

Prospects for Resilience Quantification Using Dynamical Systems Theory

Kate Meyer
Minnesota Math Climate Seminar
March 22nd, 2016

Outline

Motivations

- Ecological resilience

- Translating the qualitative to quantitative

Resilience to state variable perturbations

- Definitions from the ecology literature

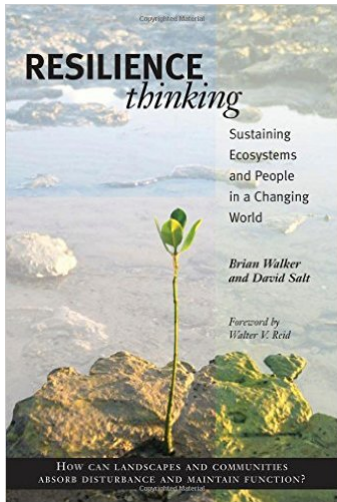
- Mathematical tools: ϵ -pseudo-orbits and intensity of attraction

Resilience to parameter perturbations

- Definition from the ecology literature

- Mathematical tools: transient and rapid parameter changes

Future directions



Resilience

“[T]he capacity of [a] system to absorb change and disturbances and still retain its basic structure and function”

– Walker & Salt (2006)



Example



Oligotrophic Lake

Nutrient Loading
Disruption to
food web

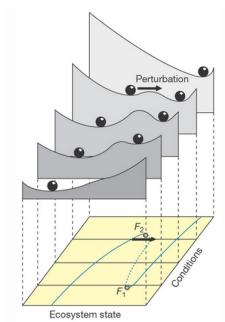


Eutrophic Lake

Images: http://resac.gis.umn.edu/water/regional_water_clarity/content/project_summary.htm



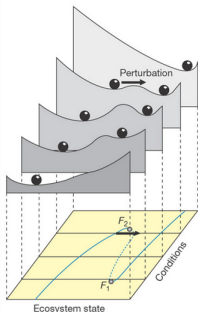
Translating to Dynamical Systems



Scheffer et al. (2001)



Translating to Dynamical Systems



Ecosystems (2001) 4: 765–781
DOI: 10.1007/s10021-001-0045-9

ECOSYSTEMS
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MINIREVIEW

From Metaphor to Measurement: Resilience of What to What?

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ABSTRACT

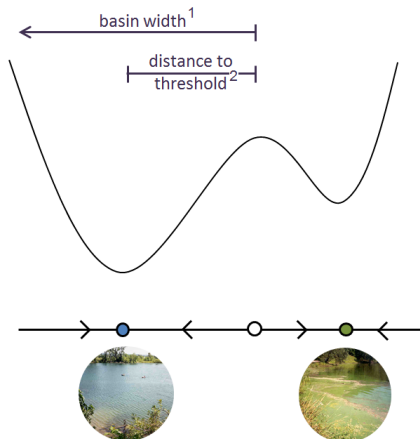
Resilience is the magnitude of disturbance that can be tolerated before a socioecological system (SES) moves to a different region of state space controlled by a different set of processes. Resilience has multiple levels of meaning: as a metaphor related to sustainability, as a property of dynamic models, and as a measurable quantity that can be assessed in field studies of SES. The operational indicators of resilience have, however, received little attention in the literature. To assess a system's resilience, one must specify which system configuration and which

Examples are soil phosphorus content in lake districts woody vegetation cover in rangelands, and property rights systems that affect land use in both lake districts and rangelands. (b) The ability of an SES to self-organize is related to the extent to which reorganization is endogenous rather than forced by external drivers. Self-organization is enhanced by coevolved ecosystem components and the presence of social networks that facilitate innovative problem solving. (c) The adaptive capacity of an SES is related to the existence of mechanisms for the evolu-

Scheffer et al. (2001)

Carpenter et al. (2001)

