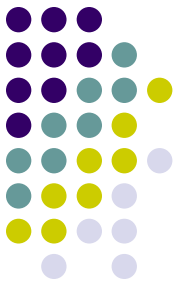


Inhibitory

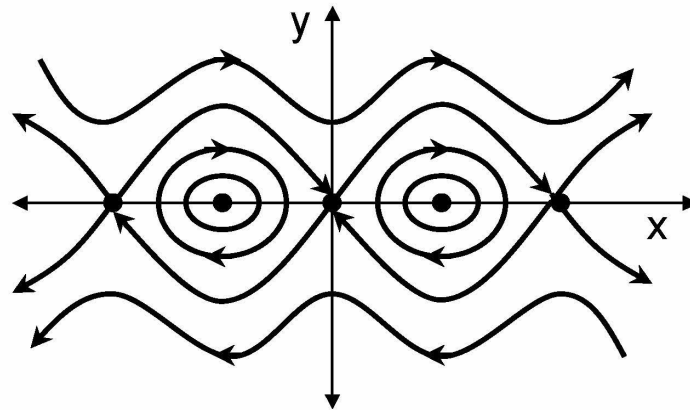


Destexhe, Mainen, Sejnowsky, Neural Comp., 1994

Dynamical systems

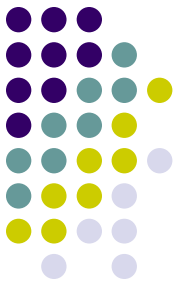


$$x'' + \sin(x) = 0 \quad \begin{aligned} x' &= y \\ y' &= -\sin(x). \end{aligned}$$

