

B5.6: Nonlinear Dynamics, Bifurcations and Chaos

- Lecturer: Radek Erban
- Lectures: Wednesdays 4pm and Fridays 4pm (lecture room L2)
- Prerequisites: This course builds on ten Prelims and Part A courses. Students taking this course should have mastered the material in Part A courses on Differential Equations and Complex Analysis, and Prelims courses covering Probability, Computational Mathematics, Introductory Calculus, Multivariable Calculus, Fourier Series and PDEs, Geometry, Dynamics and Constructive Mathematics.
- Problem Sheet 0: You were asked to solve it before our first lecture. The solutions to Problem Sheet 0 will be provided in our first lecture (today).
- Classes: The course is accompanied by four Problem Sheets (labelled 1, 2, 3, and 4), which will be discussed in your classes, and by Problem Sheet 0. Three classes, covering Problem Sheets 1, 2 and 3, are scheduled in Hilary Term. Your last class will be in Trinity Term and will cover Problem Sheet 4, your vacation work.

MSc Students: A2 Mathematical Methods II

- The first 8 lectures of this course is part of the core syllabus for the MSc in Mathematical Modelling and Scientific Computing A2 Mathematical Methods.
- Course synopsis is divided into two parts **Lectures 1-8** and **Lectures 9-16**
- **Lectures 1-8:**

Discrete-time (maps) and continuous-time (differential equations) dynamical systems. Notion of flows, stability of fixed points, Lyapunov function, invariant manifolds, stable manifold theorem, notion of hyperbolicity, center manifold. Chemical reaction networks. Stable, unstable and center subspaces. Poincaré-Bendixson theorem. Periodic solutions, stable and unstable limit cycles. Introduction to bifurcation theory, covering saddle-node, transcritical, supercritical pitchfork and subcritical pitchfork bifurcations. Extended center manifold. Logistic map. Periodic points of maps. Stability of N-cycles. Period-doubling bifurcation. Sharkovsky's theorem. Invariant distribution.

Part B examination, MSc examination and exam preparation

- Past Part B and MSc exam papers are available here:

www.maths.ox.ac.uk/members/students/undergraduate-courses/examinations-assessments/past-papers

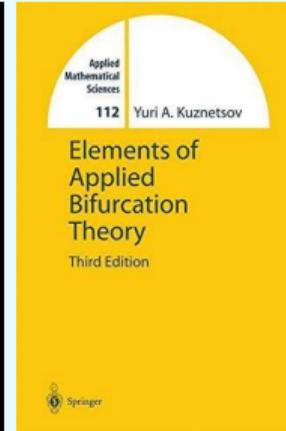
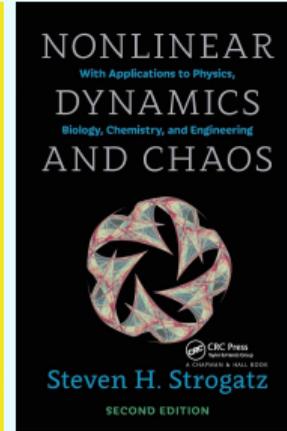
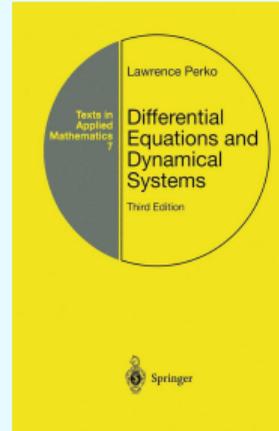
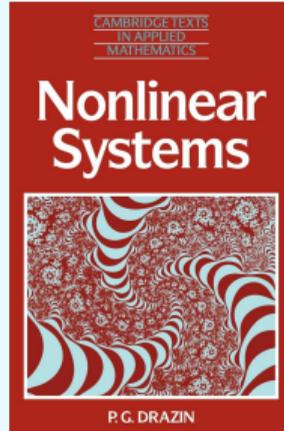
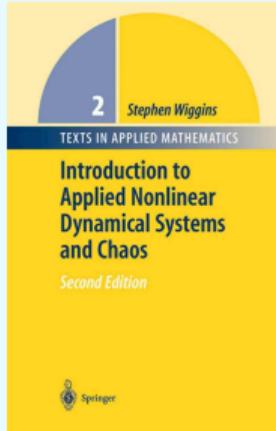
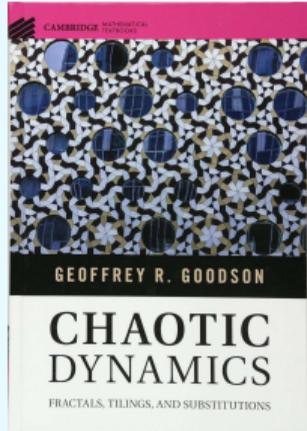
www.maths.ox.ac.uk/members/students/postgraduate-courses/msc-mmsc/core-courses

Please note that this course was called B5.6 *Nonlinear Systems* in some previous years. It was renamed to B5.6 *Nonlinear Dynamics, Bifurcations and Chaos* to give a clearer sense of what the course covers.

- First Notice to Candidates: “It must be stressed that to preserve the independence of the examiners, candidates are not allowed to make contact directly about matters relating to the content or marking of papers.”
- 2024 Examiner’s Report: “Most of the candidates demonstrated good understanding of the bookwork material. The exam was well-balanced with all three questions having the similar level of difficulty. While some candidates submitted some incomplete solutions, they often achieved at least 40% of raw marks. Other candidates also made successful attempts at more advanced parts of each question. In fact, each question received one complete solution (getting the perfect raw mark of 25), illustrating the solvability of each question under the exam conditions.”

Reading List and Lecture Notes

The B5.6 course material can be introduced with different levels of mathematical rigour, ranging from the 'definition-theorem-proof approach' to an example-based course covering dynamical systems appearing in applications. There are 6 books in the Reading List:



Students interested in building further theory with more proofs could like [Wiggins] or [Perko], or [Kuznetsov] (for bifurcations), or [Goodson] (for maps), while [Drazin] or [Strogatz] could be more appreciated by students interested in applications.

My slides will be uploaded to the course website after each lecture.

Introduction: main questions of course B5.6

Discrete-time dynamical system: Let $\mathbf{F} : \Omega \times \Theta \rightarrow \Omega$, where $\Omega \subset \mathbb{R}^n$ and $\Theta \subset \mathbb{R}^m$.

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Let $\mathbf{x}_0 \in \Omega$, $\boldsymbol{\mu} \in \Theta$ and $\mathbf{x}(t) \in \Omega$ be a solution of the ODE

$$\frac{d\mathbf{x}}{dt} = \mathbf{f}(\mathbf{x}; \boldsymbol{\mu}) \quad \text{with the initial condition} \quad \mathbf{x}(0) = \mathbf{x}_0$$

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We want to find \mathbf{x} as a function of t and sketch the phase plane or phase space.

What is the behaviour of $\mathbf{x}(t)$ as $t \rightarrow \infty$?

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Linear example (Question 1(a) on Problem Sheet 0):

$$\mathbf{x}_{k+1} = M\mathbf{x}_k \quad \text{for} \quad M = \begin{pmatrix} 2 & 1 & -1 \\ 1 & -1 & 2 \\ -1 & 1 & 2 \end{pmatrix} \quad \text{and} \quad \mathbf{x}_0 = \begin{pmatrix} 8 \\ 1 \\ 3 \end{pmatrix}$$

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Closed form formula for solutions [Prelims Probability and Calculus courses]:

$$\mathbf{x}_k = 3^k \begin{pmatrix} 2 \\ -1 \\ -3 \end{pmatrix} + (-2)^k \begin{pmatrix} 1 \\ -3 \\ 1 \end{pmatrix} + 2^k \begin{pmatrix} 5 \\ 5 \\ 5 \end{pmatrix}$$

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Linear example (Question 2(a) on Problem Sheet 0): $\mathbf{x}_{k+1} = M\mathbf{x}_k$ for $M = \begin{pmatrix} 1 & 2 \\ -1 & \mu \end{pmatrix}$

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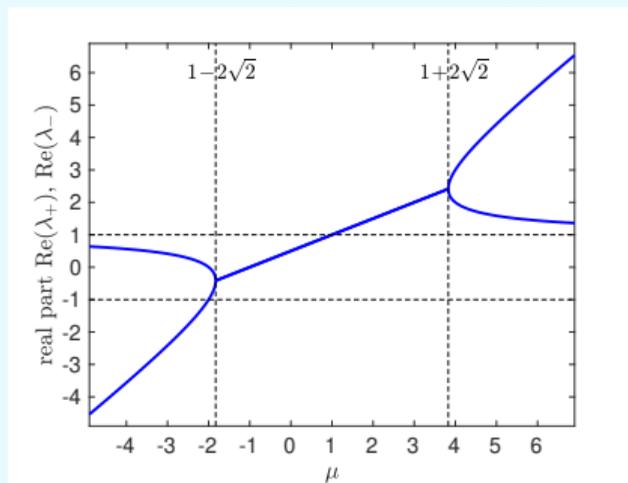
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eigenvalues of M are

$$\lambda_{\pm} = \frac{1 + \mu \pm \sqrt{(\mu - 1 + 2\sqrt{2})(\mu - 1 - 2\sqrt{2})}}{2}$$

general solution $\lambda_+^k \mathbf{v}_+ + \lambda_-^k \mathbf{v}_-$



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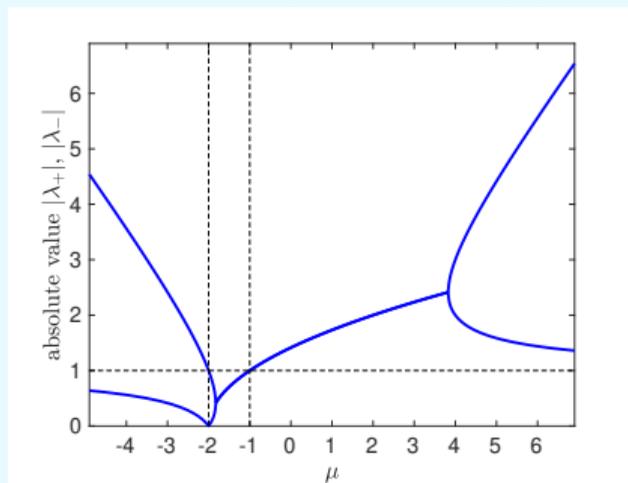
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$$\lim_{k \rightarrow \infty} \|\mathbf{x}_k\| = \infty \text{ for } \mu \in (-1, \infty)$$

$$\lim_{k \rightarrow \infty} \|\mathbf{x}_k\| = 0 \text{ for } \mu \in (-2, -1)$$



Nonlinear example: logistic map

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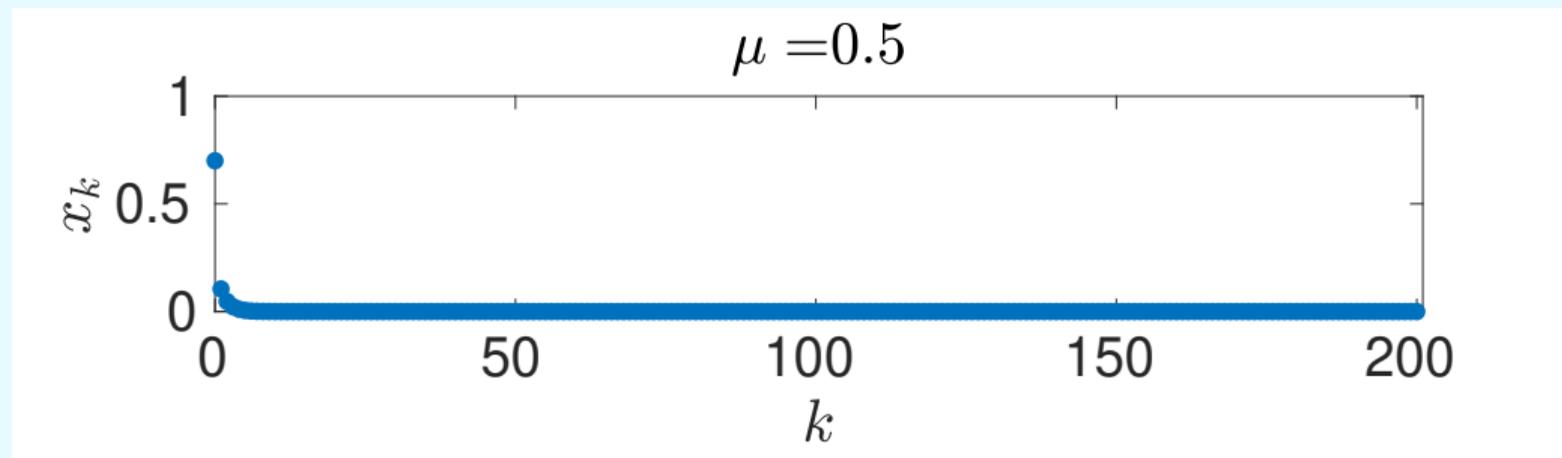
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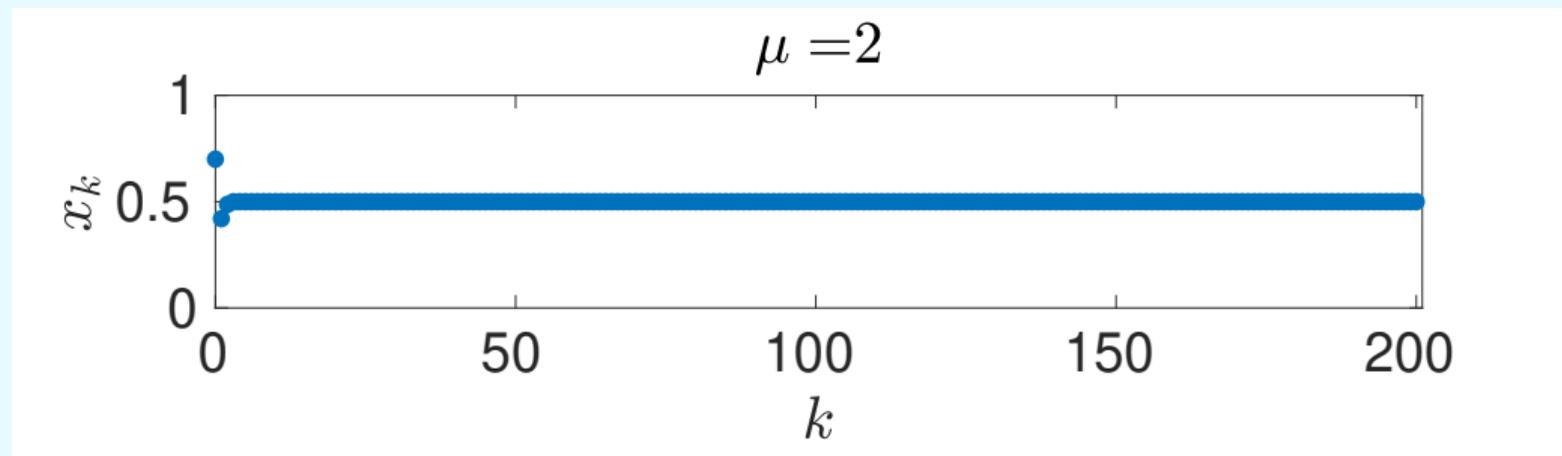


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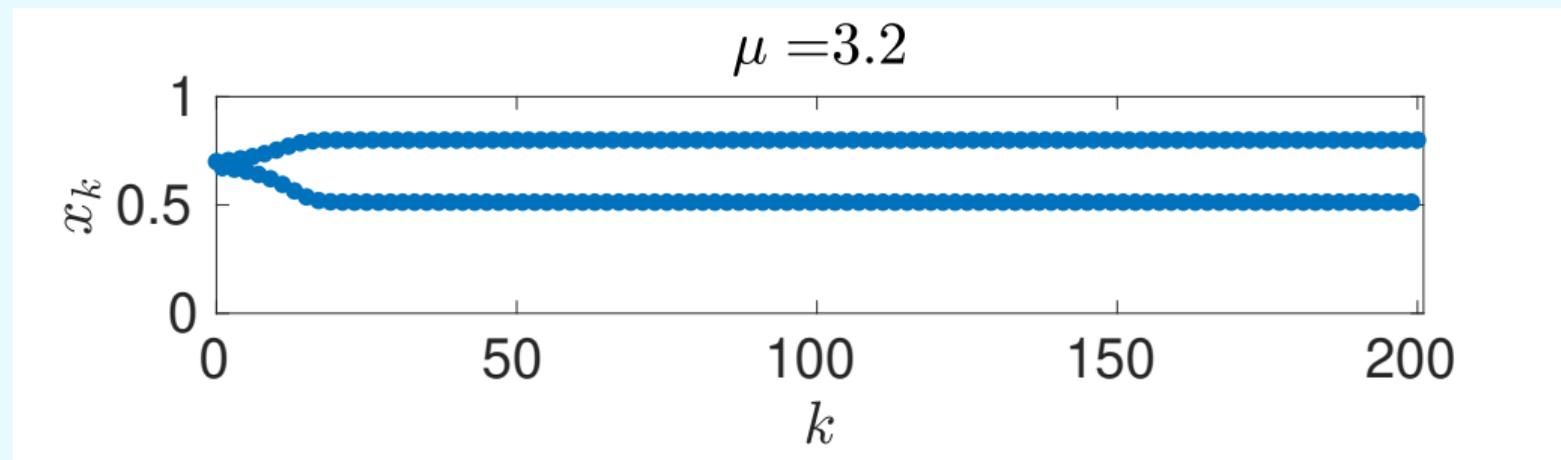


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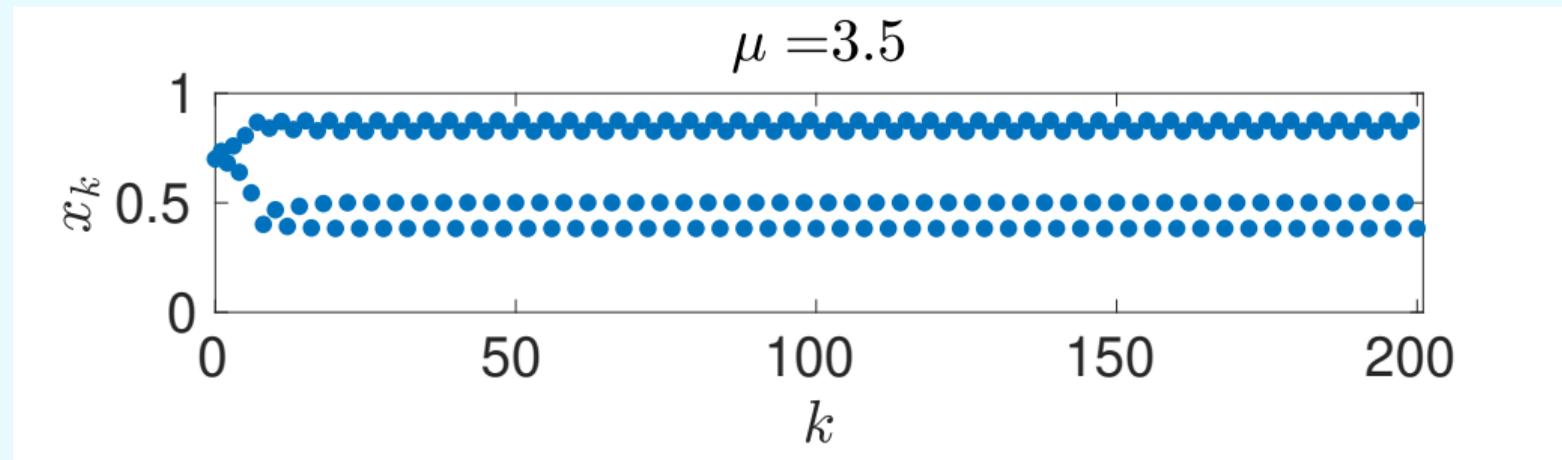


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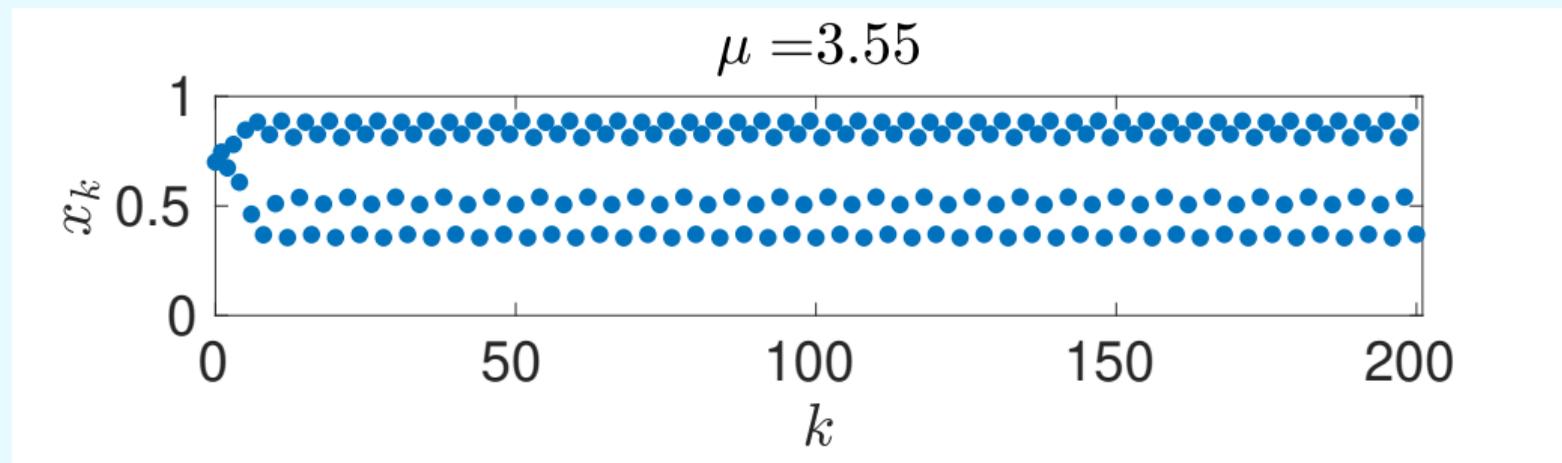


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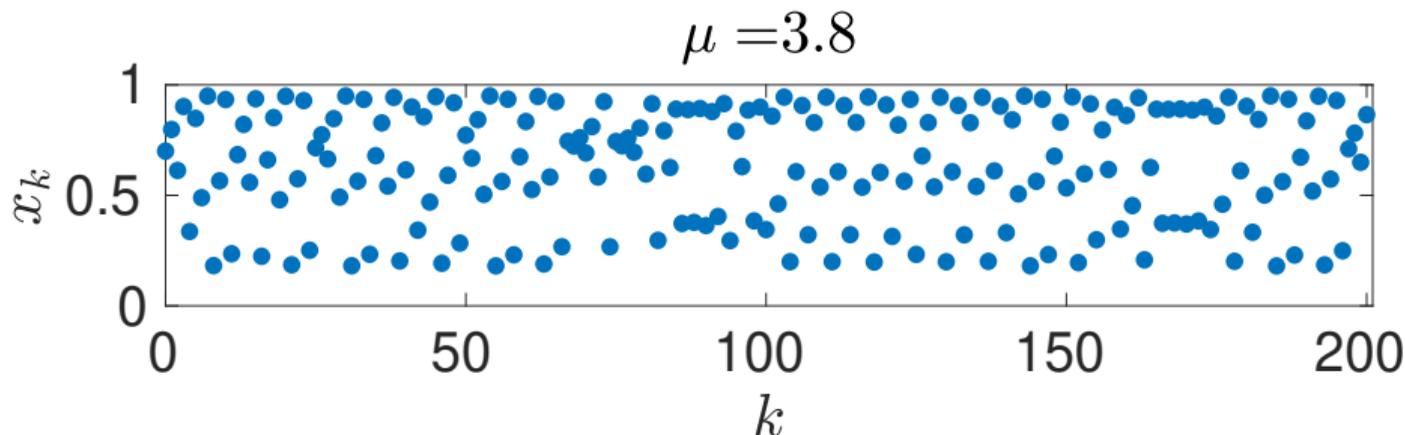


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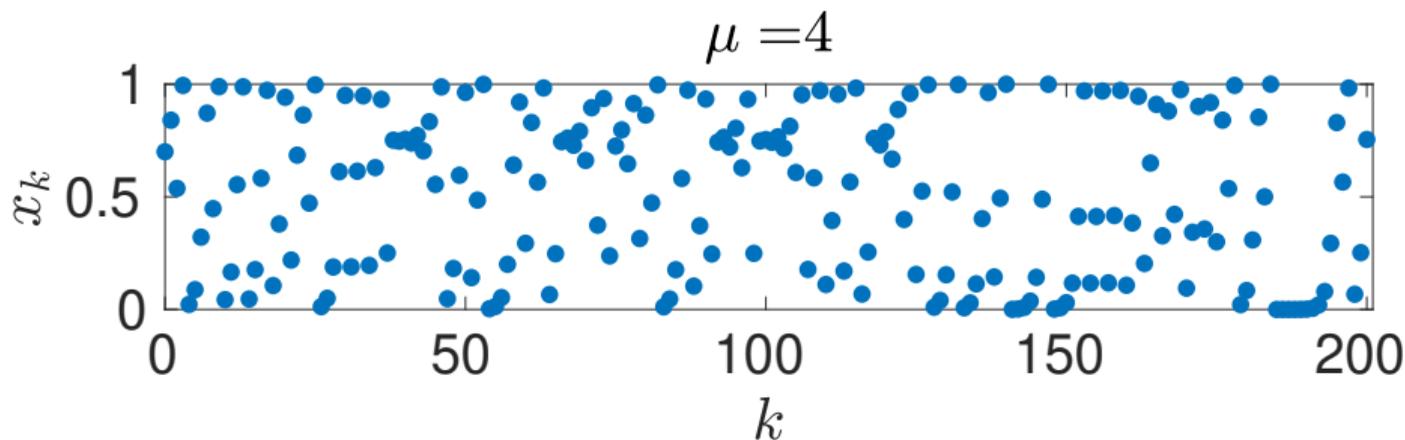


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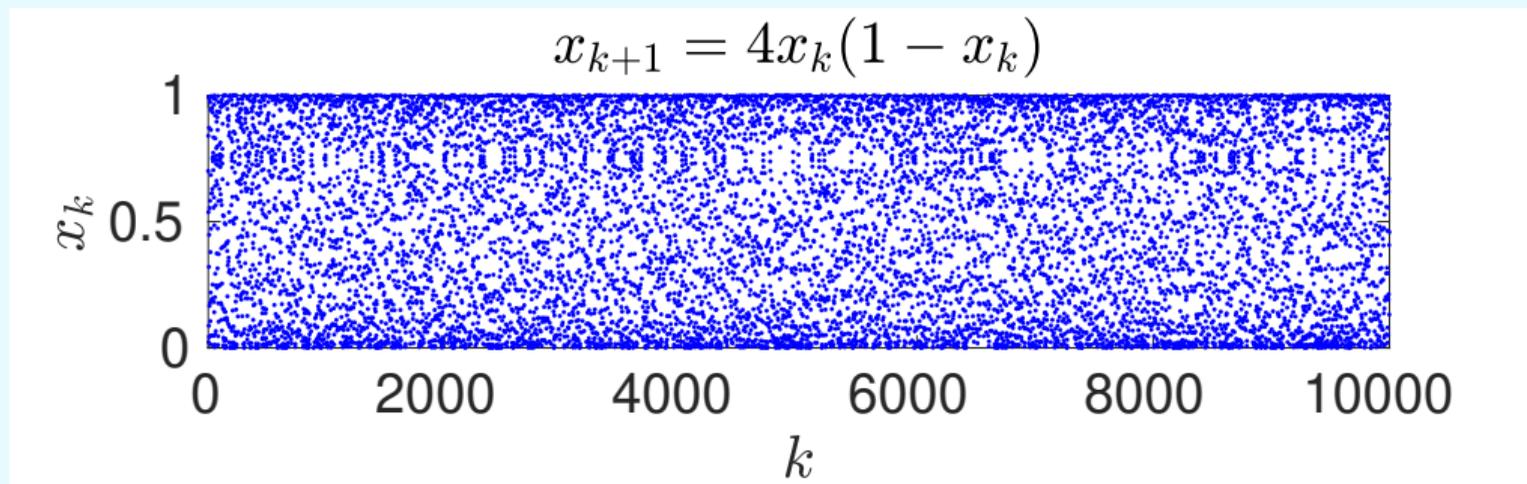


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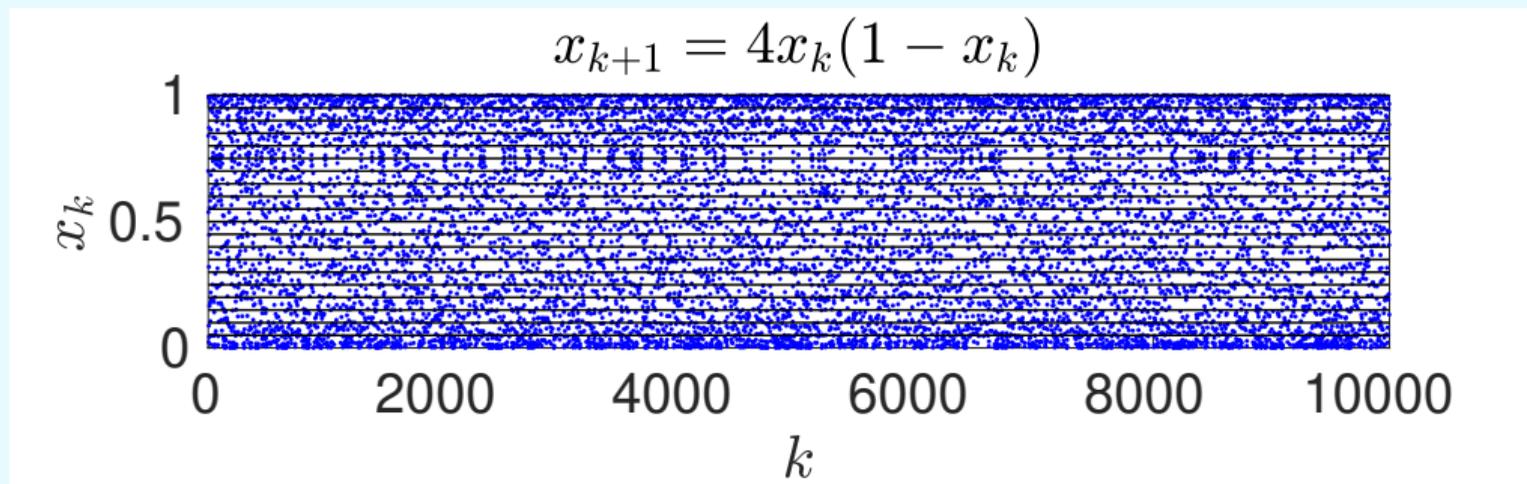


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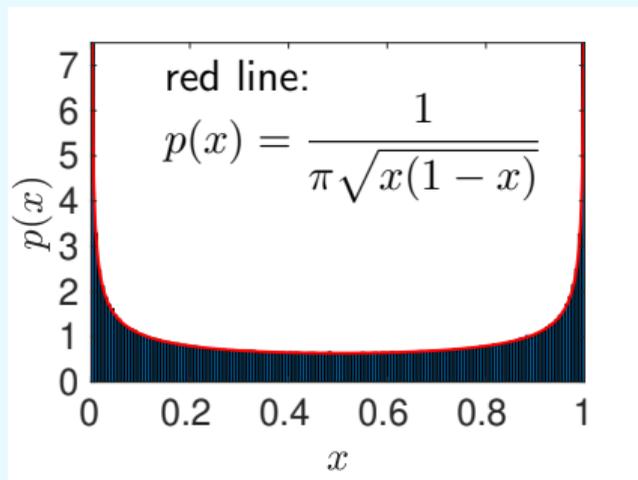


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Histogram of values x_k , for $k = 0, 1, 2, \dots, 10^6$ (blue bars): $x_{k+1} = 4 x_k (1 - x_k)$



Problem Sheet 0 Question 4:

Let X_k be a continuous random variable on interval $[0, 1]$ with the probability density function $p(x)$. Then the random variable $X_{k+1} = F(X_k) = 4 X_k (1 - X_k)$ has the same probability density function $p(x)$.

[Prelims Probability and Calculus]

Prelims Probability and Calculus: Problem Sheet 0 Question 4

Let X be a continuous random variable on interval $[0, 1]$ with the probability density function $p : [0, 1] \rightarrow [0, \infty)$ given by $p(x) = 1/(\pi\sqrt{x(1-x)})$. Let $F : [0, 1] \rightarrow [0, 1]$ be defined by $F(x) = 4x(1-x)$. Then the cumulative distribution function of $F(X)$ is

$$\begin{aligned}\mathbb{P}(F(X) < x) &= \mathbb{P}\left(X < \frac{1}{2}(1 - \sqrt{1-x})\right) + \mathbb{P}\left(X > \frac{1}{2}(1 + \sqrt{1-x})\right) \\ &= \int_0^{\frac{1}{2}(1-\sqrt{1-x})} p(z) \, dz + \int_{\frac{1}{2}(1+\sqrt{1-x})}^1 p(z) \, dz\end{aligned}$$

Prelims Probability and Calculus: Problem Sheet 0 Question 4

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Prelims Probability and Calculus: Problem Sheet 0 Question 4

Let X be a continuous random variable on interval $[0, 1]$ with the probability density function $p : [0, 1] \rightarrow [0, \infty)$ given by $p(x) = 1/(\pi\sqrt{x(1-x)})$. Let $F : [0, 1] \rightarrow [0, 1]$ be defined by $F(x) = 4x(1-x)$. Then the cumulative distribution function of $F(X)$ is

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Consequently, the probability density function of $F(X)$ is:

$$\frac{d}{dx} \mathbb{P}(F(X) < x) = -\frac{2}{\pi} \frac{d}{dx} \sin^{-1}(\sqrt{1-x}) = \frac{1}{\pi\sqrt{x(1-x)}} = p(x)$$

B5.6 covers nonlinear dynamics (linear systems were in Prelims/Part A)

Continuous-time dynamical system: Let $\mathbf{f} : \Omega \times \Theta \rightarrow \mathbb{R}^n$, where $\Omega \subset \mathbb{R}^n$ and $\Theta \subset \mathbb{R}^m$.

Let $\mathbf{x}_0 \in \Omega$, $\boldsymbol{\mu} \in \Theta$ and $\mathbf{x}(t) \in \Omega$ be a solution of the ODE

$$\frac{d\mathbf{x}}{dt} = \mathbf{f}(\mathbf{x}; \boldsymbol{\mu}) \quad \text{with the initial condition} \quad \mathbf{x}(0) = \mathbf{x}_0$$

We want to find \mathbf{x} as a function of t and sketch the phase plane or phase space.

What is the behaviour of $\mathbf{x}(t)$ as $t \rightarrow \infty$?

How do our answers depend on the initial value \mathbf{x}_0 ?

How does the behaviour of $\mathbf{x}(t)$ depend on parameters $\boldsymbol{\mu}$?

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Linear example (Question 1(b) on Problem Sheet 0):

$$\frac{d\mathbf{x}}{dt} = M\mathbf{x} \quad \text{for} \quad M = \begin{pmatrix} 2 & 1 & -1 \\ 1 & -1 & 2 \\ -1 & 1 & 2 \end{pmatrix} \quad \text{and} \quad \mathbf{x}(0) = \begin{pmatrix} 8 \\ 1 \\ 3 \end{pmatrix}$$

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Closed form solution formula [Prelims Calculus and Part A Differential Equations courses]:

$$\mathbf{x}(t) = e^{3t} \begin{pmatrix} 2 \\ -1 \\ -3 \end{pmatrix} + e^{-2t} \begin{pmatrix} 1 \\ -3 \\ 1 \end{pmatrix} + e^{2t} \begin{pmatrix} 5 \\ 5 \\ 5 \end{pmatrix}$$

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Linear example (Question 2(b) on Problem Sheet 0): $\frac{d\mathbf{x}}{dt} = M\mathbf{x}$ for $M = \begin{pmatrix} 1 & 2 \\ -1 & \mu \end{pmatrix}$

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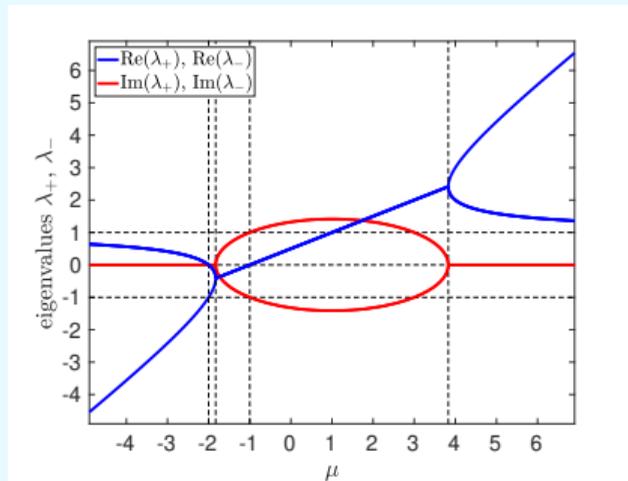
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eigenvalues of M are

$$\lambda_{\pm} = \frac{1 + \mu \pm \sqrt{(\mu - 1 + 2\sqrt{2})(\mu - 1 - 2\sqrt{2})}}{2}$$

$[0, 0]$ is the only critical point



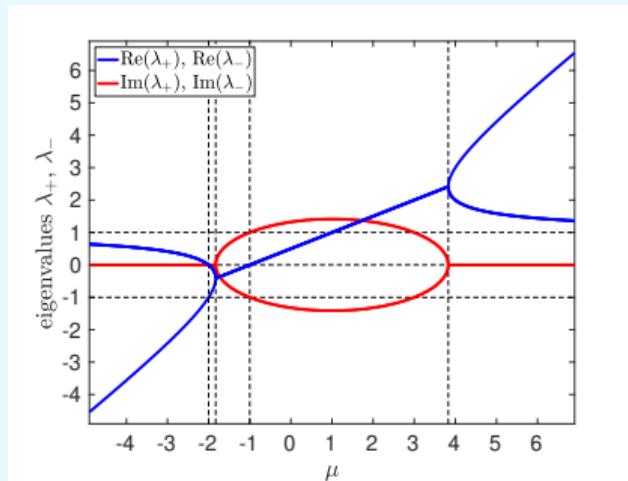
[Part A Differential Equations 1]

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$[0, 0]$ is the only critical point which is

saddle for $\mu < -2$

stable node for $-2 < \mu < 1 - 2\sqrt{2}$

stable spiral for $1 - 2\sqrt{2} < \mu < -1$

unstable spiral for $-1 < \mu < 1 + 2\sqrt{2}$

unstable node for $\mu > 1 + 2\sqrt{2}$

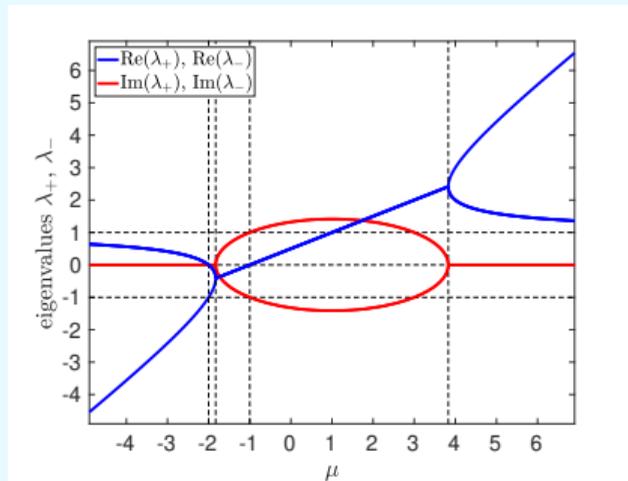
[Part A Differential Equations 1]

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Linear example (Question 2(b) on Problem Sheet 0): $\frac{d\mathbf{x}}{dt} = M\mathbf{x}$ for $M = \begin{pmatrix} 1 & 2 \\ -1 & \mu \end{pmatrix}$



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$$\lambda_{\pm} = \frac{1 + \mu \pm \sqrt{(\mu - 1 + 2\sqrt{2})(\mu - 1 - 2\sqrt{2})}}{2}$$

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unstable spiral for $-1 < \mu < 1 + 2\sqrt{2}$

unstable node for $\mu > 1 + 2\sqrt{2}$

center for $\mu = -1$, stable/unstable inflected node for $\mu = 1 \pm 2\sqrt{2}$

[Part A Differential Equations 1]

Nonlinear example: Problem Sheet 0 Question 5

Let $\mu \in \mathbb{R}$ be a parameter. Consider a planar autonomous ODE system given by:

$$\begin{aligned}\frac{dx}{dt} &= x - \mu y + y^2(1 - x) - x^3 \\ \frac{dy}{dt} &= \mu x - xy(1 + x) + y - y^3\end{aligned}$$

Nonlinear example: Problem Sheet 0 Question 5

Let $\mu \in (-1, 1)$ be a parameter. Consider a planar autonomous ODE system given by:

$$\begin{aligned}\frac{dx}{dt} &= x - \mu y + y^2(1 - x) - x^3 \\ \frac{dy}{dt} &= \mu x - xy(1 + x) + y - y^3\end{aligned}$$

Part A Differential Equations 1: linearized system next to the critical point $[x_c, y_c]$

$$\frac{d}{dt} \begin{pmatrix} x - x_c \\ y - y_c \end{pmatrix} = M \begin{pmatrix} x - x_c \\ y - y_c \end{pmatrix} \quad \text{where} \quad M = \begin{pmatrix} 1 - y_c^2 - 3x_c^2 & -\mu + 2y_c(1 - x_c) \\ \mu - y_c - 2x_c y_c & -x_c(1 + x_c) + 1 - 3y_c^2 \end{pmatrix}$$

$$[0, 0]: \text{unstable spiral} \quad M = \begin{pmatrix} 1 & -\mu \\ \mu & 1 \end{pmatrix} \quad \text{eigenvalues: } 1 \pm \mu i$$

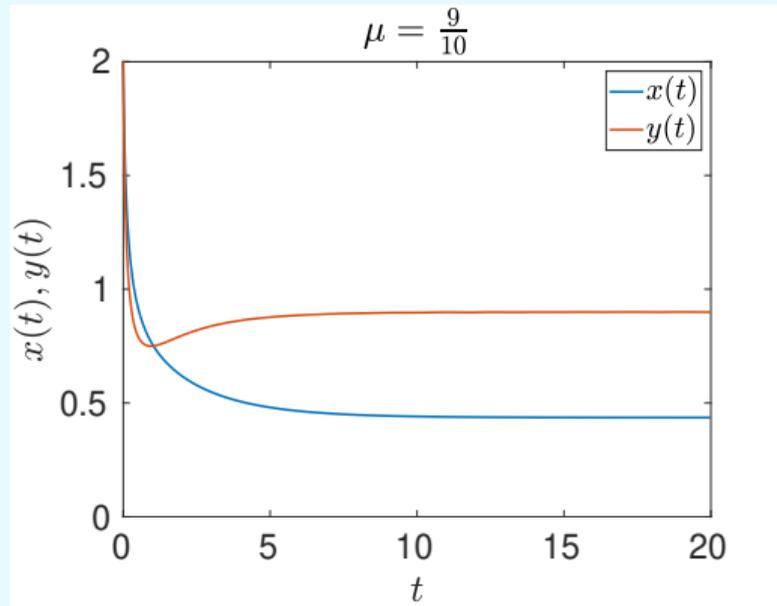
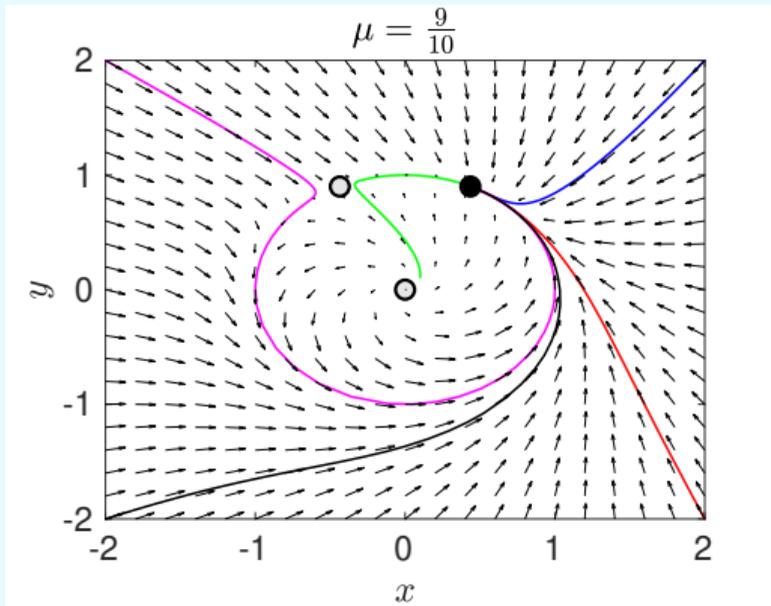
$$[\sqrt{1 - \mu^2}, \mu]: \text{stable node} \quad M = \begin{pmatrix} -2 + 2\mu^2 & \mu - 2\mu\sqrt{1 - \mu^2} \\ -2\mu\sqrt{1 - \mu^2} & -2\mu^2 - \sqrt{1 - \mu^2} \end{pmatrix} \quad \text{eigenvalues: } -2, -\sqrt{1 - \mu^2}$$

$$[-\sqrt{1 - \mu^2}, \mu]: \text{saddle} \quad M = \begin{pmatrix} -2 + 2\mu^2 & \mu + 2\mu\sqrt{1 - \mu^2} \\ 2\mu\sqrt{1 - \mu^2} & -2\mu^2 + \sqrt{1 - \mu^2} \end{pmatrix} \quad \text{eigenvalues: } -2, \sqrt{1 - \mu^2}$$

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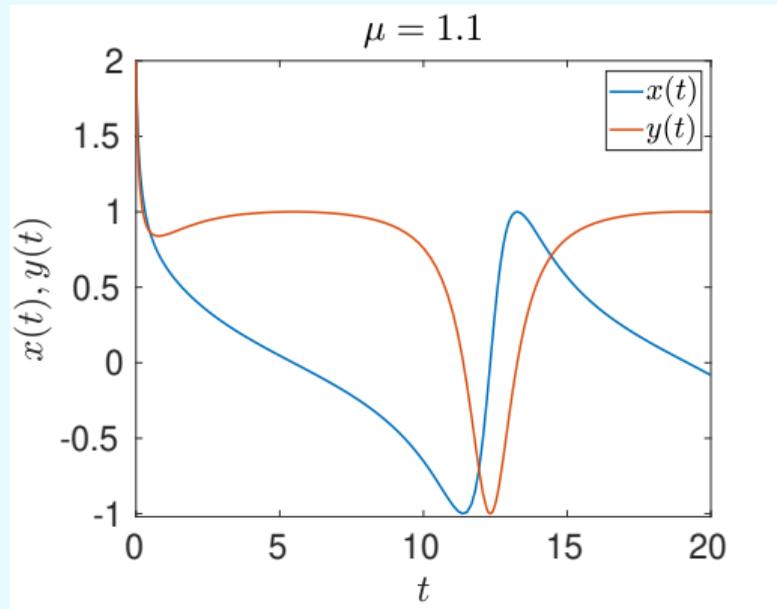
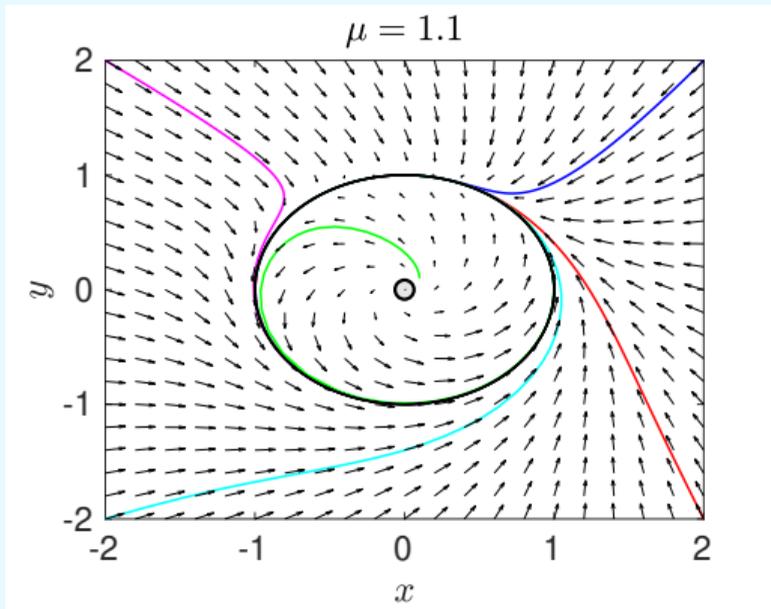
$$\begin{aligned}\frac{dx}{dt} &= x - \mu y + y^2(1 - x) - x^3 \\ \frac{dy}{dt} &= \mu x - xy(1 + x) + y - y^3\end{aligned}$$



Nonlinear example: Problem Sheet 0 Question 5

Let $\mu > 1$ be a parameter. Consider a planar autonomous ODE system given by:

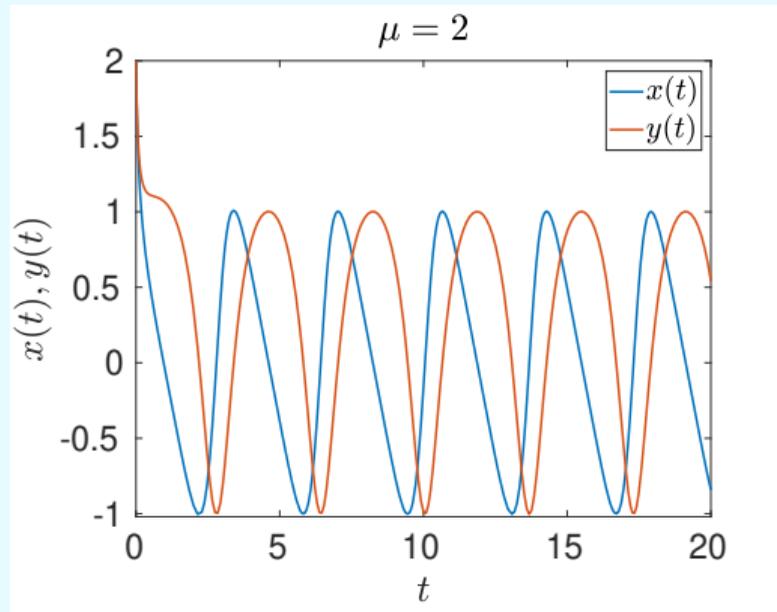
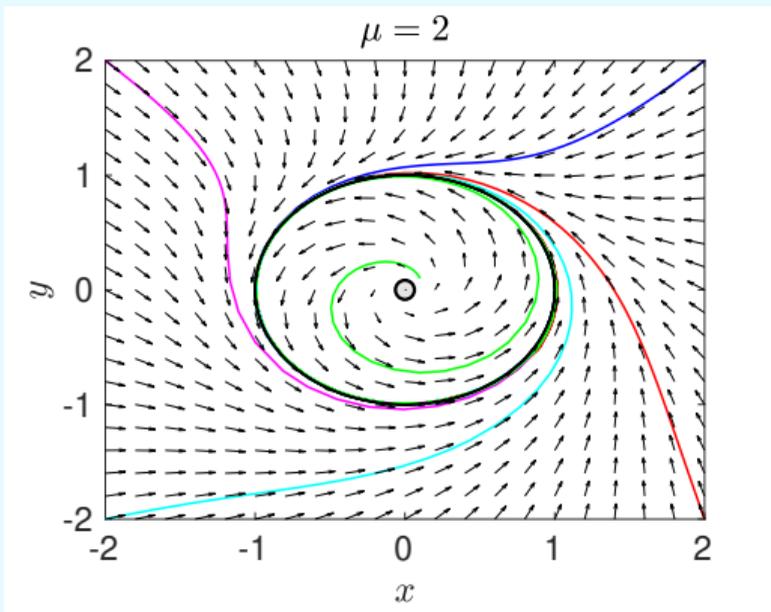
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Nonlinear example: Problem Sheet 0 Question 5

Let $\mu \in \mathbb{R}$ be a parameter. Consider a planar autonomous ODE system given by:

$$\begin{aligned}\frac{dx}{dt} &= x - \mu y + y^2(1 - x) - x^3 \\ \frac{dy}{dt} &= \mu x - xy(1 + x) + y - y^3\end{aligned}$$

Prelims Calculus: We transform the ODEs to polar coordinates by using variables $r(t)$ and $\theta(t)$, where $x(t) = r(t) \cos \theta(t)$ and $y(t) = r(t) \sin \theta(t)$. We obtain

$$\frac{dr}{dt} = r(1 - r^2)$$

We conclude that $r(t) \rightarrow 1$ as $t \rightarrow \infty$ for any initial condition satisfying $r(0) > 0$.

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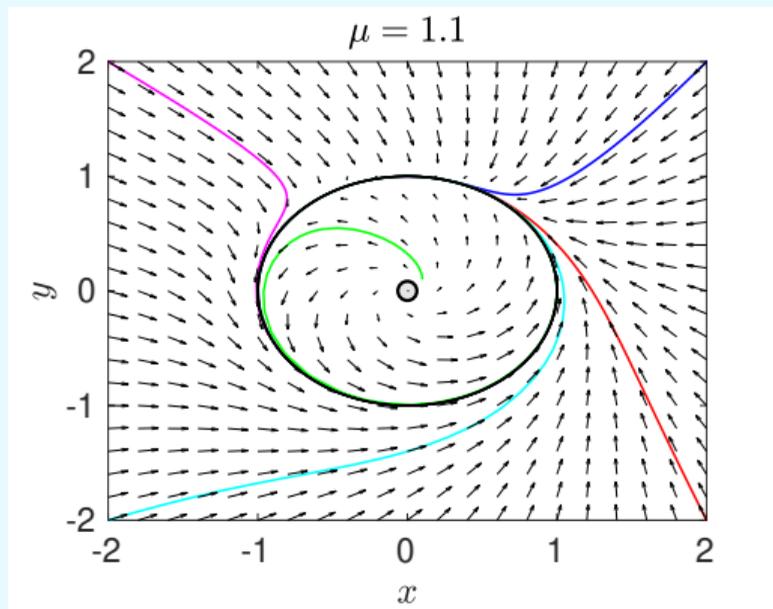
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$$\frac{d\theta}{dt} = \mu - y = \mu - r \sin(\theta)$$

If $\mu > 1$, then $d\theta/dt > \mu - 1 > 0$.



Nonlinear example: Problem Sheet 0 Question 5

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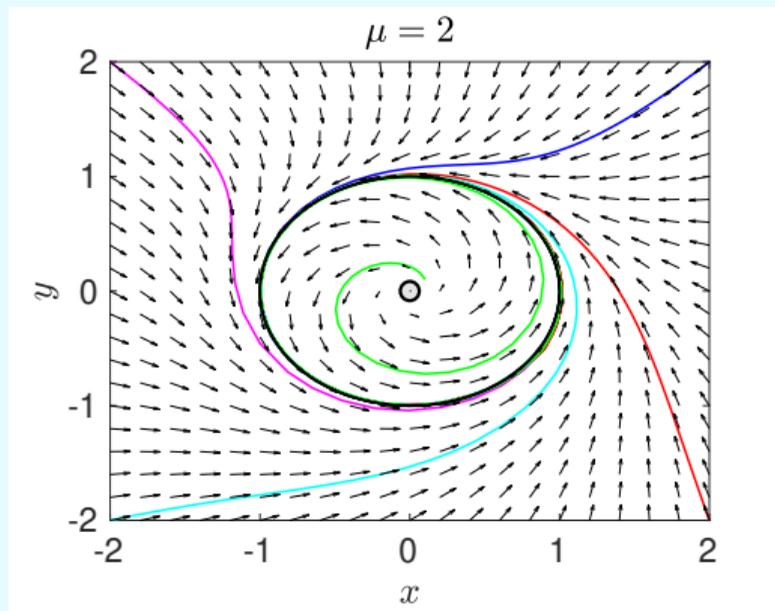
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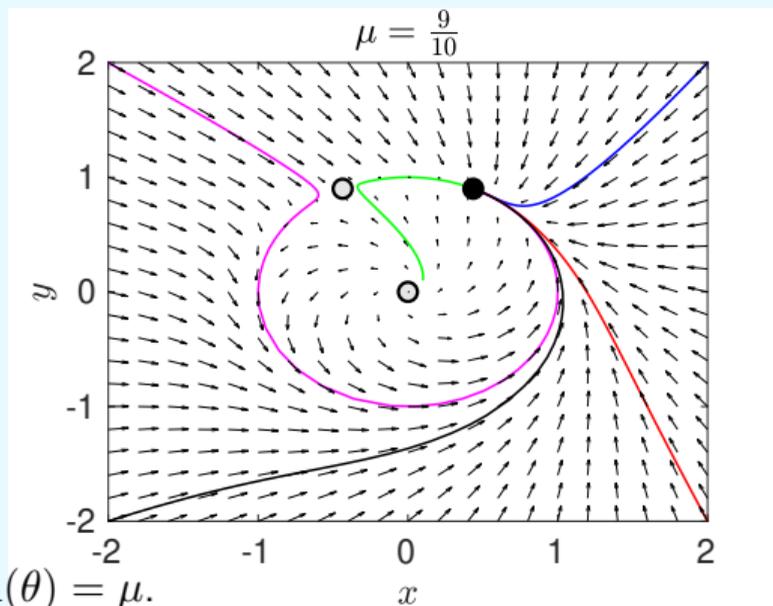
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$$\frac{d\theta}{dt} = \mu - y = \mu - r \sin(\theta)$$

If $|\mu| < 1$, then $d\theta/dt = 0$ for $r = 1$ and $\sin(\theta) = \mu$.



ODEs and Chaos

Continuous-time dynamical system: Let $\mathbf{f} : \Omega \times \Theta \rightarrow \mathbb{R}^n$, where $\Omega \subset \mathbb{R}^n$ and $\Theta \subset \mathbb{R}^m$.

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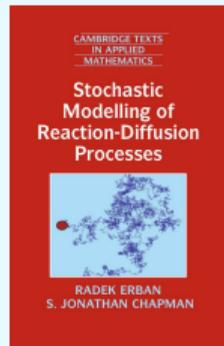
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- In course B5.6, we will focus on chaotic solutions of ODEs for $n = 3$, but chaos is common for $n \gg 3$. Examples are discussed in course B5.1 Stochastic Modelling of Biological Processes.
[video of molecular dynamics simulation of ions in water]

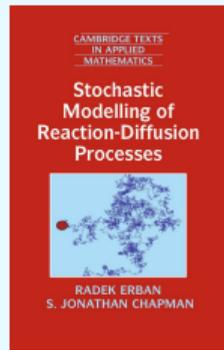


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[video of molecular dynamics simulation of ions in water]
- B5.6: we will consider relatively simple ODEs (small n , polynomials):
(1) good for developing general theory; (2) there are also interesting applications



Chemical reaction networks

Consider a well-stirred (well-mixed) chemical system with n chemical species X_1, X_2, \dots, X_n which are subject to ℓ chemical reactions.

Let $x_i(t)$ be the concentration of the chemical species X_i , $i = 1, 2, \dots, n$.

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The time evolution of concentration $x_1(t)$ is given by the ODE
$$\frac{dx_1}{dt} = \sum_{j=1}^{\ell} c_j r_j,$$

where r_j is the rate of the j th reaction and c_j is the change in the number of molecules of X_1 corresponding to the occurrence of one j -th reaction, i.e. it is the difference between the number (stoichiometric coefficient) in front of X_1 on the right hand side of the reaction and the corresponding stoichiometric coefficient on the left hand side. The rate $r_j \equiv r_j(t)$ is computed as a product of the rate constant and the concentrations of the reactants (mass action kinetics).

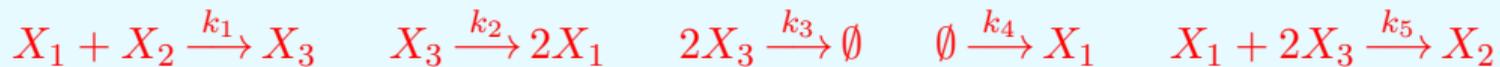
Chemical reaction networks

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Example: system of $n = 3$ chemical species which are subject to $\ell = 5$ reactions:



$$c_1 = -1$$

$$c_2 = 2$$

$$c_3 = 0$$

$$c_4 = 1$$

$$c_5 = -1$$

$$r_1 = k_1 x_1 x_2$$

$$r_2 = k_2 x_3$$

$$r_3 = k_3 x_3^2$$

$$r_4 = k_4$$

$$r_5 = k_5 x_1 x_3^2$$

$$\frac{dx_1}{dt} = -k_1 x_1 x_2 + 2k_2 x_3 + k_4 - k_5 x_1 x_3^2$$

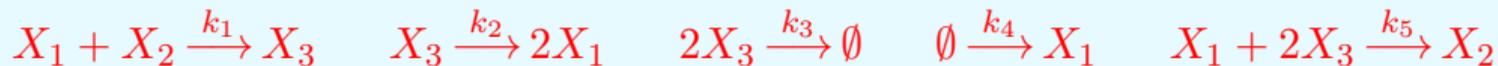
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$$r_1 = k_1 x_1 x_2 \quad r_2 = k_2 x_3 \quad r_3 = k_3 x_3^2 \quad r_4 = k_4 \quad r_5 = k_5 x_1 x_3^2$$

$$\frac{dx_1}{dt} = -k_1 x_1 x_2 + 2k_2 x_3 + k_4 - k_5 x_1 x_3^2$$

The units of $x_i(t)$ are usually moles (or number of molecules) per unit of volume, k_1 and k_3 have units of $[\text{m}^3 \text{sec}^{-1}]$, k_2 is in $[\text{sec}^{-1}]$, k_4 is in $[\text{m}^{-3} \text{sec}^{-1}]$ and k_5 is in $[\text{m}^6 \text{sec}^{-1}]$, but we will assume that $x_i(t)$ and all parameters are dimensionless.

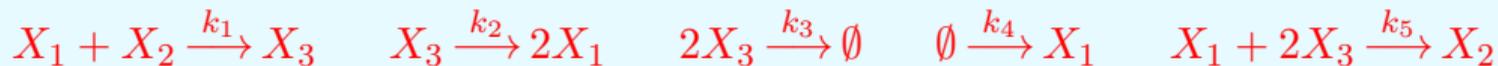
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The time evolution of concentration $x_2(t)$ is given by the ODE $\frac{dx_2}{dt} = \sum_{j=1}^{\ell} c_j r_j$,

Example: system of $n = 3$ chemical species which are subject to $\ell = 5$ reactions:



$$c_1 = -1 \quad c_2 = 0 \quad c_3 = 0 \quad c_4 = 0 \quad c_5 = 1$$

$$r_1 = k_1 x_1 x_2 \quad r_2 = k_2 x_3 \quad r_3 = k_3 x_3^2 \quad r_4 = k_4 \quad r_5 = k_5 x_1 x_3^2$$

$$\frac{dx_1}{dt} = -k_1 x_1 x_2 + 2k_2 x_3 + k_4 - k_5 x_1 x_3^2$$

similarly for $x_2(t)$: $\frac{dx_2}{dt} = -k_1 x_1 x_2 + k_5 x_1 x_3^2$

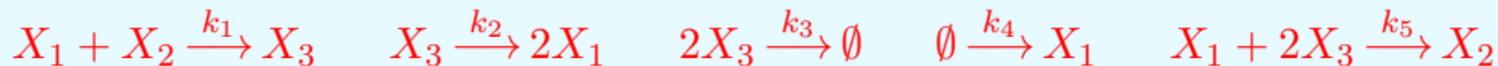
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Example: system of $n = 3$ chemical species which are subject to $\ell = 5$ reactions:



$$c_1 = 1 \quad c_2 = -1 \quad c_3 = -2 \quad c_4 = 0 \quad c_5 = -2$$

$$r_1 = k_1 x_1 x_2 \quad r_2 = k_2 x_3 \quad r_3 = k_3 x_3^2 \quad r_4 = k_4 \quad r_5 = k_5 x_1 x_3^2$$

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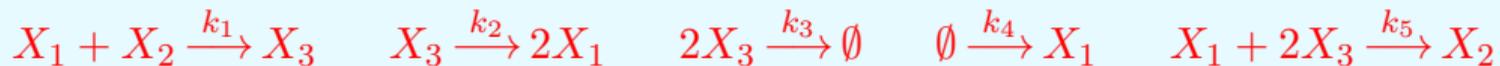
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other examples: Questions 3, 4 and 6 on Problem Sheet 1

Course B5.6: ODEs with relatively small n and simple right hand sides (often polynomials). They appear in applications as (i) models of (bio)chemical systems; or (ii) they can also be constructed in experiments (synthetic biology, DNA computing).

Polynomials can also approximate more complicated right hand sides of ODEs (stable manifold, center manifold, bifurcations). Let us go back to some theory.

B5.6: Nonlinear Dynamics, Bifurcations and Chaos (Lecture 2)

- summary of Lecture 1: we discussed
Discrete-time (maps) and continuous-time (differential equations) dynamical systems.
Chemical reaction networks. (Questions 3, 4 and 6 on Problem Sheet 1)
- today: we will continue in our discussion of Problem Sheet 1

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- course synopsis of **Lectures 1-8** (taken by both Part B and MSc students): Discrete-time (maps) and continuous-time (differential equations) dynamical systems. Notion of flows, stability of fixed points, Lyapunov function, invariant manifolds, stable manifold theorem, notion of hyperbolicity, center manifold. Chemical reaction networks. Stable, unstable and center subspaces. Poincaré-Bendixson theorem. Periodic solutions, stable and unstable limit cycles. Introduction to bifurcation theory, covering saddle-node, transcritical, supercritical pitchfork and subcritical pitchfork bifurcations. Extended center manifold. Logistic map. Periodic points of maps. Stability of N-cycles. Period-doubling bifurcation. Sharkovsky's theorem. Invariant distribution.

The flow defined by an ODE

Continuous-time dynamical system: Let $\mathbf{f} : \Omega \times \Theta \rightarrow \mathbb{R}^n$, where $\Omega \subset \mathbb{R}^n$ and $\Theta \subset \mathbb{R}^m$.

Let $\mathbf{x}_0 \in \Omega$, $\boldsymbol{\mu} \in \Theta$ and $\mathbf{x}(t) \in \Omega$ be a solution of the ODE

$$\frac{d\mathbf{x}}{dt} = \mathbf{f}(\mathbf{x}; \boldsymbol{\mu}) \quad \text{with the initial condition} \quad \mathbf{x}(0) = \mathbf{x}_0$$

Then we define the *flow* $\phi_t : \Omega \rightarrow \Omega$ by

$$\phi_t(\mathbf{x}_0) = \phi(t, \mathbf{x}_0) = \mathbf{x}(t)$$

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Example: Question 1(b) on Problem Sheet 0 for general initial condition $\mathbf{x}_0 \in \mathbb{R}^3$:

$$\frac{d\mathbf{x}}{dt} = M\mathbf{x} \quad \text{for} \quad M = \begin{pmatrix} 2 & 1 & -1 \\ 1 & -1 & 2 \\ -1 & 1 & 2 \end{pmatrix}$$

Then

$$\phi_t(\mathbf{x}_0) = \phi(t, \mathbf{x}_0) = \exp[Mt] \mathbf{x}_0 = \left(\sum_{j=0}^{\infty} \frac{M^j t^j}{j!} \right) \mathbf{x}_0$$

where we have used the definition of the matrix exponential: $\exp[A] = \sum_{j=0}^{\infty} \frac{A^j}{j!}$

The flow defined by an ODE

Considering the linear system of ODEs given by $\frac{d\mathbf{x}}{dt} = M\mathbf{x}$ where matrix $M \in \mathbb{R}^{n \times n}$, the flow $\phi_t : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is given by $\phi_t = \exp[Mt]$.

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In particular, the properties of the matrix exponential imply that the flow ϕ_t satisfies

$$(a) \phi_0 = I$$

$$(b) \phi_s \circ \phi_t = \phi_{s+t}$$

$$(c) \phi_t \circ \phi_{-t} = \phi_{-t} \circ \phi_t = I$$

where $I \in \mathbb{R}^{n \times n}$ is the identity matrix.

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For linear systems, the properties (a)–(c) mean:

$$(a) \phi_0(\mathbf{x}) = \mathbf{x} \text{ for all } \mathbf{x} \in \mathbb{R}^n;$$

$$(b) \phi_s(\phi_t(\mathbf{x})) = \phi_{s+t}(\mathbf{x}) \text{ for all } s, t \in \mathbb{R} \text{ and } \mathbf{x} \in \mathbb{R}^n;$$

$$(c) \phi_t(\phi_{-t}(\mathbf{x})) = \phi_{-t}(\phi_t(\mathbf{x})) = \mathbf{x} \text{ for all } t \in \mathbb{R} \text{ and } \mathbf{x} \in \mathbb{R}^n.$$

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Considering the linear system of ODEs given by $\frac{d\mathbf{x}}{dt} = M\mathbf{x}$ where matrix $M \in \mathbb{R}^{n \times n}$, the flow $\phi_t : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is given by $\phi_t = \exp[Mt]$.

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- (c) $\phi_t(\phi_{-t}(\mathbf{x})) = \phi_{-t}(\phi_t(\mathbf{x})) = \mathbf{x}$ for all $t \in \mathbb{R}$ and $\mathbf{x} \in \mathbb{R}^n$.

Nonlinear ODE system: $\frac{d\mathbf{x}}{dt} = \mathbf{f}(\mathbf{x}; \boldsymbol{\mu})$

Part A Differential Equations 1: Picard's existence theorem implies the global existence and uniqueness of solutions for $\mathbf{f} \in C^1(\mathbb{R}^n \times \mathbb{R}^m)$ which satisfies the global Lipschitz condition $|\mathbf{f}(\mathbf{x}; \boldsymbol{\mu}) - \mathbf{f}(\mathbf{y}; \boldsymbol{\mu})| \leq C|\mathbf{x} - \mathbf{y}|$ for all $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$, $\boldsymbol{\mu} \in \mathbb{R}^m$.

Then $\phi_t : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is defined for all $t \in \mathbb{R}$ and ϕ_t satisfies the properties (a)–(c).

The flow defined by an ODE

Considering the linear system of ODEs given by $\frac{d\mathbf{x}}{dt} = M\mathbf{x}$ where matrix $M \in \mathbb{R}^{n \times n}$, the flow $\phi_t : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is given by $\phi_t = \exp[Mt]$.

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where $I \in \mathbb{R}^{n \times n}$ is the identity matrix.

Note: Assuming the global Lipschitz condition could exclude some interesting ODEs. Our assumptions on $\Omega \subset \mathbb{R}^n$, $\Theta \subset \mathbb{R}^m$ and $\mathbf{f} : \Omega \times \Theta \rightarrow \mathbb{R}^n$ could be relaxed. In some cases, we would only get the local existence of solutions to the nonlinear ODE system

$$\frac{d\mathbf{x}}{dt} = \mathbf{f}(\mathbf{x}; \boldsymbol{\mu})$$

ϕ_t would not be defined for all $t \in \mathbb{R}$ and ϕ_t would only satisfy properties (a)–(c) where it is defined.

Let us illustrate this with an example with $n = 1$.

The flow defined by an ODE: nonlinear example

Consider the ODE $\frac{dx}{dt} = x^2$ (it does not satisfy the global Lipschitz condition).

The flow defined by an ODE: nonlinear example

Consider the ODE $\frac{dx}{dt} = x^2$ (it does not satisfy the global Lipschitz condition).

Given the initial condition $x(0) = x_0 \in \mathbb{R}$, we can solve this ODE to obtain

$$x(t) = \frac{x_0}{1 - t x_0} \quad \text{for } t \in I(x_0),$$

where $I(x_0)$ is the *maximal interval of existence* given by $I(0) = \mathbb{R}$,

$$I(x) = \left(-\infty, \frac{1}{x}\right) \quad \text{for } x > 0, \quad \text{and} \quad I(x) = \left(\frac{1}{x}, \infty\right) \quad \text{for } x < 0.$$

The flow defined by an ODE: nonlinear example

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In particular, the flow ϕ_t is defined as the mapping $\phi : Q \rightarrow \mathbb{R}$, where

$$Q = \{(t, x) \mid x \in \mathbb{R} \text{ and } t \in I(x)\} \quad \text{and} \quad \phi_t(x) = \phi(t, x) = \frac{x}{1 - tx}.$$

Problem Sheet 1 Question 7: We can rescale time to get a topologically equivalent ODE system which has $I(x) = \mathbb{R}$.

In general, the time along trajectories can be rescaled without affecting the phase portrait. In what follows, we will assume that ϕ_t is defined for all $t \in \mathbb{R}$ and $\phi \in C^1(\mathbb{R} \times \Omega)$ for any considered parameter values $\mu \in \Theta$.

Equilibrium points, flow, trajectory - summary

Given $\mathbf{f} : \Omega \times \Theta \rightarrow \mathbb{R}^n$, where $\Omega \subset \mathbb{R}^n$, $\Theta \subset \mathbb{R}^m$, and $\boldsymbol{\mu} \in \Theta$, we consider ODE system

$$\frac{d\mathbf{x}}{dt} = \mathbf{f}(\mathbf{x}; \boldsymbol{\mu})$$

- \mathbf{x}_c is an *equilibrium point* or *critical point* or *fixed point* if $\mathbf{f}(\mathbf{x}_c; \boldsymbol{\mu}) = 0$
- the *flow* of the ODE is the map $\phi_t : \Omega \rightarrow \Omega$ such that $\phi_t(\mathbf{x}_0) = \phi(t, \mathbf{x}_0) = \mathbf{x}(t; \mathbf{x}_0)$, where $\mathbf{x}(t; \mathbf{x}_0) \in \Omega$ is the solution with the initial condition $\mathbf{x}(0) = \mathbf{x}_0 \in \Omega$
- an *orbit* or *trajectory* based on \mathbf{x}_0 is the curve $\Gamma_{\mathbf{x}_0} \subset \Omega$ defined by

$$\Gamma_{\mathbf{x}_0} = \{ \mathbf{x}(t; \mathbf{x}_0) \mid t \in I(\mathbf{x}_0) \} ,$$

where $I(\mathbf{x}_0)$ is the maximum interval of existence (WLOG we assume $I(\mathbf{x}_0) = \mathbb{R}$)

- $S \subset \Omega$ is an *invariant set* if $\phi_t(S) \subset S$ for all $t \in \mathbb{R}$

Equilibrium points: stability

Given $\mathbf{f} : \Omega \times \Theta \rightarrow \mathbb{R}^n$, where $\Omega \subset \mathbb{R}^n$, $\Theta \subset \mathbb{R}^m$, and $\boldsymbol{\mu} \in \Theta$, we consider ODE system

$$\frac{d\mathbf{x}}{dt} = \mathbf{f}(\mathbf{x}; \boldsymbol{\mu})$$

- \mathbf{x}_c is an *equilibrium point* or *critical point* or *fixed point* if $\mathbf{f}(\mathbf{x}_c; \boldsymbol{\mu}) = 0$
- \mathbf{x}_c is *stable* if
 $\forall \varepsilon > 0 \exists \delta > 0$ such that $\forall \mathbf{x}_0 \in B_\delta(\mathbf{x}_c)$ and $t \geq 0$ we have $\phi_t(\mathbf{x}) \in B_\varepsilon(\mathbf{x}_c)$
where the open ball of radius r is defined by $B_r(\mathbf{x}_c) = \{\mathbf{x} \in \mathbb{R}^n \mid \|\mathbf{x} - \mathbf{x}_c\| < r\}$
- \mathbf{x}_c is *asymptotically stable* if (i) it is stable; and
(ii) $\exists \delta > 0$ such that $\phi_t(\mathbf{x}_0) \rightarrow \mathbf{x}_c$ for all $\mathbf{x}_0 \in B_\delta(\mathbf{x}_c)$

Equilibrium points: stability, Lyapunov function

Given $\mathbf{f} : \Omega \times \Theta \rightarrow \mathbb{R}^n$, where $\Omega \subset \mathbb{R}^n$, $\Theta \subset \mathbb{R}^m$, and $\boldsymbol{\mu} \in \Theta$, we consider ODE system

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- \mathbf{x}_c is *asymptotically stable* if (i) it is stable; and
(ii) $\exists \delta > 0$ such that $\phi_t(\mathbf{x}_0) \rightarrow \mathbf{x}_c$ for all $\mathbf{x}_0 \in B_\delta(\mathbf{x}_c)$
- Lyapunov function: $V \in C^1(A)$, where $A \subset \Omega \subset \mathbb{R}^n$ is open and $\mathbf{x}_c \in A$
 $V(\mathbf{x}) > 0$ for $\mathbf{x} \neq \mathbf{x}_c$ and $V(\mathbf{x}_c) = 0$
if $dV/dt \leq 0$ for all $\mathbf{x} \in A \setminus \{\mathbf{x}_c\}$, then \mathbf{x}_c is stable
if $dV/dt < 0$ for all $\mathbf{x} \in A \setminus \{\mathbf{x}_c\}$, then \mathbf{x}_c is asymptotically stable

Problem Sheet 1 Question 5: proving stability by finding a suitable Lyapunov function

Equilibrium points: linearization

Given $\mathbf{f} : \Omega \times \Theta \rightarrow \mathbb{R}^n$, where $\Omega \subset \mathbb{R}^n$, $\Theta \subset \mathbb{R}^m$, and $\boldsymbol{\mu} \in \Theta$, we consider ODE system

$$\frac{d\mathbf{x}}{dt} = \mathbf{f}(\mathbf{x}; \boldsymbol{\mu})$$

- \mathbf{x}_c is an *equilibrium point* or *critical point* or *fixed point* if $\mathbf{f}(\mathbf{x}_c; \boldsymbol{\mu}) = 0$

- linearization at \mathbf{x}_c is given by

$$\frac{d\mathbf{x}}{dt} = M\mathbf{x}$$

$$M = D\mathbf{f}(\mathbf{x}_c) = \begin{pmatrix} \frac{\partial f_1}{\partial x_1}(\mathbf{x}_c) & \frac{\partial f_1}{\partial x_2}(\mathbf{x}_c) & \cdots & \frac{\partial f_1}{\partial x_n}(\mathbf{x}_c) \\ \frac{\partial f_2}{\partial x_1}(\mathbf{x}_c) & \frac{\partial f_2}{\partial x_2}(\mathbf{x}_c) & \cdots & \frac{\partial f_2}{\partial x_n}(\mathbf{x}_c) \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial f_n}{\partial x_1}(\mathbf{x}_c) & \frac{\partial f_n}{\partial x_2}(\mathbf{x}_c) & \cdots & \frac{\partial f_n}{\partial x_n}(\mathbf{x}_c) \end{pmatrix}$$

where M is the Jacobian matrix

- equilibrium point \mathbf{x}_c is called

hyperbolic: if none of the eigenvalues of the matrix $D\mathbf{f}(\mathbf{x}_c)$ have zero real part

sink: if all of the eigenvalues of the matrix $D\mathbf{f}(\mathbf{x}_c)$ have negative real part

source: if all of the eigenvalues of the matrix $D\mathbf{f}(\mathbf{x}_c)$ have positive real part

saddle: if it is a hyperbolic equilibrium point and $D\mathbf{f}(\mathbf{x}_c)$ has at least one eigenvalue with a positive real part and at least one with a negative real part

Invariant manifolds

stable manifold theorem:

- the nonlinear system has locally similar behaviour close to a hyperbolic critical point
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What is a manifold?

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What is a manifold? Wiggins [page 29], Perko [page 107], Kuznetsov [page 598]

- linear settings: a linear vector subspace of \mathbb{R}^n
- nonlinear settings: a surface embedded in \mathbb{R}^n which can be locally represented as a graph

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- linear settings: a linear vector subspace of \mathbb{R}^n
- nonlinear settings: a surface embedded in \mathbb{R}^n which can be locally represented as a graph
- there is also the center manifold (invariant manifold that appears in the center manifold theorem), but we will start with the stable manifold theorem

Stable manifold theorem: linear systems

Consider the linear system $\frac{d\mathbf{x}}{dt} = M\mathbf{x}$ where $M \in \mathbb{R}^{n \times n}$ and none of the eigenvalues of $M \in \mathbb{R}^{n \times n}$ have zero real part.

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Assume that M is diagonalizable (semi-simple) and denote its eigenvalues and eigenvectors by $\lambda_j = a_j + ib_j$ and $\mathbf{w}_j = \mathbf{u}_j + i\mathbf{v}_j$, where $a_j, b_j \in \mathbb{R}$, $\mathbf{u}_j, \mathbf{v}_j \in \mathbb{R}^n$, for $j = 1, 2, \dots, n$. Then we define

stable subspace: $E^s = \text{span}\{\mathbf{u}_j, \mathbf{v}_j \mid a_j < 0\}$

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Example (Question 1(b) on Problem Sheet 0): $\lambda_1 = -2$, $\lambda_2 = 2$ and $\lambda_3 = 3$

$$M = \begin{pmatrix} 2 & 1 & -1 \\ 1 & -1 & 2 \\ -1 & 1 & 2 \end{pmatrix} \quad \mathbf{w}_1 = \begin{pmatrix} 1 \\ -3 \\ 1 \end{pmatrix}, \quad \mathbf{w}_2 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}, \quad \mathbf{w}_3 = \begin{pmatrix} 2 \\ -1 \\ -3 \end{pmatrix}$$

Then we have $E^s = \text{span}\left\{\begin{pmatrix} 1 \\ -3 \\ 1 \end{pmatrix}\right\}$, $E^u = \text{span}\left\{\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 2 \\ -1 \\ -3 \end{pmatrix}\right\}$

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Denote the eigenvalues and generalized eigenvectors of M by

$$\lambda_j = a_j + i b_j \text{ and } \mathbf{w}_j = \mathbf{u}_j + i \mathbf{v}_j,$$

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remarks: (1) if λ is an eigenvalue of matrix $M \in \mathbb{R}^{n \times n}$ of algebraic multiplicity $m \leq n$, then for $k = 1, 2, \dots, m$, any nonzero solution \mathbf{v} of $(A - \lambda I)^k \mathbf{v} = \mathbf{0}$ is called a generalized eigenvector of M

(2) if some eigenvalues of $M \in \mathbb{R}^{n \times n}$ have zero real part, we also define
center subspace: $E^c = \text{span}\{\mathbf{u}_j, \mathbf{v}_j \mid a_j = 0\}$

examples: Question 1 on Problem Sheet 1

Stable manifold theorem: linear systems

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Moreover, the solution is given by $\mathbf{x}(t) = \sum_{j=1}^n c_j e^{\lambda_j t} \mathbf{w}_j$ which implies:

if $\mathbf{x}(t) = \mathbf{x}_0 \in E^s$, then $\lim_{t \rightarrow \infty} \mathbf{x}(t) = \mathbf{0}$ and $\lim_{t \rightarrow -\infty} \|\mathbf{x}(t)\| = \infty$

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Question 2 on Problem Sheet 1: this is also true for non-diagonalizable matrix M (the nonlinear system has locally similar behaviour close to a hyperbolic critical point)

Stable manifold theorem

Given C^1 vector field $\mathbf{f} : \Omega \times \Theta \rightarrow \mathbb{R}^n$, where $\Omega \subset \mathbb{R}^n$, $\Theta \subset \mathbb{R}^m$, and $\boldsymbol{\mu} \in \Theta$, we consider ODE system

$$\frac{d\mathbf{x}}{dt} = \mathbf{f}(\mathbf{x}; \boldsymbol{\mu})$$

WLOG, assume that $\mathbf{0} \in \Omega$ is the hyperbolic critical point, *i.e.* $\mathbf{f}(\mathbf{0}; \boldsymbol{\mu}) = \mathbf{0}$ and matrix $D\mathbf{f}(\mathbf{0})$ has k eigenvalues with negative real part and $n - k$ eigenvalues with positive real part. In particular, our discussion of linear systems is applicable to the linear system $\frac{d\mathbf{x}}{dt} = M\mathbf{x}$ with $M = D\mathbf{f}(\mathbf{0})$.

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Then there exists (local results):

- a k -dimensional differentiable manifold M_{loc}^s tangent to the stable subspace E^s of the linear system at $\mathbf{0}$ such that for all $t \geq 0$, we have $\phi_t(M_{\text{loc}}^s) \subset M_{\text{loc}}^s$ and for all $\mathbf{x}_0 \in M_{\text{loc}}^s$, we have

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- an $(n - k)$ -dimensional differentiable manifold M_{loc}^u tangent to the unstable subspace E^u of the linear system at $\mathbf{0}$ such that for all $t \leq 0$, we have $\phi_t(M_{\text{loc}}^u) \subset M_{\text{loc}}^u$ and for all $\mathbf{x}_0 \in M_{\text{loc}}^u$, we have

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B5.6: Nonlinear Dynamics, Bifurcations and Chaos (Lecture 3)

- summary of Lecture 2: we discussed
Notion of flows, stability of fixed points, Lyapunov function, invariant manifolds, stable manifold theorem, notion of hyperbolicity. Stable, unstable and center subspaces. (Questions 1, 2, 5 and 7 on Problem Sheet 1)
- today: we will continue in our discussion of Problem Sheet 1

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- course synopsis of **Lectures 1-8** (taken by both Part B and MSc students):
Discrete-time (maps) and continuous-time (differential equations) dynamical systems. Notion of flows, stability of fixed points, Lyapunov function, invariant manifolds, stable manifold theorem, notion of hyperbolicity, center manifold. Chemical reaction networks. Stable, unstable and center subspaces. Poincaré-Bendixson theorem. Periodic solutions, stable and unstable limit cycles. Introduction to bifurcation theory, covering saddle-node, transcritical, supercritical pitchfork and subcritical pitchfork bifurcations. Extended center manifold. Logistic map. Periodic points of maps. Stability of N-cycles. Period-doubling bifurcation. Sharkovsky's theorem. Invariant distribution.

Stable manifold theorem (last slide of Lecture 2)

Given C^1 vector field $\mathbf{f} : \Omega \times \Theta \rightarrow \mathbb{R}^n$, where $\Omega \subset \mathbb{R}^n$, $\Theta \subset \mathbb{R}^m$, and $\boldsymbol{\mu} \in \Theta$, we consider ODE system

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Stable manifold theorem: example

example: $\frac{dx_1}{dt} = -x_1 - x_2^2$

$$\frac{dx_2}{dt} = x_2 + x_1^2$$

Stable manifold theorem: example

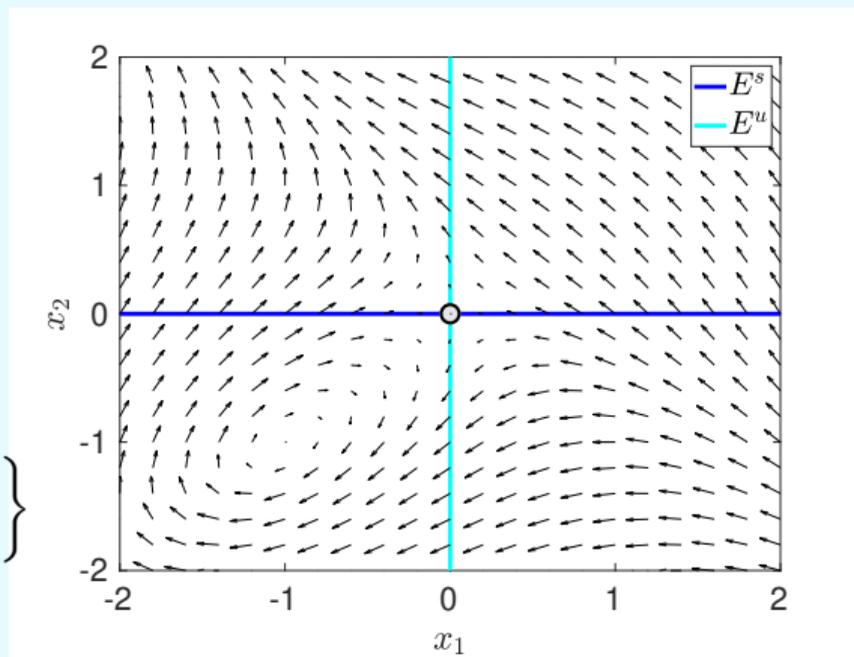
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$\mathbf{0} = [0, 0]$ is a fixed point

$$D\mathbf{f}(\mathbf{0}) = \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix}$$

$$E^s = \text{span} \left\{ \begin{pmatrix} 1 \\ 0 \end{pmatrix} \right\}, \quad E^u = \text{span} \left\{ \begin{pmatrix} 0 \\ 1 \end{pmatrix} \right\}$$



Stable manifold theorem: example

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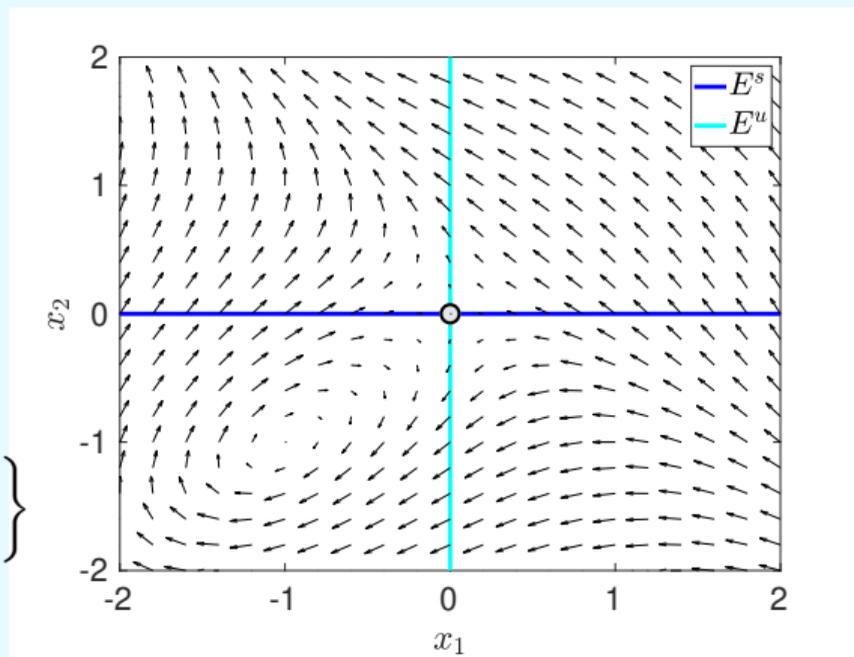
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M_{loc}^s is of the form $x_2 = c_1 x_1^2 + \mathcal{O}(x_1^3)$

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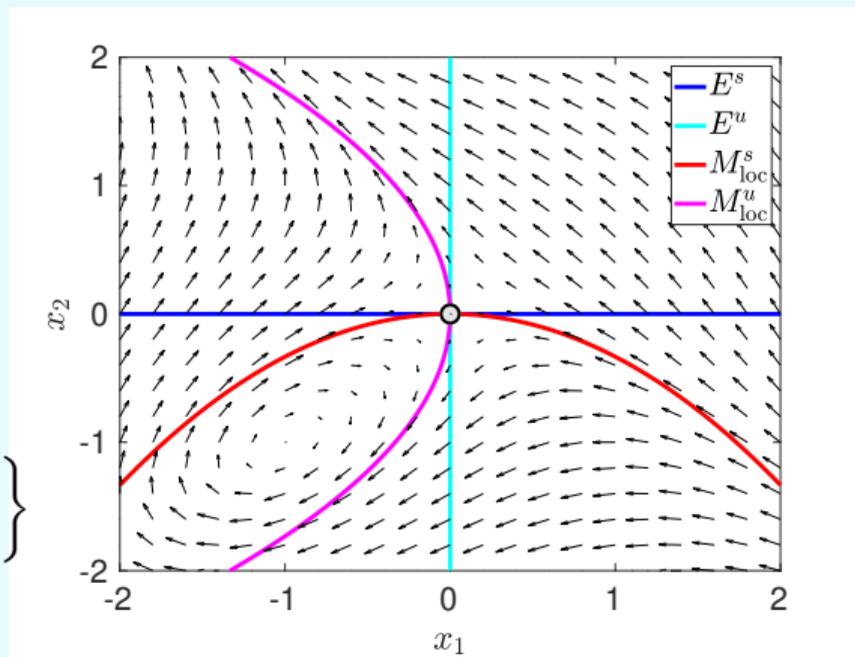
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differentiating these approximations, we get $c_1 = c_2 = -\frac{1}{3}$, i.e.