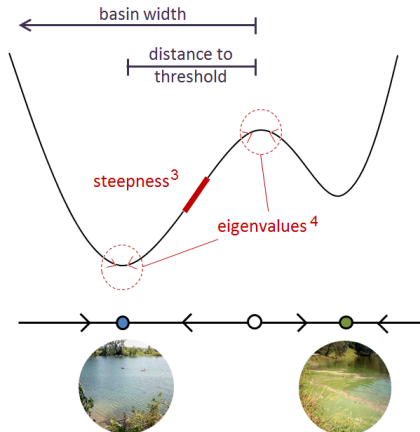
Resilience to state variable perturbations



¹ Levin & Lubchenco (2008), Walker et al. (2004)

² Beisner et al. (2003a)

Resilience to state variable perturbations

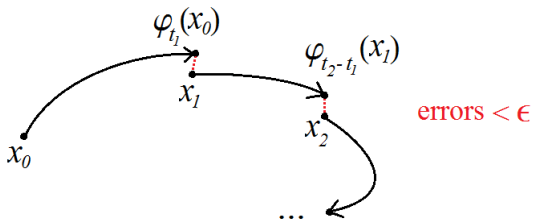


³ Beisner et al. (2003b), Levin & Lubchenco (2008)

⁴ Ives & Carpenter (2007), Pimm (1984)

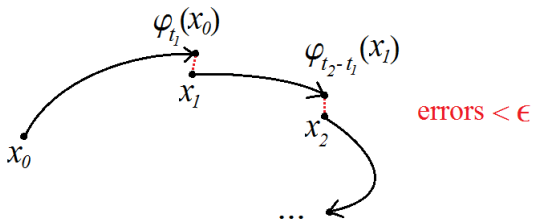
ϵ -pseudo-orbits

Classically defined for flows:



ϵ -pseudo-orbits

Classically defined for flows:



Fix a disturbance period, τ :

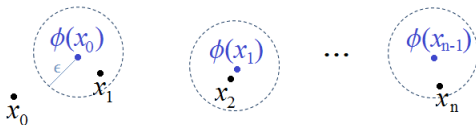
$$\varphi(t, x) : \mathbb{R} \times X \rightarrow X \quad \rightsquigarrow \quad \varphi^\tau(x) : X \rightarrow X$$



Following McGehee (1988): Let X be a locally compact metric space and $\phi : X \rightarrow X$ a continuous map.

Definition

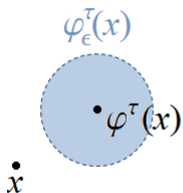
An ϵ -**pseudo-orbit** of length n is a sequence of points (x_0, x_1, \dots, x_n) in the state space X satisfying $d(\phi(x_{k-1}), x_k) < \epsilon$, for $k = 1, 2, \dots, n$



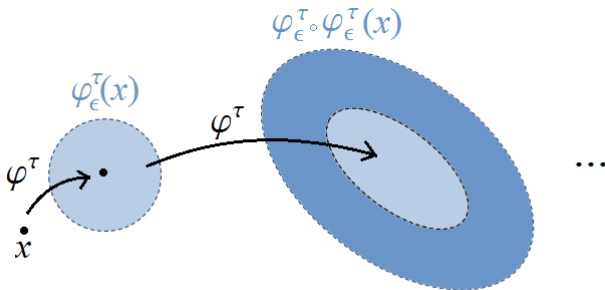
Consider the multi-valued map

$$\varphi_\epsilon^\tau : X \rightarrow \mathcal{P}(X)$$

$$x \mapsto \{y \in X : d(y, \varphi^\tau(x)) < \epsilon\}$$



Let
$$P_\epsilon(S) = \bigcup_{n=0}^{\infty} (\varphi_\epsilon^\tau)^n(S)$$



Chain intensity of attraction

Definition

The **chain intensity** of an attractor A is

$$\mu(A) \equiv \sup\{\epsilon : P_\epsilon(A) \subset K \subset \mathcal{D}(A), \text{ where } K \text{ is compact}\}$$



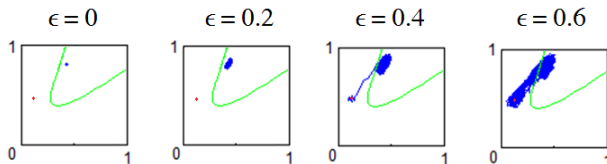
Chain intensity of attraction

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Chain intensity is amenable to numerical approximation:

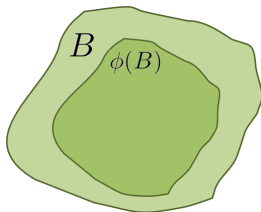


- simulated ϵ -pseudo-orbits
- boundary of domain of attraction

Attractor blocks

Definition

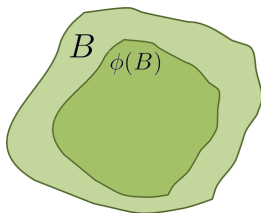
An **attractor block** for a map ϕ is a nonempty, compact set B such that $\phi(B) \subset \text{int}(B)$.



Attractor blocks

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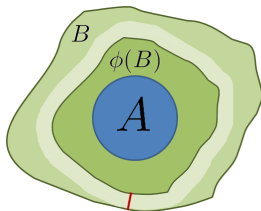
Theorem

For a continuous map on a locally compact metric space,

- every attractor block contains an attractor in its interior, and
- all attractors are the maximal invariant set inside some attractor block.

Definition

$$\beta(B) \equiv \sup\{\epsilon : \varphi_\epsilon^T(B) \subset B\}$$



Intensity of attraction

Definition

The **intensity** of A is

$$\nu(A) \equiv \sup\{\beta(B) : B \text{ is an attractor block associated with } A\}$$

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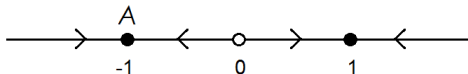
Theorem

$$\mu(A) = \nu(A)$$

Towards a resilience profile

Example

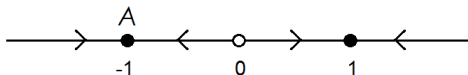
$$\dot{x} = x - x^3$$



Towards a resilience profile

Example

$$\dot{x} = x - x^3$$

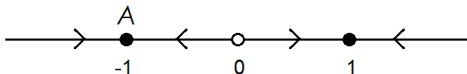


$$\mu^{\tau}(A) = \nu^{\tau}(A) = \sup_{x \in (-1, 0)} \{x - \varphi^{\tau}(x)\}$$

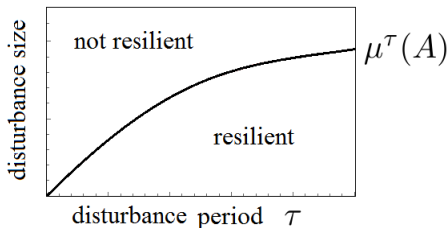
Towards a resilience profile

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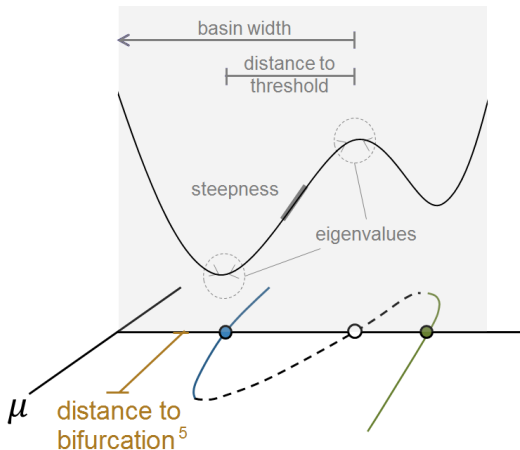
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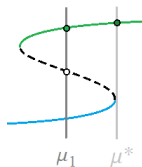
Quantifying resilience to parameter changes



Time-dependent parameter changes

Transient parameter change

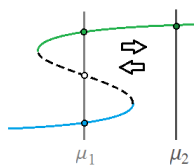
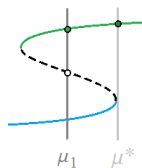
Rate-induced tipping