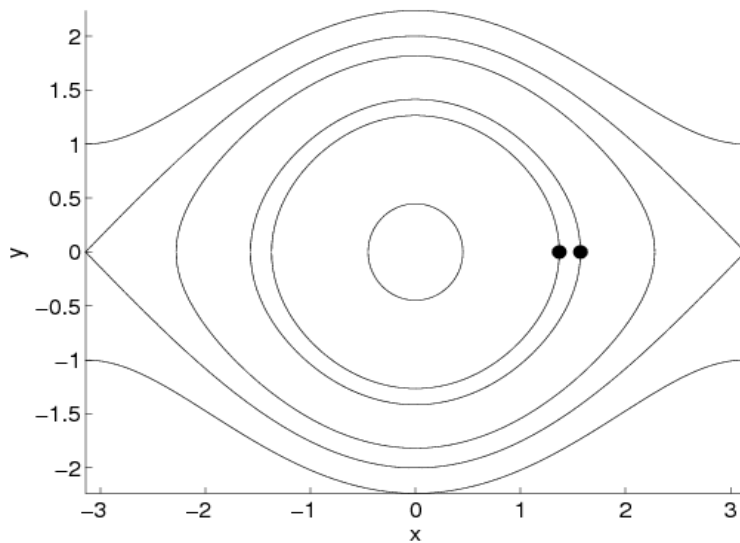


1. Equilibria: centre (stable), saddle (unstable)
2. Periodic orbits
3. Heteroclinic/Homoclinic trajectories

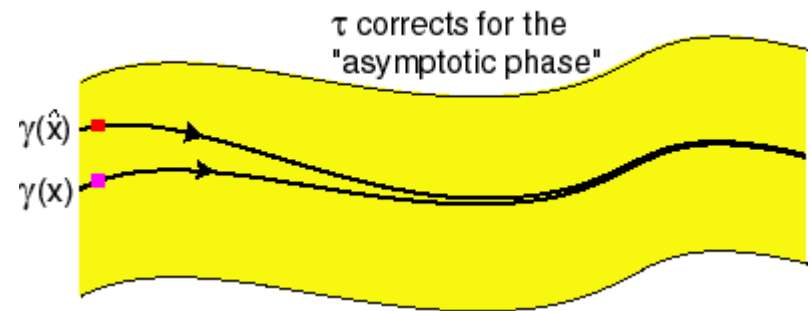
Stability



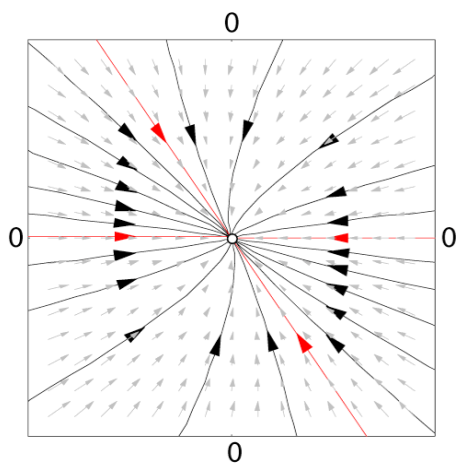
Lyapunov



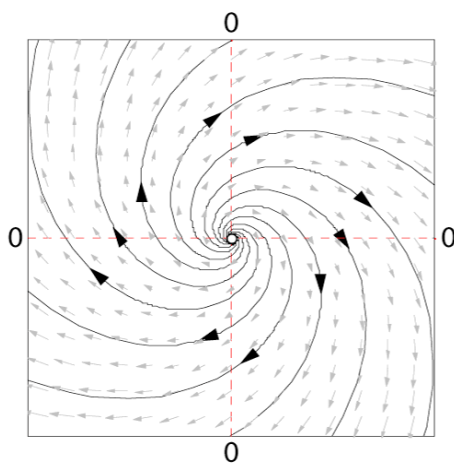
Stability



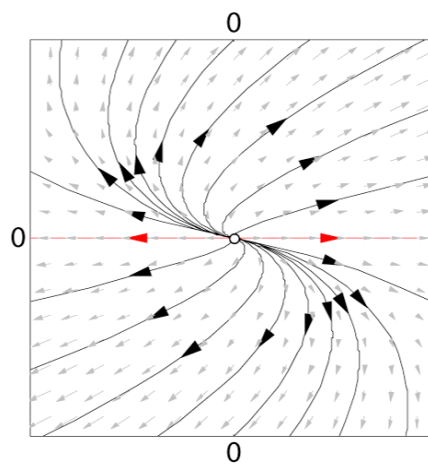
Equilibria in 2D



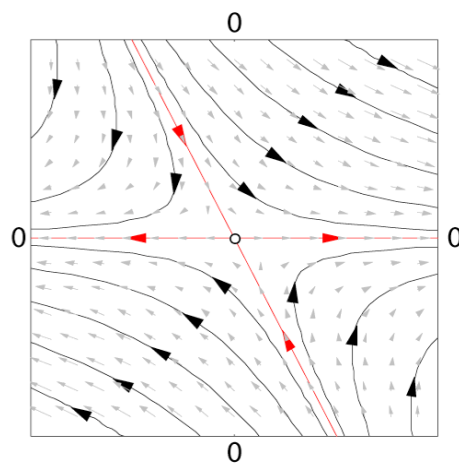
Stable node



Stable focus

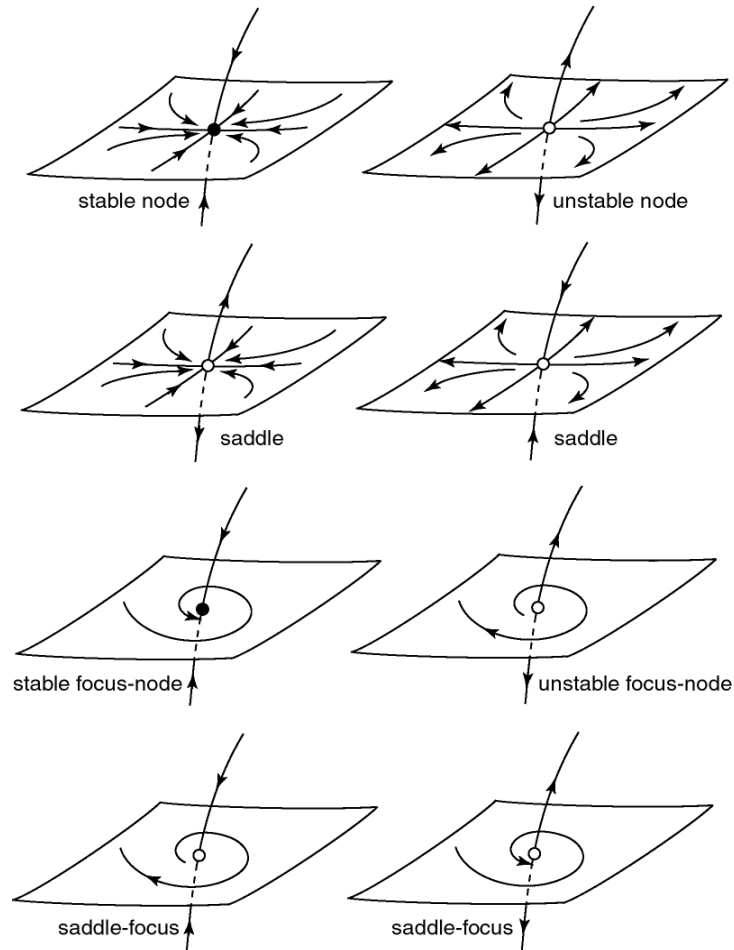
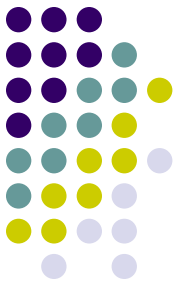