M_{loc}^s is of the form $x_2 = -\frac{x_1^2}{3} + \mathcal{O}(x_1^3)$ and M_{loc}^u is of the form $x_1 = -\frac{x_2^2}{3} + \mathcal{O}(x_2^3)$



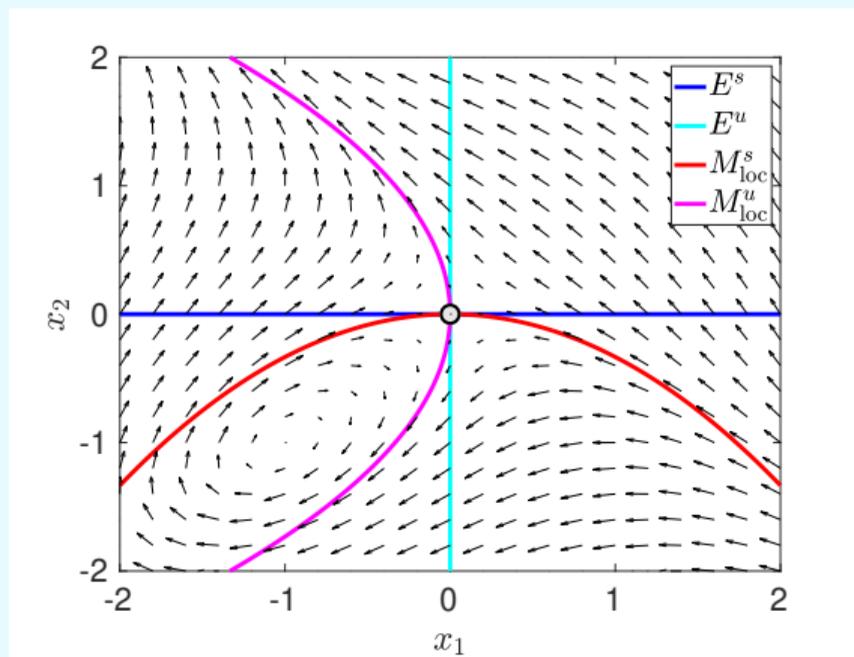
Global stable and unstable manifolds

global stable and unstable manifolds: $M^s = \bigcup_{t \leq 0} \phi_t(M_{\text{loc}}^s)$ and $M^u = \bigcup_{t \geq 0} \phi_t(M_{\text{loc}}^u)$

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Global stable and unstable manifolds

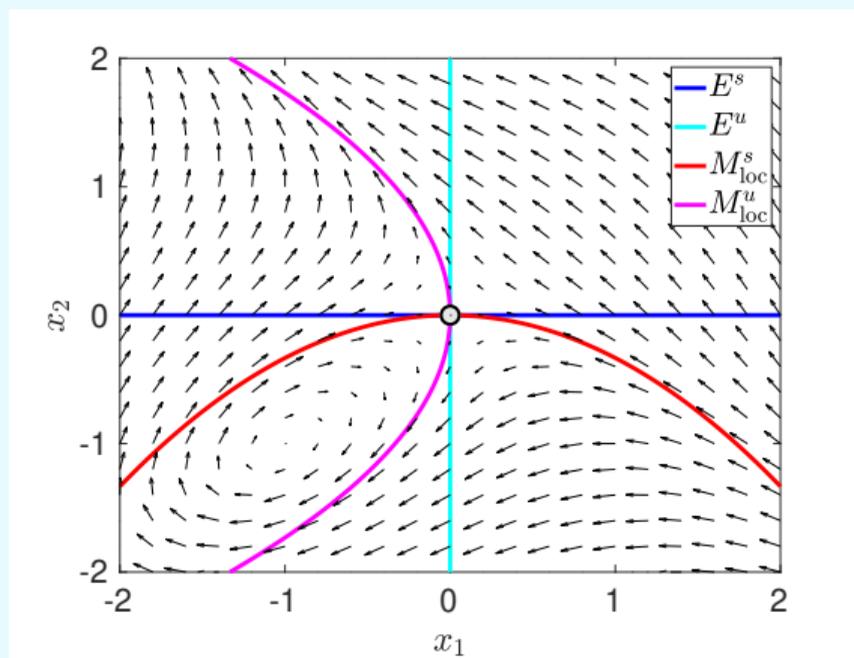
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example:
$$\frac{dx_1}{dt} = -x_1 - x_2^2$$
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observe that $A = 3x_1x_2 + x_1^3 + x_2^3$
is time independent:

$$\begin{aligned} \frac{dA}{dt} &= 3(x_2 + x_1^2) \frac{dx_1}{dt} + 3(x_1 + x_2^2) \frac{dx_2}{dt} \\ &= 3(x_2 + x_1^2)(-x_1 - x_2^2) \\ &\quad + 3(x_1 + x_2^2)(x_2 + x_1^2) = 0 \end{aligned}$$

consequently, both stable and unstable manifolds satisfy $A = 3x_1x_2 + x_1^3 + x_2^3 = 0$



Global stable and unstable manifolds

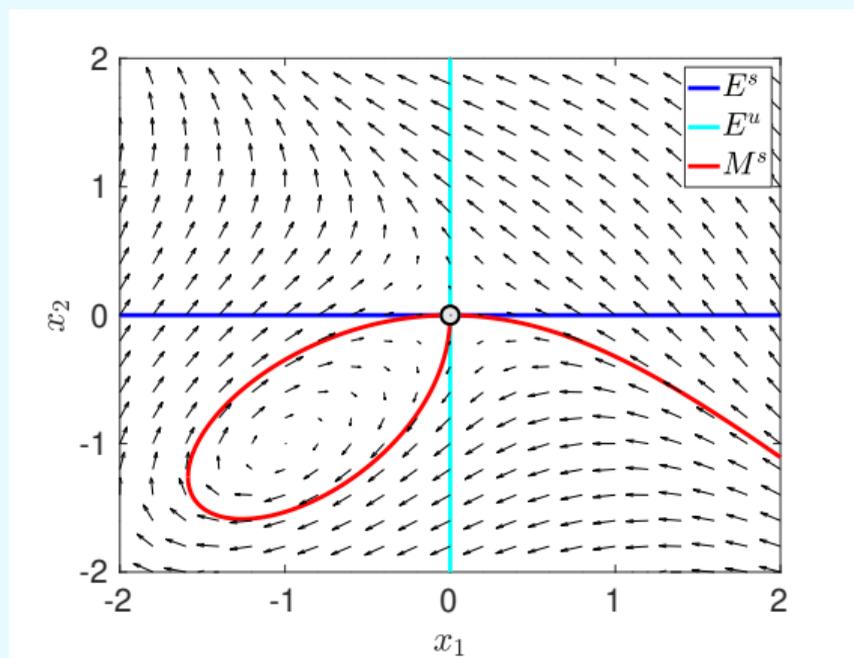
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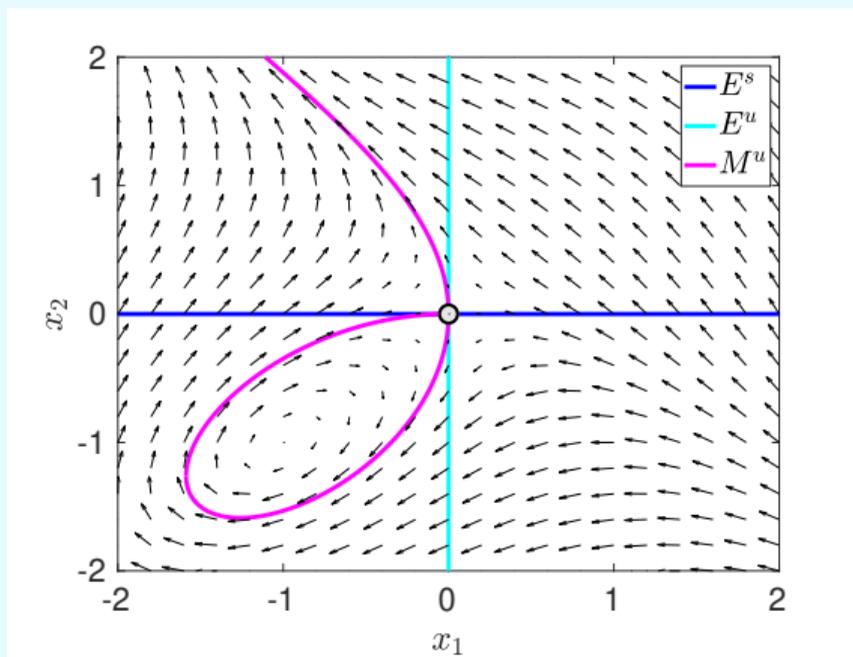
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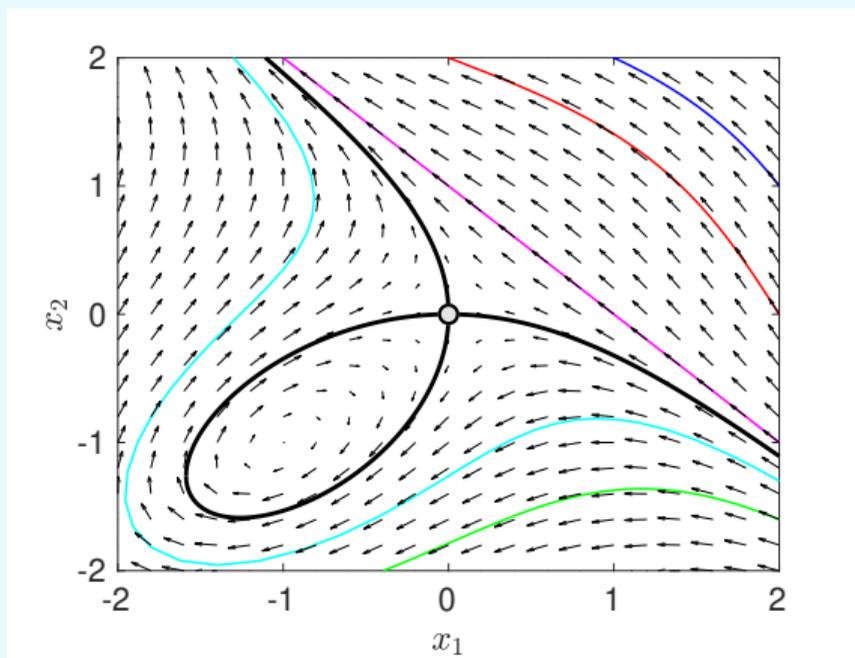
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Periodic solutions (closed orbits)

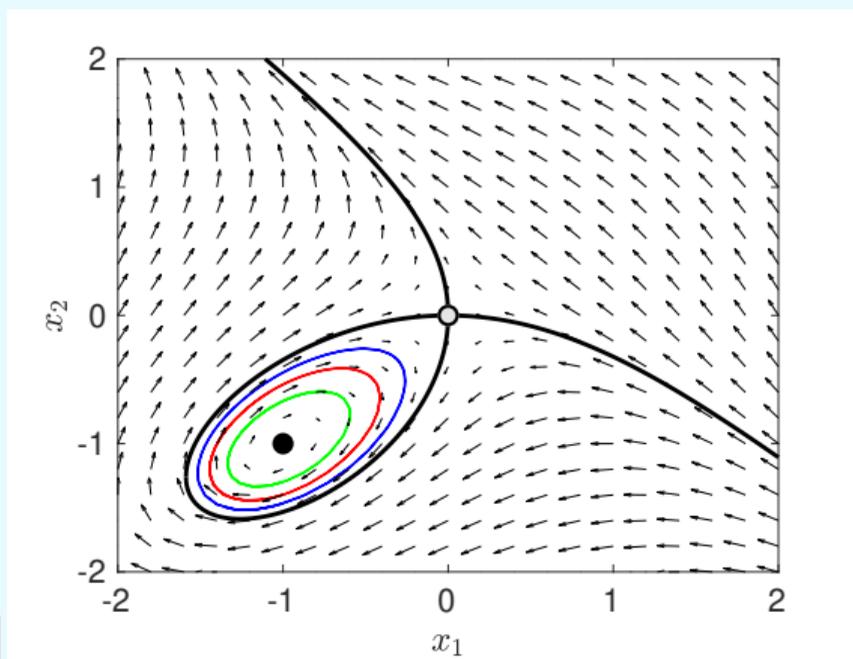
[Part A Differential Equations 1]: The existence of suitable A is one possible approach to prove the existence of periodic solutions (closed orbits) in planar systems.

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periodic solutions around point $[-1, -1]$,
satisfy $A = 3x_1x_2 + x_1^3 + x_2^3 = c$ for $c \in (0, 1)$



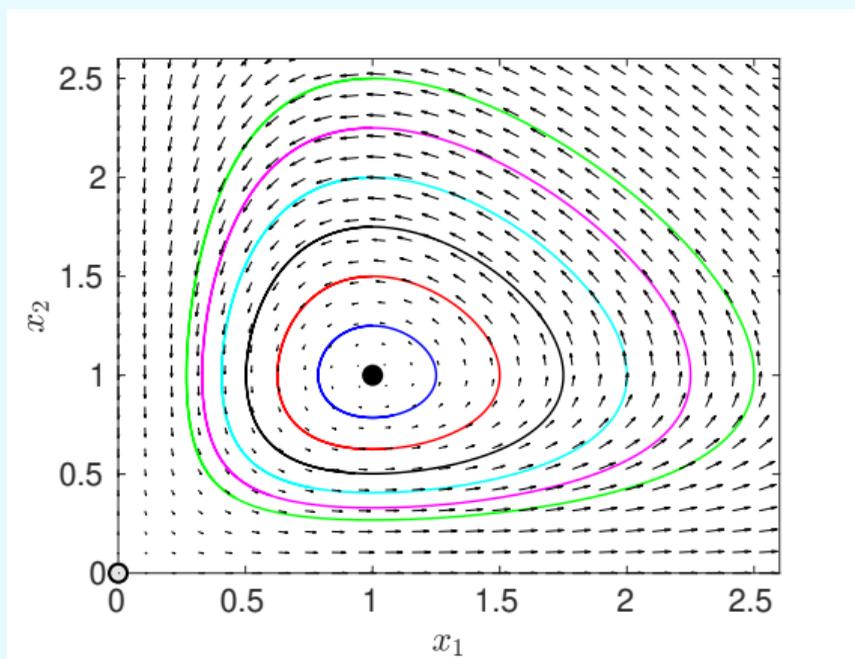
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Lotka-Volterra predator-prey equations



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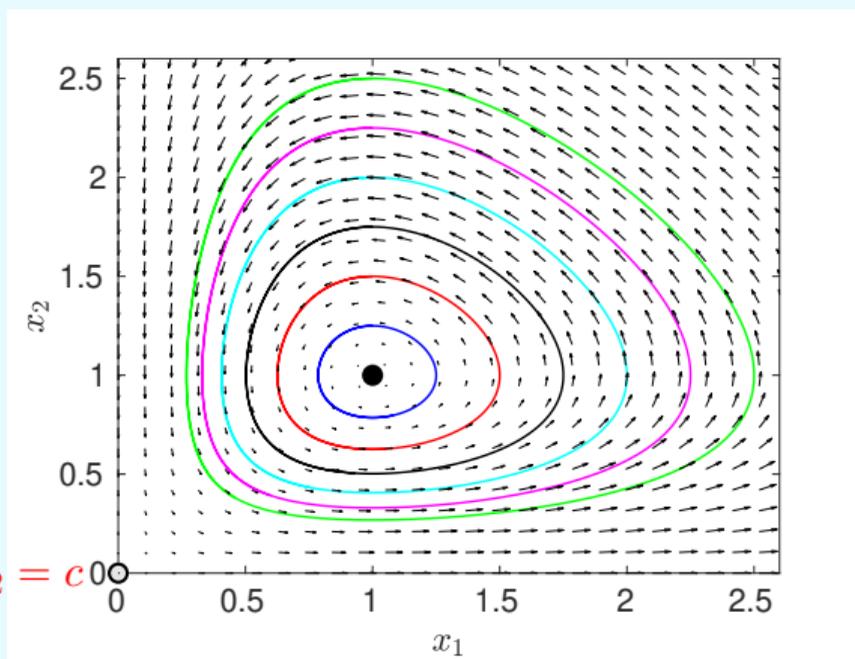
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[Part A Differential Equations 1]:

see pages 39-41 of your lecture notes from last year



Periodic solutions (closed orbits)

[Part A Differential Equations 1]: The existence of suitable A is one possible approach to prove the existence of periodic solutions (closed orbits) in planar systems.

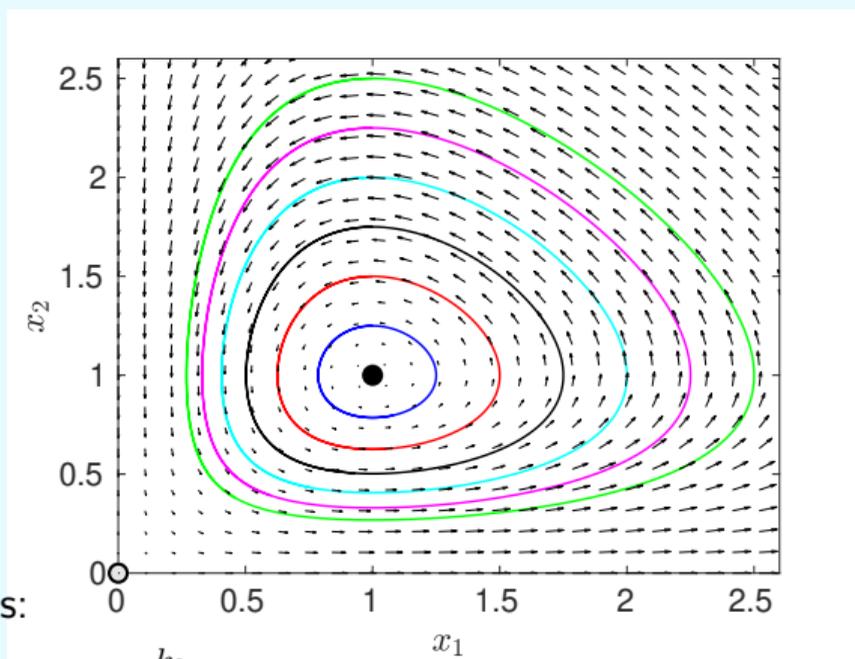
example:
$$\frac{dx_1}{dt} = x_1 - x_1x_2$$
$$\frac{dx_2}{dt} = -x_2 + x_1x_2$$

$A = \log(x_1) - x_1 + \log(x_2) - x_2$
is time independent

Note: Lotka-Volterra ODE system also describes a system of $n = 2$ chemical species X_1 and X_2 which are subject to the following $\ell = 3$ chemical reactions:



where the values of the rate constants are: $k_1 = k_2 = k_3 = 1$



Poincaré-Bendixson theorem ($n = 2$)

Given C^1 vector field $\mathbf{f} : \Omega \times \Theta \rightarrow \mathbb{R}^2$, where $\Omega \subset \mathbb{R}^2$, $\Theta \subset \mathbb{R}^m$, and $\boldsymbol{\mu} \in \Theta$, we consider the planar ODE system

$$\frac{d\mathbf{x}}{dt} = \mathbf{f}(\mathbf{x}; \boldsymbol{\mu})$$

Suppose that $R \subset \Omega$ is compact (*i.e.* closed and bounded) and

- R does not contain any fixed points
- there exists $\mathbf{x}_0 \in R$ such that $\phi_t(\mathbf{x}_0) \in R$ for all $t \geq 0$, *i.e.* the trajectory is confined in R for $t \geq 0$

Poincaré-Bendixson theorem: Then either $\Gamma_{\mathbf{x}_0}$ is a closed orbit, or $\phi_t(\mathbf{x}_0)$ spirals toward a closed orbit as $t \rightarrow \infty$. In either case, **R contains a closed orbit.**

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application of the Poincaré-Bendixson theorem

- we need to find a *trapping region*: compact connected subset of Ω such that the vector field $\mathbf{f}(\mathbf{x}; \boldsymbol{\mu})$ points 'inward' everywhere on the boundary

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- we need to show that any fixed point in the trapping region is unstable, and remove its small neighbourhood to construct R

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application of the Poincaré-Bendixson theorem [Question 6 on Problem Sheet 1]

- we need to find a *trapping region*: compact connected subset of Ω such that the vector field $\mathbf{f}(\mathbf{x}; \boldsymbol{\mu})$ points 'inward' everywhere on the boundary [Question 6(c)]
- we need to show that any fixed point in the trapping region is unstable, and remove its small neighbourhood to construct R [Question 6(d)]

Poincaré-Bendixson theorem: example

$$\frac{dx_1}{dt} = x_1 + x_2 - x_1^3$$

$$\frac{dx_2}{dt} = -x_1 - x_2^3$$

Poincaré-Bendixson theorem: example

$$\frac{dx_1}{dt} = x_1 + x_2 - x_1^3 = f_1(x_1, x_2)$$

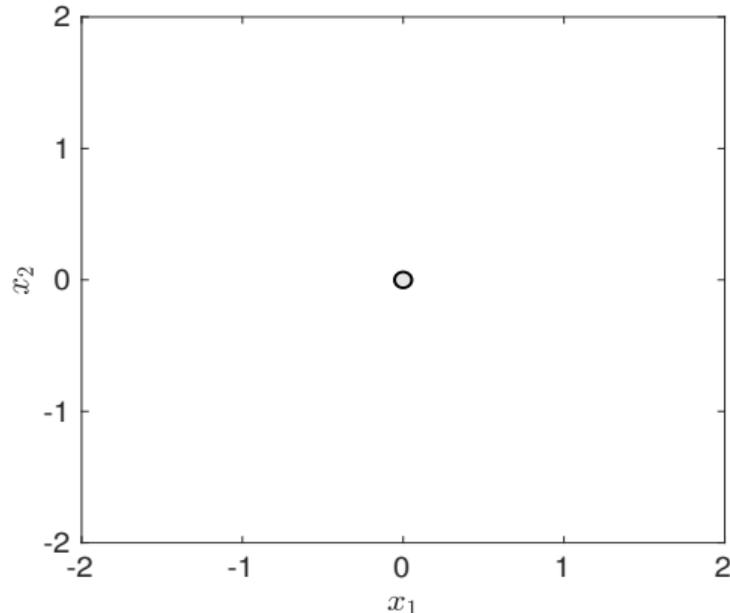
$$\frac{dx_2}{dt} = -x_1 - x_2^3 = f_2(x_1, x_2)$$

solving $f_1(x_1, x_2) = f_2(x_1, x_2) = 0$, we get $\mathbf{0} = [0, 0]$ as the only fixed point

$$D\mathbf{f}(\mathbf{0}) = \begin{pmatrix} 1 & 1 \\ -1 & 0 \end{pmatrix}$$

eigenvalues $\lambda_{\pm} = (1 \pm i\sqrt{3})/2$

$\mathbf{0} = [0, 0]$ is unstable (spiral)



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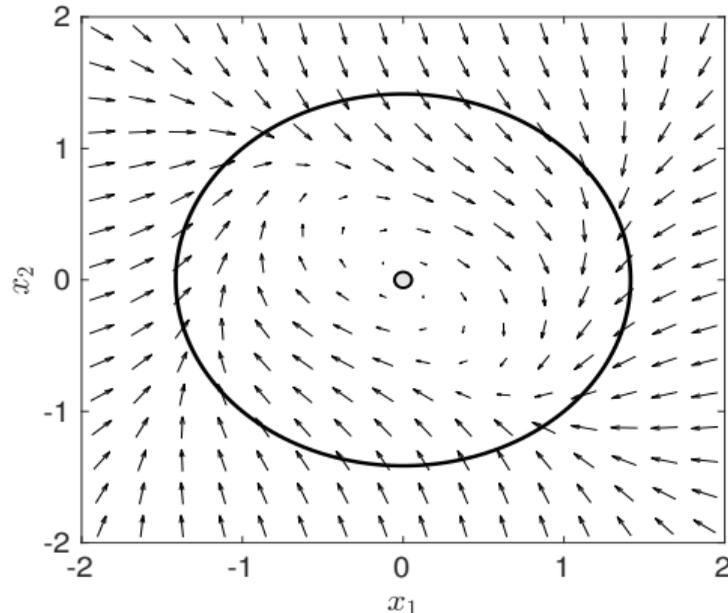
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trapping region: circle with boundary $\mathbf{z}(s) = [\sqrt{2} \cos(s), \sqrt{2} \sin(s)]$ for $s \in [0, 2\pi]$
and inward pointing normal $\mathbf{n}(s) = [-\cos(s), -\sin(s)]$

we have $\mathbf{n}(s) \cdot \mathbf{f}(\mathbf{z}(s)) = \sqrt{2}(2(\cos^4(s) + \sin^4(s)) - \cos^2(s)) > 0$



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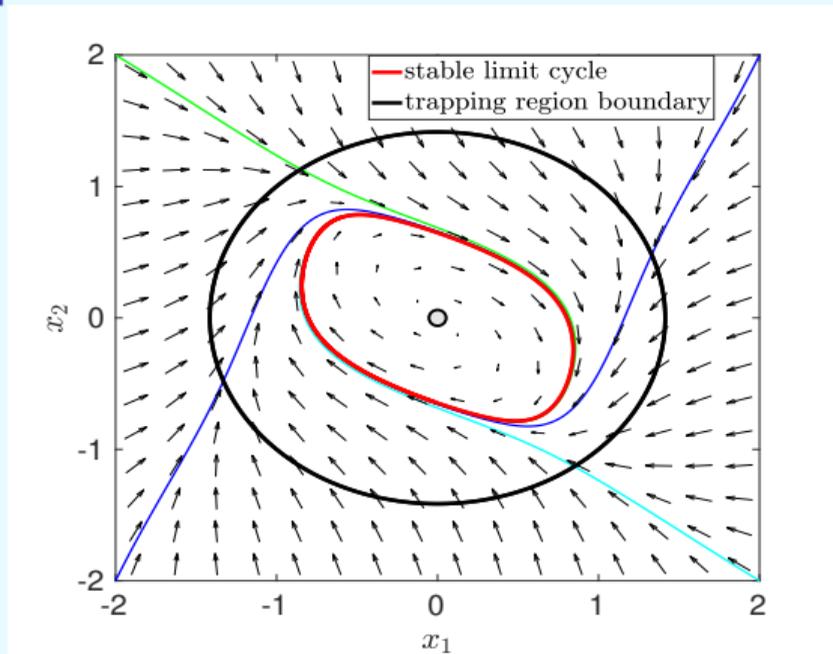
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\implies the Poincaré-Bendixson theorem implies the existence of a closed orbit



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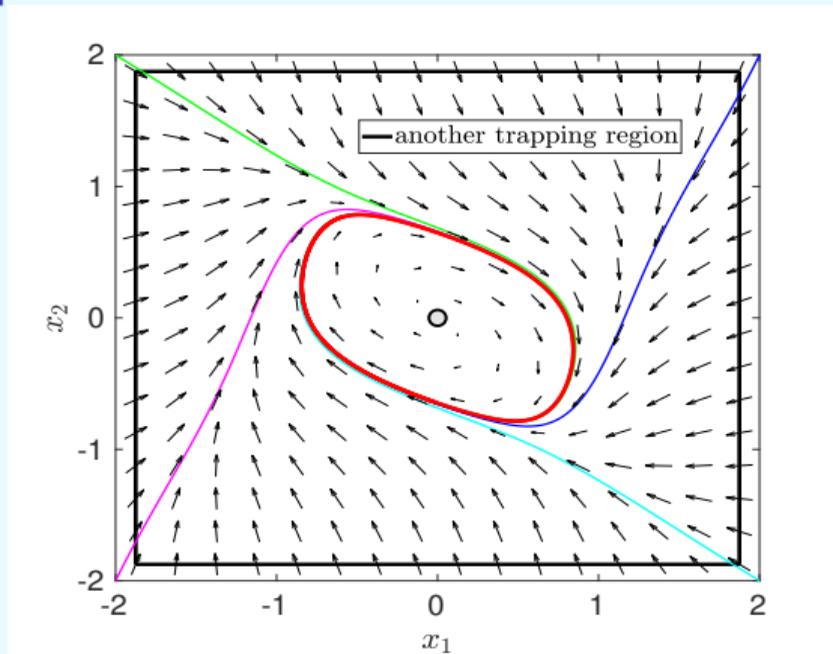
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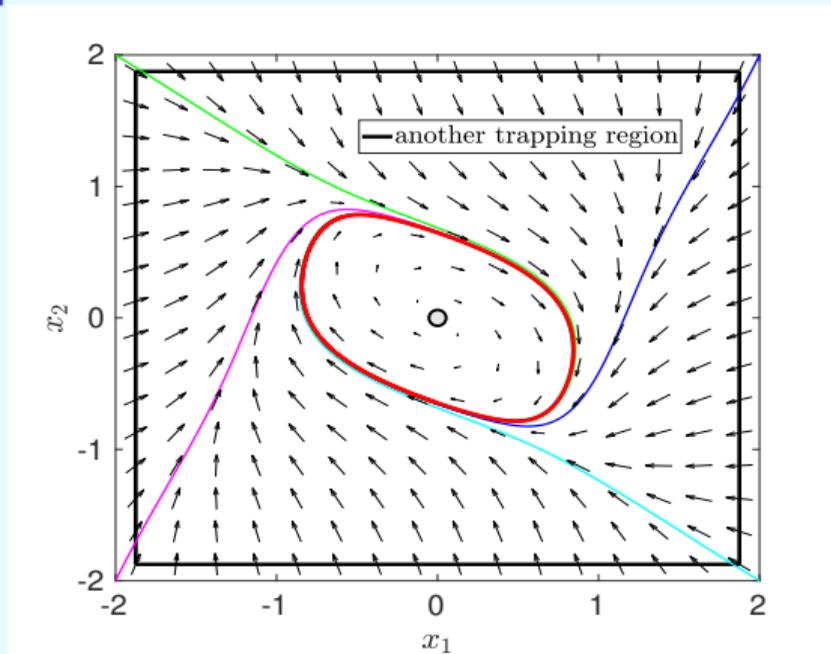
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[Question 6 on Problem Sheet 1]: you could choose a trapping region as a polygon (specify its vertices, parameterize its edges and verify that the vector field $\mathbf{f}(\mathbf{x}; \boldsymbol{\mu})$ points inward everywhere on its boundary)



Center manifold

Given C^r vector field $\mathbf{f} : \Omega \times \Theta \rightarrow \mathbb{R}^n$, where $\Omega \subset \mathbb{R}^n$, $\Theta \subset \mathbb{R}^m$, and $\boldsymbol{\mu} \in \Theta$, we consider ODE system

$$\frac{d\mathbf{x}}{dt} = \mathbf{f}(\mathbf{x}; \boldsymbol{\mu})$$

Assume that $\mathbf{x}_c \in \Omega$ is the critical point, *i.e.* $\mathbf{f}(\mathbf{x}_c; \boldsymbol{\mu}) = \mathbf{0}$ and matrix $D\mathbf{f}(\mathbf{x}_c)$ has $k > 0$ eigenvalues with zero real part and $n - k$ eigenvalues with non-zero real part.

Then there exists a k -dimensional C^r -manifold M_{loc}^c tangent to center subspace E^s of the linear system at \mathbf{x}_c such that for all $t \geq 0$, we have $\phi_t(M_{\text{loc}}^c) \subset M_{\text{loc}}^c$.

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- If the unstable manifold is non-empty, then the fixed point \mathbf{x}_c is unstable.

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- If the unstable manifold is non-empty, then the fixed point \mathbf{x}_c is unstable.
- Suppose the unstable manifold is empty and the system has both a non-empty stable and center manifold. Then the stability of the fixed point \mathbf{x}_c is governed by the dynamics on the center manifold.

B5.6: Nonlinear Dynamics, Bifurcations and Chaos (Lecture 4)

- summary of Lecture 3: we discussed
Invariant manifolds, stable manifold theorem. Poincaré-Bendixson theorem. Periodic solutions, stable and unstable limit cycles. ([Question 6 on Problem Sheet 1](#))
- today: we will conclude our discussion of Problem Sheet 1

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Invariant manifolds, stable manifold theorem. Poincaré-Bendixson theorem. Periodic solutions, stable and unstable limit cycles. (Question 6 on Problem Sheet 1)
- today: we will conclude our discussion of Problem Sheet 1
- course synopsis of **Lectures 1-8** (taken by both Part B and MSc students):
Discrete-time (maps) and continuous-time (differential equations) dynamical systems. Notion of flows, stability of fixed points, Lyapunov function, invariant manifolds, stable manifold theorem, notion of hyperbolicity, center manifold. Chemical reaction networks. Stable, unstable and center subspaces. Poincaré-Bendixson theorem. Periodic solutions, stable and unstable limit cycles. Introduction to bifurcation theory, covering saddle-node, transcritical, supercritical pitchfork and subcritical pitchfork bifurcations. Extended center manifold. Logistic map. Periodic points of maps. Stability of N-cycles. Period-doubling bifurcation. Sharkovsky's theorem. Invariant distribution.

Stable manifold theorem (last slide of Lecture 2)

Given C^1 vector field $\mathbf{f} : \Omega \times \Theta \rightarrow \mathbb{R}^n$, where $\Omega \subset \mathbb{R}^n$, $\Theta \subset \mathbb{R}^m$, and $\boldsymbol{\mu} \in \Theta$, we consider ODE system

$$\frac{d\mathbf{x}}{dt} = \mathbf{f}(\mathbf{x}; \boldsymbol{\mu})$$

WLOG, assume that $\mathbf{0} \in \Omega$ is the **hyperbolic critical point**, i.e. $\mathbf{f}(\mathbf{0}; \boldsymbol{\mu}) = \mathbf{0}$ and matrix $D\mathbf{f}(\mathbf{0})$ has k eigenvalues with negative real part and $n - k$ eigenvalues with positive real part. In particular, our discussion of linear systems is applicable to the linear system $\frac{d\mathbf{x}}{dt} = M\mathbf{x}$ with $M = D\mathbf{f}(\mathbf{0})$.

Then there exists (local results):

- a k -dimensional differentiable manifold M_{loc}^s tangent to the stable subspace E^s of the linear system at $\mathbf{0}$ such that for all $t \geq 0$, we have $\phi_t(M_{\text{loc}}^s) \subset M_{\text{loc}}^s$ and for all $\mathbf{x}_0 \in M_{\text{loc}}^s$, we have

$$\lim_{t \rightarrow \infty} \phi_t(\mathbf{x}_0) = \mathbf{0}$$

- an $(n - k)$ -dimensional differentiable manifold M_{loc}^u tangent to the unstable subspace E^u of the linear system at $\mathbf{0}$ such that for all $t \leq 0$, we have $\phi_t(M_{\text{loc}}^u) \subset M_{\text{loc}}^u$ and for all $\mathbf{x}_0 \in M_{\text{loc}}^u$, we have

$$\lim_{t \rightarrow -\infty} \phi_t(\mathbf{x}_0) = \mathbf{0}$$

Examples of non-hyperbolic critical points ($n = 2$)

$$\frac{dx_1}{dt} = x_2 - 3x_1^3$$

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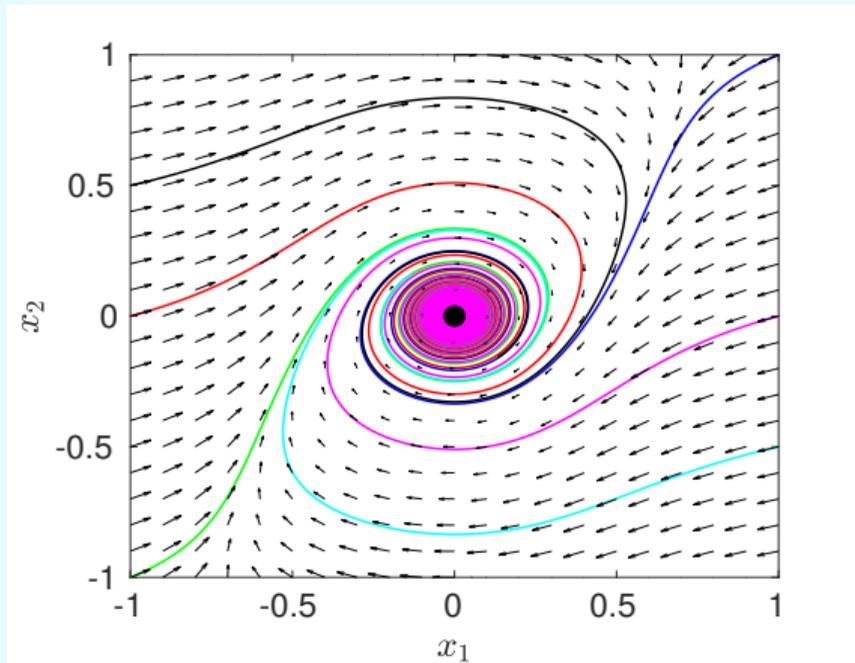
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$\mathbf{0} = [0, 0]$ is a critical point

$$D\mathbf{f}(\mathbf{0}) = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}$$

eigenvalues $\lambda_{\pm} = \pm i$

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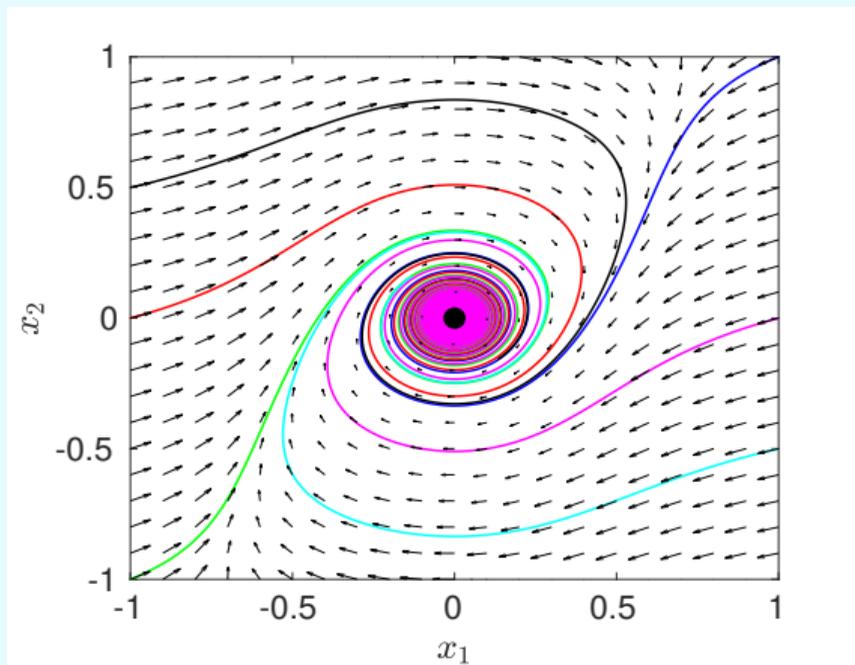
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[Part A Differential Equations 1:] examples of ODEs where the critical point of the linearized system is a center (periodic solutions), and the non-linear system either has periodic solutions (Lotka-Volterra equations), or does not have periodic solutions the critical point can be stable or unstable (examples above), center or **focus**

Examples of non-hyperbolic critical points ($n = 2$)

$$\frac{dx_1}{dt} = x_2 + 3x_1^3$$

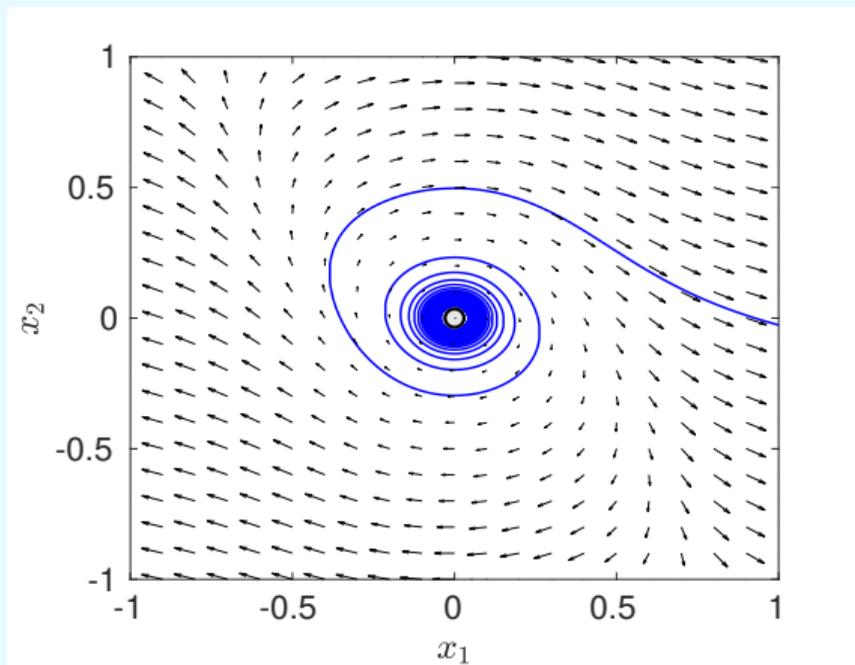
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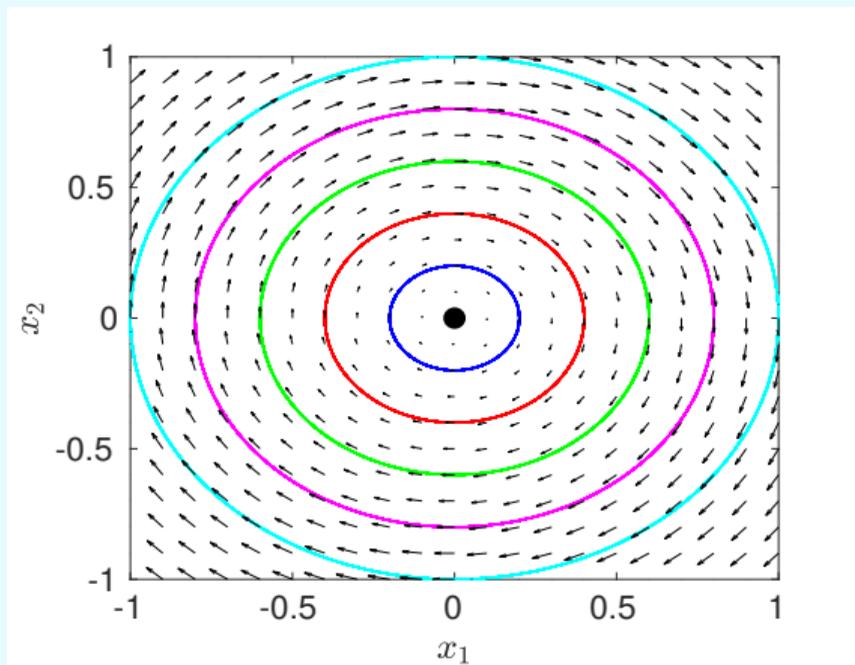
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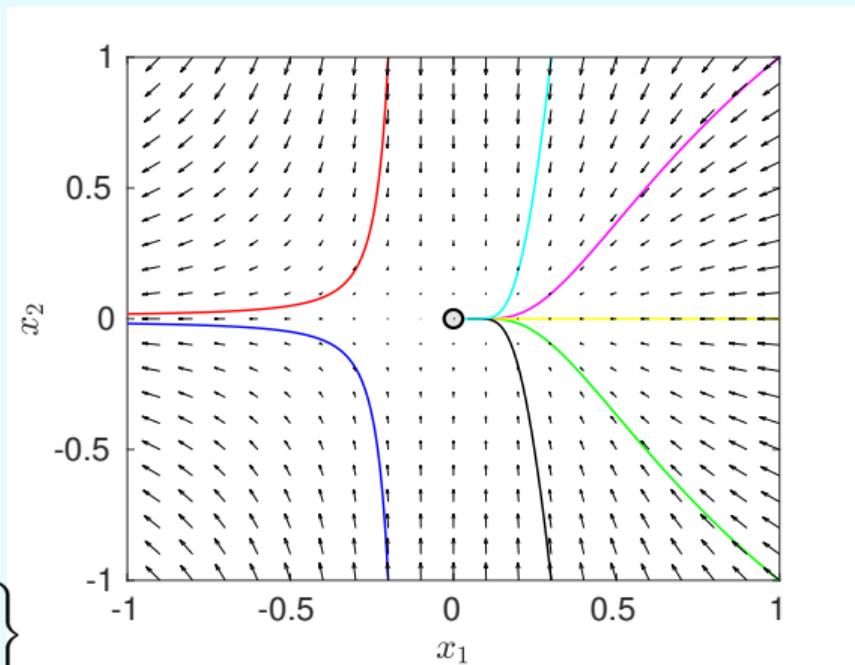
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$$E^c = \text{span} \left\{ \begin{pmatrix} 1 \\ 0 \end{pmatrix} \right\}, \quad E^s = \text{span} \left\{ \begin{pmatrix} 0 \\ 1 \end{pmatrix} \right\}$$

$\mathbf{0} = [0, 0]$ is a **saddle-node**



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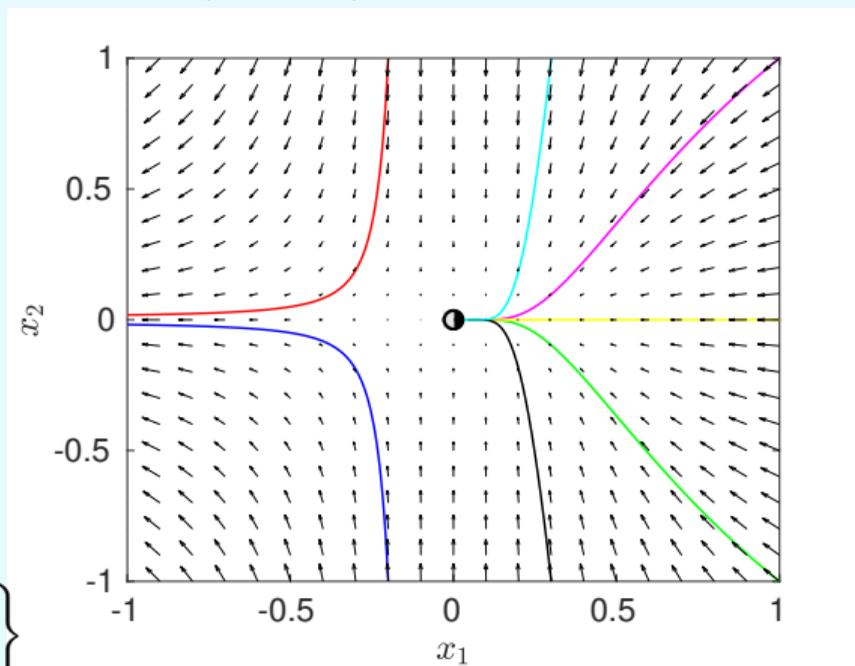
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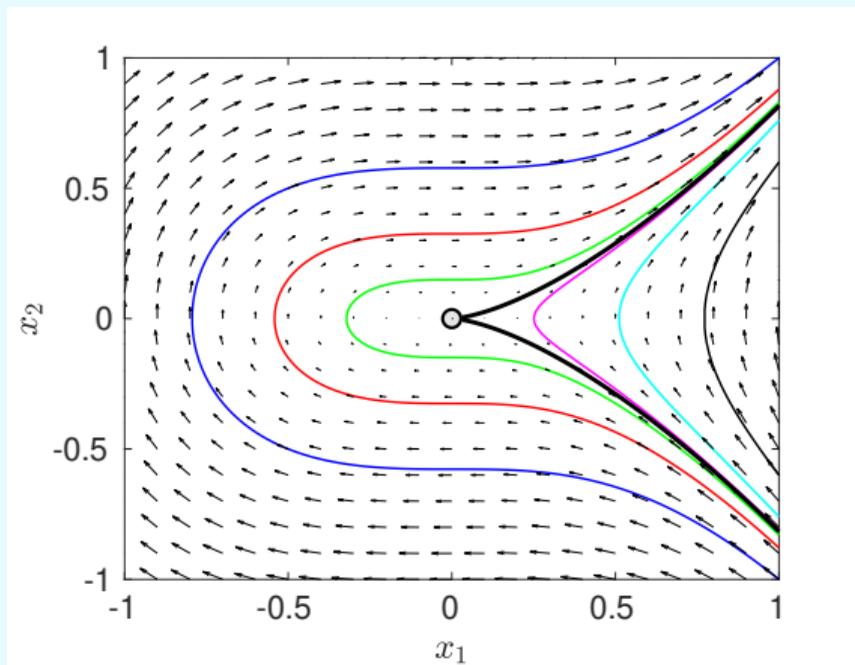
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eigenvalue $\lambda = 0$

$\mathbf{0} = [0, 0]$ is a **cusp**



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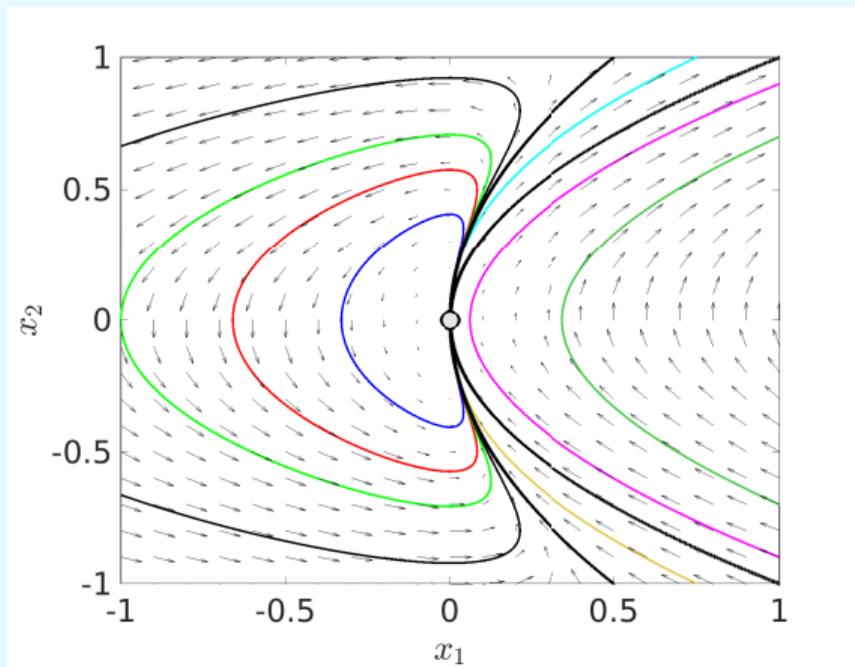
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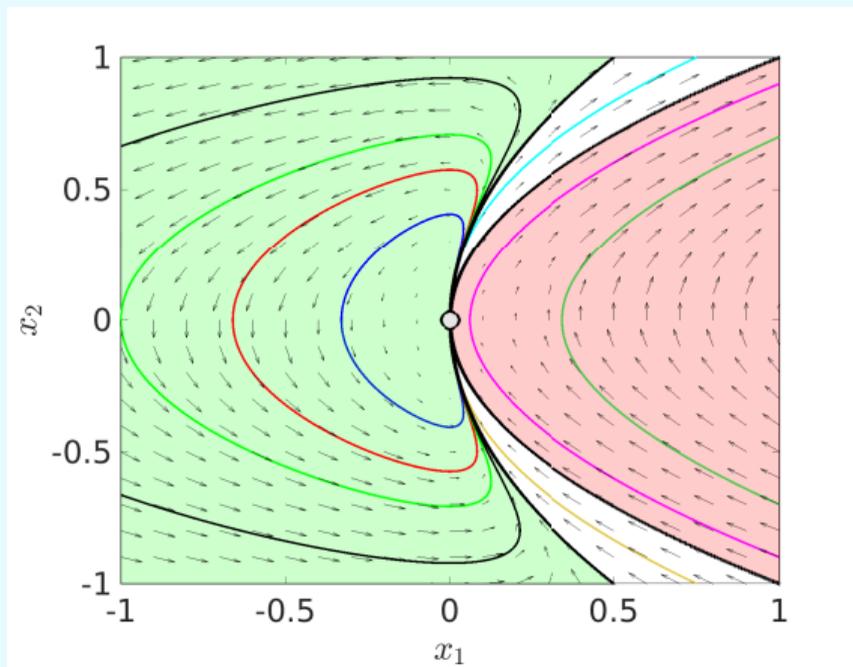
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green - elliptic sector

red - hyperbolic sector

white - two parabolic sectors

Examples of non-hyperbolic critical points ($n = 2$)

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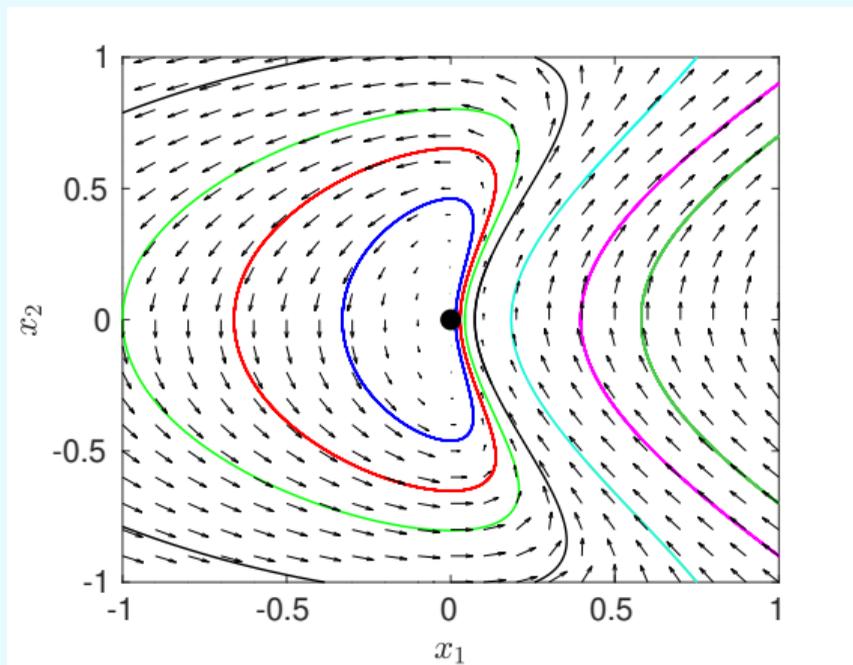
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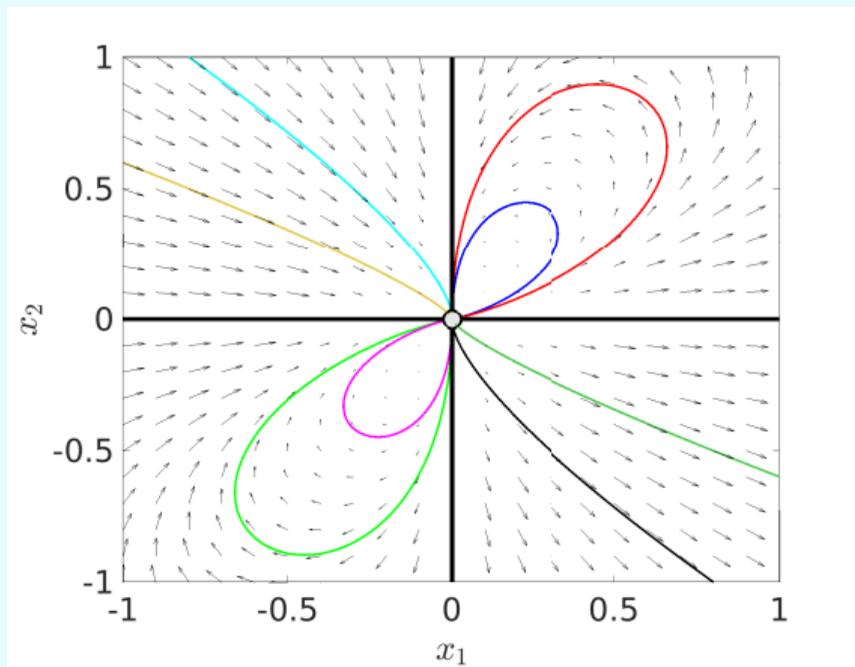
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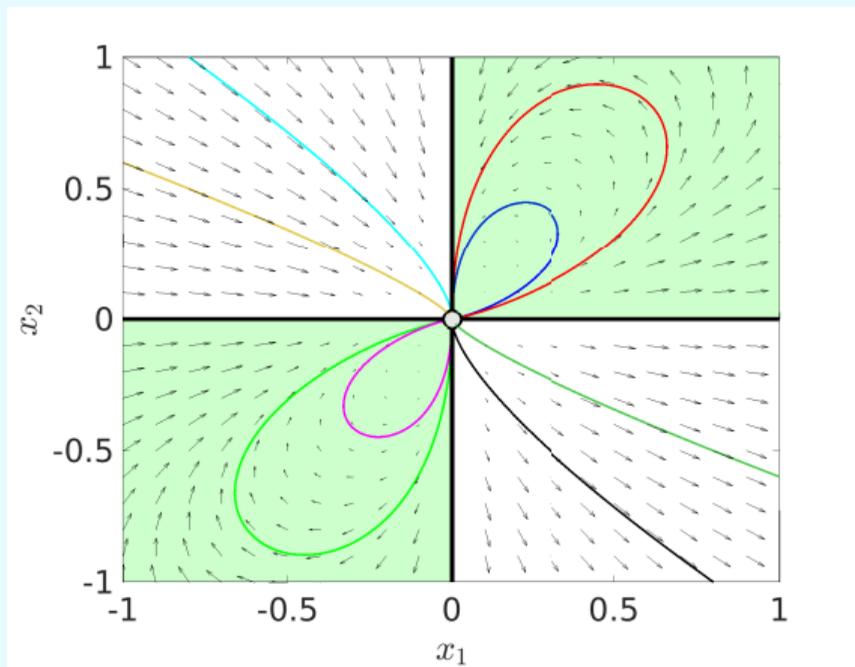
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green - two elliptic sectors
white - two parabolic sectors

Center manifold (last slide of Lecture 3)

Given C^r vector field $\mathbf{f} : \Omega \times \Theta \rightarrow \mathbb{R}^n$, where $\Omega \subset \mathbb{R}^n$, $\Theta \subset \mathbb{R}^m$, and $\boldsymbol{\mu} \in \Theta$, we consider ODE system

$$\frac{d\mathbf{x}}{dt} = \mathbf{f}(\mathbf{x}; \boldsymbol{\mu})$$

Assume that $\mathbf{x}_c \in \Omega$ is the critical point, *i.e.* $\mathbf{f}(\mathbf{x}_c; \boldsymbol{\mu}) = \mathbf{0}$ and matrix $D\mathbf{f}(\mathbf{x}_c)$ has $k > 0$ eigenvalues with zero real part and $n - k$ eigenvalues with non-zero real part.

Then there exists a k -dimensional C^r -manifold M_{loc}^c tangent to center subspace E^s of the linear system at \mathbf{x}_c such that for all $t \geq 0$, we have $\phi_t(M_{\text{loc}}^c) \subset M_{\text{loc}}^c$.

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- If the unstable manifold is non-empty, then the fixed point \mathbf{x}_c is unstable.
- Suppose the unstable manifold is empty and the system has both a non-empty stable and center manifold. Then the stability of the fixed point \mathbf{x}_c is governed by the dynamics on the center manifold.

Reduction to the center manifold

example: $\frac{dx_1}{dt} = x_1^2(x_2 - x_1^3)$

$$\frac{dx_2}{dt} = x_1^2 - x_2$$

Reduction to the center manifold

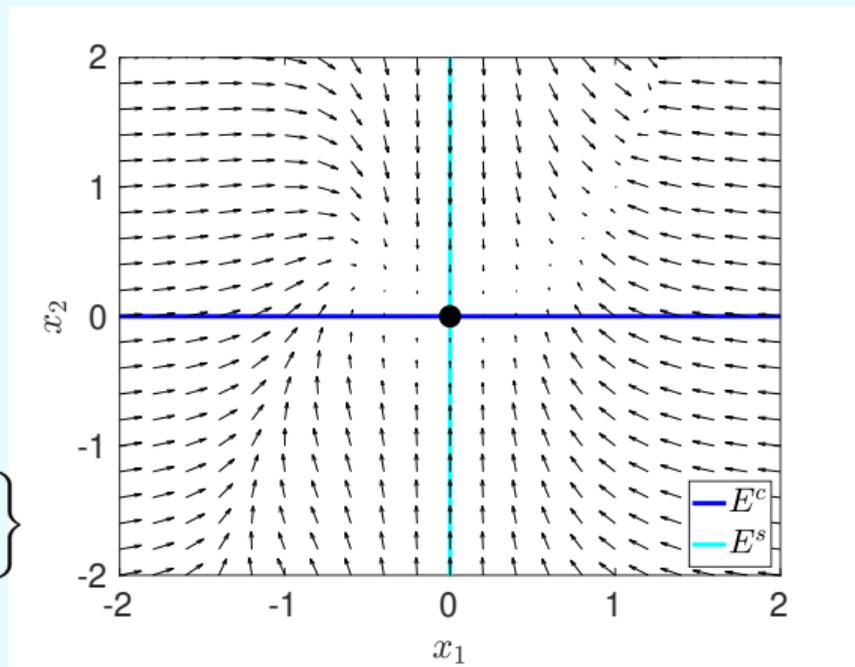
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Reduction to the center manifold

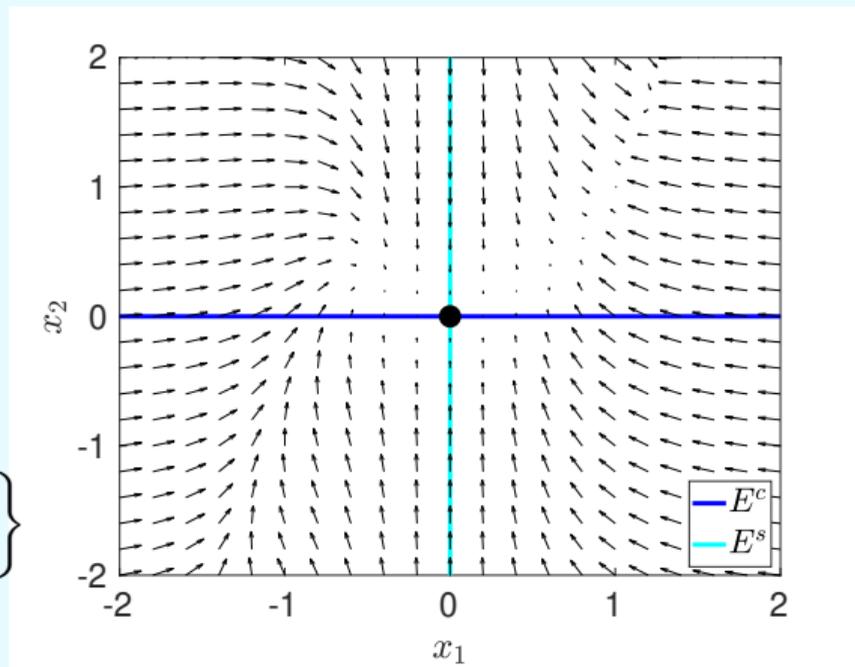
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Warning: On the center linear subspace, we have $x_2 = 0$. Substituting $x_2 = 0$ into the first equation gives $dx_1/dt = -x_1^5$, but this does not mean that the origin is stable!

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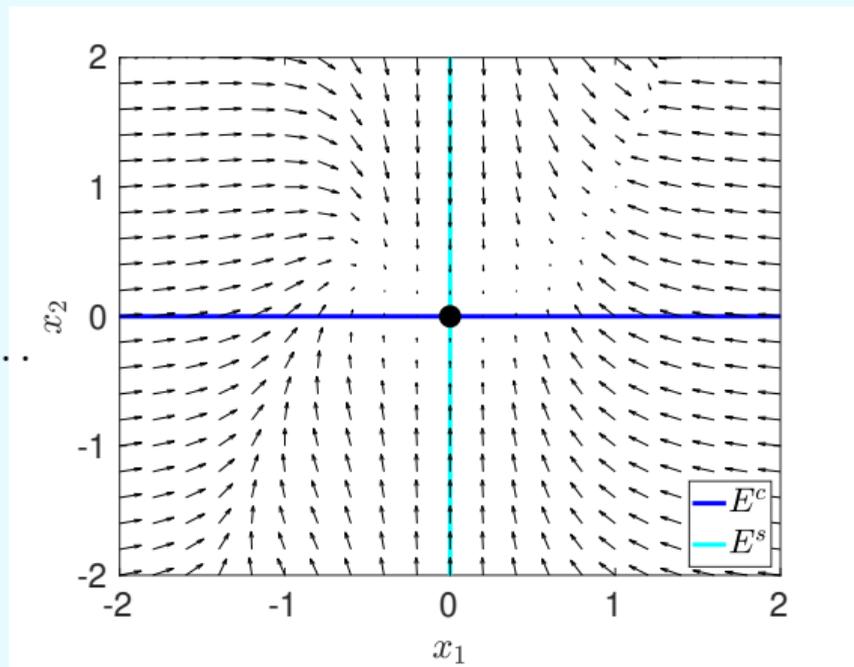
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M_{loc}^c is of the form

$$x_2 = h(x_1) = h_2 x_1^2 + h_3 x_1^3 + h_4 x_1^4 + \dots$$

- differentiating, we get

$$\frac{dx_2}{dt} = (2h_2 x_1 + 3h_3 x_1^2 + \dots) \frac{dx_1}{dt}$$



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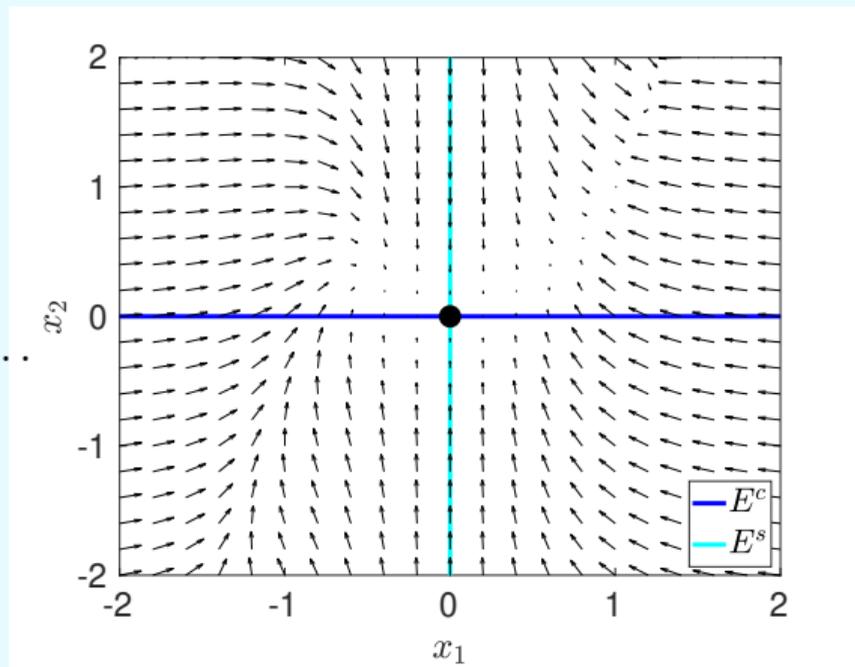
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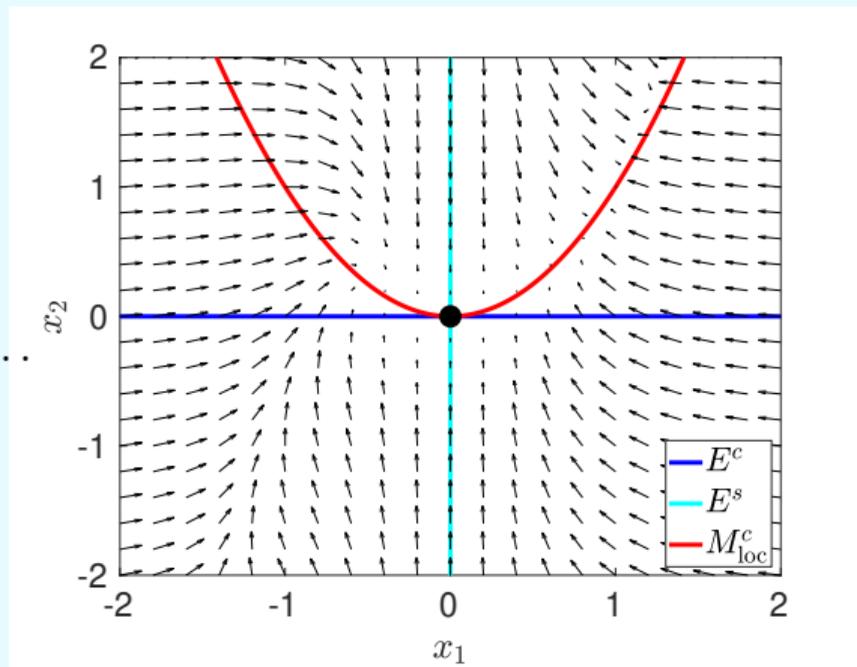
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we have $x_1^2 - h(x_1) = (2h_2x_1 + 3h_3x_1^2 + \dots) x_1^2(h(x_1) - x_1^3)$

- equating coefficients of powers of x_1 gives $h_2 = 1$, $h_3 = 0$ and $h_4 = 0$

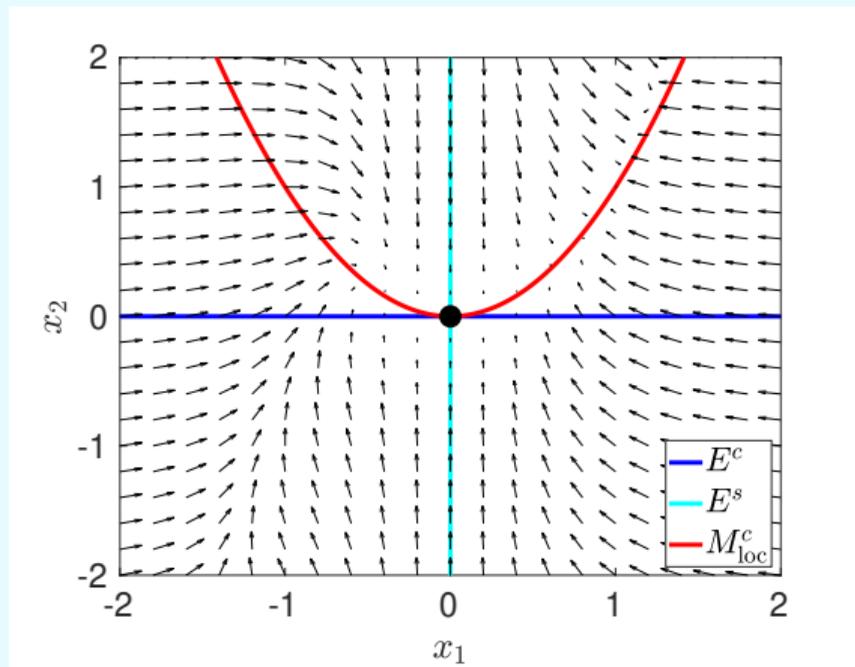


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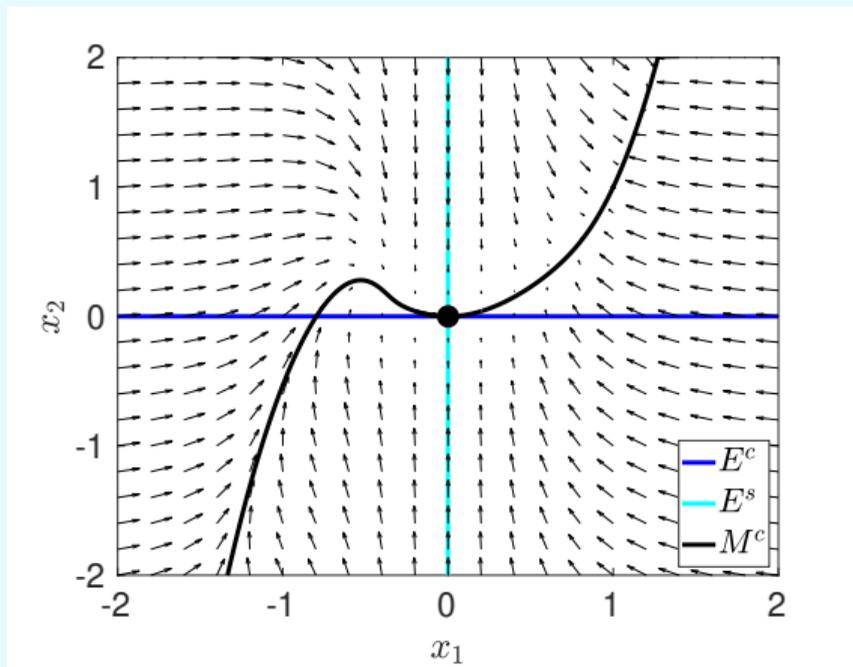
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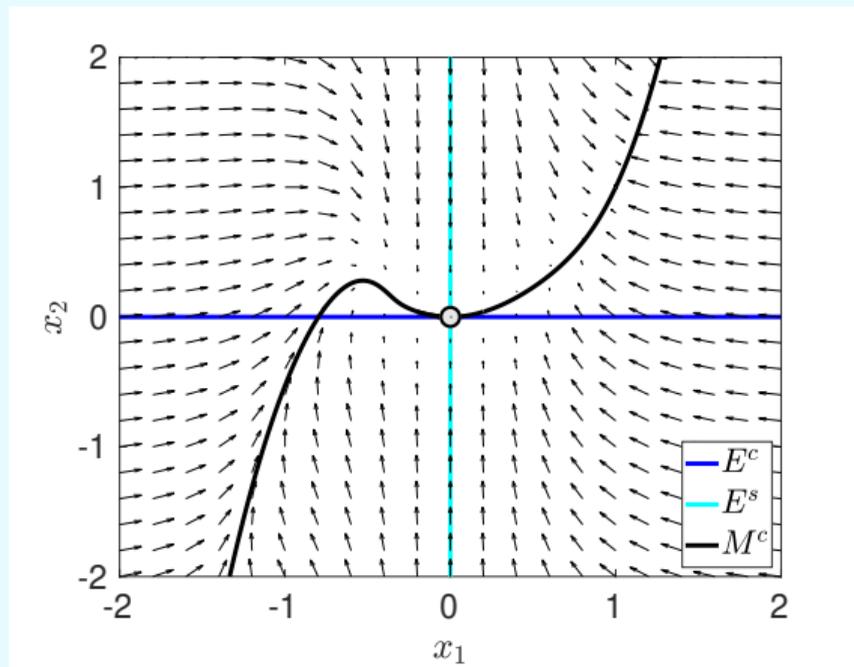
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which implies that the origin is unstable



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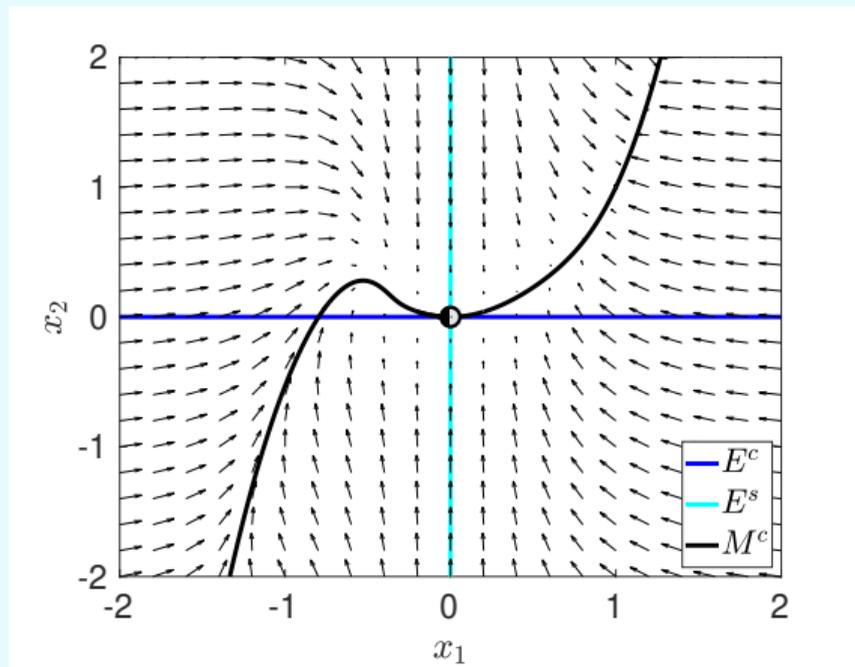
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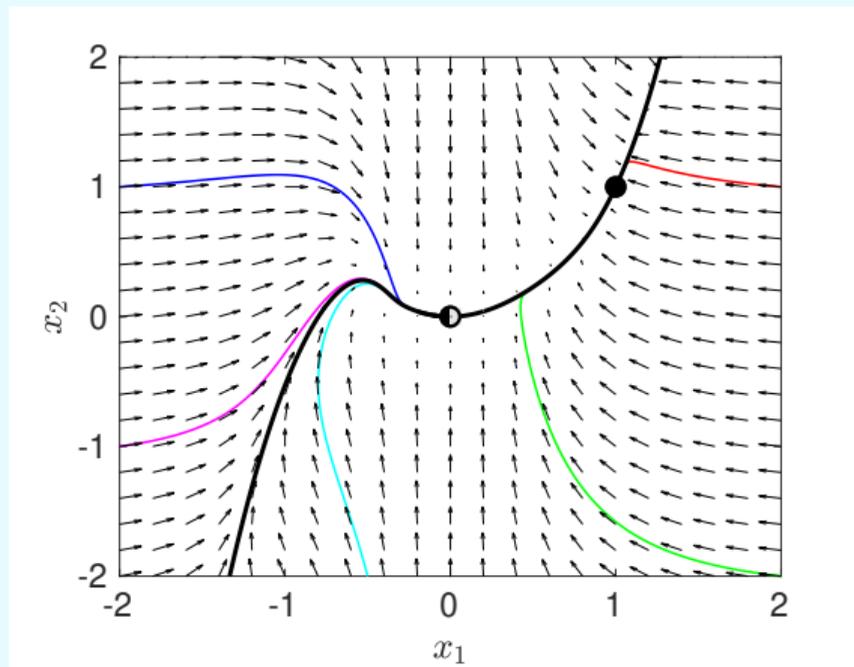
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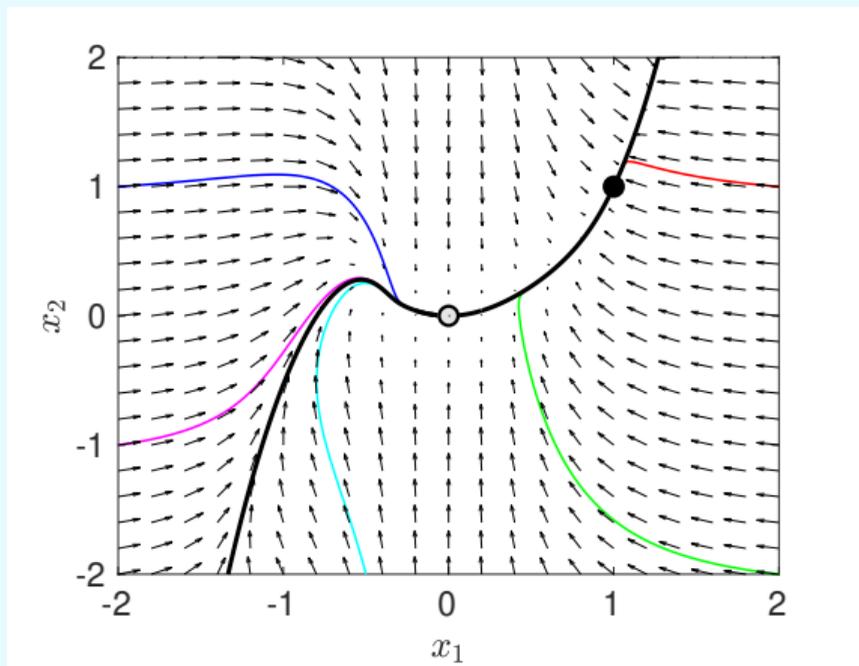
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another example: [Question 3 on Problem Sheet 1](#)



Example 2: reduction to the center manifold

$$\frac{dx_1}{dt} = x_2 - x_1 + 2x_2(x_1 + x_2) - (x_1 + x_2)^6$$

$$\frac{dx_2}{dt} = x_1 - x_2 - 2x_1(x_1 + x_2) - (x_1 + x_2)^6$$

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consider fixed point $\mathbf{x}_c = \mathbf{0} = [0, 0]$

$$D\mathbf{f}(\mathbf{0}) = \begin{pmatrix} -1 & 1 \\ 1 & -1 \end{pmatrix}, \quad E^c = \text{span} \left\{ \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right\}, \quad E^s = \text{span} \left\{ \begin{pmatrix} 1 \\ -1 \end{pmatrix} \right\}$$

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new variables: $y_1 = x_1 + x_2$ and $y_2 = x_1 - x_2$

$$\frac{dy_1}{dt} = -2y_1y_2 - 2y_1^6 \qquad \frac{dy_2}{dt} = -2y_2 + 2y_1^2$$

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Warning: On the center linear subspace, we have $y_2 = 0$. Substituting $y_2 = 0$ into the first equation gives $dy_1/dt = -2y_1^6$, but this does not mean that the origin is unstable! In fact, the center manifold can be calculated as $y_2 = h(y_1) = y_1^2 + \mathcal{O}(y_1^4)$, which implies that the dynamics on the center manifold is

$$\frac{dy_1}{dt} = -2y_1h(y_1) - 2y_1^6 = -2y_1^3 + \mathcal{O}(y_1^5) \quad \text{which implies that the origin is stable.}$$

B5.6: Nonlinear Dynamics, Bifurcations and Chaos (Lecture 5)

- summary of Lecture 4: we concluded our discussion of Problem Sheet 1
- today: we will start with our discussion of Problem Sheet 2 (covered in Lectures 5–8)
- course synopsis of **Lectures 1-8** (taken by both Part B and MSc students):
Discrete-time (maps) and continuous-time (differential equations) dynamical systems. Notion of flows, stability of fixed points, Lyapunov function, invariant manifolds, stable manifold theorem, notion of hyperbolicity, center manifold. Chemical reaction networks. Stable, unstable and center subspaces. Poincaré-Bendixson theorem. Periodic solutions, stable and unstable limit cycles. [Introduction to bifurcation theory, covering saddle-node, transcritical, supercritical pitchfork and subcritical pitchfork bifurcations.](#) Extended center manifold. Logistic map. Periodic points of maps. Stability of N-cycles. Period-doubling bifurcation. Sharkovsky's theorem. Invariant distribution.

Bifurcations

Continuous-time dynamical system: Let $\mathbf{f} : \Omega \times \Theta \rightarrow \mathbb{R}^n$, where $\Omega \subset \mathbb{R}^n$ and $\Theta \subset \mathbb{R}^m$. Let $\mathbf{x}_0 \in \Omega$, $\boldsymbol{\mu} \in \Theta$ and $\mathbf{x}(t) \in \Omega$ be a solution of the ODE

$$\frac{d\mathbf{x}}{dt} = \mathbf{f}(\mathbf{x}; \boldsymbol{\mu}) \quad \text{with the initial condition} \quad \mathbf{x}(0) = \mathbf{x}_0 \in \Omega$$

Bifurcations: The qualitative structure of the flow can change as parameters $\boldsymbol{\mu}$ are varied. For example, critical points (fixed points) can be created or destroyed, or limit cycles can be created or destroyed. The parameter values at which these qualitative changes in the dynamics occur are called bifurcation points.

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Some bifurcations can occur for $n = 1$, so we start with them.

- saddle-node bifurcation
- transcritical bifurcation
- supercritical pitchfork bifurcation
- subcritical pitchfork bifurcation

Saddle-node bifurcation

example: $\frac{dx_1}{dt} = \mu + x_1^2$

$$f(x_1; \mu) = \mu + x_1^2$$

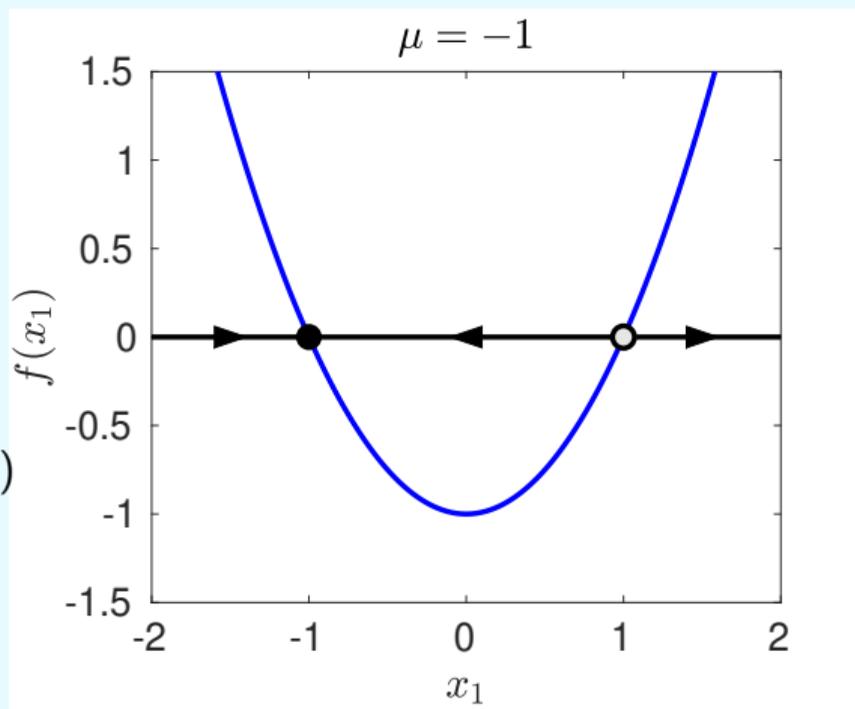
Saddle-node bifurcation

example: $\frac{dx_1}{dt} = \mu + x_1^2$

$$f(x_1; \mu) = \mu + x_1^2$$

$$\mu < 0$$

two fixed points at $x_1 = -\sqrt{-\mu}$ (stable)
and $x_1 = \sqrt{-\mu}$ (unstable)



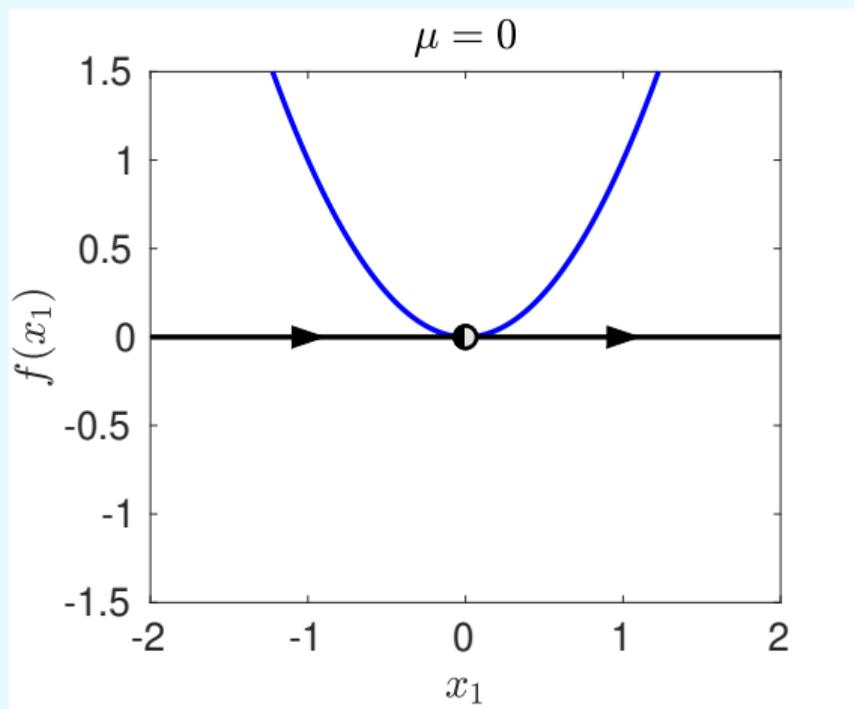
Saddle-node bifurcation

example: $\frac{dx_1}{dt} = \mu + x_1^2$

$$f(x_1; \mu) = \mu + x_1^2$$

as μ approaches zero from below,
the two fixed points $-\sqrt{-\mu}$ and $\sqrt{-\mu}$
move toward each other

$\mu = 0$: the fixed points coalesce into
a half-stable fixed point at $x_1 = 0$

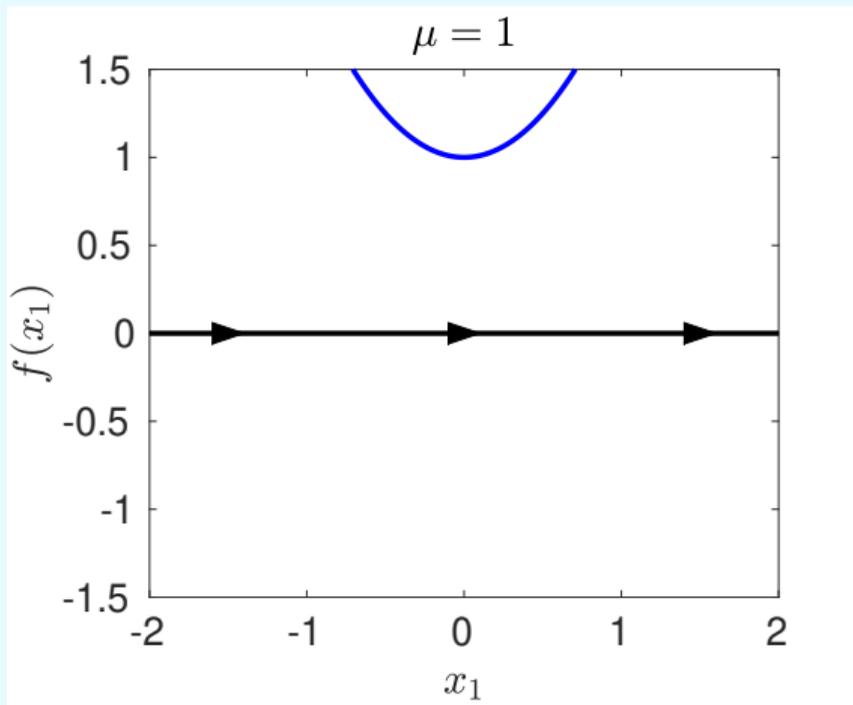


Saddle-node bifurcation

example: $\frac{dx_1}{dt} = \mu + x_1^2$

$$f(x_1; \mu) = \mu + x_1^2$$

$\mu > 0$: no fixed points



Saddle-node bifurcation

example: $\frac{dx_1}{dt} = \mu + x_1^2$

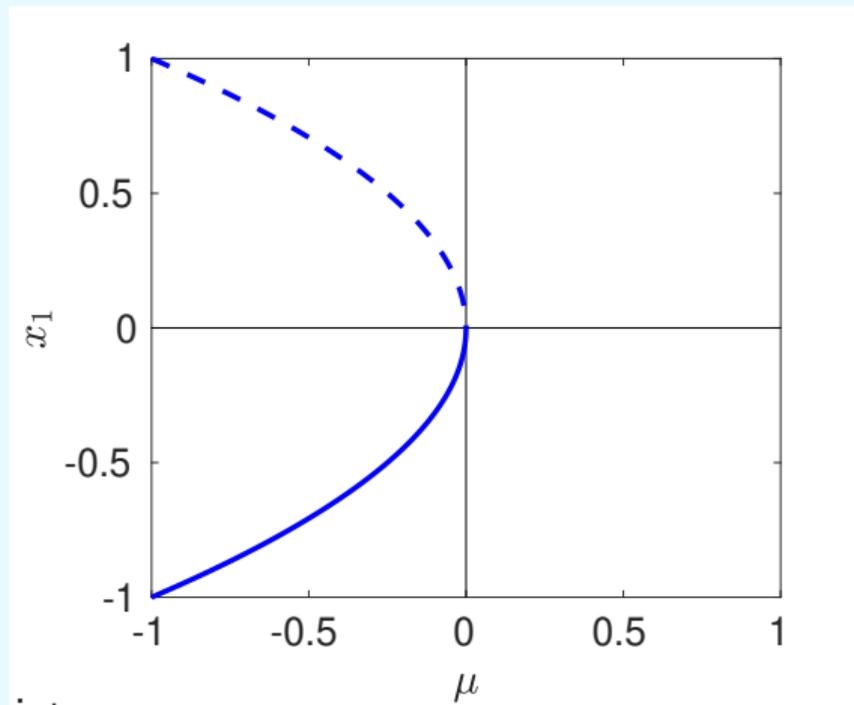
bifurcation diagram

saddle node bifurcation is a simple mechanism by which critical points can be created or destroyed

terminology:

critical point: fixed point, equilibrium point

saddle-node bifurcation: fold bifurcation, turning point bifurcation

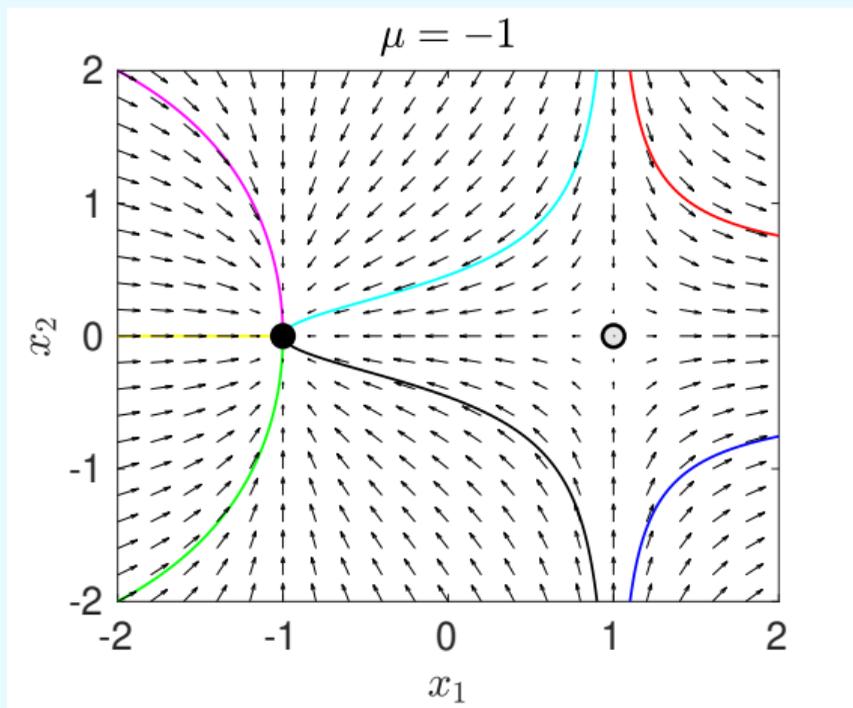


Saddle-node bifurcation

example:
$$\frac{dx_1}{dt} = \mu + x_1^2$$
$$\frac{dx_2}{dt} = -x_2$$

$$\mu < 0$$

two fixed points at
 $\mathbf{x} = [-\sqrt{-\mu}, 0]$ stable node
and $\mathbf{x} = [\sqrt{-\mu}, 0]$ saddle

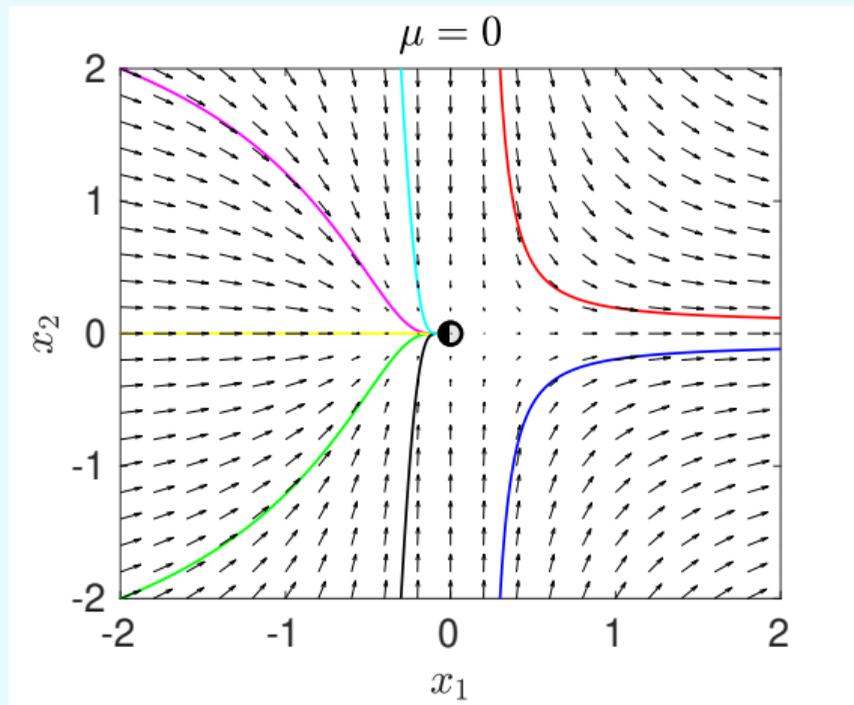


Saddle-node bifurcation

example:
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as μ approaches zero from below, the two fixed points $[-\sqrt{-\mu}, 0]$ and $[\sqrt{-\mu}, 0]$ move toward each other

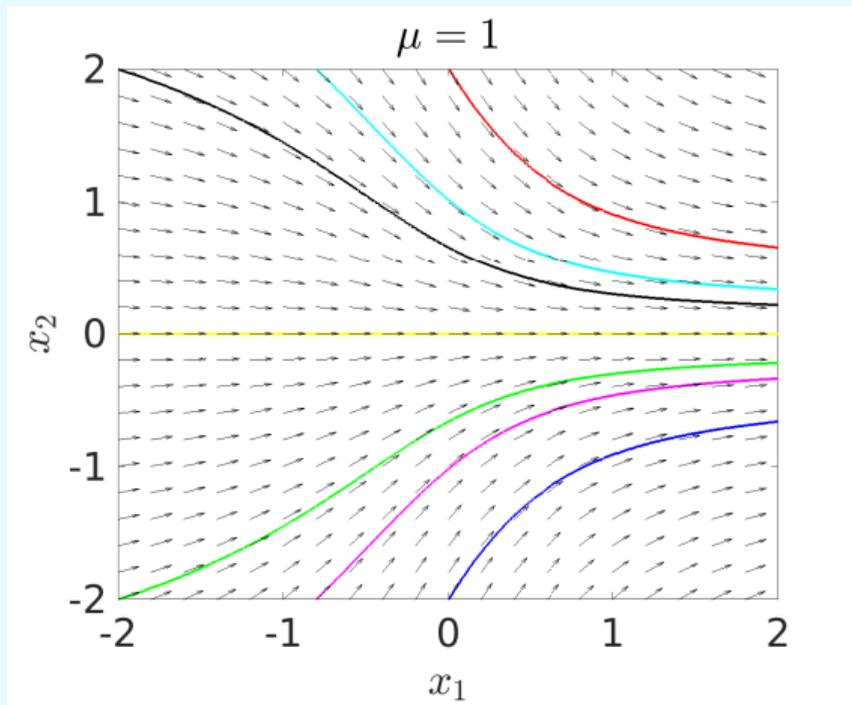
$\mu = 0$: the fixed points coalesce into a (saddle-node) fixed point at $\mathbf{x} = [0, 0]$



Saddle-node bifurcation

example:
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$$\frac{dx_2}{dt} = -x_2$$

