

Designing noisy networks

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Task switching

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Switch cost

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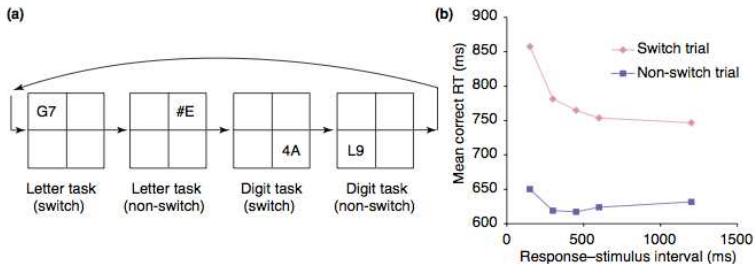
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Can we model this in a useful way?

Image from: Rogers, R.D. and Monsell, S. (1995) The costs of a predictable switch between simple cognitive tasks. *J. Exp. Psychol. Gen.* 124, 207–231

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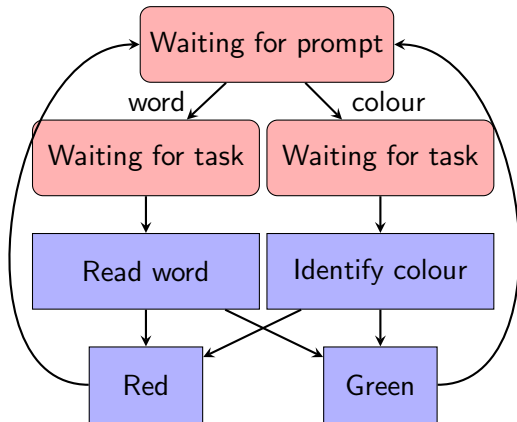
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Heteroclinic cycles

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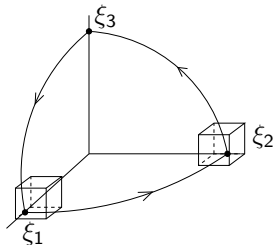
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Summary

Consider the ODE:

$$\dot{x} = f(x), \quad x \in \mathbb{R}^n. \quad (1)$$

- An **equilibrium** ξ of (1) satisfies $f(\xi) = 0$.
- A solution ϕ_j of (1) is a **heteroclinic connection** from ξ_j to ξ_{j+1} , if it is backward asymptotic to ξ_j and forward asymptotic to ξ_{j+1} .
- A **heteroclinic cycle** is a set of equilibria $\{\xi_1, \dots, \xi_m\}$ and orbits $\{\phi_1, \dots, \phi_m\}$, where ϕ_j is a heteroclinic connection between ξ_j and ξ_{j+1} , and $\xi_1 \equiv \xi_{m+1}$.



Noisy heteroclinic cycles

Stone and Holmes, 1990

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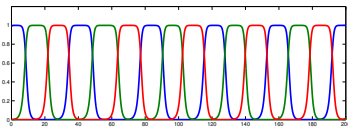
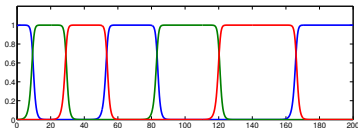
Summary

- Consider additive noise to a heteroclinic cycle, i.e. the SDE

$$dx_t = f(x_t) + \eta dW_t, \quad x_t \in \mathbb{R}^n$$

- W_t is n -dimensional Brownian motion
- η is noise amplitude.
- Mean passage time past an equilibrium

$$T \sim \frac{\log \eta}{\lambda_u}$$



Heteroclinic networks

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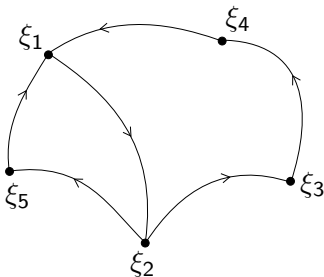
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- A heteroclinic network is a connected union of heteroclinic cycles.