Unstable node

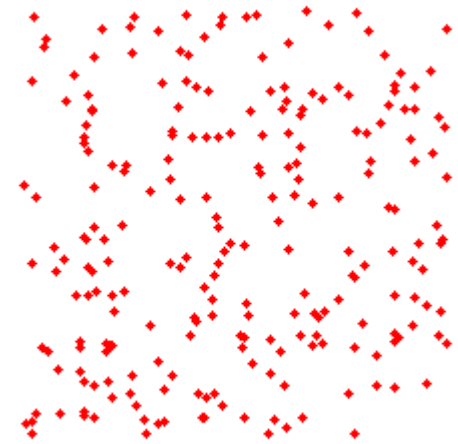
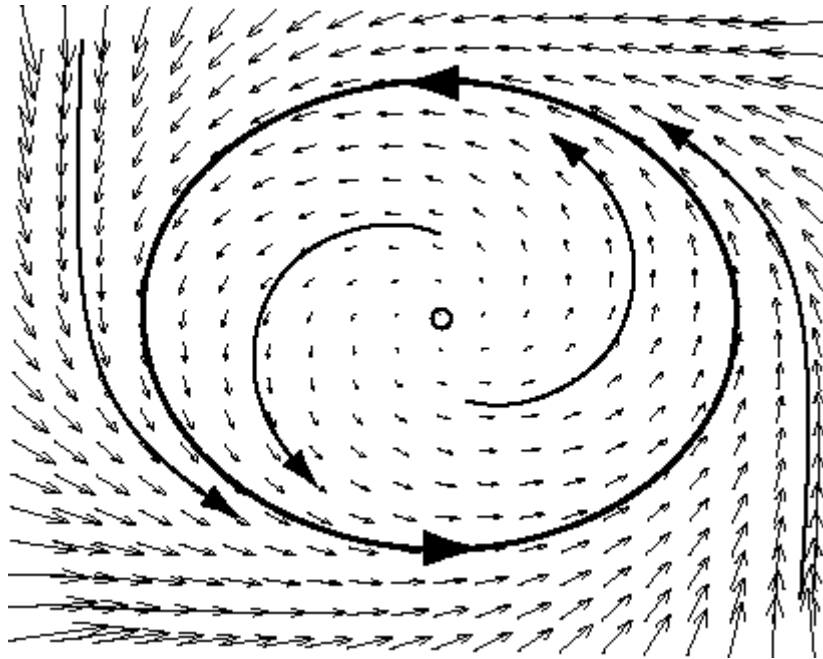
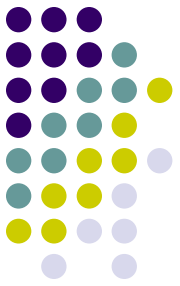


Saddle

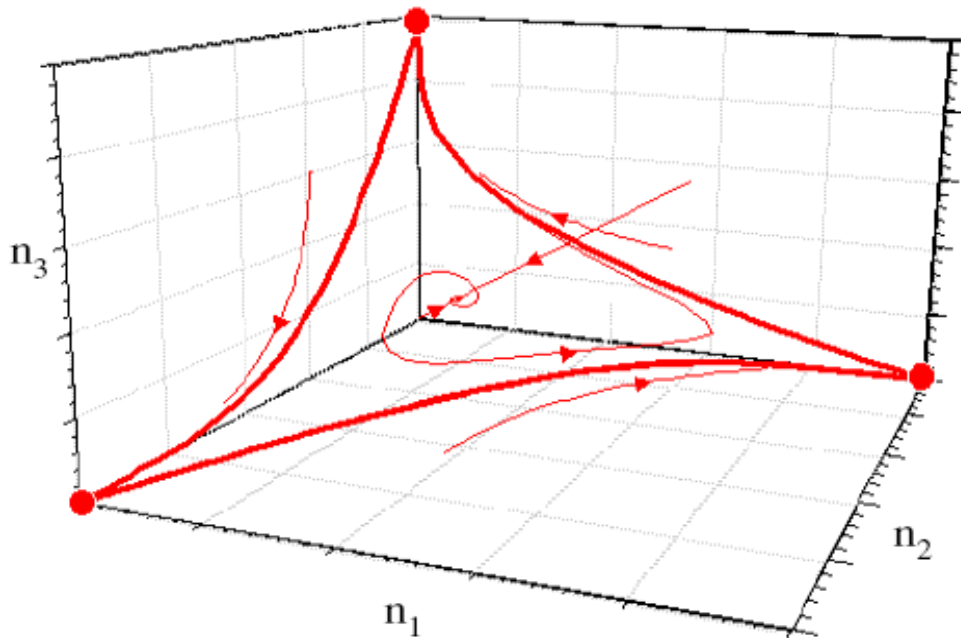
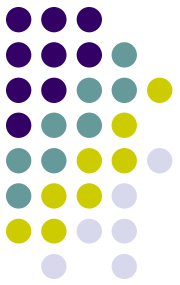
Equilibria in 3D