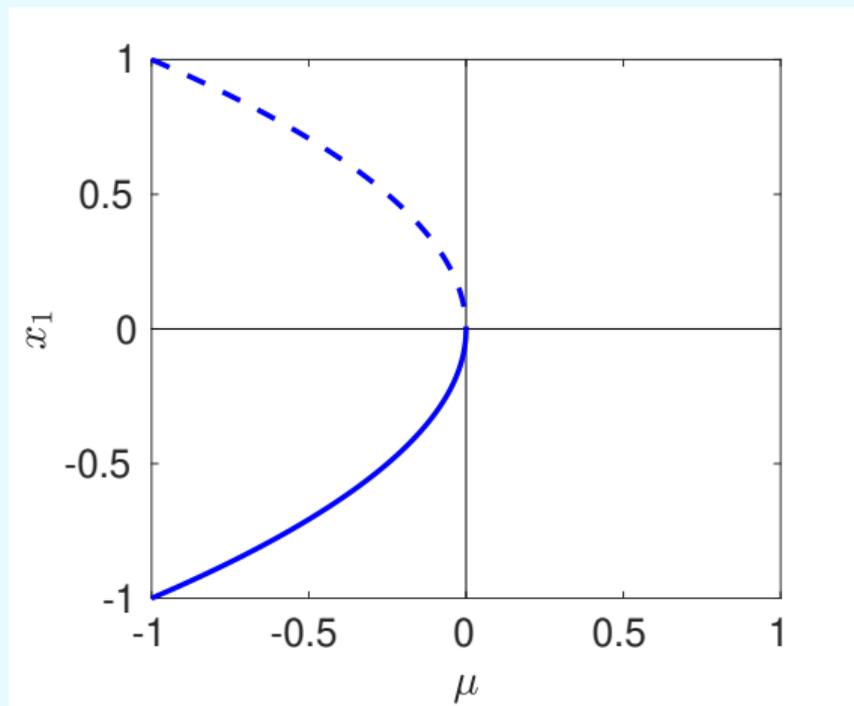
$\mu > 0$: no fixed points



Saddle-node bifurcation

example: $\frac{dx_1}{dt} = \mu + x_1^2$
 $\frac{dx_2}{dt} = -x_2$

bifurcation diagram



Saddle-node bifurcation

example: $\frac{dx_1}{dt} = \mu + x_1^2$

$$f(x_1; \mu) = \mu + x_1^2$$

bifurcation diagram

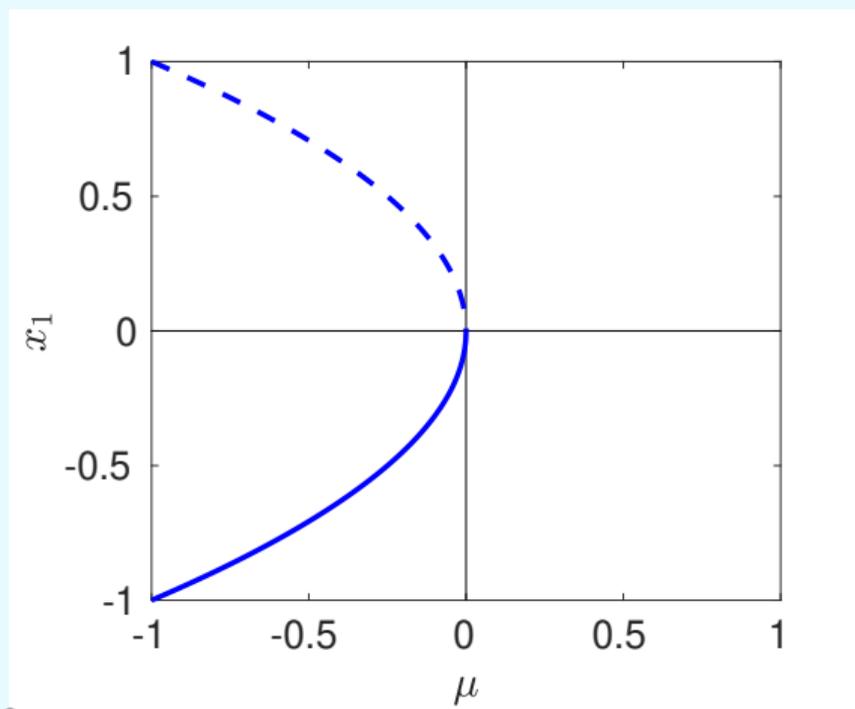
saddle node bifurcation is a general mechanism by which critical points can be created or destroyed

if it occurs at $x_1 = x_c$ and $\mu = \mu_c$, we have

$$f(x_c; \mu_c) = 0 \text{ and } \frac{\partial f}{\partial x_1}(x_c; \mu_c) = 0$$

Taylor expansion:

$$f(x_1; \mu) = (\mu - \mu_c) \frac{\partial f}{\partial \mu}(x_c; \mu_c) + (x_1 - x_c)^2 \frac{1}{2} \frac{\partial^2 f}{\partial x_1^2}(x_c; \mu_c) + \dots \quad (\text{normal form})$$



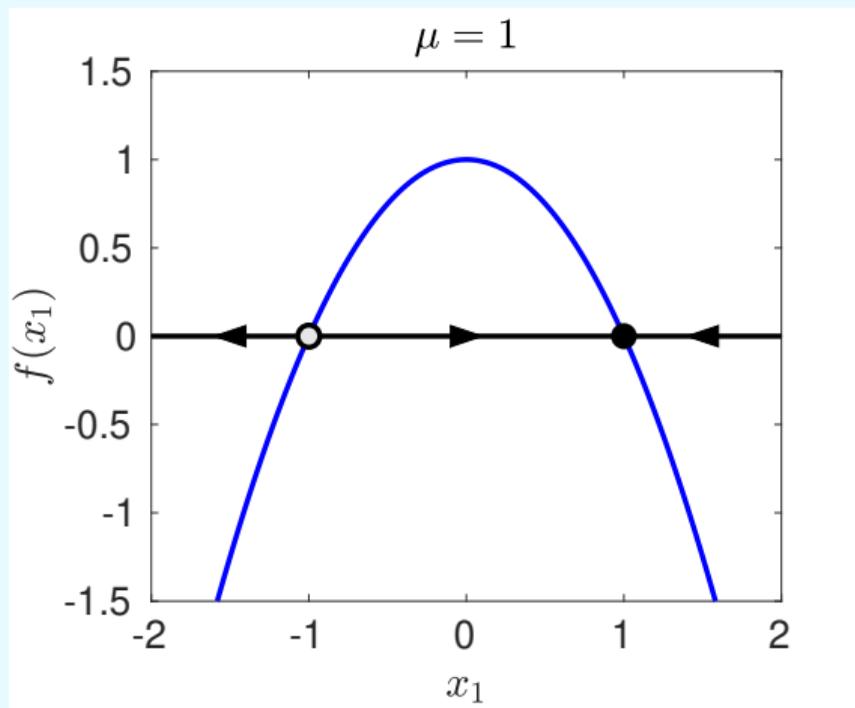
Saddle-node bifurcation

example: $\frac{dx_1}{dt} = \mu - x_1^2$

$$f(x_1; \mu) = \mu - x_1^2$$

$$\mu > 0$$

two fixed points at $x_1 = \sqrt{\mu}$ (stable)
and $x_1 = -\sqrt{\mu}$ (unstable)



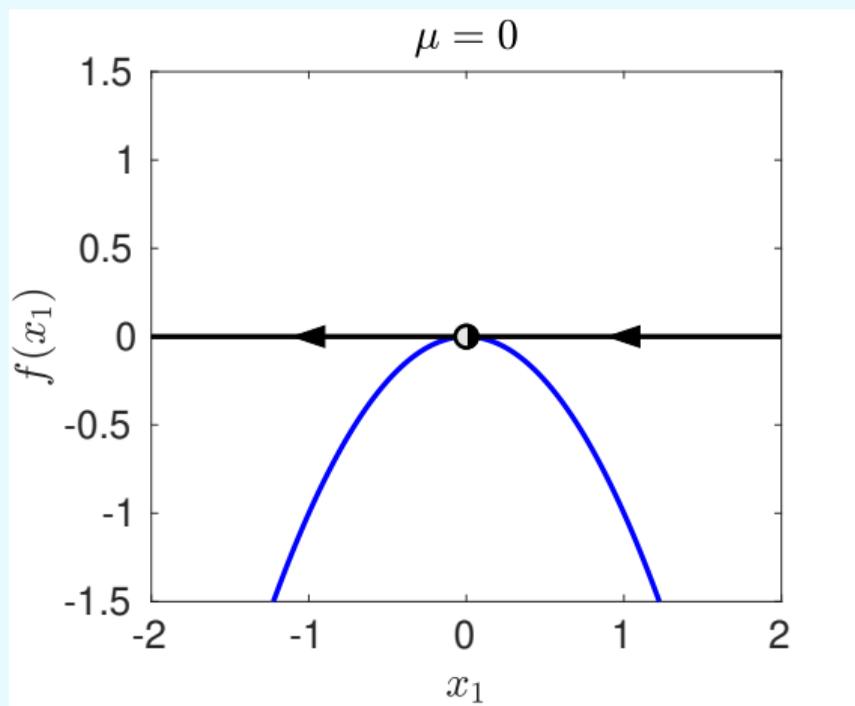
Saddle-node bifurcation

example: $\frac{dx_1}{dt} = \mu - x_1^2$

$$f(x_1; \mu) = \mu - x_1^2$$

as μ approaches zero from above, the two fixed points $-\sqrt{\mu}$ and $\sqrt{\mu}$ move toward each other

$\mu = 0$: the fixed points coalesce into a half-stable fixed point at $x_1 = 0$

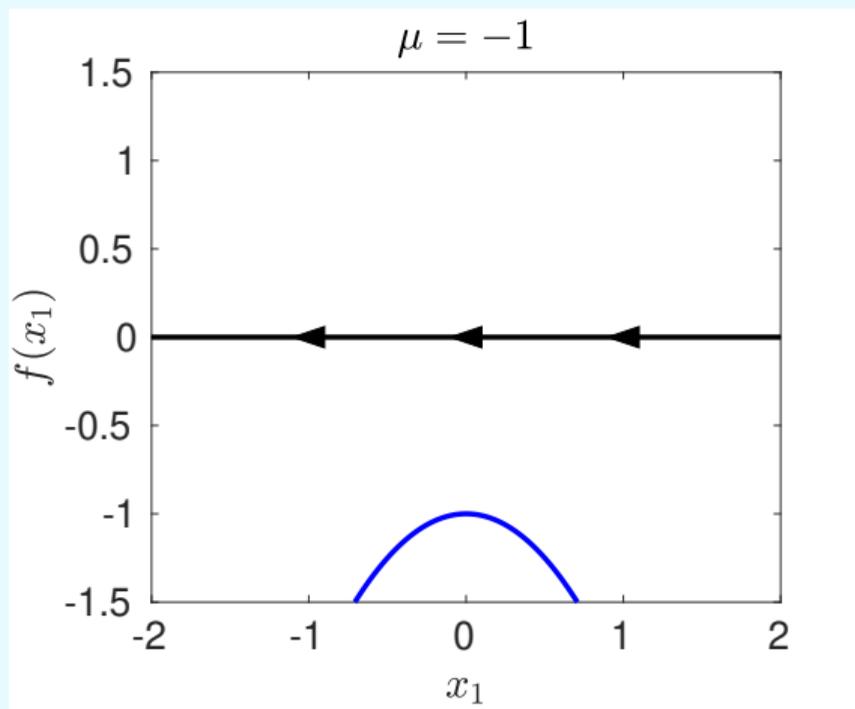


Saddle-node bifurcation

example: $\frac{dx_1}{dt} = \mu - x_1^2$

$$f(x_1; \mu) = \mu - x_1^2$$

$\mu < 0$: no fixed points



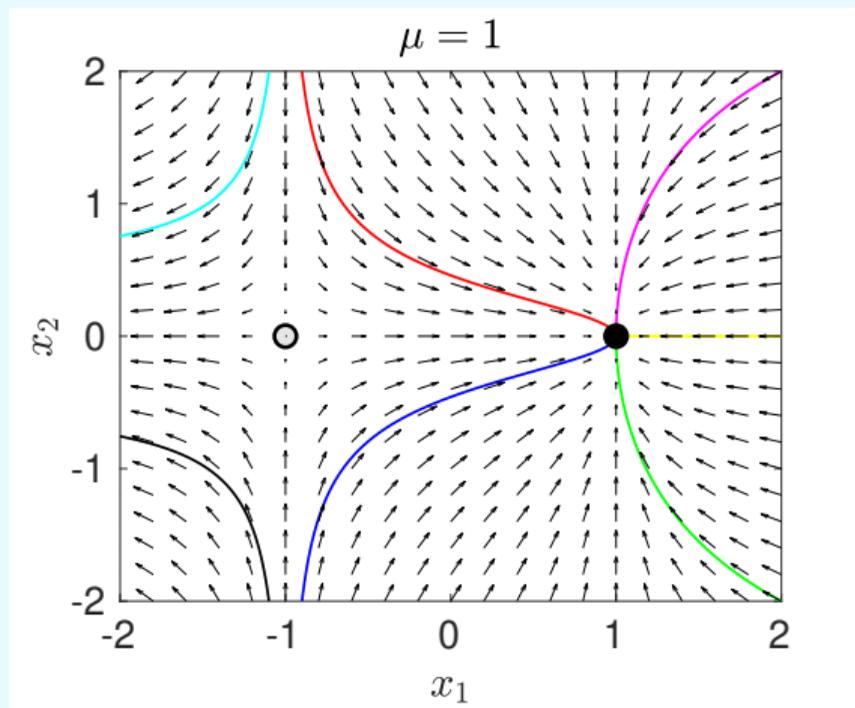
Saddle-node bifurcation

example:
$$\frac{dx_1}{dt} = \mu - x_1^2$$
$$\frac{dx_2}{dt} = -x_2$$

$$\mu > 0$$

two fixed points at
 $\mathbf{x} = [-\sqrt{\mu}, 0]$ saddle
and

$\mathbf{x} = [\sqrt{\mu}, 0]$ stable node

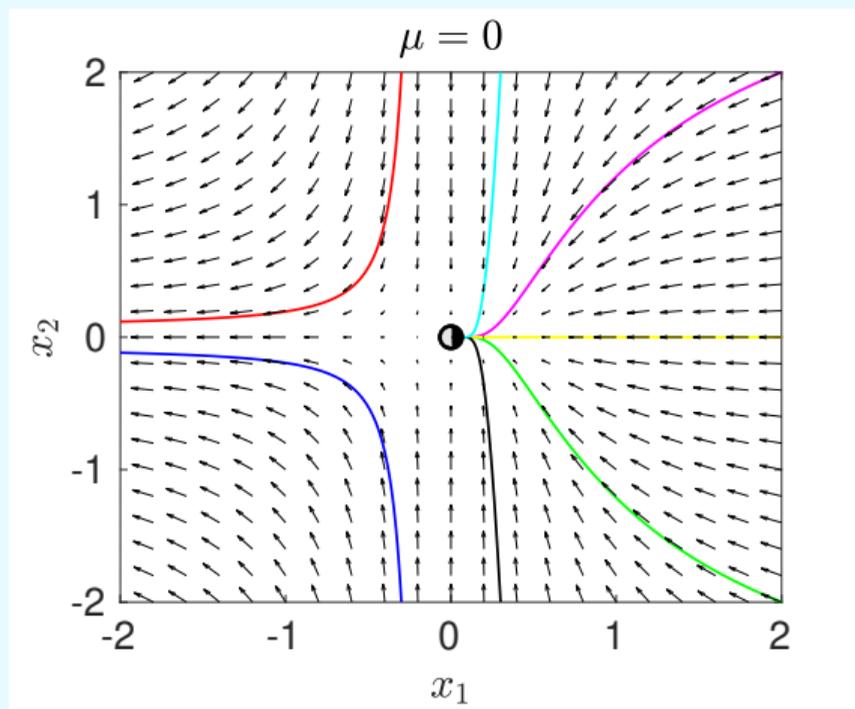


Saddle-node bifurcation

example:
$$\frac{dx_1}{dt} = \mu - x_1^2$$
$$\frac{dx_2}{dt} = -x_2$$

as μ approaches zero from above, the two fixed points $[-\sqrt{\mu}, 0]$ and $[\sqrt{\mu}, 0]$ move toward each other

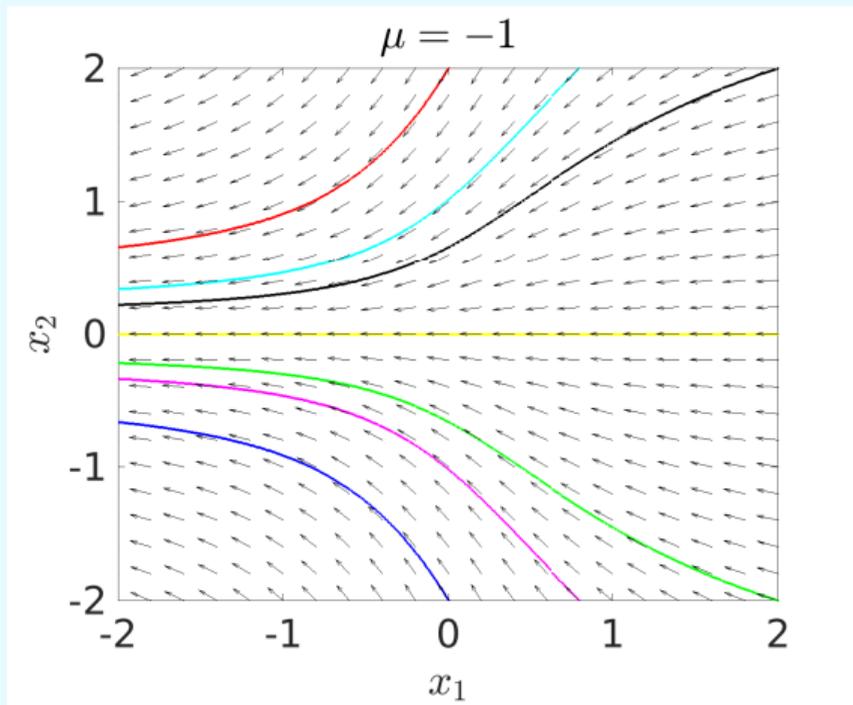
$\mu = 0$: the fixed points coalesce into a (saddle-node) fixed point at $\mathbf{x} = [0, 0]$



Saddle-node bifurcation

example:
$$\frac{dx_1}{dt} = \mu - x_1^2$$
$$\frac{dx_2}{dt} = -x_2$$

$\mu < 0$: no fixed points



Saddle-node bifurcation

example: $\frac{dx_1}{dt} = \mu - x_1^2$

$$f(x_1; \mu) = \mu - x_1^2$$

bifurcation diagram

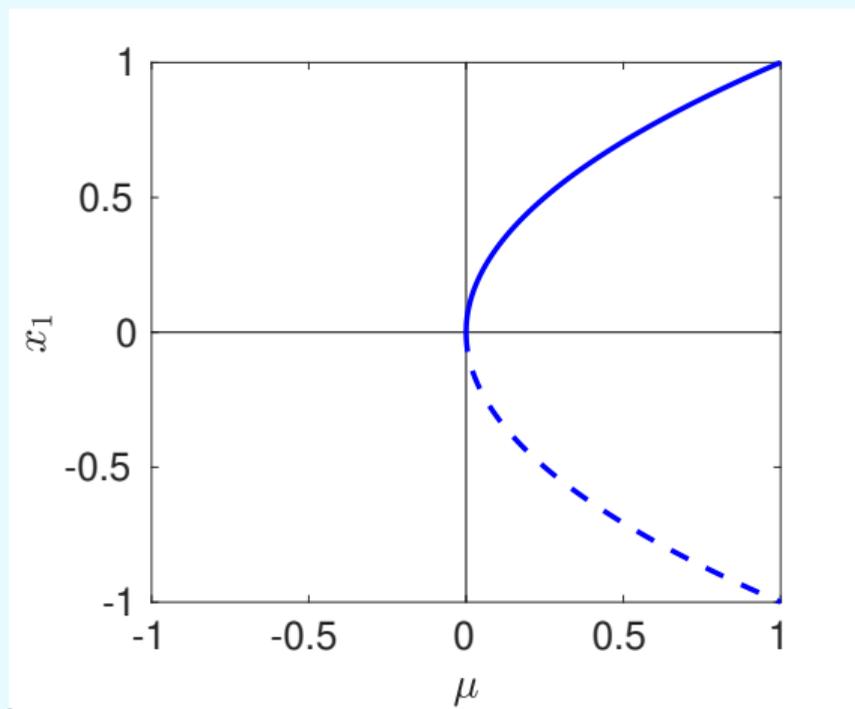
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Taylor expansion:

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Transcritical bifurcation

example: $\frac{dx_1}{dt} = \mu x_1 - x_1^2$

$$f(x_1; \mu) = \mu x_1 - x_1^2$$

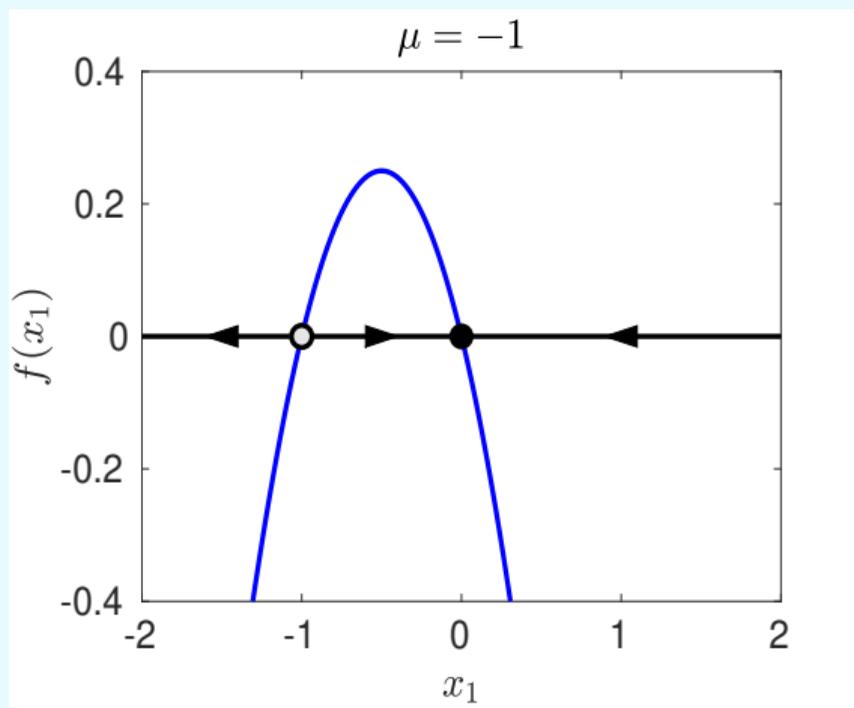
Transcritical bifurcation

example: $\frac{dx_1}{dt} = \mu x_1 - x_1^2$

$$f(x_1; \mu) = \mu x_1 - x_1^2$$

$$\mu < 0$$

two fixed points at $x_1 = \mu$ (**unstable**)
and $x_1 = 0$ (**stable**)



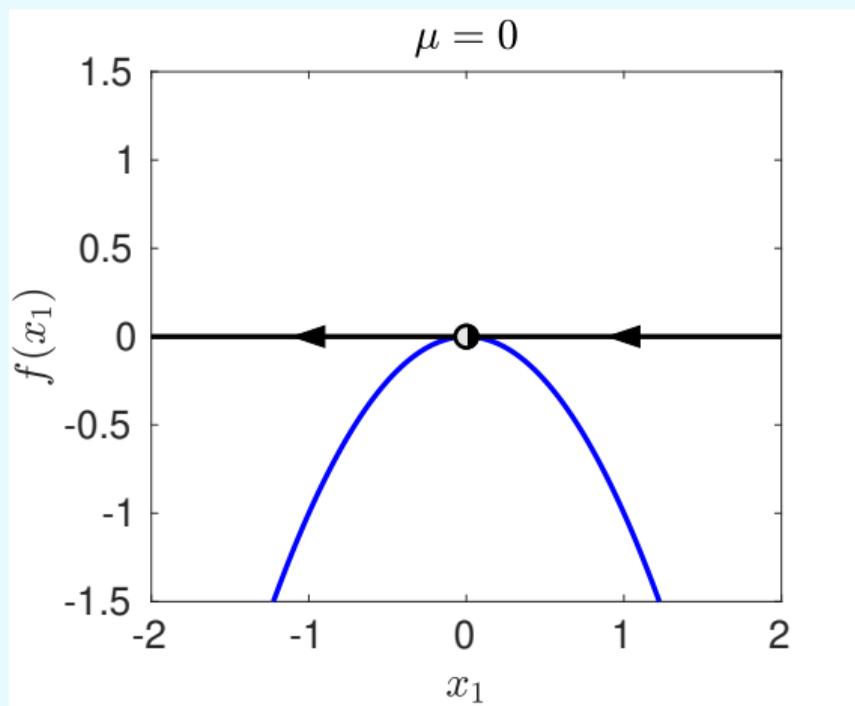
Transcritical bifurcation

example: $\frac{dx_1}{dt} = \mu x_1 - x_1^2$

$$f(x_1; \mu) = \mu x_1 - x_1^2$$

as μ approaches zero,
the two fixed points μ and 0
move toward each other

$\mu = 0$: the fixed points coalesce into
a half-stable fixed point at $x_1 = 0$



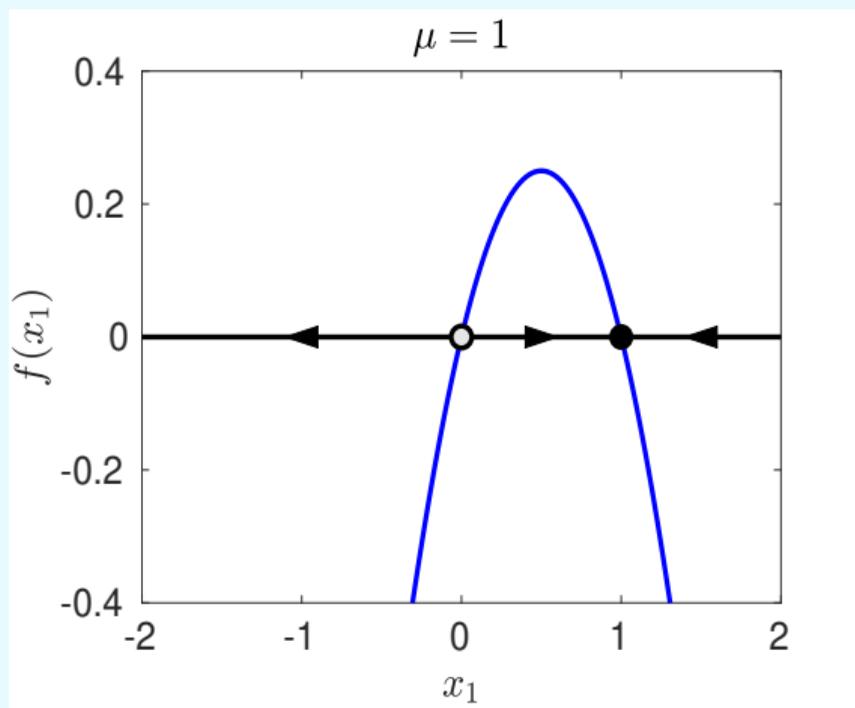
Transcritical bifurcation

example: $\frac{dx_1}{dt} = \mu x_1 - x_1^2$

$$f(x_1; \mu) = \mu x_1 - x_1^2$$

$$\mu > 0$$

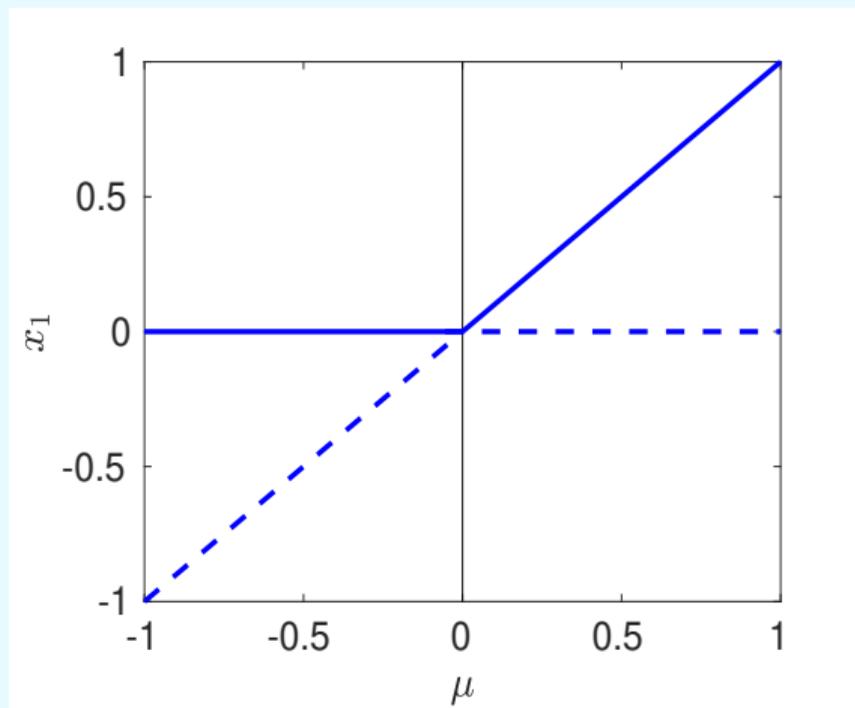
two fixed points at $x_1 = \mu$ (**stable**)
and $x_1 = 0$ (**unstable**)



Transcritical bifurcation

example: $\frac{dx_1}{dt} = \mu x_1 - x_1^2$

bifurcation diagram



Transcritical bifurcation

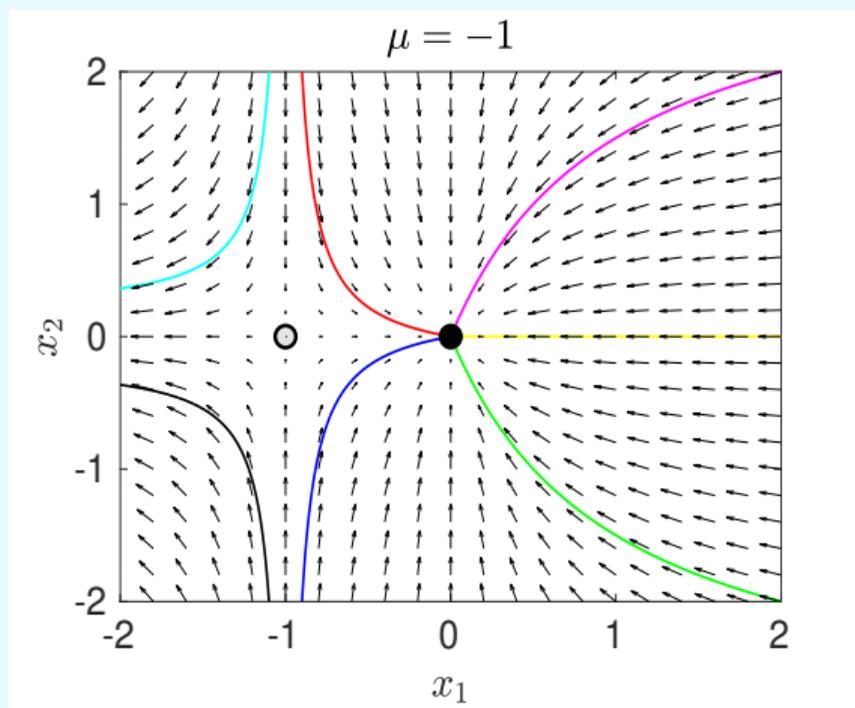
example:
$$\frac{dx_1}{dt} = \mu x_1 - x_1^2$$
$$\frac{dx_2}{dt} = -x_2$$

$$\mu < 0$$

two fixed points at

$\mathbf{x} = [\mu, 0]$ saddle (unstable)

and $\mathbf{x} = [0, 0]$ stable node

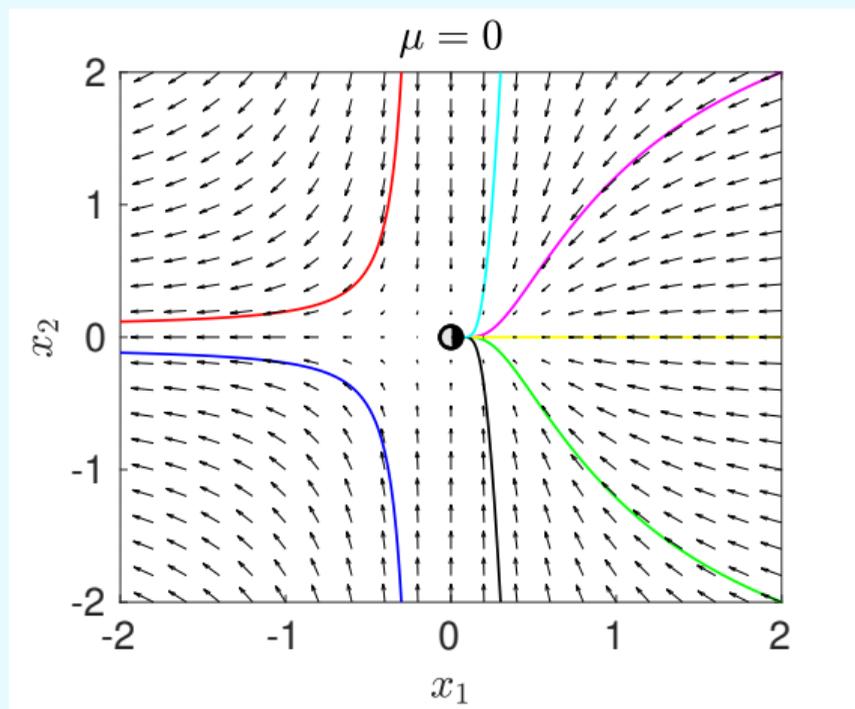


Transcritical bifurcation

example:
$$\frac{dx_1}{dt} = \mu x_1 - x_1^2$$
$$\frac{dx_2}{dt} = -x_2$$

as μ approaches zero,
the two fixed points $[\mu, 0]$ and
 $[0, 0]$ move toward each other

$\mu = 0$: the fixed points coalesce into
a (saddle-node) fixed point at $\mathbf{x} = [0, 0]$



Transcritical bifurcation

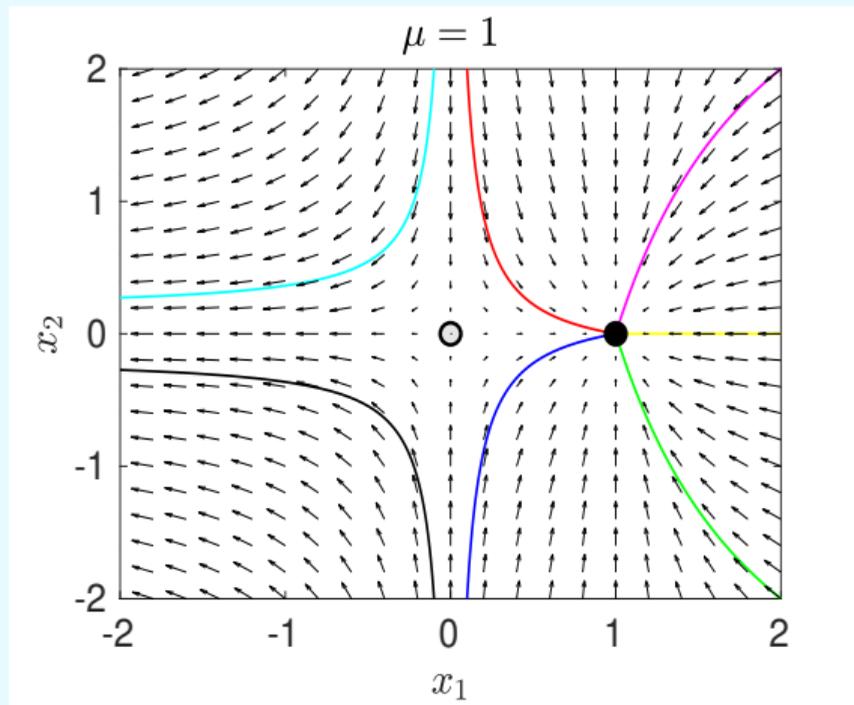
example:
$$\frac{dx_1}{dt} = \mu x_1 - x_1^2$$
$$\frac{dx_2}{dt} = -x_2$$

$\mu > 0$

two fixed points at

$\mathbf{x} = [\mu, 0]$ **stable node**

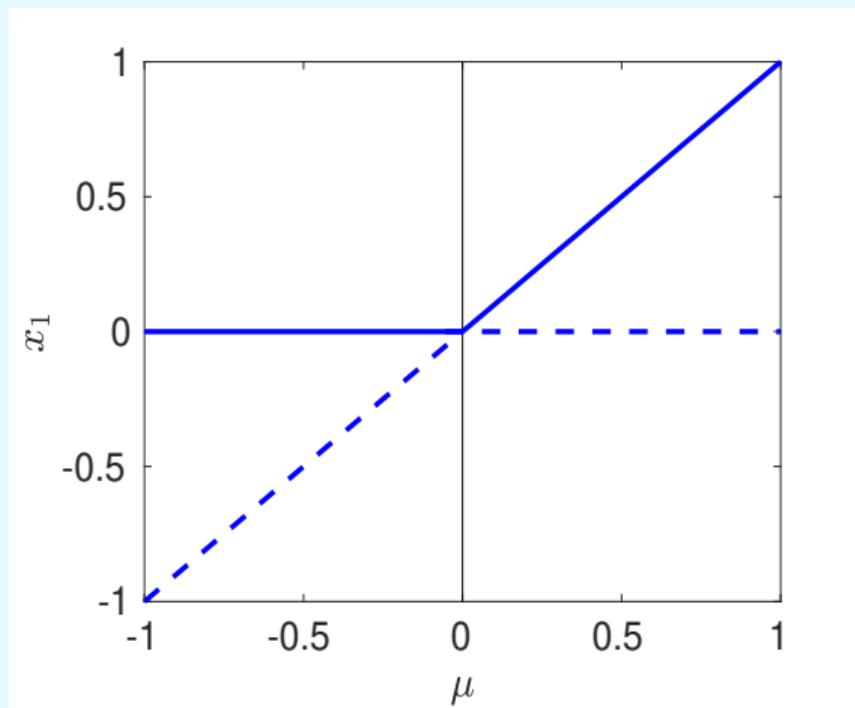
and $\mathbf{x} = [0, 0]$ **saddle (unstable)**



Transcritical bifurcation

example:
$$\frac{dx_1}{dt} = \mu x_1 - x_1^2$$
$$\frac{dx_2}{dt} = -x_2$$

bifurcation diagram

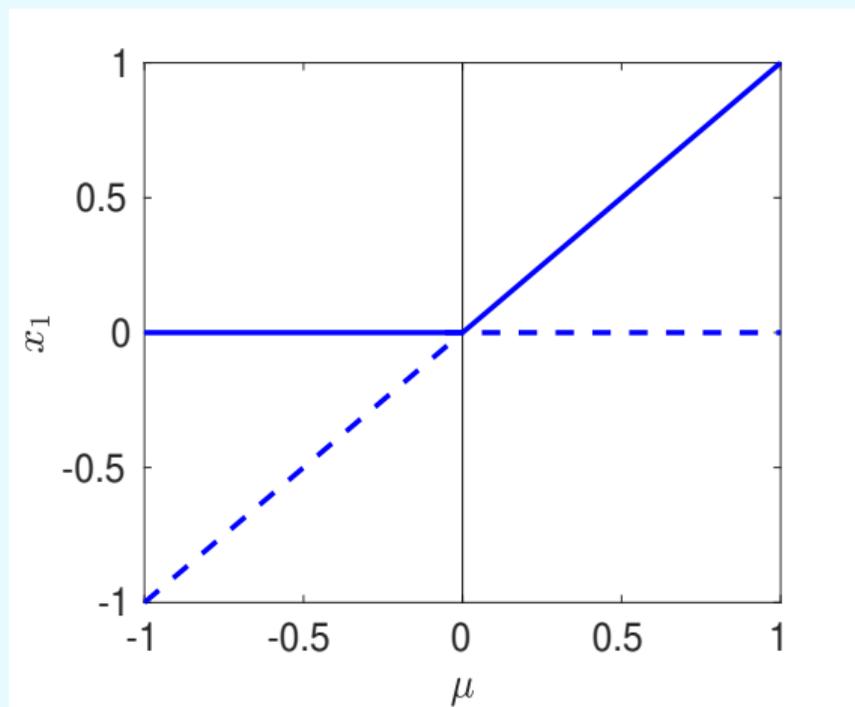


Transcritical bifurcation

example: $\frac{dx_1}{dt} = \mu x_1 - x_1^2$

$$f(x_1; \mu) = \mu x_1 - x_1^2$$

bifurcation diagram



Other examples of ODE systems with bifurcations:

Questions 1, 2, 5 and 6 on Problem Sheet 2

Supercritical pitchfork bifurcation

example: $\frac{dx_1}{dt} = \mu x_1 - x_1^3$

$$f(x_1; \mu) = \mu x_1 - x_1^3$$

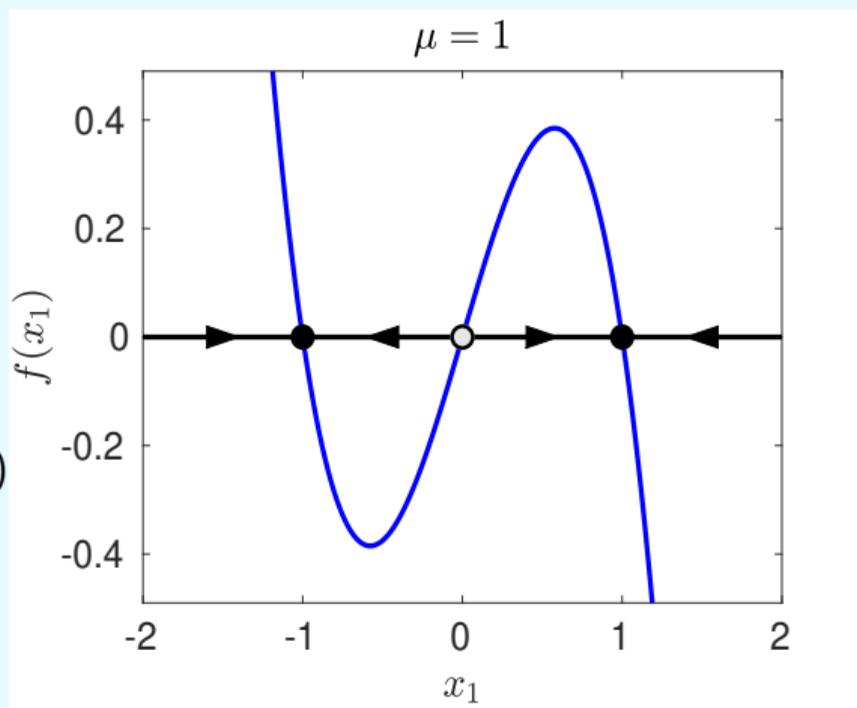
Supercritical pitchfork bifurcation

example: $\frac{dx_1}{dt} = \mu x_1 - x_1^3$

$$f(x_1; \mu) = \mu x_1 - x_1^3$$

$$\mu > 0$$

three fixed points at $x_1 = \pm\sqrt{\mu}$ (**stable**)
and $x_1 = 0$ (**unstable**)



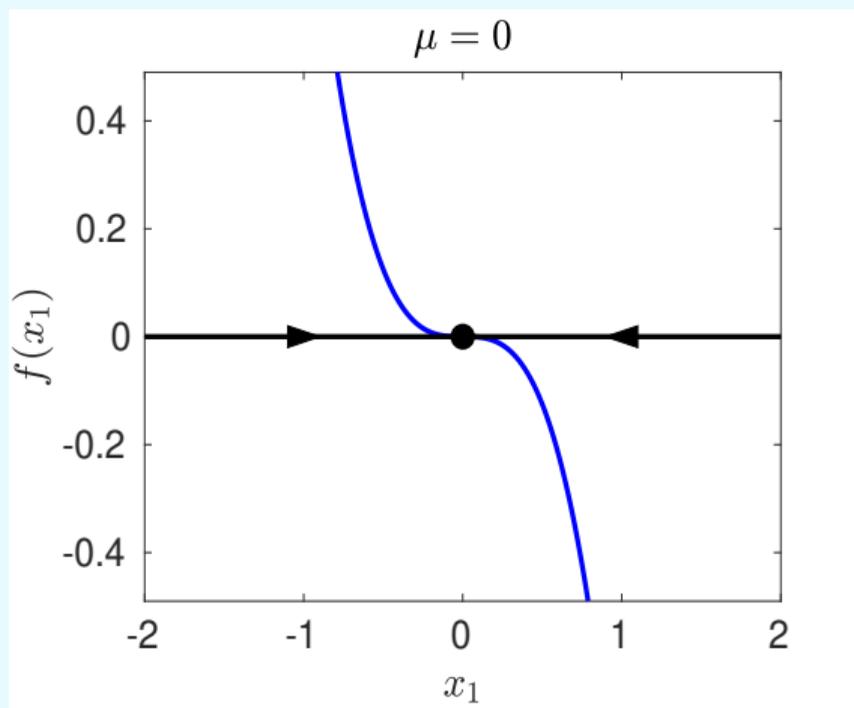
Supercritical pitchfork bifurcation

example: $\frac{dx_1}{dt} = \mu x_1 - x_1^3$

$$f(x_1; \mu) = \mu x_1 - x_1^3$$

as μ approaches zero from above,
two fixed points $\sqrt{\mu}$ and $-\sqrt{\mu}$
move toward the third one

$\mu = 0$: the fixed points coalesce into
a stable fixed point at $x_1 = 0$

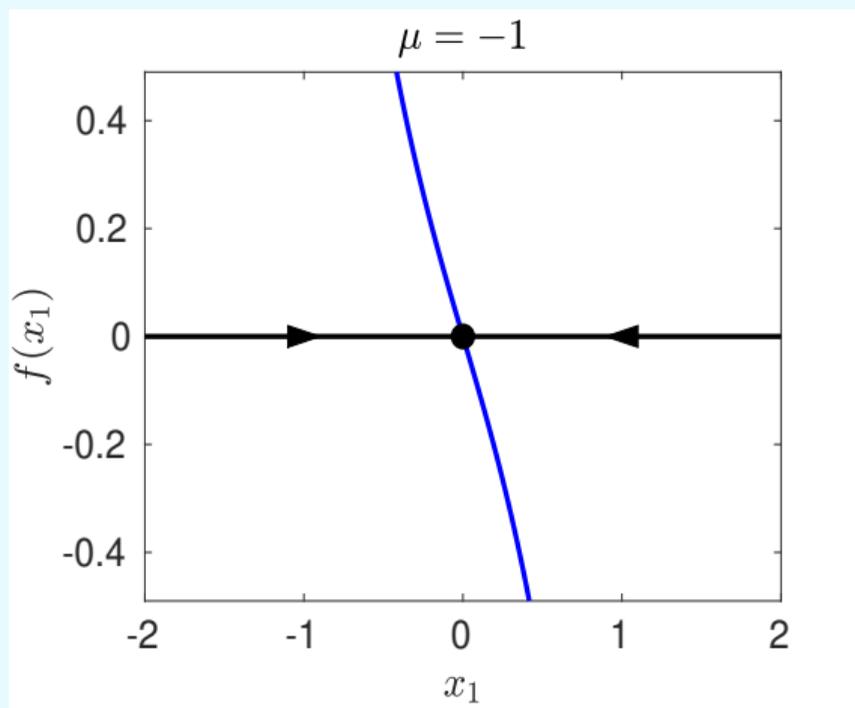


Supercritical pitchfork bifurcation

example: $\frac{dx_1}{dt} = \mu x_1 - x_1^3$

$$f(x_1; \mu) = \mu x_1 - x_1^3$$

$\mu < 0$: one stable fixed point at $x_1 = 0$

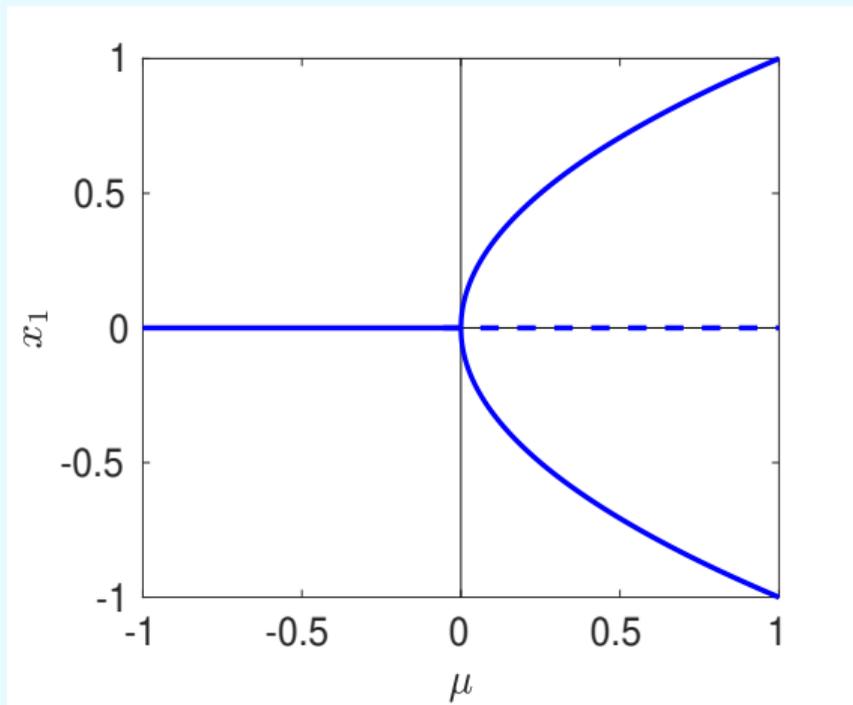


Supercritical pitchfork bifurcation

example: $\frac{dx_1}{dt} = \mu x_1 - x_1^3$

$$f(x_1; \mu) = \mu x_1 - x_1^3$$

bifurcation diagram



Supercritical pitchfork bifurcation

example:
$$\frac{dx_1}{dt} = \mu x_1 - x_1^3$$
$$\frac{dx_2}{dt} = -x_2$$

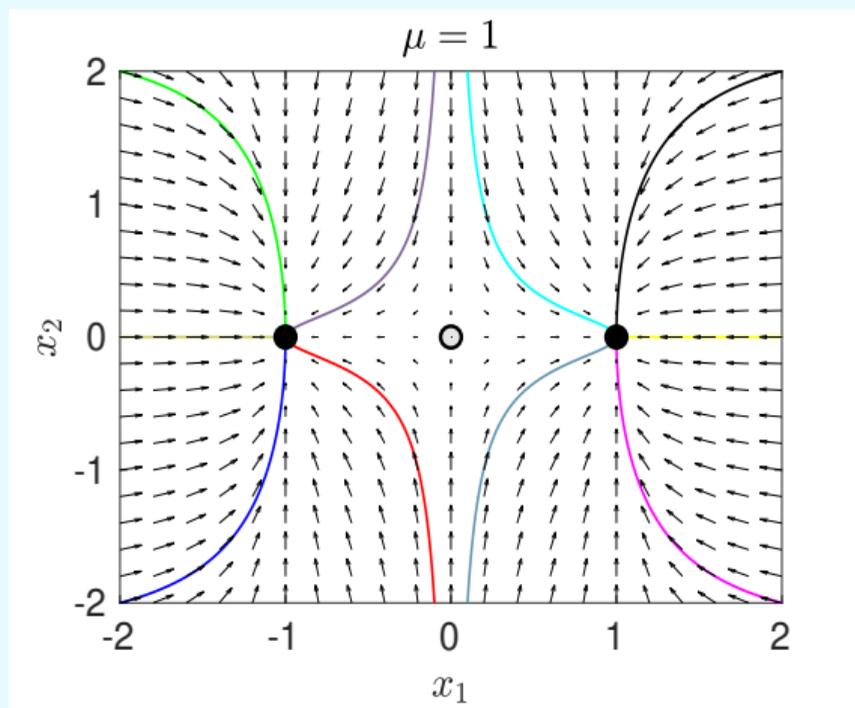
$\mu > 0$

three fixed points at

$\mathbf{x} = [-\sqrt{\mu}, 0]$ (stable node)

$\mathbf{x} = [0, 0]$ (saddle)

$\mathbf{x} = [\sqrt{\mu}, 0]$ (stable node)

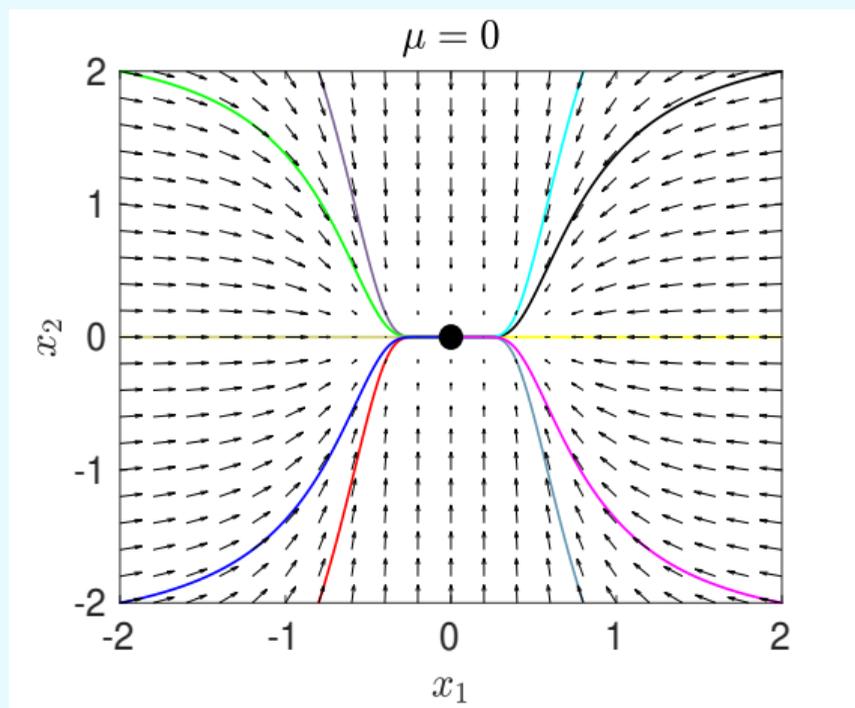


Supercritical pitchfork bifurcation

example:
$$\frac{dx_1}{dt} = \mu x_1 - x_1^3$$
$$\frac{dx_2}{dt} = -x_2$$

as μ approaches zero from above,
two fixed points $[-\sqrt{\mu}, 0]$ and $[\sqrt{\mu}, 0]$
move toward the third one

$\mu = 0$: the fixed points coalesce into
a stable fixed point at $\mathbf{x} = [0, 0]$

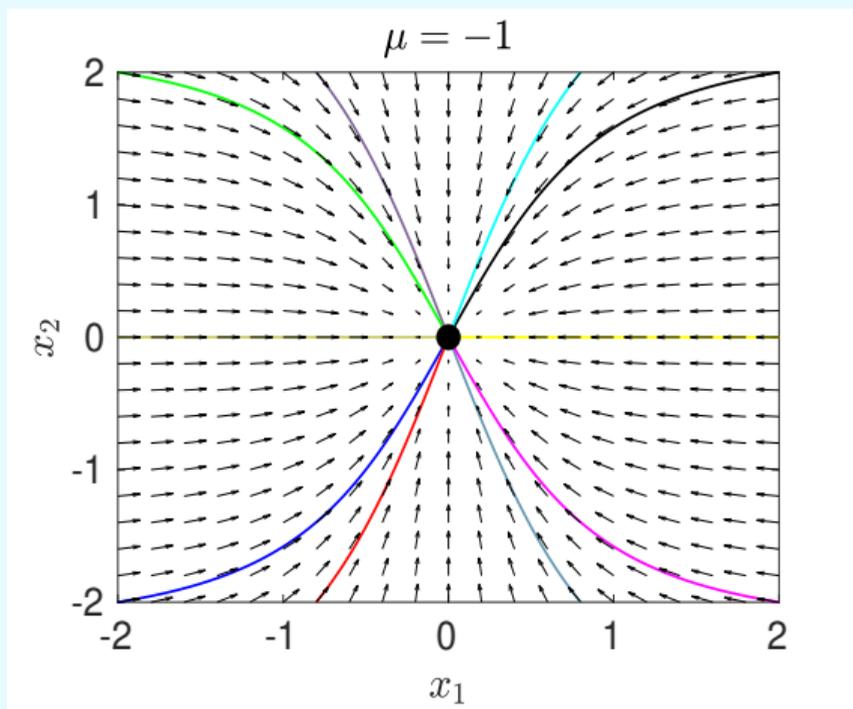


Supercritical pitchfork bifurcation

example:
$$\frac{dx_1}{dt} = \mu x_1 - x_1^3$$
$$\frac{dx_2}{dt} = -x_2$$

$\mu < 0$:

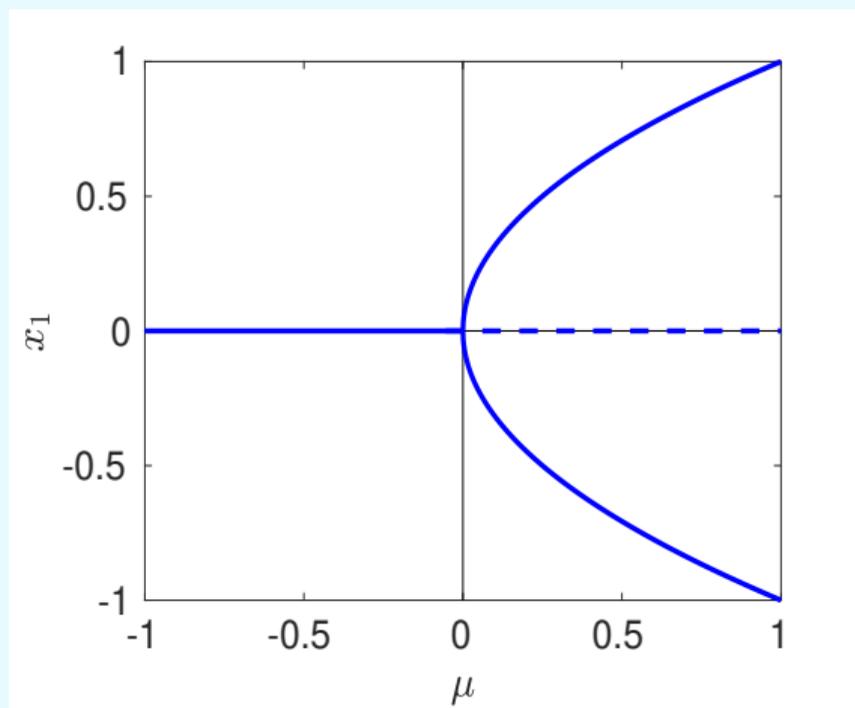
one stable fixed point at $\mathbf{x} = [0, 0]$



Supercritical pitchfork bifurcation

example:
$$\frac{dx_1}{dt} = \mu x_1 - x_1^3$$
$$\frac{dx_2}{dt} = -x_2$$

bifurcation diagram



Subcritical pitchfork bifurcation

example: $\frac{dx_1}{dt} = \mu x_1 + x_1^3$

$$f(x_1; \mu) = \mu x_1 + x_1^3$$

Subcritical pitchfork bifurcation

example: $\frac{dx_1}{dt} = \mu x_1 + x_1^3$

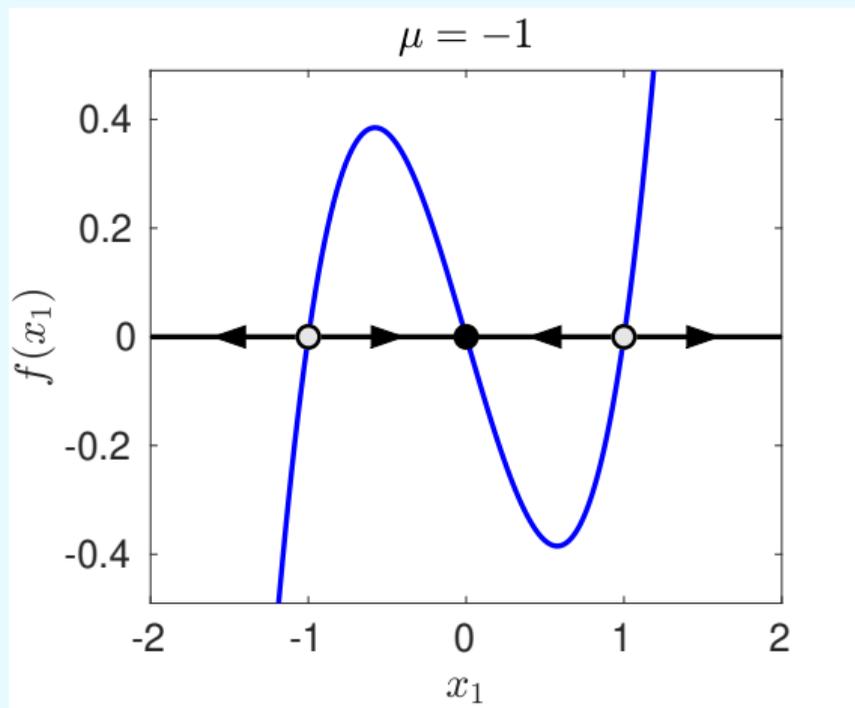
$$f(x_1; \mu) = \mu x_1 + x_1^3$$

$$\mu < 0$$

three fixed points at

$$x_1 = \pm\sqrt{-\mu} \text{ (unstable)}$$

and $x_1 = 0$ (stable)



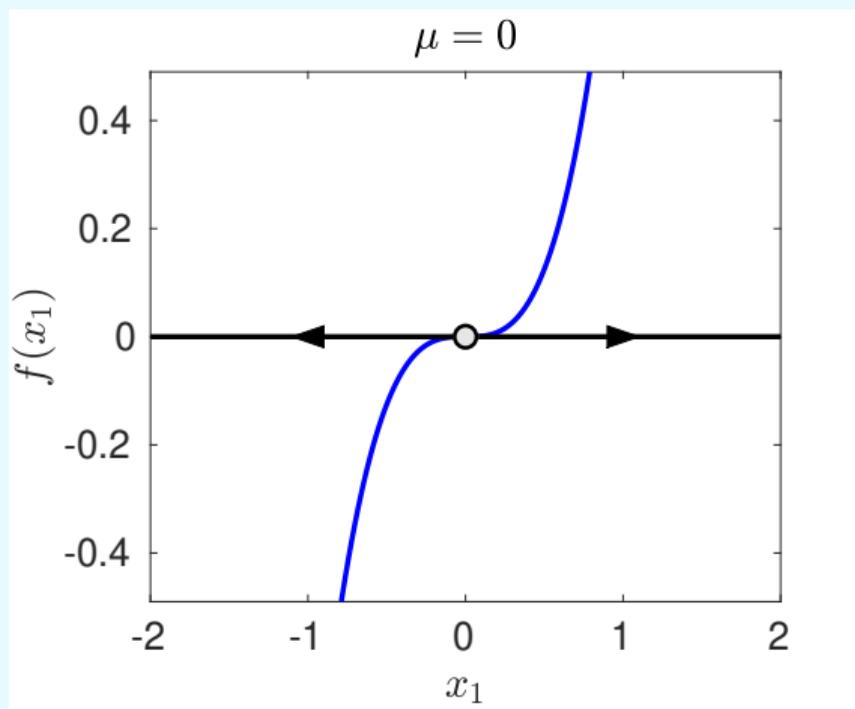
Subcritical pitchfork bifurcation

example: $\frac{dx_1}{dt} = \mu x_1 + x_1^3$

$$f(x_1; \mu) = \mu x_1 + x_1^3$$

as μ approaches zero from below,
two fixed points $-\sqrt{-\mu}$ and $\sqrt{-\mu}$
move toward the third one

$\mu = 0$: the fixed points coalesce into
an unstable fixed point at $x_1 = 0$



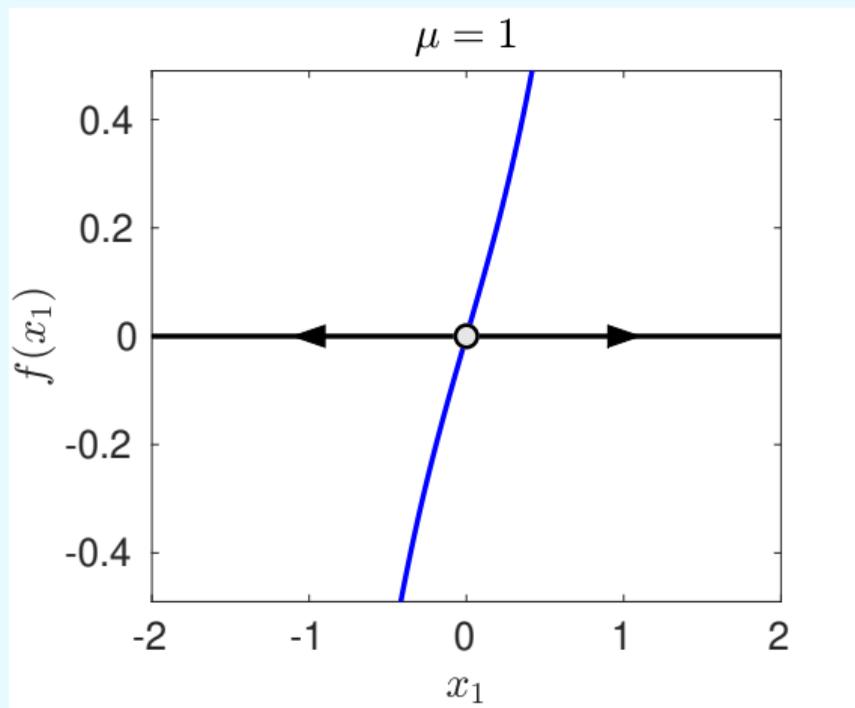
Subcritical pitchfork bifurcation

example: $\frac{dx_1}{dt} = \mu x_1 + x_1^3$

$$f(x_1; \mu) = \mu x_1 + x_1^3$$

$\mu > 0$:

one unstable fixed point at $x_1 = 0$

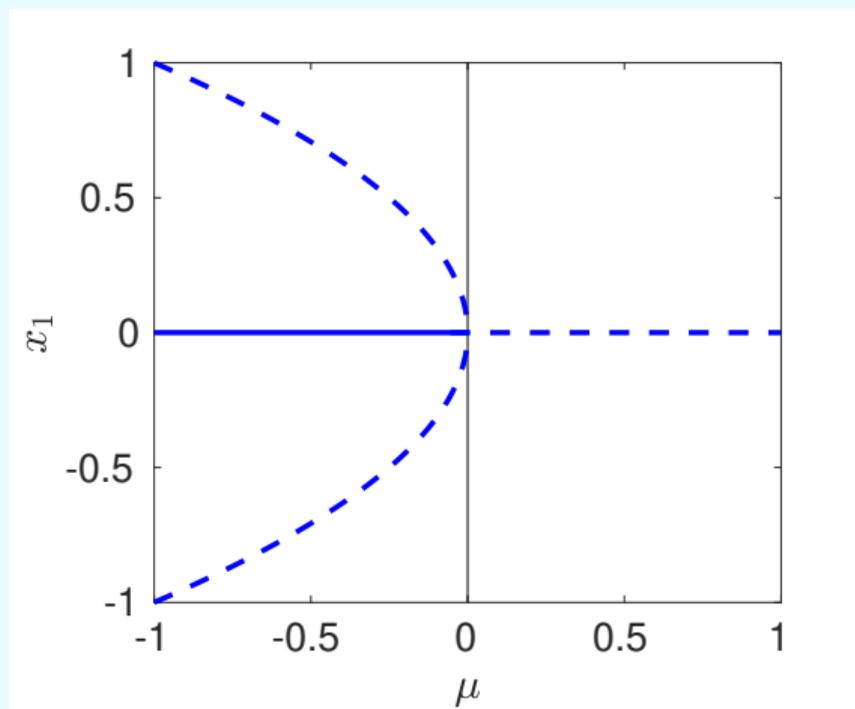


Subcritical pitchfork bifurcation

example: $\frac{dx_1}{dt} = \mu x_1 + x_1^3$

$$f(x_1; \mu) = \mu x_1 + x_1^3$$

bifurcation diagram



Subcritical pitchfork bifurcation

example:
$$\frac{dx_1}{dt} = \mu x_1 + x_1^3$$
$$\frac{dx_2}{dt} = -x_2$$

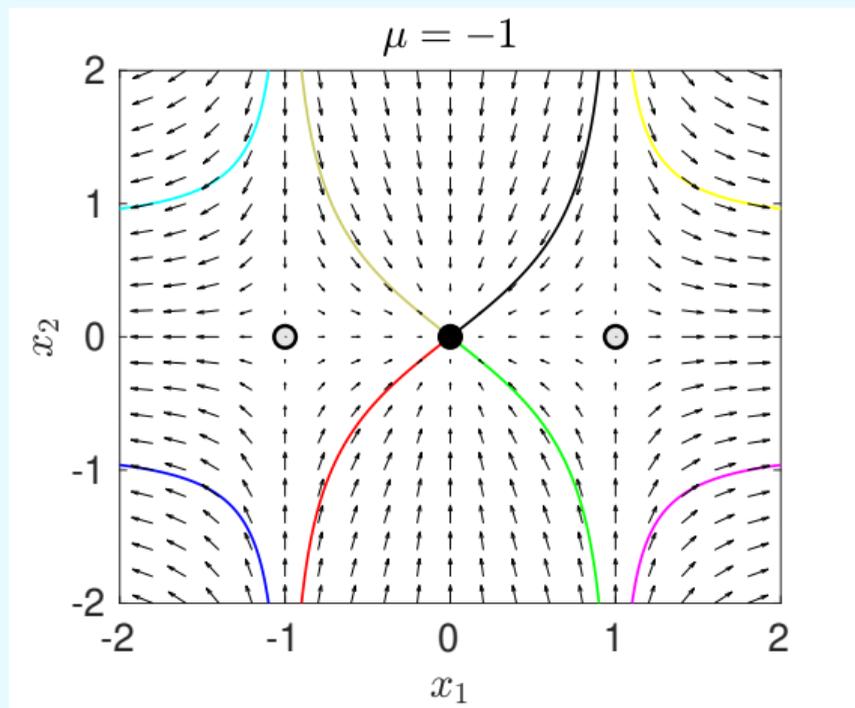
$$\mu < 0$$

three fixed points at

$$\mathbf{x} = [-\sqrt{-\mu}, 0] \text{ (saddle)}$$

$$\mathbf{x} = [0, 0] \text{ (stable node)}$$

$$\mathbf{x} = [\sqrt{-\mu}, 0] \text{ (saddle)}$$

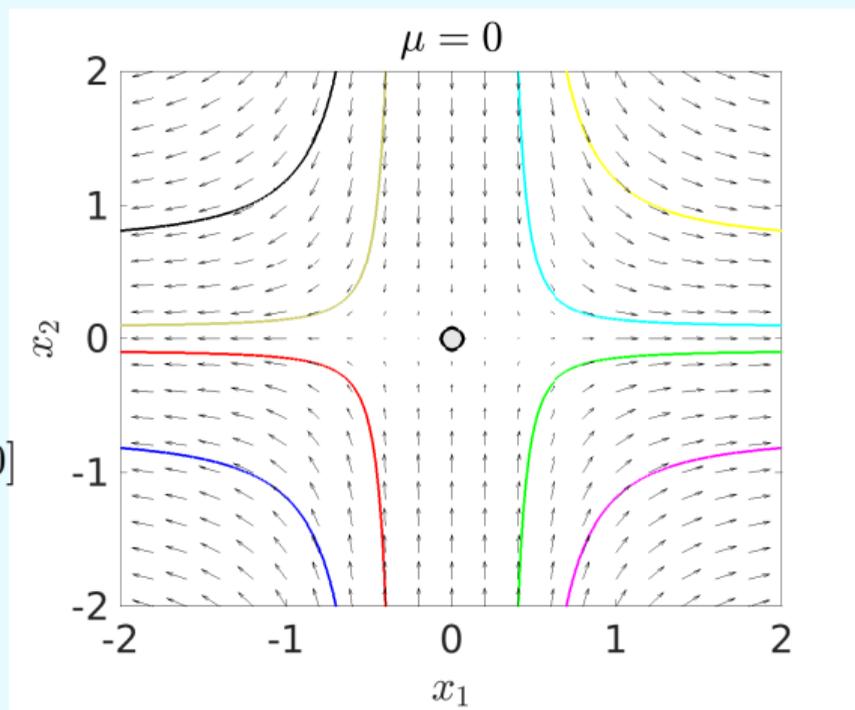


Subcritical pitchfork bifurcation

example:
$$\frac{dx_1}{dt} = \mu x_1 + x_1^3$$
$$\frac{dx_2}{dt} = -x_2$$

as μ approaches zero from below, two fixed points $[-\sqrt{-\mu}, 0]$ and $[\sqrt{-\mu}, 0]$ move toward the third one

$\mu = 0$: the fixed points coalesce into an unstable fixed point at $\mathbf{x} = [0, 0]$

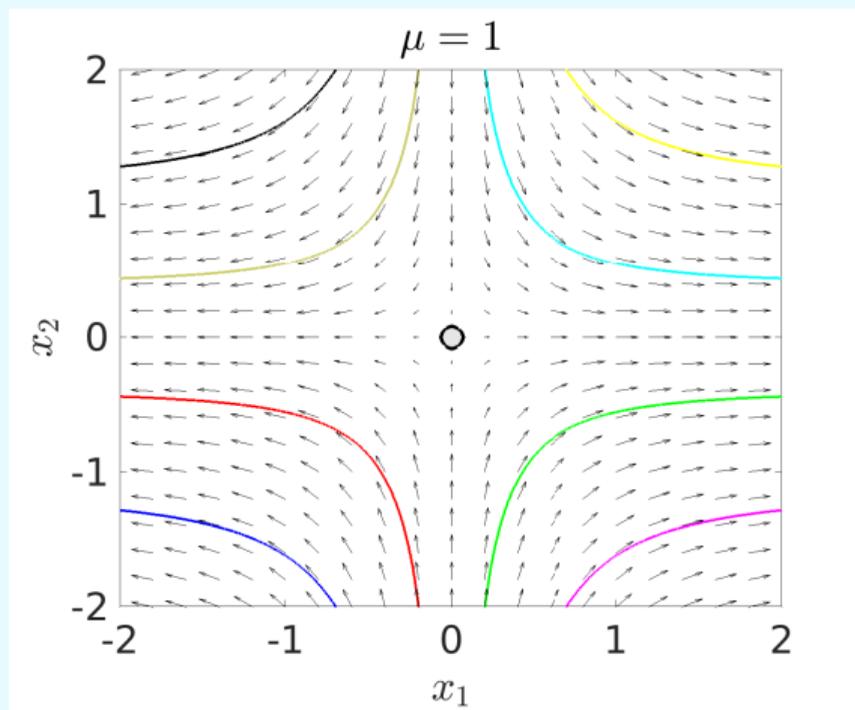


Subcritical pitchfork bifurcation

example:
$$\frac{dx_1}{dt} = \mu x_1 + x_1^3$$
$$\frac{dx_2}{dt} = -x_2$$

$\mu > 0$:

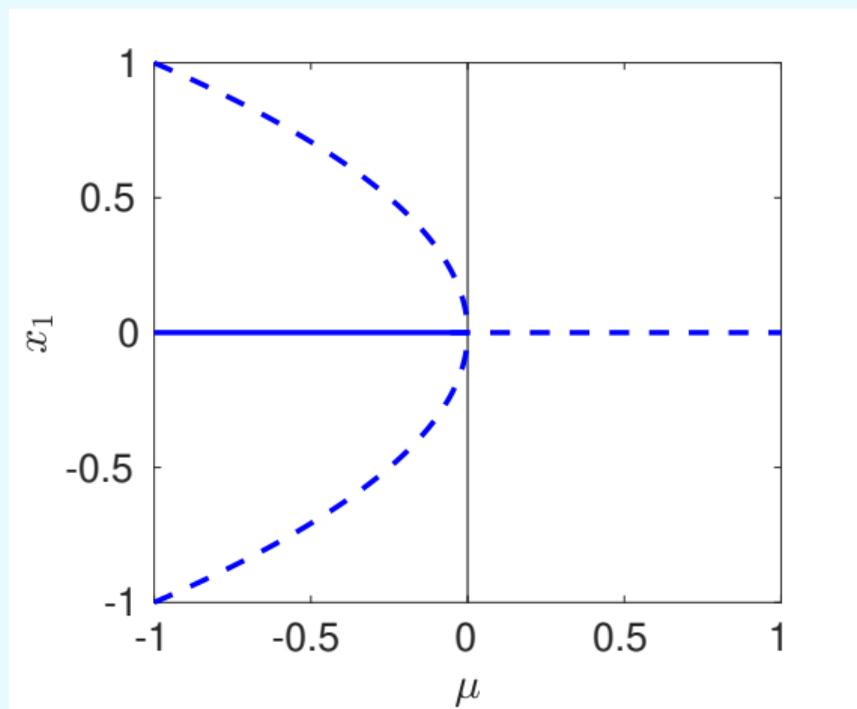
one unstable fixed point at $\mathbf{x} = [0, 0]$



Subcritical pitchfork bifurcation

example:
$$\frac{dx_1}{dt} = \mu x_1 + x_1^3$$
$$\frac{dx_2}{dt} = -x_2$$

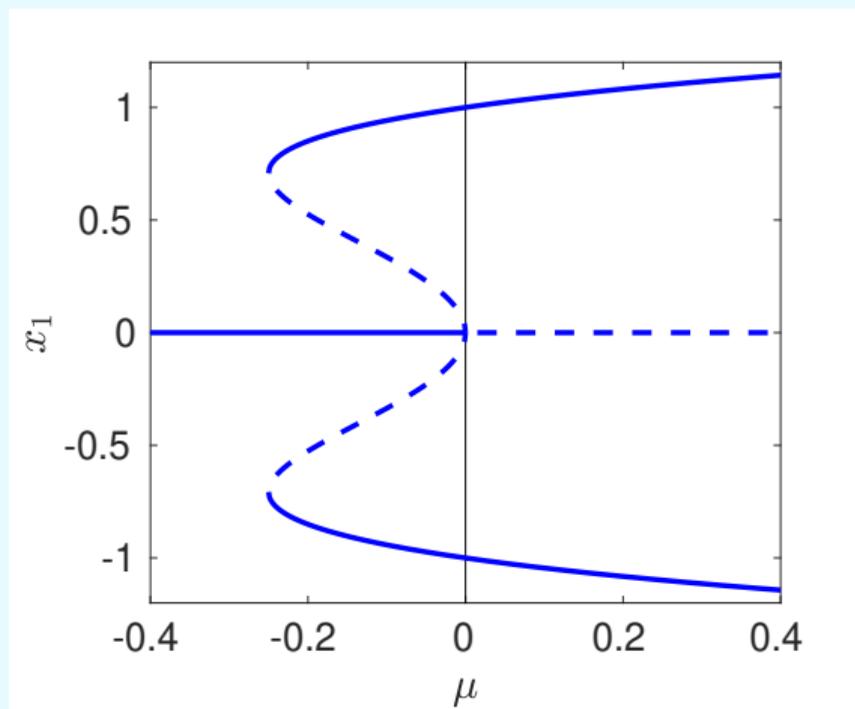
bifurcation diagram



Subcritical pitchfork bifurcation

example:
$$\frac{dx_1}{dt} = \mu x_1 + x_1^3 - x_1^5$$
$$\frac{dx_2}{dt} = -x_2$$

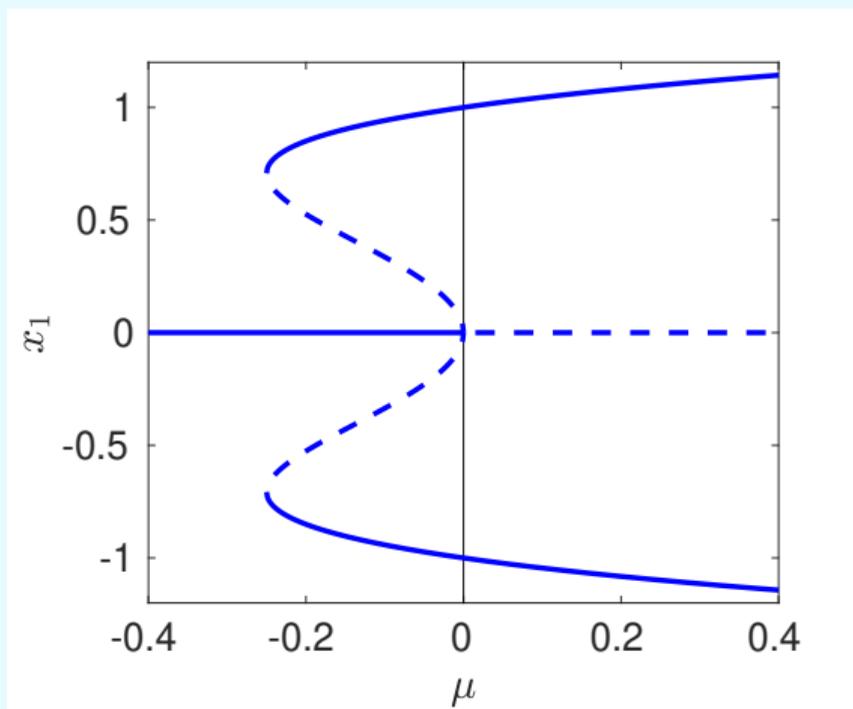
bifurcation diagram



Subcritical pitchfork bifurcation

example:
$$\frac{dx_1}{dt} = \mu x_1 + x_1^3 - x_1^5$$
$$\frac{dx_2}{dt} = -x_2$$

bifurcation diagram



Other examples of ODEs with bifurcations:

Questions 1, 2, 5 and 6 on Problem Sheet 2

Extended center manifold

example: $\frac{dx_1}{dt} = x_2^2 - x_1$

$$\frac{dx_2}{dt} = \mu x_2 - x_1 x_2$$

Extended center manifold

example: $\frac{dx_1}{dt} = x_2^2 - x_1$

$$\frac{dx_2}{dt} = \mu x_2 - x_1 x_2$$

$[x_1, x_2] = [0, 0]$ is a critical point

linearized system $\frac{d\mathbf{x}}{dt} = M\mathbf{x}$

$$\text{for } M = \begin{pmatrix} -1 & 0 \\ 0 & \mu \end{pmatrix}$$

Extended center manifold

example: $\frac{dx_1}{dt} = x_2^2 - x_1$

$$\frac{dx_2}{dt} = \mu x_2 - x_1 x_2$$

$$\frac{d\mu}{dt} = 0$$

$[x_1, x_2, \mu] = [0, 0, 0]$ is a critical point

linearized system $\frac{d\mathbf{x}}{dt} = M\mathbf{x}$

$$\text{for } M = \begin{pmatrix} -1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

Extended center manifold

example: $\frac{dx_1}{dt} = x_2^2 - x_1$

$$\frac{dx_2}{dt} = \mu x_2 - x_1 x_2$$

$$\frac{d\mu}{dt} = 0$$

$[x_1, x_2, \mu] = [0, 0, 0]$ is a critical point

linearized system $\frac{d\mathbf{x}}{dt} = M\mathbf{x}$

$$\text{for } M = \begin{pmatrix} -1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

the center manifold is given by

$$x_1 = h(x_2, \mu) = c_{20} x_2^2 + c_{11} \mu x_2 + c_{02} \mu^2 + \mathcal{O}(x_2^3, x_2^2 \mu, x_2 \mu^2, \mu^3)$$

Extended center manifold

example: $\frac{dx_1}{dt} = x_2^2 - x_1$

$$\frac{dx_2}{dt} = \mu x_2 - x_1 x_2$$

$$\frac{d\mu}{dt} = 0$$

$[x_1, x_2, \mu] = [0, 0, 0]$ is a critical point

linearized system $\frac{d\mathbf{x}}{dt} = M\mathbf{x}$

$$\text{for } M = \begin{pmatrix} -1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

the center manifold is given by

$$x_1 = h(x_2, \mu) = c_{20} x_2^2 + c_{11} \mu x_2 + c_{02} \mu^2 + \mathcal{O}(x_2^3, x_2^2 \mu, x_2 \mu^2, \mu^3)$$

$$\frac{dx_1}{dt} = \frac{\partial h}{\partial x_2}(x_2, \mu) \frac{dx_2}{dt} + \frac{\partial h}{\partial \mu}(x_2, \mu) \frac{d\mu}{dt}$$

Extended center manifold

example: $\frac{dx_1}{dt} = x_2^2 - x_1$

$$\frac{dx_2}{dt} = \mu x_2 - x_1 x_2$$

$$\frac{d\mu}{dt} = 0$$

$[x_1, x_2, \mu] = [0, 0, 0]$ is a critical point

linearized system $\frac{d\mathbf{x}}{dt} = M\mathbf{x}$

$$\text{for } M = \begin{pmatrix} -1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

the center manifold is given by

$$x_1 = h(x_2, \mu) = c_{20} x_2^2 + c_{11} \mu x_2 + c_{02} \mu^2 + \mathcal{O}(x_2^3, x_2^2 \mu, x_2 \mu^2, \mu^3)$$

$$x_2^2 - x_1 = \frac{dx_1}{dt} = \frac{\partial h}{\partial x_2}(x_2, \mu) \frac{dx_2}{dt} + \frac{\partial h}{\partial \mu}(x_2, \mu) \frac{d\mu}{dt} = \frac{\partial h}{\partial x_2}(x_2, \mu) (\mu x_2 - x_1 x_2)$$

Extended center manifold

example: $\frac{dx_1}{dt} = x_2^2 - x_1$

$$\frac{dx_2}{dt} = \mu x_2 - x_1 x_2$$

$$\frac{d\mu}{dt} = 0$$

$[x_1, x_2, \mu] = [0, 0, 0]$ is a critical point

linearized system $\frac{d\mathbf{x}}{dt} = M\mathbf{x}$

$$\text{for } M = \begin{pmatrix} -1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

the center manifold is given by

$$x_1 = h(x_2, \mu) = c_{20} x_2^2 + c_{11} \mu x_2 + c_{02} \mu^2 + \mathcal{O}(x_2^3, x_2^2 \mu, x_2 \mu^2, \mu^3)$$

$$x_2^2 - x_1 = \frac{dx_1}{dt} = \frac{\partial h}{\partial x_2}(x_2, \mu) \frac{dx_2}{dt} + \frac{\partial h}{\partial \mu}(x_2, \mu) \frac{d\mu}{dt} = \frac{\partial h}{\partial x_2}(x_2, \mu) (\mu x_2 - x_1 x_2)$$

$$x_2^2 - c_{20} x_2^2 - c_{11} \mu x_2 - c_{02} \mu^2 = (2 c_{20} x_2 + c_{11} \mu) \mu x_2 + \mathcal{O}(x_2^3, x_2^2 \mu, x_2 \mu^2, \mu^3)$$

Extended center manifold

example: $\frac{dx_1}{dt} = x_2^2 - x_1$

$$\frac{dx_2}{dt} = \mu x_2 - x_1 x_2$$

$$\frac{d\mu}{dt} = 0$$

$[x_1, x_2, \mu] = [0, 0, 0]$ is a critical point

linearized system $\frac{d\mathbf{x}}{dt} = M\mathbf{x}$

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$$c_{20} = 1, \quad c_{11} = 0, \quad c_{02} = 0$$

center manifold: $x_1 = x_2^2 + \mathcal{O}(x_2^3, x_2^2 \mu, x_2 \mu^2, \mu^3)$

the dynamics on the center manifold: $\frac{dx_2}{dt} = \mu x_2 - x_2^3 + \mathcal{O}(x_2^4, x_2^3 \mu, x_2^2 \mu^2, x_2 \mu^3, \mu^4)$

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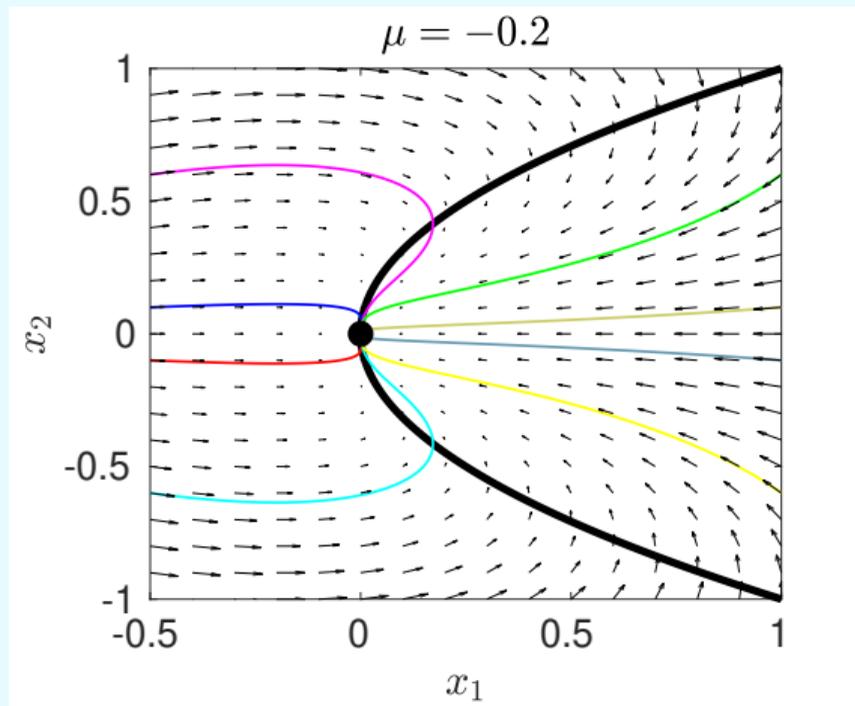
center manifold:

$$x_1 = x_2^2 + \dots$$

dynamics on the center manifold:

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supercritical pitchfork bifurcation



Extended center manifold

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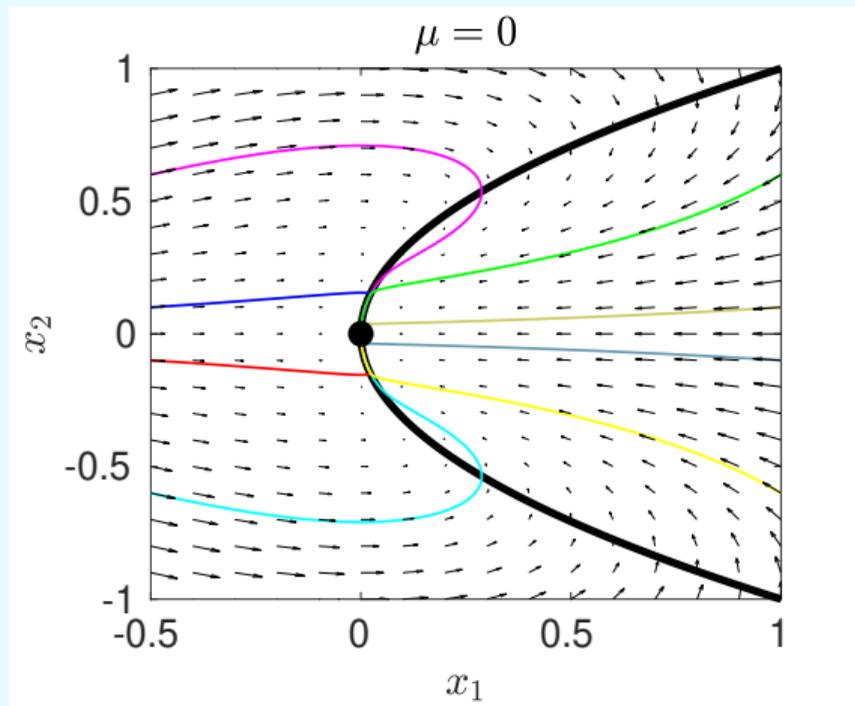
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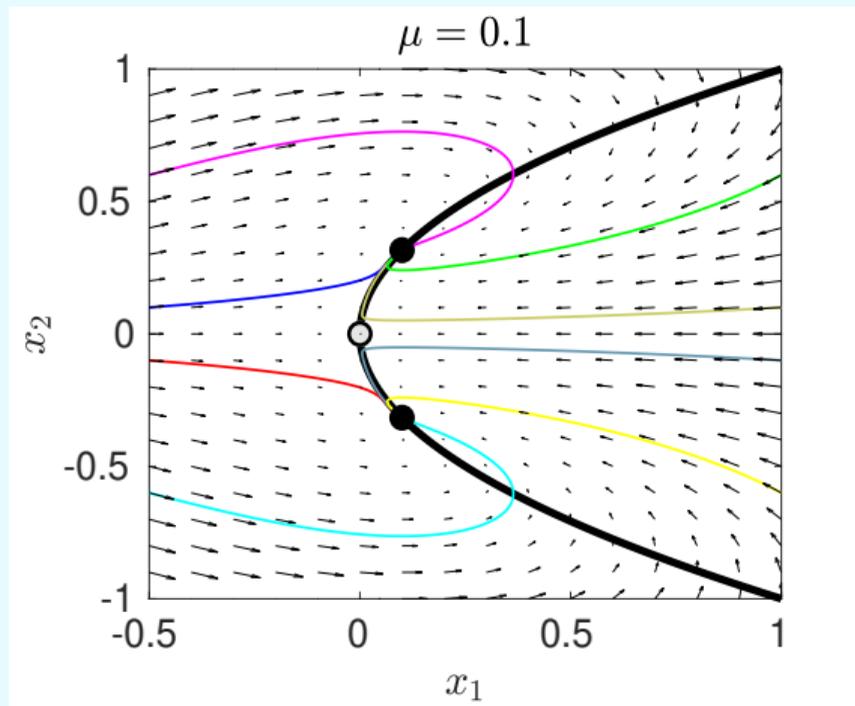
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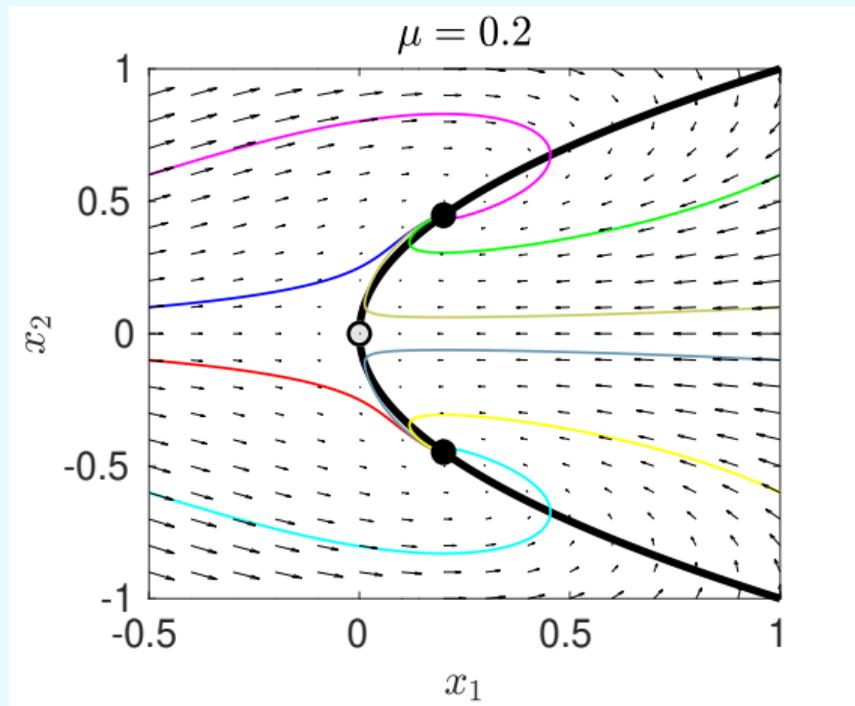
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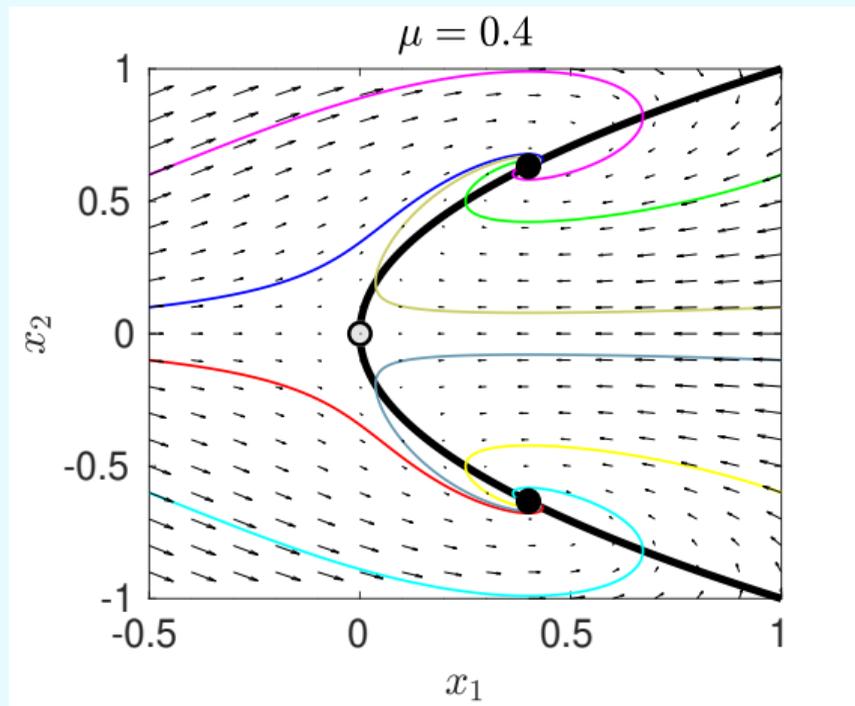
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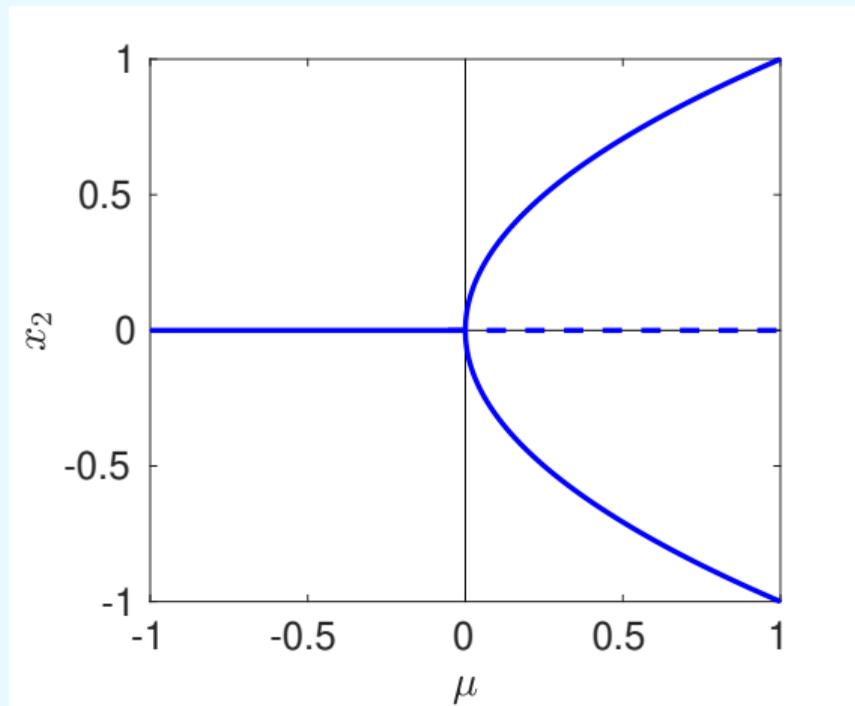
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Other examples of the extended center manifold calculation in ODEs with bifurcations:

[Questions 2, 5 and 6\(e\) on Problem Sheet 2](#)

Bifurcations of continuous-time dynamical systems – summary

Continuous-time dynamical system: Let $\mathbf{f} : \Omega \times \Theta \rightarrow \mathbb{R}^n$, where $\Omega \subset \mathbb{R}^n$ and $\Theta \subset \mathbb{R}^m$.