- Stability conditions of the network as a whole may be quite complicated.
- Nearby trajectories may switch between different sub-cycles of the network.
- Noise can have unexpected effects.

Excitable networks

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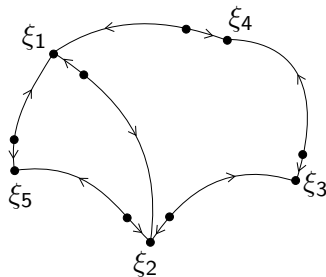
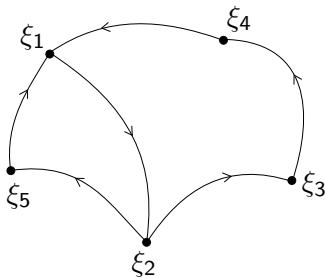
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- Consider as a pitchfork bifurcation from a heteroclinic network.
- Original equilibria are now all stable.
- Small amplitude perturbations can push trajectories between equilibria.



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Designing networks

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Consider using heteroclinic networks to model neural processes. Design a graph structure describing the process, each node corresponds to an equilibrium.

Some questions:

- For a given graph, can we construct an ODE which contains that graph embedded as a heteroclinic or excitable network?
- Can we control the residence times near the equilibria?
- Can we control the switching probabilities between nodes?

Network construction

Ashwin and P, 2014, 2015

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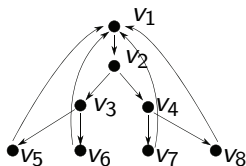
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- Consider a graph with n_v vertices and n_e edges.
- Several construction methods:
 - Simplex network: in \mathbb{R}^{n_v} , each vertex is an equilibrium on a coordinate axes, connections live in two-dimensional coordinate planes.
 - Cylinder network: in \mathbb{R}^{n_e+1} , vertices lie in a line, each connecting orbit lives in a two-dimensional plane.
 - Excitation-inhibition network: in $\mathbb{R}^{n_e+n_v}$, reminiscent of neuronal systems.
 - Others...

Network construction: Time series

Ashwin and P, 2014, 2015

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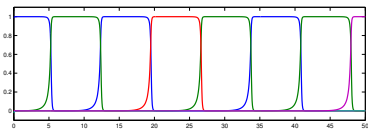
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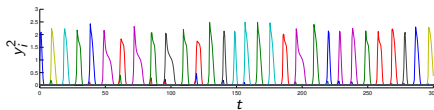
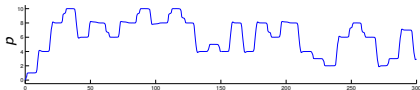
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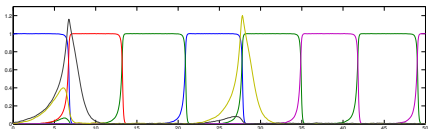
■ Simplex method:



■ Cylinder method:



■ Excitation-inhibition:



Network statistics: Noisy Kirk and Silber network

Armbruster, Stone and Kirk, 2003

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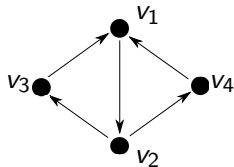
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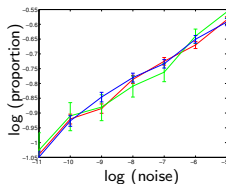
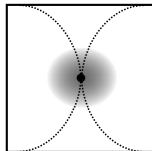
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- How does noise affect the dynamics near the network?
 - Residence times
 - Switching rates
- Consider a probability density function of trajectories, assume it is centered at the origin.
- Proportion of times each cycle visited proportional to shaded area.