Limit cycles



Heteroclinic sequences



Lottka-Volterra model:
competition between three species

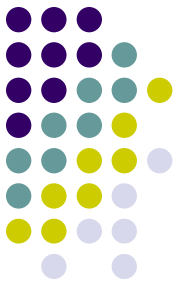
$$\dot{a}_1 = a_1[1 - (a_1 + \rho_{12}a_2 + \rho_{13}a_3)]$$

$$\dot{a}_2 = a_2[1 - (a_2 + \rho_{21}a_1 + \rho_{23}a_3)]$$

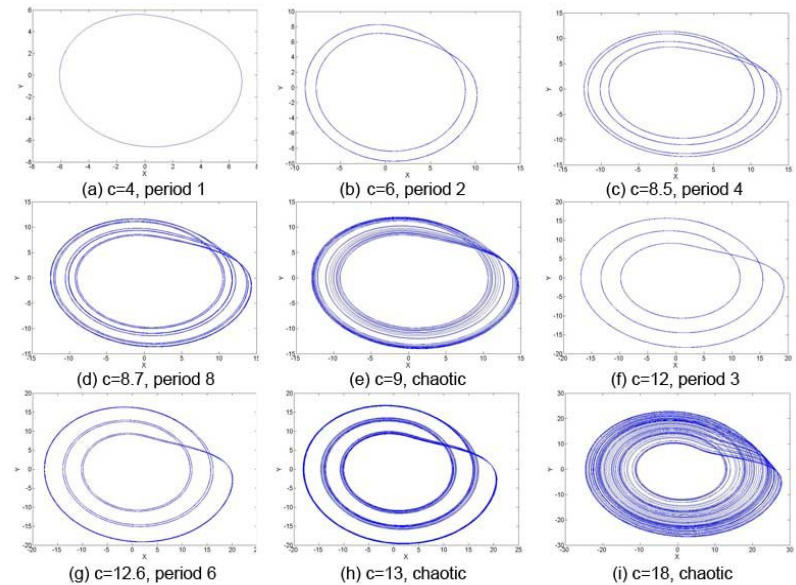
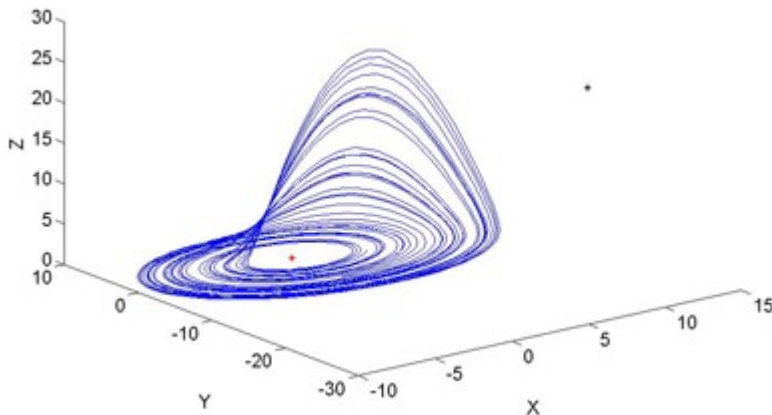
$$\dot{a}_3 = a_3[1 - (a_3 + \rho_{31}a_1 + \rho_{32}a_2)]$$

May and Leonard, 1975

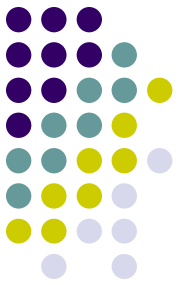
Chaotic attractors


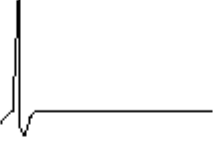

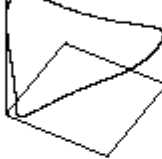
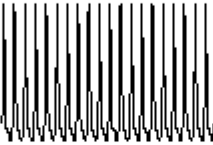