Let $\mathbf{x}_0 \in \Omega$, $\boldsymbol{\mu} \in \Theta$ and $\mathbf{x}(t) \in \Omega$ be a solution of the ODE

$$\frac{d\mathbf{x}}{dt} = \mathbf{f}(\mathbf{x}; \boldsymbol{\mu}) \quad \text{with the initial condition} \quad \mathbf{x}(0) = \mathbf{x}_0 \in \Omega$$

We have discussed bifurcations of fixed points, which can occur for $n \geq 1$ and $m \geq 1$ (so, they can be explained on examples with $n = 1$ and $m = 1$):

- saddle-node bifurcation
- transcritical bifurcation
- pitchfork bifurcation (supercritical, subcritical)

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- saddle-node bifurcation
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We will discuss later in the course:

- bifurcations of limit cycles ($n > 1$)
- bifurcations with more than one parameter ($m > 1$)

Next, we will discuss bifurcations of discrete-time dynamical systems.

B5.6: Nonlinear Dynamics, Bifurcations and Chaos (Lecture 6)

- summary of Lecture 5: we discussed Introduction to bifurcation theory, covering saddle-node, transcritical, supercritical pitchfork and subcritical pitchfork bifurcations. Extended center manifold.
(Questions 1, 2, 5 and 6 on Problem Sheet 2)
- today: we will continue in our discussion of Problem Sheet 2

B5.6: Nonlinear Dynamics, Bifurcations and Chaos (Lecture 6)

- summary of Lecture 5: we discussed Introduction to bifurcation theory, covering saddle-node, transcritical, supercritical pitchfork and subcritical pitchfork bifurcations. Extended center manifold. (Questions 1, 2, 5 and 6 on Problem Sheet 2)
- today: we will continue in our discussion of Problem Sheet 2
- course synopsis of **Lectures 1-8** (taken by both Part B and MSc students): Discrete-time (maps) and continuous-time (differential equations) dynamical systems. Notion of flows, stability of fixed points, Lyapunov function, invariant manifolds, stable manifold theorem, notion of hyperbolicity, center manifold. Chemical reaction networks. Stable, unstable and center subspaces. Poincaré-Bendixson theorem. Periodic solutions, stable and unstable limit cycles. Introduction to bifurcation theory, covering saddle-node, transcritical, supercritical pitchfork and subcritical pitchfork bifurcations. Extended center manifold. Logistic map. Periodic points of maps. Stability of N-cycles. Period-doubling bifurcation. Sharkovsky's theorem. Invariant distribution.

Fixed points

Discrete-time dynamical system: Let $\mathbf{F} : \Omega \times \Theta \rightarrow \Omega$, where $\Omega \subset \mathbb{R}^n$ and $\Theta \subset \mathbb{R}^m$.

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- Prelims Constructive Mathematics: we considered $n = 1$ where $x_{k+1} = F(x_k)$
 - $\alpha \in \mathbb{R}$ is a fixed point if $\alpha = F(\alpha)$
 - if $|F'(\alpha)| < 1$, then α is asymptotically stable

Example

$$x_{k+1} = 1 - 6x_k + 15x_k^2 - 10x_k^3$$

Example

$$x_{k+1} = 1 - 6x_k + 15x_k^2 - 10x_k^3$$

$$F(x) = 1 - 6x + 15x^2 - 10x^3$$

fixed points: solving $F(\alpha) = \alpha$

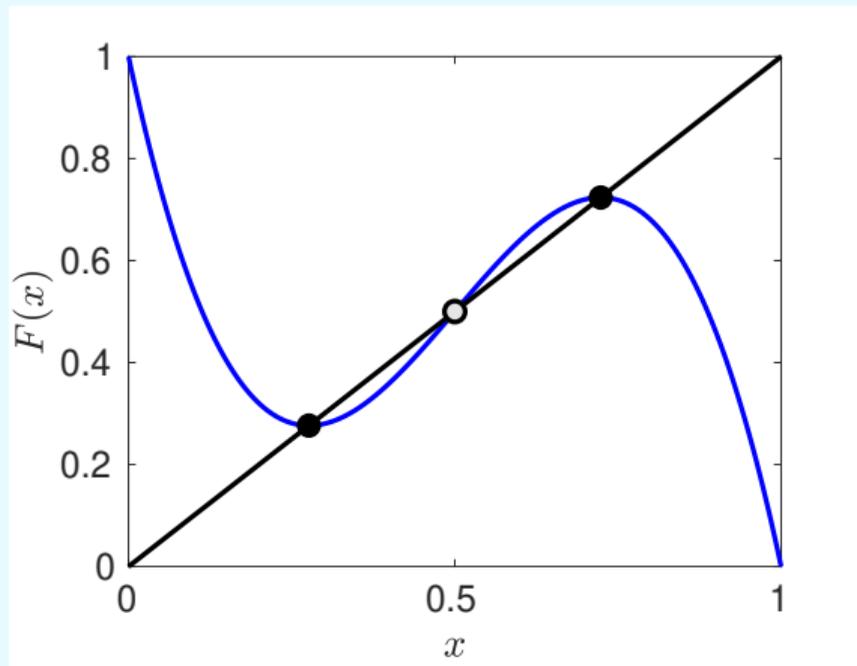
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fixed points: solving $F(\alpha) = \alpha$, we get

$$\alpha_1 = \frac{1}{2} - \frac{\sqrt{5}}{10}, \quad \alpha_2 = \frac{1}{2}, \quad \alpha_3 = \frac{1}{2} + \frac{\sqrt{5}}{10}$$



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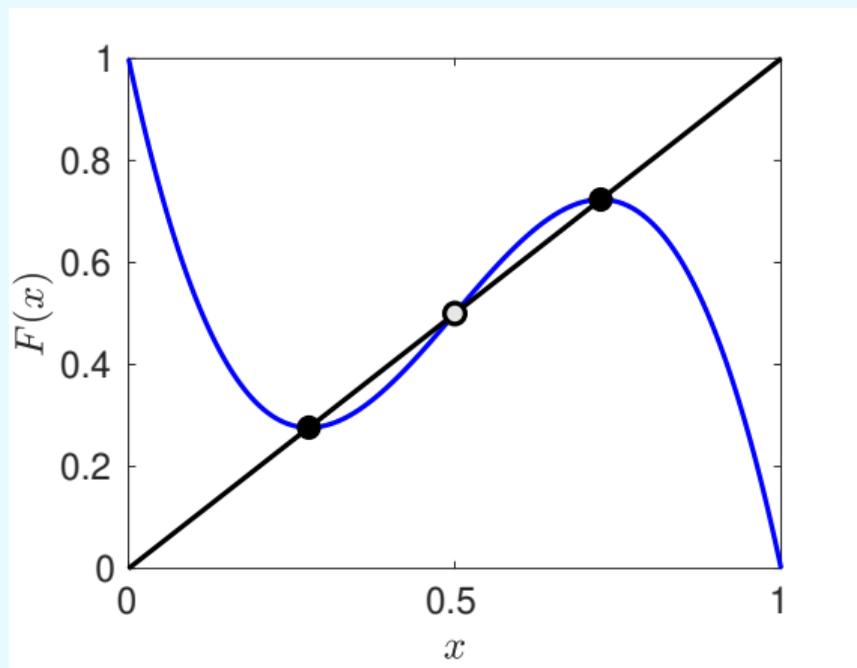
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$$F'(x) = -6 + 30x - 30x^2$$

$$F'(\alpha_1) = F'(\alpha_3) = 0, \quad F'(\alpha_2) = \frac{3}{2}$$

α_1 and α_3 are asymptotically stable

α_2 is unstable



Example

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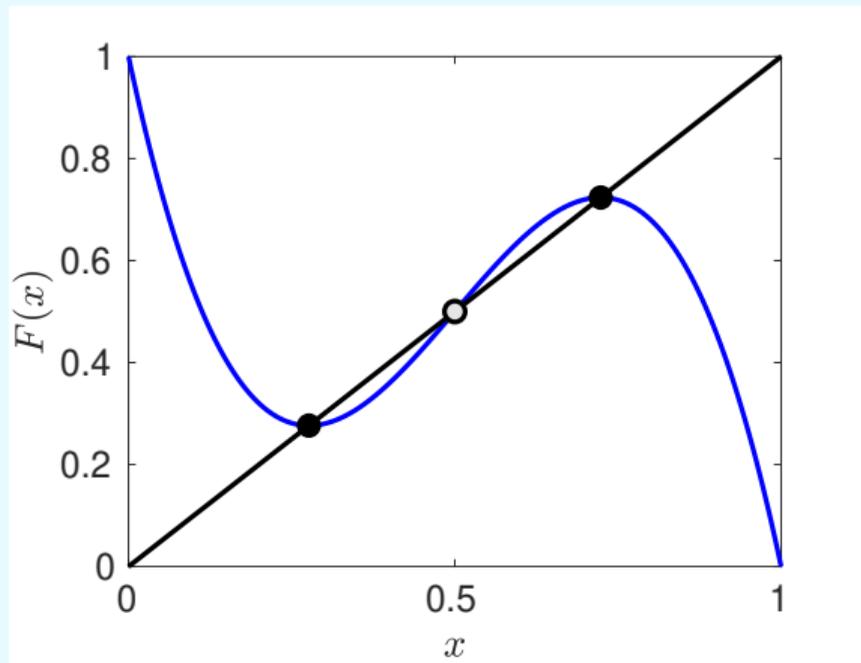
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α_2 is unstable

fixed point α with $|F'(\alpha)| < 1$ is asymptotically stable

fixed point α with $|F'(\alpha)| > 1$ is unstable

fixed point α with $|F'(\alpha)| = 0$ is called *super-attracting* because $|F'(\alpha)| = 0$ gives very fast convergence to the fixed point for nearby points



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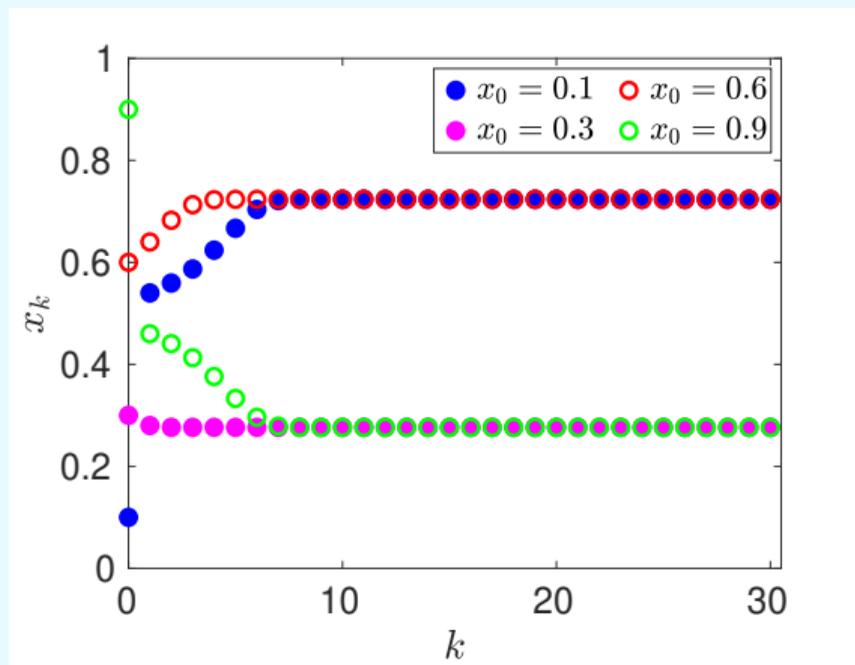
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- using notation $\mathbf{F}(\mathbf{x}; \boldsymbol{\mu}) = \mathbf{F}_\mu(\mathbf{x})$, we observe that $\mathbf{x}_1 = \mathbf{F}_\mu(\mathbf{x}_0)$

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$$\mathbf{x}_k = \mathbf{F}_\mu^{(k)}(\mathbf{x}_0)$$

which we can also use in above definitions.

Fixed points

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Fixed points, periodic points and N -cycles

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- $\boldsymbol{\alpha} \in \Omega$ is a *periodic point* with *period* $N \in \mathbb{N}$ if

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and the set $\left\{ \boldsymbol{\alpha}, \mathbf{F}_{\boldsymbol{\mu}}(\boldsymbol{\alpha}), \mathbf{F}_{\boldsymbol{\mu}}^{(2)}(\boldsymbol{\alpha}), \dots, \mathbf{F}_{\boldsymbol{\mu}}^{(N-1)}(\boldsymbol{\alpha}) \right\}$ is called an N -cycle

Fixed points, periodic points and N -cycles

Discrete-time dynamical system: Let $\mathbf{F} : \Omega \times \Theta \rightarrow \Omega$, where $\Omega \subset \mathbb{R}^n$ and $\Theta \subset \mathbb{R}^m$.

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- periodic point $\boldsymbol{\alpha} \in \Omega$ is *stable* if it is a *stable* fixed point of $\mathbf{F}_{\boldsymbol{\mu}}^{(N)}$ (resp. *asymptotically stable*, *unstable*)
- to find periodic points and the corresponding N -cycles, we need to solve $\boldsymbol{\alpha} = \mathbf{F}_{\boldsymbol{\mu}}^{(N)}(\boldsymbol{\alpha})$ and we also need to exclude solutions with some lesser period

Question 4 on Problem Sheet 2

Fixed points, periodic points, N -cycles, orbits and bifurcations

Discrete-time dynamical system: Let $\mathbf{F} : \Omega \times \Theta \rightarrow \Omega$, where $\Omega \subset \mathbb{R}^n$ and $\Theta \subset \mathbb{R}^m$.

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- given $\mathbf{x}_0 \in \Omega$, the *orbit* of map $\mathbf{F}_{\boldsymbol{\mu}}$ is the set

$$\left\{ \mathbf{x}_0, \mathbf{F}_{\boldsymbol{\mu}}(\mathbf{x}_0), \mathbf{F}_{\boldsymbol{\mu}}^{(2)}(\mathbf{x}_0), \mathbf{F}_{\boldsymbol{\mu}}^{(3)}(\mathbf{x}_0), \mathbf{F}_{\boldsymbol{\mu}}^{(4)}(\mathbf{x}_0), \dots \right\} = \{ \mathbf{x}_0, \mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3, \mathbf{x}_4, \dots \}$$

Fixed points, periodic points, N -cycles, orbits and bifurcations

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- if \mathbf{x}_0 is a periodic point with period N , then its orbit is a finite set (N -cycle)
- if orbit is a finite set, then it is (eventually) periodic,
i.e. there exists $j \in \mathbb{N}_0$ such that $\mathbf{F}_{\boldsymbol{\mu}}^{(j)}(\mathbf{x}_0)$ is a *periodic point* with *period* $N \in \mathbb{N}$

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- if orbit is an infinite set, then it can approach a fixed point or an N -cycle, or it can be chaotic - we will illustrate this on examples with $n = 1$ and $m = 1$
- bifurcations: the qualitative behaviour of orbits can change as parameters $\boldsymbol{\mu}$ are varied (for example, fixed points or N -cycles can be created or destroyed, or their stability changes); the parameter values at which these qualitative changes in the dynamics occur are called bifurcation points

Example

$$x_{k+1} = (1 - x_k)(1 - 5x_k + \mu x_k^2)$$

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fixed points: $F(x; \mu) = x$

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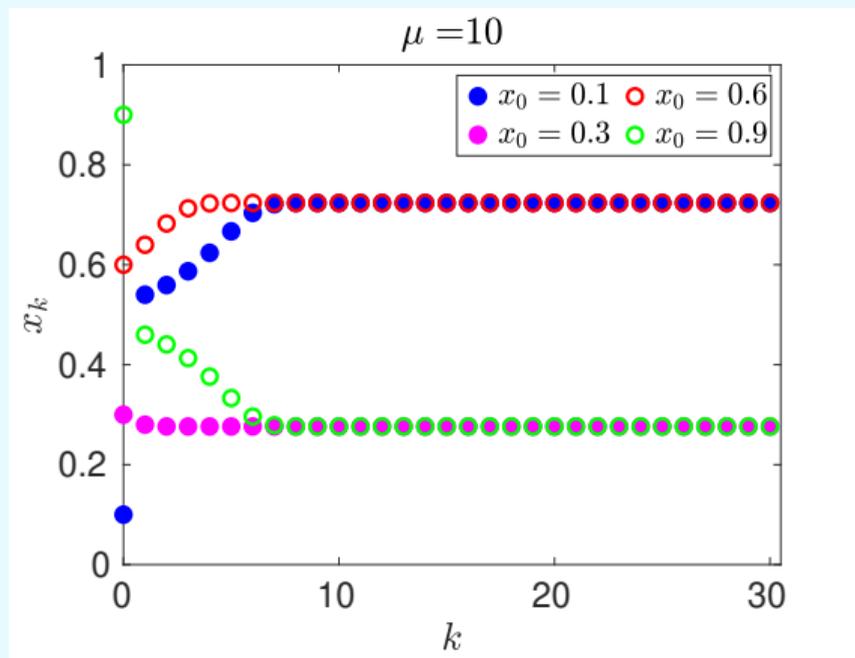
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our previous example: $\mu = 10$

$$x_{k+1} = 1 - 6x_k + 15x_k^2 - 10x_k^3$$

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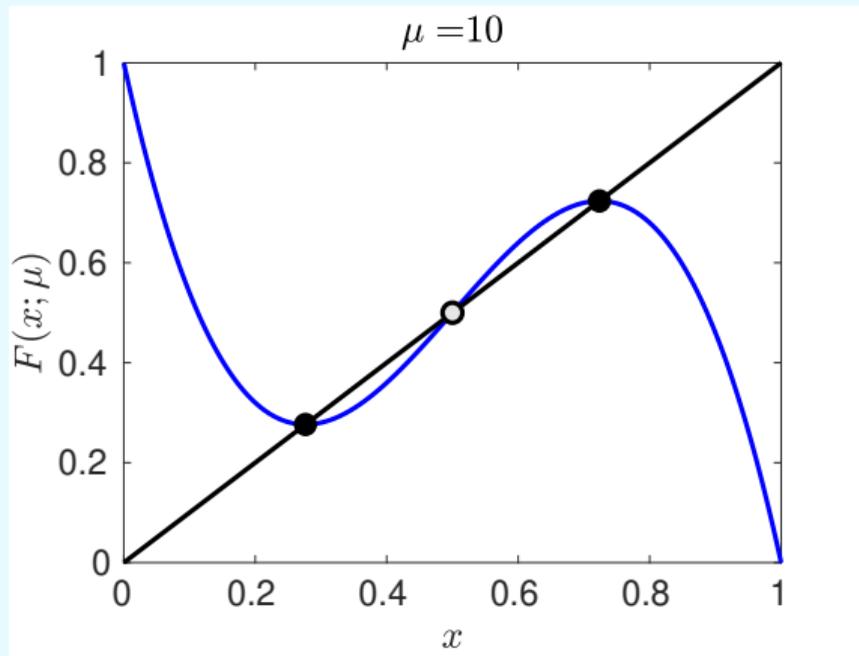
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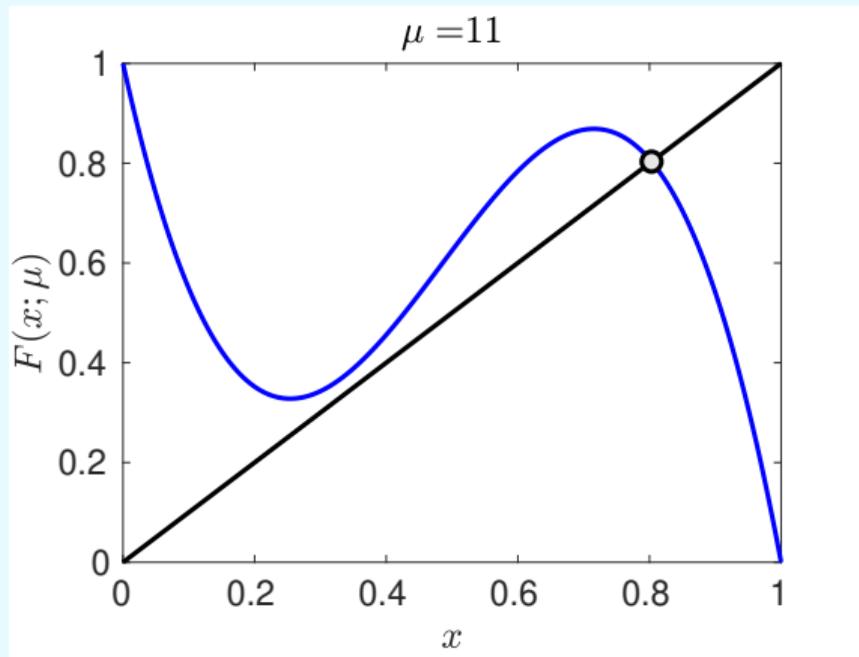
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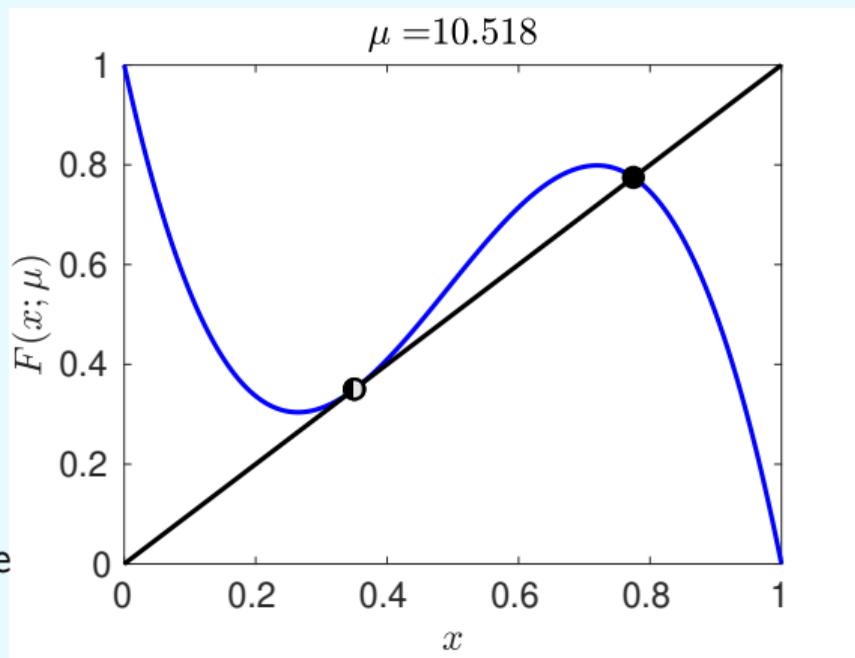
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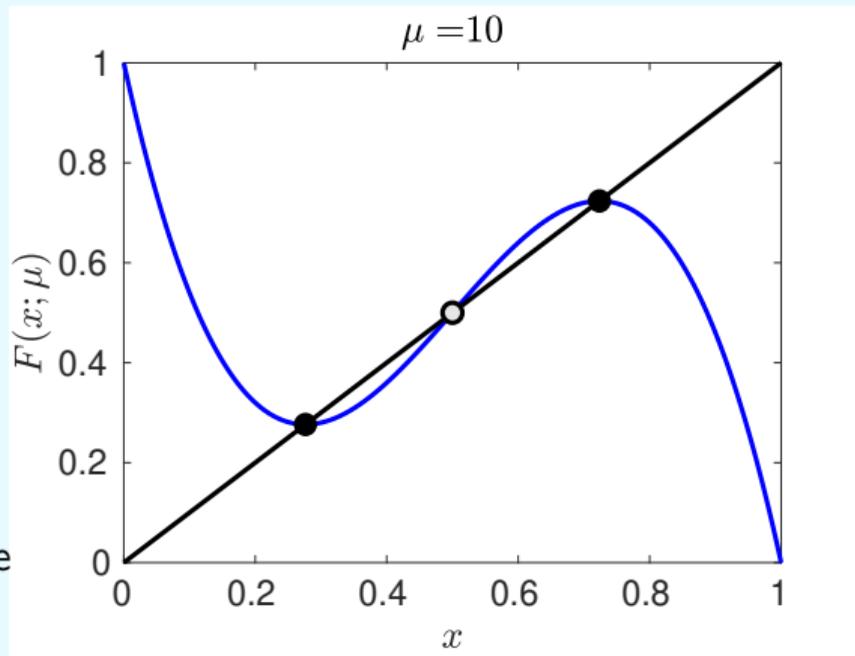
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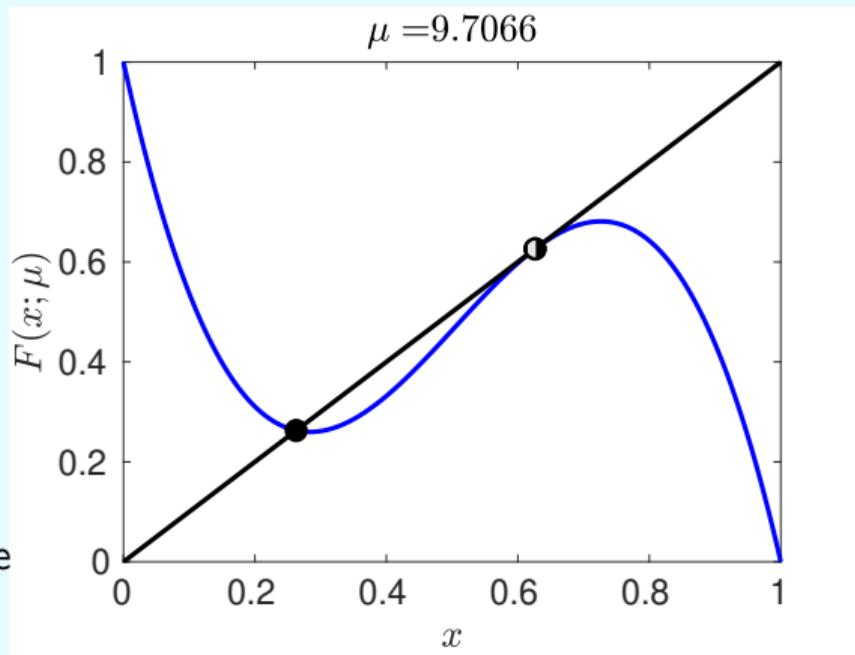
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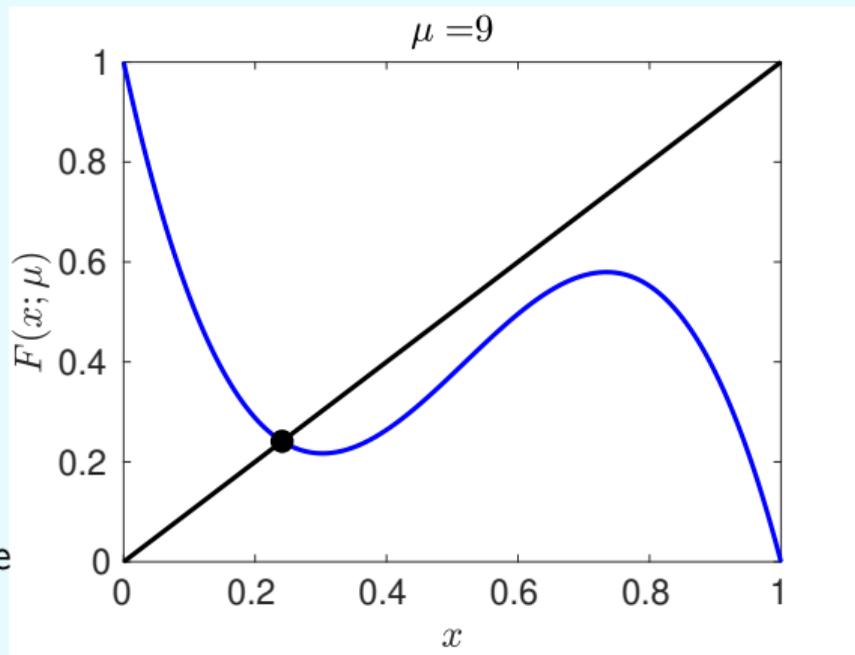
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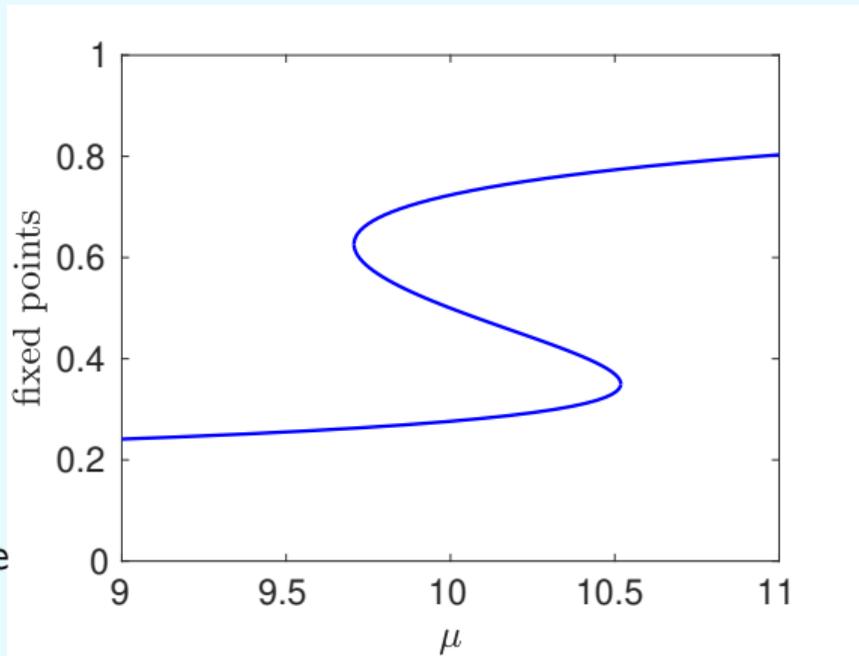
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$$\mu(x) = \frac{7x - 5x^2 - 1}{(1 - x)x^2}$$

$$\mu_1 \text{ and } \mu_2 \text{ can be found by solving } 0 = \mu'(x) = \frac{2 - 10x + 14x^2 - 5x^3}{(1 - x)^2 x^3} = 0$$



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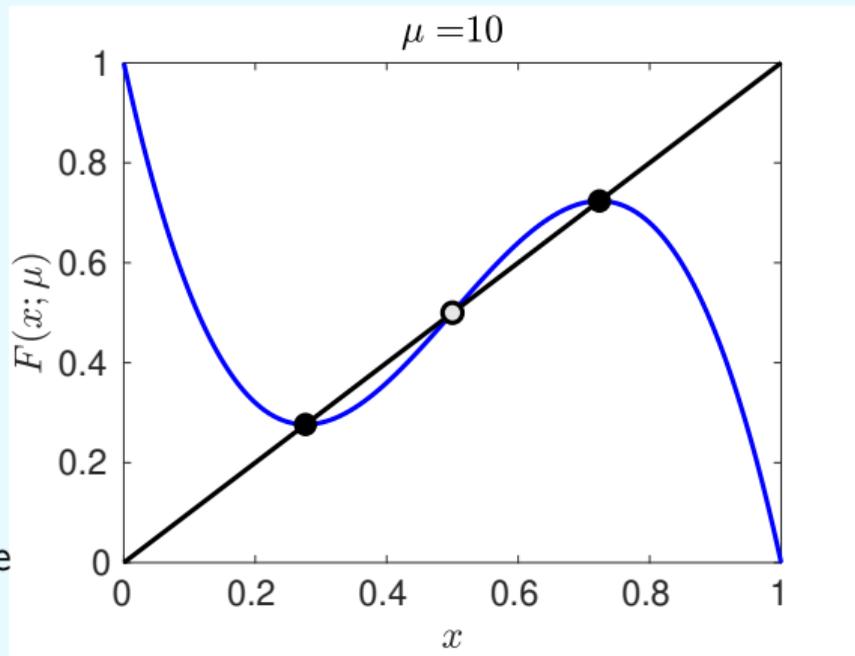
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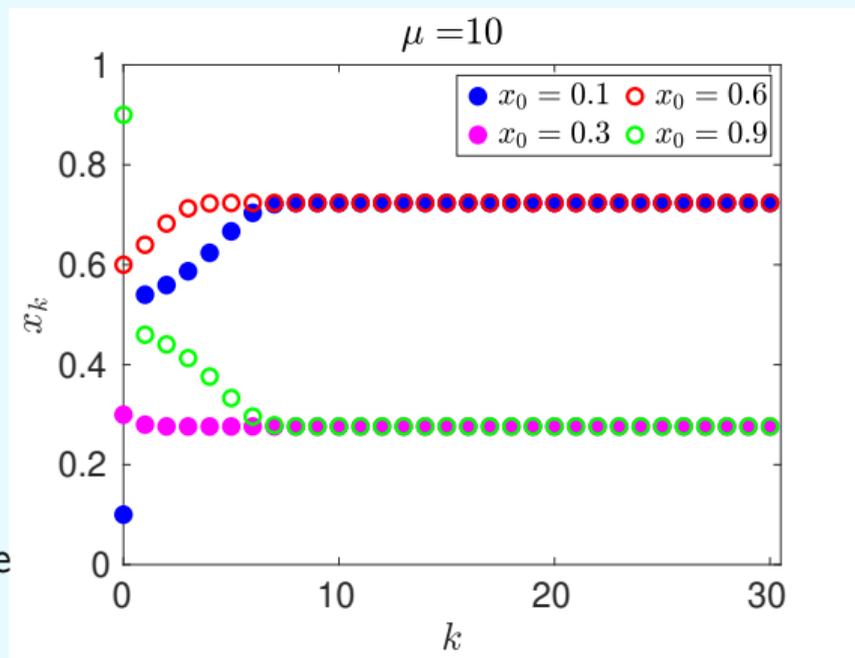
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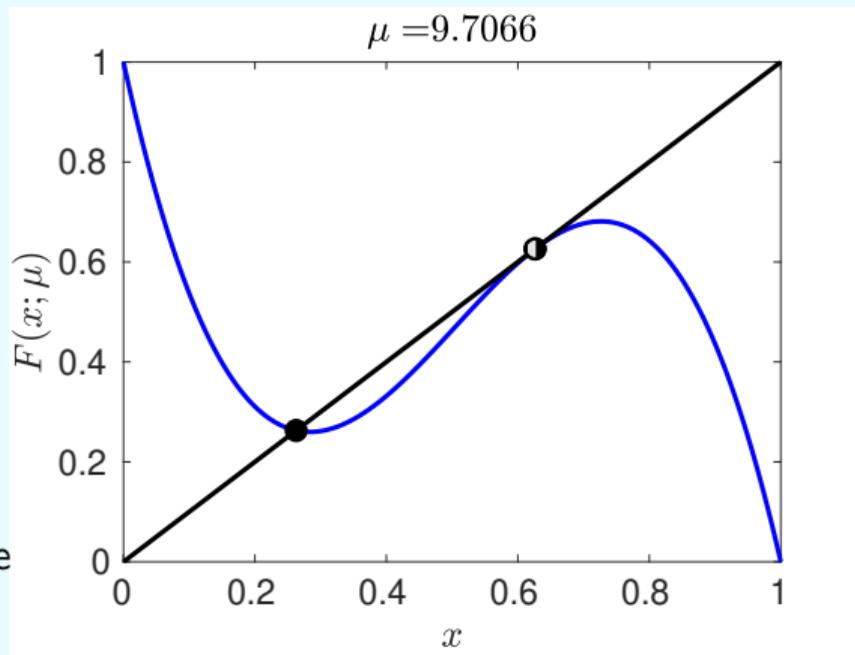
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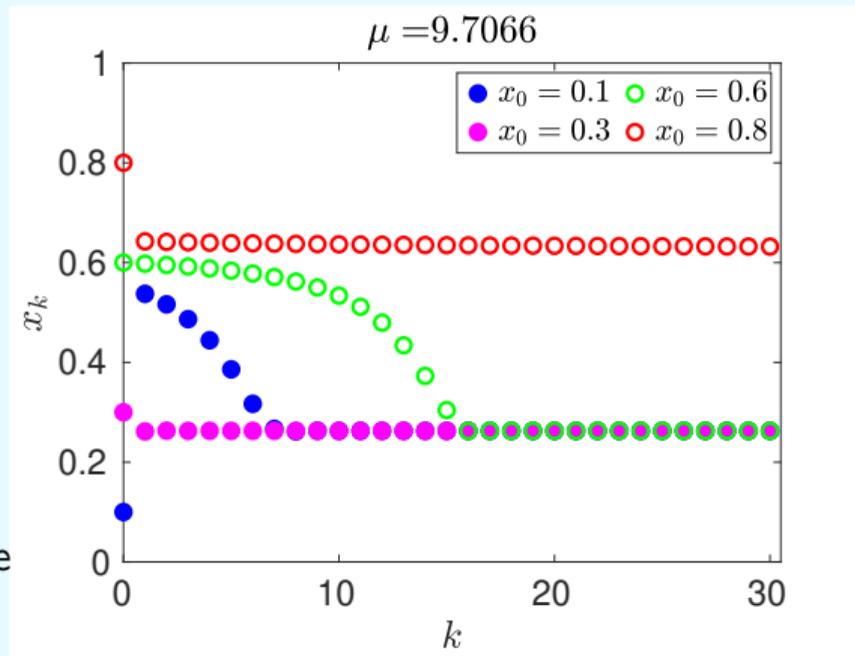
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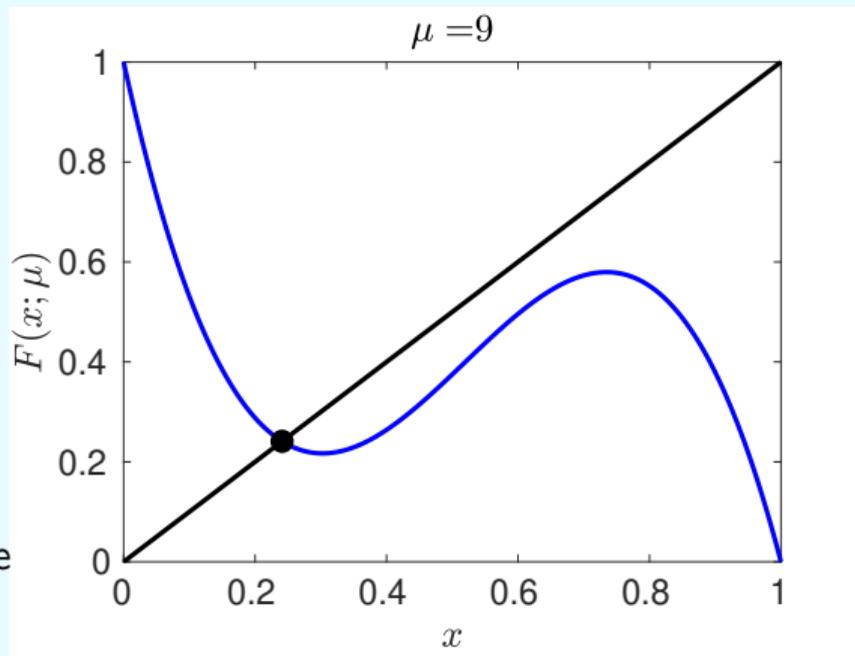
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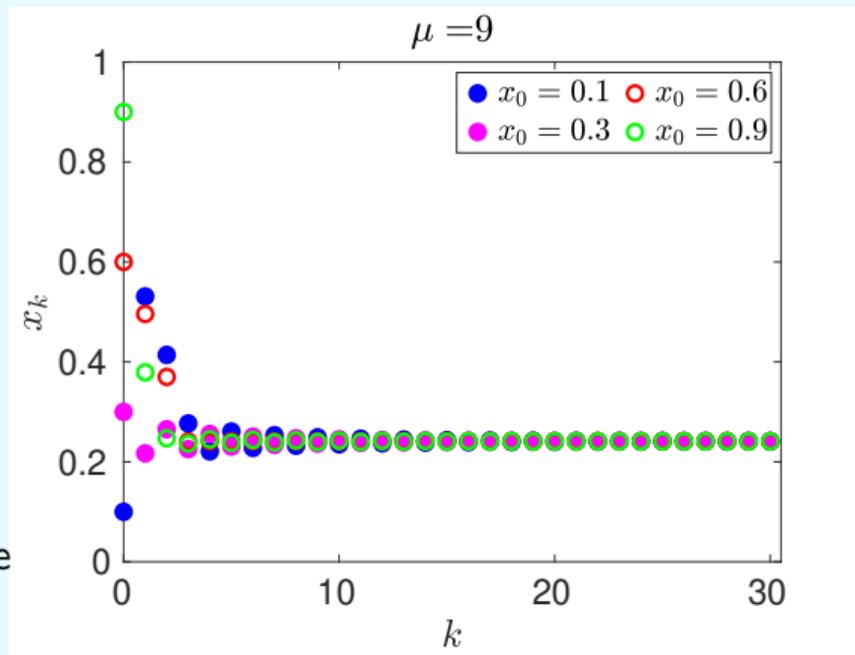
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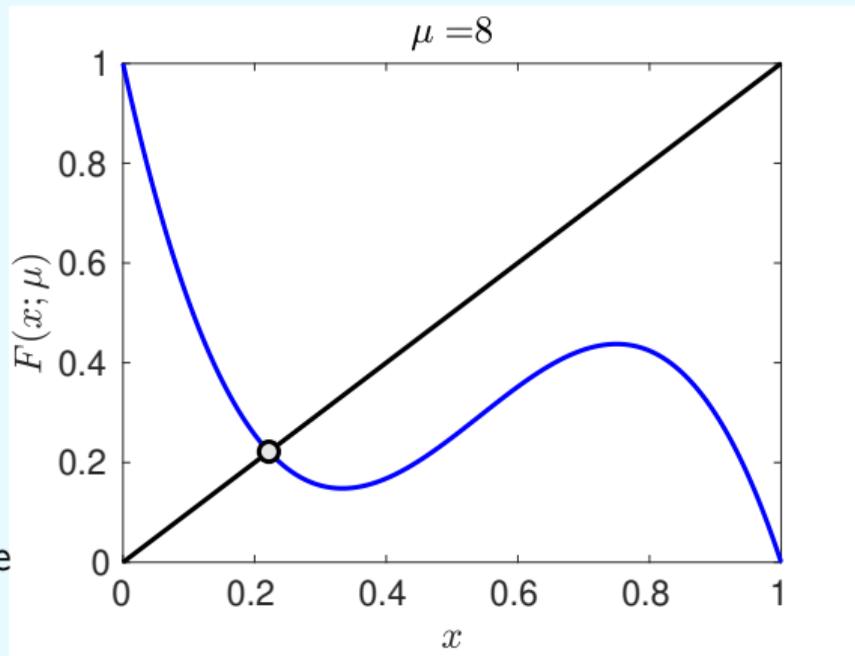
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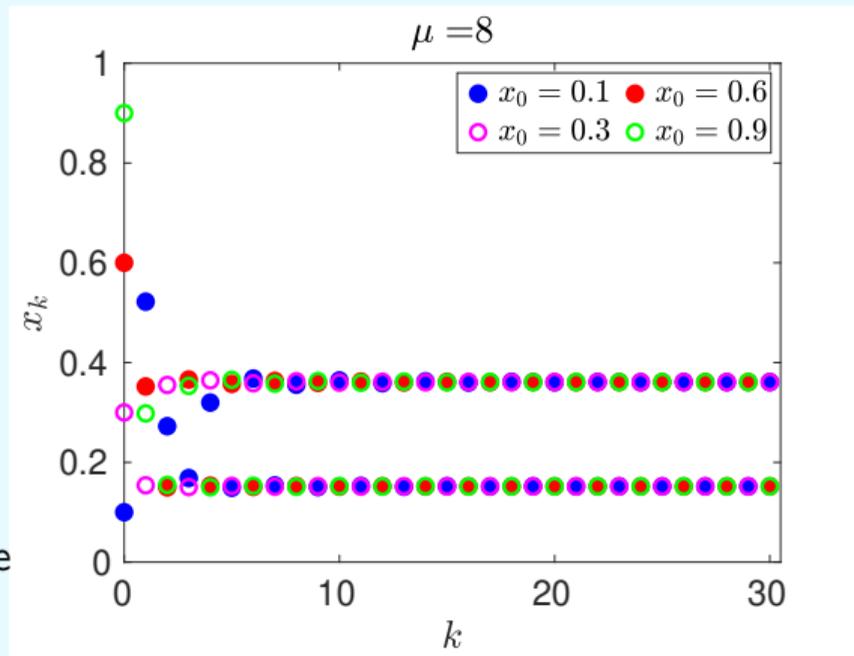
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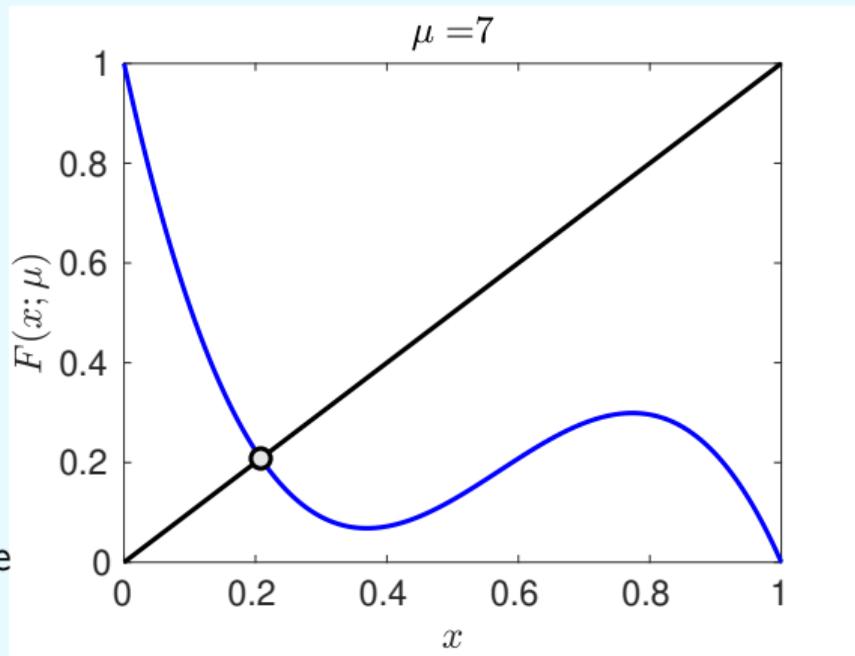
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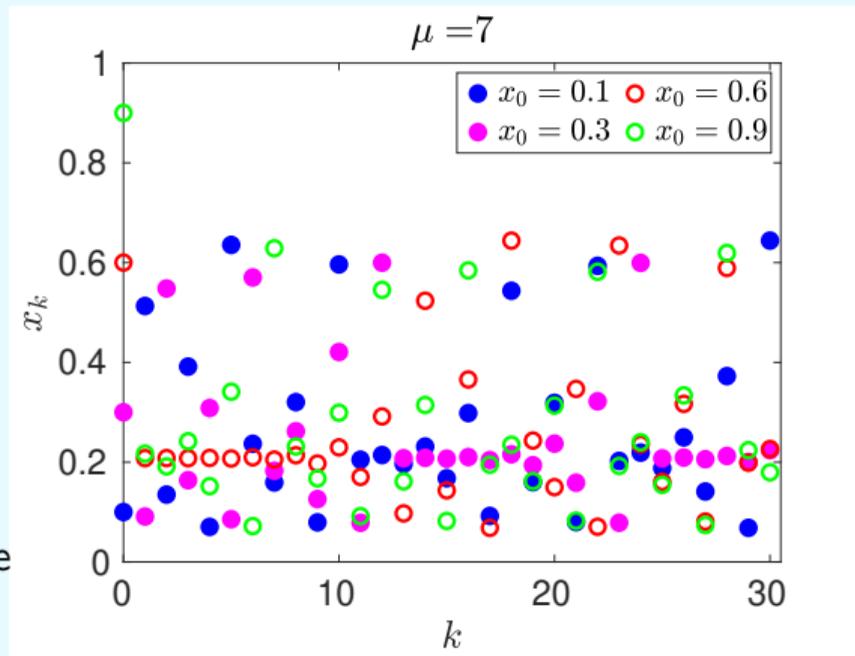
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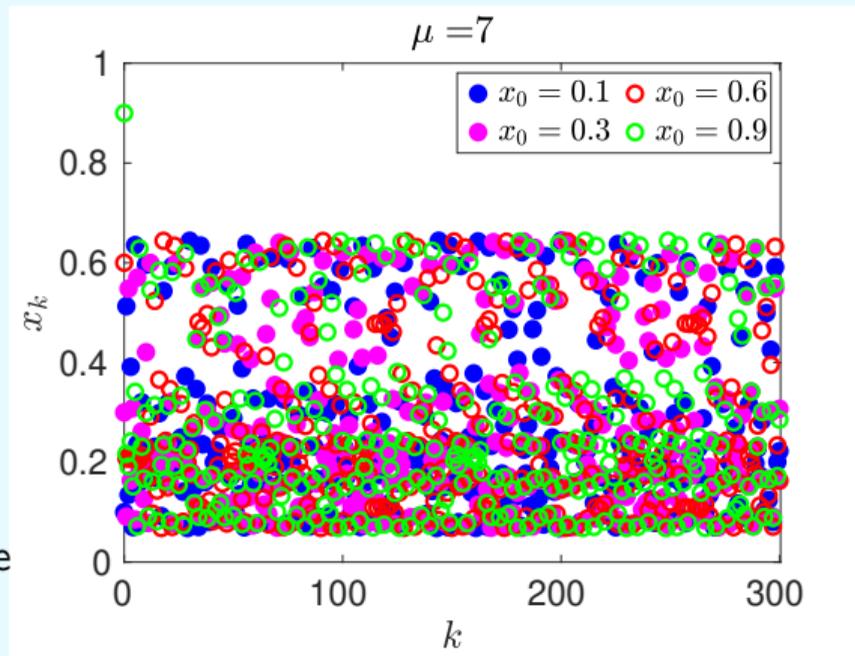
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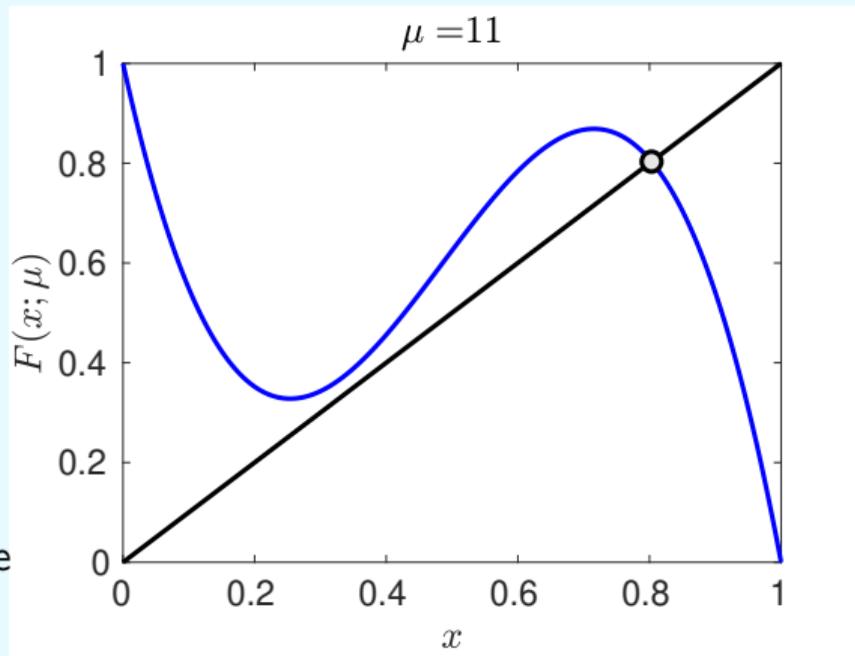
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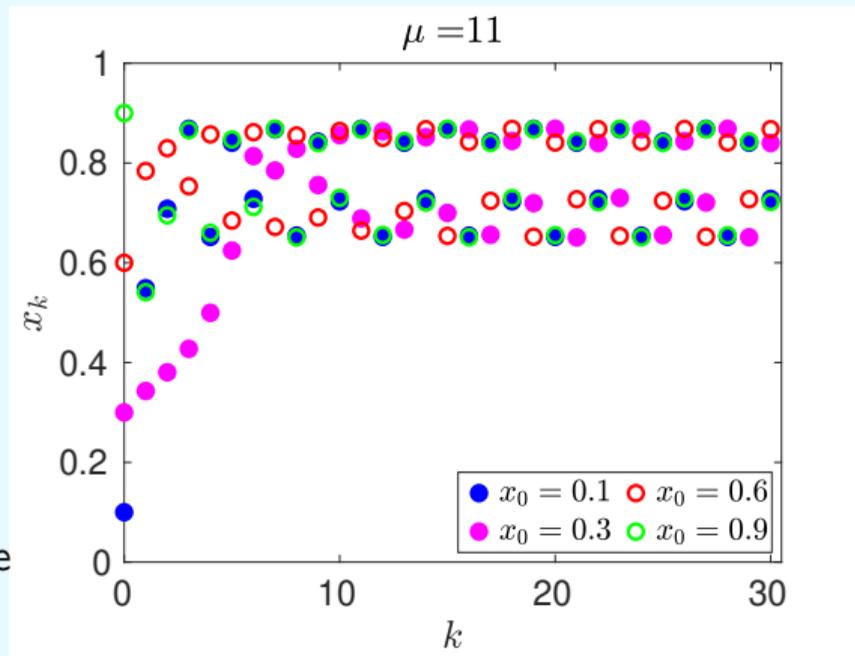
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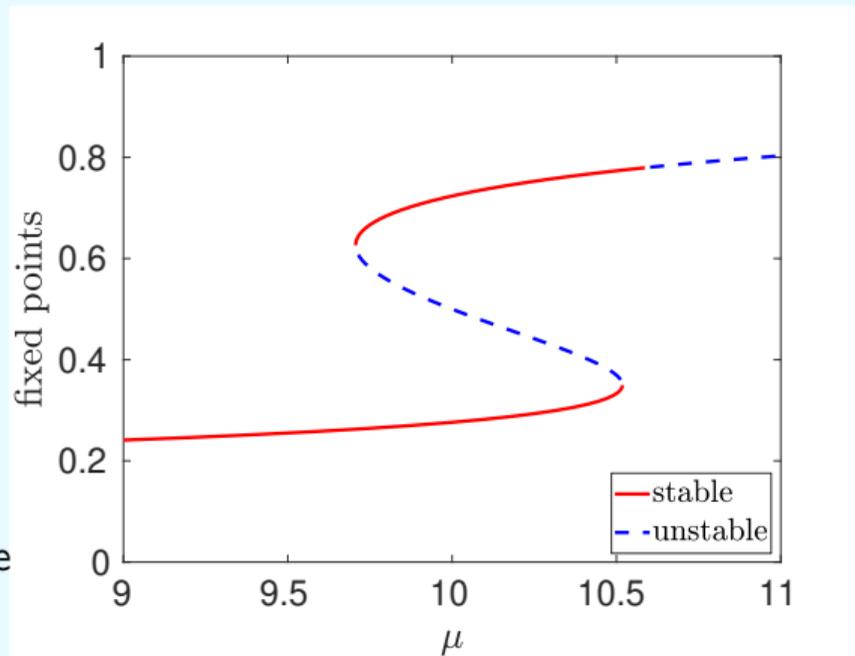
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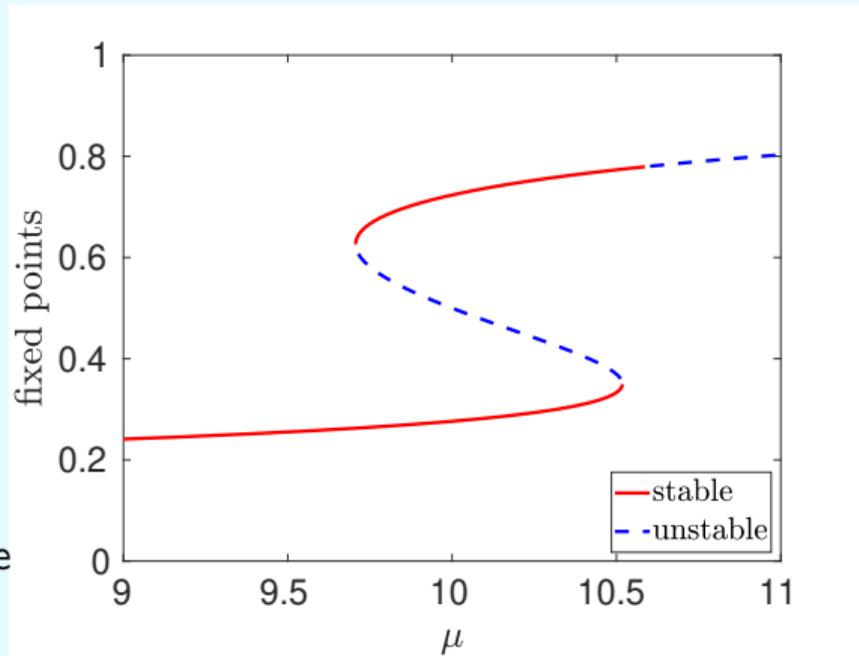
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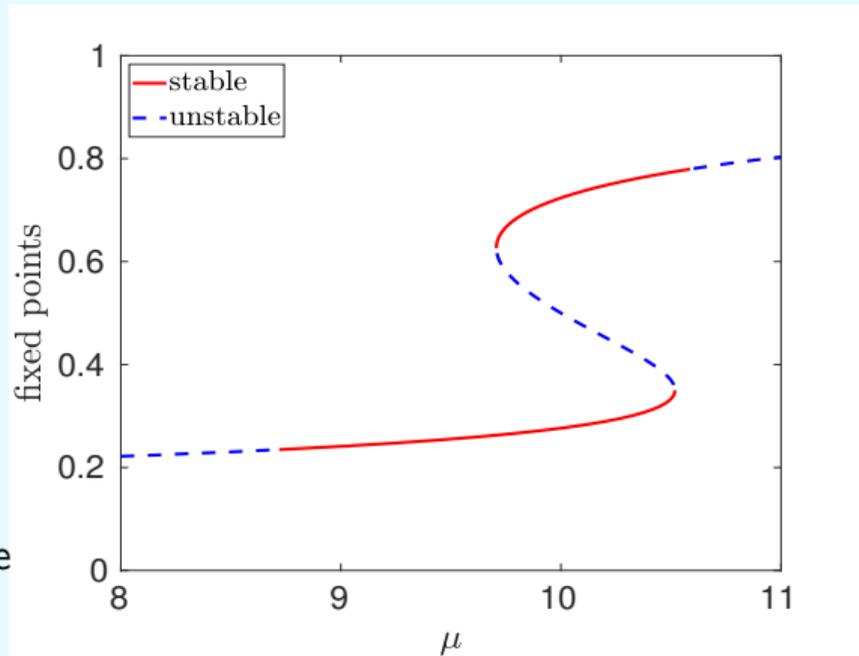
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Stability of N -cycles

Discrete-time dynamical system ($n = 1, m = 1$):

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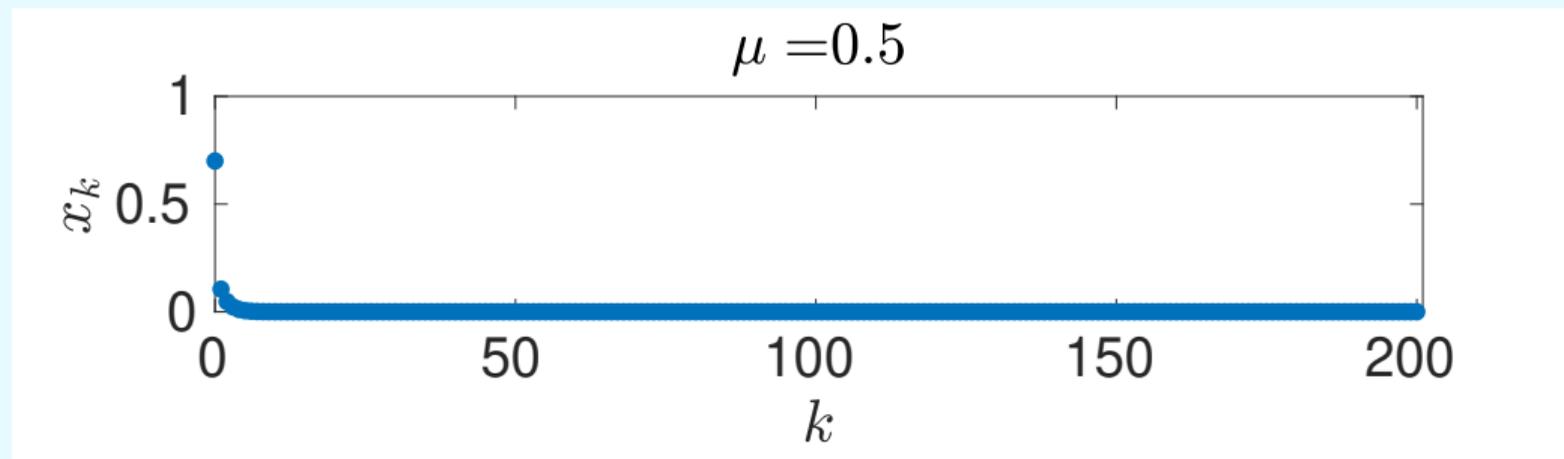
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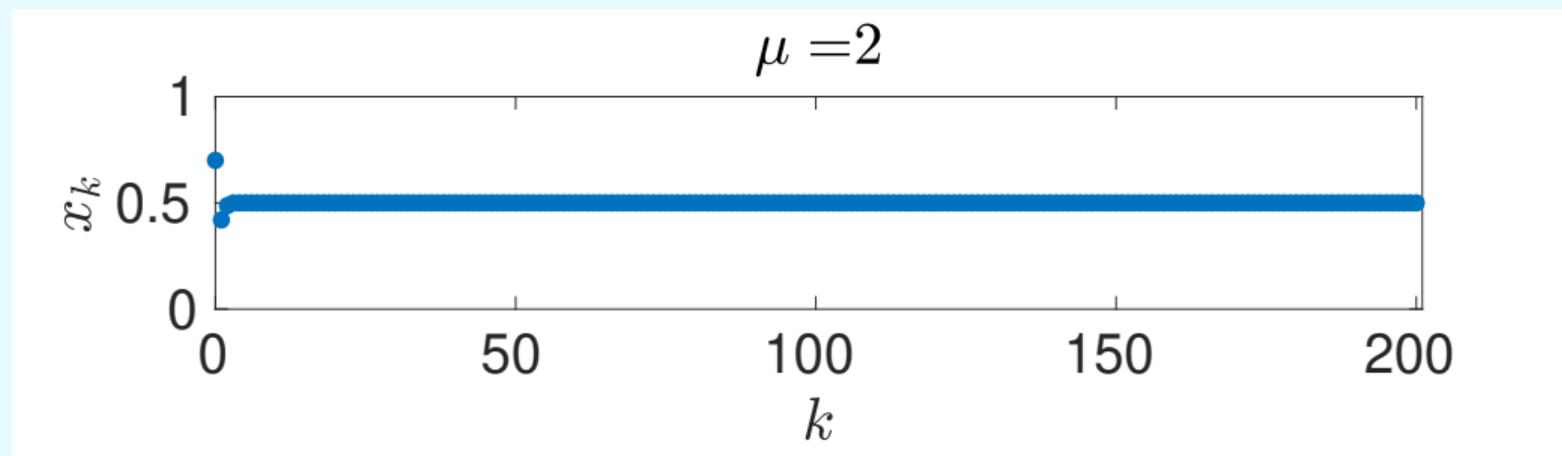
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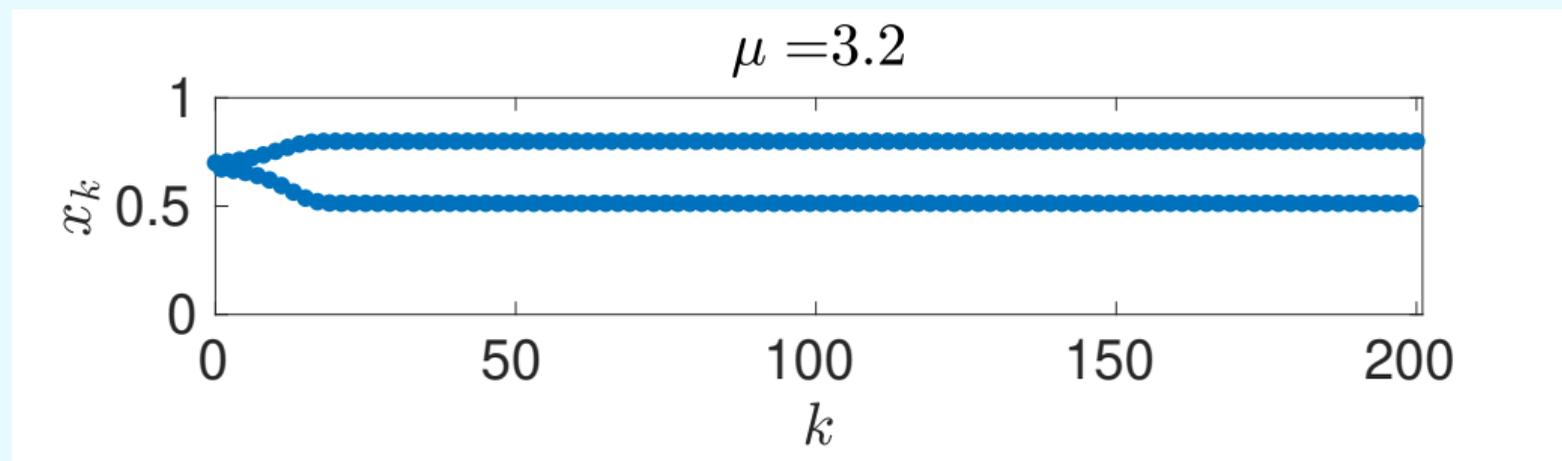
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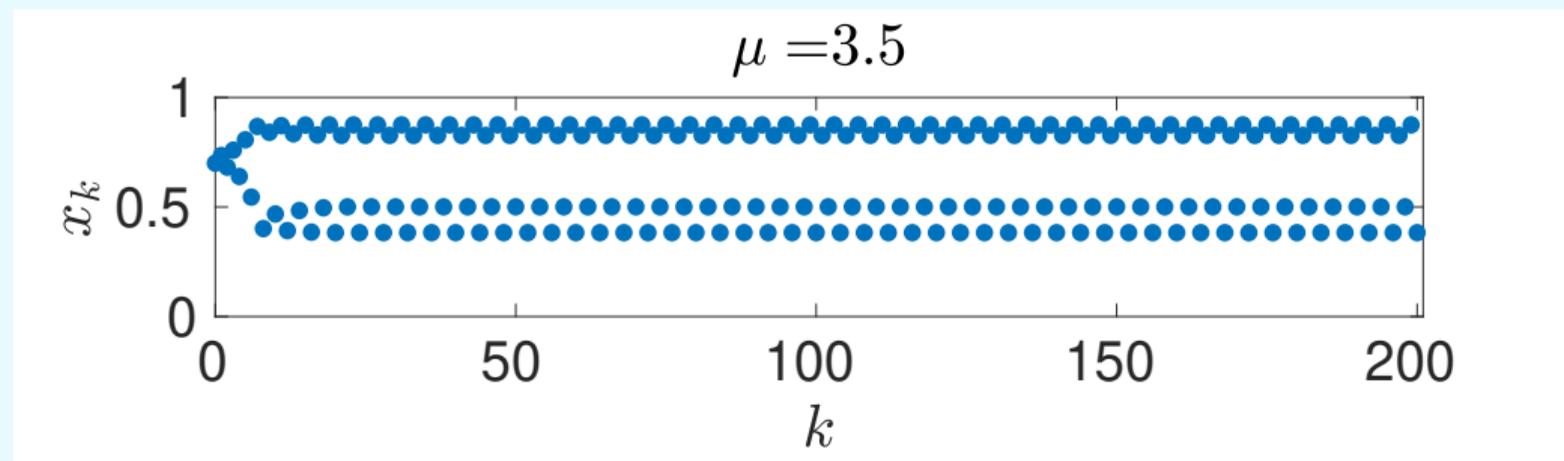
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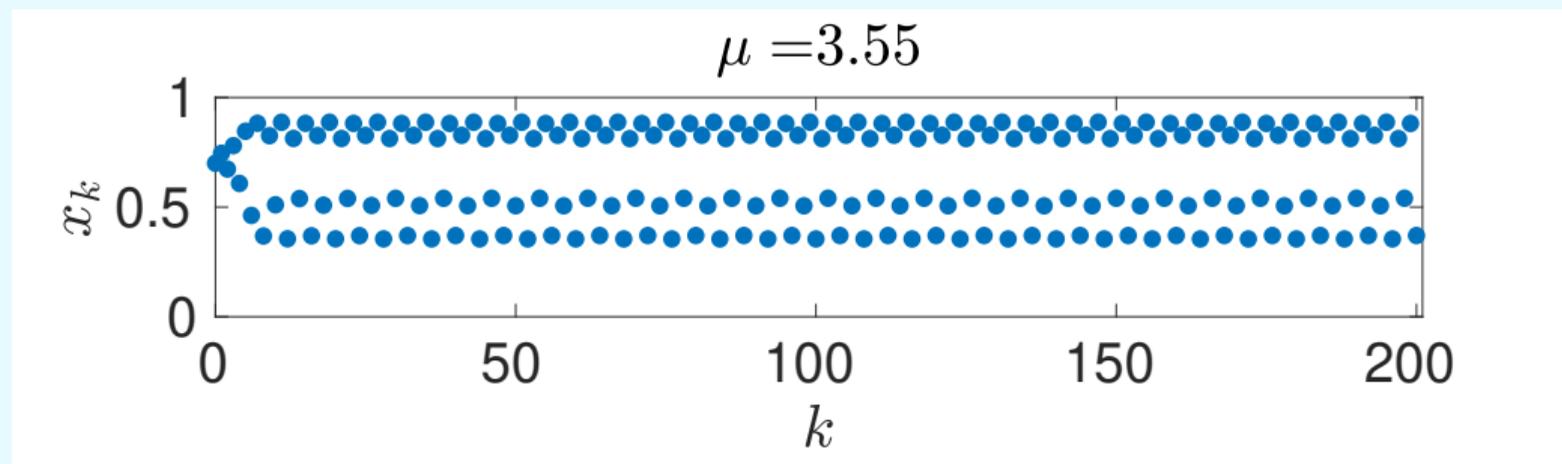
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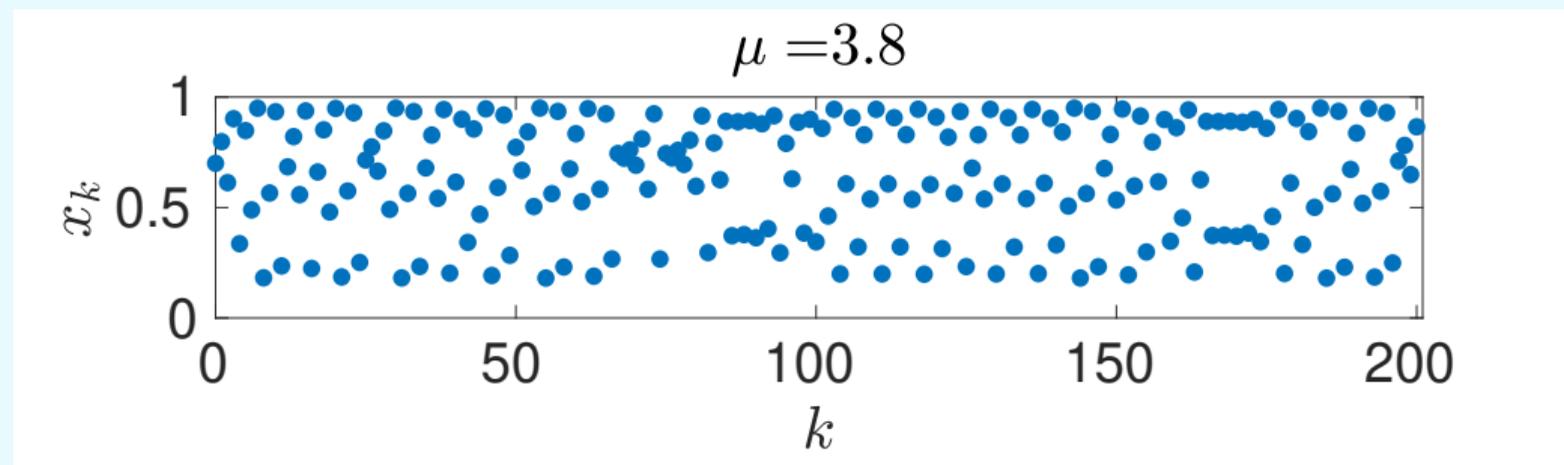
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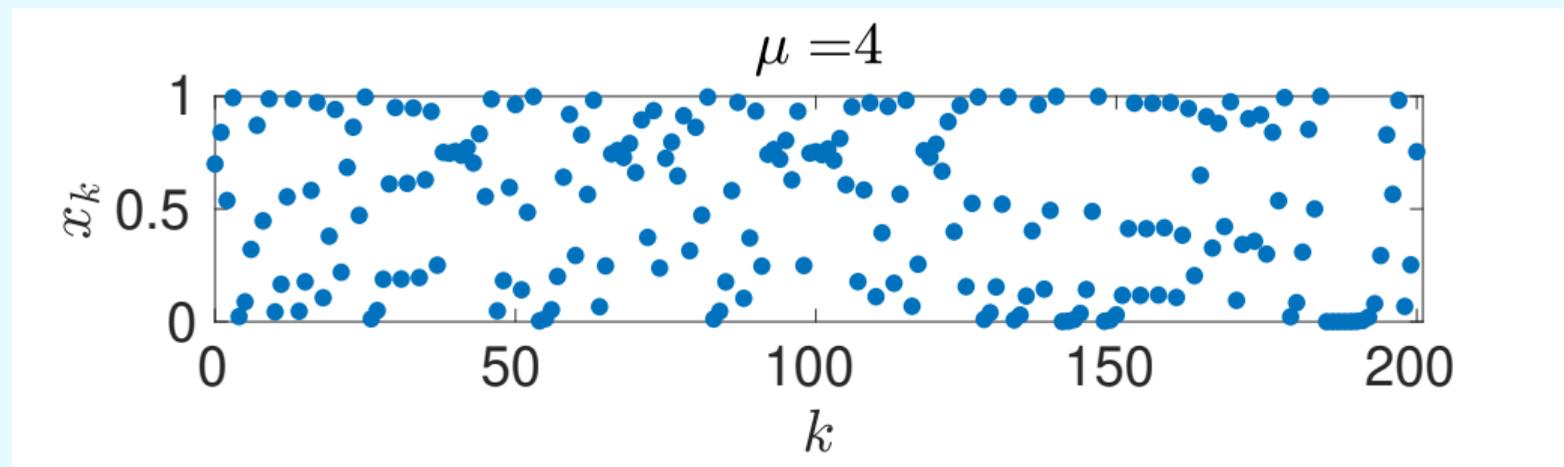
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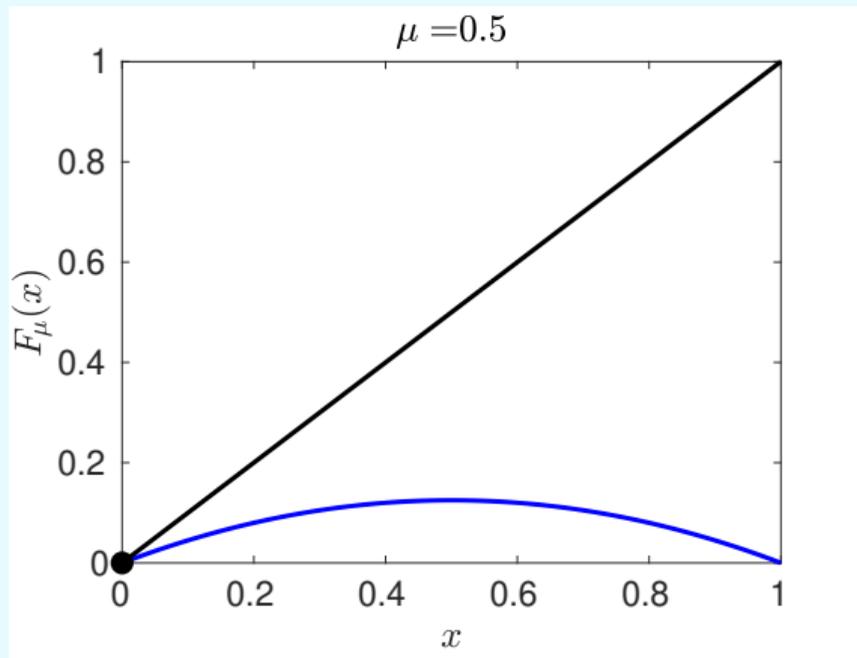
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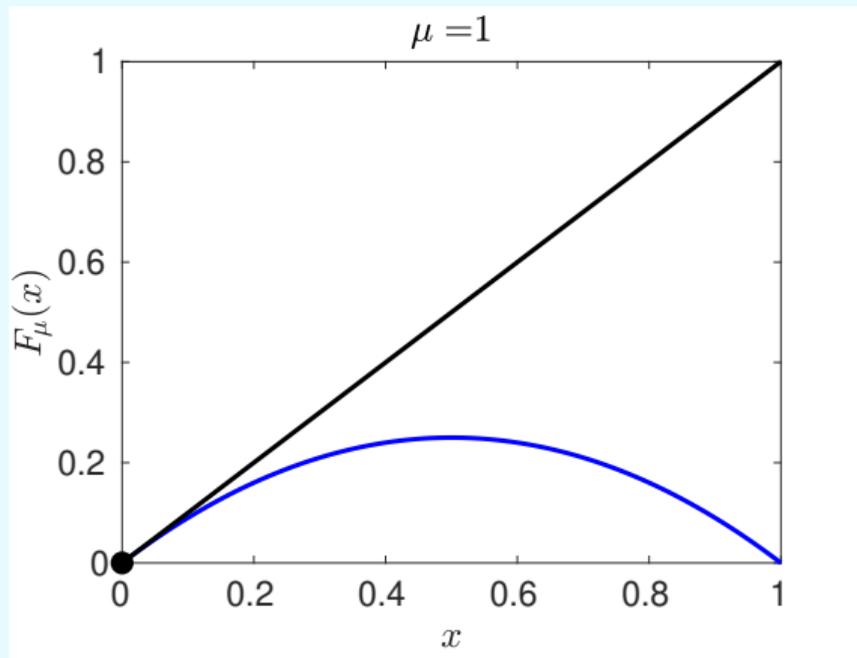
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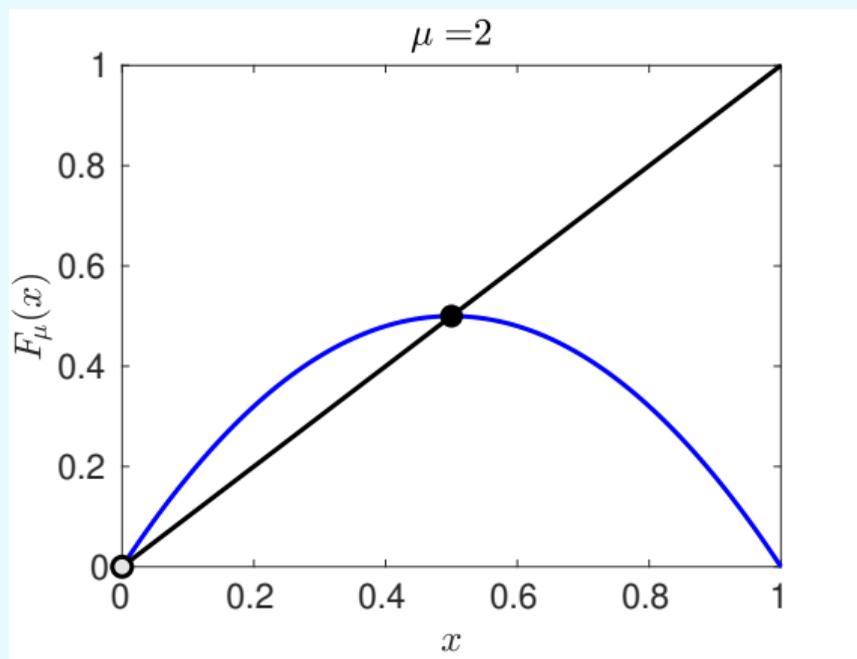
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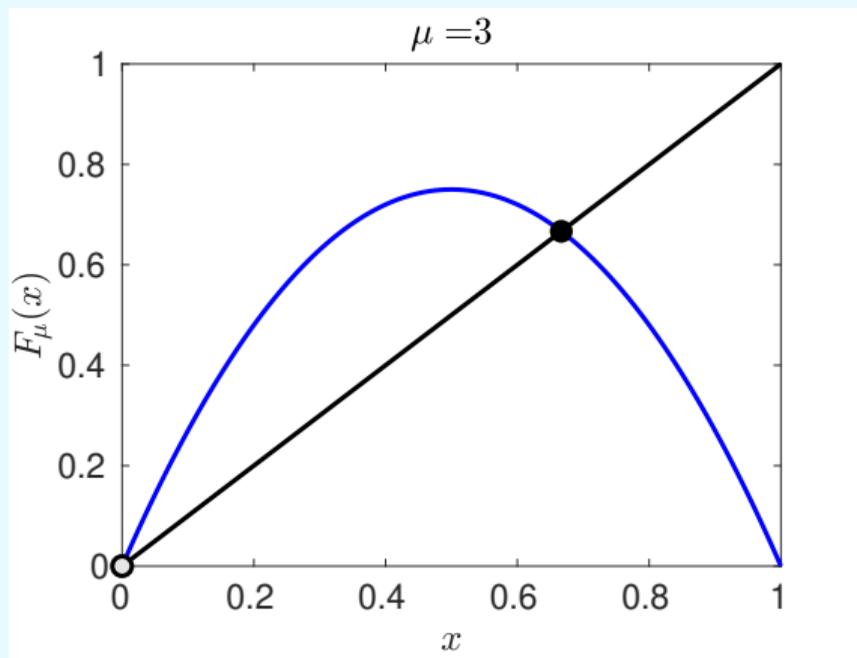
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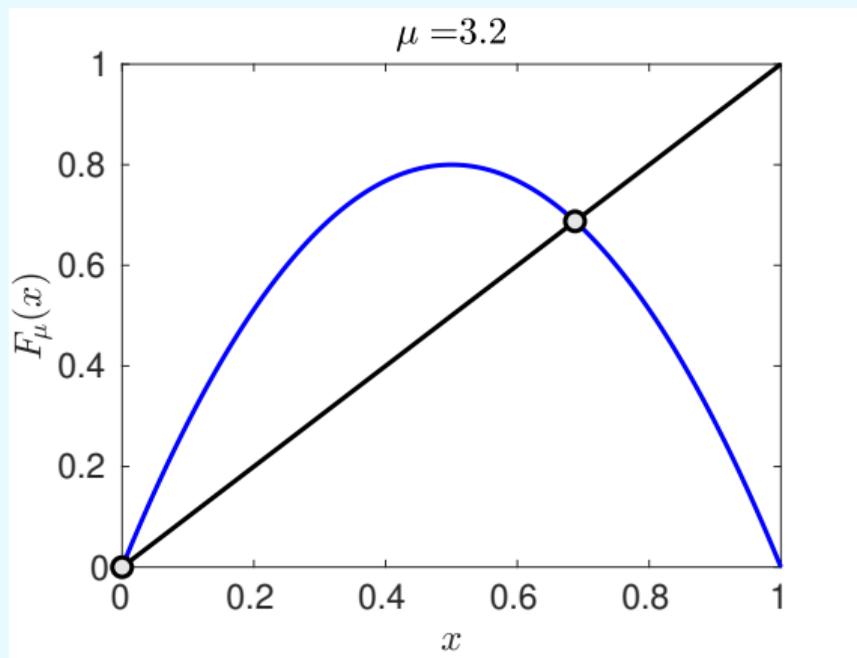
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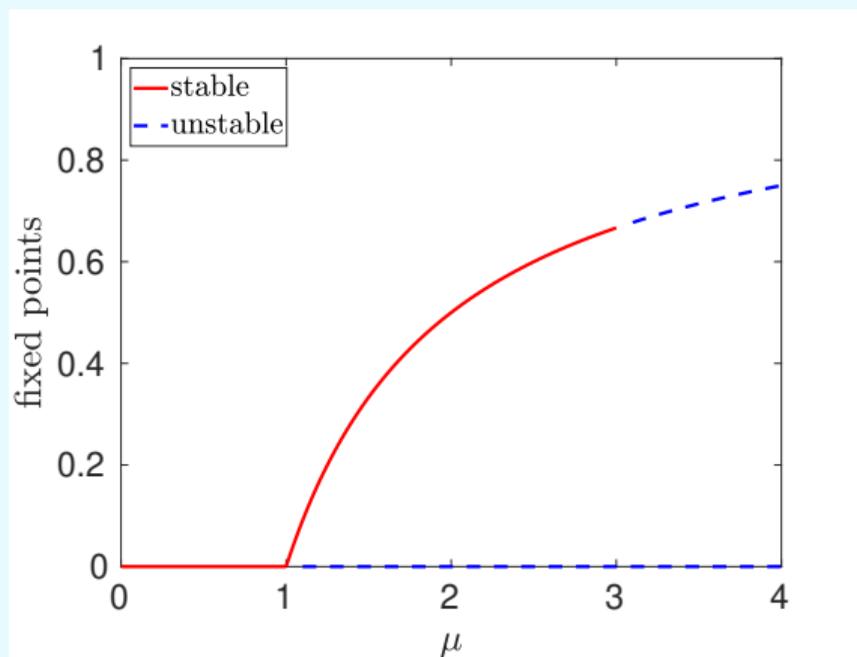
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B5.6: Nonlinear Dynamics, Bifurcations and Chaos (Lecture 7)

- summary of Lecture 6: we discussed Discrete-time (maps) dynamical systems. Fixed points. Periodic points of maps. N -cycles.
(Questions 3 and 4 on Problem Sheet 2)
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- course synopsis of **Lectures 1-8** (taken by both Part B and MSc students):
Discrete-time (maps) and continuous-time (differential equations) dynamical systems. Notion of flows, stability of fixed points, Lyapunov function, invariant manifolds, stable manifold theorem, notion of hyperbolicity, center manifold. Chemical reaction networks. Stable, unstable and center subspaces. Poincaré-Bendixson theorem. Periodic solutions, stable and unstable limit cycles. Introduction to bifurcation theory, covering saddle-node, transcritical, supercritical pitchfork and subcritical pitchfork bifurcations. Extended center manifold. Logistic map. Periodic points of maps. Stability of N -cycles. Period-doubling bifurcation. Sharkovsky's theorem. Invariant distribution.