Time-dependent parameter changes

Transient parameter change

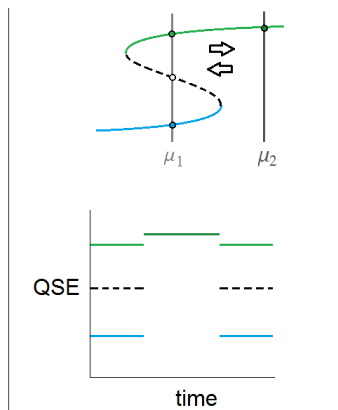
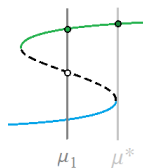
Rate-induced tipping



Time-dependent parameter changes

Transient parameter change

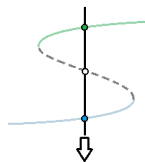
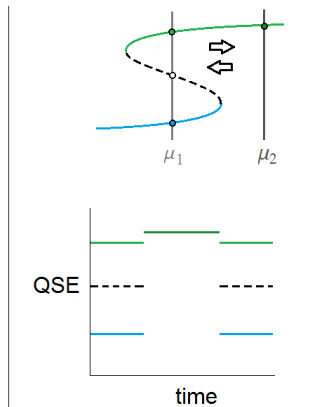
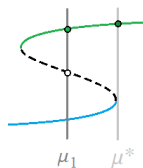
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Time-dependent parameter changes

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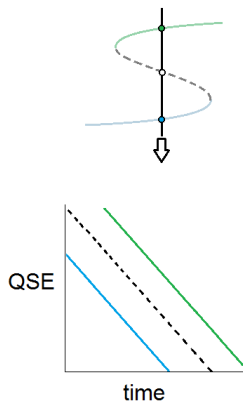
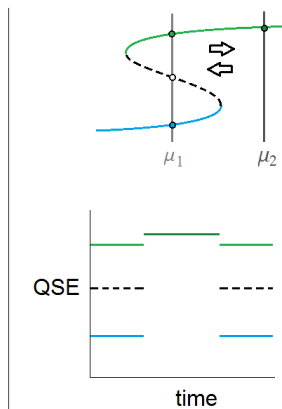
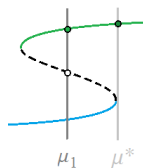
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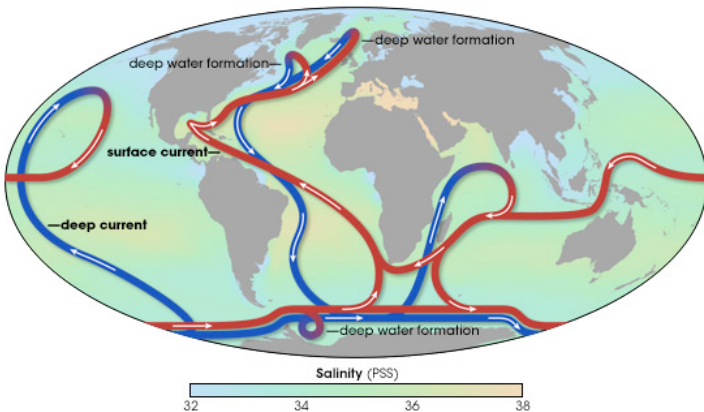
Time-dependent parameter changes

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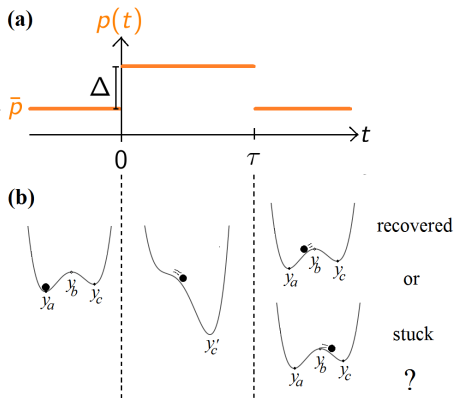
Example: transient change in salinity forcing of ocean circulation

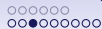


http://earthobservatory.nasa.gov/Features/Paleoclimatology_Evidence/paleoclimatology_evidence_2.php