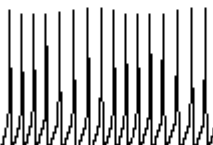






Rossler attractor

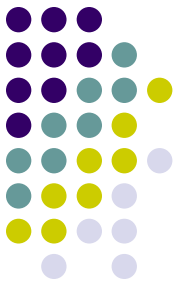


Hodgkin-Huxley model



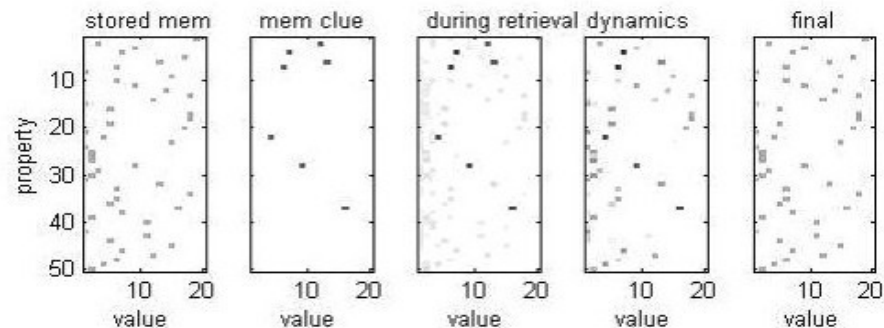
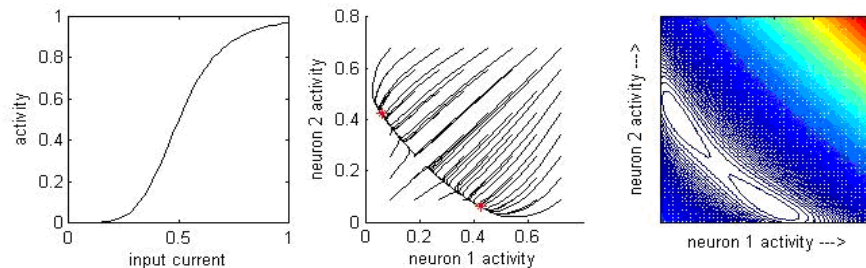
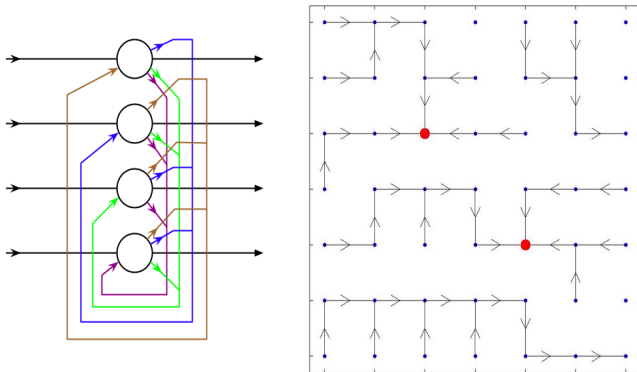
| Attractors | Dynamics | Trajectories in State Space | Time Series | Topological Structure | Dimension | Lyapunov Spectrum | Poincare Section |
|-------------------|----------------|---|--|-----------------------------|-------------|--|---|
| Equilibrium Point | Static |  |  | Point | 0 | $\lambda_1 < 0$ |  |
| Limit Cycle | Periodic |  |  | \mathbb{R}/\mathbb{Z} | 1 | $\lambda_1 = 0$ $\lambda_i < 0$ ($i \neq 1$) |  |
| Torus | Quasi-Periodic |  |  | $\mathbb{R}^k/\mathbb{Z}^k$ | k | $\lambda_i = 0$ ($i=1,2,\dots,k$) $\lambda_i < 0$ (otherwise) |  |
| Strange Attractor | Chaotic |  |  | Fractal | Real Number | $\lambda_i > 0$ ($i=1,2,\dots,n$) $\lambda_i = 0$ ($i=n+1,\dots,m$) $\lambda_i < 0$ (otherwise) |  |

Computing with attractors

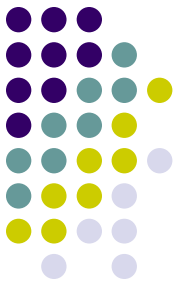


Associative memory: multi-stable system with convergence to one of the limiting sets from some initial condition/under some stimulus

Hopfield network

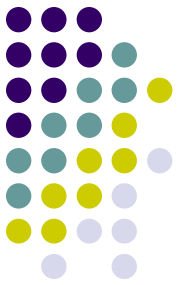


Transient dynamics

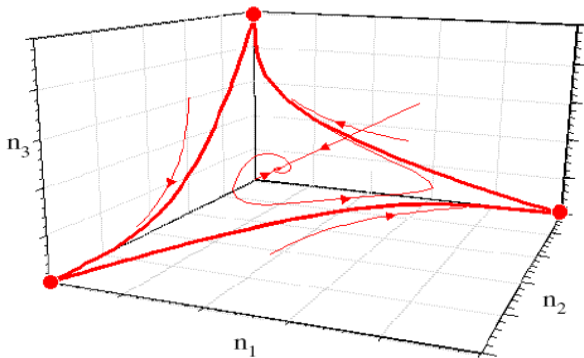


- Computation by attractors ignores dynamics, hence, is slow
- Equilibria or limit cycles cannot be realistically reached on often short timescales of neural computation
- No need for a classical attractor state
- No need for “resetting”