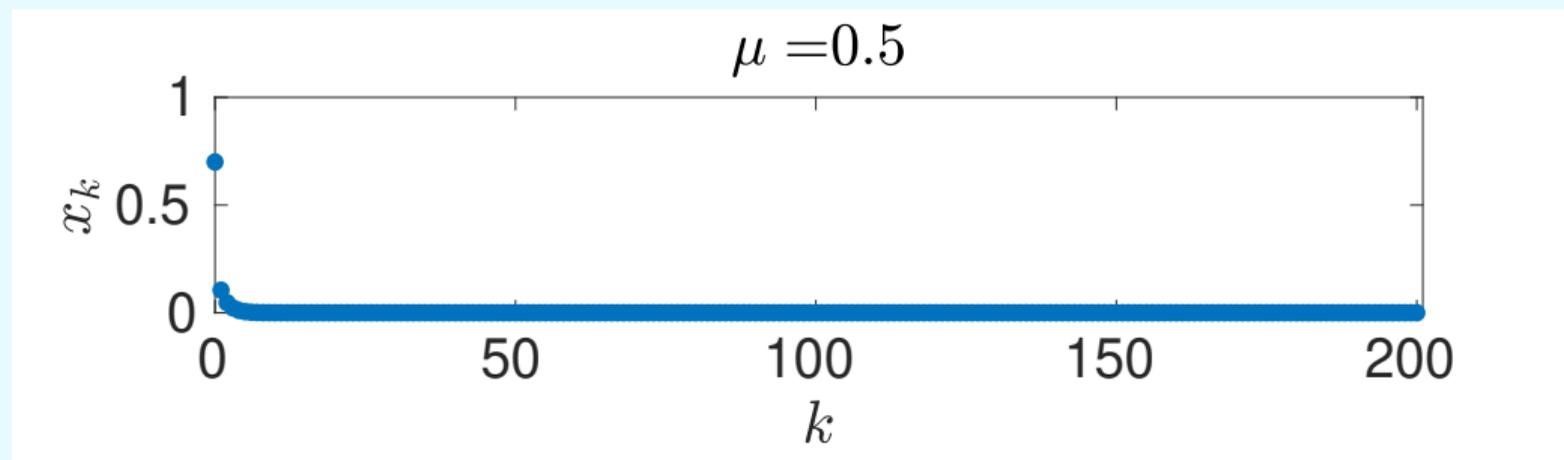
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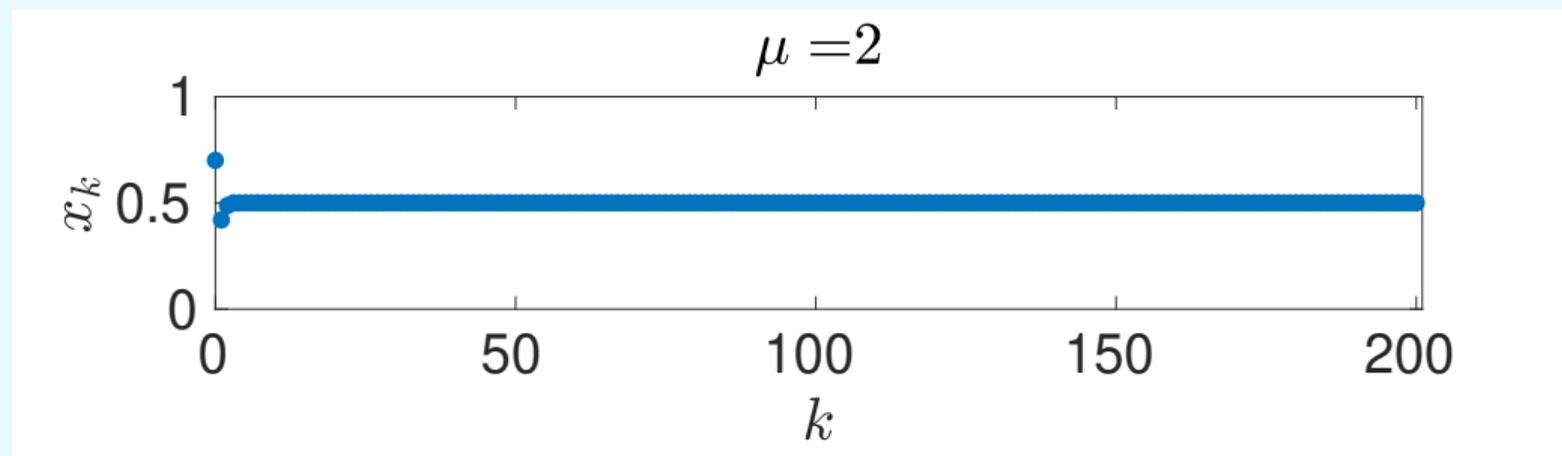
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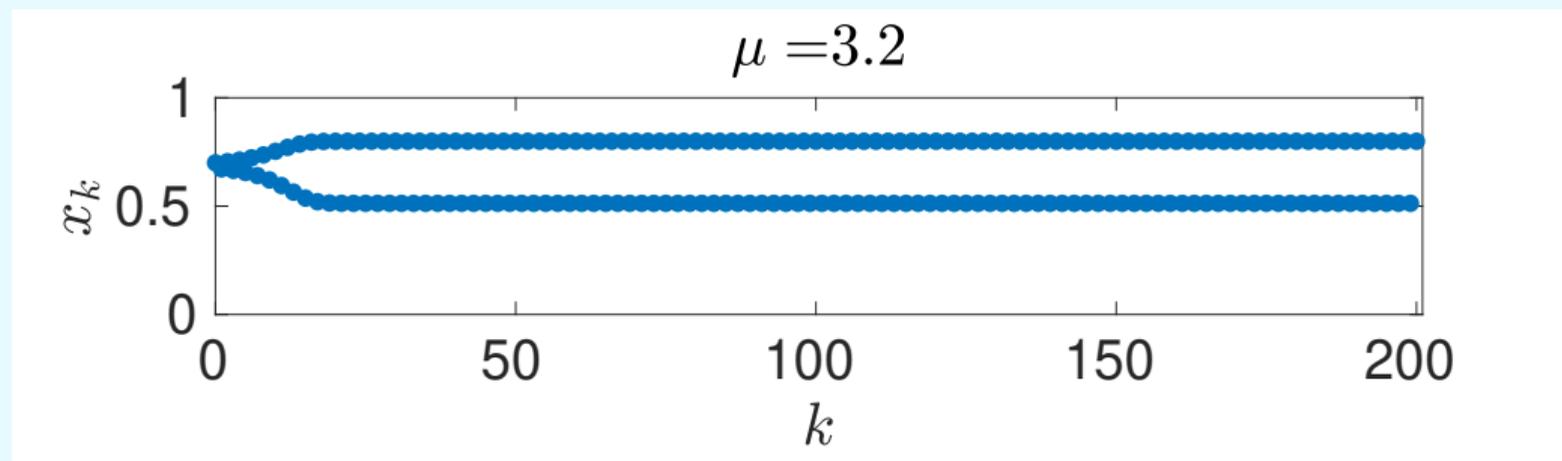
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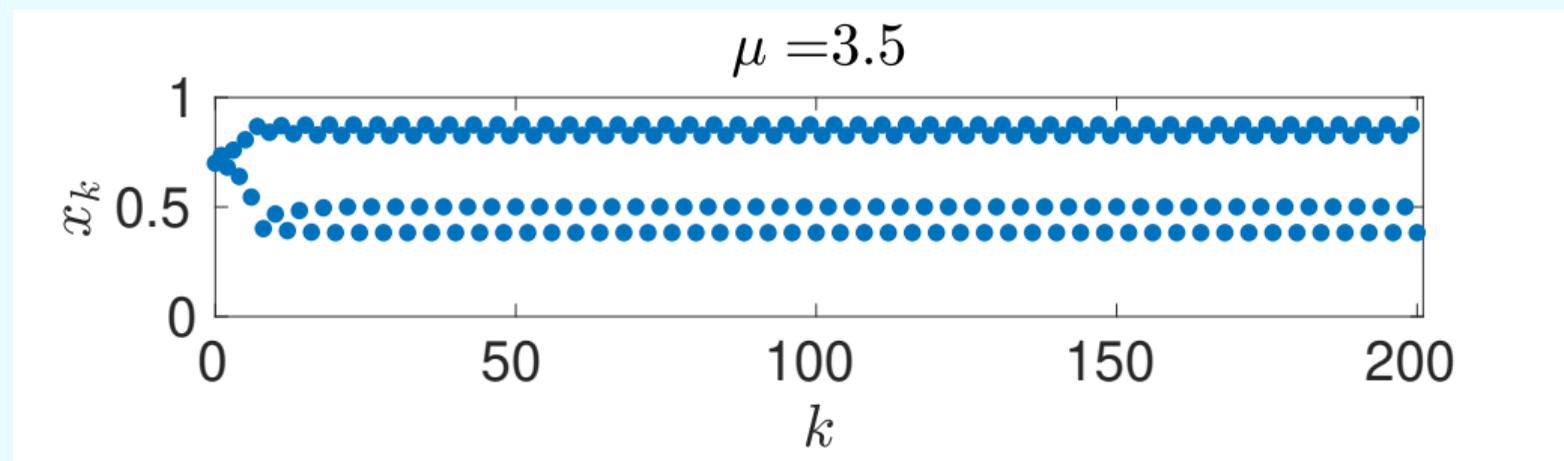
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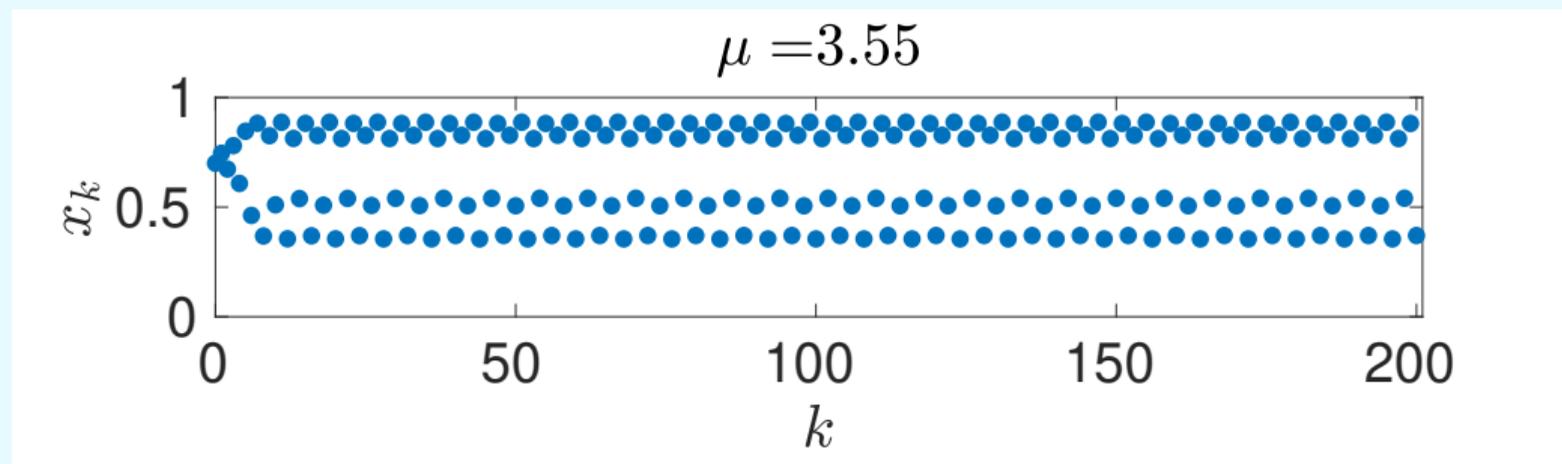
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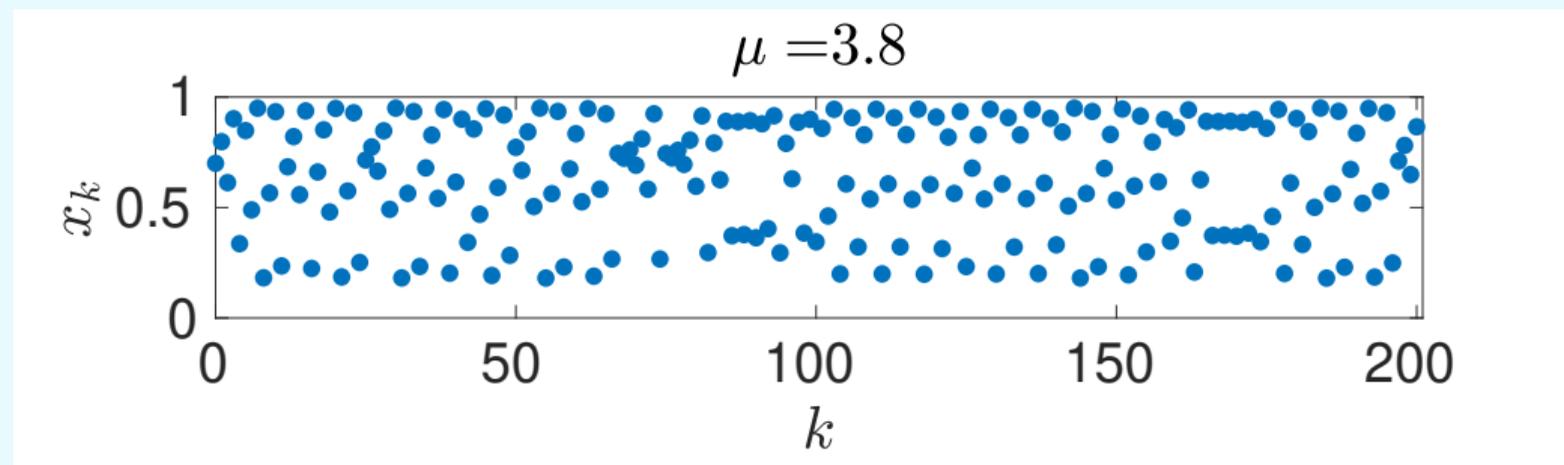
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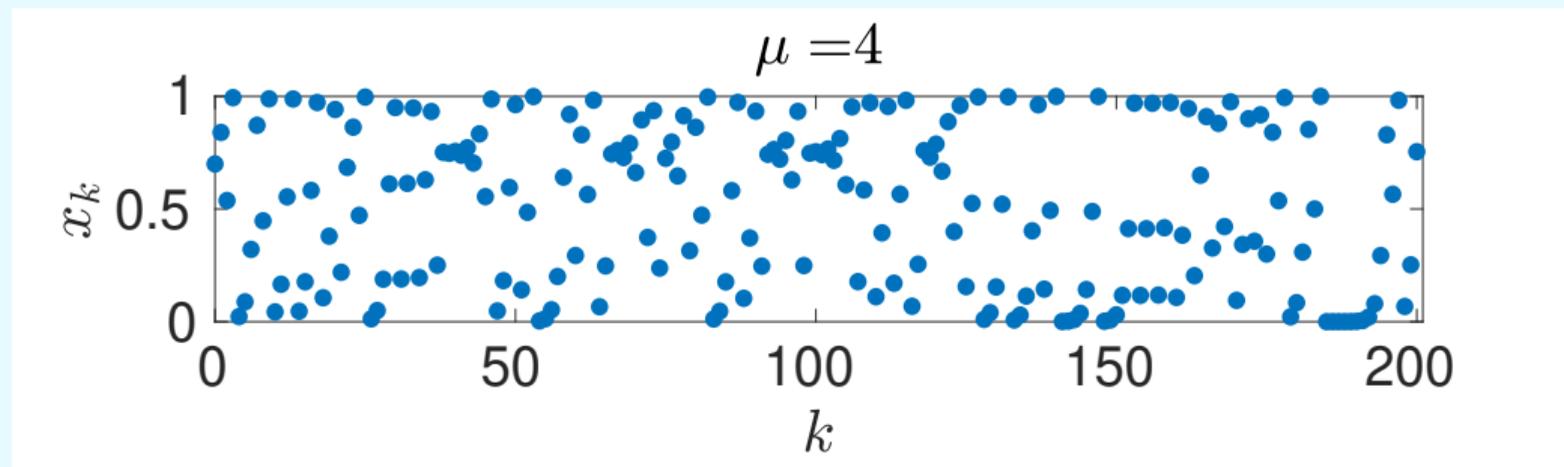
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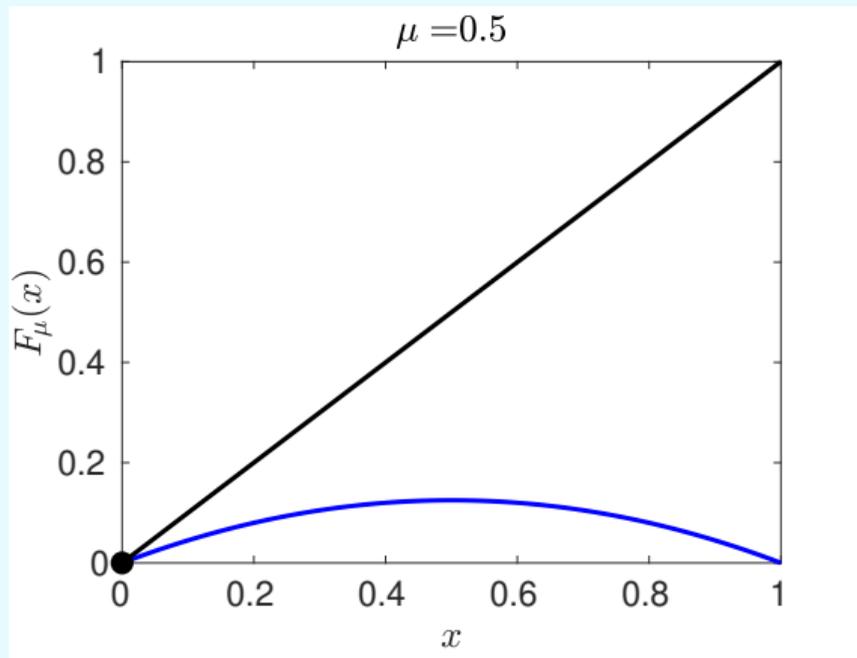
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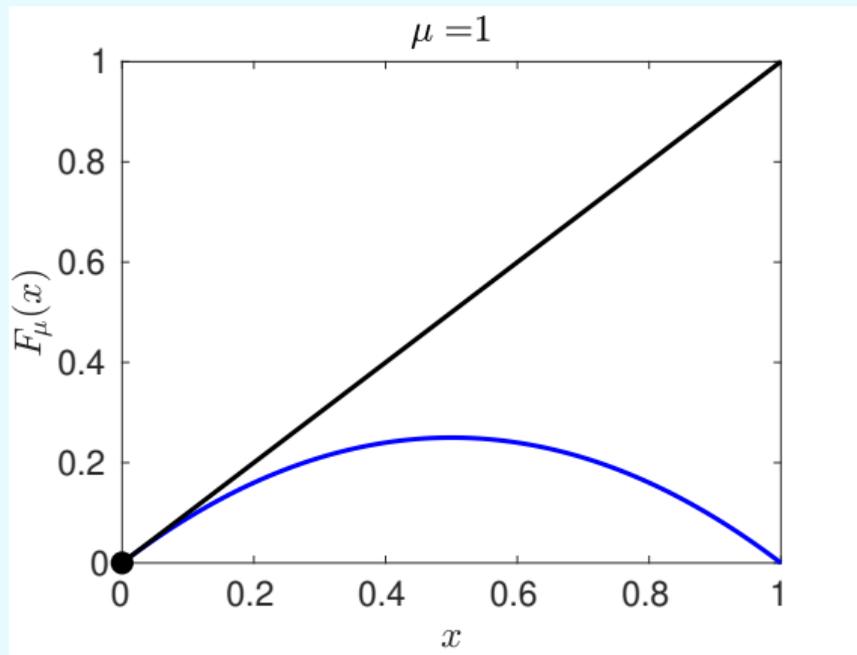
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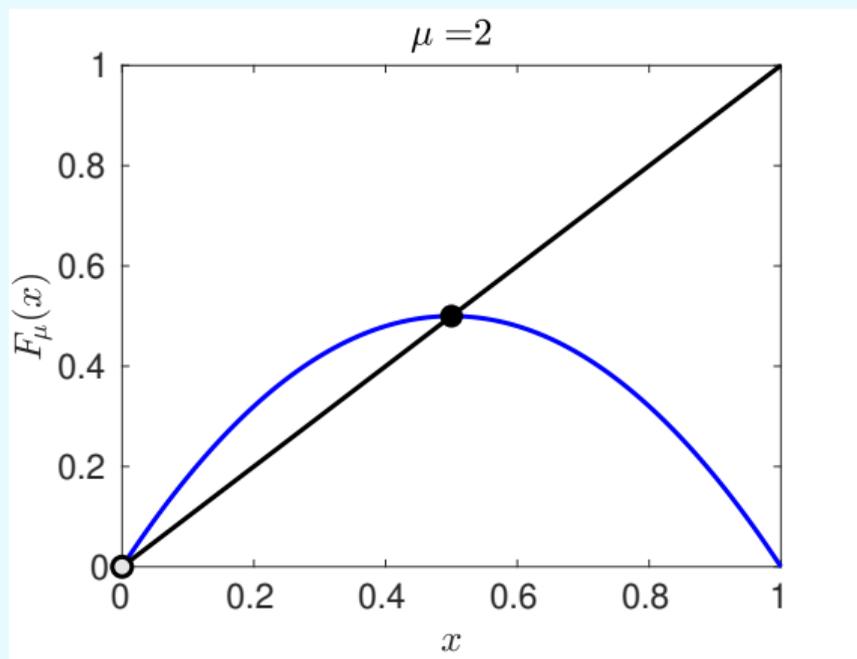
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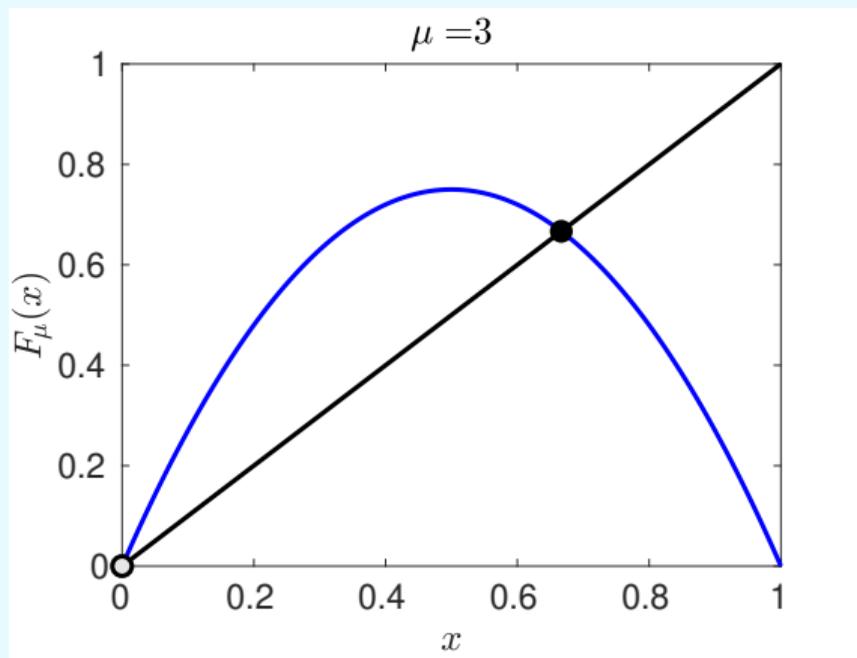
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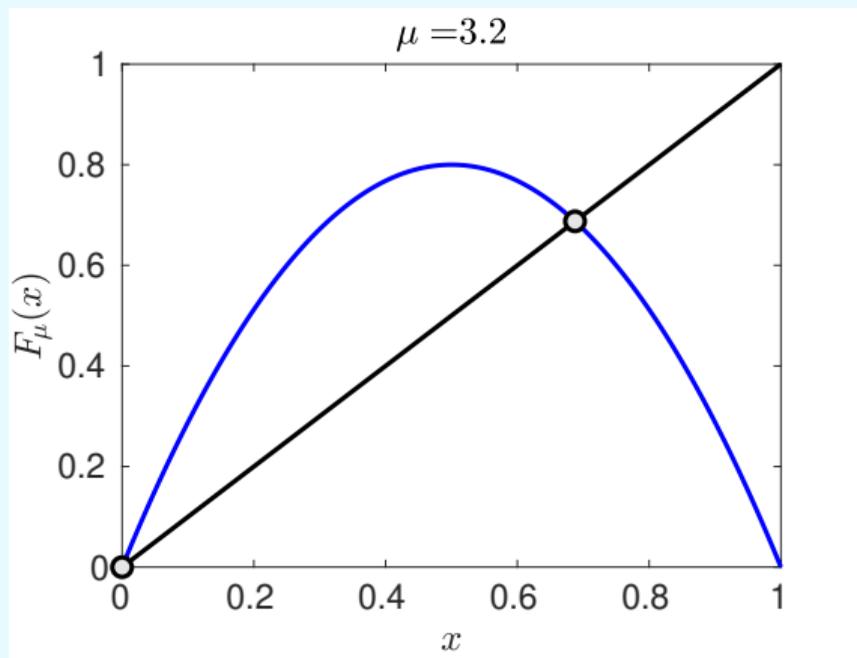
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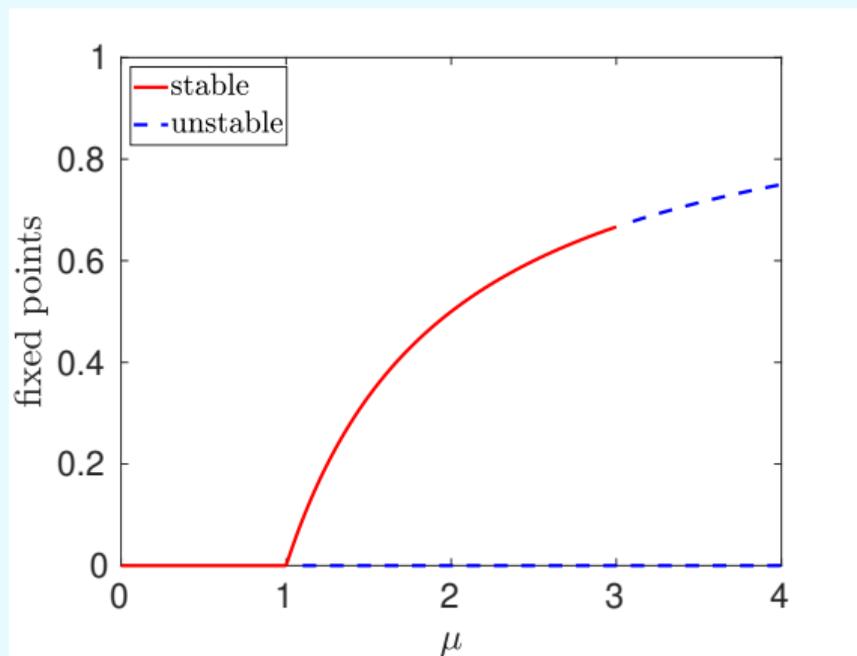
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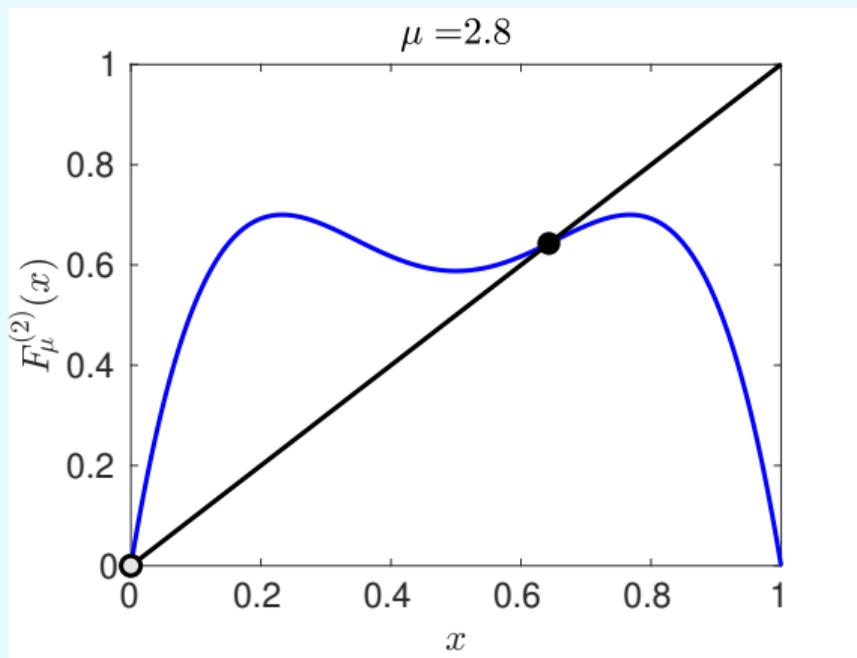


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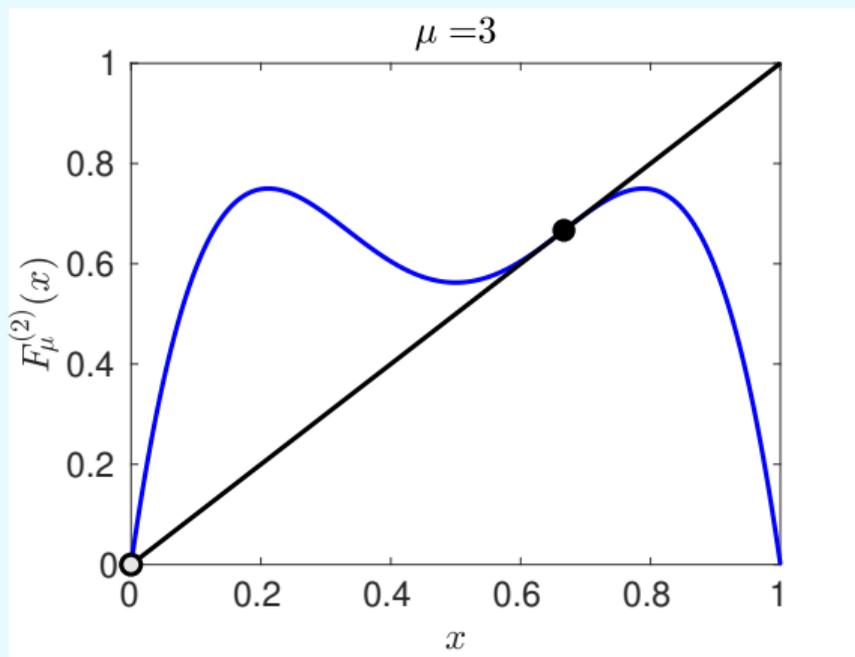
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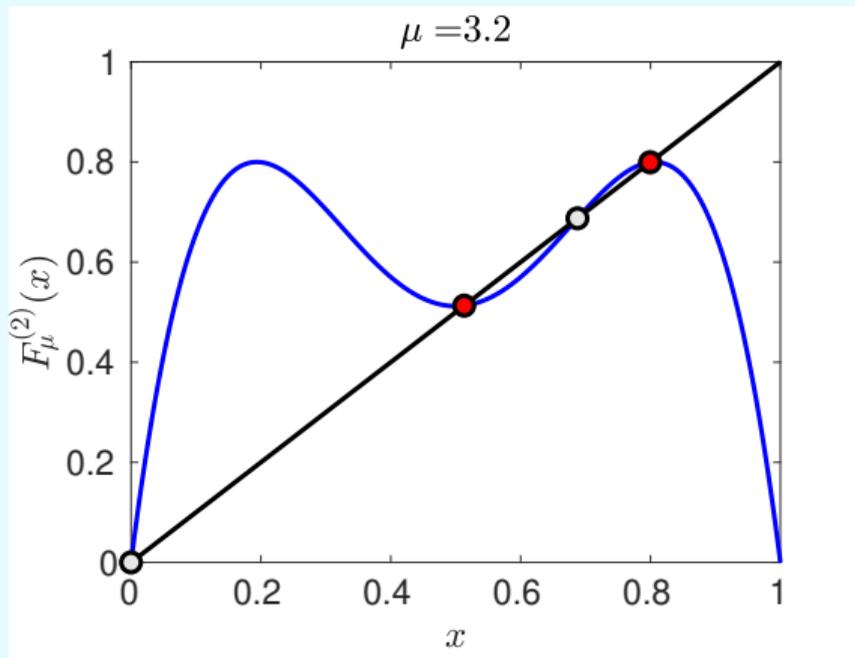
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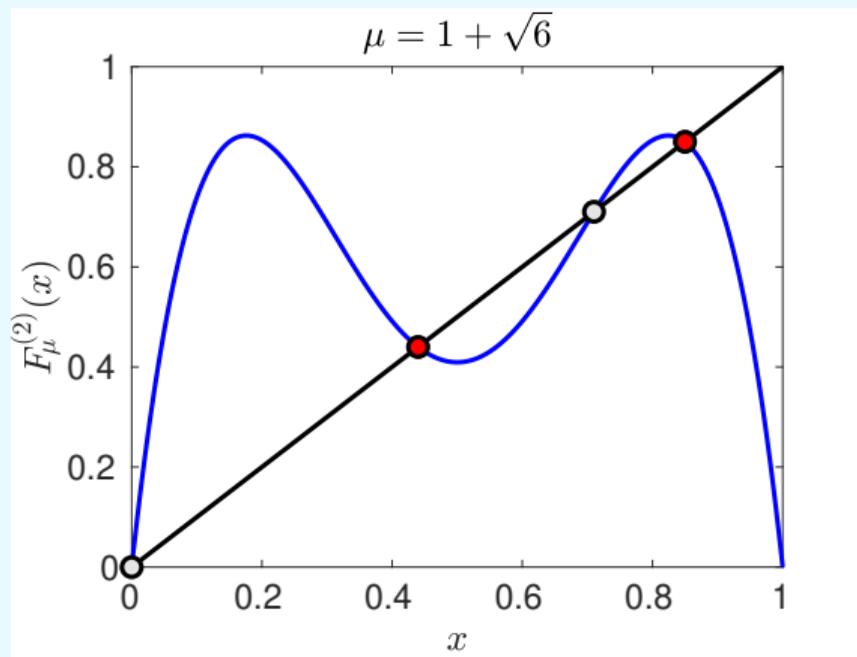
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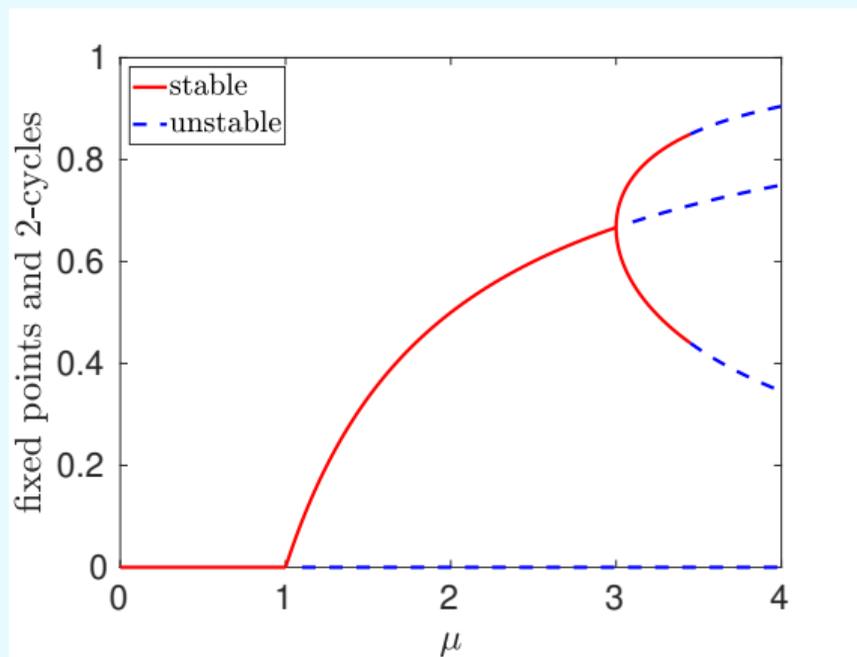
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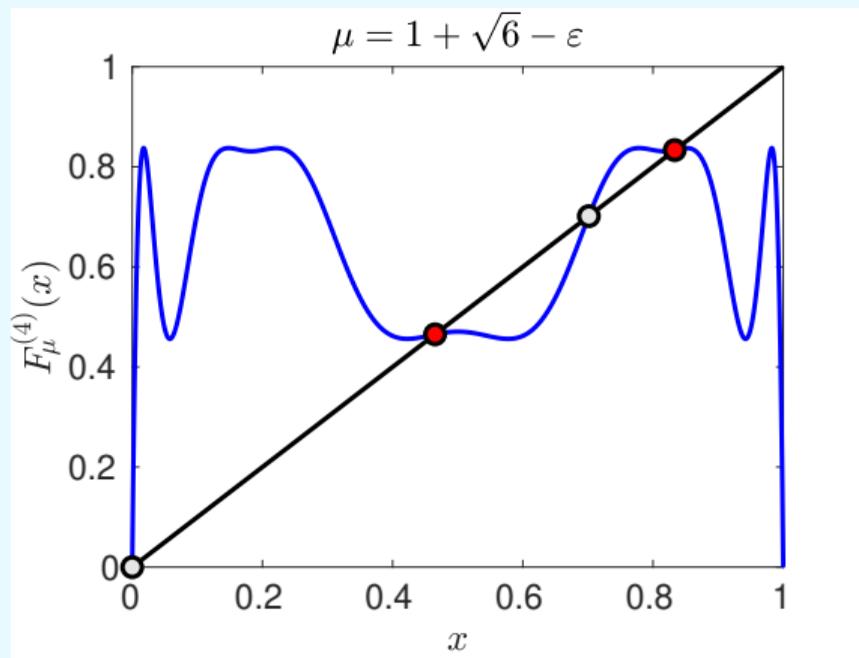
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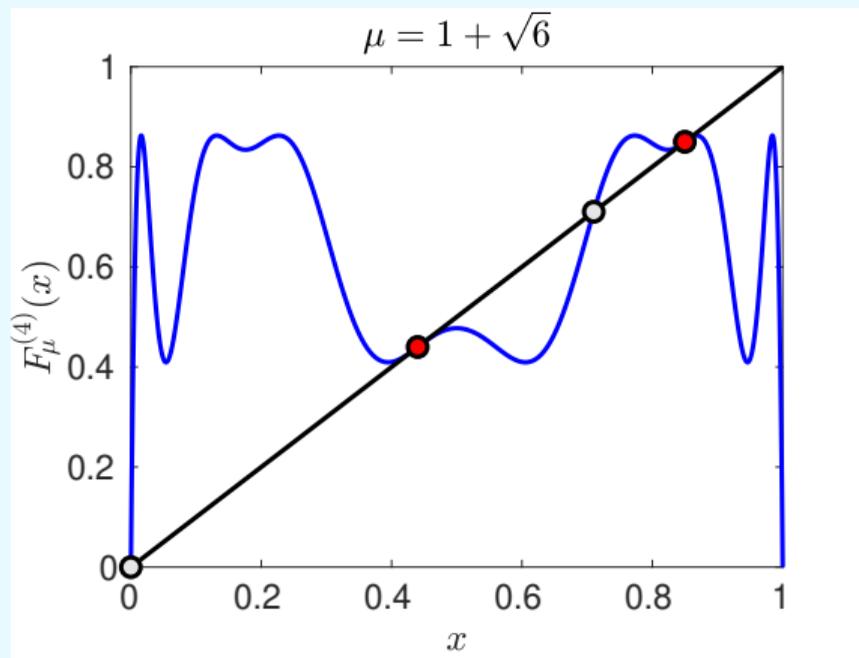
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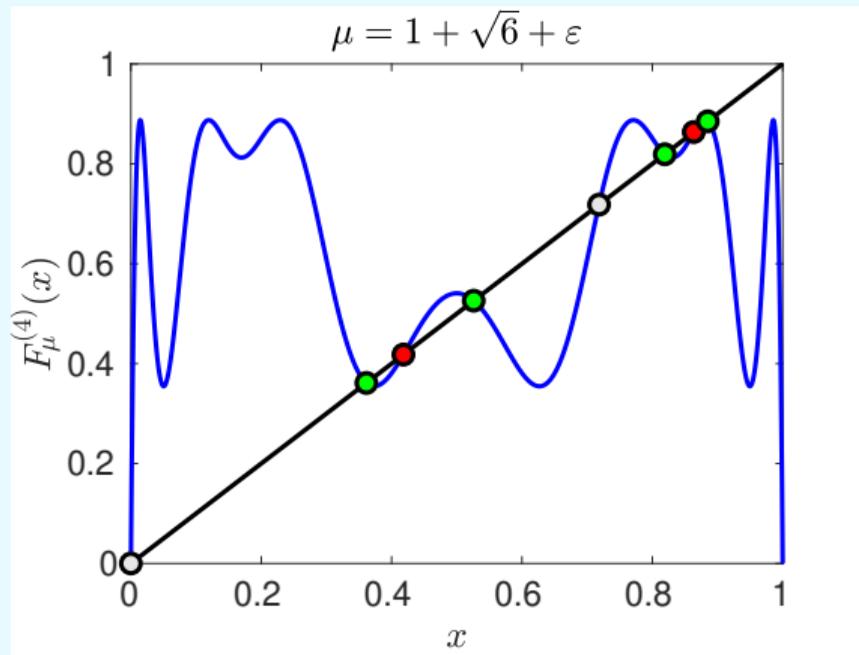


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4-cycle exists and is asymptotically stable for $\mu \in (1 + \sqrt{6}, 3.544090 \dots]$



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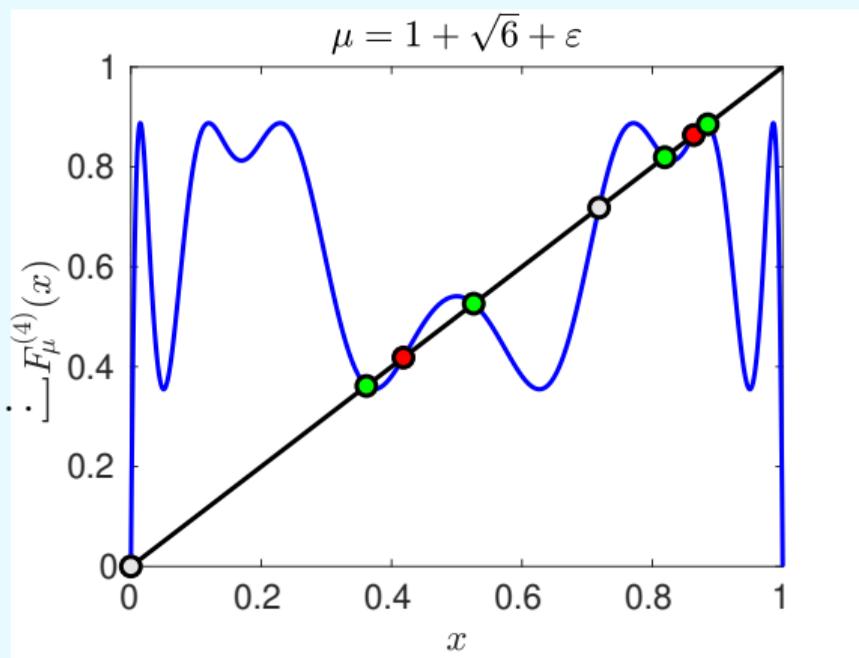
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8-cycle exists and is asymptotically stable for $\mu \in (3.544090 \dots, 3.564407 \dots]$

this is called the **period doubling route to chaos**

additional example:

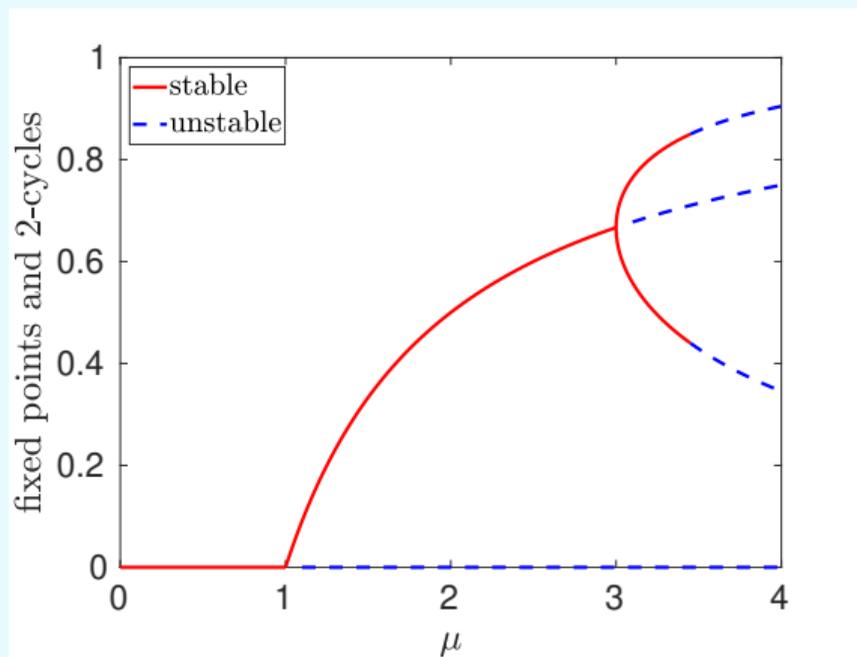
Question 3 on Problem Sheet 2



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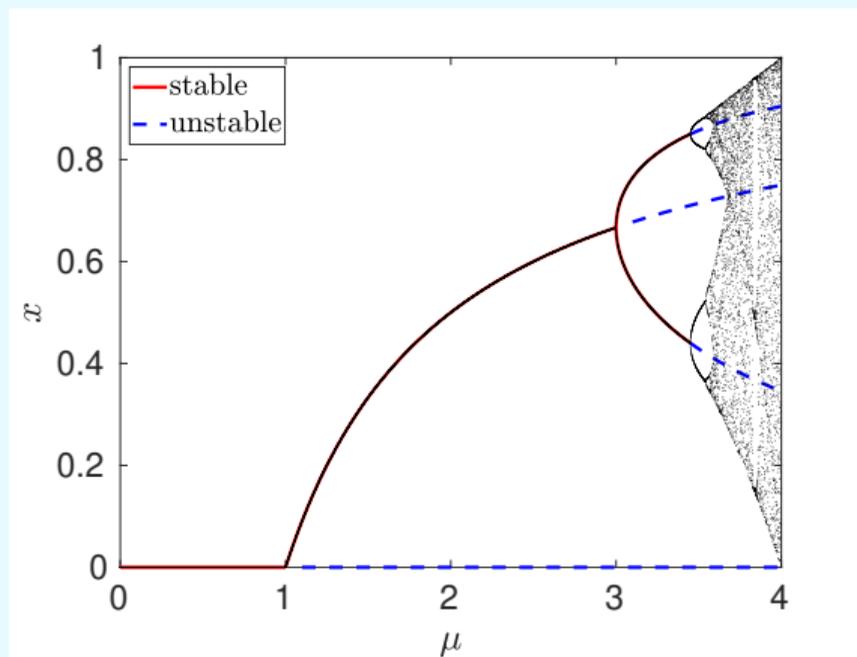
bifurcation diagram



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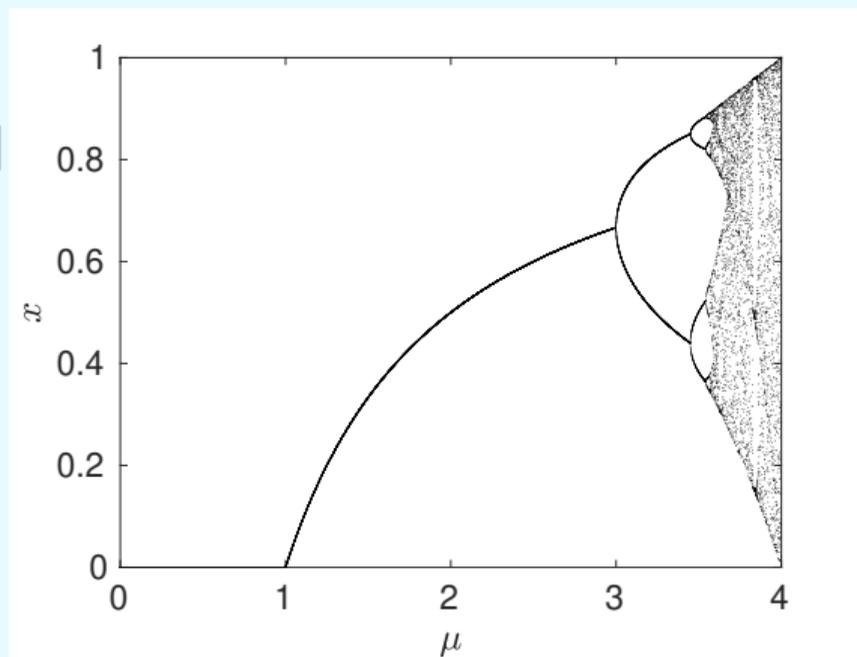
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asymptotically stable 2-cycle exists
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16-cycle, 32-cycle, 64-cycle, ...



period doubling route to chaos

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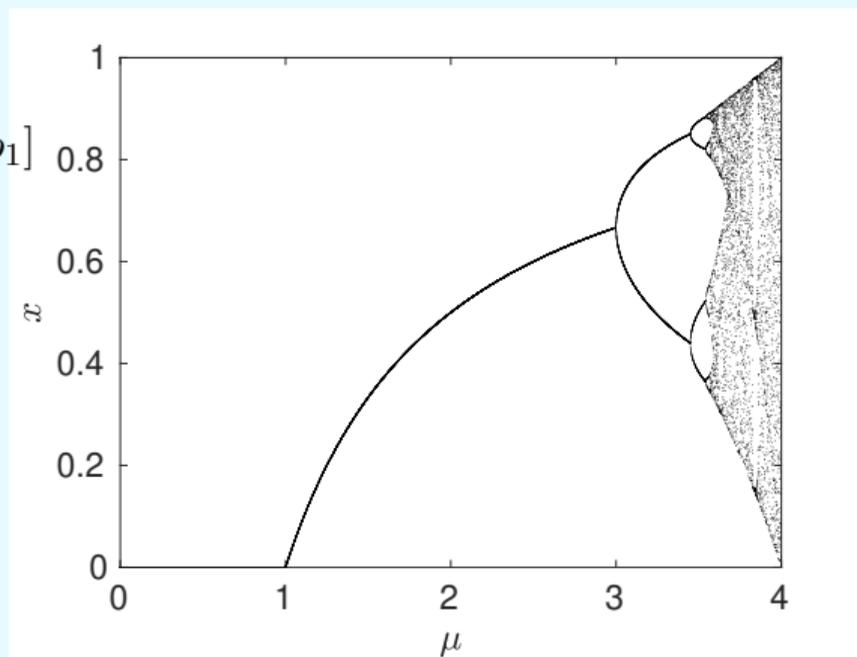
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asymptotically stable 2-cycle exists
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asymptotically stable 2^k -cycle exists
for $\mu \in (b_k, b_{k+1}]$

Feigenbaum's constant:

$$\lim_{k \rightarrow \infty} \frac{b_k - b_{k-1}}{b_{k+1} - b_k} = 4.6692016 \dots$$



additional example: [Question 3 on Problem Sheet 2](#)

Example (from Lecture 6)

$$x_{k+1} = (1 - x_k)(1 - 5x_k + \mu x_k^2)$$

$$F(x; \mu) = (1 - x)(1 - 5x + \mu x^2)$$

If $\mu \in \Theta = [6.3, 11.8]$,
then $F(x; \mu) \in [0, 1]$ for all $x \in [0, 1]$.

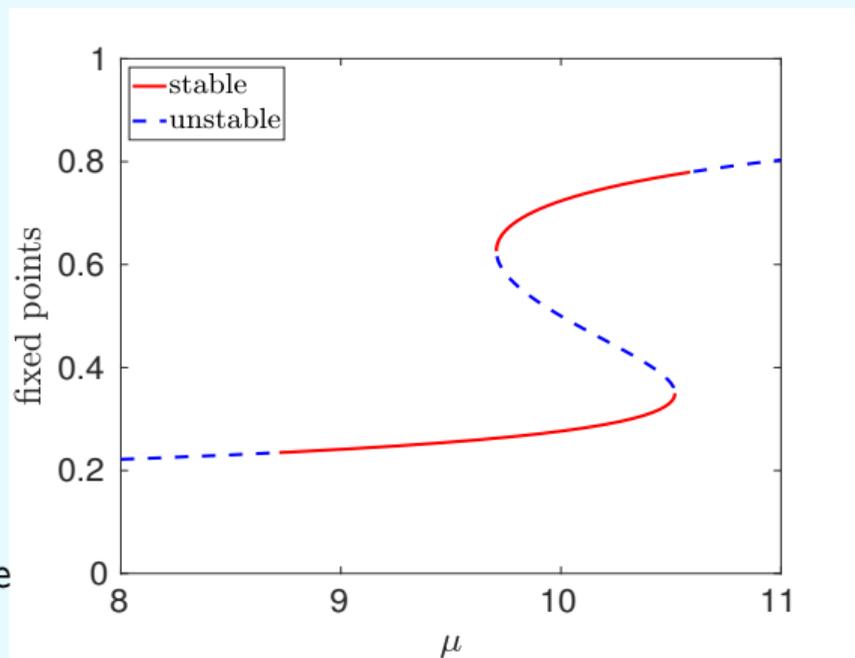
We have studied dynamics
of $F : \Omega \times \Theta \rightarrow \Omega$, where $\Omega = [0, 1]$.

three fixed points for $\mu \in (\mu_1, \mu_2)$ where
 $\mu_1 = 9.7066\dots$ and $\mu_2 = 10.518\dots$

one fixed point for $\mu < \mu_1$ and $\mu > \mu_2$

we have **saddle-node bifurcations** at $\mu = \mu_1$ and $\mu = \mu_2$

we also have **period doubling bifurcations** at $\mu \approx 8.71988\dots$ and $\mu \approx 10.5877\dots$



Example (from Lecture 6) – bifurcation diagram

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If $\mu \in \Theta = [6.3, 11.8]$,
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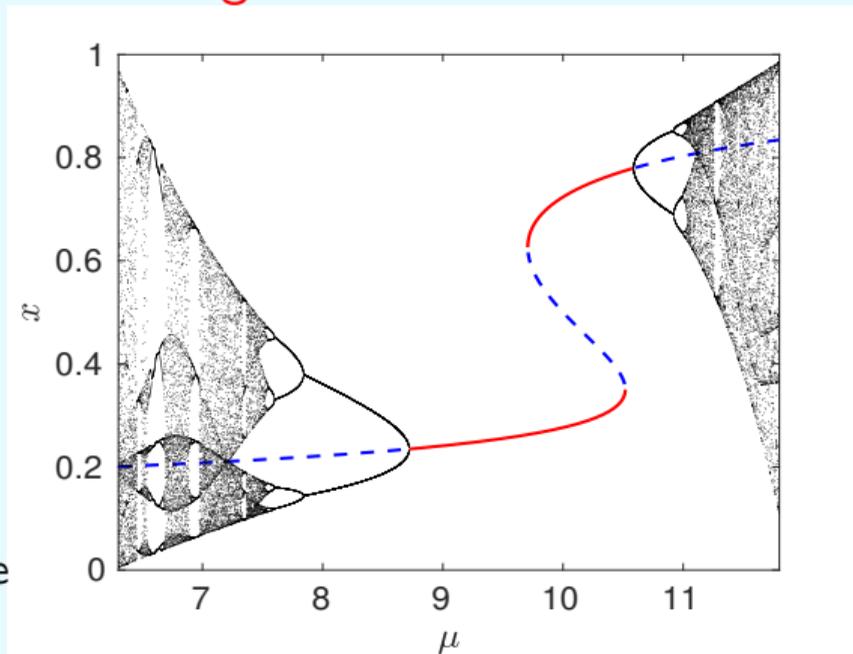
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3-cycles: logistic map $x_{k+1} = \mu x_k (1 - x_k)$

asymptotically stable 2^k -cycle exists

for $\mu \in (b_k, b_{k+1}]$ where

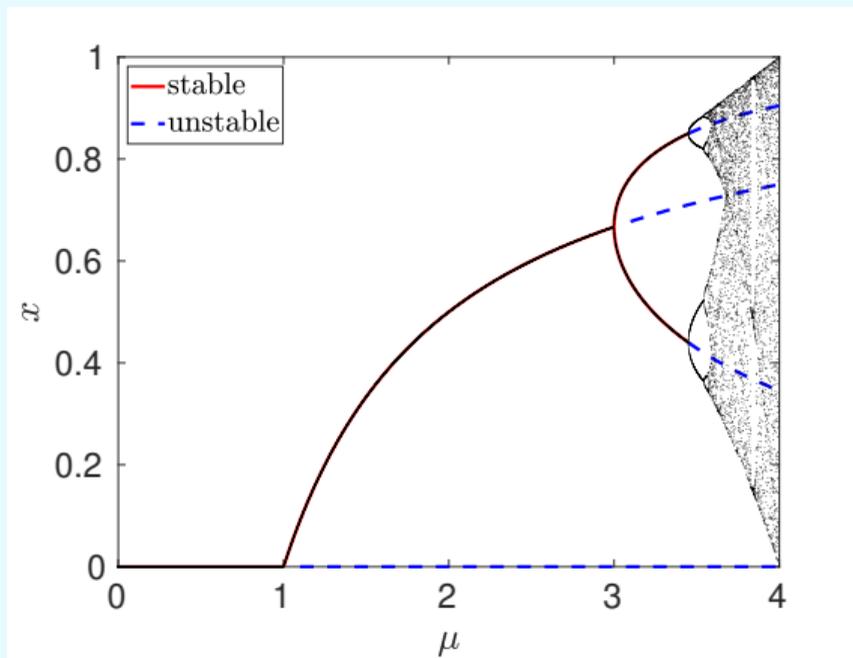
$$b_1 = 3$$

$$b_2 = 1 + \sqrt{6}$$

$$b_3 = 3.544090\dots$$

$$b_4 = 3.564407\dots$$

$$\lim_{k \rightarrow \infty} b_k = 3.56994567\dots$$



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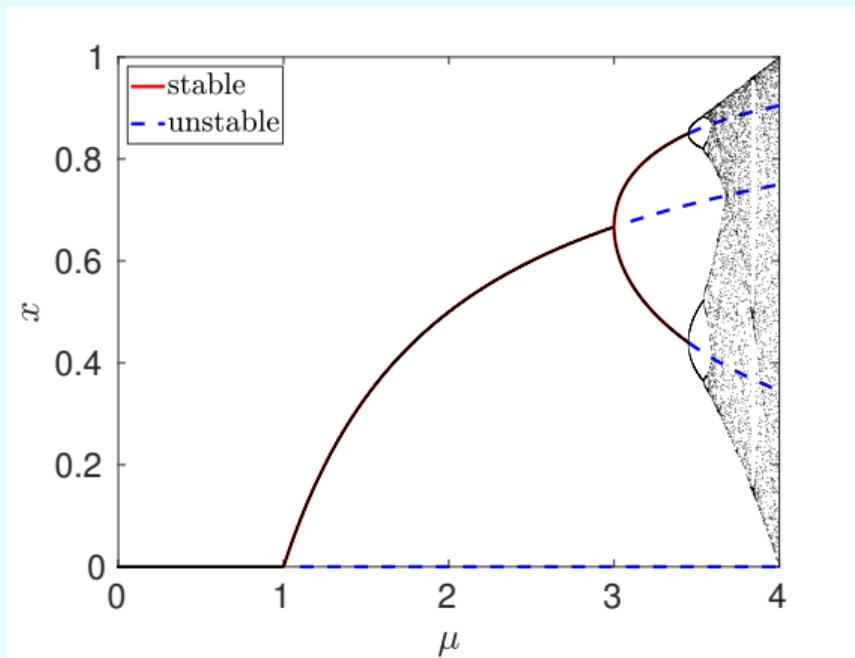
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3-cycles: solve $x = F_\mu^{(3)}(x)$



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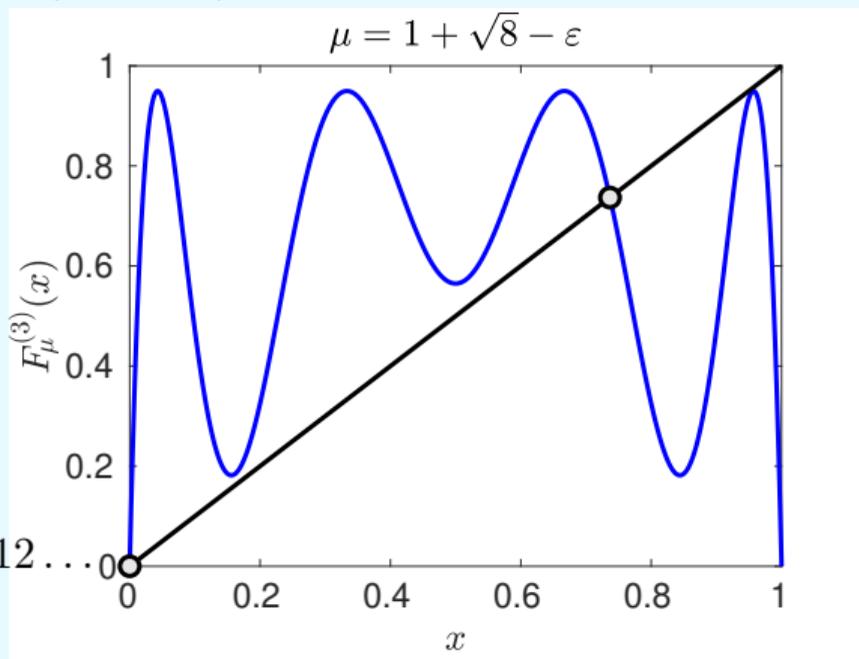
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no 3-cycles for $\mu < 1 + \sqrt{8} = 3.82842712\dots$



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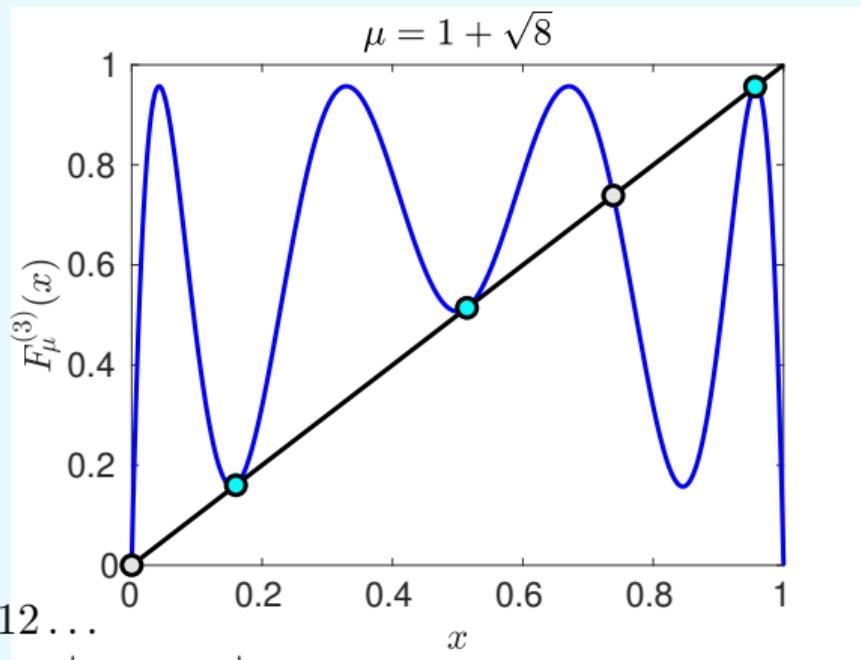
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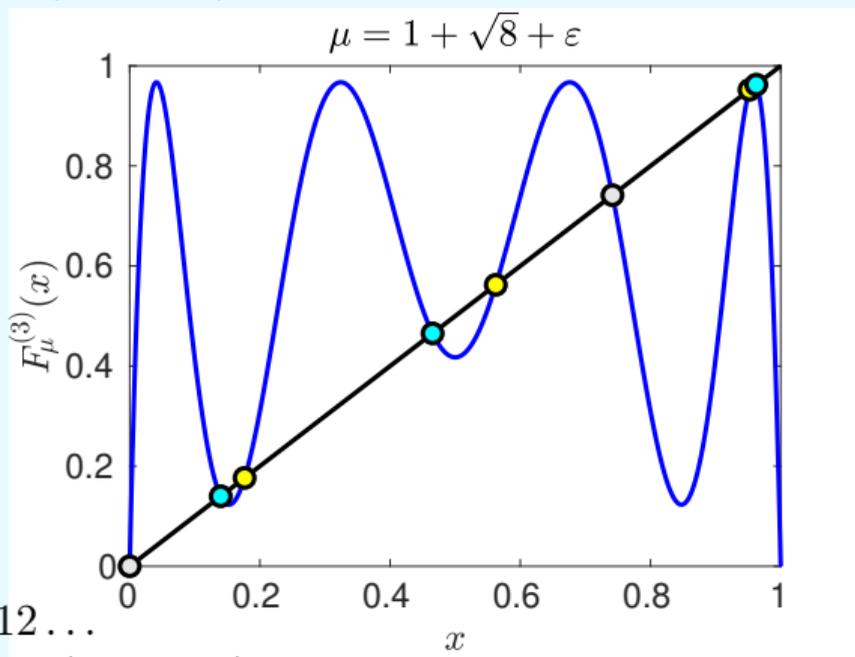
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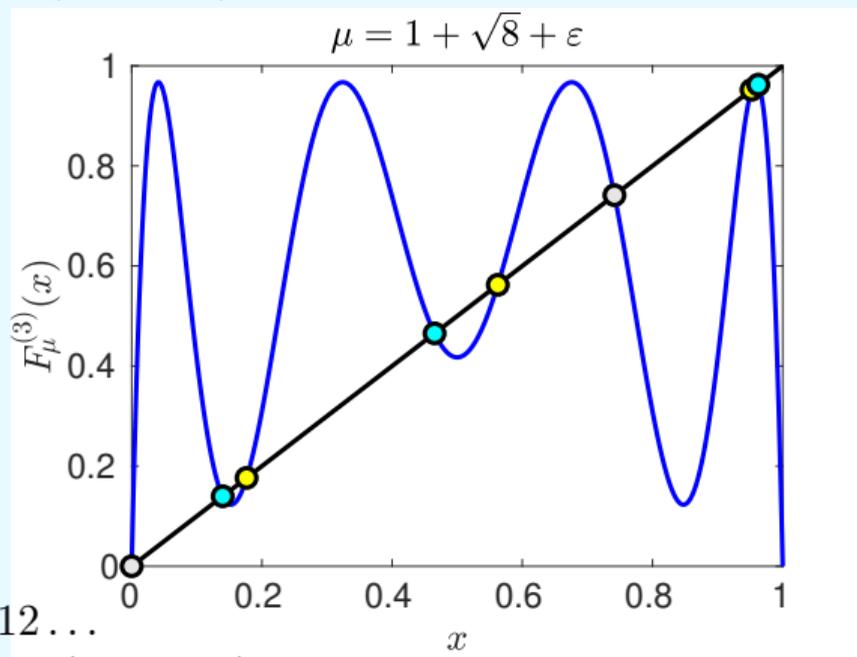
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'cyan 3-cycle' is stable for $\mu = 1 + \sqrt{8} + \varepsilon$ for sufficiently small ε

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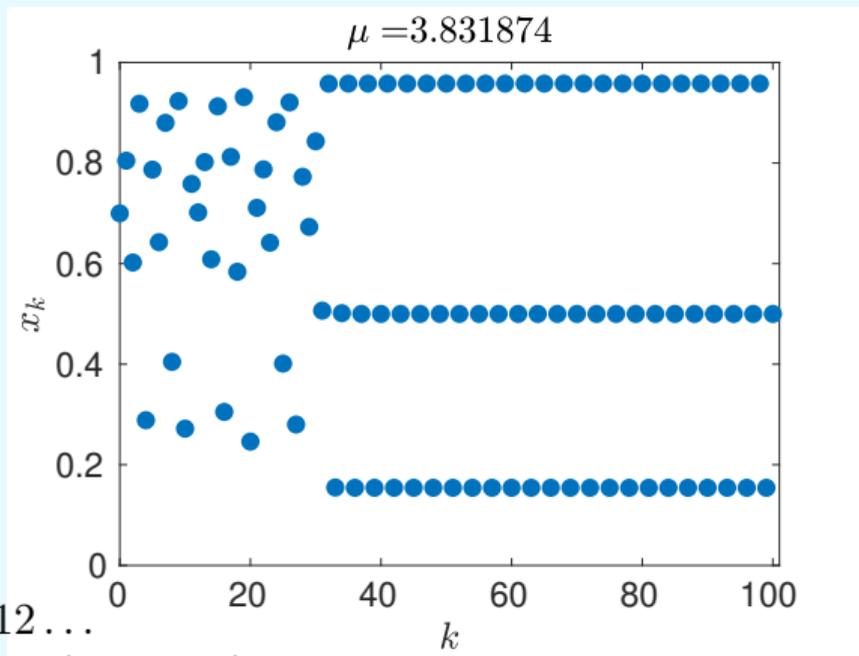
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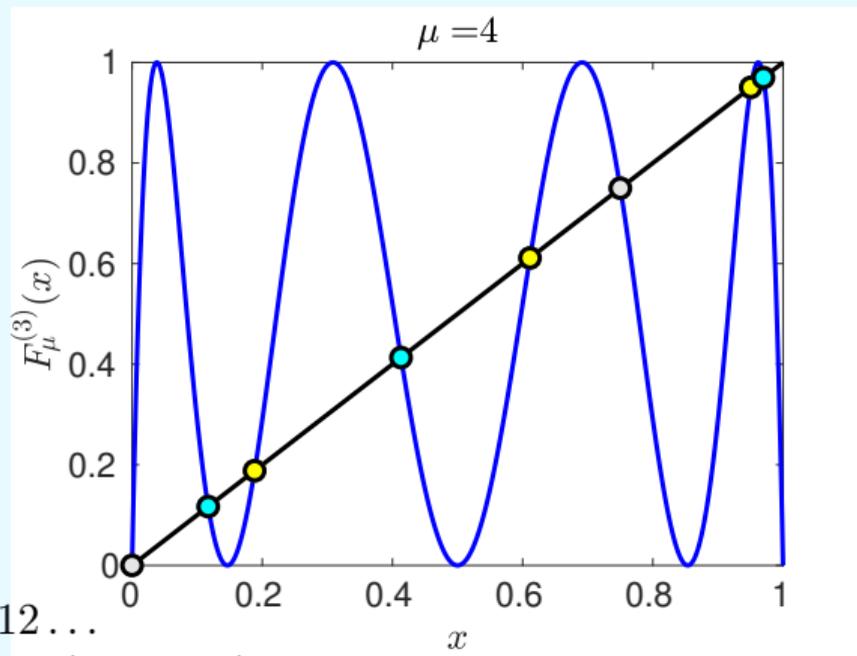
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two 3-cycles for $\mu \in (1 + \sqrt{8}, 4]$... 'cyan 3-cycle' and 'yellow 3-cycle'

Question 4 on Problem Sheet 2: closed formulas for both 3-cycles derived for $\mu = 4$

both 3-cycles are unstable because $F_\mu^{(3)}(c_1) = F'_\mu(c_1) F'_\mu(c_2) F'_\mu(c_3) = \pm 2^3 = \pm 8$



Sharkovsky's Theorem

Sharkovsky's ordering:

$$3 \triangleright 5 \triangleright 7 \triangleright \dots \triangleright 2 \times 3 \triangleright 2 \times 5 \triangleright \dots \triangleright 2^2 \times 3 \triangleright 2^2 \times 5 \triangleright \dots \triangleright 2^3 \times 3 \triangleright 2^3 \times 5 \triangleright \dots \\ \dots \triangleright 2^n \times 3 \triangleright 2^n \times 5 \triangleright \dots \triangleright 2^n \triangleright 2^{n-1} \triangleright 2^3 \triangleright 2^2 \triangleright 2 \triangleright 1$$

Sharkovsky's Theorem (1964):

Let $\Omega = [a, b] \subset \mathbb{R}$ be an interval and $F : \Omega \rightarrow \Omega$ be continuous.

If F has a point of period n , then it has points of period k for all $k \in \mathbb{N}$ with $n \triangleright k$.

Sharkovsky's Theorem

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We have shown that the logistic map $x_{k+1} = \mu x_k (1 - x_k)$ has 3-cycles (points of period 3) for any $\mu \in [1 + \sqrt{8}, 4]$.

Sharkovsky's theorem implies that the logistic map has points of period k (i.e. k -cycles) for all $k \in \mathbb{N}$ for $\mu \in [1 + \sqrt{8}, 4]$.

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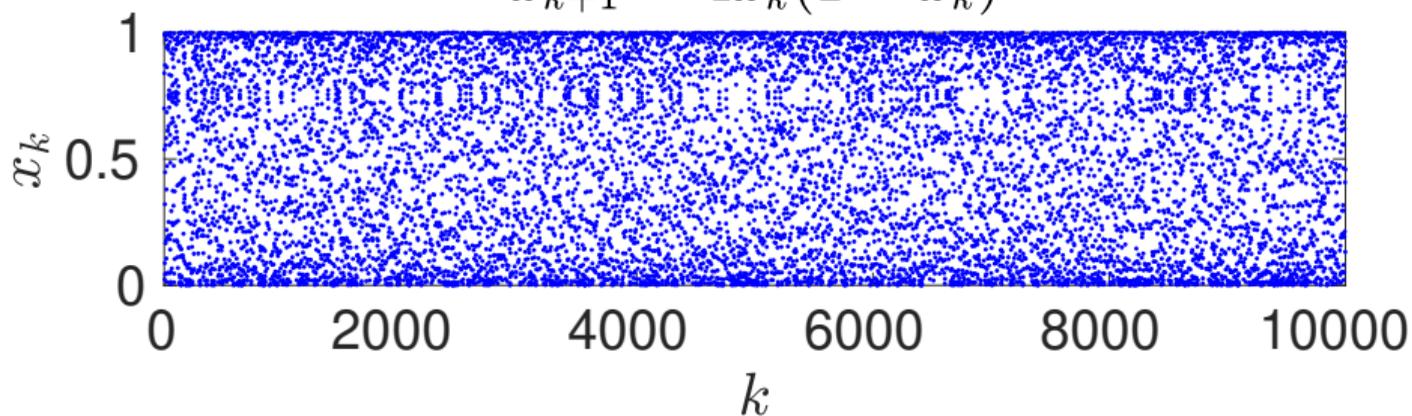
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Question 4 on Problem Sheet 2: closed formulas for k -cycles can be derived for $\mu = 4$, we can also show that k -cycles are unstable by calculating the corresponding derivatives

Invariant distribution (Question 7 on Problem Sheet 2)

Questions 3 and 4 on Problem Sheet 0: Starting with $x_0 = 0.7$, we obtain x_k as:

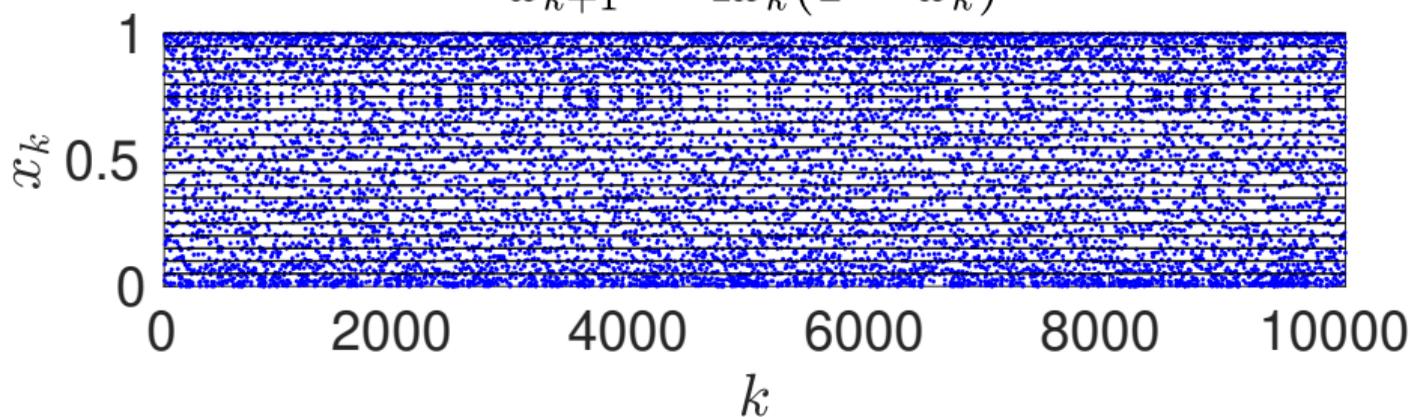
$$x_{k+1} = 4x_k(1 - x_k)$$



Invariant distribution (Question 7 on Problem Sheet 2)

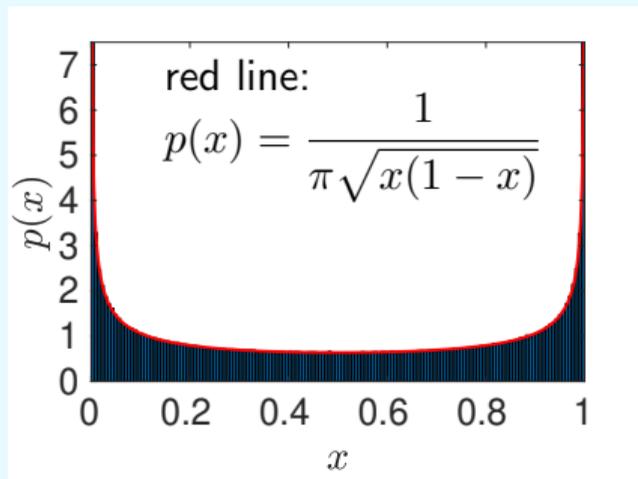
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Invariant distribution (Question 7 on Problem Sheet 2)

Histogram of values x_k , for $k = 0, 1, 2, \dots, 10^6$ (blue bars): $x_{k+1} = 4x_k(1 - x_k)$



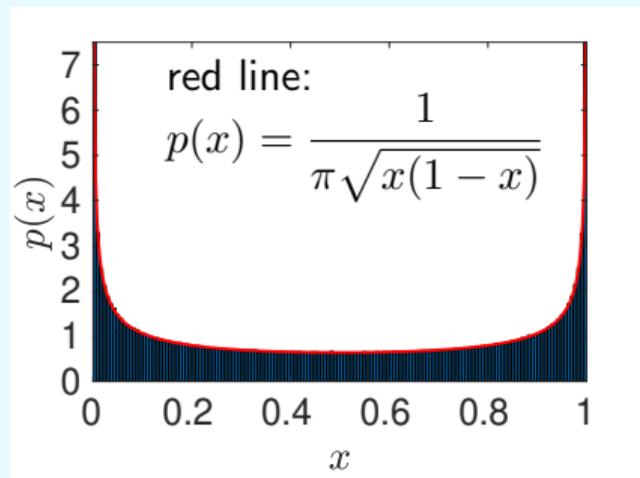
Question 4 on Problem Sheet 0:

Let X_k be a continuous random variable on interval $[0, 1]$ with the probability density function $p(x)$. Then the random variable $X_{k+1} = F(X_k) = 4X_k(1 - X_k)$ has the same probability density function $p(x)$.

[Prelims Probability and Calculus]

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[Prelims Probability and Calculus]

invariant distribution $p(x)$: if the random variable X is distributed according to $p(x)$, then the random variable $F(X)$ is also distributed according to $p(x)$

Question 7 on Problem Sheet 2: calculate invariant distributions for some other chaotic maps and compare them with the histograms of orbits

theoretical justification is given by ergodic theory (Birkhoff ergodic theorem)

B5.6: Nonlinear Dynamics, Bifurcations and Chaos (Lecture 8)

- summary of Lecture 7: we discussed
Logistic map. Periodic points of maps. Stability of N -cycles. Period-doubling bifurcation. Sharkovsky's theorem.
(Questions 3 and 4 on Problem Sheet 2)
- today: we will conclude our discussion of Problem Sheet 2

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- course synopsis of **Lectures 1-8** (taken by both Part B and MSc students):
Discrete-time (maps) and continuous-time (differential equations) dynamical systems. Notion of flows, stability of fixed points, Lyapunov function, invariant manifolds, stable manifold theorem, notion of hyperbolicity, center manifold. Chemical reaction networks. Stable, unstable and center subspaces. Poincaré-Bendixson theorem. Periodic solutions, stable and unstable limit cycles. [Introduction to bifurcation theory, covering saddle-node, transcritical, supercritical pitchfork and subcritical pitchfork bifurcations. Extended center manifold. Logistic map. Periodic points of maps. Stability of \$N\$ -cycles. Period-doubling bifurcation. Sharkovsky's theorem. Invariant distribution.](#)

Problem Sheet 2: bifurcations of continuous-time dynamical systems

Continuous-time dynamical system: Let $\mathbf{f} : \Omega \times \Theta \rightarrow \mathbb{R}^n$, where $\Omega \subset \mathbb{R}^n$ and $\Theta \subset \mathbb{R}^m$.

Let $\mathbf{x}_0 \in \Omega$, $\boldsymbol{\mu} \in \Theta$ and $\mathbf{x}(t) \in \Omega$ be a solution of the ODE

$$\frac{d\mathbf{x}}{dt} = \mathbf{f}(\mathbf{x}; \boldsymbol{\mu}) \quad \text{with the initial condition} \quad \mathbf{x}(0) = \mathbf{x}_0 \in \Omega$$

Problem Sheet 2: bifurcations of continuous-time dynamical systems

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Questions 1, 2, 5 and 6 on Problem Sheet 2 cover bifurcations of fixed points, which can occur for $n \geq 1$ and $m \geq 1$:

- saddle-node bifurcation
- transcritical bifurcation
- supercritical pitchfork bifurcation
- subcritical pitchfork bifurcation

We have explained them in our lectures on examples with $n = 1, 2$ and $m = 1$.

Next, we will discuss some additional examples to help you solve Problem Sheet 2, including examples with $m \geq 2$ and $n = 3$.

Example: $n = 1, m = 2$

$$\frac{dx_1}{dt} = \mu_2 + \mu_1 x_1 - x_1^3$$

$$f(x_1; \boldsymbol{\mu}) = \mu_2 + \mu_1 x_1 - x_1^3$$

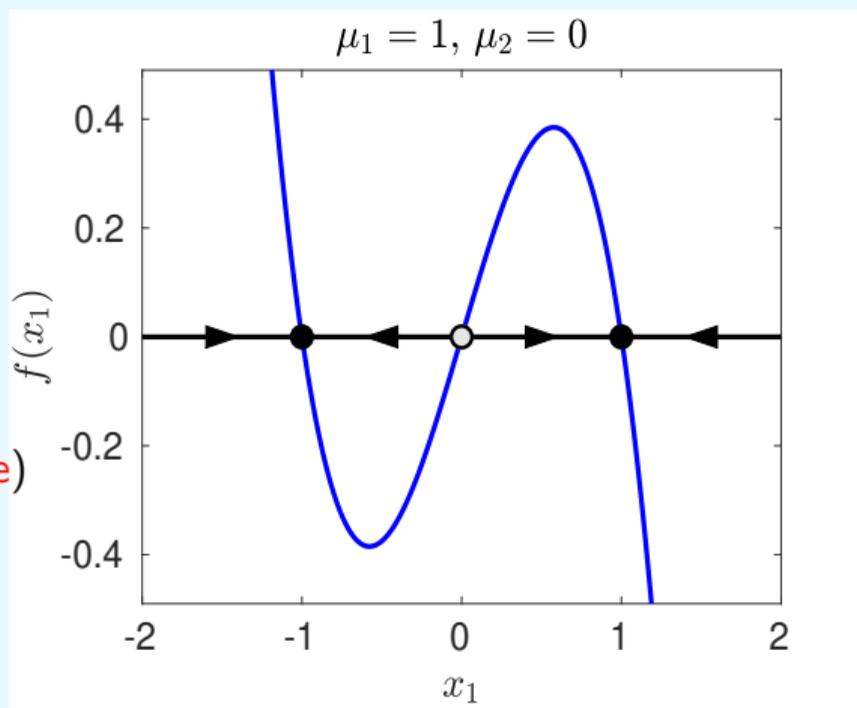
$\mu_2 = 0$: supercritical pitchfork bifurcation

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$$\mu_1 > 0, \mu_2 = 0$$

three fixed points at $x_1 = \pm\sqrt{\mu_1}$ (stable)
and $x_1 = 0$ (unstable)



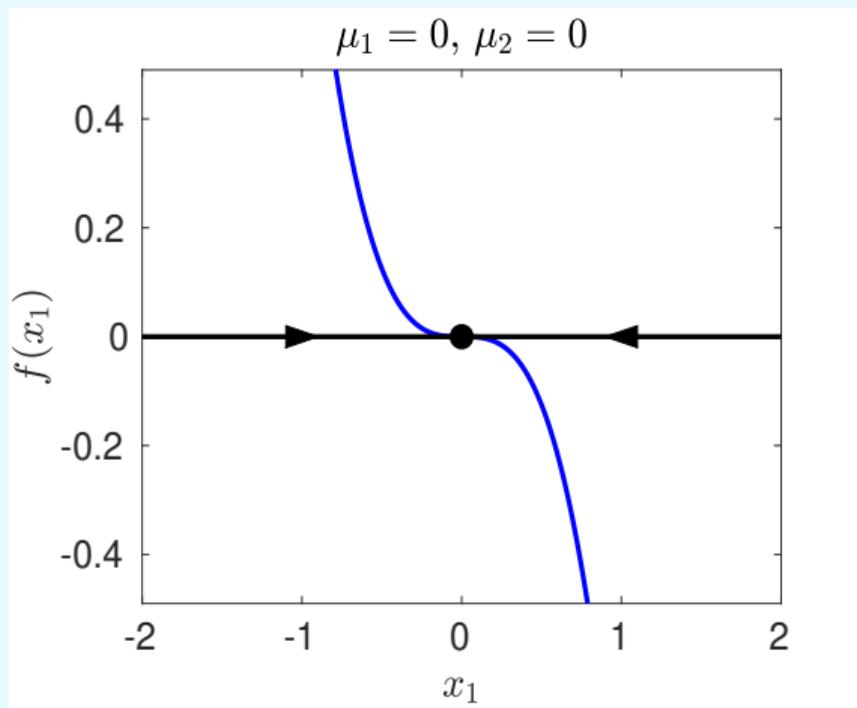
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as μ_1 approaches zero from above,
two fixed points $\sqrt{\mu_1}$ and $-\sqrt{\mu_1}$
move toward the third one

$\mu_1 = 0$: the fixed points coalesce into
a stable fixed point at $x_1 = 0$

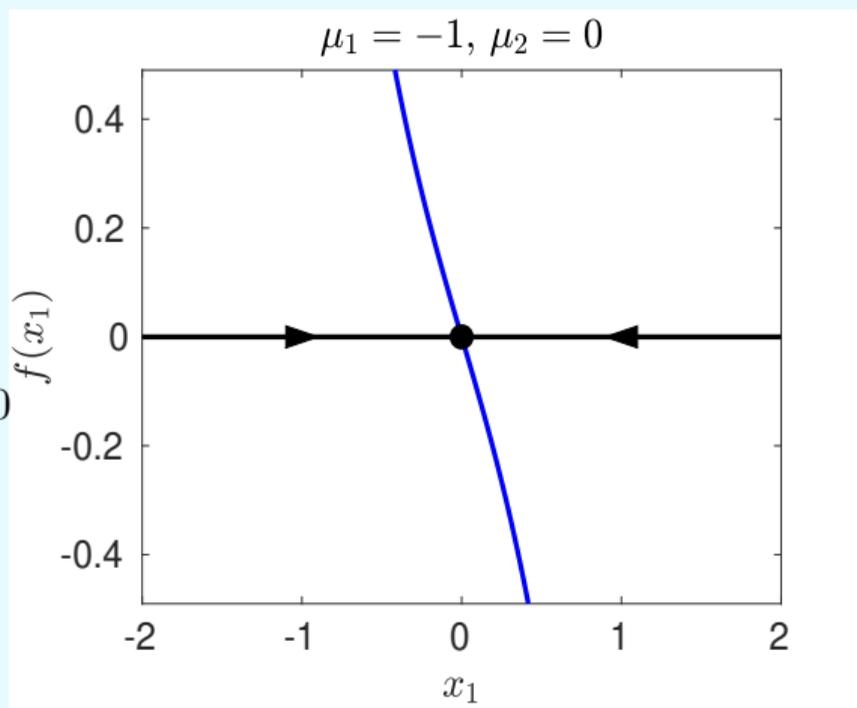


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$\mu_1 < 0$: one stable fixed point at $x_1 = 0$

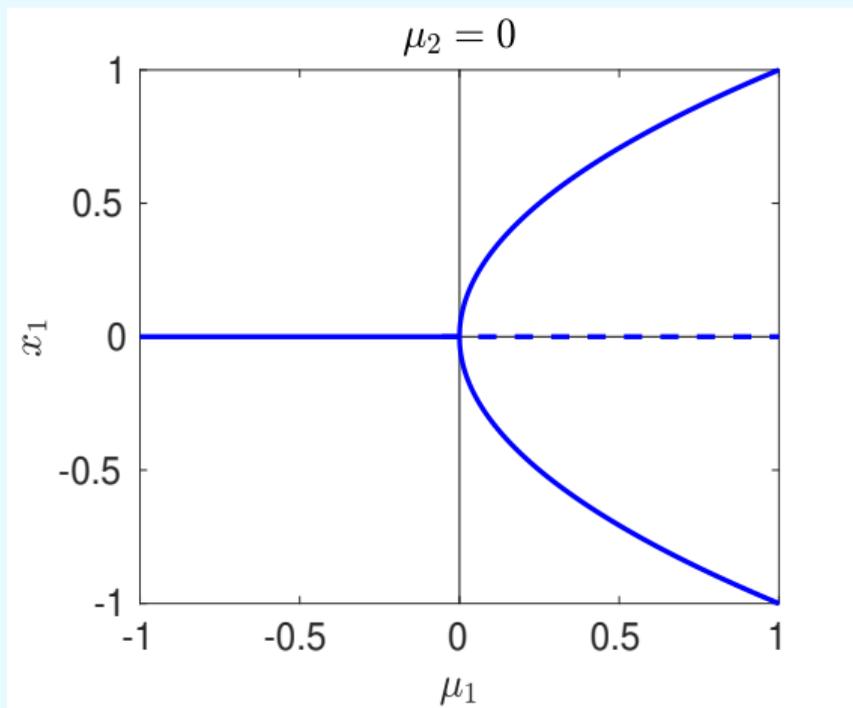


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bifurcation diagram



$\mu_2 = 0$: supercritical pitchfork bifurcation ($n = 2$)

$$\begin{aligned}\frac{dx_1}{dt} &= \mu_2 + \mu_1 x_1 - x_1^3 \\ \frac{dx_2}{dt} &= -x_2\end{aligned}$$

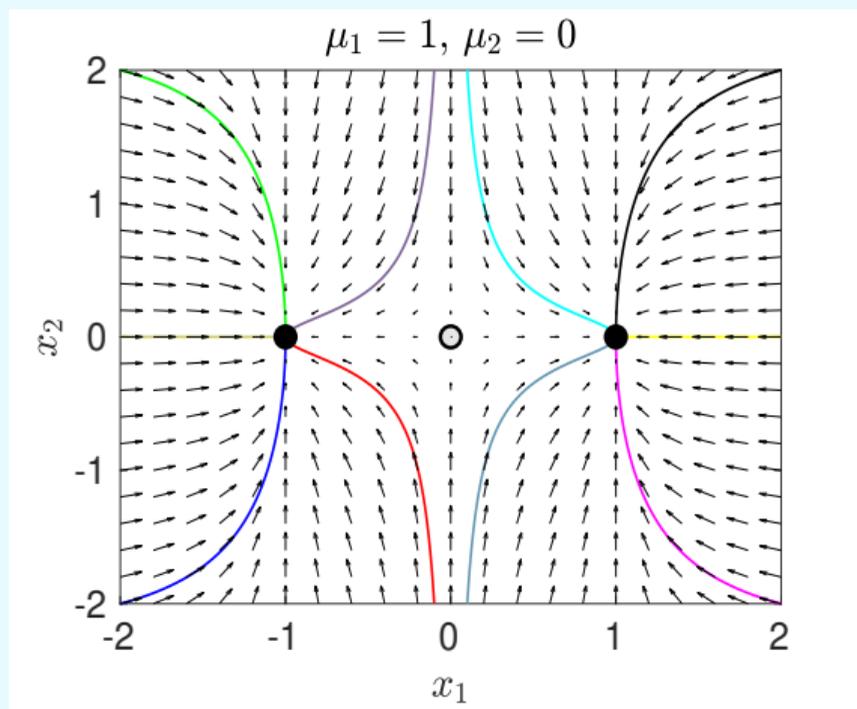
$$\mu_1 > 0, \mu_2 = 0$$

three fixed points at

$$\mathbf{x} = [-\sqrt{\mu_1}, 0] \text{ (stable node)}$$

$$\mathbf{x} = [0, 0] \text{ (saddle)}$$

$$\mathbf{x} = [\sqrt{\mu_1}, 0] \text{ (stable node)}$$

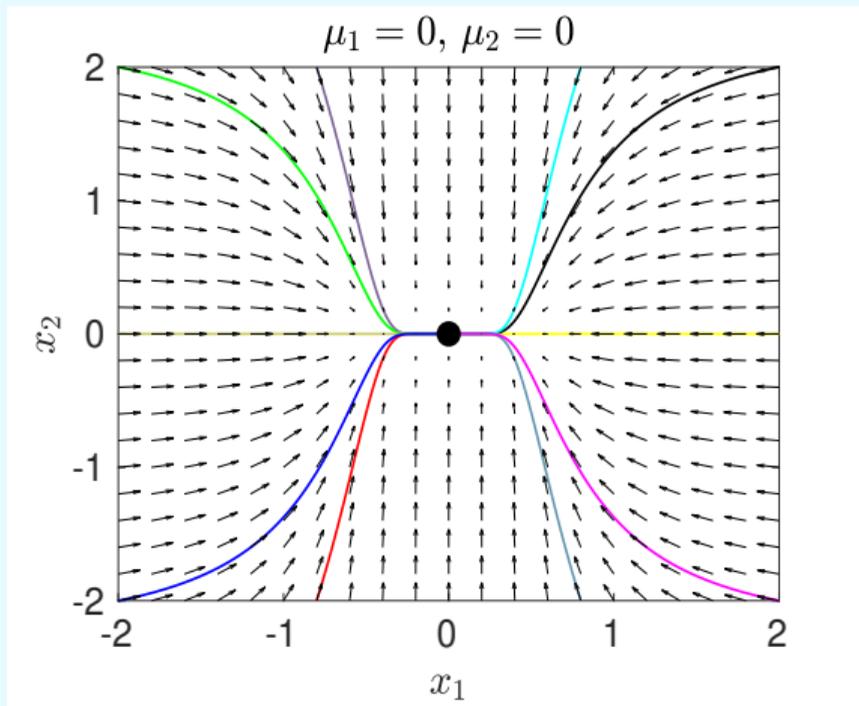


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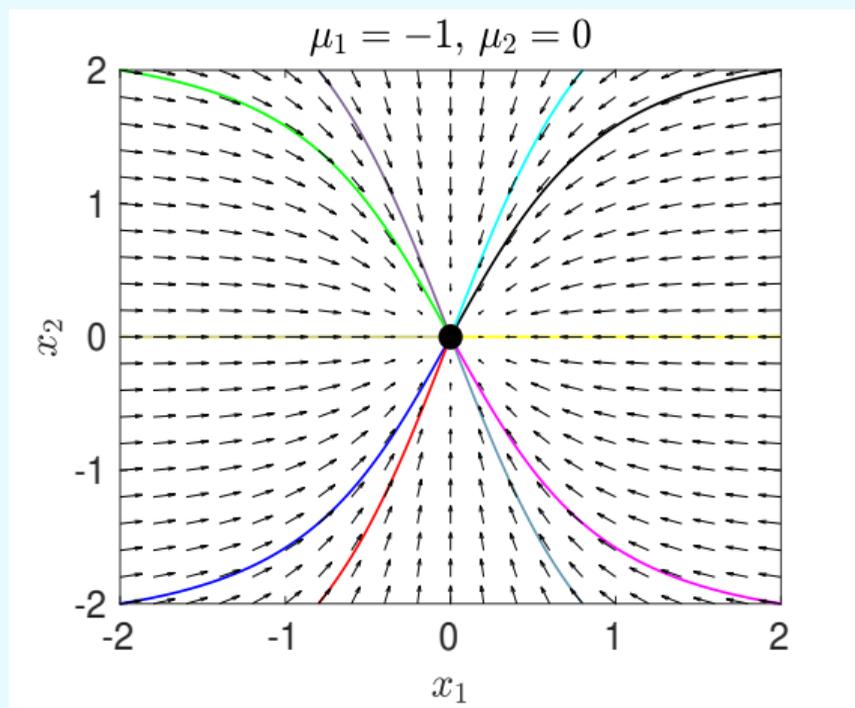
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$$\frac{dx_1}{dt} = \mu_2 + \mu_1 x_1 - x_1^3$$
$$\frac{dx_2}{dt} = -x_2$$