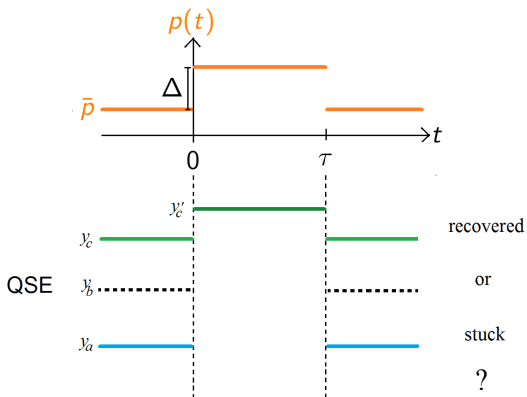
Cessi's 1D model

$$\dot{y} = \bar{p} + \hat{p}(t) - y [1 + \mu^2(1 - y)^2], \text{ where } \hat{p}(t) = \begin{cases} \Delta & 0 \leq t \leq \tau \\ 0 & \text{o'wise} \end{cases}$$

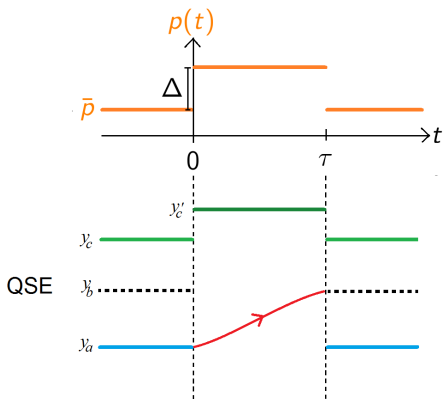




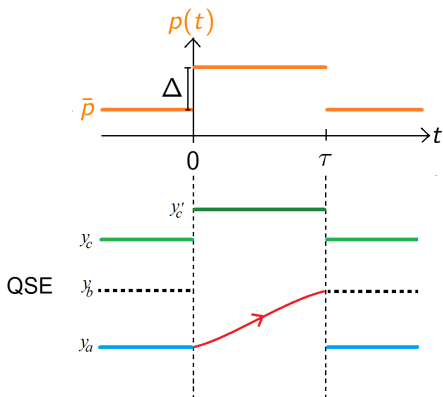
Cessi's 1D model



Calculating critical time in terms of Δ



Calculating critical time in terms of Δ



$$\tau_{\text{crit}}(\Delta) = \int_0^{\tau_{\text{crit}}} dt = \int_{y_a}^{y_b} \frac{dy}{\bar{p} + \Delta - y[1 + \mu^2(1 - y)^2]}$$

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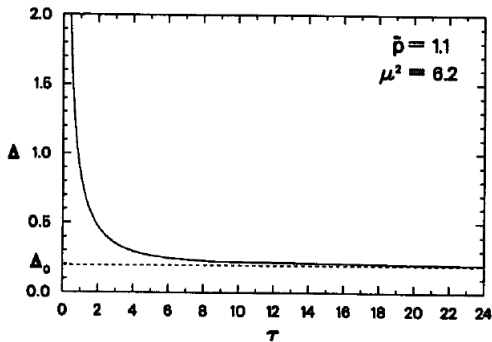
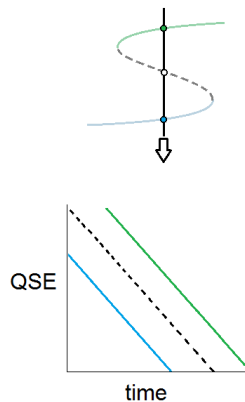
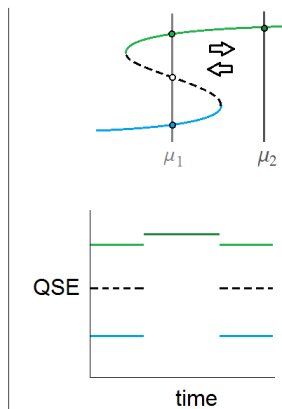
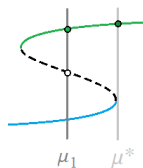