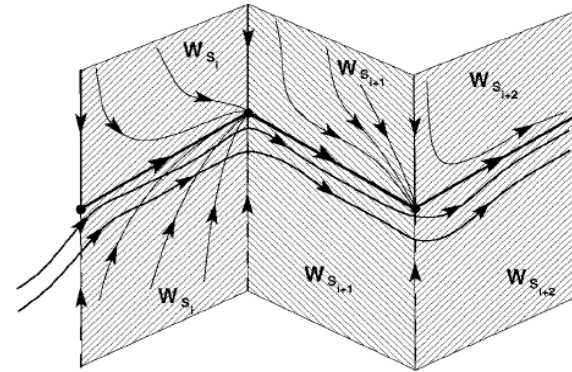
Transient dynamics



Small systems - periodic

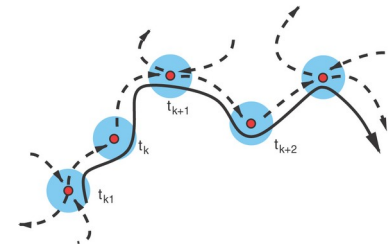


Large systems - sequential

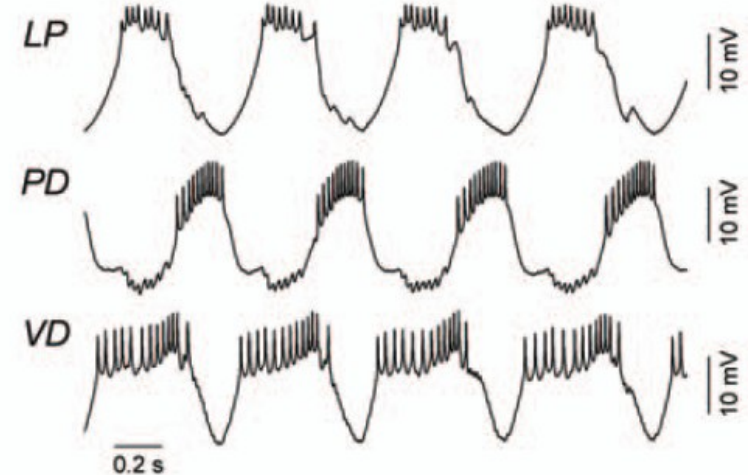
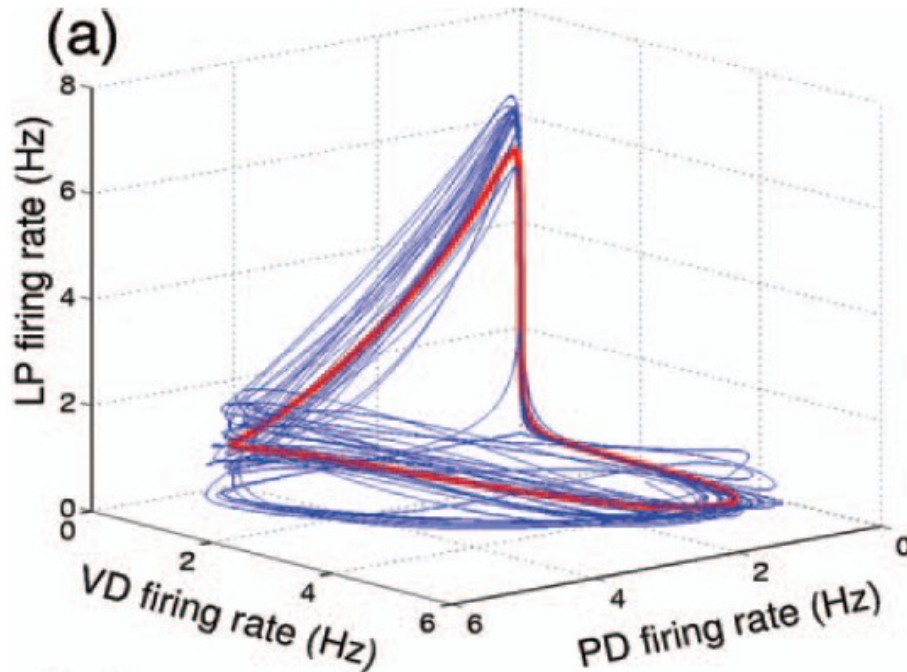
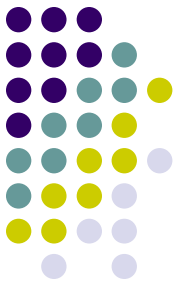