$\mu_1 < 0$:
one stable fixed point at $\mathbf{x} = [0, 0]$

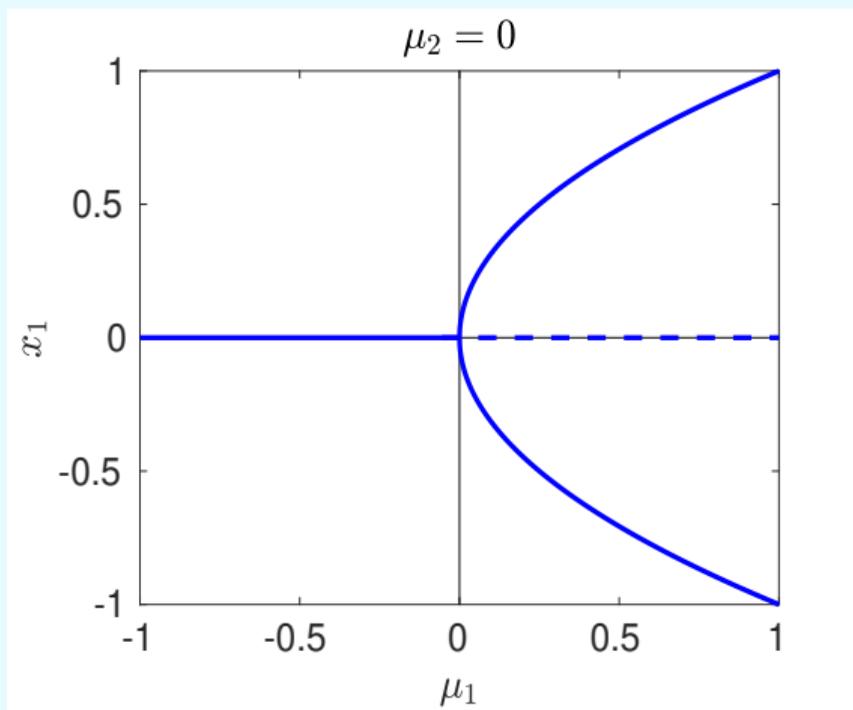


$\mu_2 = 0$: supercritical pitchfork bifurcation ($n = 2$)

$$\frac{dx_1}{dt} = \mu_2 + \mu_1 x_1 - x_1^3$$

$$\frac{dx_2}{dt} = -x_2$$

bifurcation diagram



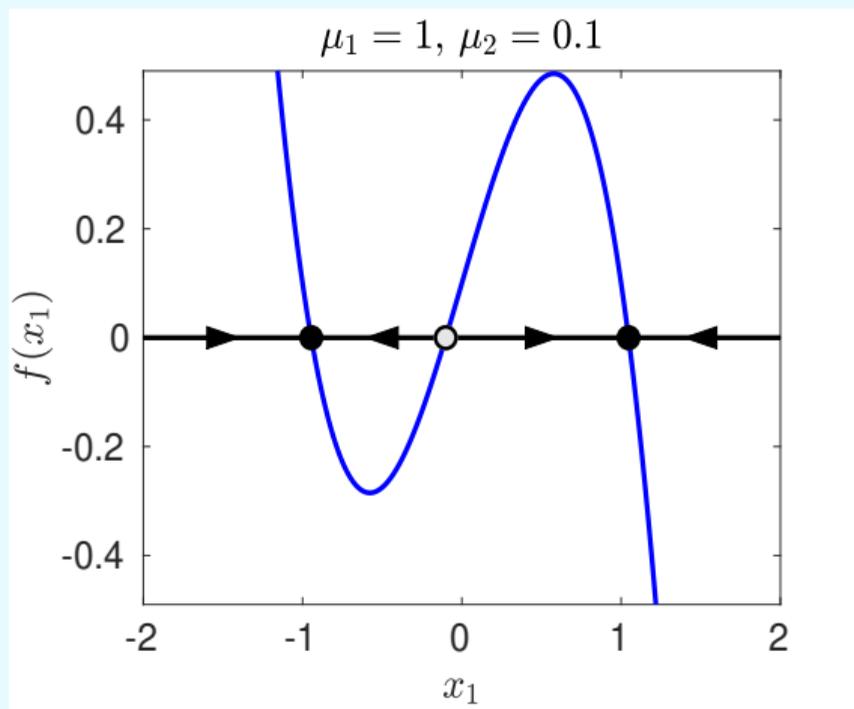
Cusp catastrophe ($\mu_2 = 0.1$)

$$\frac{dx_1}{dt} = \mu_2 + \mu_1 x_1 - x_1^3$$

$$f(x_1; \mu) = \mu_2 + \mu_1 x_1 - x_1^3$$

$$\mu_1 = 1, \mu_2 = 0.1$$

three fixed points given as
solutions of $\mu_2 + \mu_1 x_1 - x_1^3 = 0$



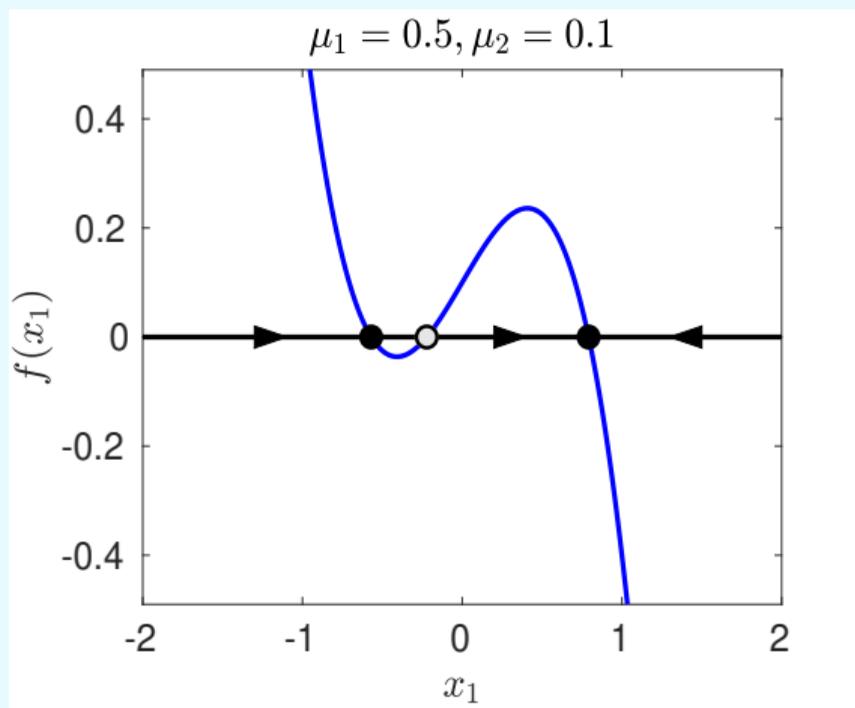
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$$f(x_1; \mu) = \mu_2 + \mu_1 x_1 - x_1^3$$

$$\mu_1 = 0.5, \mu_2 = 0.1$$

three fixed points given as
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Cusp catastrophe ($\mu_2 = 0.1$)

$$\frac{dx_1}{dt} = \mu_2 + \mu_1 x_1 - x_1^3$$

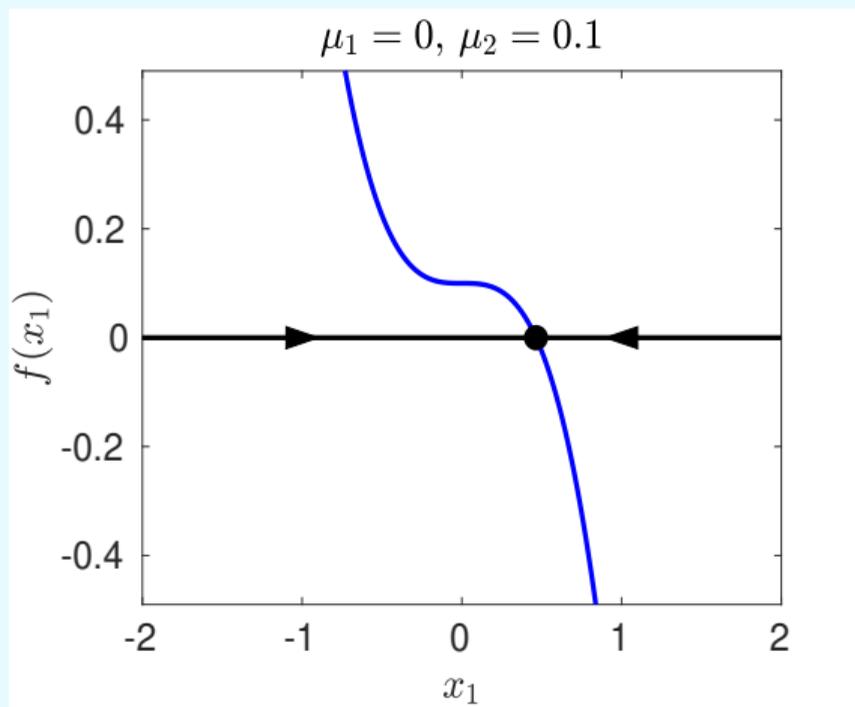
$$f(x_1; \boldsymbol{\mu}) = \mu_2 + \mu_1 x_1 - x_1^3$$

as μ_1 approaches the bifurcation value

$$\mu_c = \left(\frac{27\mu_2^2}{4} \right)^{1/3}$$

from above, two (smaller) fixed points
move toward each other
(saddle-node bifurcation)

$\mu_1 < \mu_c$: one stable fixed point

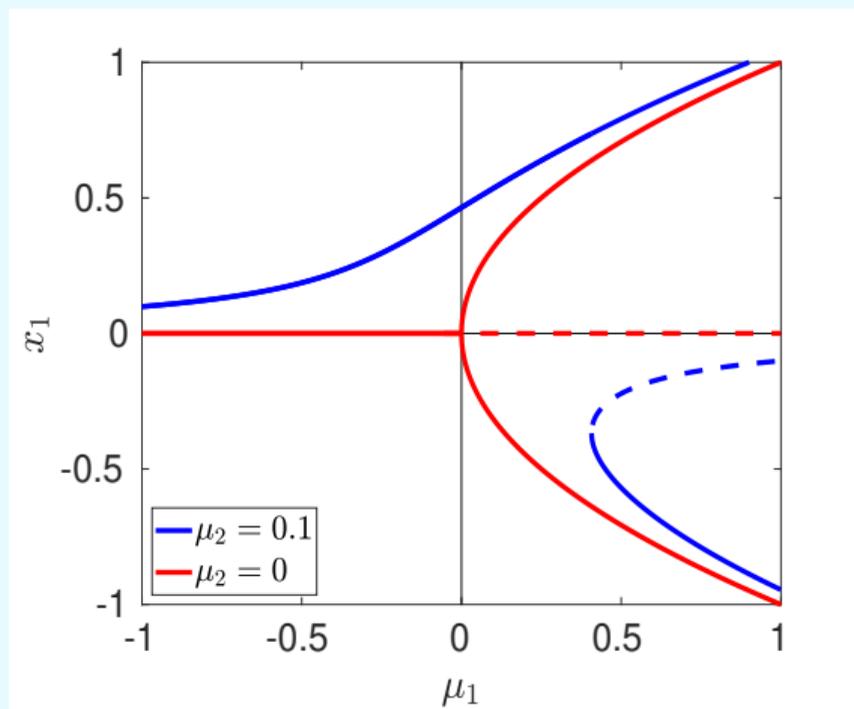


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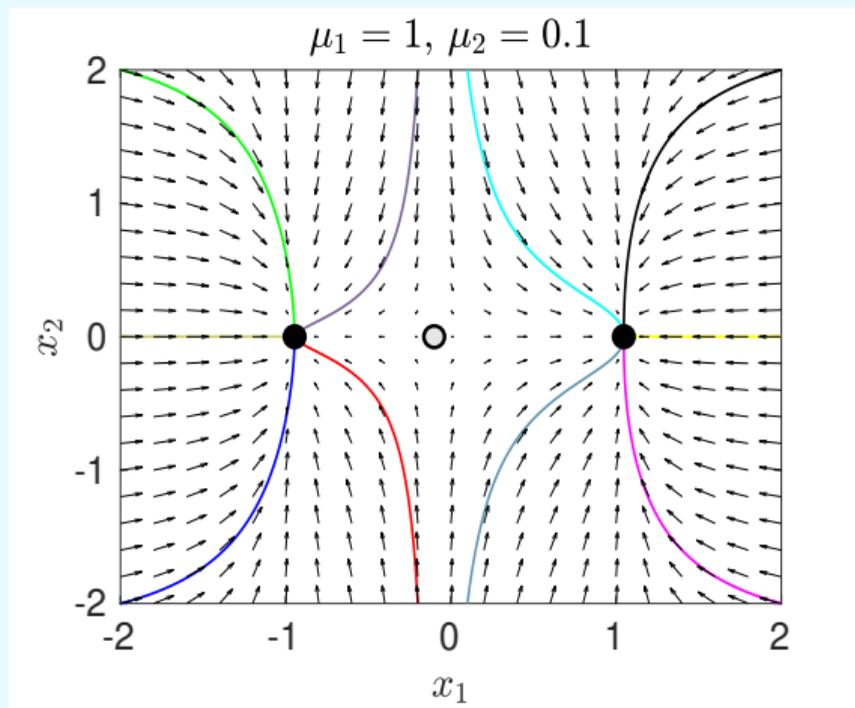
bifurcation diagram



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$$\frac{dx_1}{dt} = \mu_2 + \mu_1 x_1 - x_1^3$$
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$\mu_1 > \mu_c$: three fixed points given as solutions of $\mu_2 + \mu_1 x_1 - x_1^3 = 0$

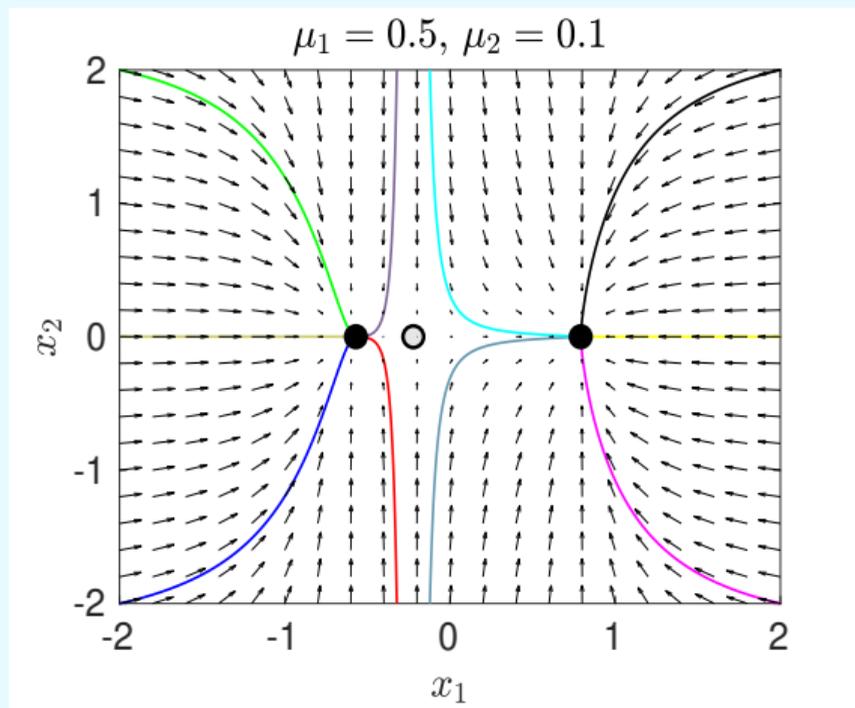


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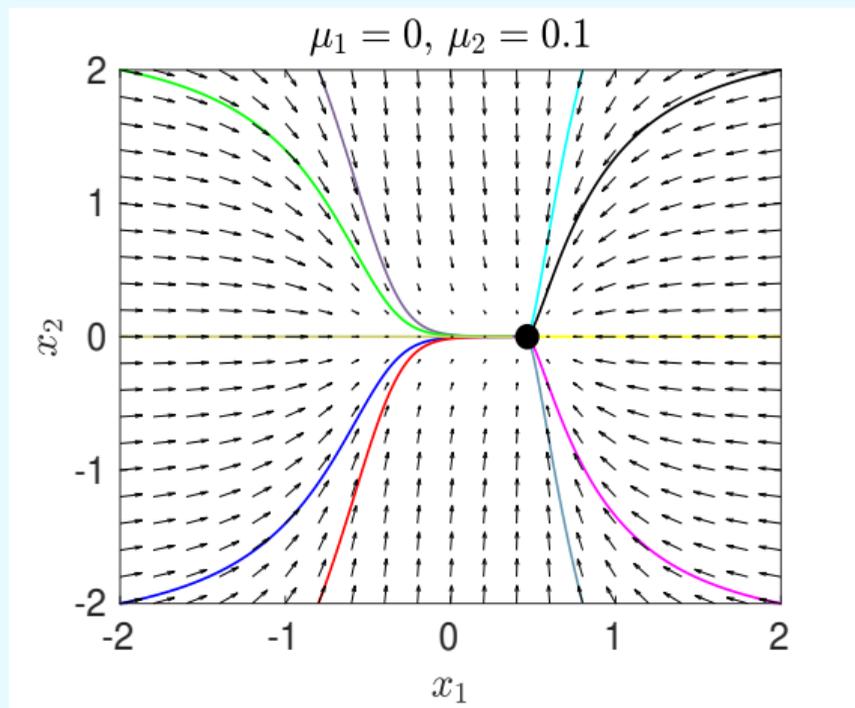
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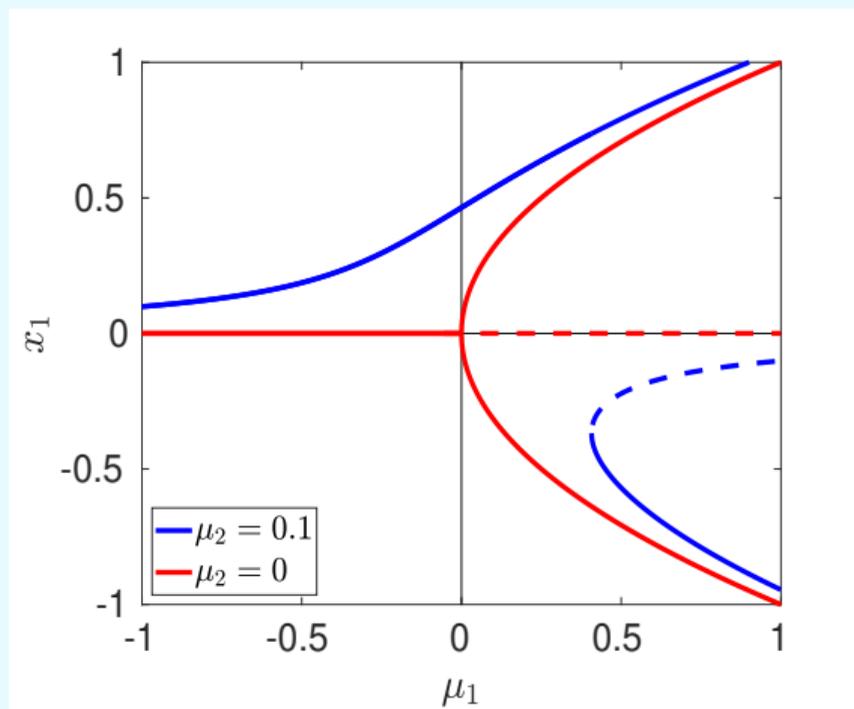
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Cusp catastrophe

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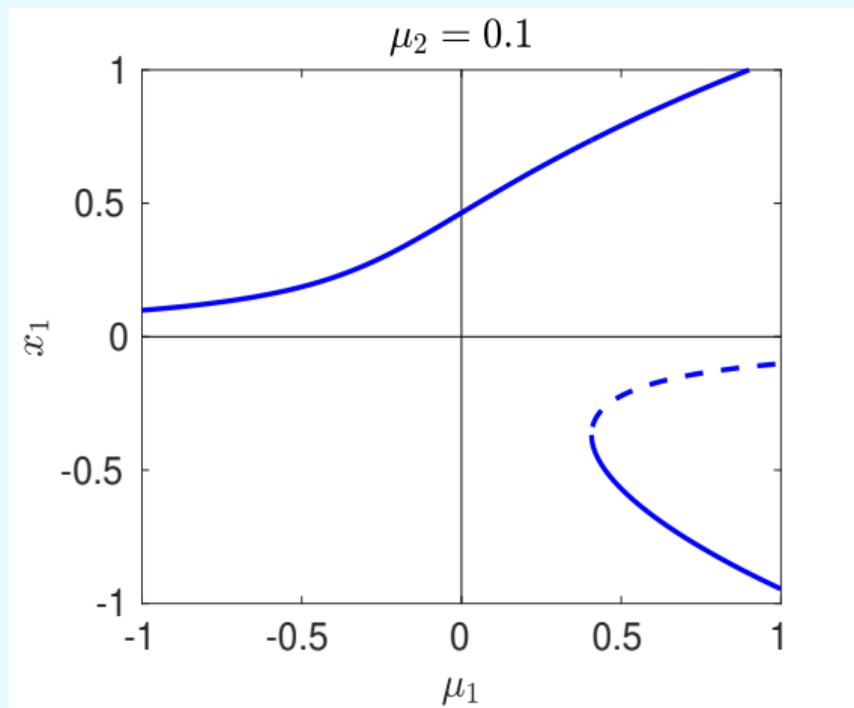


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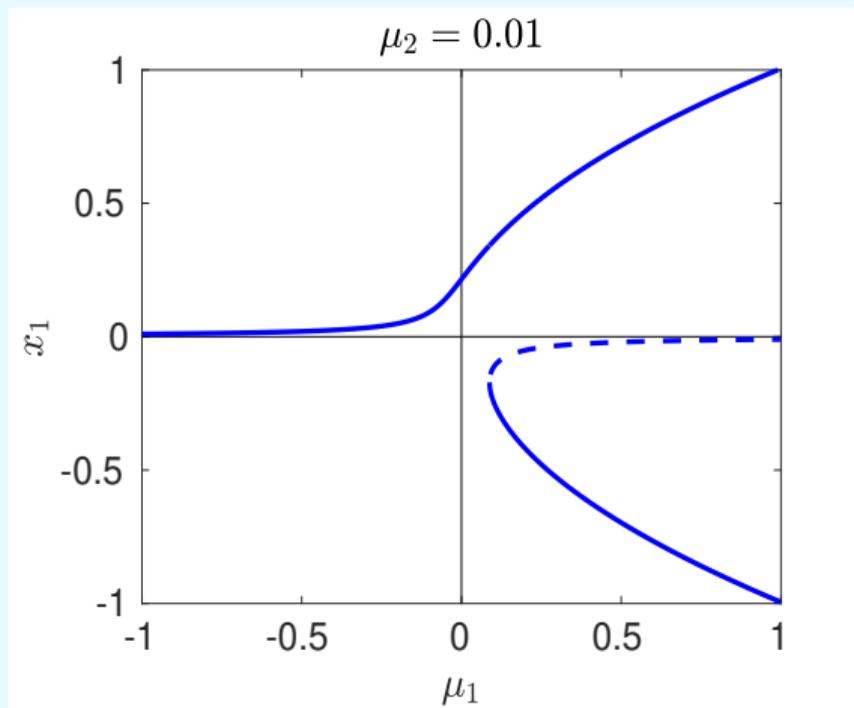


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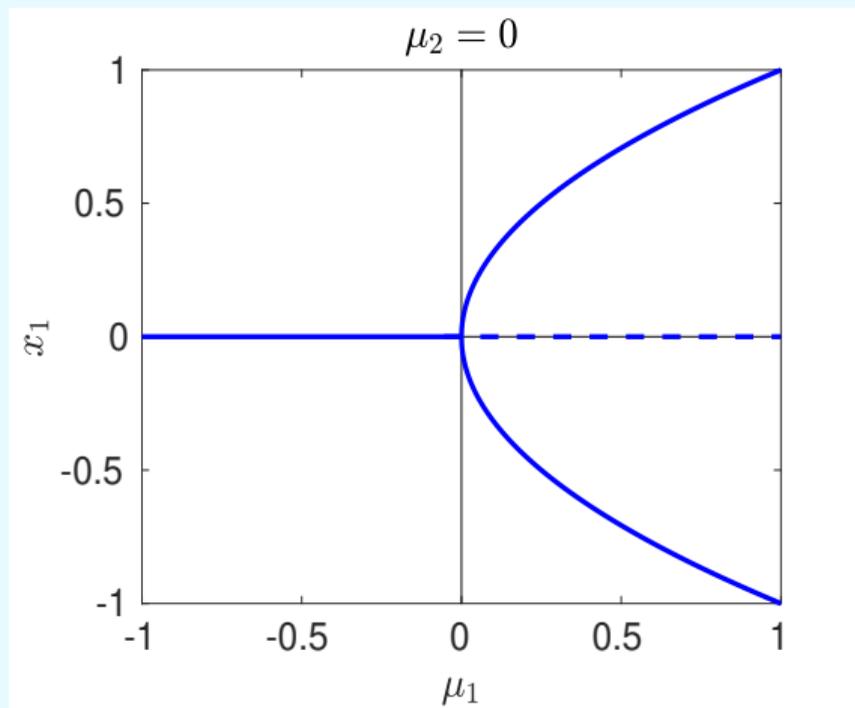


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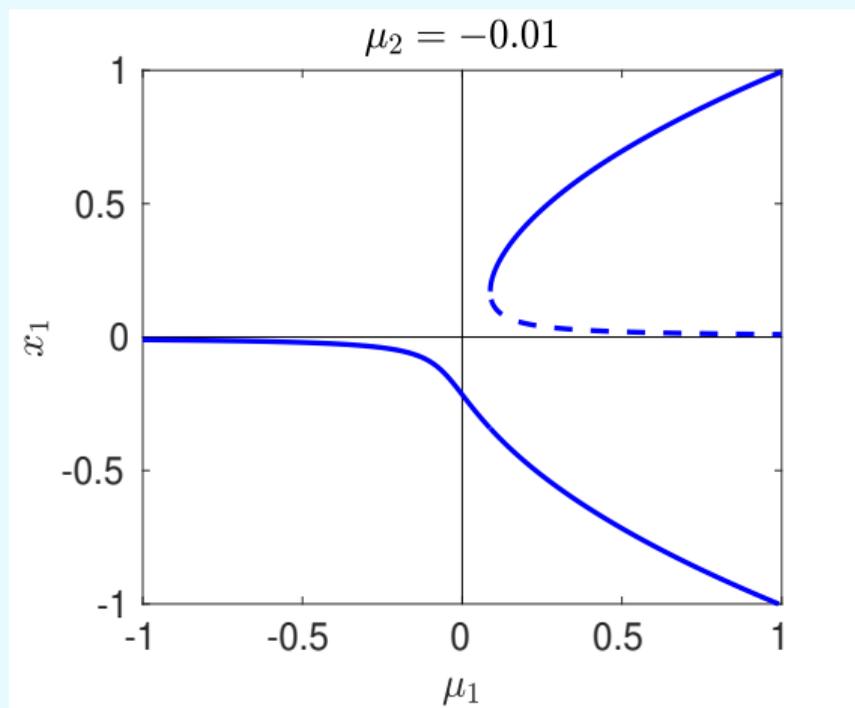


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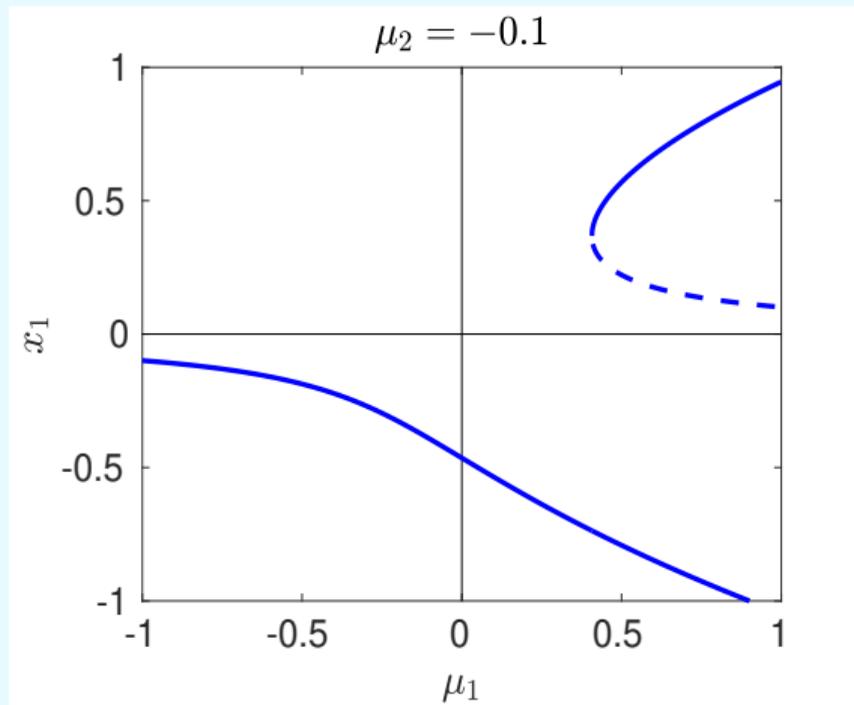


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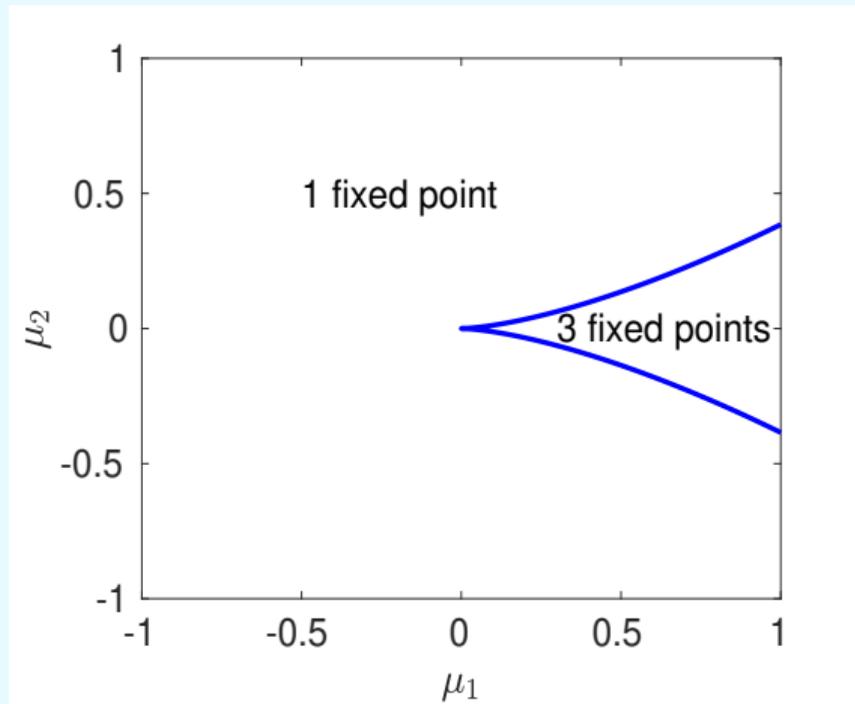
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Example: $n = 2, m = 2$

$$\frac{dx_1}{dt} = \mu_2 + \mu_1 x_1 - x_1^2$$

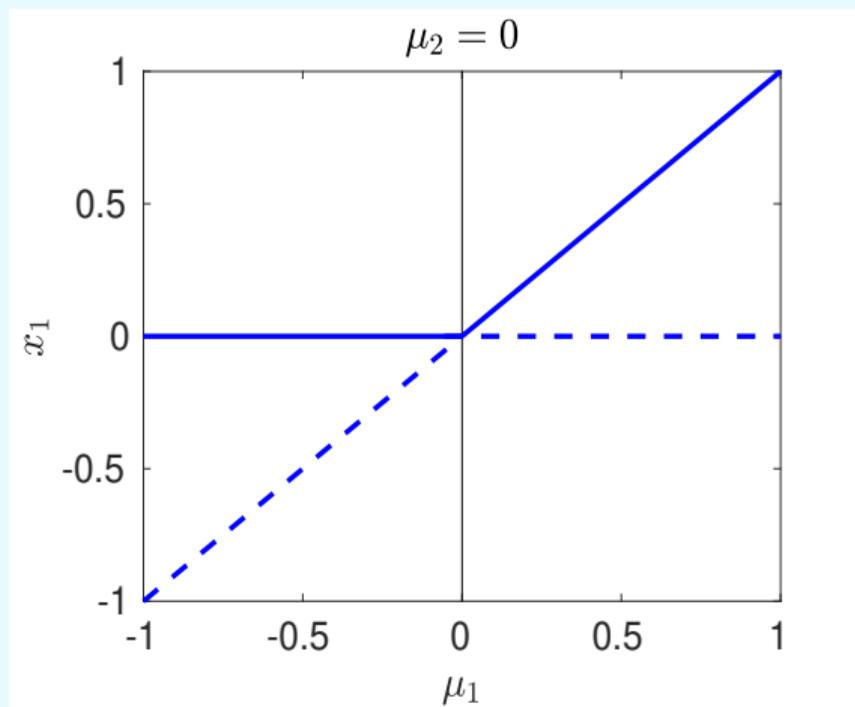
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Example: $n = 2, m = 2$

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$\mu_2 = 0$: transcritical bifurcation



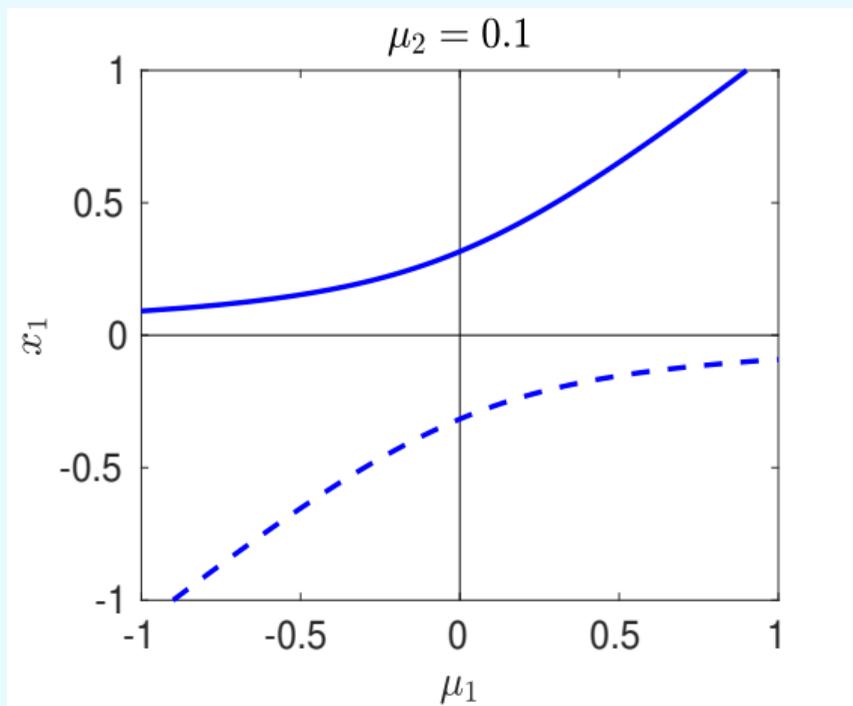
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The saddle-node bifurcation is robust under small changes of parameters, but transcritical and pitchfork bifurcations change under small perturbations.



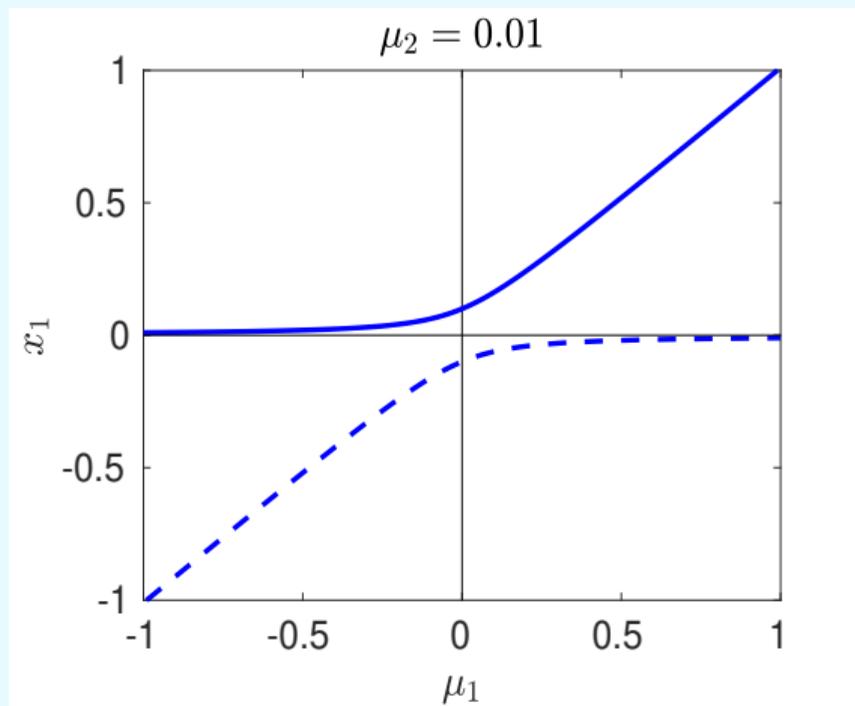
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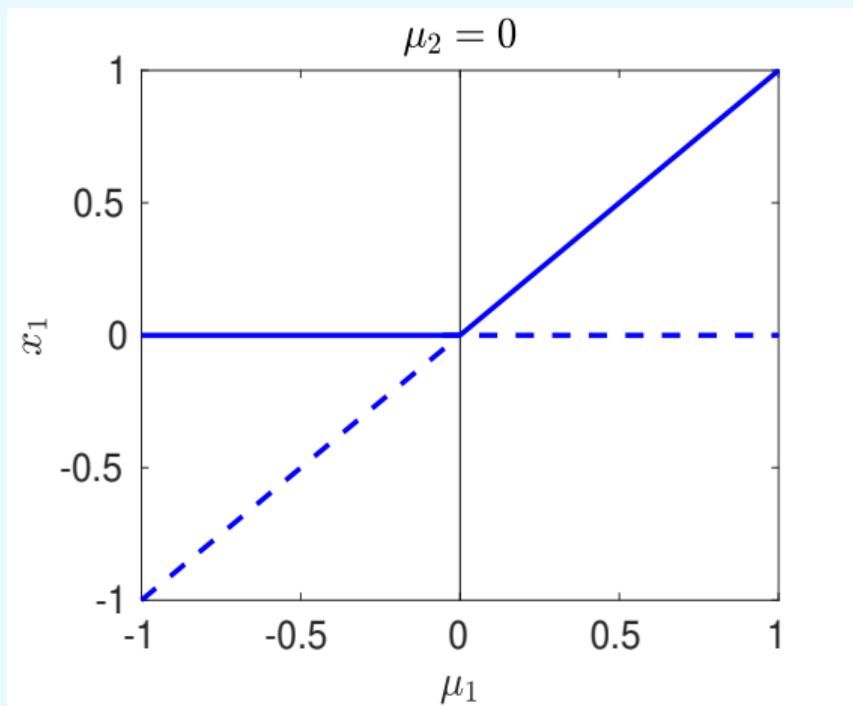
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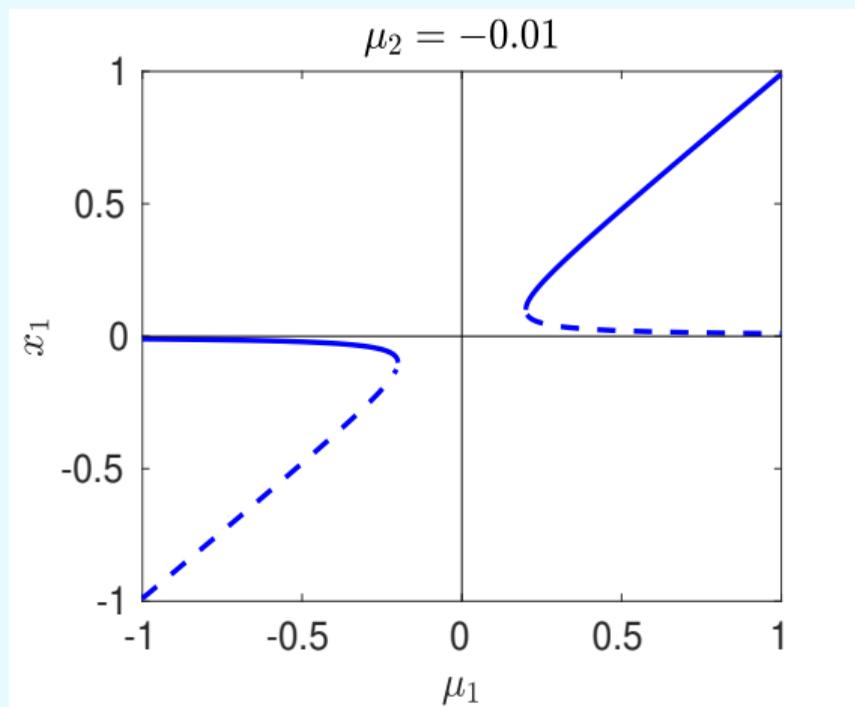
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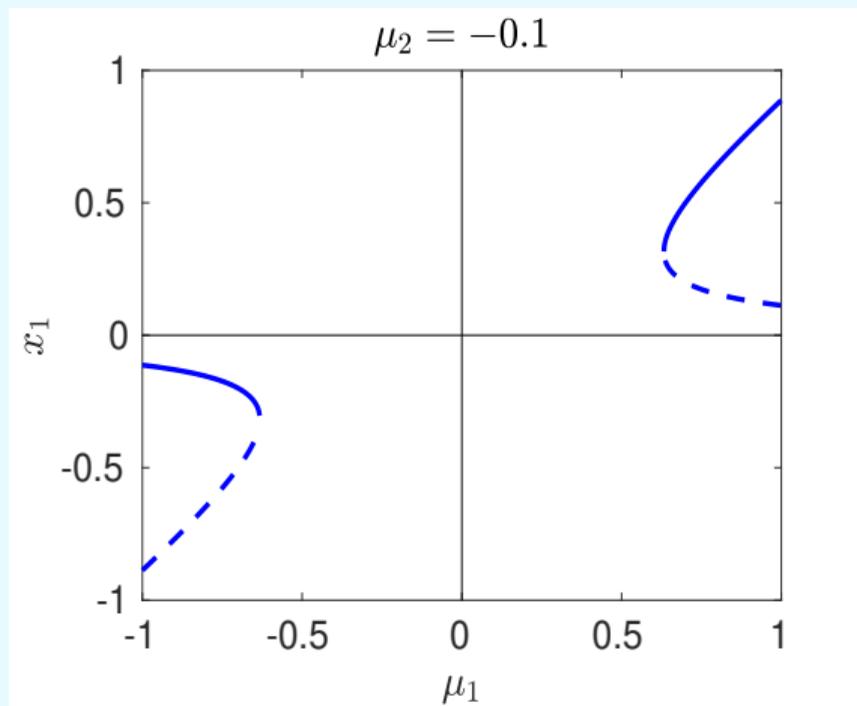
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The saddle-node bifurcation is robust under small changes of parameters, but transcritical and pitchfork bifurcations change under small perturbations.



Example with $n = 3$, $m = 3$: Lorenz equations

$$\frac{dx_1}{dt} = \mu_2 (x_2 - x_1)$$

$$\frac{dx_2}{dt} = \mu_1 x_1 - x_2 - x_1 x_3$$

$$\frac{dx_3}{dt} = x_1 x_2 - \mu_3 x_3$$

Lorenz equations: summary of our 3D visualization of dynamics

$$\frac{dx_1}{dt} = \mu_2 (x_2 - x_1)$$

$$\frac{dx_2}{dt} = \mu_1 x_1 - x_2 - x_1 x_3$$

$$\frac{dx_3}{dt} = x_1 x_2 - \mu_3 x_3$$

We will study the Lorenz system again in the second part of our course, when we will discuss chaos in ODEs.

TODAY: we illustrated the convergence to fixed points, limit cycles, chaos, strange attractor and transient chaos

Lorenz equations: summary of our 3D visualization of dynamics

$$\frac{dx_1}{dt} = 10(x_2 - x_1)$$

$$\frac{dx_2}{dt} = \mu_1 x_1 - x_2 - x_1 x_3$$

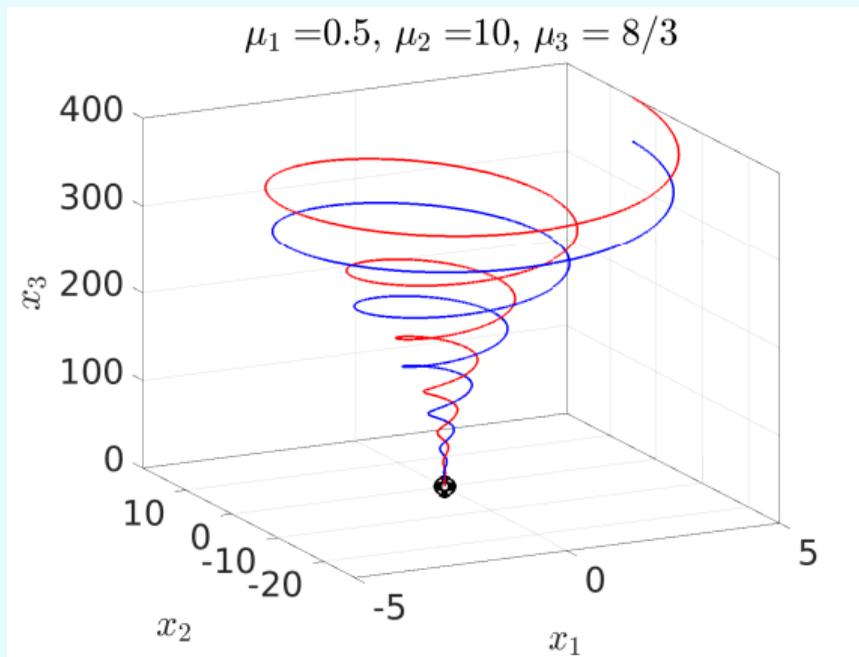
$$\frac{dx_3}{dt} = x_1 x_2 - \frac{8x_3}{3}$$

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TODAY: we illustrated the convergence to fixed points, limit cycles, chaos, strange attractor and transient chaos

by observing trajectories in the phase space for different values of the parameters we varied μ_1 , while we fixed the values of parameters μ_2 and μ_3 :

$$\mu_2 = 10 \text{ and } \mu_3 = \frac{8}{3} \quad (\text{Lorenz used } \mu_1 = 28 \text{ to get chaos})$$



Lorenz equations: summary of our 3D visualization of dynamics

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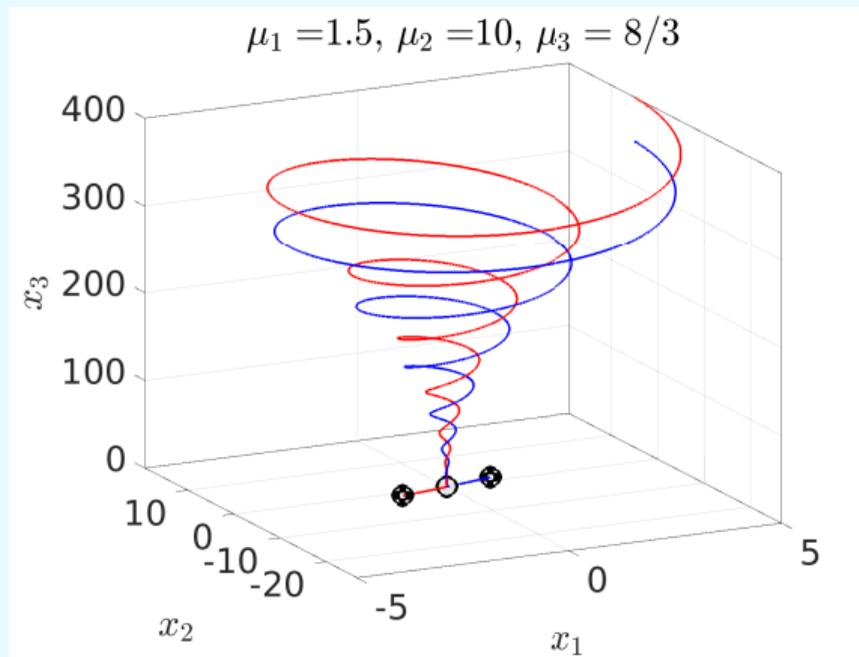
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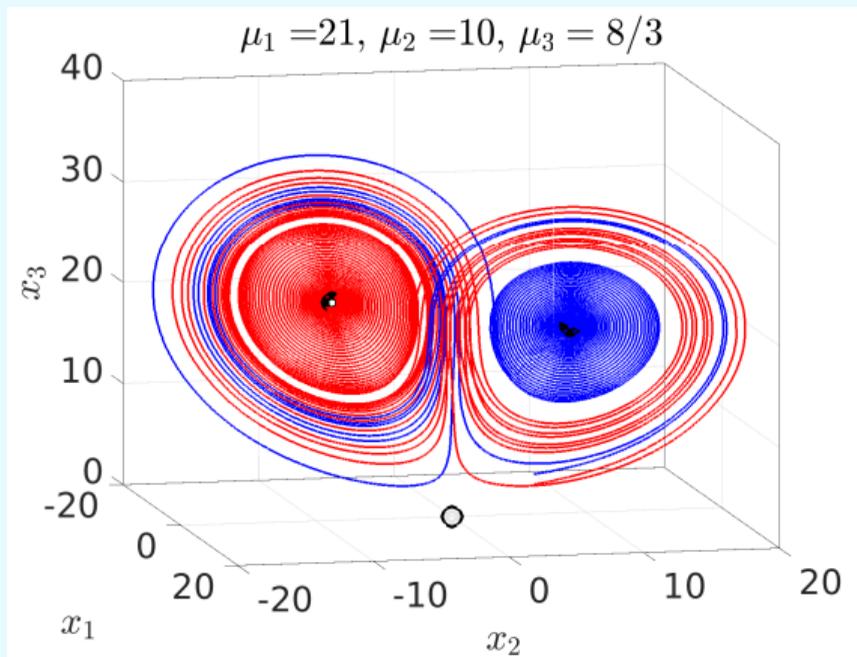
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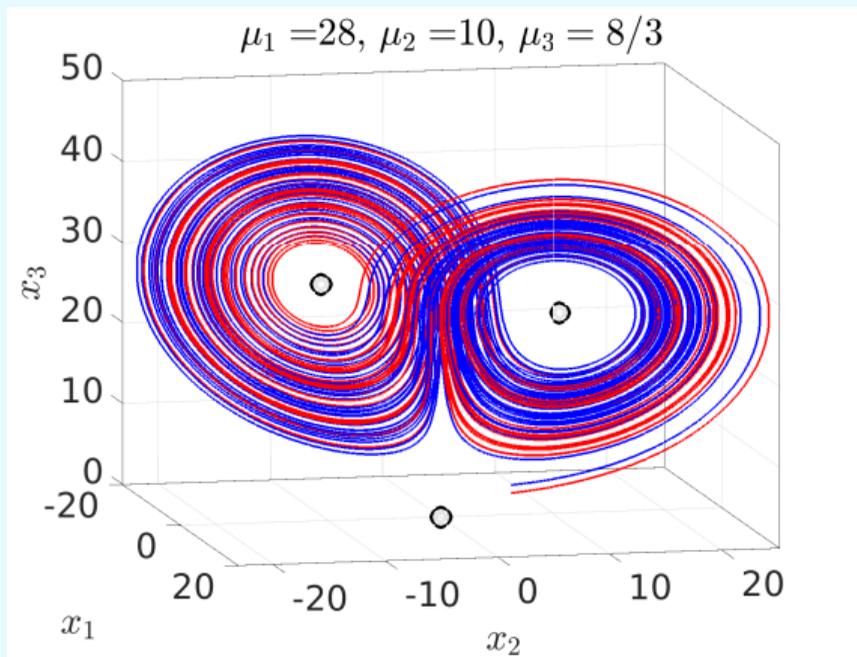
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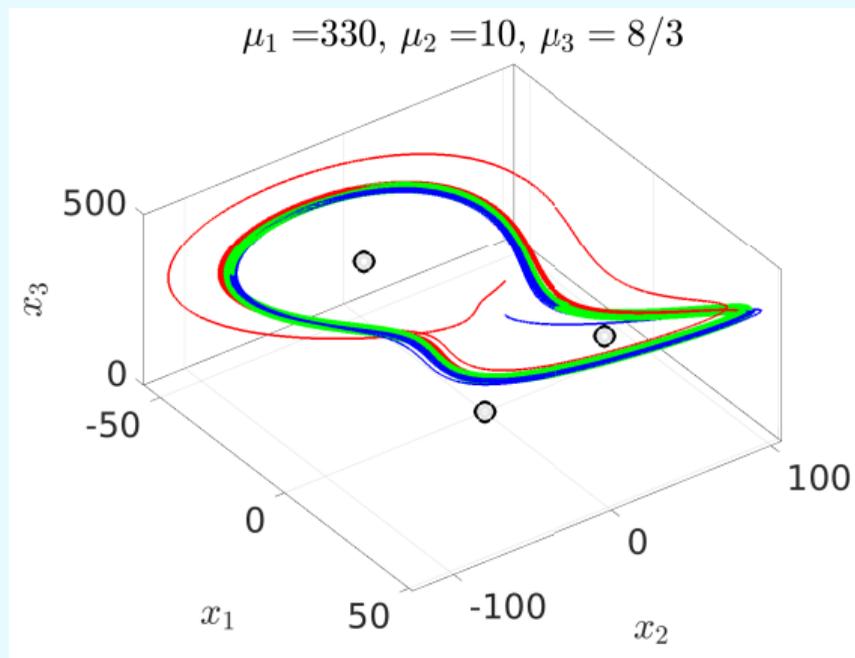
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Lorenz equations: Question 6 on Problem Sheet 2

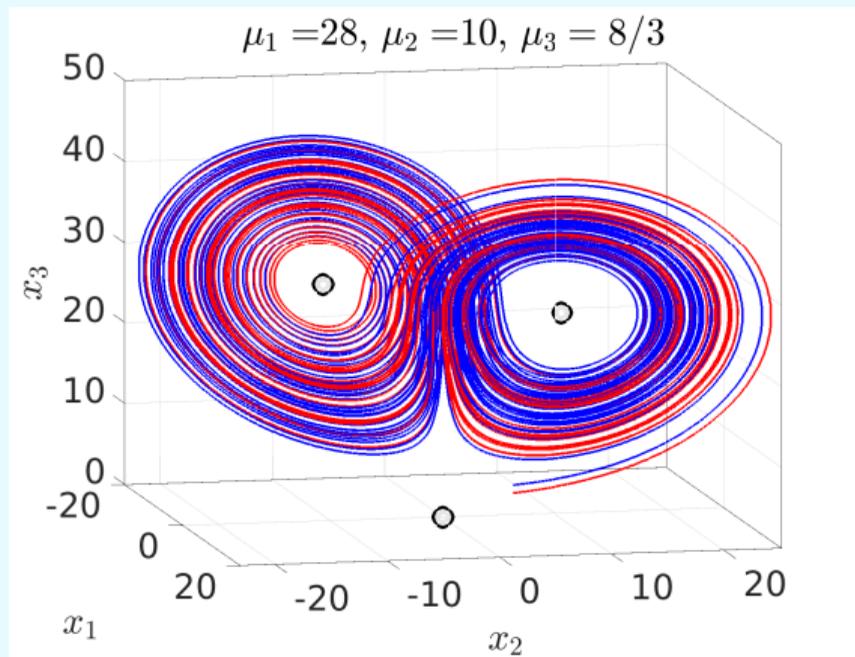
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Question 6 on Problem Sheet 2:

we use the Lorenz system to further practice some techniques studied in Lectures 1-8



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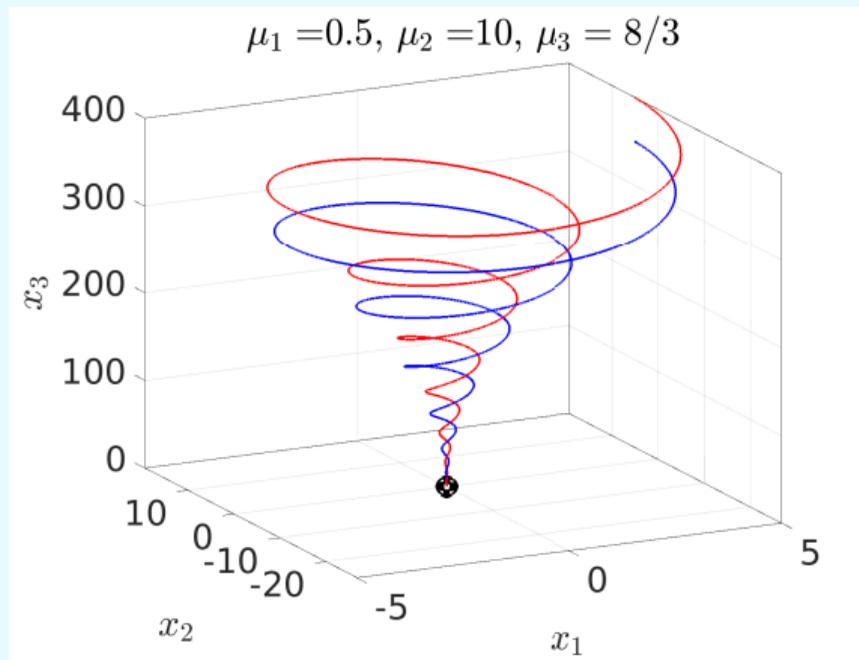
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Question 6 on Problem Sheet 2:

we use the Lorenz system to further practice some techniques studied in Lectures 1-8 including:

- finding the Lyapunov function to prove the global stability of the fixed point at the origin $\mathbf{0} = [0, 0, 0]$ for $\mu_1 < 1$



[Question 6(d)]

Lorenz equations: Question 6 on Problem Sheet 2

$$\frac{dx_1}{dt} = \mu_2 (x_2 - x_1)$$

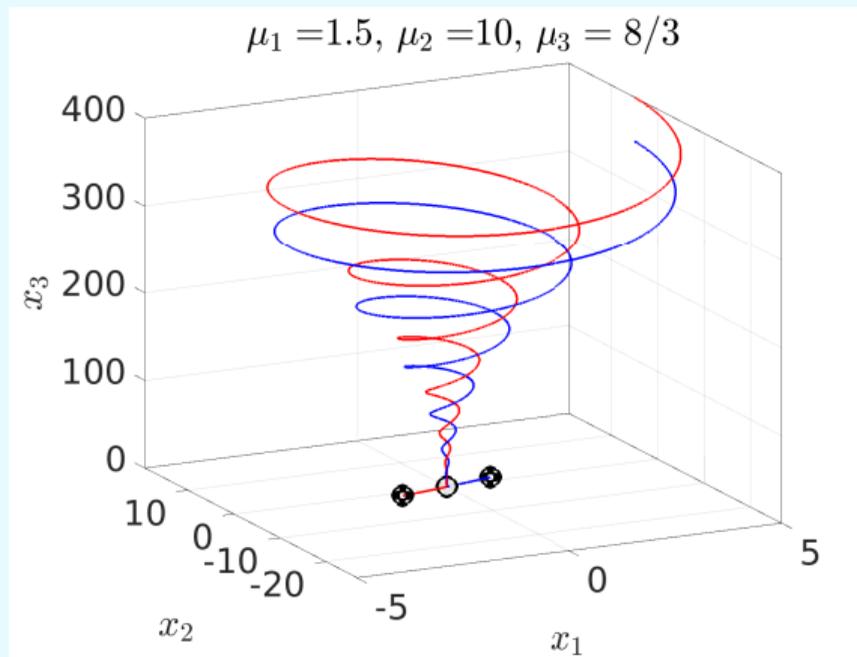
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Question 6 on Problem Sheet 2:

we use the Lorenz system to further practice some techniques studied in Lectures 1-8 including:

- finding the Lyapunov function to prove the global stability of the fixed point at the origin $\mathbf{0} = [0, 0, 0]$ for $\mu_1 < 1$ [Question 6(d)]
- using the extended center manifold theory to analyze the bifurcation at $\mu_1 = 1$, calculating the center manifold and the dynamics on it [Question 6(e)]



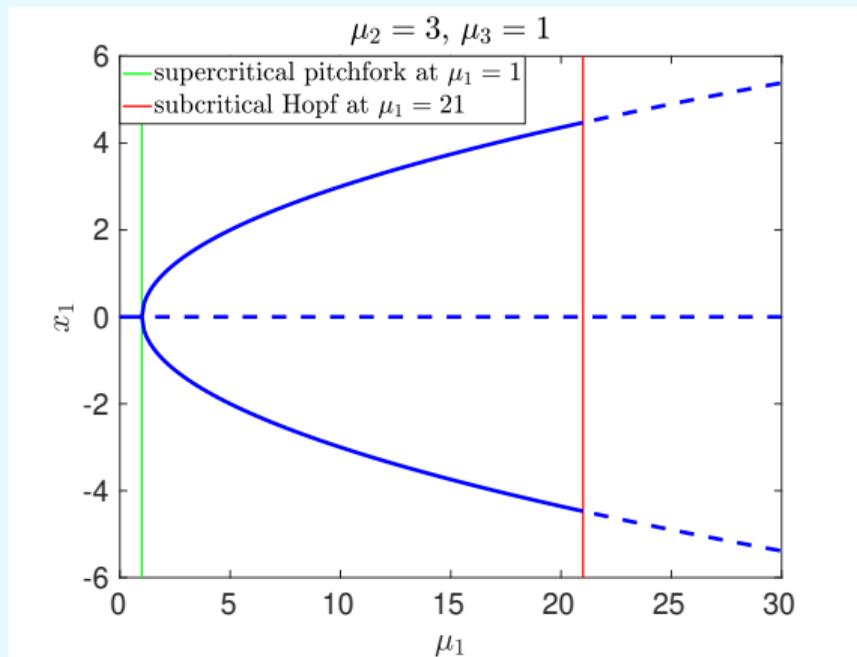
Lorenz equations: Question 6 on Problem Sheet 2

$$\frac{dx_1}{dt} = 3(x_2 - x_1)$$

$$\frac{dx_2}{dt} = \mu_1 x_1 - x_2 - x_1 x_3$$

$$\frac{dx_3}{dt} = x_1 x_2 - x_3$$

- fixed points $\mathbf{x}_{c1} = \mathbf{0} = [0, 0, 0]$
 $\mathbf{x}_{c2} = [\sqrt{\mu_1 - 1}, \sqrt{\mu_1 - 1}, \mu_1 - 1]$
 $\mathbf{x}_{c3} = [-\sqrt{\mu_1 - 1}, -\sqrt{\mu_1 - 1}, \mu_1 - 1]$
 \mathbf{x}_{c2} and \mathbf{x}_{c3} only exist for $\mu_1 > 1$



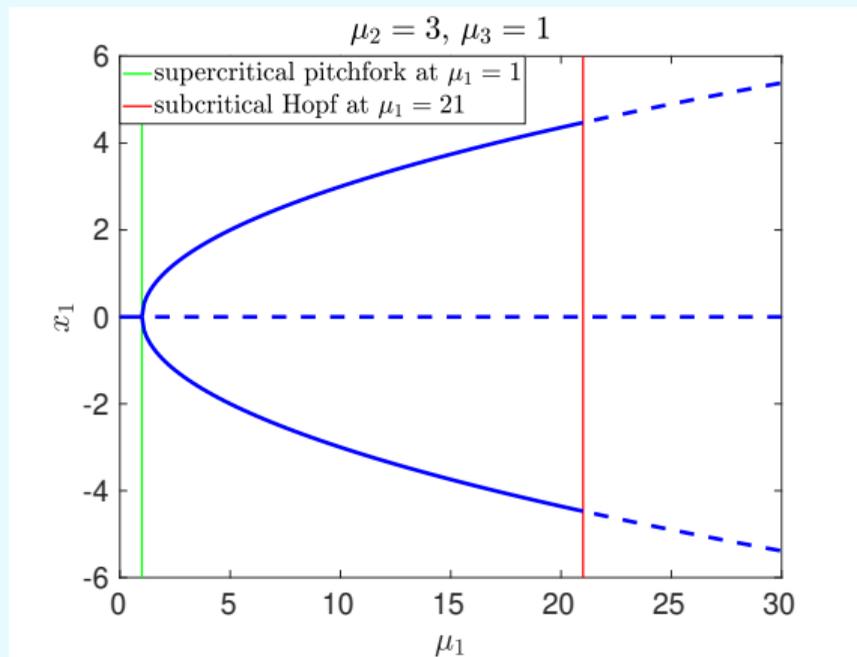
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- fixed points $\mathbf{x}_{c1} = \mathbf{0} = [0, 0, 0]$
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 \mathbf{x}_{c2} and \mathbf{x}_{c3} only exist for $\mu_1 > 1$
- supercritical pitchfork bifurcation
at $\mu_1 = 1$ ($\mathbf{x}_{c1} = \mathbf{0}$ is stable for $\mu_1 < 1$ and unstable for $\mu_1 > 1$)



[Question 6(e)]

Lorenz equations: Question 6 on Problem Sheet 2

$$\frac{dx_1}{dt} = 3(x_2 - x_1)$$

$$\frac{dx_2}{dt} = \mu_1 x_1 - x_2 - x_1 x_3$$

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- fixed points $\mathbf{x}_{c1} = \mathbf{0} = [0, 0, 0]$

$$\mathbf{x}_{c2} = [\sqrt{\mu_1 - 1}, \sqrt{\mu_1 - 1}, \mu_1 - 1]$$

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\mathbf{x}_{c2} and \mathbf{x}_{c3} only exist for $\mu_1 > 1$

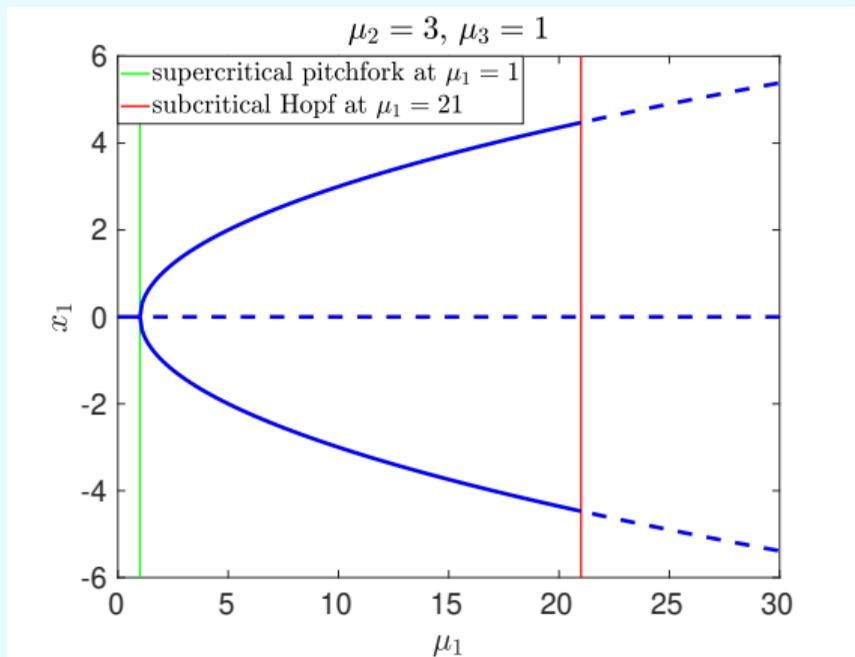
- supercritical pitchfork bifurcation

at $\mu_1 = 1$ ($\mathbf{x}_{c1} = \mathbf{0}$ is stable for $\mu_1 < 1$ and unstable for $\mu_1 > 1$)

[Question 6(e)]

- \mathbf{x}_{c2} and \mathbf{x}_{c3} are stable for $\mu_1 < 21$ and unstable for $\mu_1 > 21$

[Hopf bifurcations will be discussed in the second part of the course]



Lorenz equations: Hints for Question 6(e) on Problem Sheet 2

Denote $\nu = \mu - 1$ and consider the extended system

$$\frac{dx_1}{dt} = 3x_2 - 3x_1$$

$$\frac{dx_2}{dt} = \nu x_1 + x_1 - x_2 - x_1 x_3$$

$$\frac{dx_3}{dt} = x_1 x_2 - x_3$$

$$\frac{d\nu}{dt} = 0$$

Linearize about the critical point $\mathbf{x}_c = \mathbf{0}$, i.e. $[x_1, x_2, x_3, \nu] = [0, 0, 0, 0]$.

$$M = D\mathbf{f}(\mathbf{x}_c) = \begin{pmatrix} -3 & 3 & 0 & 0 \\ 1 & -1 & 0 & 0 \\ 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix},$$

Calculate eigenvalues: 0 (with multiplicity 2), -1 and -4 .

Lorenz equations: Hints for Question 6(e) on Problem Sheet 2

Introduce new variables

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ \nu \end{pmatrix} = \begin{pmatrix} 1 & 3 & 0 & 0 \\ 1 & -1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ \nu \end{pmatrix}, \quad \text{with inverse} \quad \begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ \nu \end{pmatrix} = \begin{pmatrix} 1/4 & 3/4 & 0 & 0 \\ 1/4 & -1/4 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ \nu \end{pmatrix}.$$

Then, we have

$$\frac{d}{dt} \begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ \nu \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & -4 & 0 & 0 \\ 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ \nu \end{pmatrix} + \frac{1}{4} \begin{pmatrix} 3\nu(y_1 + 3y_2) - 3(y_1 + 3y_2)y_3 \\ -\nu(y_1 + 3y_2) + (y_1 + 3y_2)y_3 \\ 4(y_1 + 3y_2)(y_1 - y_2) \\ 0 \end{pmatrix}.$$

Thus, a center manifold can be represented as a graph over the y_1 and ν variables

$$M_{\text{loc}}^c = \left\{ (y_1, y_2, y_3, \nu) \in \mathbb{R}^4 \mid y_2 = h_1(y_1, \nu), y_3 = h_2(y_1, \nu), \text{ with} \right. \\ \left. h_i(0, 0) = 0, \nabla h_i(0, 0) = [0, 0], \text{ for } i = 1, 2 \right\}.$$

Lorenz equations: Hints for Question 6(e) on Problem Sheet 2

Assume

$$\begin{aligned}y_2 &= h_1(y_1, \nu) = c_{120} y_1^2 + c_{111} y_1 \nu + c_{102} \nu^2 + \dots, \\y_3 &= h_2(y_1, \nu) = c_{220} y_1^2 + c_{211} y_1 \nu + c_{202} \nu^2 + \dots\end{aligned}$$

Differentiate and keep only quadratic terms to deduce

$$c_{120} = c_{102} = c_{211} = c_{202} = 0, \quad c_{111} = -\frac{1}{16}, \quad c_{220} = 1,$$

giving

$$\begin{aligned}y_2 &= h_1(y_1, \nu) = -\frac{y_1 \nu}{16} + \dots, \\y_3 &= h_2(y_1, \nu) = y_1^2 + \dots\end{aligned}$$

Conclude that the dynamics on the center manifold is given by

$$\frac{dy_1}{dt} = \frac{3y_1}{4} (\nu - y_1^2 + \dots)$$

i.e. we have a supercritical pitchfork bifurcation at the origin for $\mu = 1$ (which corresponds to $\nu = 0$).

Lorenz equations: Question 6 on Problem Sheet 2

$$\frac{dx_1}{dt} = 3(x_2 - x_1)$$

$$\frac{dx_2}{dt} = \mu_1 x_1 - x_2 - x_1 x_3$$

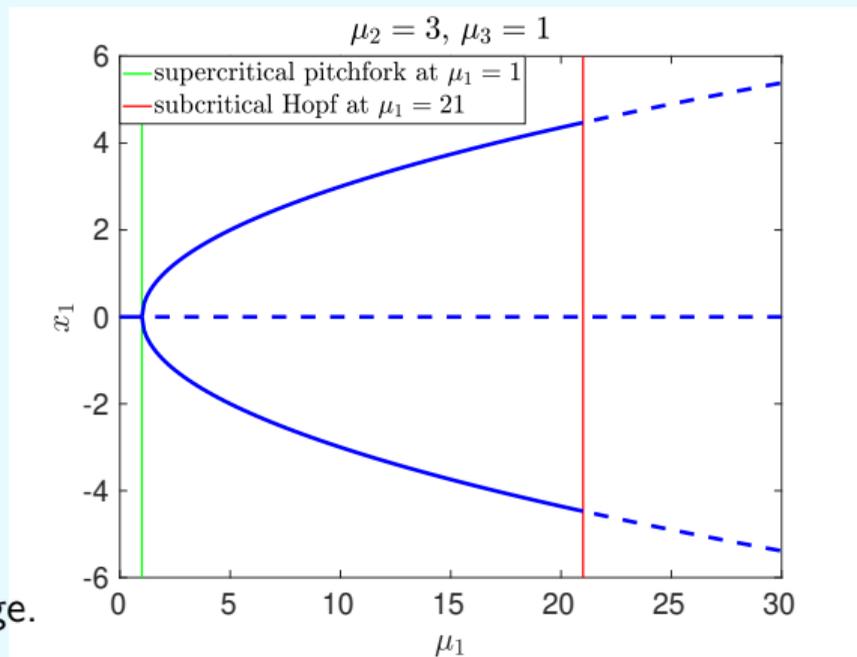
$$\frac{dx_3}{dt} = x_1 x_2 - x_3$$

Question 6(c) on Problem Sheet 2:

All trajectories eventually enter and remain inside a large sphere of the form

$$x_1^2 + x_2^2 + (x_3 - \mu_1 - 3)^2 = C(\mu_1)$$

where constant $C(\mu_1)$ is sufficiently large.



Lorenz equations: Question 6 on Problem Sheet 2

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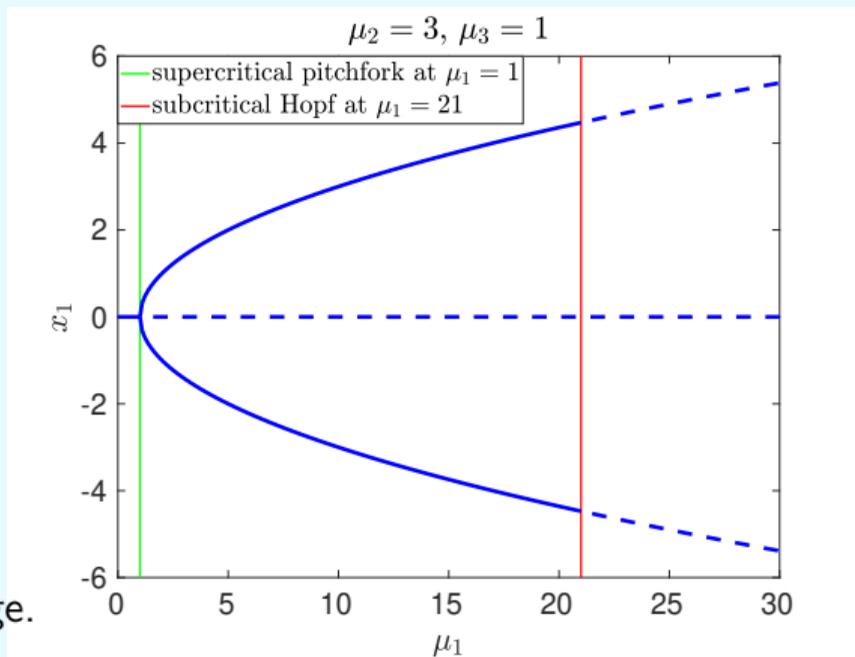
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Question 6(a): Let $U \equiv U(0) \subset \mathbb{R}^3$ be a compact connected subset of initial conditions. Let $U(t) = \phi_t(U)$ and $v(t) = |U(t)| = |\phi_t(U)|$ be the volume of $U(t)$.

Then

$$\lim_{t \rightarrow \infty} v(t) = 0$$

Lorenz equations: Question 6 on Problem Sheet 2

$$\frac{dx_1}{dt} = 3(x_2 - x_1)$$

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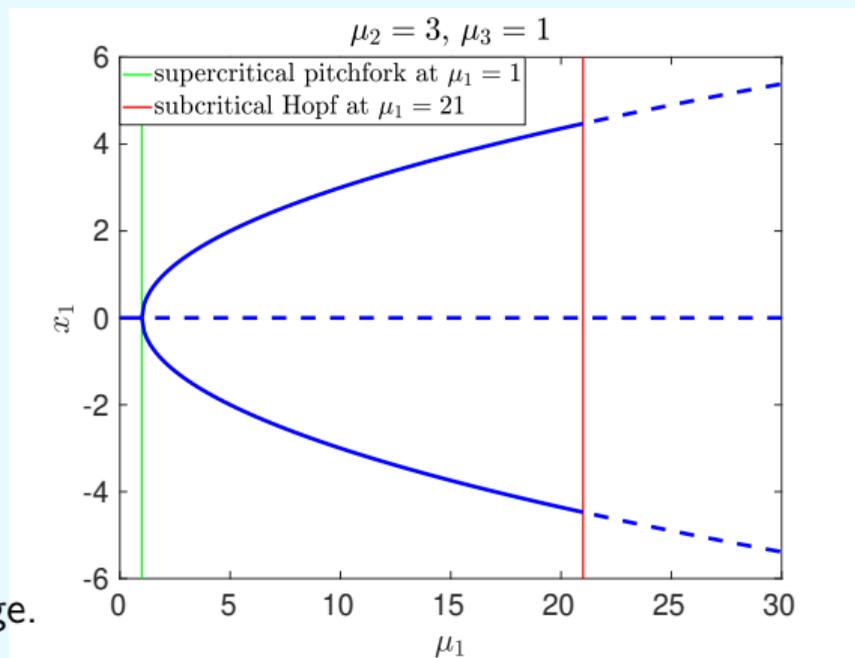
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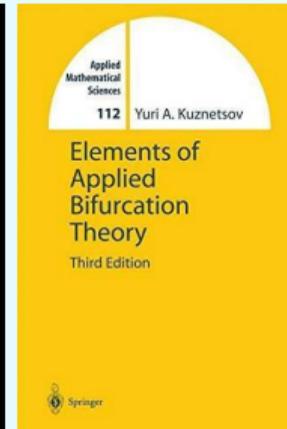
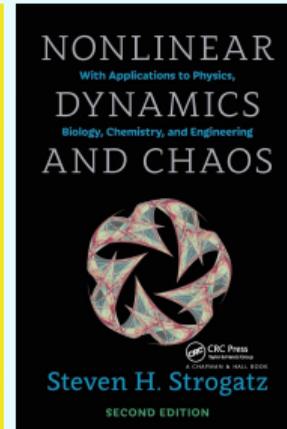
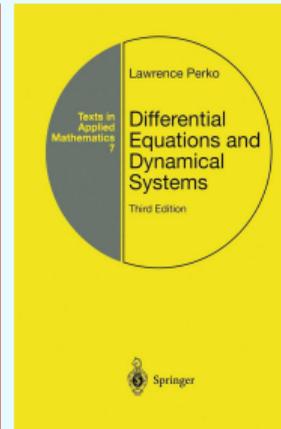
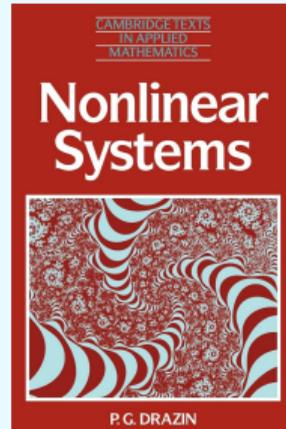
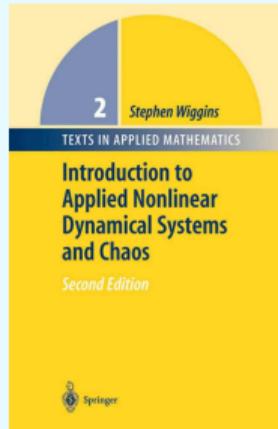
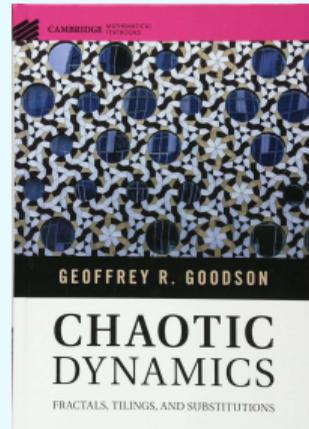
We have a trapping region and volume contraction, but there are no stable critical points or stable limit cycles for some parameter values!

We will continue with the analysis of the Lorenz system in the second part of the course after we introduce Hopf bifurcations.

End of Lecture 8: Goodbye, MSc students!

Course synopsis of **Lectures 1-8** (taken by both Part B and MSc students)

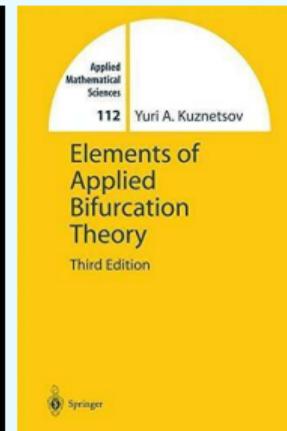
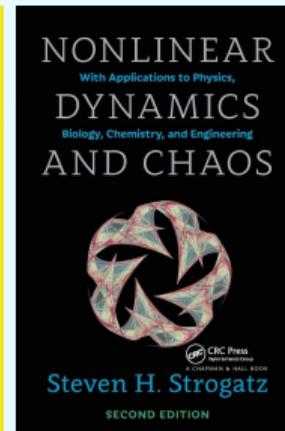
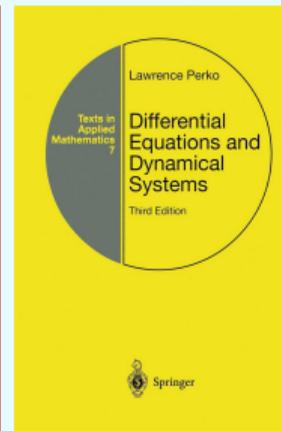
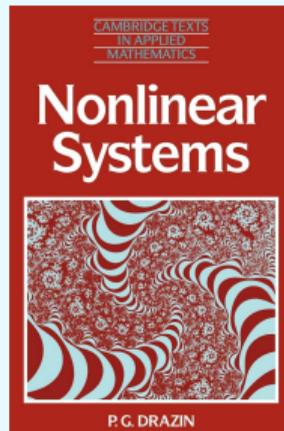
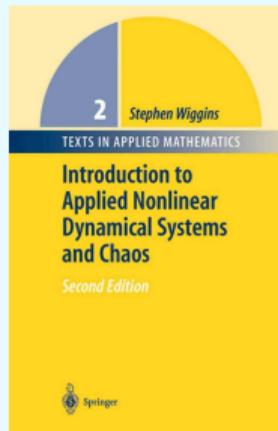
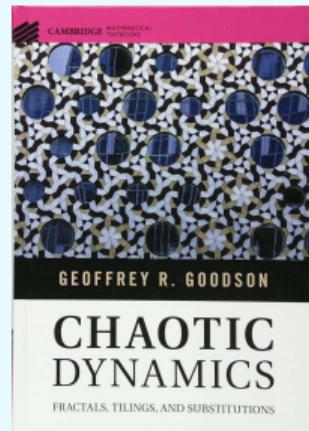
Discrete-time (maps) and continuous-time (differential equations) dynamical systems. Notion of flows, stability of fixed points, Lyapunov function, invariant manifolds, stable manifold theorem, notion of hyperbolicity, center manifold. Chemical reaction networks. Stable, unstable and center subspaces. Poincaré-Bendixson theorem. Periodic solutions, stable and unstable limit cycles. Introduction to bifurcation theory, covering saddle-node, transcritical, supercritical pitchfork and subcritical pitchfork bifurcations. Extended center manifold. Logistic map. Periodic points of maps. Stability of N -cycles. Period-doubling bifurcation. Sharkovsky's theorem. Invariant distribution.



Next week: Part B students only

Course synopsis of **Lectures 9-16** (Problem Sheets 3 and 4)

Bifurcations of limit cycles, covering supercritical and subcritical Hopf bifurcations, saddle-node bifurcation of cycles, infinite-period (SNIC) bifurcation and homoclinic (saddle-loop) bifurcation. Oscillations in chemical reaction networks. Weakly nonlinear oscillators. Poincaré-Lindstedt method. Conservative and non-conservative systems. Liénard systems, van der Pol oscillator. Hilbert's 16th problem. Lorenz equations. Lorenz map. Poincaré section. Poincaré map. Converse of Sharkovsky's theorem. Bernoulli shift map, symbolic dynamics. Tent map. Dynamics on metric spaces, sensitive dependence on initial conditions, transitivity, conjugate maps, chaotic dynamics.



B5.6: Nonlinear Dynamics, Bifurcations and Chaos (Lecture 9)

- summary of Lecture 8: we discussed
Extended center manifold. Bifurcations of fixed points. Lorenz equations: analysis using techniques in Lectures 1-8 (taken by both Part B and MSc students)
- today: we will start our discussion of Problem Sheet 3 (Part B students only)

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Bifurcations

Continuous-time dynamical system: Let $\mathbf{f} : \Omega \times \Theta \rightarrow \mathbb{R}^n$, where $\Omega \subset \mathbb{R}^n$ and $\Theta \subset \mathbb{R}^m$. Let $\mathbf{x}_0 \in \Omega$, $\boldsymbol{\mu} \in \Theta$ and $\mathbf{x}(t) \in \Omega$ be a solution of the ODE

$$\frac{d\mathbf{x}}{dt} = \mathbf{f}(\mathbf{x}; \boldsymbol{\mu}) \quad \text{with the initial condition} \quad \mathbf{x}(0) = \mathbf{x}_0 \in \Omega$$

Bifurcations: The qualitative structure of the flow can change as parameters $\boldsymbol{\mu}$ are varied. For example, critical points (fixed points) can be created or destroyed, or limit cycles can be created or destroyed. The parameter values at which these qualitative changes in the dynamics occur are called bifurcation points.