Figure 3 from Cessi (1994)

Time-dependent parameter changes

Transient parameter change

Rate-induced tipping



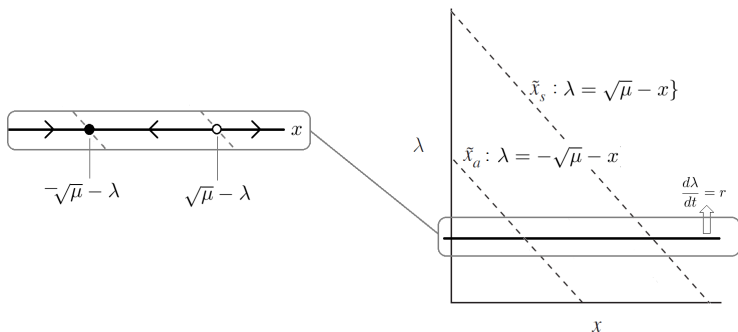
Rate-induced tipping

An example from Ashwin et al. (2012):

$$\frac{dx}{dt} = (x + \lambda)^2 - \mu \quad (\mu > 0 \text{ fixed})$$

with

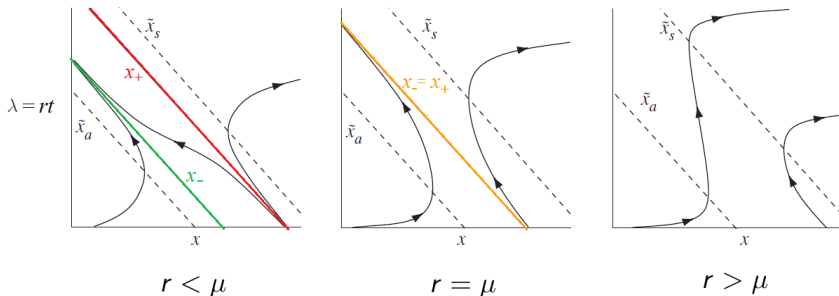
$$\frac{d\lambda}{dt} = r$$





$$\frac{dx}{dt} = (x + \lambda)^2 - \mu \quad (\mu > 0 \text{ fixed})$$

$$\frac{d\lambda}{dt} = r$$



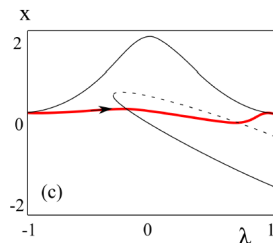
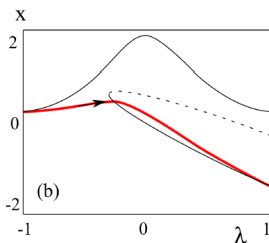
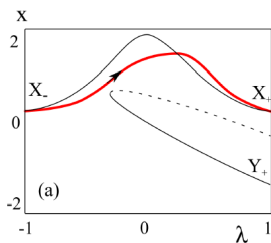
adapted from Figure 3 of Ashwin et al. (2012)

Asymptotically autonomous parameter shifts

An example from Ashwin et al. (2015):

$$\frac{dx}{dt} = - \left((x - 0.25 + b\lambda)^2 - 0.4 \tanh(\lambda + 0.3) \right) (x - K / \cosh(3\lambda))$$

$$\frac{d\lambda}{dt} = -r(\lambda + 1)(\lambda - 1)$$



Future directions

- Develop a continuous analogue to intensity of attraction
- Generalize Cessi's method to higher dimensions
- Generalize Cessi's method to smooth parameter changes

Develop a continuous analogue to intensity of attraction

	Maps	Flows
Space X	locally compact, metric	locally compact, metric
Disturbed Trajectories	$\phi_\epsilon(x) : X \rightarrow \mathcal{P}(X)$	$\varphi_r(t, x) : \mathbb{R} \times X \rightarrow \mathcal{P}(X)$ flow on $f + g$, $ g < r$
Region of Accessibility	$P_\epsilon(A)$	$P_r(A)$
Chain Intensity	$\sup\{\epsilon : P_\epsilon(A) \subset K \subset \mathcal{D}(A)\}$	$\sup\{r : P_r(A) \subset K \subset \mathcal{D}(A)\}$
Attractor Block	compact, nonempty, $\phi(B) \subset \text{int}(B)$	compact, nonempty, $\varphi^t(B) \subset \text{int}(B) \forall t > 0$
Intensity	$\sup\{\beta(B)\}$	