- Stability
- Robustness
- Variability

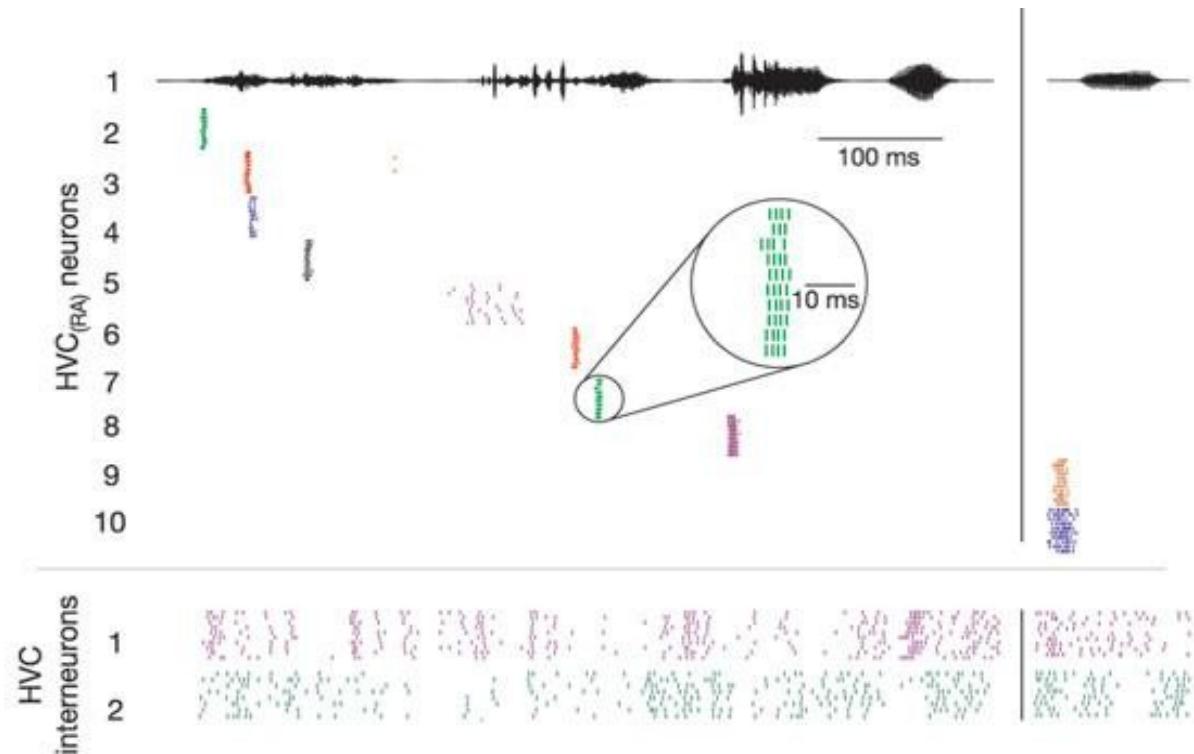
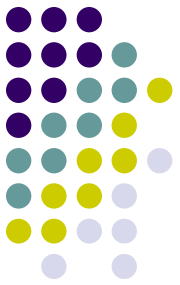
Decision-making



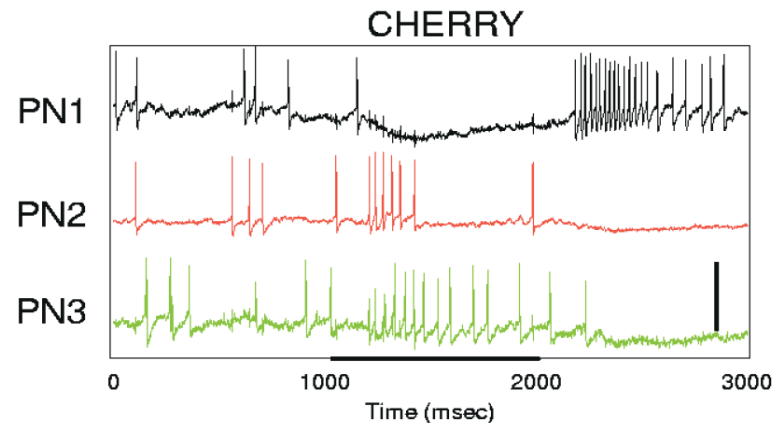
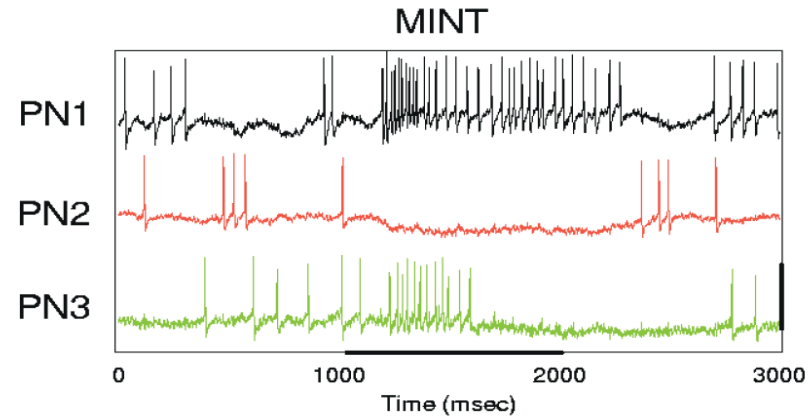
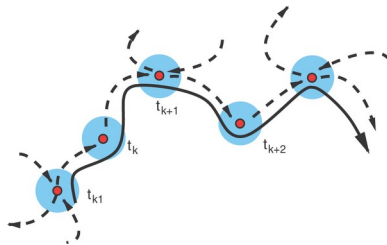
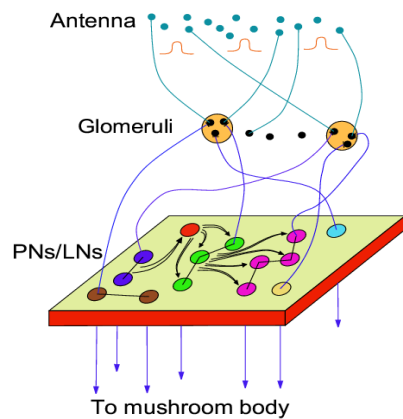
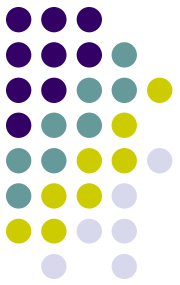
Pyloric rhythm of lobster stomatogastric ganglion