Bifurcations of fixed points

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Problem Sheet 2: bifurcations of fixed points

they can occur for $n \geq 1$, we studied examples with $n = 1$, $n = 2$ and $n = 3$

- saddle-node bifurcation
- transcritical bifurcation
- supercritical pitchfork bifurcation
- subcritical pitchfork bifurcation

Bifurcations of limit cycles

Continuous-time dynamical system: Let $\mathbf{f} : \Omega \times \Theta \rightarrow \mathbb{R}^n$, where $\Omega \subset \mathbb{R}^n$ and $\Theta \subset \mathbb{R}^m$. Let $\mathbf{x}_0 \in \Omega$, $\boldsymbol{\mu} \in \Theta$ and $\mathbf{x}(t) \in \Omega$ be a solution of the ODE

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Problem Sheet 3: bifurcations of limit cycles

they can occur for $n \geq 2$, we will first explain them on the case $n = 2$

- supercritical Hopf bifurcation
- subcritical Hopf bifurcation
- saddle-node bifurcation of cycles
- infinite-period bifurcation (SNIC, SNIPER)
- homoclinic bifurcation (saddle-loop bifurcation)

Supercritical Hopf bifurcation

example:

$$\frac{dx_1}{dt} = \mu x_1 - x_2 - x_1(x_1^2 + x_2^2)$$

$$\frac{dx_2}{dt} = x_1 + \mu x_2 - x_2(x_1^2 + x_2^2)$$

Supercritical Hopf bifurcation

example:

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fixed point at $\mathbf{0} = [0, 0]$

linearization $D\mathbf{f}(\mathbf{0}) = \begin{pmatrix} \mu & -1 \\ 1 & \mu \end{pmatrix}$

eigenvalues $\lambda_{\pm} = \mu \pm i$

Supercritical Hopf bifurcation

example:

$$\frac{dx_1}{dt} = \mu x_1 - x_2 - x_1(x_1^2 + x_2^2)$$

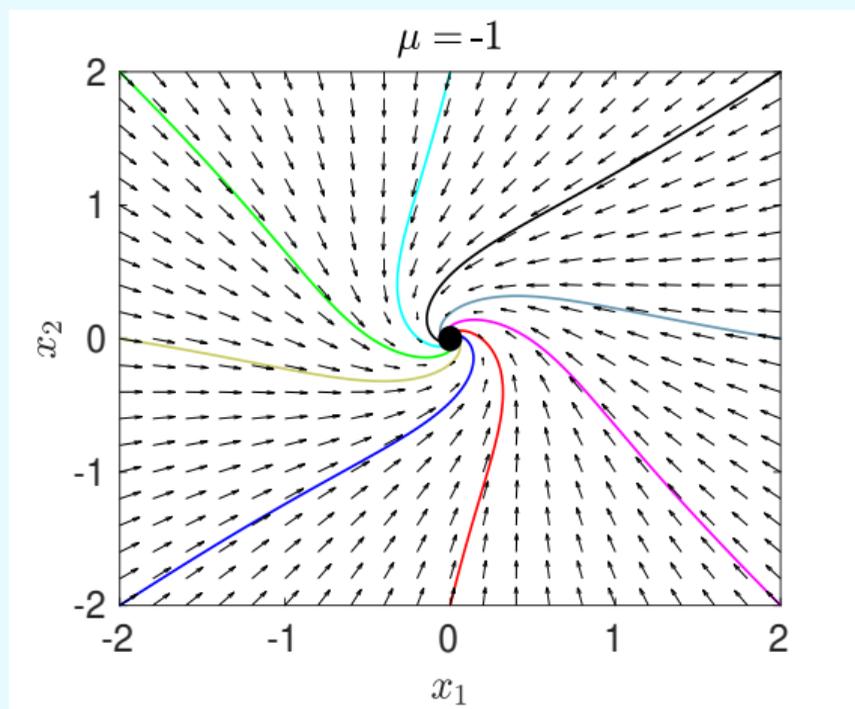
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Supercritical Hopf bifurcation

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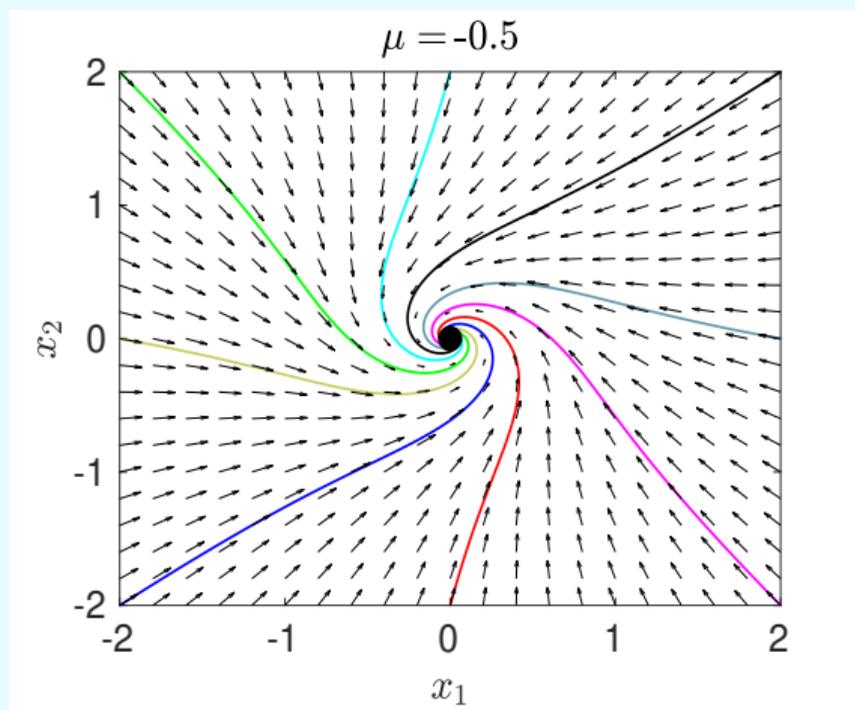
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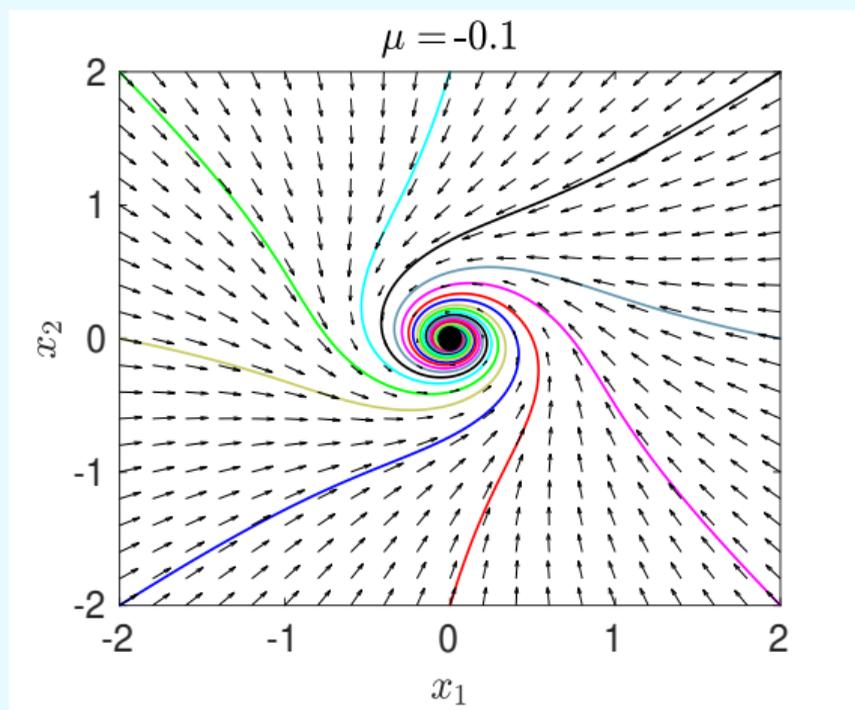
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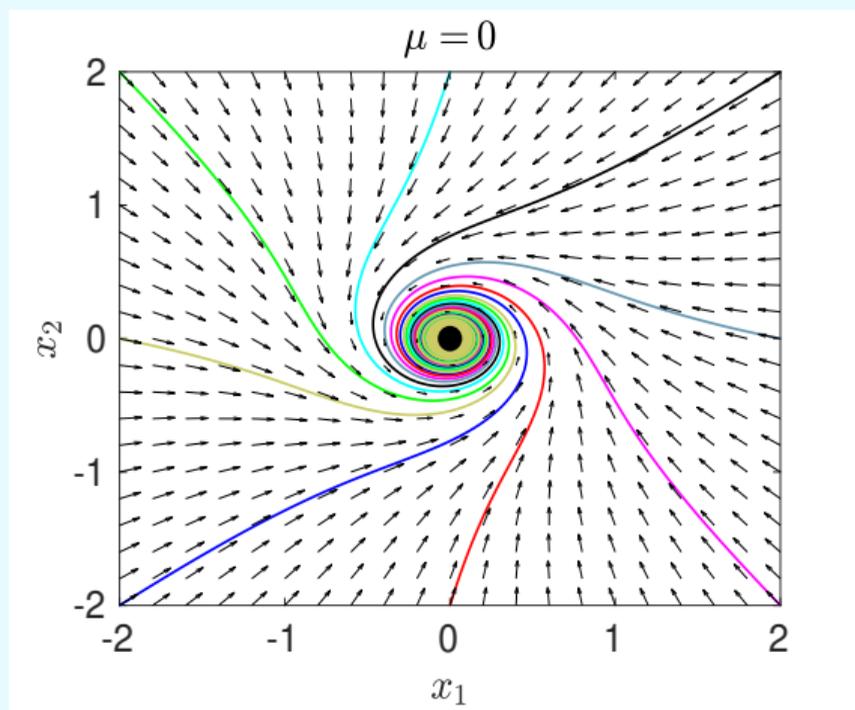
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$$\text{linearization } D\mathbf{f}(\mathbf{0}) = \begin{pmatrix} \mu & -1 \\ 1 & \mu \end{pmatrix}$$

eigenvalues $\lambda_{\pm} = \mu \pm i$

as μ increases from negative to positive values, eigenvalues cross the imaginary axis from left to right

$\mu = 0$: fixed point $\mathbf{0} = [0, 0]$ is a still **stable** spiral, though a very weak one



Supercritical Hopf bifurcation

example:

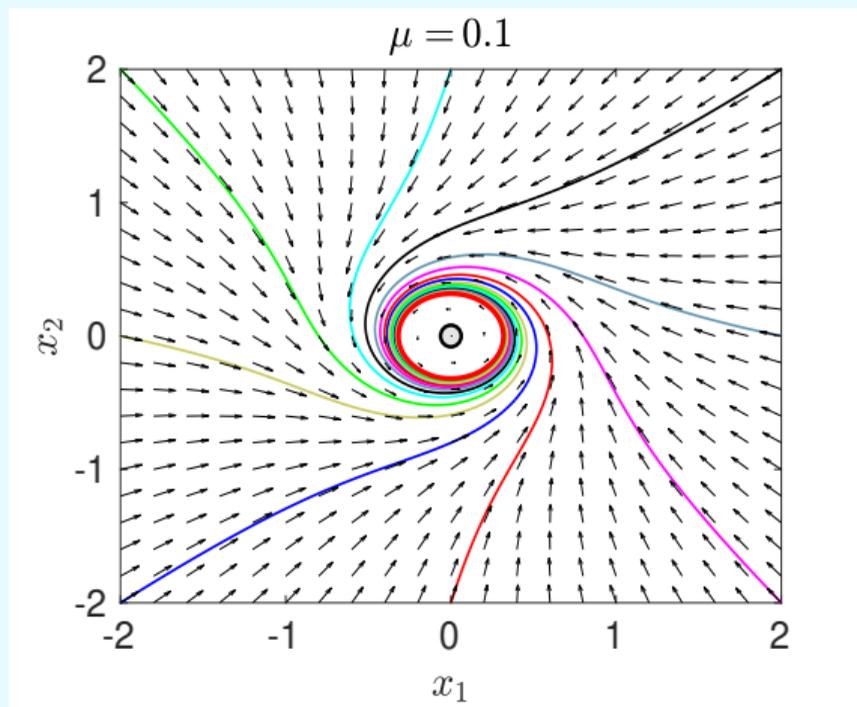
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fixed point at $\mathbf{0} = [0, 0]$

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eigenvalues $\lambda_{\pm} = \mu \pm i$



$\mu > 0$: fixed point $\mathbf{0} = [0, 0]$ is an **unstable** spiral

stable circular limit cycle of radius $r = \sqrt{\mu}$

Supercritical Hopf bifurcation

example:

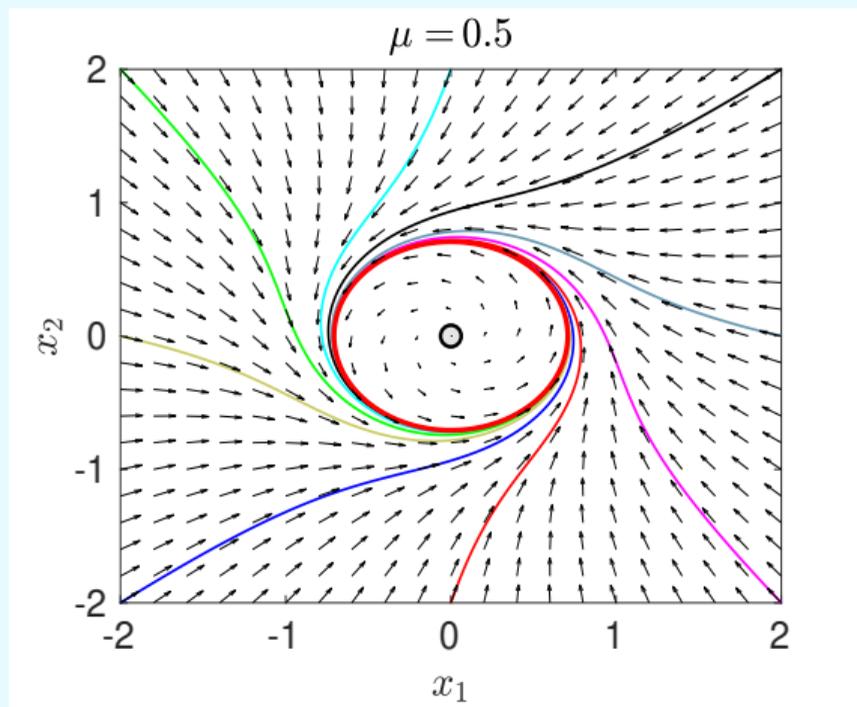
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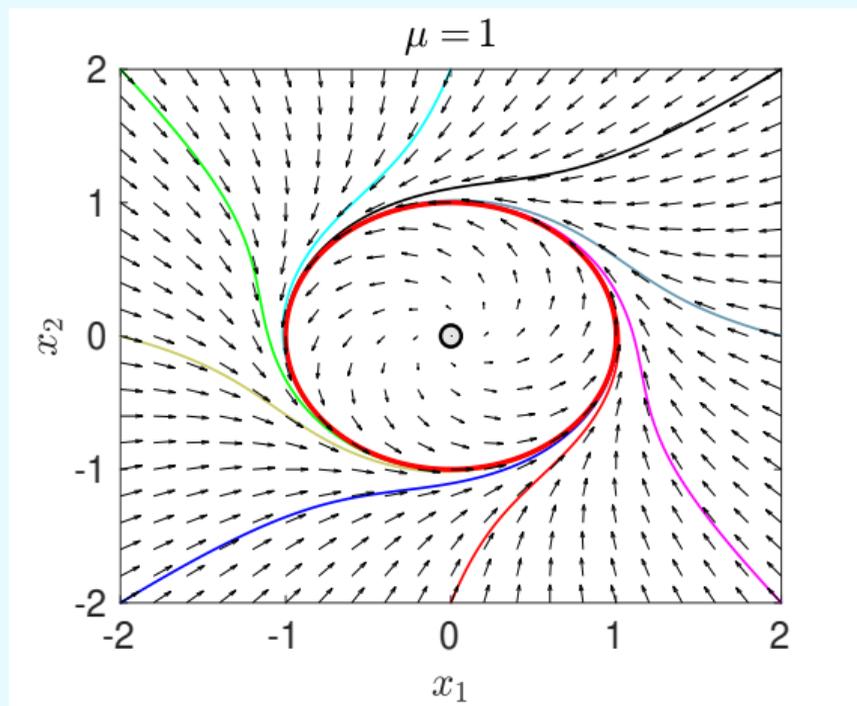
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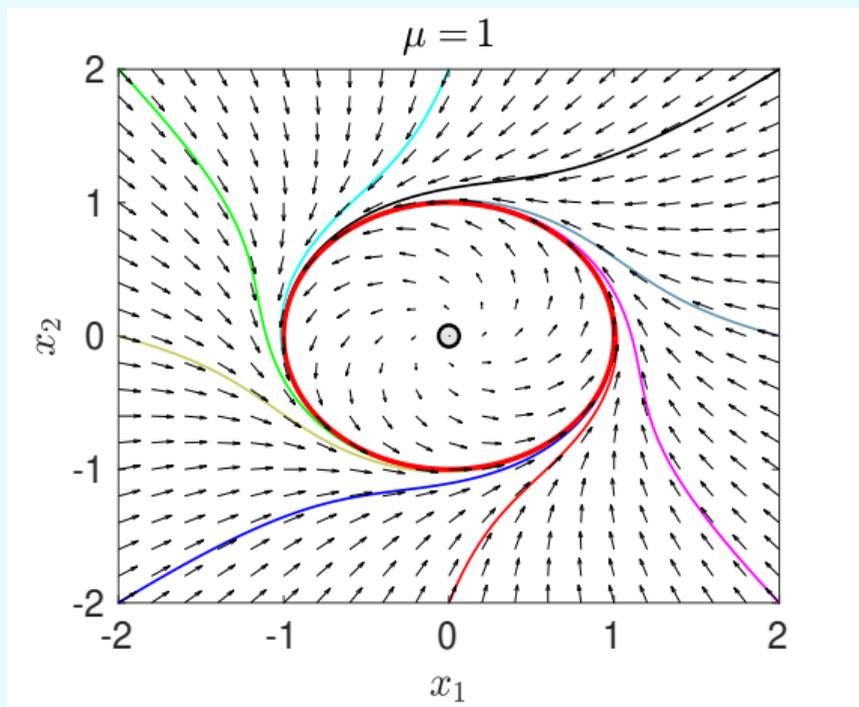
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We transform the ODEs to polar coordinates by using variables $r(t)$ and $\theta(t)$, where $x_1(t) = r(t) \cos \theta(t)$ and $x_2(t) = r(t) \sin \theta(t)$. We obtain

$$\frac{dr}{dt} = r(\mu - r^2) \qquad \frac{d\theta}{dt} = 1$$



Supercritical Hopf bifurcation

example:

$$\frac{dx_1}{dt} = \mu x_1 - x_2 - x_1(x_1^2 + x_2^2)$$

$$\frac{dx_2}{dt} = x_1 + \mu x_2 - x_2(x_1^2 + x_2^2)$$

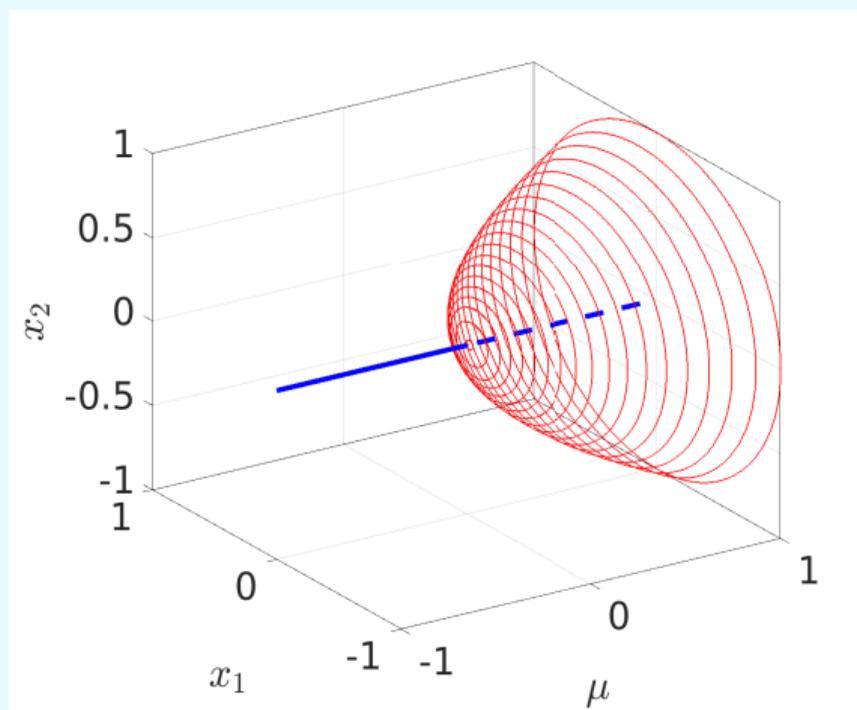
fixed point at $\mathbf{0} = [0, 0]$

$$\text{linearization } D\mathbf{f}(\mathbf{0}) = \begin{pmatrix} \mu & -1 \\ 1 & \mu \end{pmatrix}$$

eigenvalues $\lambda_{\pm} = \mu \pm i$

bifurcation diagram

[show 3D animation]



Hopf bifurcation - general case

general case: eigenvalues $\lambda(\mu) = \alpha(\mu) \pm i\omega(\mu)$ with $\alpha(0) = 0$ and $\omega(0) \neq 0$

behaviour close to the fixed point: normal form (in polar coordinates)

$$\frac{dr}{dt} = \alpha(\mu)r + a(\mu)r^3 + \mathcal{O}(r^5)$$

$$\frac{d\theta}{dt} = \omega(\mu) + b(\mu)r^2 + \mathcal{O}(r^4)$$

Taylor expanding:
$$\frac{dr}{dt} = \alpha'(0)\mu r + a(0)r^3 + \mathcal{O}(\mu^2 r, \mu r^3, r^5)$$

$$\frac{d\theta}{dt} = \omega(0) + \omega'(0)\mu + b(0)r^2 + \mathcal{O}(\mu^2, \mu r^2, r^4)$$

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$$\frac{d\theta}{dt} = \omega(0) + \omega'(0)\mu + b(0)r^2 + \mathcal{O}(\mu^2, \mu r^2, r^4)$$

our previous example: $\alpha'(0) = 1$, $a(0) = -1$, $\omega(0) = 1$, $\omega'(0) = b(0) = 0$

supercritical Hopf bifurcation: $a(0) < 0$ (periodic orbit is asymptotically stable)

subcritical Hopf bifurcation: $a(0) > 0$ (periodic orbit is unstable)

Supercritical Hopf bifurcation

general case: $a(0) < 0$

$$\frac{dr}{dt} = \alpha'(0) \mu r + a(0) r^3$$

$$\frac{d\theta}{dt} = \omega(0) + \omega'(0) \mu + b(0) r^2$$

Supercritical Hopf bifurcation

general case: $a(0) < 0$

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eigenvalues $\lambda_{\pm} = \alpha'(0) \mu \pm i \omega(0)$

Supercritical Hopf bifurcation

general case: $a(0) < 0$

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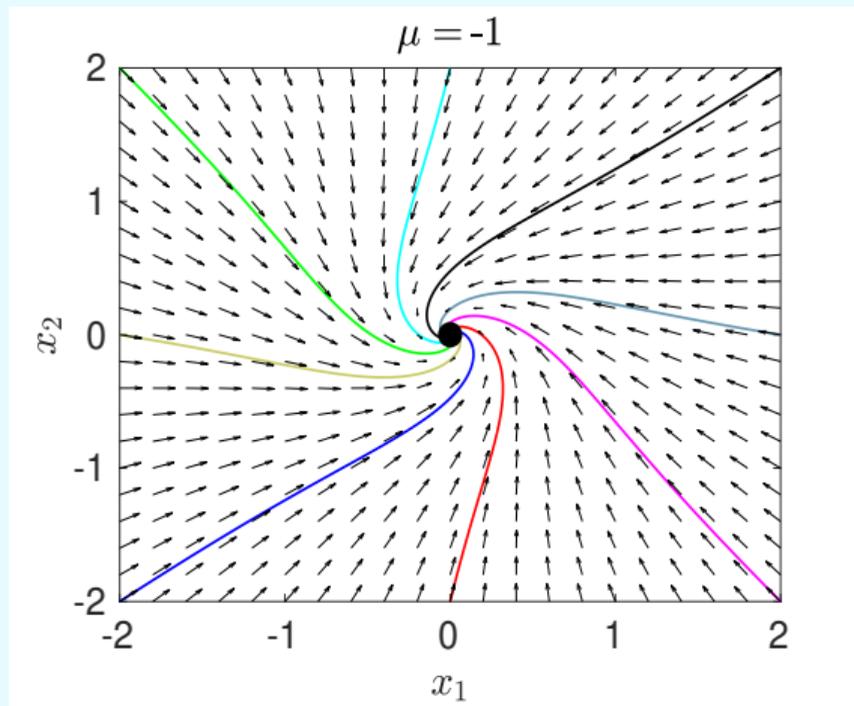
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stable circular limit cycle of radius $r = \sqrt{\mu}$



Supercritical Hopf bifurcation

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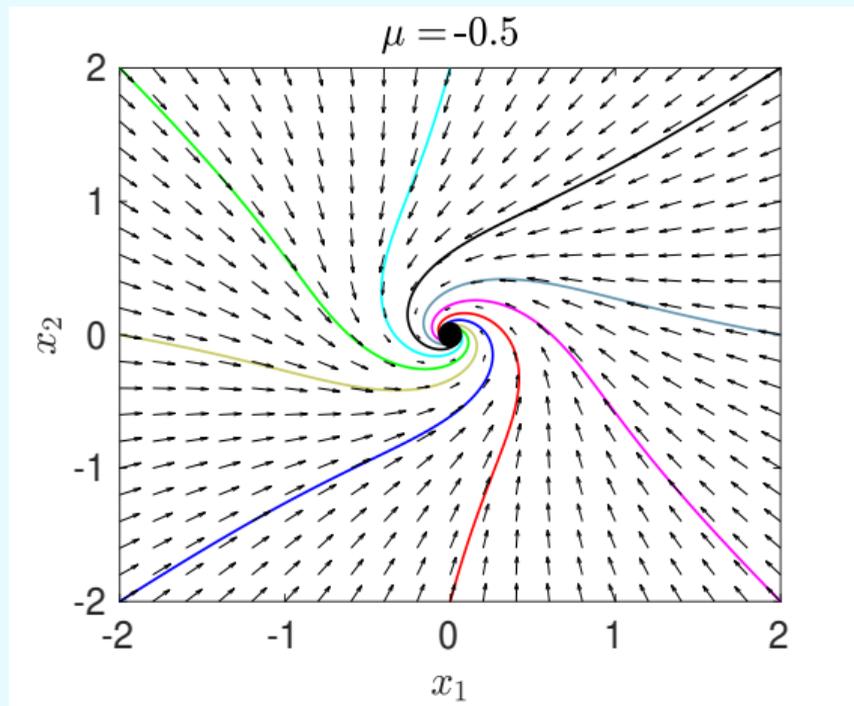
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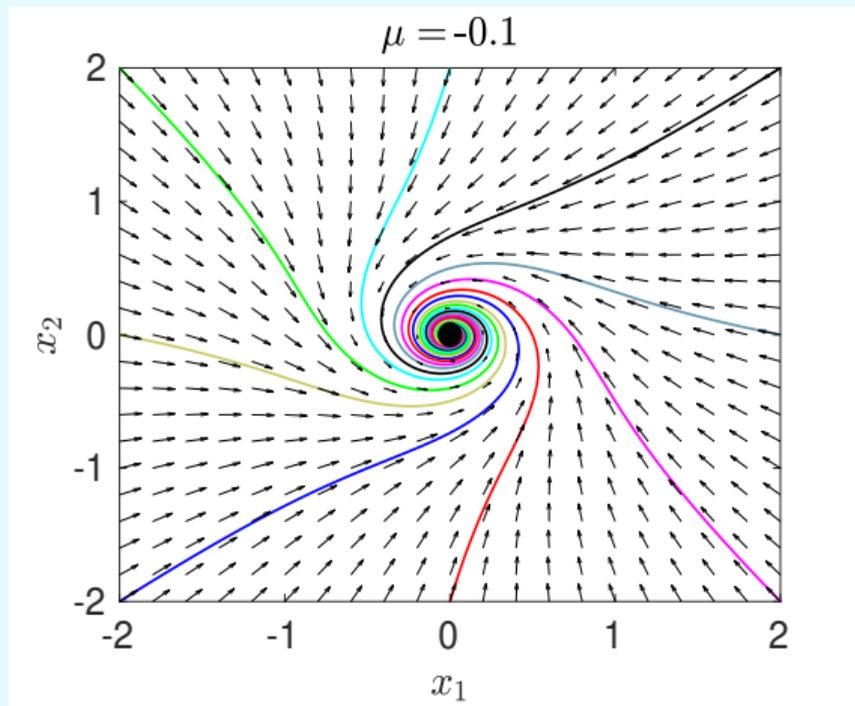
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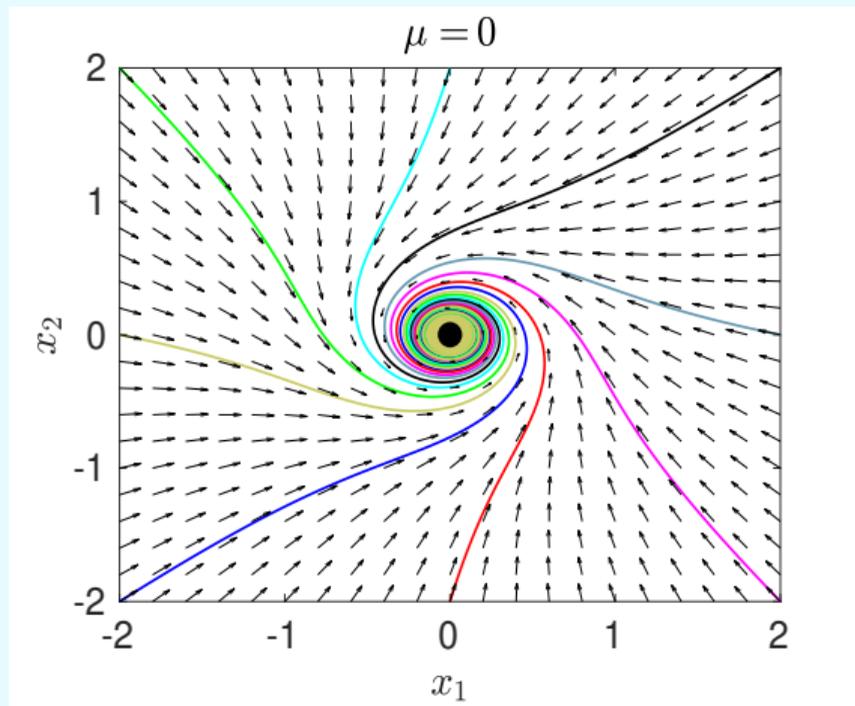
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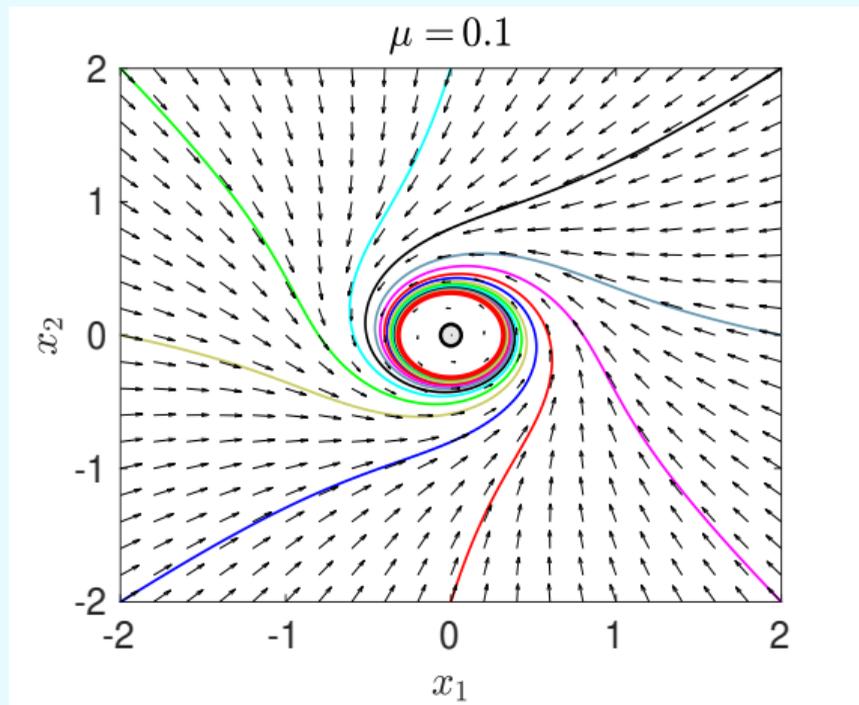
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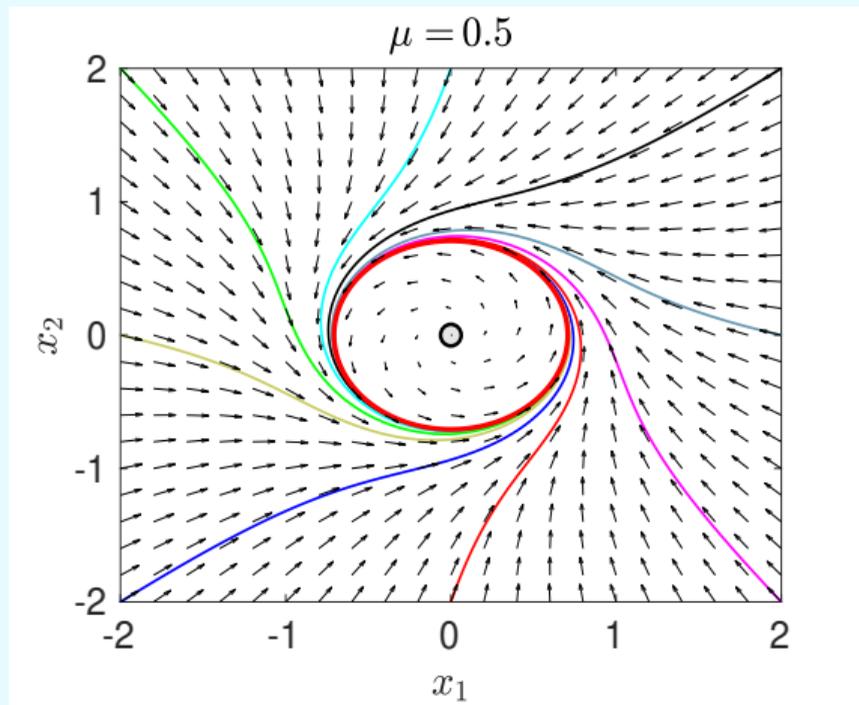
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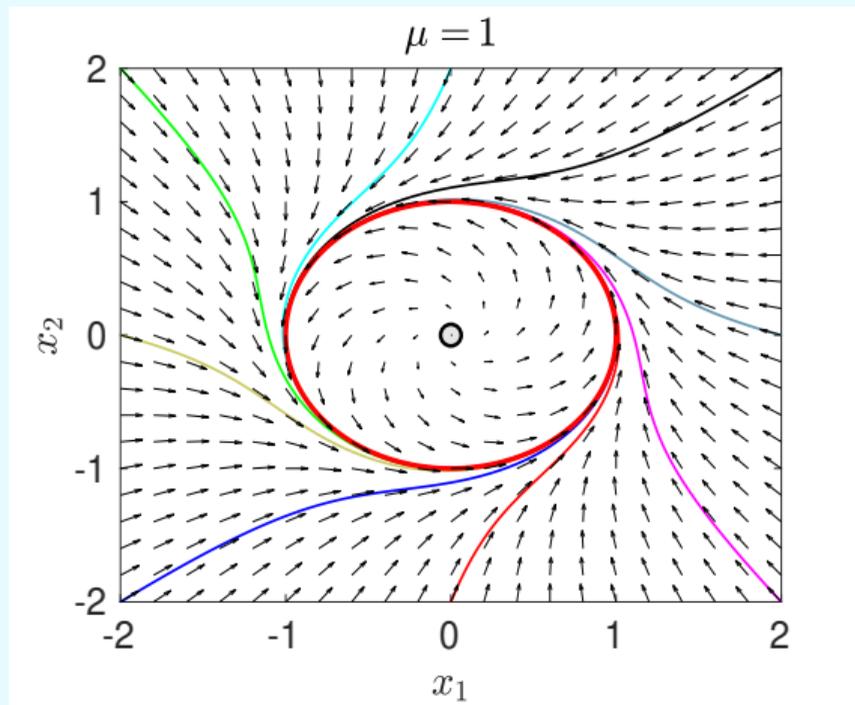
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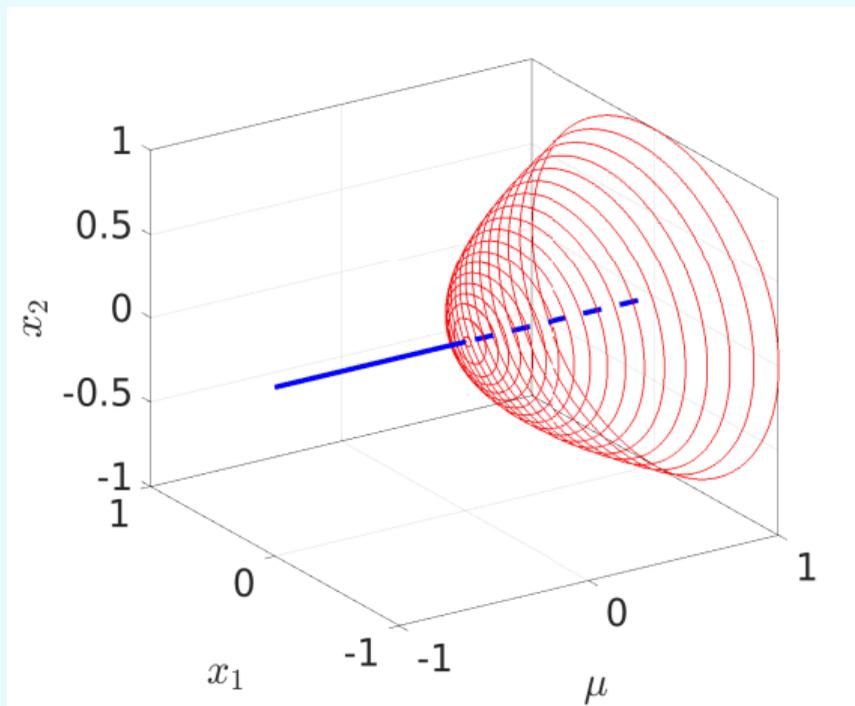
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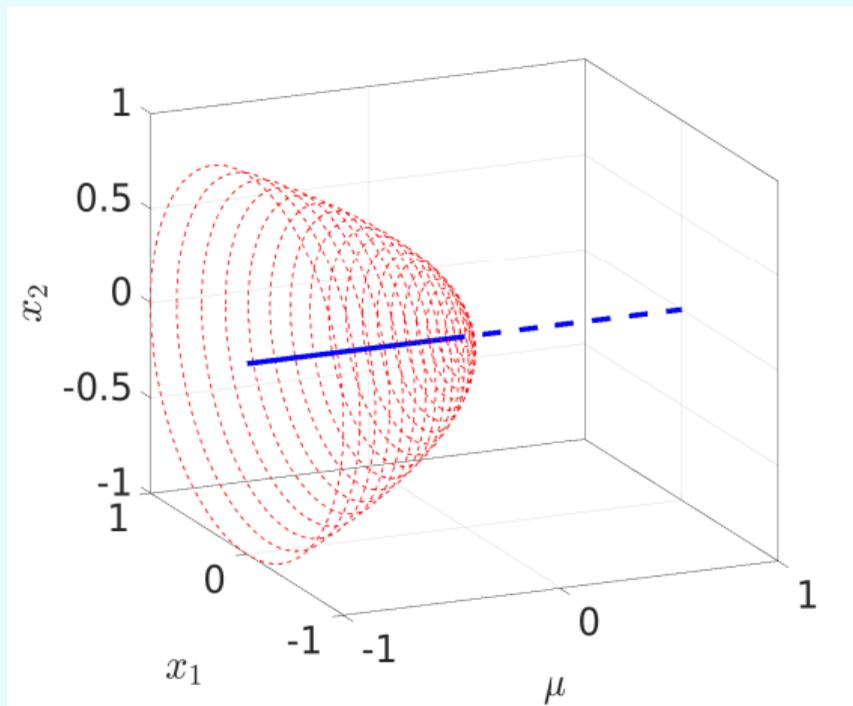
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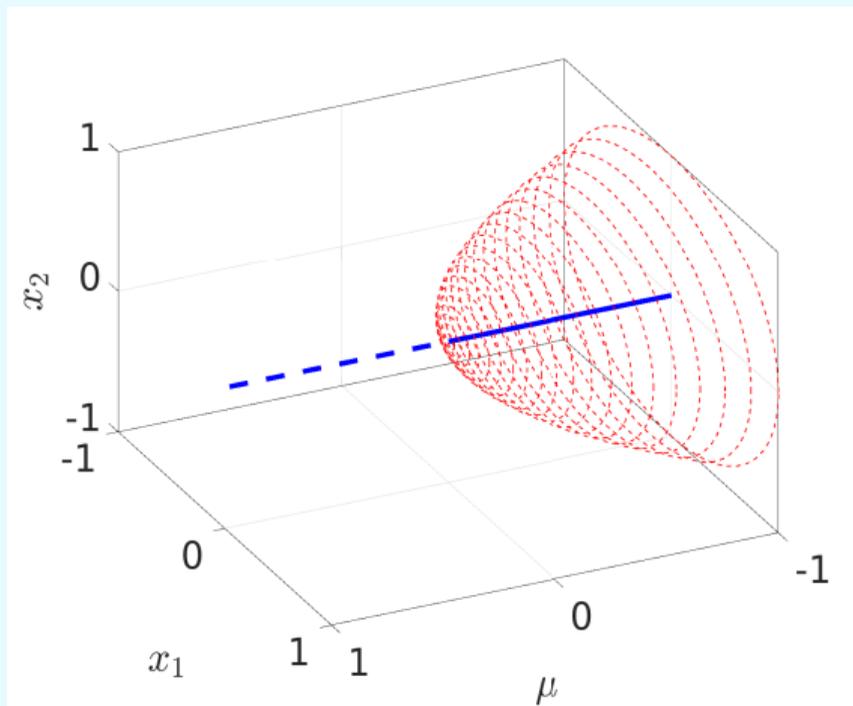
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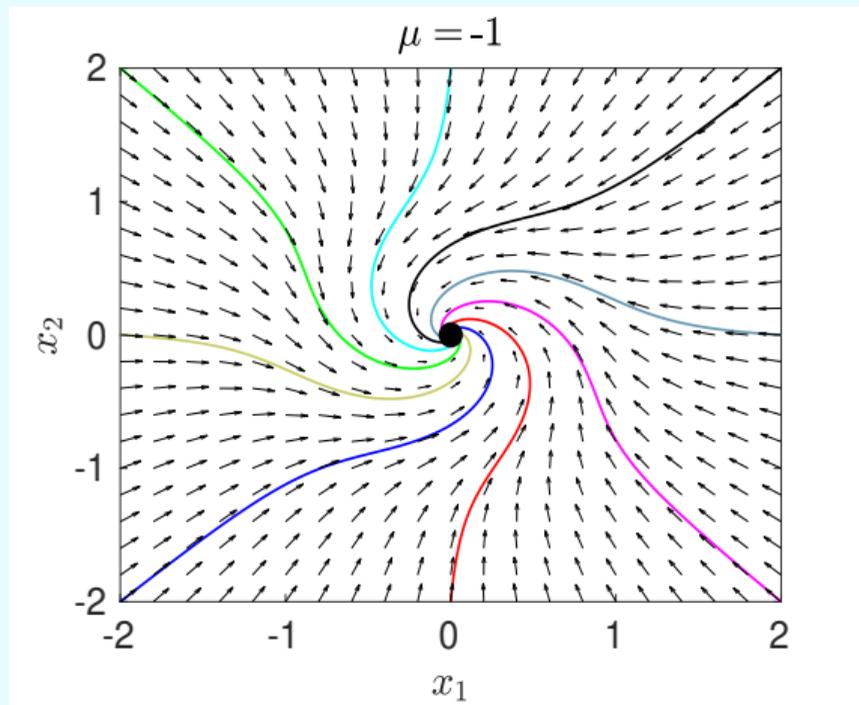
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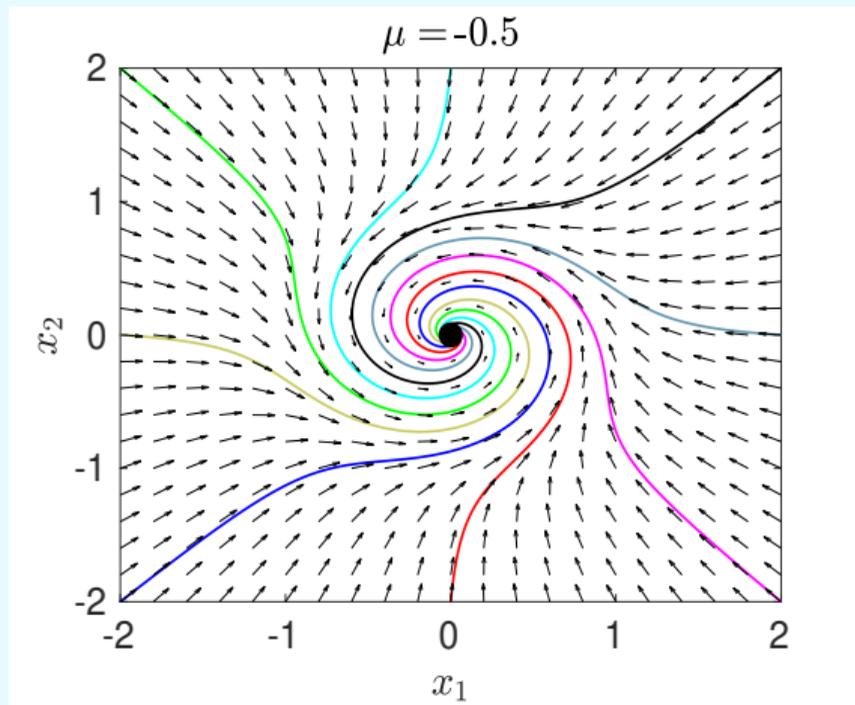
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Saddle-node bifurcation of cycles

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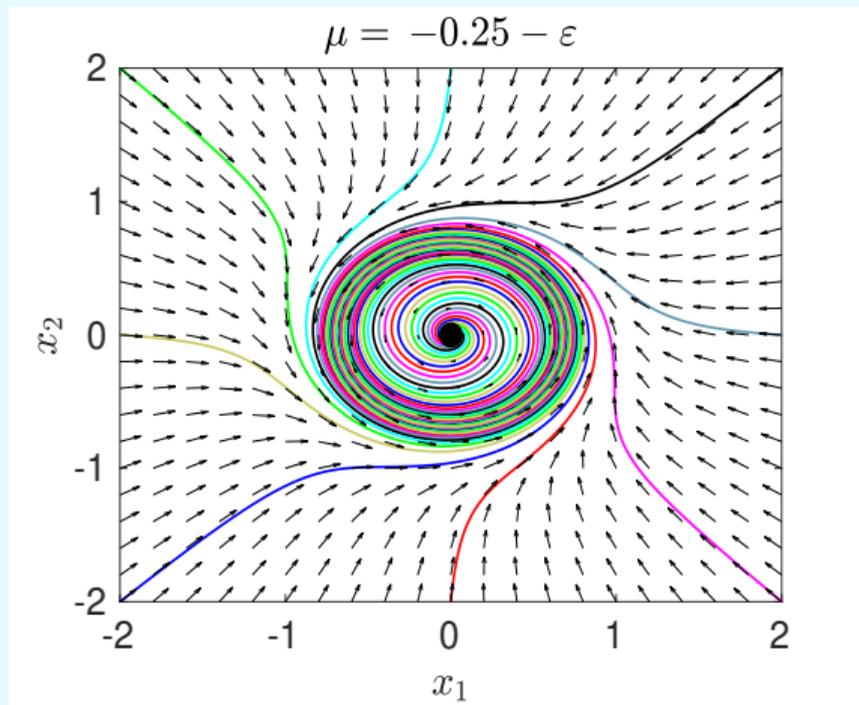
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saddle-node bifurcation of cycles at $\mu = -1/4$: a half-stable cycle appears, it splits into a pair of limit cycles for $\mu > -1/4$, one stable, one unstable



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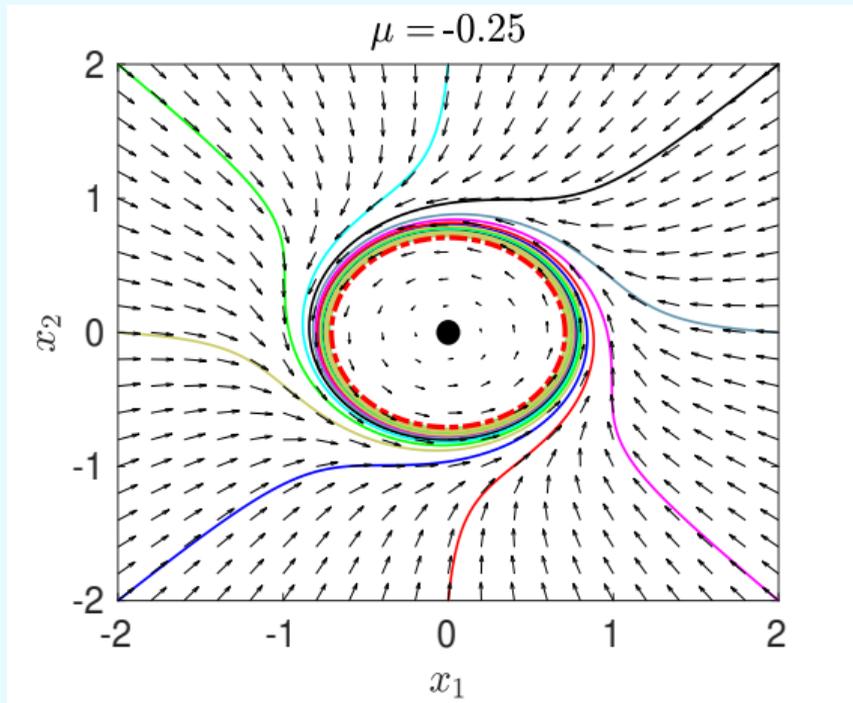
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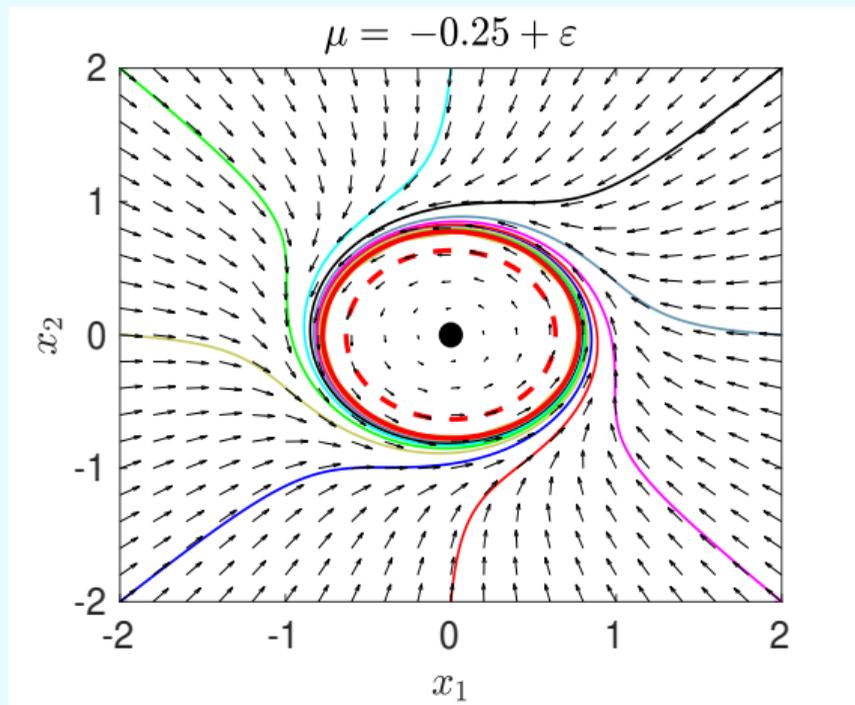
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saddle-node bifurcation of cycles at $\mu = -1/4$: viewed in the other direction, a stable and unstable cycle collide and disappear as μ decreases through $\mu = -1/4$



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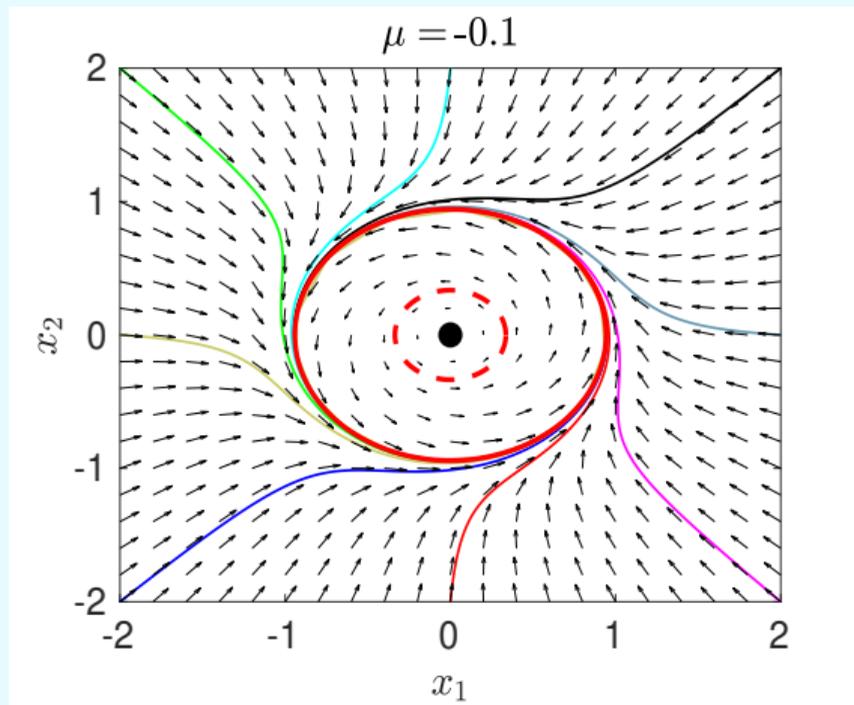
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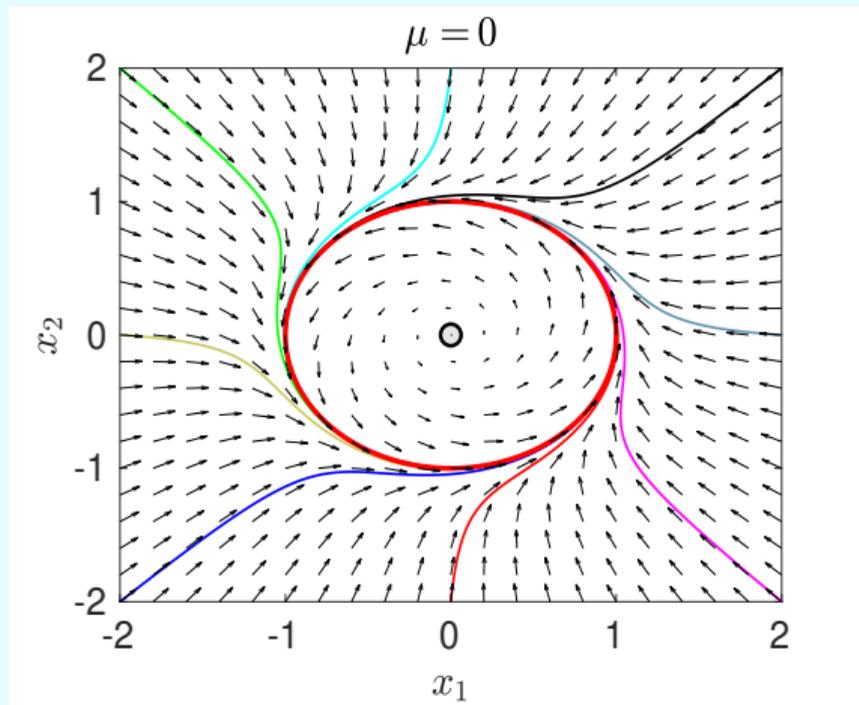
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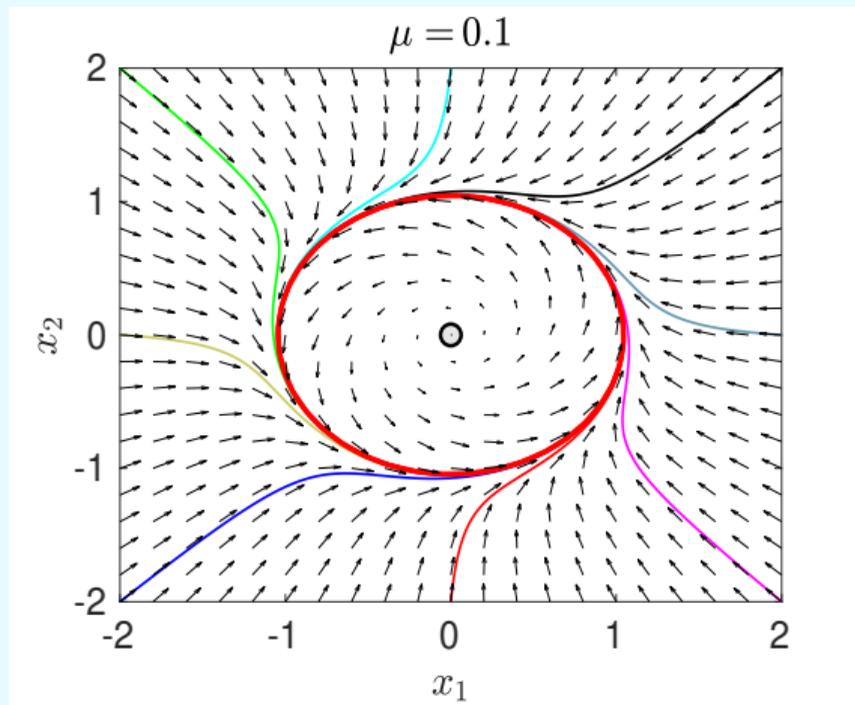
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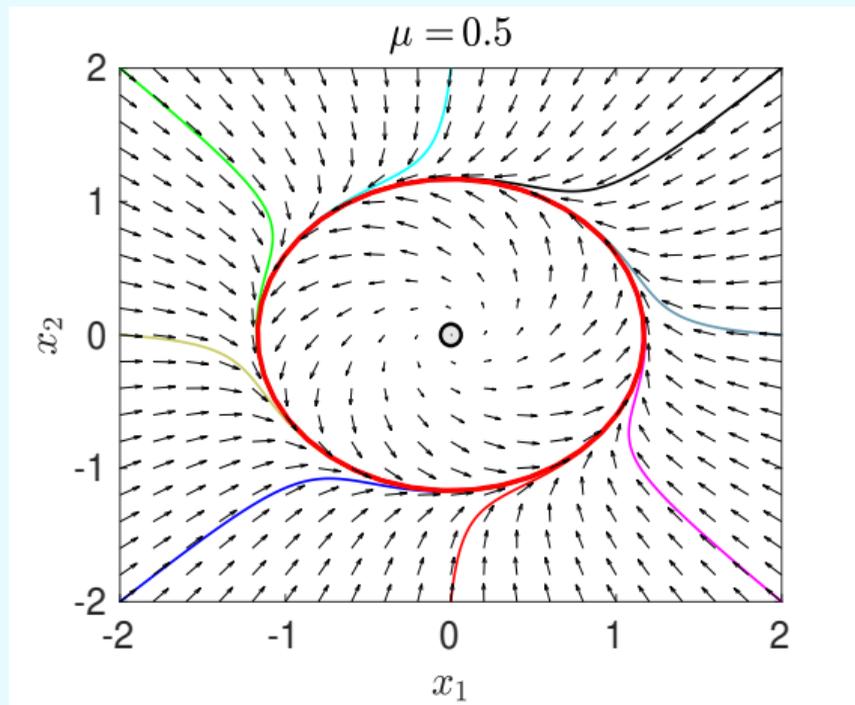
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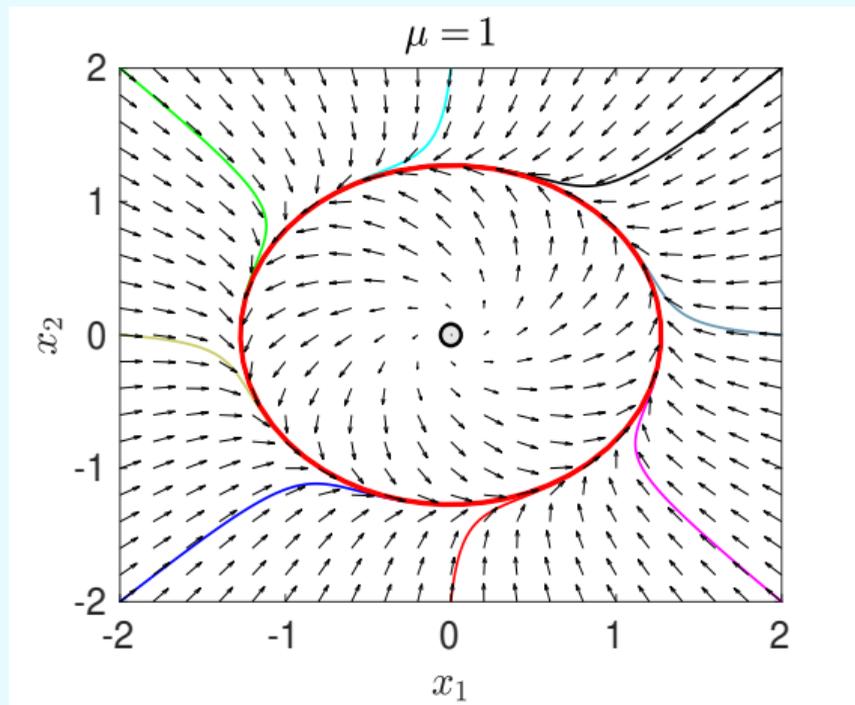
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subcritical Hopf bifurcation at $\mu = 0$



Subcritical Hopf bifurcation and saddle-node bifurcation of cycles

general case: $a(0) > 0$

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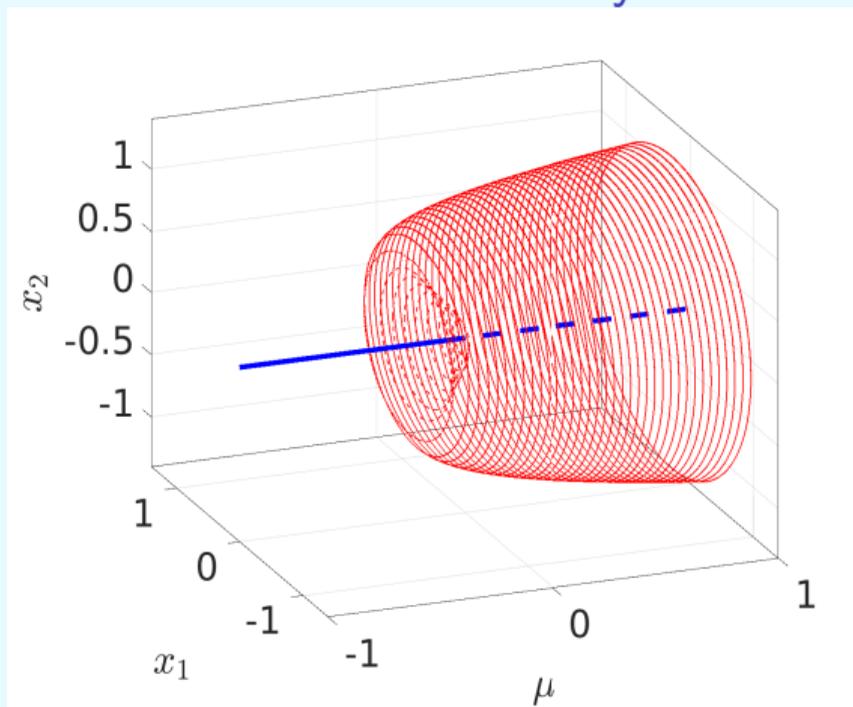
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subcritical Hopf bifurcation at $\mu = 0$

saddle-node bifurcation of cycles at $\mu = -1/4$



Subcritical Hopf bifurcation

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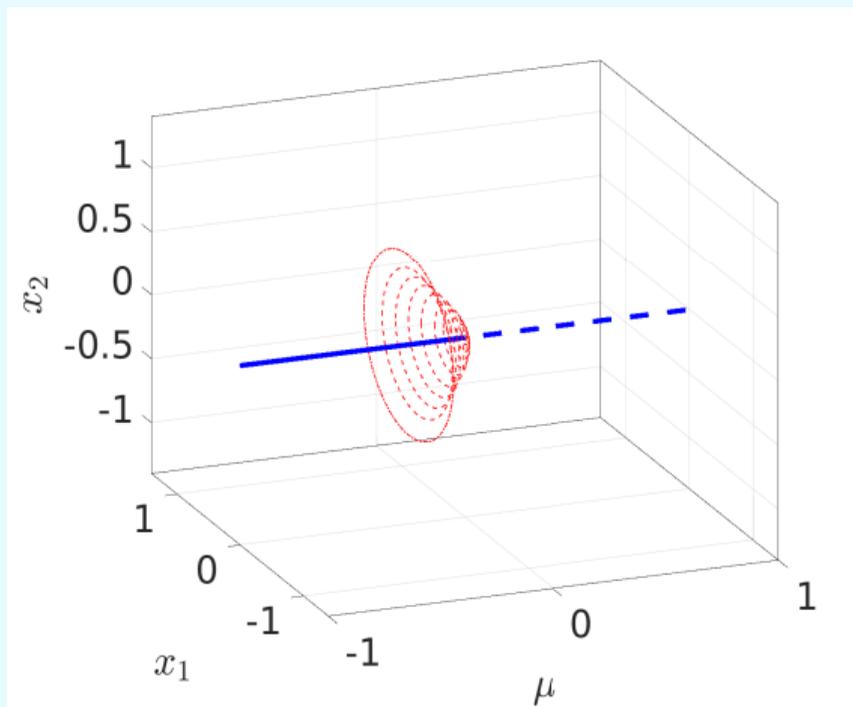
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subcritical Hopf bifurcation at $\mu = 0$



Transforming into normal form

example:

$$\frac{dx_1}{dt} = \mu x_1 - x_2 - x_1^3$$

$$\frac{dx_2}{dt} = x_1$$

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$$\frac{dx_1}{dt} = \mu x_1 - x_2 - x_1^3$$

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$z = x_1 + ix_2$, $z^* = x_1 - ix_2$, $\mu = 0$:

$$\frac{dz}{dt} = iz - \frac{(z + z^*)^3}{8} = iz - \frac{1}{8} z^3 - \frac{3}{8} z^2 z^* - \frac{3}{8} z (z^*)^2 - \frac{1}{8} (z^*)^3$$

Transforming into normal form

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near identity change of coordinates: $w = z + Az^3 + Bz^2z^* + Cz(z^*)^2 + D(z^*)^3$

$$\frac{dw}{dt} = iw - \frac{(w + w^*)^3}{8} + 2iAw^3 - 2iCw(w^*)^2 - 4iD(w^*)^3 + \mathcal{O}(|w|^4)$$

$$\frac{dw}{dt} = iw - \frac{3}{8}w^2w^* + \mathcal{O}(|w|^4) = iw - \frac{3}{8}w|w|^2 + \mathcal{O}(|w|^4)$$

$a(0) = -3/8 < 0$, supercritical Hopf bifurcation

Transforming into normal form

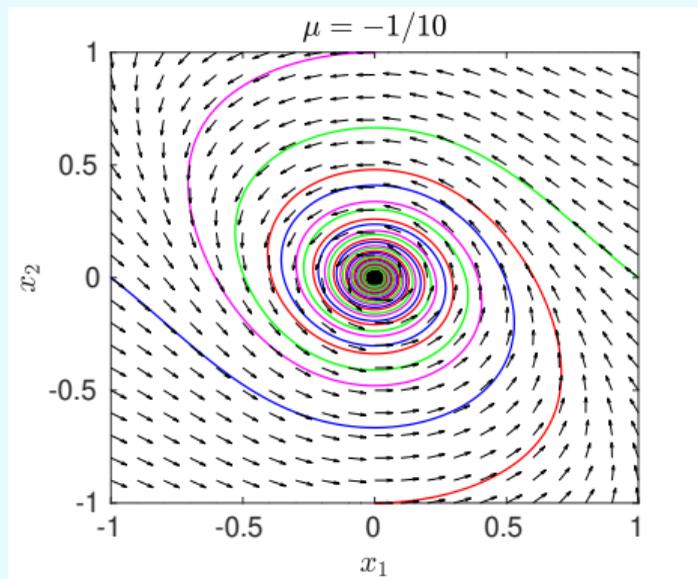
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$$\frac{dw}{dt} = iw - \frac{3}{8}w^2w^* + \mathcal{O}(|w|^4) = iw - \frac{3}{8}w|w|^2 + \mathcal{O}(|w|^4)$$

$a(0) = -3/8 < 0$, supercritical Hopf bifurcation

Transforming into normal form

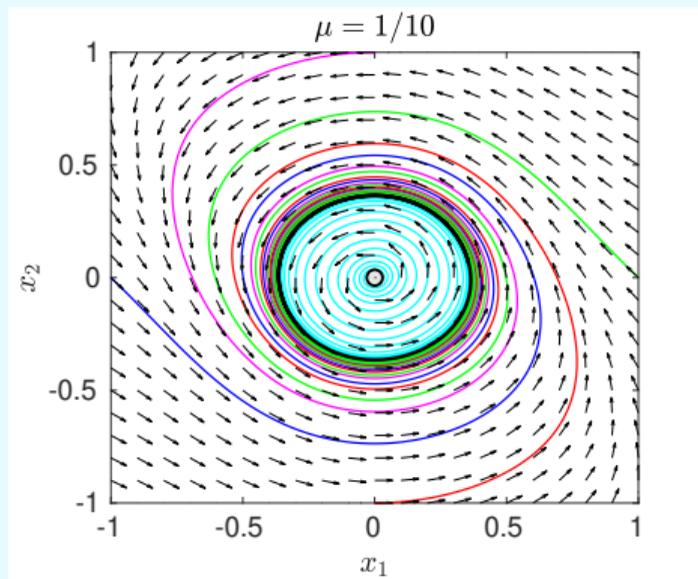
example:

$$\frac{dx_1}{dt} = \mu x_1 - x_2 - x_1^3$$

$$\frac{dx_2}{dt} = x_1$$

$$z = x_1 + ix_2, \quad z^* = x_1 - ix_2, \quad \mu = 0:$$

$$\frac{dz}{dt} = iz - \frac{(z + z^*)^3}{8}$$



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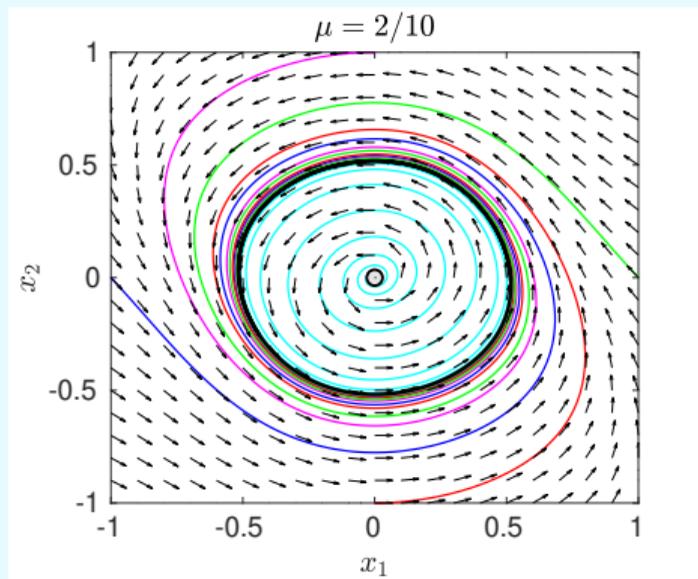
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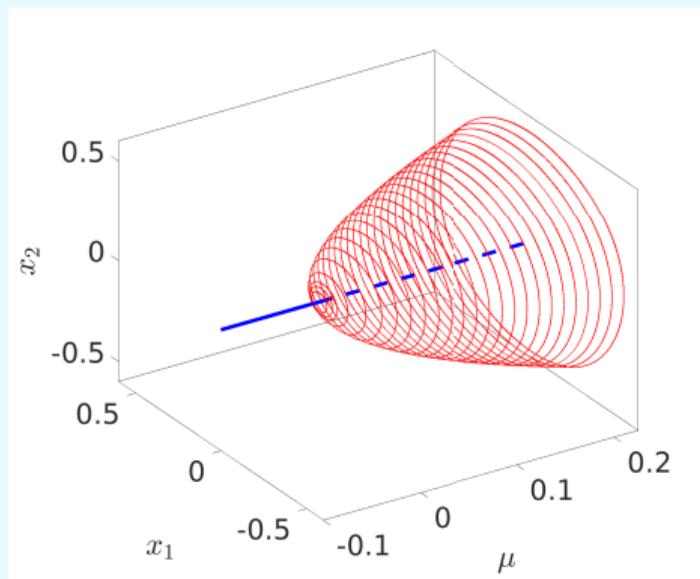
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B5.6: Nonlinear Dynamics, Bifurcations and Chaos (Lecture 10)

- summary of Lecture 9: we discussed Bifurcations of limit cycles, covering supercritical and subcritical Hopf bifurcations, saddle-node bifurcation of cycles.
- today: we will continue in our discussion of Problem Sheet 3

B5.6: Nonlinear Dynamics, Bifurcations and Chaos (Lecture 10)

- summary of Lecture 9: we discussed Bifurcations of limit cycles, covering supercritical and subcritical Hopf bifurcations, saddle-node bifurcation of cycles.
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- course synopsis of **Lectures 9-16**:
Bifurcations of limit cycles, covering supercritical and subcritical Hopf bifurcations, saddle-node bifurcation of cycles, infinite-period (SNIC) bifurcation and homoclinic (saddle-loop) bifurcation. Oscillations in chemical reaction networks. Weakly nonlinear oscillators. Poincaré-Lindstedt method. Conservative and non-conservative systems. Liénard systems, van der Pol oscillator. Hilbert's 16th problem. Lorenz equations. Lorenz map. Poincaré section. Poincaré map. Converse of Sharkovsky's theorem. Bernoulli shift map, symbolic dynamics. Tent map. Dynamics on metric spaces, sensitive dependence on initial conditions, transitivity, conjugate maps, chaotic dynamics.

Question 6 on Problem Sheet 1

System of $n = 2$ chemical species X_1 and X_2 which are subject to $\ell = 4$ reactions:



Question 6 on Problem Sheet 1

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Assuming mass action kinetics, concentrations $x_1(t)$ and $x_2(t)$ evolve by

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Question 6 on Problem Sheet 1

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$$\frac{dx_2}{dt} = -k_1 x_1^2 x_2 + k_4$$

Using $k_1 = k_2 = 1$, $k_3 = \mu$ and $k_4 = 2$, we get: $\frac{dx_1}{dt} = x_1^2 x_2 + 1 - \mu x_1$

$$\frac{dx_2}{dt} = -x_1^2 x_2 + 2$$

Question 6 on Problem Sheet 1: We considered $\mu = 9$. We showed that the fixed point $[1/3, 18]$ is unstable and we found a trapping region (closed bounded connected set such that the vector field points inward everywhere on its boundary). We applied the Poincaré-Bendixson theorem to prove that there exists a periodic solution.

Question 6 on Problem Sheet 1

$$\frac{dx_1}{dt} = x_1^2 x_2 + 1 - \mu x_1$$

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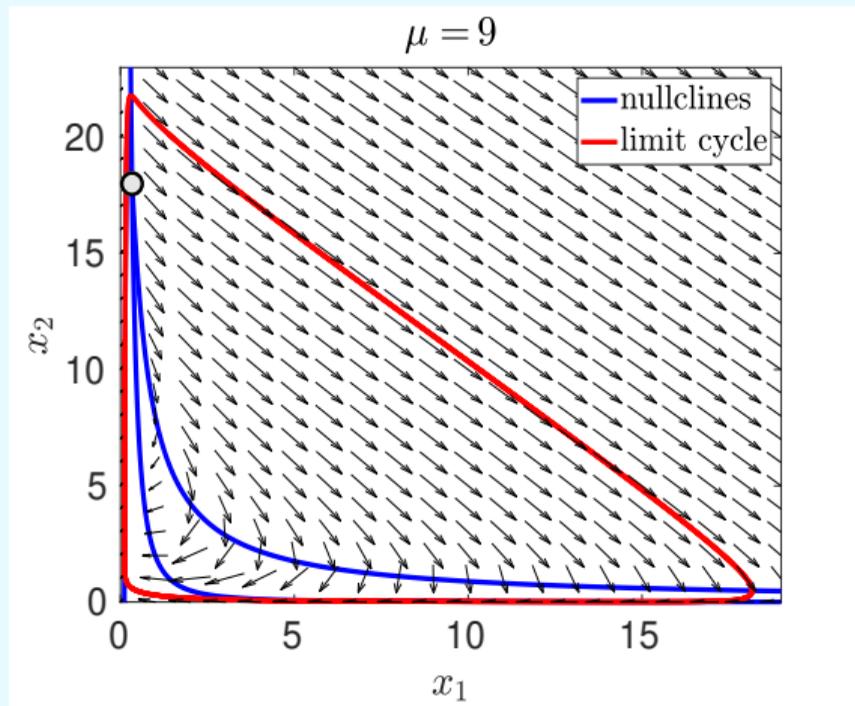
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$$\mu = 9$$



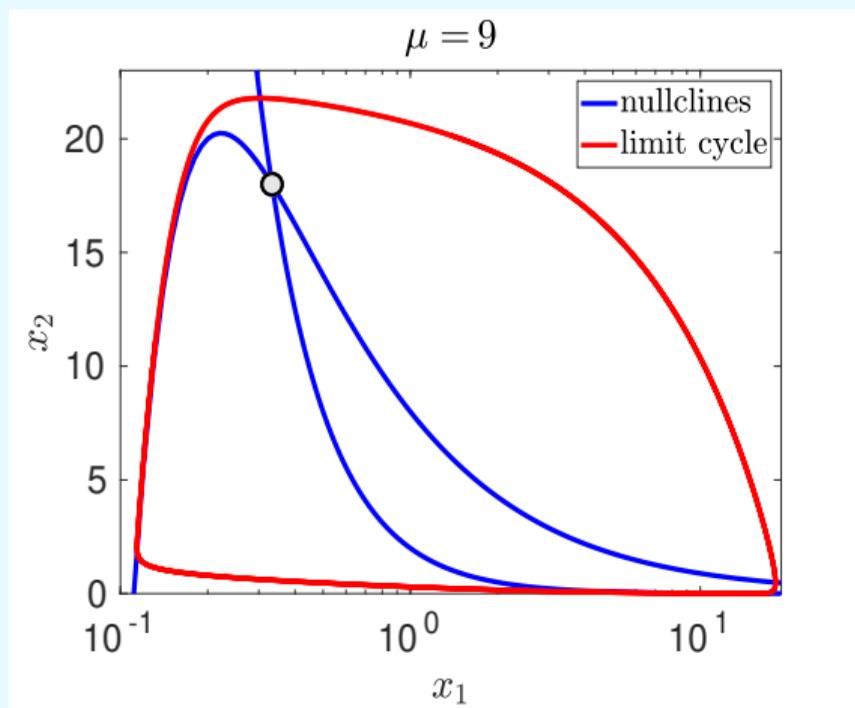
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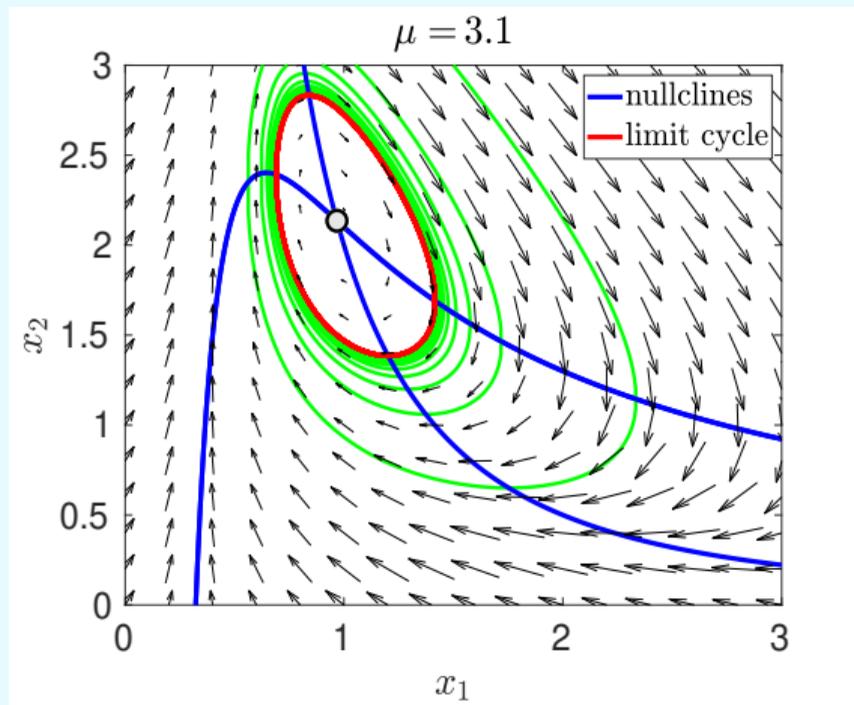


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We decrease the value of parameter μ and the limit cycle shrinks.

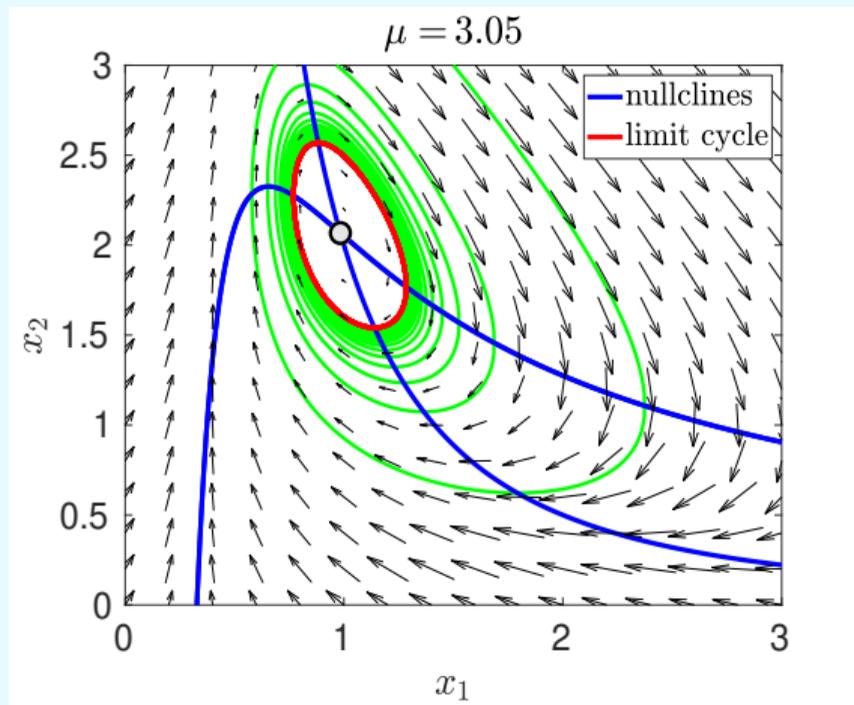


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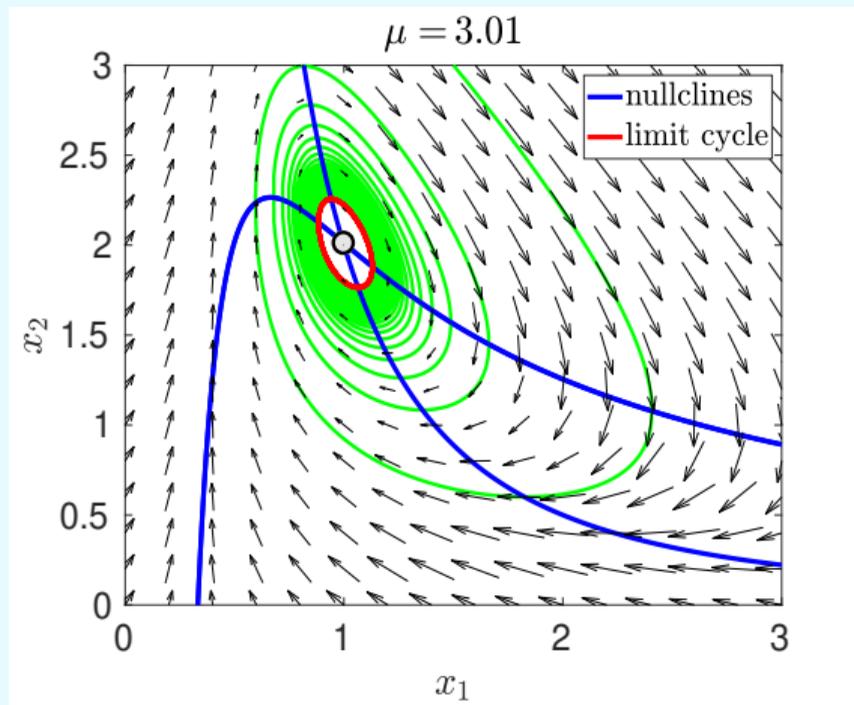


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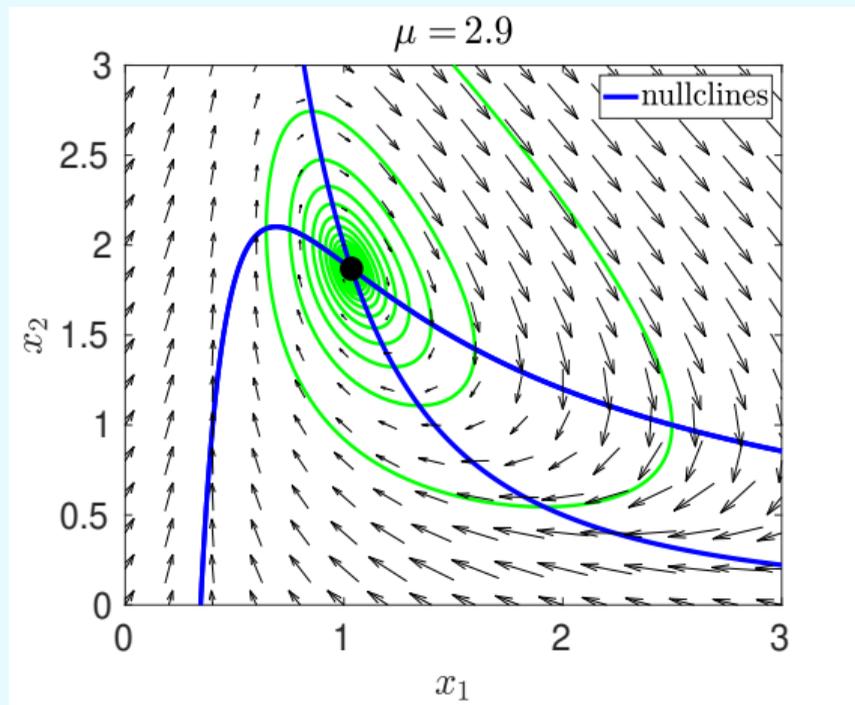


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There is no limit cycle for $\mu < 3$.



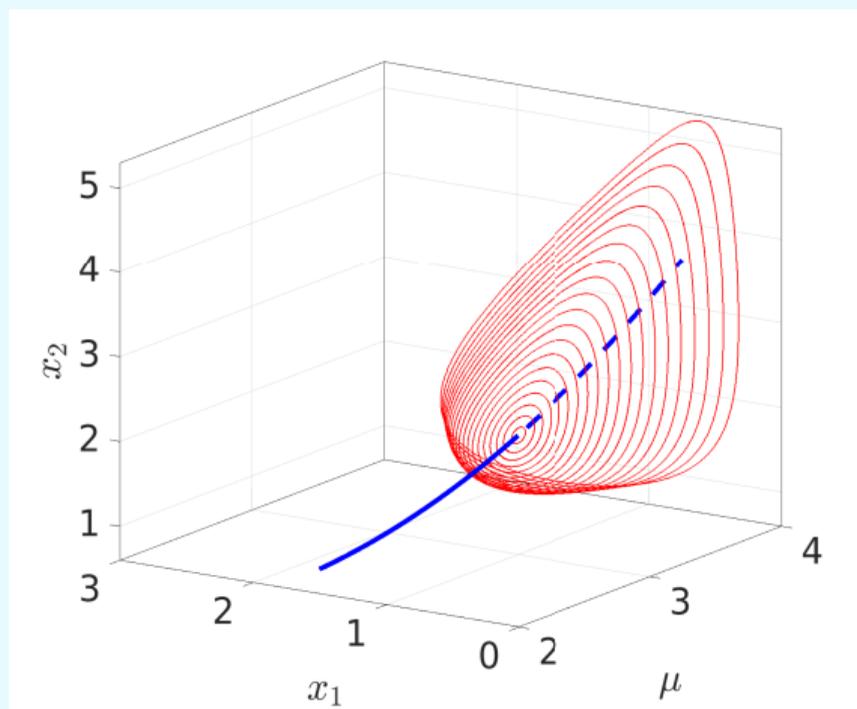
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bifurcation diagram

[show 3D animation]



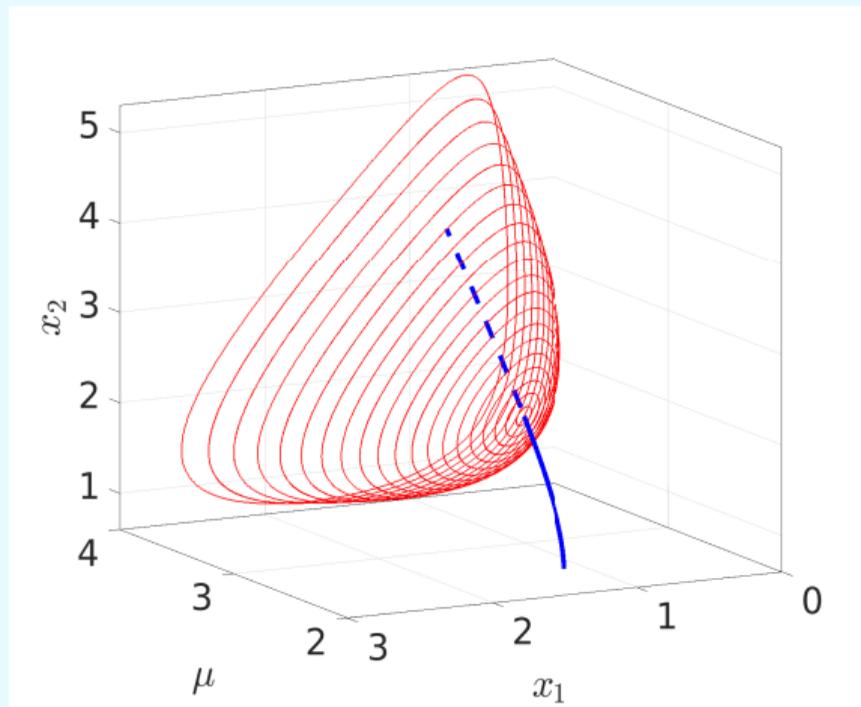
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bifurcation diagram

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Bifurcation analysis of chemical system in Question 6 on Problem Sheet 1

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fixed point at $\mathbf{x}_c = \left[\frac{3}{\mu}, \frac{2\mu^2}{9} \right]$

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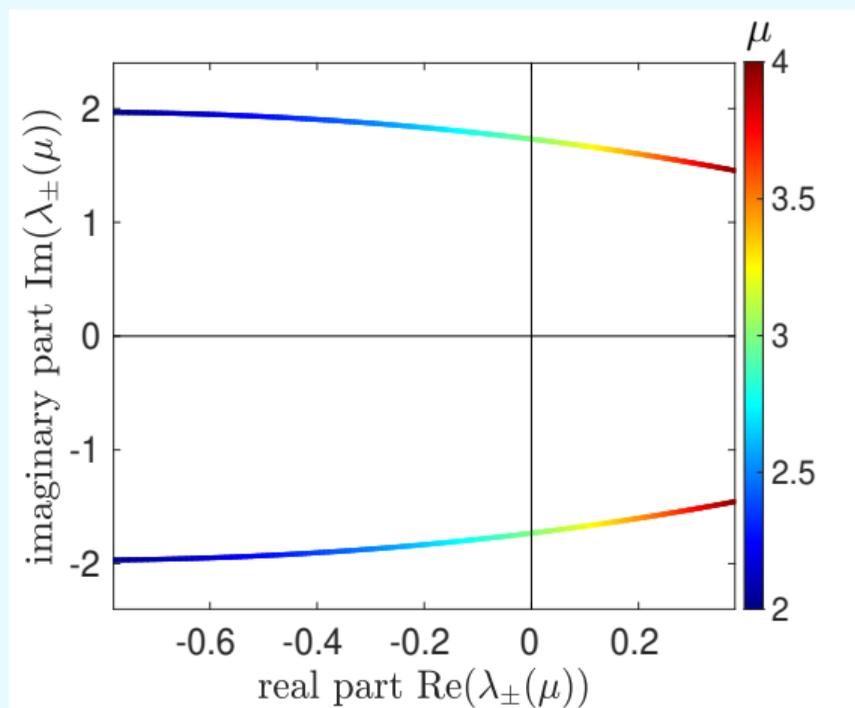
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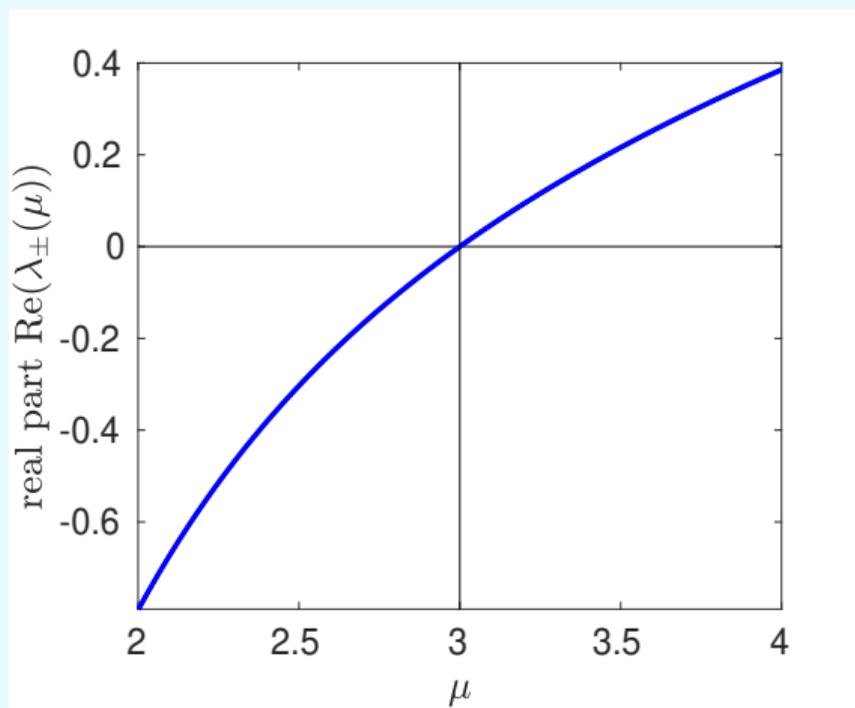
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using new variables $\bar{x}_1 = x_1 - \frac{3}{\mu}$, $\bar{x}_2 = x_2 - \frac{2\mu^2}{9}$, $\bar{\mu} = \frac{\mu - 3}{3}$, we obtain

$$\frac{d\bar{x}_1}{dt} = (1 + \bar{\mu})\bar{x}_1 + \frac{1}{(1 + \bar{\mu})^2}\bar{x}_2 + 2(1 + \bar{\mu})^2\bar{x}_1^2 + \frac{2}{1 + \bar{\mu}}\bar{x}_1\bar{x}_2 + \bar{x}_1^2\bar{x}_2$$

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$$\frac{d}{dt} \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} = M(\bar{\mu}) \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} + g(\bar{x}_1, \bar{x}_2; \bar{\mu}) \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

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at $\bar{\mu} = 0$, we have $M(0) = \begin{pmatrix} 1 & 1 \\ -4 & -1 \end{pmatrix}$

eigenvalues $\lambda_{\pm} = \pm i\sqrt{3}$, eigenvectors $\mathbf{v}_{\pm} = \begin{pmatrix} 1 \\ -4 \end{pmatrix} \pm i \begin{pmatrix} \sqrt{3} \\ 0 \end{pmatrix}$,

Bifurcation analysis of chemical system in Question 6 on Problem Sheet 1

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at $\bar{\mu} = 0$, we have $M(0) = \begin{pmatrix} 1 & 1 \\ -4 & -1 \end{pmatrix}$

eigenvalues $\lambda_{\pm} = \pm i\sqrt{3}$, eigenvectors $\mathbf{v}_{\pm} = \begin{pmatrix} 1 \\ -4 \end{pmatrix} \pm i \begin{pmatrix} \sqrt{3} \\ 0 \end{pmatrix}$, change of variables

$$\begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} = \begin{pmatrix} 1 & \sqrt{3} \\ -4 & 0 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} \text{ with inverse } \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \frac{1}{4\sqrt{3}} \begin{pmatrix} 0 & -\sqrt{3} \\ 4 & 1 \end{pmatrix} \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix}$$

Bifurcation analysis of chemical system in Question 6 on Problem Sheet 1

$$\frac{d}{dt} \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} = M(0) \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} + g(\bar{x}_1, \bar{x}_2; 0) \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

Bifurcation analysis of chemical system in Question 6 on Problem Sheet 1

$$\frac{d}{dt} \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} = M(0) \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} + g(\bar{x}_1, \bar{x}_2; 0) \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

change of variables: $\begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} = \begin{pmatrix} 1 & \sqrt{3} \\ -4 & 0 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix}$ and $\begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \frac{1}{4\sqrt{3}} \begin{pmatrix} 0 & -\sqrt{3} \\ 4 & 1 \end{pmatrix} \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix}$

Bifurcation analysis of chemical system in Question 6 on Problem Sheet 1

$$\frac{d}{dt} \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} = M(0) \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} + g(\bar{x}_1, \bar{x}_2; 0) \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

change of variables: $\begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} = \begin{pmatrix} 1 & \sqrt{3} \\ -4 & 0 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix}$ and $\begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \frac{1}{4\sqrt{3}} \begin{pmatrix} 0 & -\sqrt{3} \\ 4 & 1 \end{pmatrix} \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix}$

$$\frac{d}{dt} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \frac{1}{4\sqrt{3}} \begin{pmatrix} 0 & -\sqrt{3} \\ 4 & 1 \end{pmatrix} \frac{d}{dt} \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix}$$

Bifurcation analysis of chemical system in Question 6 on Problem Sheet 1

$$\frac{d}{dt} \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} = M(0) \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} + g(\bar{x}_1, \bar{x}_2; 0) \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

change of variables: $\begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} = \begin{pmatrix} 1 & \sqrt{3} \\ -4 & 0 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix}$ and $\begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \frac{1}{4\sqrt{3}} \begin{pmatrix} 0 & -\sqrt{3} \\ 4 & 1 \end{pmatrix} \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix}$

$$\frac{d}{dt} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \frac{1}{4\sqrt{3}} \begin{pmatrix} 0 & -\sqrt{3} \\ 4 & 1 \end{pmatrix} \frac{d}{dt} \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix}$$

$$\frac{d}{dt} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \frac{1}{4\sqrt{3}} \begin{pmatrix} 0 & -\sqrt{3} \\ 4 & 1 \end{pmatrix} M(0) \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} + \frac{1}{4\sqrt{3}} \begin{pmatrix} 0 & -\sqrt{3} \\ 4 & 1 \end{pmatrix} g(\bar{x}_1, \bar{x}_2; 0) \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

Bifurcation analysis of chemical system in Question 6 on Problem Sheet 1

$$\frac{d}{dt} \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} = M(0) \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} + g(\bar{x}_1, \bar{x}_2; 0) \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

change of variables: $\begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} = \begin{pmatrix} 1 & \sqrt{3} \\ -4 & 0 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix}$ and $\begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \frac{1}{4\sqrt{3}} \begin{pmatrix} 0 & -\sqrt{3} \\ 4 & 1 \end{pmatrix} \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix}$

$$\frac{d}{dt} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \frac{1}{4\sqrt{3}} \begin{pmatrix} 0 & -\sqrt{3} \\ 4 & 1 \end{pmatrix} \frac{d}{dt} \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix}$$

$$\frac{d}{dt} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \frac{1}{4\sqrt{3}} \begin{pmatrix} 0 & -\sqrt{3} \\ 4 & 1 \end{pmatrix} M(0) \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} + \frac{1}{4\sqrt{3}} \begin{pmatrix} 0 & -\sqrt{3} \\ 4 & 1 \end{pmatrix} g(\bar{x}_1, \bar{x}_2; 0) \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

$$\frac{d}{dt} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \frac{1}{4\sqrt{3}} \begin{pmatrix} 0 & -\sqrt{3} \\ 4 & 1 \end{pmatrix} M(0) \begin{pmatrix} 1 & \sqrt{3} \\ -4 & 0 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} + \frac{1}{4} \begin{pmatrix} 1 \\ \sqrt{3} \end{pmatrix} g(\bar{x}_1, \bar{x}_2; 0)$$

Bifurcation analysis of chemical system in Question 6 on Problem Sheet 1

$$\frac{d}{dt} \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} = M(0) \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} + g(\bar{x}_1, \bar{x}_2; 0) \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

change of variables: $\begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} = \begin{pmatrix} 1 & \sqrt{3} \\ -4 & 0 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix}$ and $\begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \frac{1}{4\sqrt{3}} \begin{pmatrix} 0 & -\sqrt{3} \\ 4 & 1 \end{pmatrix} \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix}$

$$\frac{d}{dt} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \frac{1}{4\sqrt{3}} \begin{pmatrix} 0 & -\sqrt{3} \\ 4 & 1 \end{pmatrix} \frac{d}{dt} \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix}$$

$$\frac{d}{dt} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \frac{1}{4\sqrt{3}} \begin{pmatrix} 0 & -\sqrt{3} \\ 4 & 1 \end{pmatrix} M(0) \begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} + \frac{1}{4\sqrt{3}} \begin{pmatrix} 0 & -\sqrt{3} \\ 4 & 1 \end{pmatrix} g(\bar{x}_1, \bar{x}_2; 0) \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

$$\frac{d}{dt} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \frac{1}{4\sqrt{3}} \begin{pmatrix} 0 & -\sqrt{3} \\ 4 & 1 \end{pmatrix} M(0) \begin{pmatrix} 1 & \sqrt{3} \\ -4 & 0 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} + \frac{1}{4} \begin{pmatrix} 1 \\ \sqrt{3} \end{pmatrix} g(\bar{x}_1, \bar{x}_2; 0)$$

$$\frac{d}{dt} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \begin{pmatrix} 0 & \sqrt{3} \\ -\sqrt{3} & 0 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} + \frac{1}{4} \begin{pmatrix} 1 \\ \sqrt{3} \end{pmatrix} g(\bar{x}_1, \bar{x}_2; 0)$$

where $g(\bar{x}_1, \bar{x}_2; 0) = 2\bar{x}_1^2 + 2\bar{x}_1\bar{x}_2 + \bar{x}_1^2\bar{x}_2$
 $= -6y_1^2 - 4y_1y_2\sqrt{3} + 6y_2^2 - 4y_1^3 - 8y_1^2y_2\sqrt{3} - 12y_2^2y_1$

Bifurcation analysis of chemical system in Question 6 on Problem Sheet 1

$$\frac{d}{dt} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \begin{pmatrix} 0 & \sqrt{3} \\ -\sqrt{3} & 0 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} + \frac{1}{2} \begin{pmatrix} 1 \\ \sqrt{3} \end{pmatrix} h(y_1, y_2)$$

$$\text{where } h(y_1, y_2) = -3y_1^2 - 2y_1y_2\sqrt{3} + 3y_2^2 - 2y_1^3 - 4y_1^2y_2\sqrt{3} - 6y_2^2y_1$$

Bifurcation analysis of chemical system in Question 6 on Problem Sheet 1

$$\frac{d}{dt} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \begin{pmatrix} 0 & \sqrt{3} \\ -\sqrt{3} & 0 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} + \frac{1}{2} \begin{pmatrix} 1 \\ \sqrt{3} \end{pmatrix} h(y_1, y_2)$$

where $h(y_1, y_2) = -3y_1^2 - 2y_1y_2\sqrt{3} + 3y_2^2 - 2y_1^3 - 4y_1^2y_2\sqrt{3} - 6y_2^2y_1$

$\bar{\mu}$ close to the bifurcation point $\bar{\mu} = 0$: matrix $M(\bar{\mu}) = \begin{pmatrix} 1 + \bar{\mu} & (1 + \bar{\mu})^{-2} \\ -4(1 + \bar{\mu}) & -(1 + \bar{\mu})^{-2} \end{pmatrix}$

has eigenvalues $\lambda_{\pm}(\bar{\mu}) = \alpha(\bar{\mu}) \pm i\omega(\bar{\mu})$ where

$$\alpha(\bar{\mu}) = \frac{1}{2} \left(1 + \bar{\mu} - \frac{1}{(1 + \bar{\mu})^2} \right) \text{ and } \omega(\bar{\mu}) = -\frac{1}{2} \sqrt{\frac{14}{1 + \bar{\mu}} - 1 - 2\bar{\mu} - \bar{\mu}^2 - \frac{1}{(1 + \bar{\mu})^4}}$$

Bifurcation analysis of chemical system in Question 6 on Problem Sheet 1

$$\frac{d}{dt} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \begin{pmatrix} 0 & \sqrt{3} \\ -\sqrt{3} & 0 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} + \frac{1}{2} \begin{pmatrix} 1 \\ \sqrt{3} \end{pmatrix} h(y_1, y_2)$$

where $h(y_1, y_2) = -3y_1^2 - 2y_1y_2\sqrt{3} + 3y_2^2 - 2y_1^3 - 4y_1^2y_2\sqrt{3} - 6y_2^2y_1$

$\bar{\mu}$ close to the bifurcation point $\bar{\mu} = 0$: matrix $M(\bar{\mu}) = \begin{pmatrix} 1 + \bar{\mu} & (1 + \bar{\mu})^{-2} \\ -4(1 + \bar{\mu}) & -(1 + \bar{\mu})^{-2} \end{pmatrix}$

has eigenvalues $\lambda_{\pm}(\bar{\mu}) = \alpha(\bar{\mu}) \pm i\omega(\bar{\mu})$ where

$$\alpha(\bar{\mu}) = \frac{1}{2} \left(1 + \bar{\mu} - \frac{1}{(1 + \bar{\mu})^2} \right) \text{ and } \omega(\bar{\mu}) = -\frac{1}{2} \sqrt{\frac{14}{1 + \bar{\mu}} - 1 - 2\bar{\mu} - \bar{\mu}^2 - \frac{1}{(1 + \bar{\mu})^4}}$$

which implies $\alpha(0) = 0$, $\omega(0) = -\sqrt{3}$, $\alpha'(0) = \frac{3}{2}$ and $\omega'(0) = \frac{\sqrt{3}}{2}$

Bifurcation analysis of chemical system in Question 6 on Problem Sheet 1

$$\frac{d}{dt} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \begin{pmatrix} 0 & \sqrt{3} \\ -\sqrt{3} & 0 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} + \frac{1}{2} \begin{pmatrix} 1 \\ \sqrt{3} \end{pmatrix} h(y_1, y_2)$$

where $h(y_1, y_2) = -3y_1^2 - 2y_1y_2\sqrt{3} + 3y_2^2 - 2y_1^3 - 4y_1^2y_2\sqrt{3} - 6y_2^2y_1$

$\bar{\mu}$ close to the bifurcation point $\bar{\mu} = 0$: matrix $M(\bar{\mu}) = \begin{pmatrix} 1 + \bar{\mu} & (1 + \bar{\mu})^{-2} \\ -4(1 + \bar{\mu}) & -(1 + \bar{\mu})^{-2} \end{pmatrix}$

has eigenvalues $\lambda_{\pm}(\bar{\mu}) = \alpha(\bar{\mu}) \pm i\omega(\bar{\mu})$ where

$$\alpha(\bar{\mu}) = \frac{1}{2} \left(1 + \bar{\mu} - \frac{1}{(1 + \bar{\mu})^2} \right) \text{ and } \omega(\bar{\mu}) = -\frac{1}{2} \sqrt{\frac{14}{1 + \bar{\mu}} - 1 - 2\bar{\mu} - \bar{\mu}^2 - \frac{1}{(1 + \bar{\mu})^4}}$$

which implies $\alpha(0) = 0$, $\omega(0) = -\sqrt{3}$, $\alpha'(0) = \frac{3}{2}$ and $\omega'(0) = \frac{\sqrt{3}}{2}$

normal form in polar coordinates:

$$\frac{dr}{dt} = \alpha'(0)\bar{\mu}r + a(0)r^3 + \mathcal{O}(\bar{\mu}^2r, \bar{\mu}r^3, r^5)$$

$$\frac{d\theta}{dt} = \omega(0) + \omega'(0)\bar{\mu} + b(0)r^2 + \mathcal{O}(\bar{\mu}^2, \bar{\mu}r^2, r^4)$$

Bifurcation analysis of chemical system in Question 6 on Problem Sheet 1

$$\frac{d}{dt} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \begin{pmatrix} 0 & \sqrt{3} \\ -\sqrt{3} & 0 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} + \frac{1}{2} \begin{pmatrix} 1 \\ \sqrt{3} \end{pmatrix} h(y_1, y_2)$$

where $h(y_1, y_2) = -3y_1^2 - 2y_1y_2\sqrt{3} + 3y_2^2 - 2y_1^3 - 4y_1^2y_2\sqrt{3} - 6y_2^2y_1$

$\bar{\mu}$ close to the bifurcation point $\bar{\mu} = 0$: matrix $M(\bar{\mu}) = \begin{pmatrix} 1 + \bar{\mu} & (1 + \bar{\mu})^{-2} \\ -4(1 + \bar{\mu}) & -(1 + \bar{\mu})^{-2} \end{pmatrix}$

has eigenvalues $\lambda_{\pm}(\bar{\mu}) = \alpha(\bar{\mu}) \pm i\omega(\bar{\mu})$ where

$$\alpha(\bar{\mu}) = \frac{1}{2} \left(1 + \bar{\mu} - \frac{1}{(1 + \bar{\mu})^2} \right) \text{ and } \omega(\bar{\mu}) = -\frac{1}{2} \sqrt{\frac{14}{1 + \bar{\mu}} - 1 - 2\bar{\mu} - \bar{\mu}^2 - \frac{1}{(1 + \bar{\mu})^4}}$$

which implies $\alpha(0) = 0$, $\omega(0) = -\sqrt{3}$, $\alpha'(0) = \frac{3}{2}$ and $\omega'(0) = \frac{\sqrt{3}}{2}$

normal form in polar coordinates:

$$\frac{dr}{dt} = \frac{3}{2} \bar{\mu} r + a(0) r^3 + \mathcal{O}(\bar{\mu}^2 r, \bar{\mu} r^3, r^5)$$

$$\frac{d\theta}{dt} = -\sqrt{3} + \frac{\sqrt{3}}{2} \bar{\mu} + b(0) r^2 + \mathcal{O}(\bar{\mu}^2, \bar{\mu} r^2, r^4)$$

Calculation of $a(0)$

supercritical Hopf bifurcation: $a(0) < 0$ (periodic orbit is asymptotically stable)

subcritical Hopf bifurcation: $a(0) > 0$ (periodic orbit is unstable)

Calculation of $a(0)$

supercritical Hopf bifurcation: $a(0) < 0$ (periodic orbit is asymptotically stable)

subcritical Hopf bifurcation: $a(0) > 0$ (periodic orbit is unstable)

Lemma: Assume that the ODE system with Hopf bifurcation at $\bar{\mu} = 0$ was transformed to

$$\frac{d}{dt} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \begin{pmatrix} 0 & -\omega(0) \\ \omega(0) & 0 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} + \begin{pmatrix} h_1(y_1, y_2) \\ h_2(y_1, y_2) \end{pmatrix}$$

where $h_1(y_1, y_2)$ and $h_2(y_1, y_2)$ contain only higher-order nonlinear terms that vanish at the origin. Then

$$a(0) = \frac{1}{16} \left(\frac{\partial^3 h_1}{\partial y_1^3} + \frac{\partial^3 h_1}{\partial y_1 \partial y_2^2} + \frac{\partial^3 h_2}{\partial y_1^2 \partial y_2} + \frac{\partial^3 h_2}{\partial y_2^3} \right) + \frac{1}{16 \omega(0)} \left[\frac{\partial^2 h_1}{\partial y_1 \partial y_2} \left(\frac{\partial^2 h_1}{\partial y_1^2} + \frac{\partial^2 h_1}{\partial y_2^2} \right) - \frac{\partial^2 h_2}{\partial y_1 \partial y_2} \left(\frac{\partial^2 h_2}{\partial y_1^2} + \frac{\partial^2 h_2}{\partial y_2^2} \right) - \frac{\partial^2 h_1}{\partial y_1^2} \frac{\partial^2 h_2}{\partial y_1^2} + \frac{\partial^2 h_1}{\partial y_2^2} \frac{\partial^2 h_2}{\partial y_2^2} \right]$$

where the partial derivatives are evaluated at the origin $\mathbf{0}$.