Generalize Cessi's method to higher dimensions

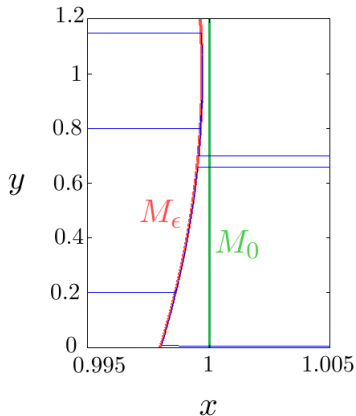
Cessi's 1D system

$$\dot{y} = p - y [1 + \mu^2(1 - y)^2]$$

came from

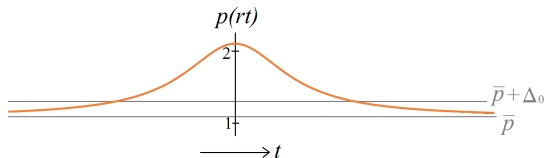
$$\begin{aligned} x' &= -(x - 1) - \epsilon x [1 + \mu^2(x - y)^2] \\ y' &= \epsilon (p - y [1 + \mu^2(x - y)^2]) \end{aligned}$$

via fast-slow reduction.



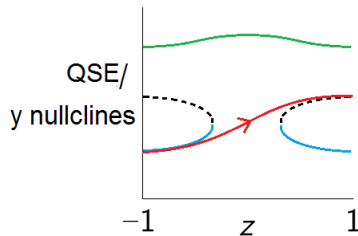
Generalize Cessi's method to smooth parameter changes

e.g. $p(rt) = \bar{p} + \frac{\Delta}{1+(rt)^2}$



Let $\dot{t} = 1$

$$z = \frac{2}{\pi} \tan^{-1}(t)$$



Motivations



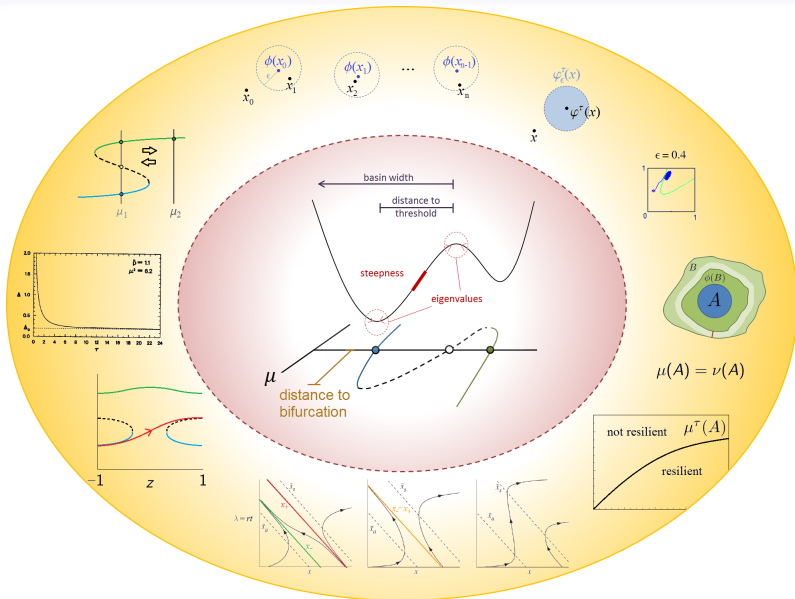
Resilience to state variable perturbations



Resilience to parameter perturbations



Future directions



References

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Motivations

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Resilience to state variable perturbations

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Resilience to parameter perturbations

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Future directions

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Thank you.

Questions?