Network statistics: Noisy Kirk and Silber network

Armbruster, Stone and Kirk, 2003

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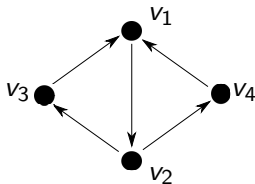
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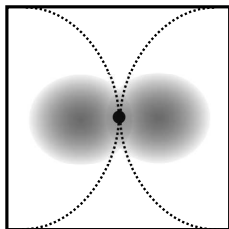
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Summary



- For certain parameter sets, noise ellipse can move into basin of attraction of one cycle or the other.
- This is termed *lift-off*.
- Lift-off can *reduce* switching, and cause *memory*.



Escape rates

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- To compute mean residence times near equilibria, we compute *escape rates* from a potential well.
- Consider

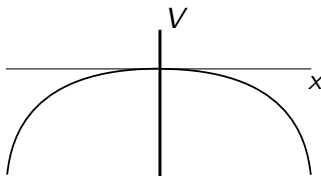
$$dx = -V'(x)dt + \eta dW_t$$

with potential

$$V(x) = \frac{\nu x^2 - x^4}{2}$$

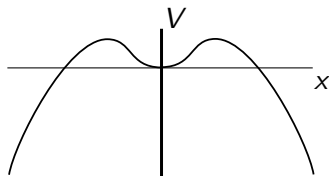
$$\nu < 0$$

Heteroclinic connection



$$\nu > 0$$

Excitable connection



Mean residence times

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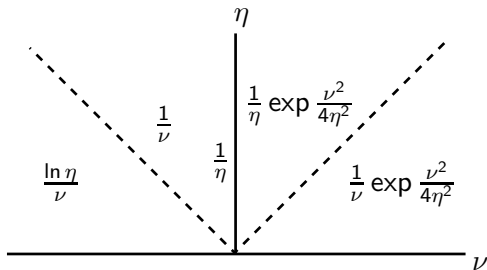
Escape rates

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Summary

- Compute the length of time spent near $x = 0$ when η and ν are small.

$$\bar{T}(\nu, \eta) = \frac{2}{\eta^2} \int_{z=0}^a \int_{y=0}^z \exp \frac{\nu(z^2 - y^2) + (y^4 - z^4)}{\eta^2} dy dz.$$



Mean residence times with forcing

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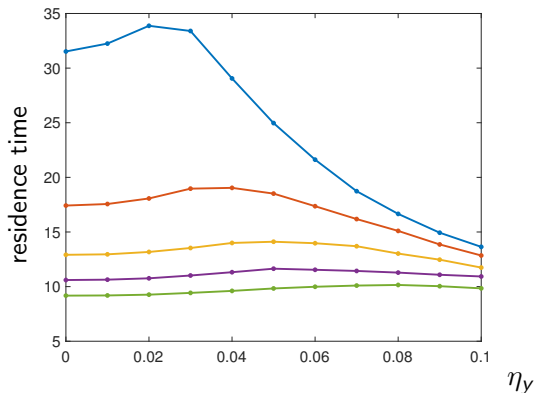
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Summary

- Consider a similar system, now non-autonomous, and 'forcing' can push the trajectories in a desired direction around the network.



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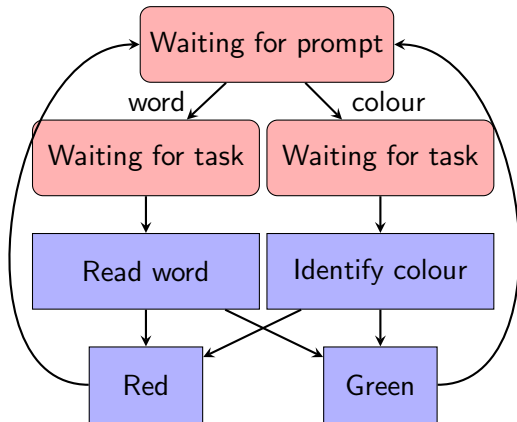
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- We can design heteroclinic and excitable networks to embed any specified graph in phase space.
- We can adjust noise parameters to fit mean residence times.
- Switching probabilities, residence time distributions, anisotropic noise....

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