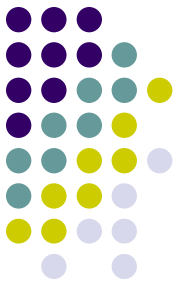
HVC Songbird patterns



Odor sensing (locust antennal lobe)



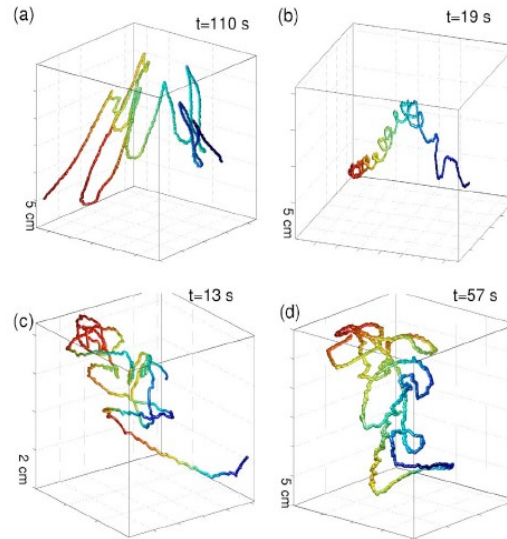
Clione's hunting



Clione (Sea angel)

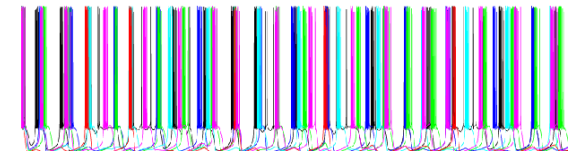
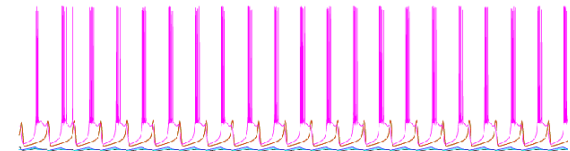
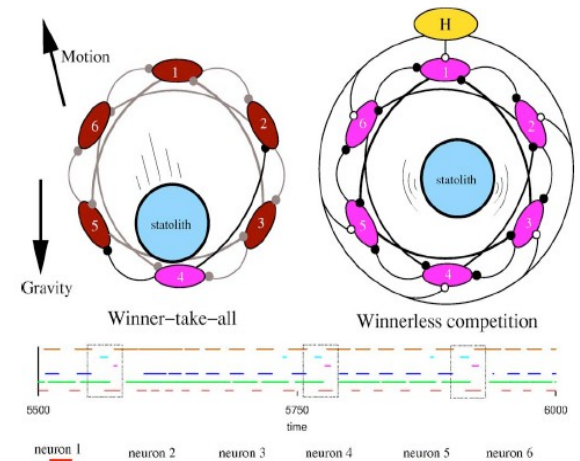


Sea-butterfly

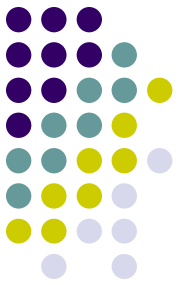


Winner-take-all

Winnerless competition



Summary



- Neurons exchange electrical signals and can be described by dynamical equations
- Neural networks: attractor computation or transient dynamics?
- Transient dynamics: robustness, reproducibility, variability
- Image: stable heteroclinic sequences
- Mathematics to follow...