Bifurcation analysis of chemical system in Question 6 on Problem Sheet 1

Our equation
$$\frac{d}{dt} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \begin{pmatrix} 0 & \sqrt{3} \\ -\sqrt{3} & 0 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} + \frac{1}{2} \begin{pmatrix} 1 \\ \sqrt{3} \end{pmatrix} h(y_1, y_2)$$

where $h(y_1, y_2) = -3y_1^2 - 2y_1y_2\sqrt{3} + 3y_2^2 - 2y_1^3 - 4y_1^2y_2\sqrt{3} - 6y_1y_2^2$

is in the form
$$\frac{d}{dt} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \begin{pmatrix} 0 & -\omega(0) \\ \omega(0) & 0 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} + \begin{pmatrix} h_1(y_1, y_2) \\ h_2(y_1, y_2) \end{pmatrix}$$

where $\omega_0 = -\sqrt{3}$, $h_1(y_1, y_2) = h(y_1, y_2)/2$ and $h_2(y_1, y_2) = \sqrt{3}h(y_1, y_2)/2$.

Bifurcation analysis of chemical system in Question 6 on Problem Sheet 1

Our equation
$$\frac{d}{dt} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \begin{pmatrix} 0 & \sqrt{3} \\ -\sqrt{3} & 0 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} + \frac{1}{2} \begin{pmatrix} 1 \\ \sqrt{3} \end{pmatrix} h(y_1, y_2)$$

where $h(y_1, y_2) = -3y_1^2 - 2y_1y_2\sqrt{3} + 3y_2^2 - 2y_1^3 - 4y_1^2y_2\sqrt{3} - 6y_1y_2^2$

is in the form
$$\frac{d}{dt} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \begin{pmatrix} 0 & -\omega(0) \\ \omega(0) & 0 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} + \begin{pmatrix} h_1(y_1, y_2) \\ h_2(y_1, y_2) \end{pmatrix}$$

where $\omega_0 = -\sqrt{3}$, $h_1(y_1, y_2) = h(y_1, y_2)/2$ and $h_2(y_1, y_2) = \sqrt{3}h(y_1, y_2)/2$.

Substituting (partial derivatives evaluated at the origin $\mathbf{0}$):

$$\frac{\partial^3 h_1}{\partial y_1^3} = -6, \quad \frac{\partial^3 h_1}{\partial y_1 \partial y_2^2} = -6, \quad \frac{\partial^3 h_2}{\partial y_1^2 \partial y_2} = -12, \quad \frac{\partial^3 h_2}{\partial y_2^3} = 0, \quad \frac{\partial^2 h_1}{\partial y_1^2} = -3,$$

$$\frac{\partial^2 h_1}{\partial y_1 \partial y_2} = -\sqrt{3}, \quad \frac{\partial^2 h_1}{\partial y_2^2} = 3, \quad \frac{\partial^2 h_2}{\partial y_1^2} = -3\sqrt{3}, \quad \frac{\partial^2 h_2}{\partial y_1 \partial y_2} = -3, \quad \frac{\partial^2 h_2}{\partial y_2^2} = 3\sqrt{3}$$

we get $a(0) = -\frac{3}{2} \implies$ **supercritical Hopf bifurcation**

Bifurcation analysis of chemical system in Question 6 on Problem Sheet 1

normal form:

$$\frac{dr}{dt} = \frac{3}{2} \bar{\mu} r - \frac{3}{2} r^3 + \dots$$

$$\frac{d\theta}{dt} = -\sqrt{3} + \frac{\sqrt{3}}{2} \bar{\mu} + \dots$$

Origin $\mathbf{0}$ is stable for $\bar{\mu} < 0 \Leftrightarrow \mu < 3$

and unstable for $\bar{\mu} > 0 \Leftrightarrow \mu > 3$

Bifurcation analysis of chemical system in Question 6 on Problem Sheet 1

normal form:

$$\frac{dr}{dt} = \frac{3}{2} \bar{\mu} r - \frac{3}{2} r^3 + \dots$$

$$\frac{d\theta}{dt} = -\sqrt{3} + \frac{\sqrt{3}}{2} \bar{\mu} + \dots$$

Origin $\mathbf{0}$ is stable for $\bar{\mu} < 0 \Leftrightarrow \mu < 3$

and unstable for $\bar{\mu} > 0 \Leftrightarrow \mu > 3$

A stable limit cycle is born with

amplitude $\sqrt{\frac{\mu - 3}{3}}$ and period $\frac{2\pi}{\sqrt{3}}$

Bifurcation analysis of chemical system in Question 6 on Problem Sheet 1

normal form:

$$\frac{dr}{dt} = \frac{3}{2} \bar{\mu} r - \frac{3}{2} r^3 + \dots$$

$$\frac{d\theta}{dt} = -\sqrt{3} + \frac{\sqrt{3}}{2} \bar{\mu} + \dots$$

Origin $\mathbf{0}$ is stable for $\bar{\mu} < 0 \Leftrightarrow \mu < 3$

and unstable for $\bar{\mu} > 0 \Leftrightarrow \mu > 3$

A stable limit cycle is born with

amplitude $\sqrt{\frac{\mu - 3}{3}}$ and period $\frac{2\pi}{\sqrt{3}}$

The limit cycle is $y_1^2 + y_2^2 = \frac{\mu - 3}{3}$ which corresponds to an ellipse in x_1 and x_2 :

$$\bar{x}_2^2 + \frac{1}{3} (4\bar{x}_1 + \bar{x}_2)^2 = \frac{16(\mu - 3)}{3}$$

Bifurcation analysis of chemical system in Question 6 on Problem Sheet 1

normal form:

$$\frac{dr}{dt} = \frac{3}{2} \bar{\mu} r - \frac{3}{2} r^3 + \dots$$

$$\frac{d\theta}{dt} = -\sqrt{3} + \frac{\sqrt{3}}{2} \bar{\mu} + \dots$$

Origin $\mathbf{0}$ is stable for $\bar{\mu} < 0 \Leftrightarrow \mu < 3$

and unstable for $\bar{\mu} > 0 \Leftrightarrow \mu > 3$

A stable limit cycle is born with

amplitude $\sqrt{\frac{\mu - 3}{3}}$ and period $\frac{2\pi}{\sqrt{3}}$

The limit cycle is $y_1^2 + y_2^2 = \frac{\mu - 3}{3}$ which corresponds to an ellipse in x_1 and x_2 :

$$\left(x_2 - \frac{2\mu^2}{9}\right)^2 + \frac{1}{3} \left(4x_1 + x_2 - \frac{12}{\mu} - \frac{2\mu^2}{9}\right)^2 = \frac{16(\mu - 3)}{3}$$

Bifurcation analysis of chemical system in Question 6 on Problem Sheet 1

normal form:

$$\frac{dr}{dt} = \frac{3}{2} \bar{\mu} r - \frac{3}{2} r^3 + \dots$$

$$\frac{d\theta}{dt} = -\sqrt{3} + \frac{\sqrt{3}}{2} \bar{\mu} + \dots$$

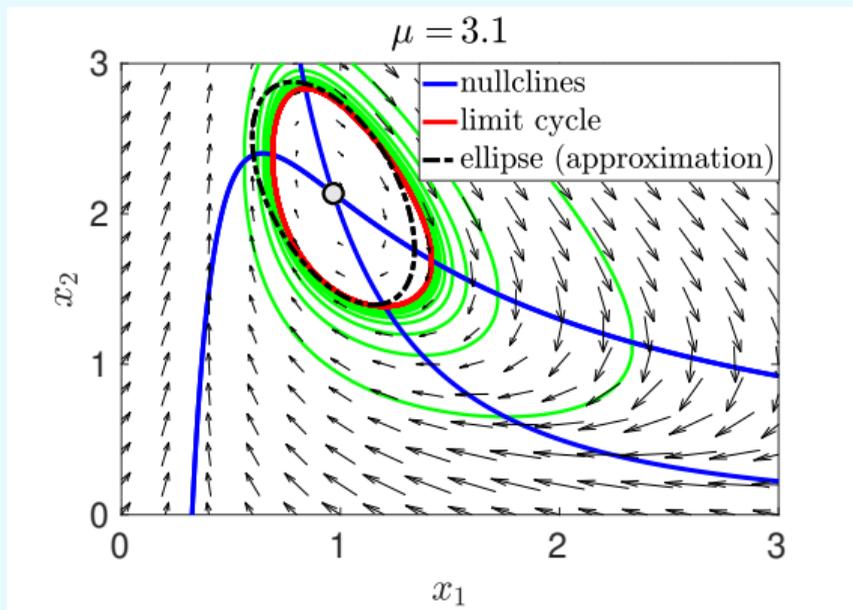
Origin $\mathbf{0}$ is stable for $\bar{\mu} < 0 \Leftrightarrow \mu < 3$
and unstable for $\bar{\mu} > 0 \Leftrightarrow \mu > 3$

A stable limit cycle is born with

amplitude $\sqrt{\frac{\mu - 3}{3}}$ and period $\frac{2\pi}{\sqrt{3}}$

The limit cycle is $y_1^2 + y_2^2 = \frac{\mu - 3}{3}$ which corresponds to an ellipse in x_1 and x_2 :

$$\left(x_2 - \frac{2\mu^2}{9}\right)^2 + \frac{1}{3} \left(4x_1 + x_2 - \frac{12}{\mu} - \frac{2\mu^2}{9}\right)^2 = \frac{16(\mu - 3)}{3}$$



Bifurcation analysis of chemical system in Question 6 on Problem Sheet 1

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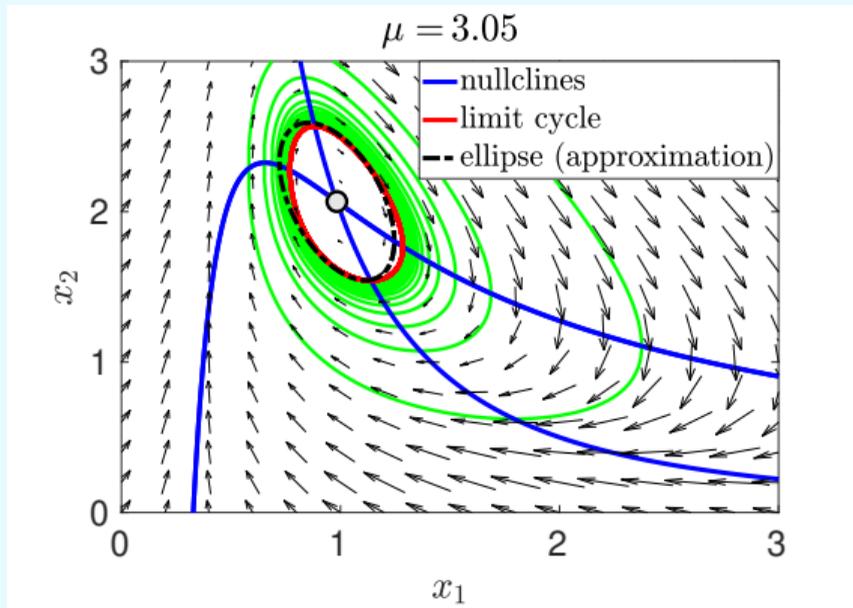
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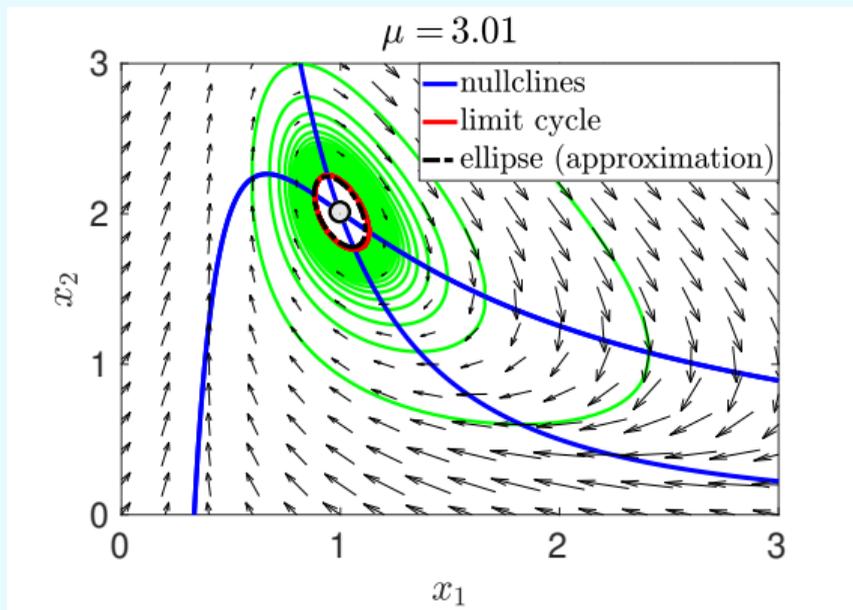
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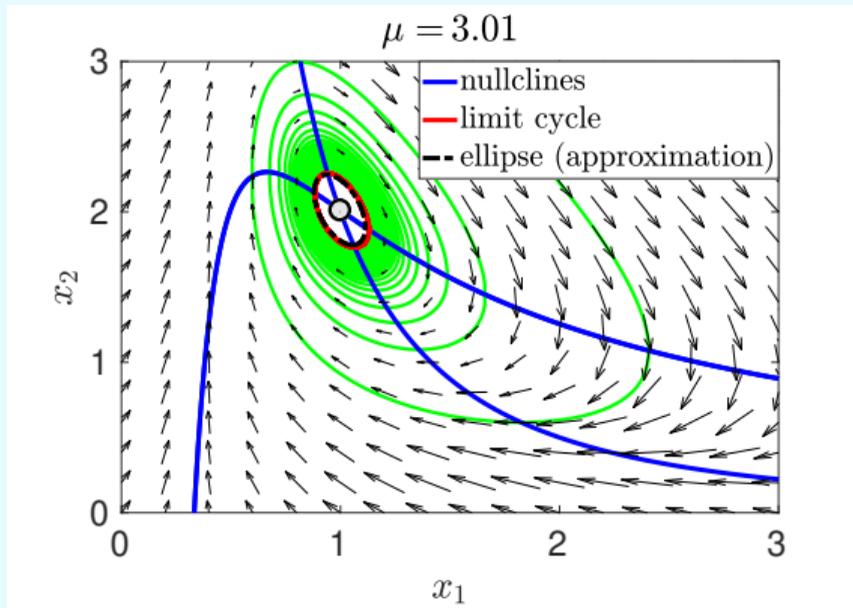
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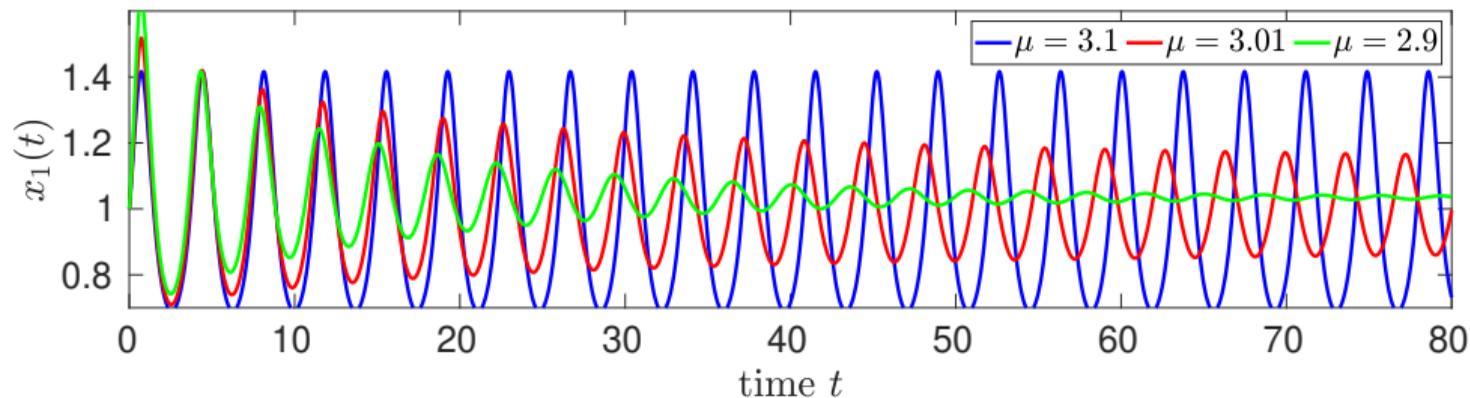
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Additional examples: [Question 2 on Problem Sheet 3](#) and

[Questions 1, 4, 5 and 6](#) (formulated in a way that the questions do not specify what bifurcations of limit cycles are there)



Bifurcation analysis of chemical system in Question 6 on Problem Sheet 1



A stable limit cycle is born with amplitude $\sqrt{\frac{\mu - 3}{3}}$ and period $\frac{2\pi}{\sqrt{3}} \approx 3.6$

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Lecture 11: we will consider global bifurcations when the amplitude will satisfy $\mathcal{O}(1)$, i.e. the amplitude of the limit cycle does not go to zero as the parameter μ approaches the bifurcation value $\mu = \mu_c$

Notes on calculation of $a(0)$

supercritical Hopf bifurcation: $a(0) < 0$ (periodic orbit is asymptotically stable)

subcritical Hopf bifurcation: $a(0) > 0$ (periodic orbit is unstable)

Lemma: Assume that the ODE system with Hopf bifurcation at $\bar{\mu} = 0$ was transformed to

$$\frac{d}{dt} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \begin{pmatrix} 0 & -\omega(0) \\ \omega(0) & 0 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} + \begin{pmatrix} h_1(y_1, y_2) \\ h_2(y_1, y_2) \end{pmatrix}$$

where $h_1(y_1, y_2)$ and $h_2(y_1, y_2)$ contain only higher-order nonlinear terms that vanish at the origin. Then

$$a(0) = \frac{1}{16} \left(\frac{\partial^3 h_1}{\partial y_1^3} + \frac{\partial^3 h_1}{\partial y_1 \partial y_2^2} + \frac{\partial^3 h_2}{\partial y_1^2 \partial y_2} + \frac{\partial^3 h_2}{\partial y_2^3} \right) + \frac{1}{16 \omega(0)} \left[\frac{\partial^2 h_1}{\partial y_1 \partial y_2} \left(\frac{\partial^2 h_1}{\partial y_1^2} + \frac{\partial^2 h_1}{\partial y_2^2} \right) - \frac{\partial^2 h_2}{\partial y_1 \partial y_2} \left(\frac{\partial^2 h_2}{\partial y_1^2} + \frac{\partial^2 h_2}{\partial y_2^2} \right) - \frac{\partial^2 h_1}{\partial y_1^2} \frac{\partial^2 h_2}{\partial y_1^2} + \frac{\partial^2 h_1}{\partial y_2^2} \frac{\partial^2 h_2}{\partial y_2^2} \right]$$

where the partial derivatives are evaluated at the origin $\mathbf{0}$.

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Questions: How do we prove our Lemma?

Use complex notation introduced in Lecture 9: $z = x_1 + ix_2$, $z^* = x_1 - ix_2$

If there are quadratic nonlinearities, remove all of them by applying the near identity change of coordinates $w = z + Az^2 + Bzz^* + Cz^2$ for a suitable choice of A , B and C .

Continue as in our example in Lecture 9, where we used the near identity change of coordinates $w = z + Az^3 + Bz^2z^* + Cz(z^*)^2 + D(z^*)^3$ to remove 3 out of 4 cubic terms.

Formula for $a(0)$ is given as the real part of the coefficient in front of $z^2z^* = z|z|^2$.

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What happens if $a(0) = 0$?

Many different possibilities.

For example, there could be no limit cycles or more than one limit cycle.

Examples with $a(0) = 0$

example with no limit cycles: $\frac{dr}{dt} = \mu r + \mu r^3$

$$\frac{d\theta}{dt} = 1$$

we have $\lambda(\mu) = \mu \pm i$ and $a(\mu) = \mu$, i.e. $a(0) = 0$

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example with two bifurcating limit cycles: $\frac{dr}{dt} = \mu^3 r - (\mu + \mu^2) r^3 + r^5$

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$\mathbf{0} = [0, 0]$ is a **stable** spiral for $\mu < 0$ and an **unstable** spiral for $\mu > 0$

$0 < \mu < 1$: there are **two limit cycles**: a stable circular limit cycle of radius $r = \mu$
and an unstable circular limit cycle of radius $r = \sqrt{\mu}$

B5.6: Nonlinear Dynamics, Bifurcations and Chaos (Lecture 11)

- summary of Lecture 10: we discussed
Bifurcations of limit cycles, covering supercritical and subcritical Hopf bifurcations.
Oscillations in chemical reaction networks.
- today: we will continue in our discussion of Problem Sheet 3

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- course synopsis of **Lectures 9-16**:
Bifurcations of limit cycles, covering supercritical and subcritical Hopf bifurcations, saddle-node bifurcation of cycles, infinite-period (SNIC) bifurcation and homoclinic (saddle-loop) bifurcation. Oscillations in chemical reaction networks. Weakly nonlinear oscillators. Poincaré-Lindstedt method. Conservative and non-conservative systems. Liénard systems, van der Pol oscillator. Hilbert's 16th problem. Lorenz equations. Lorenz map. Poincaré section. Poincaré map. Converse of Sharkovsky's theorem. Bernoulli shift map, symbolic dynamics. Tent map. Dynamics on metric spaces, sensitive dependence on initial conditions, transitivity, conjugate maps, chaotic dynamics.

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bifurcation at $\mu = \mu_c$	amplitude	period
supercritical Hopf bifurcation	$\mathcal{O}\left(\sqrt{ \mu - \mu_c }\right)$	$\mathcal{O}(1)$
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supercritical and subcritical Hopf bifurcations: we discussed them last week (including the analysis of the supercritical Hopf bifurcation in the chemical system in Question 6 on Problem Sheet 1)

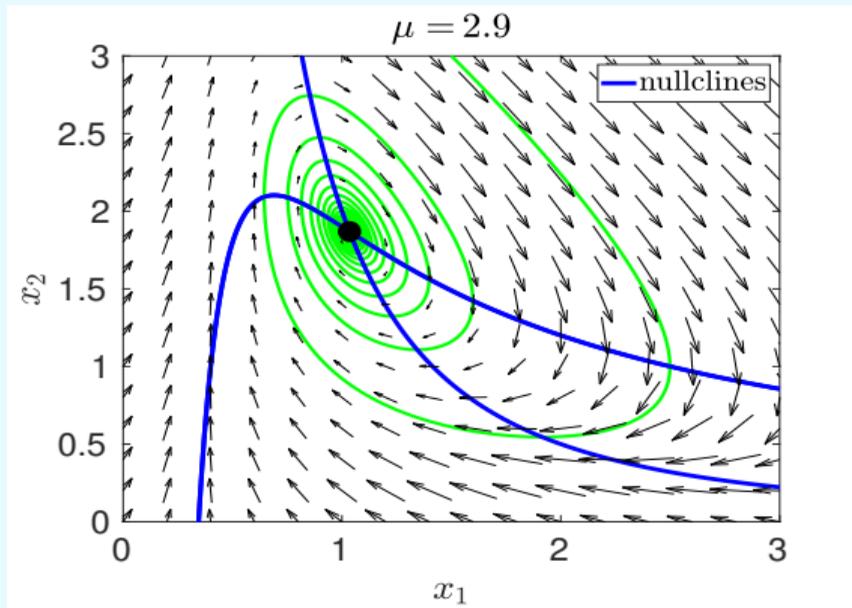
Lecture 10: chemical system in Question 6 on Problem Sheet 1

normal form (where $\bar{\mu} = (\mu - 3)/3$):

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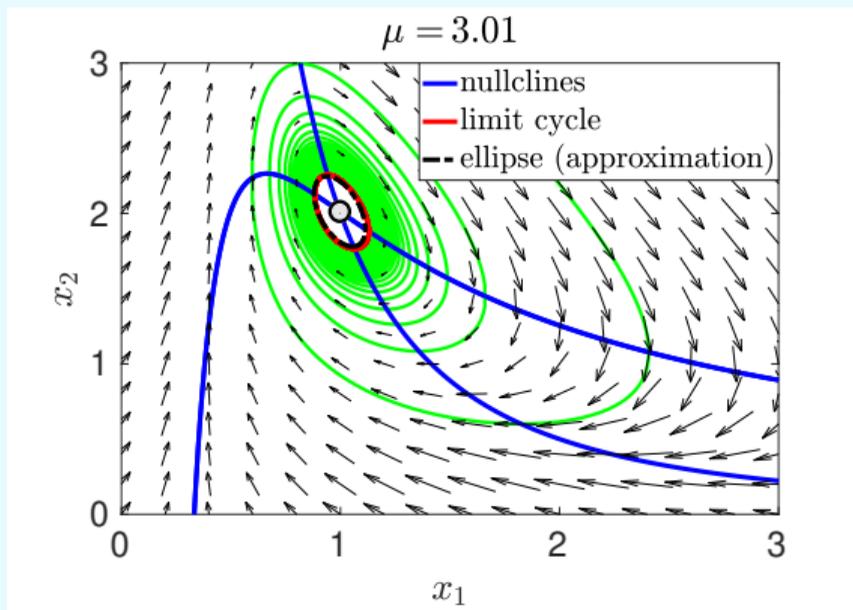
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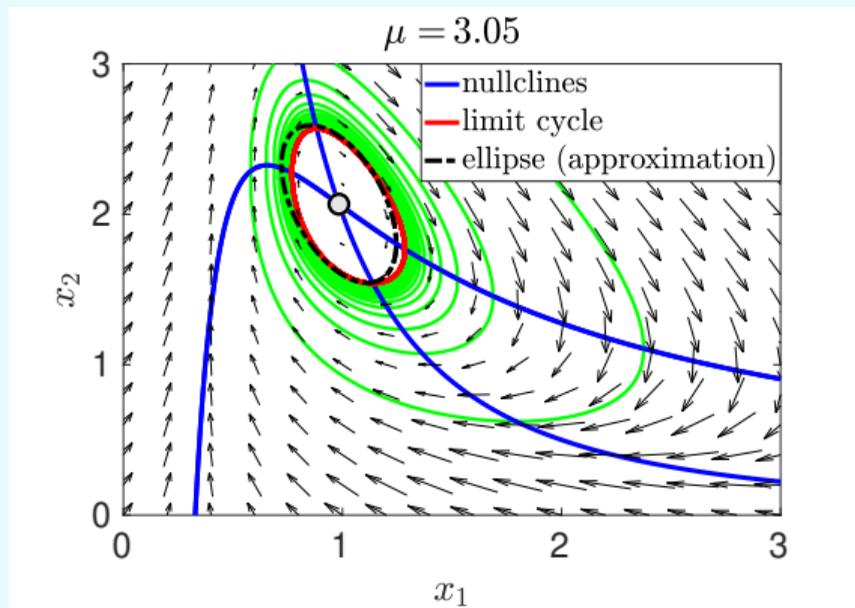
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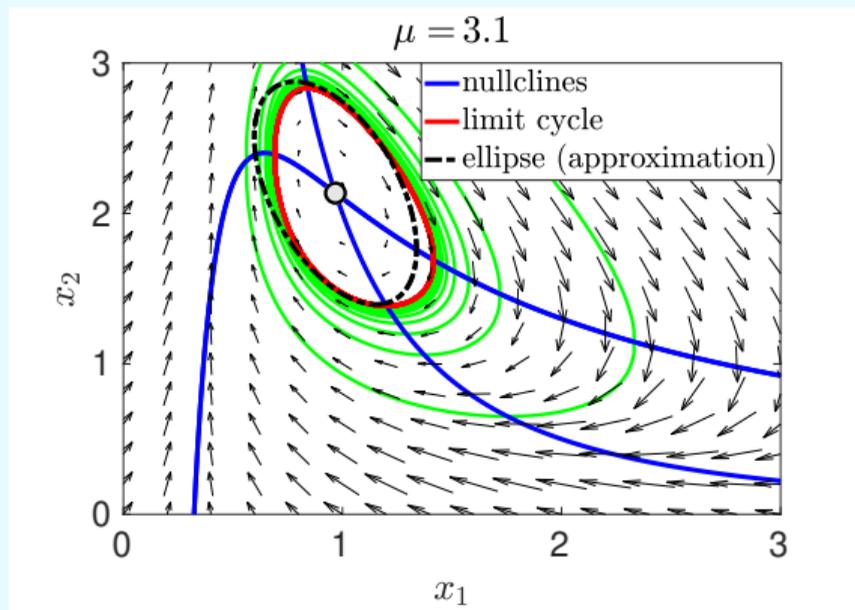
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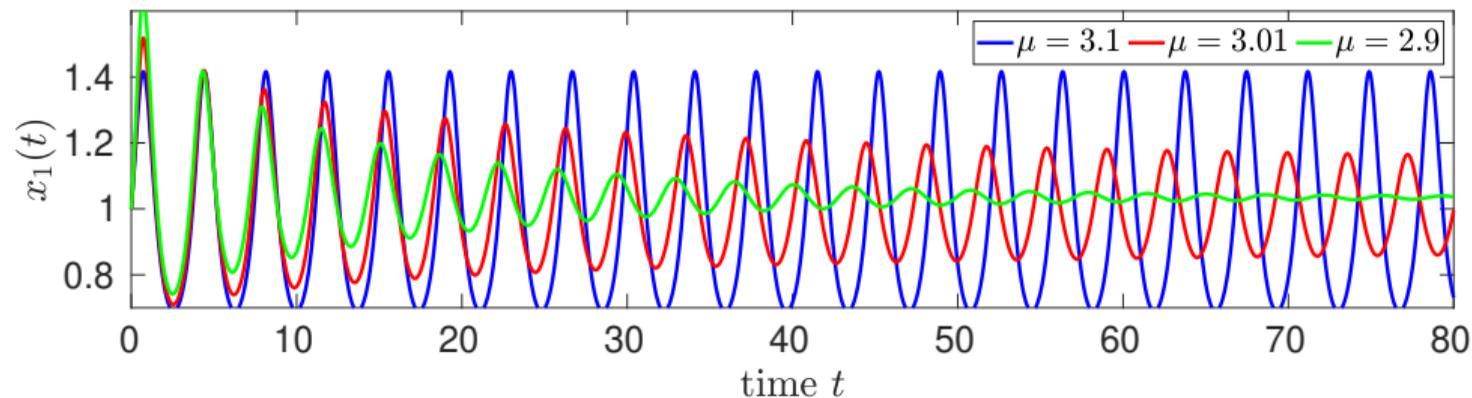
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Lecture 10: chemical system in Question 6 on Problem Sheet 1



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saddle-node bifurcation of cycles: we have already presented an example when we discussed the subcritical Hopf bifurcation

Example: saddle-node bifurcation of cycles

saddle-node bifurcation of cycles at $\mu = -1/4$: a half-stable cycle appears, it splits into a pair of limit cycles for $\mu > -1/4$, one stable, one unstable, or, viewed in the other direction, a stable and unstable cycle collide and disappear as μ decreases through $\mu = -1/4$

$$\frac{dr}{dt} = \mu r + r^3 - r^5$$

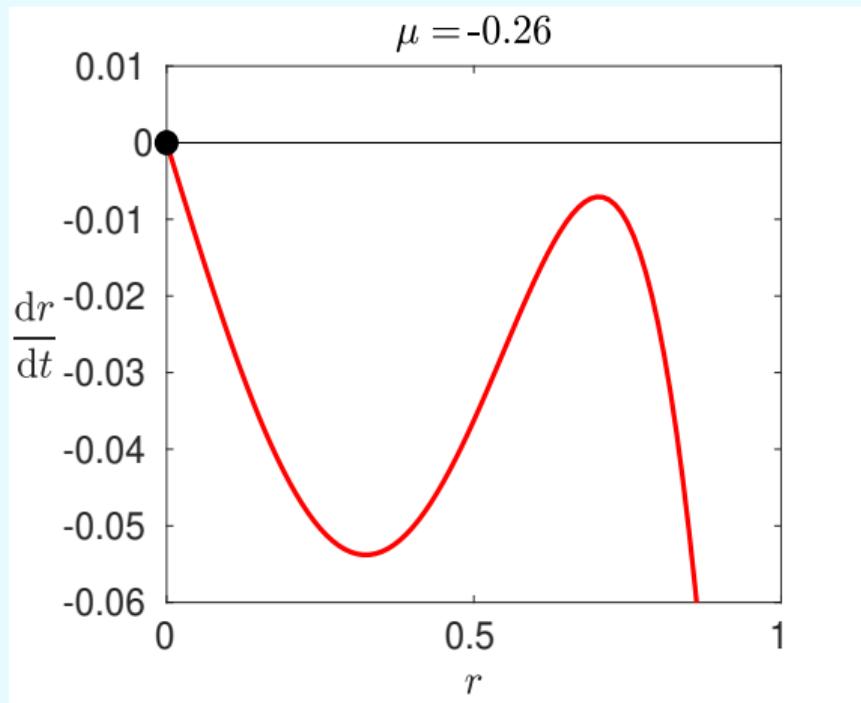
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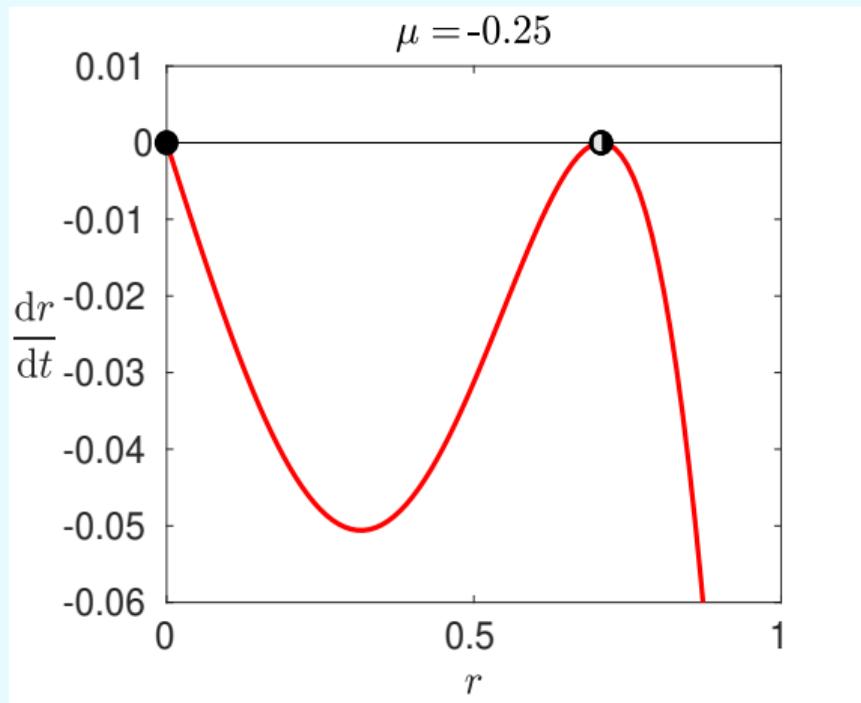


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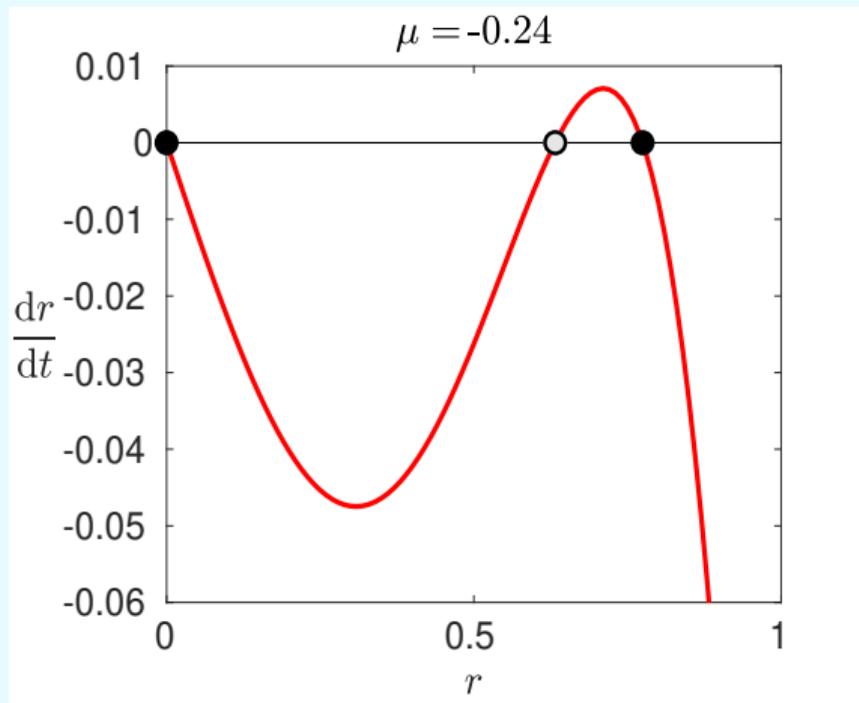


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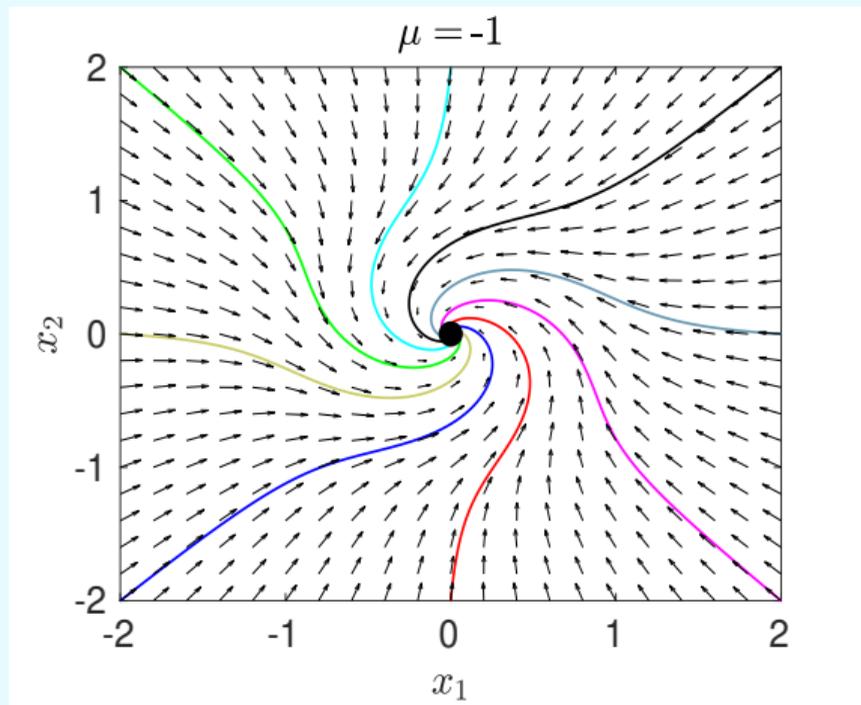
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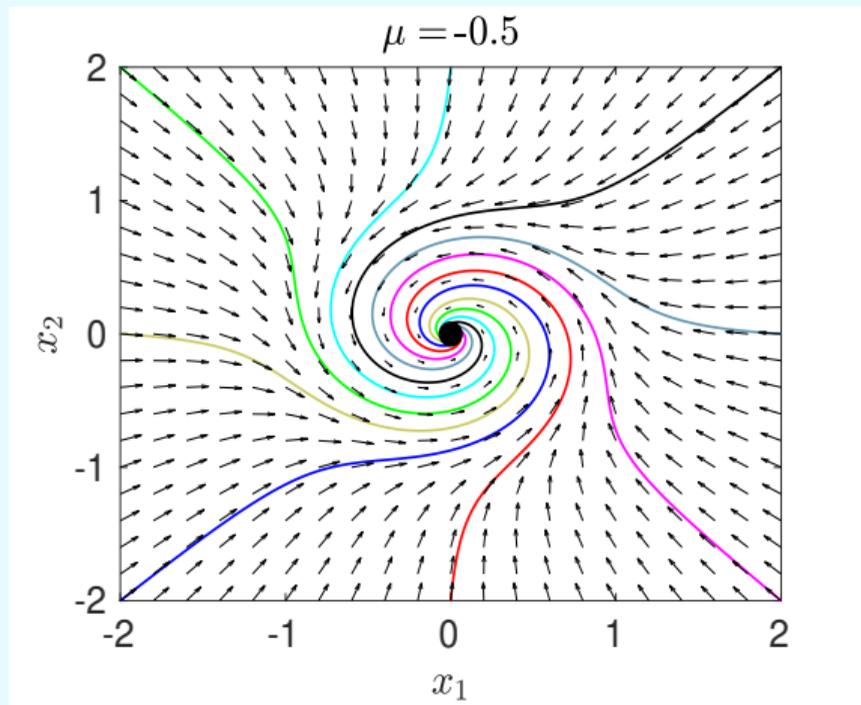
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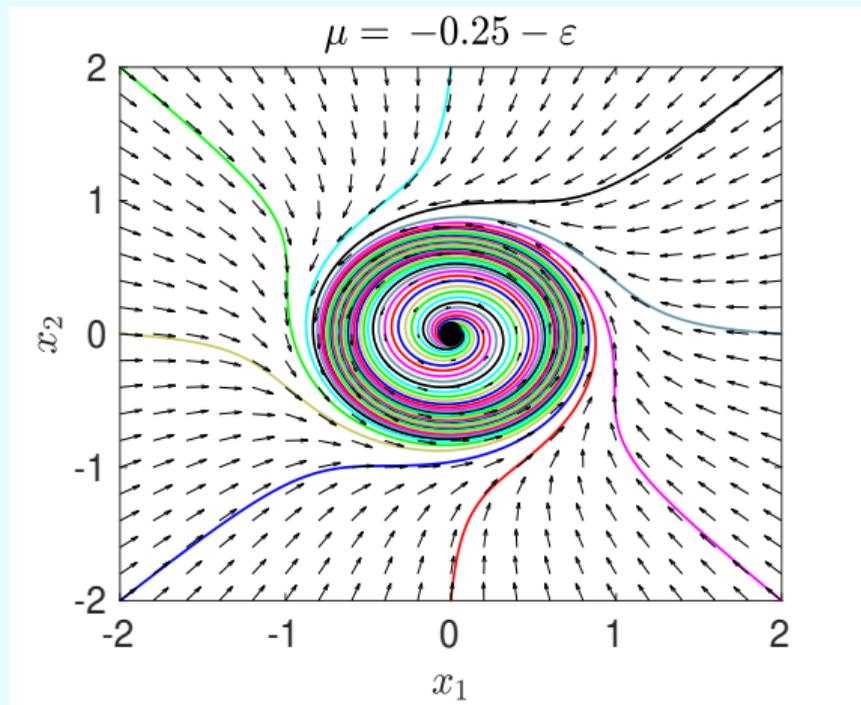
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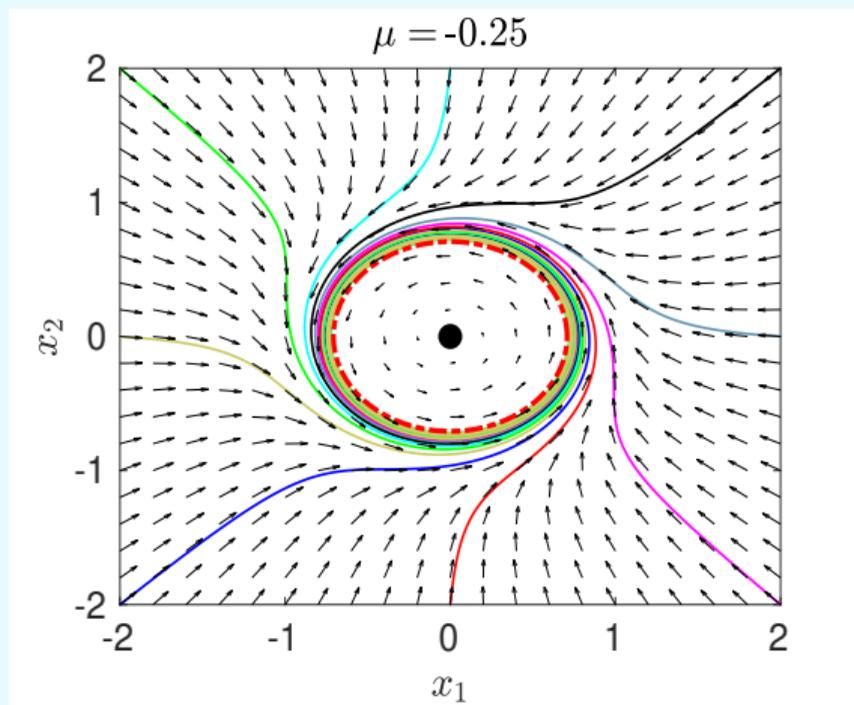
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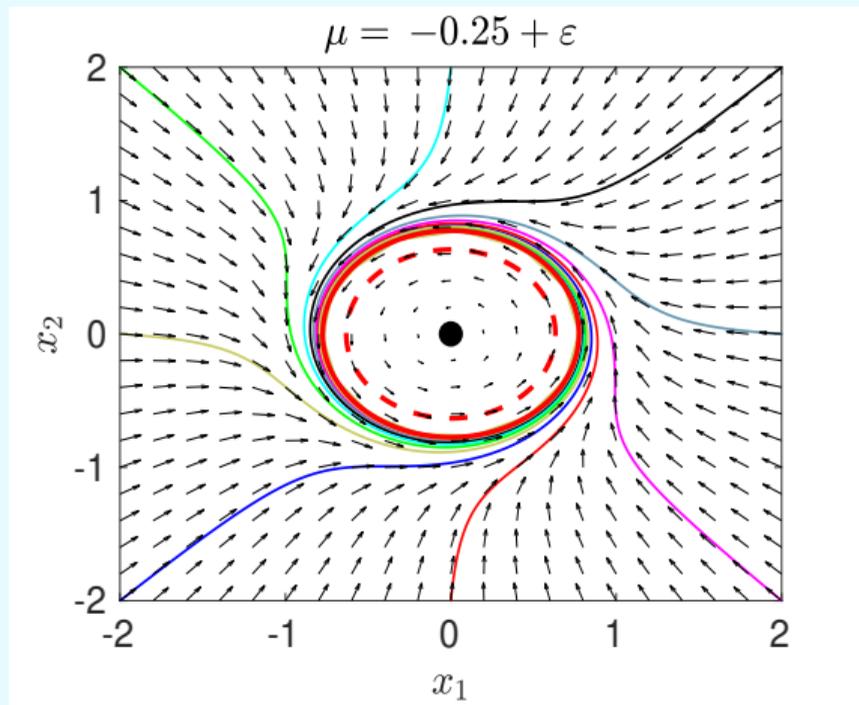
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Example: saddle-node bifurcation of cycles and subcritical Hopf bifurcation

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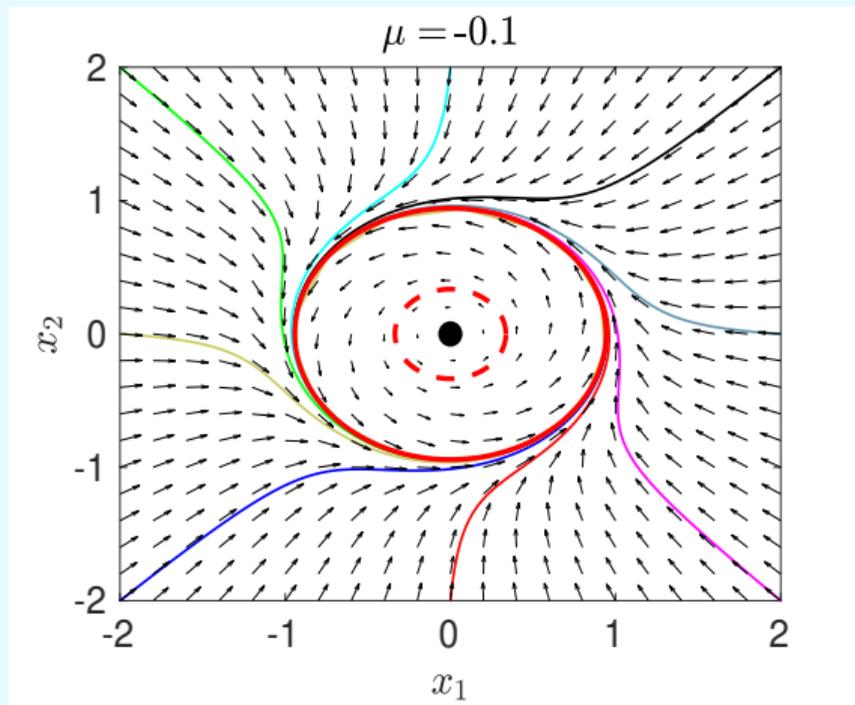
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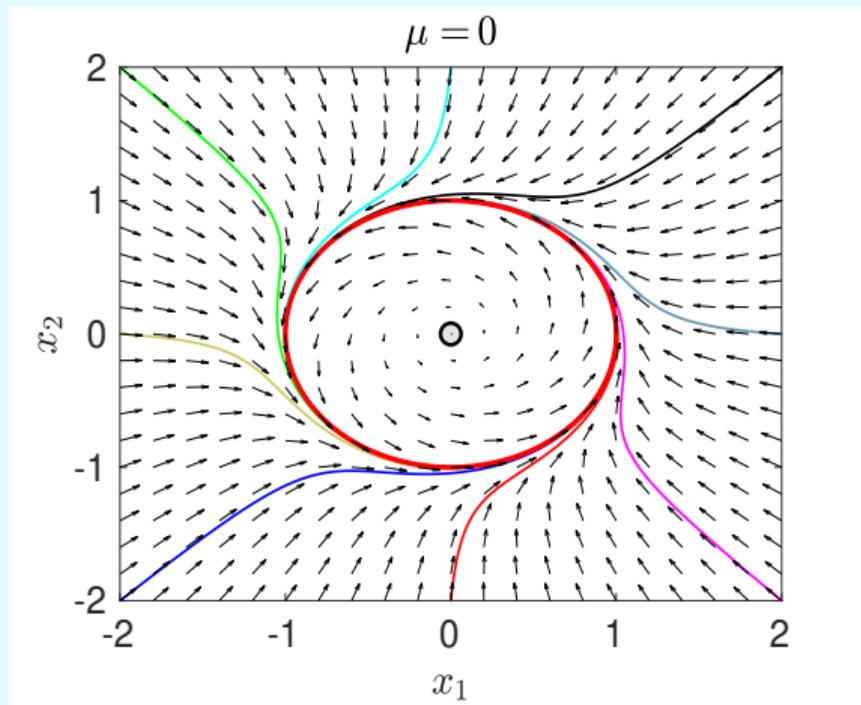
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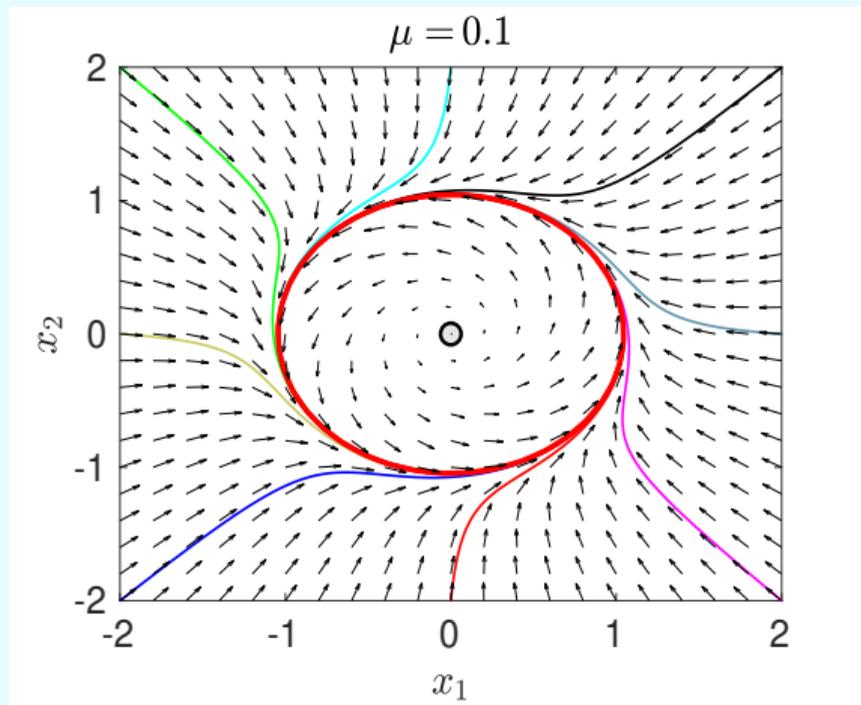
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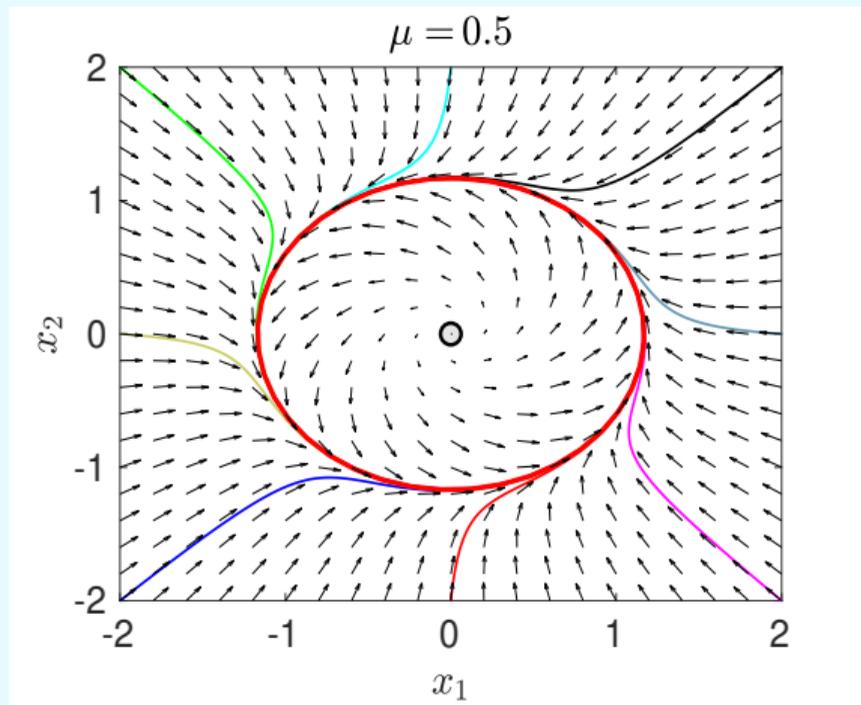
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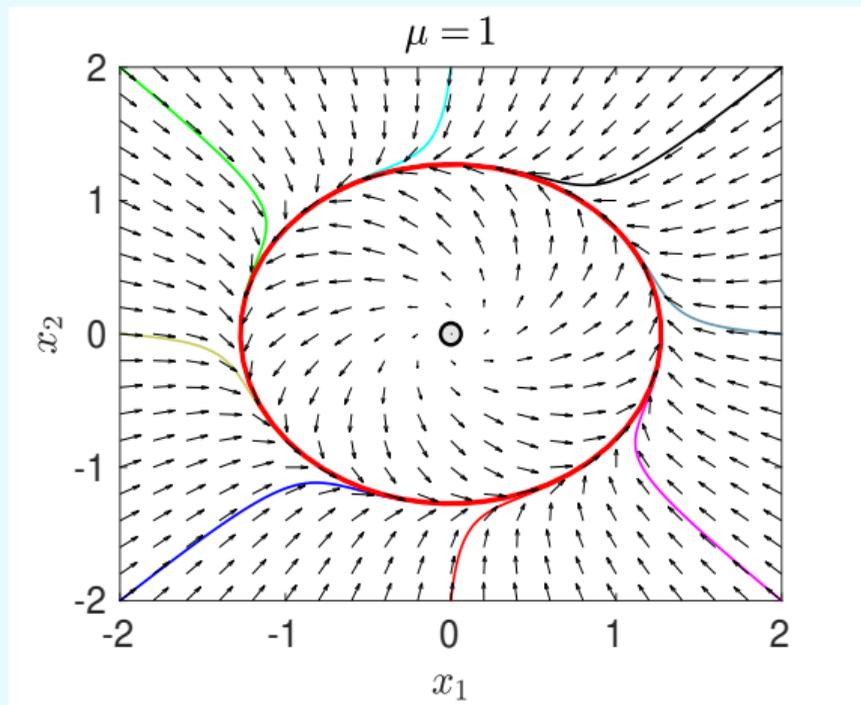
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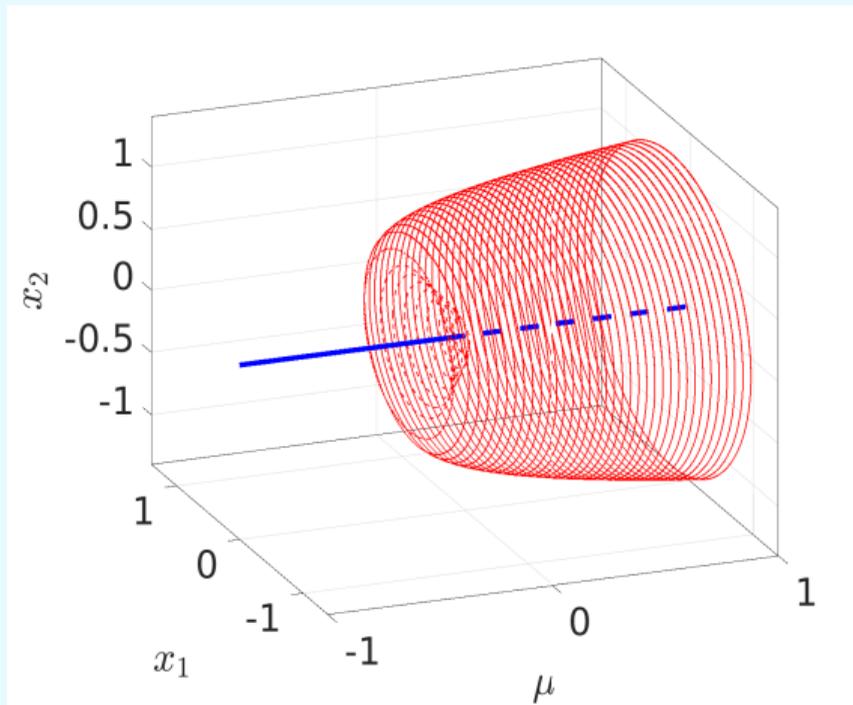
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bifurcation at $\mu = \mu_c$	amplitude	period
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infinite-period bifurcation: we have already presented an example on [Problem Sheet 0](#)

SNIC ... saddle-node bifurcation on invariant circle

SNIPER ... saddle-node infinite-period bifurcation

Example: infinite-period (SNIC, SNIPER) bifurcation

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$\mu \in (-1, 1)$: three critical points:

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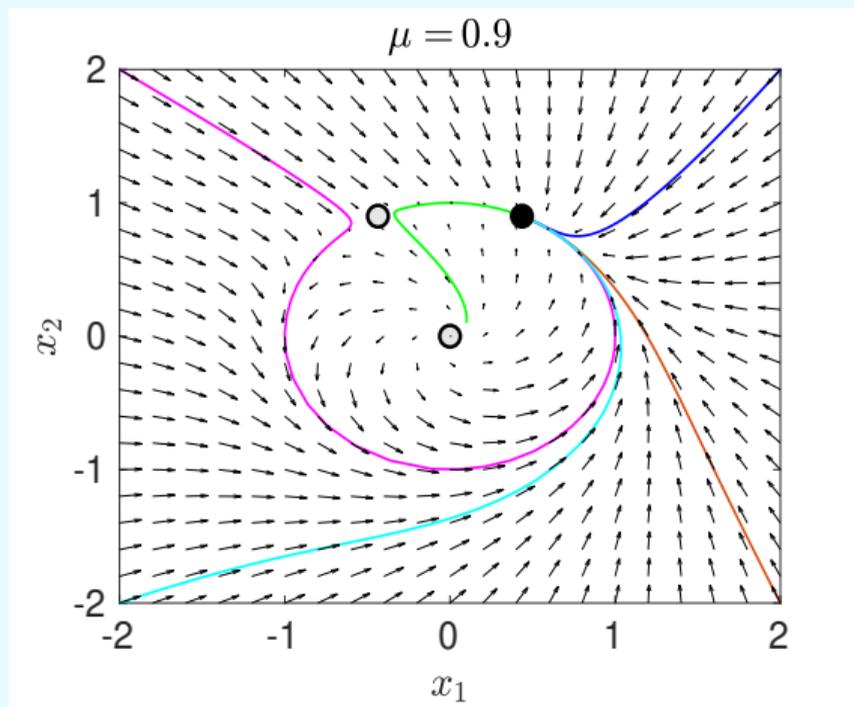
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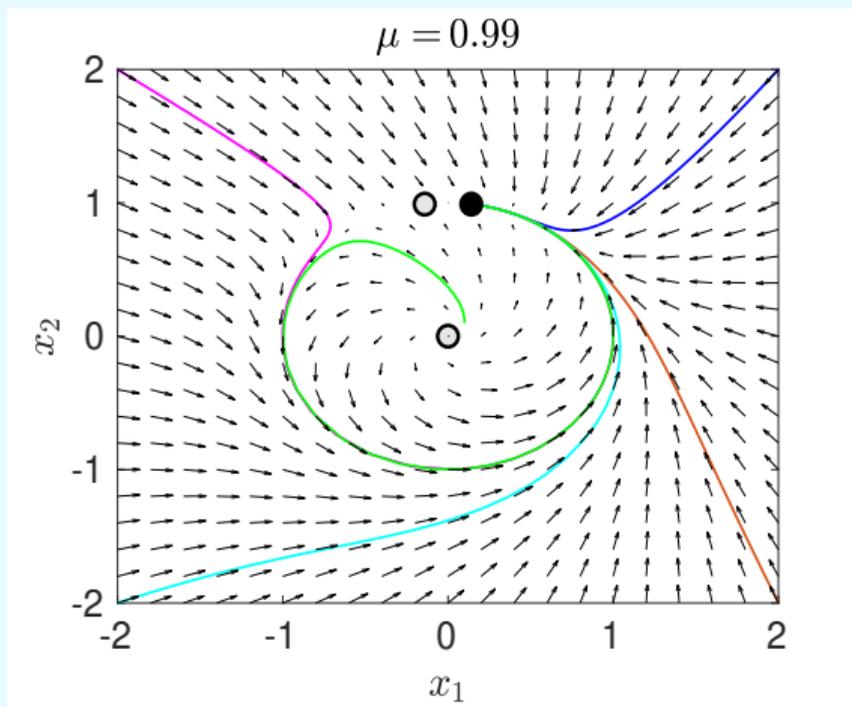
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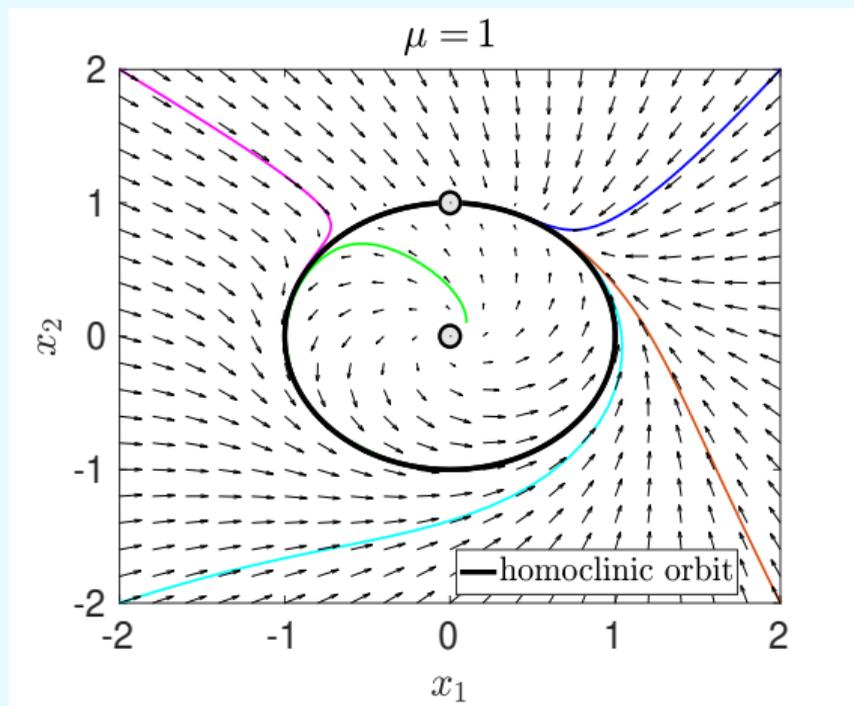
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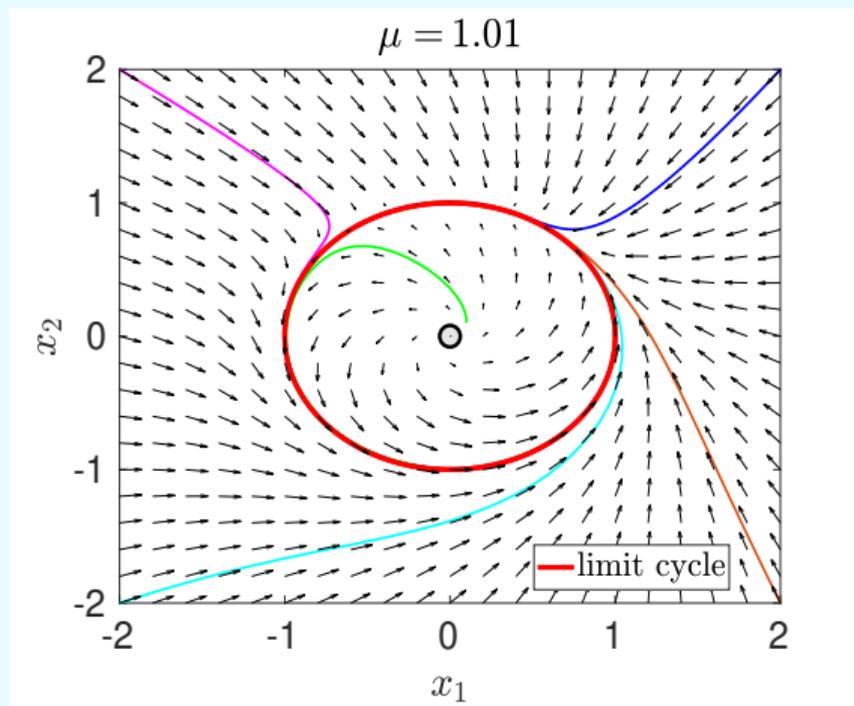
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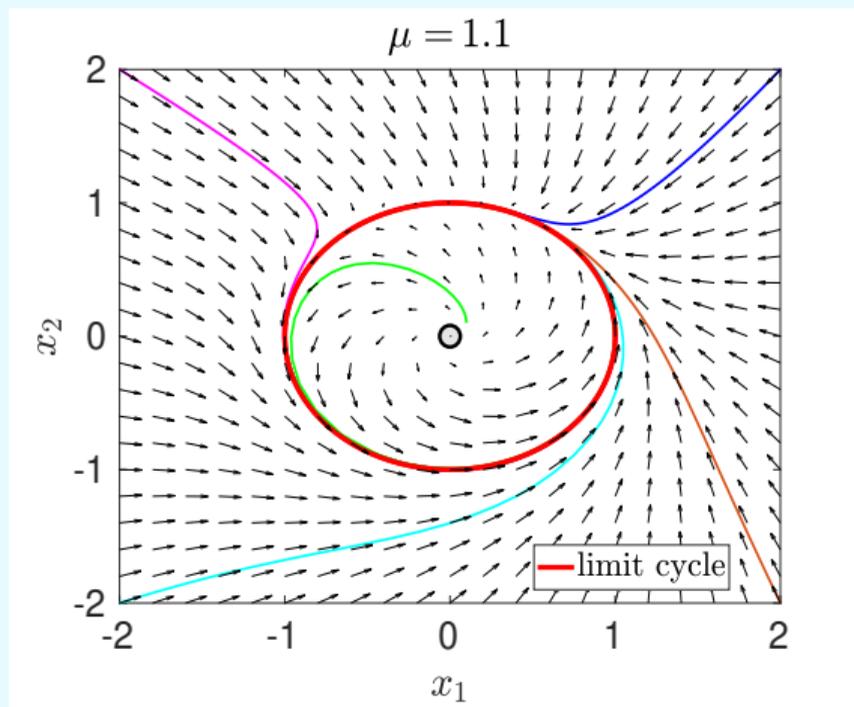
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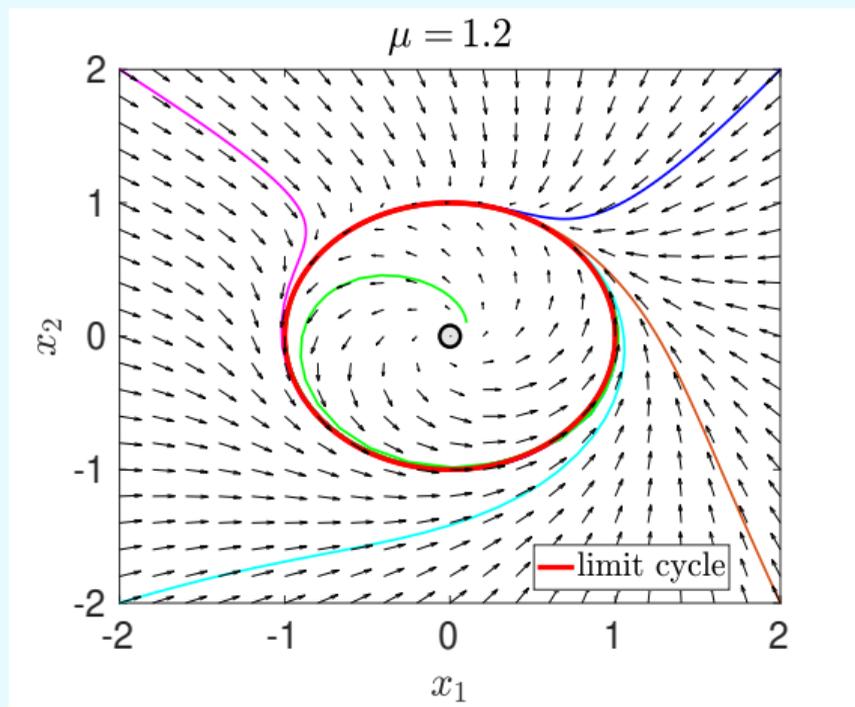
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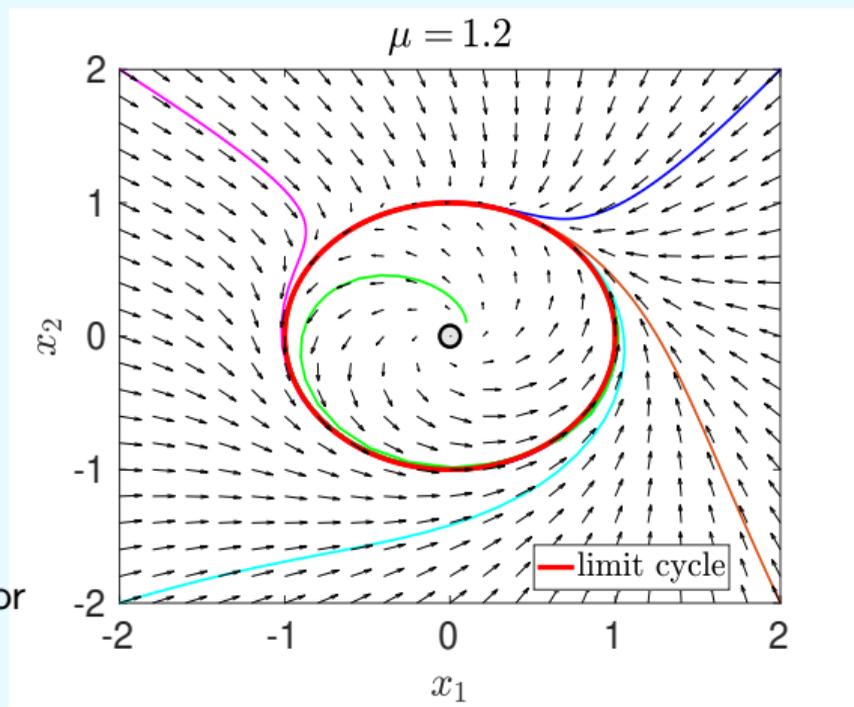
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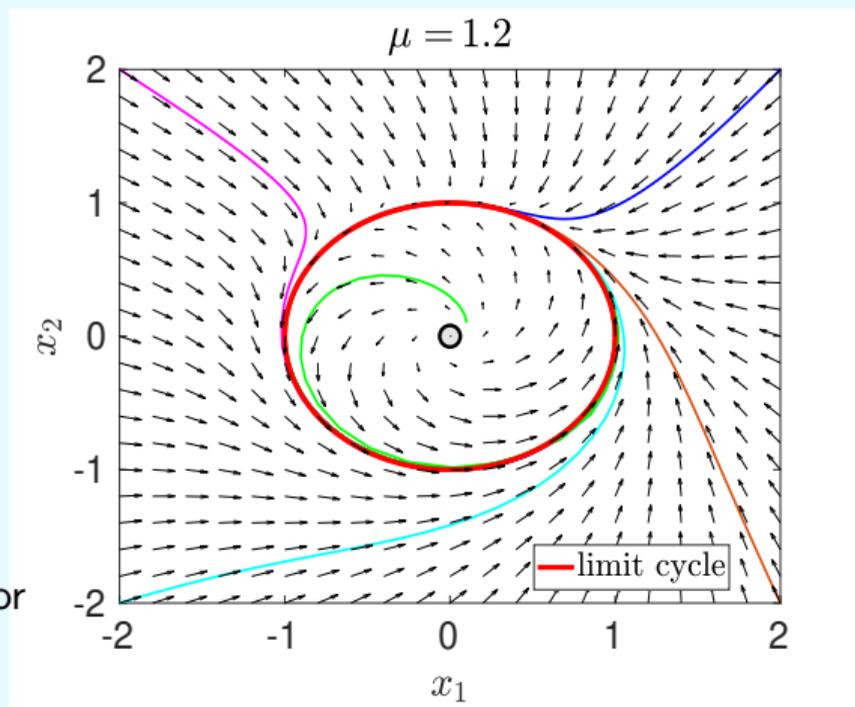
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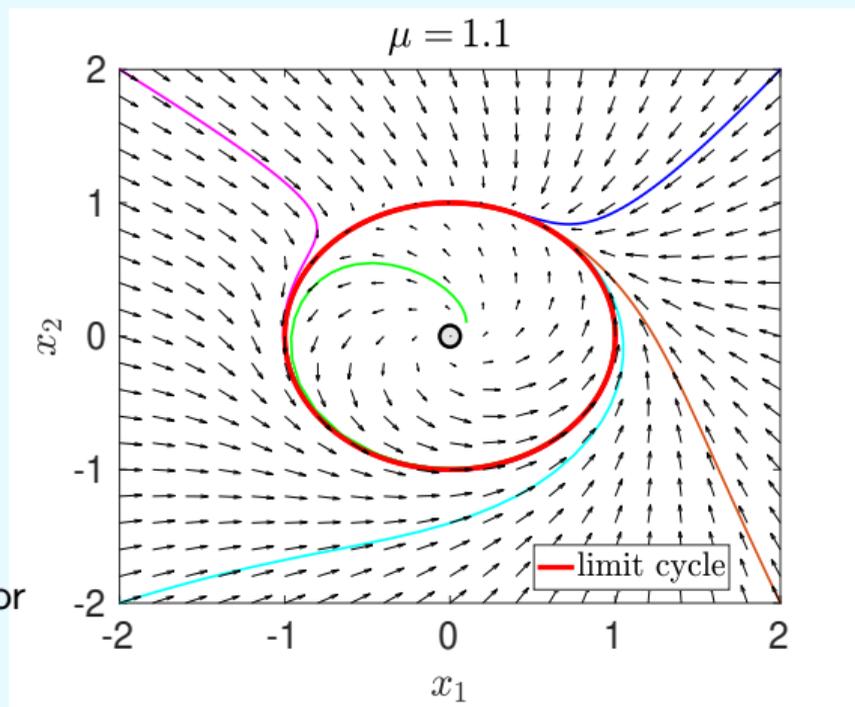
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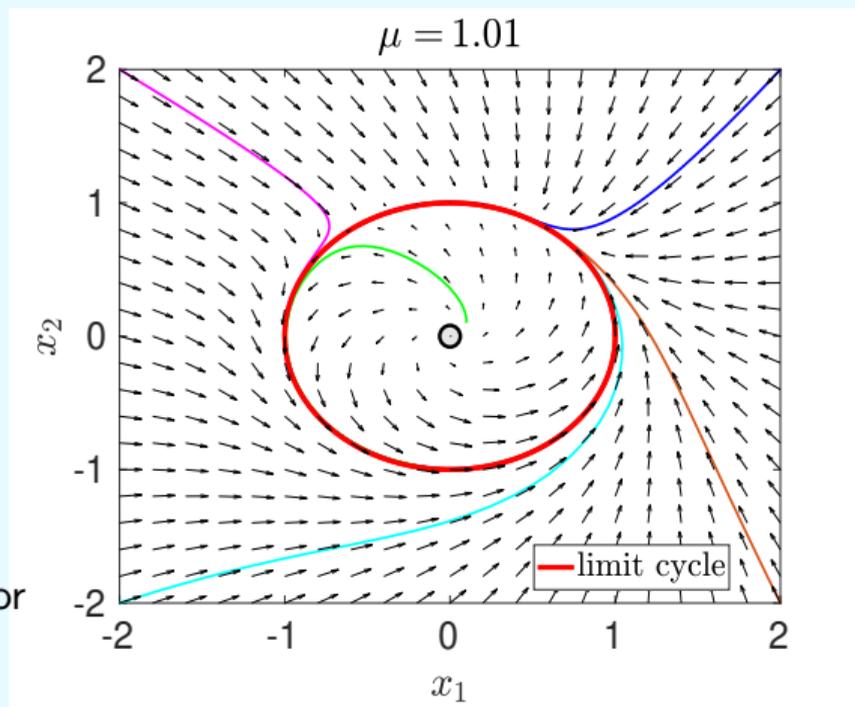
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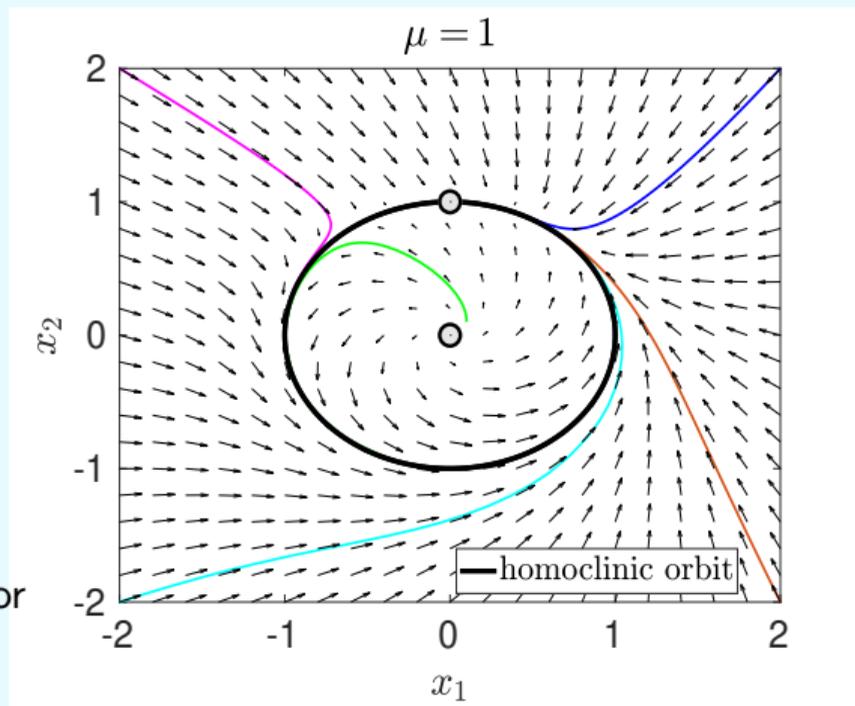
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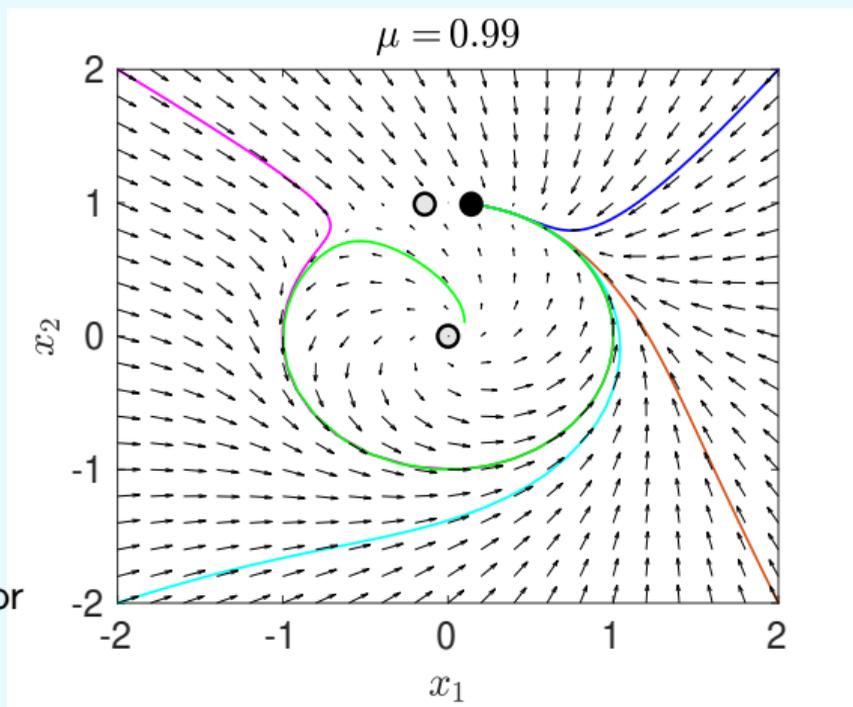
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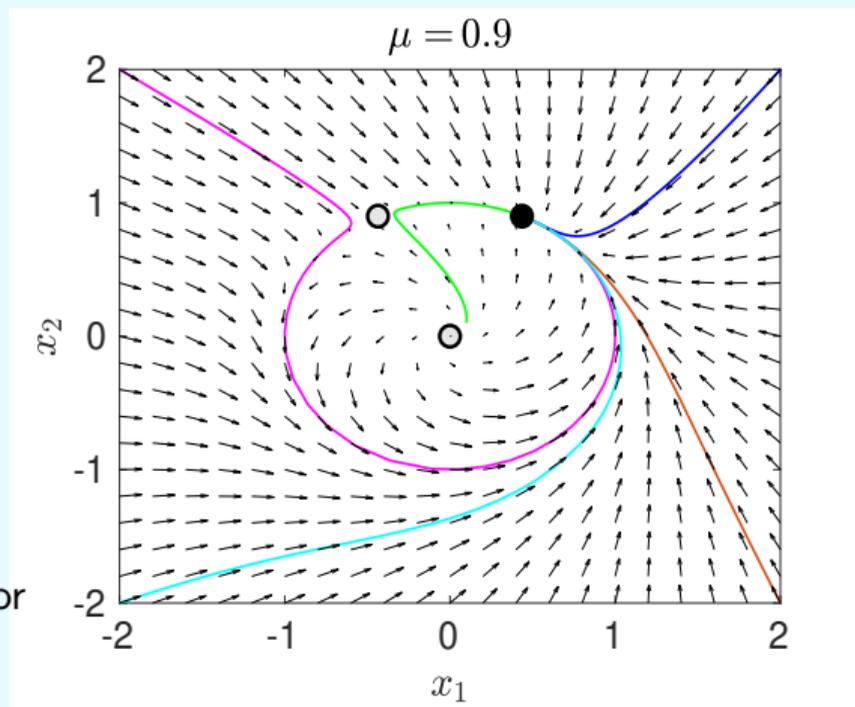
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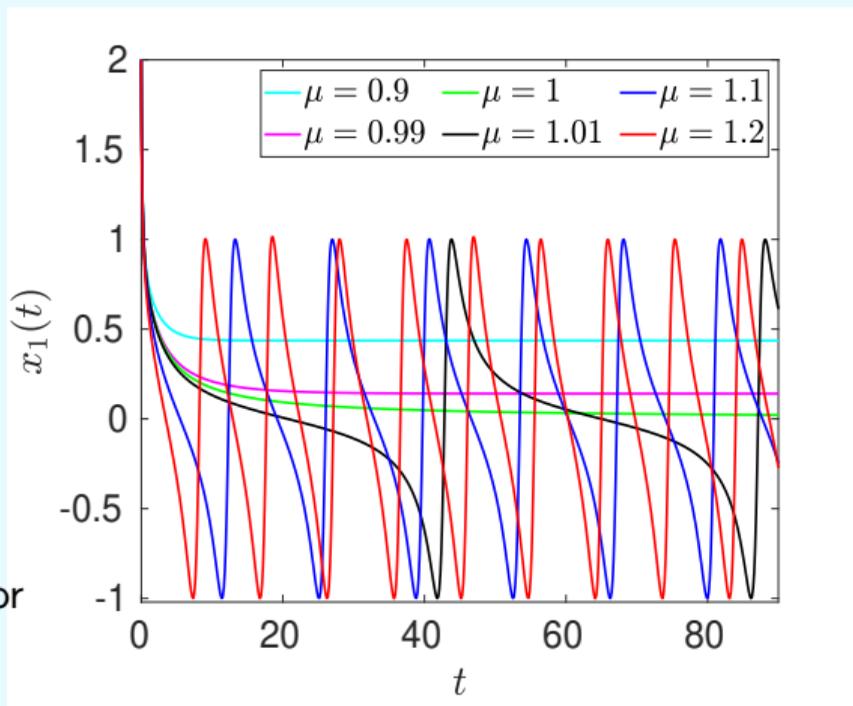
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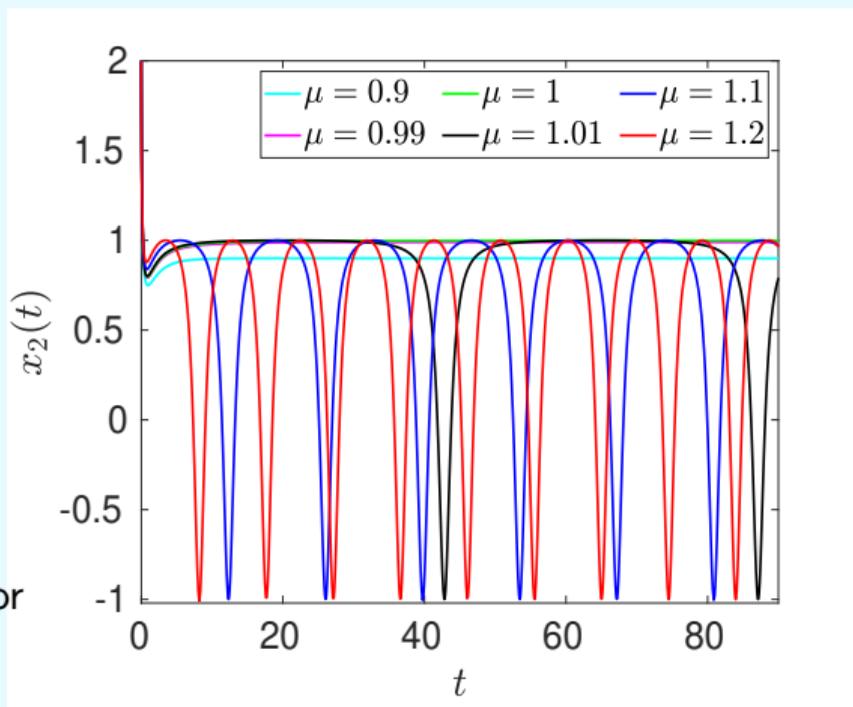
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homoclinic (saddle-loop) bifurcation	$\mathcal{O}(1)$	$\mathcal{O}(\log \mu - \mu_c)$

Bifurcations of limit cycles

bifurcation at $\mu = \mu_c$	amplitude	period
supercritical Hopf bifurcation	$\mathcal{O}\left(\sqrt{ \mu - \mu_c }\right)$	$\mathcal{O}(1)$
subcritical Hopf bifurcation	$\mathcal{O}\left(\sqrt{ \mu - \mu_c }\right)$	$\mathcal{O}(1)$
saddle-node bifurcation of cycles	$\mathcal{O}(1)$	$\mathcal{O}(1)$
infinite-period (SNIC, SNIPER)	$\mathcal{O}(1)$	$\mathcal{O}\left(\frac{1}{\sqrt{ \mu - \mu_c }}\right)$
homoclinic (saddle-loop) bifurcation	$\mathcal{O}(1)$	$\mathcal{O}(\log \mu - \mu_c)$

homoclinic bifurcation: another bifurcation when limit cycle is born with infinite period
saddle-loop bifurcation
new example

Example: homoclinic (saddle-loop) bifurcation

$$\frac{dx_1}{dt} = \mu x_1 + x_2 - x_2^2 - x_1 x_2$$

$$\frac{dx_2}{dt} = -x_1$$

Example

$$\frac{dx_1}{dt} = \mu x_1 + x_2 - x_2^2 - x_1 x_2$$

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two critical points: $\mathbf{x}_{c1} = [0, 0]$ and $\mathbf{x}_{c2} = [0, 1]$

Jacobian matrix is $D\mathbf{f}(\mathbf{x}) = \begin{pmatrix} \mu - x_2 & 1 - 2x_2 - x_1 \\ -1 & 0 \end{pmatrix}$

Example

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$$Df(\mathbf{x}_{c1}) = \begin{pmatrix} \mu & 1 \\ -1 & 0 \end{pmatrix}, \text{ eigenvalues } \lambda_{\pm} = \frac{\mu}{2} \pm \frac{\sqrt{\mu^2 - 4}}{2}$$

$\implies \mathbf{x}_{c1}$ is stable for $\mu < 0$ and unstable for $\mu > 0$

$$Df(\mathbf{x}_{c2}) = \begin{pmatrix} \mu - 1 & -1 \\ -1 & 0 \end{pmatrix}, \text{ eigenvalues } \lambda_{\pm} = \frac{\mu - 1}{2} \pm \frac{\sqrt{\mu^2 - 2\mu + 5}}{2}$$

$\implies \mathbf{x}_{c2}$ is an (unstable) saddle for all $\mu \in \mathbb{R}$

Example

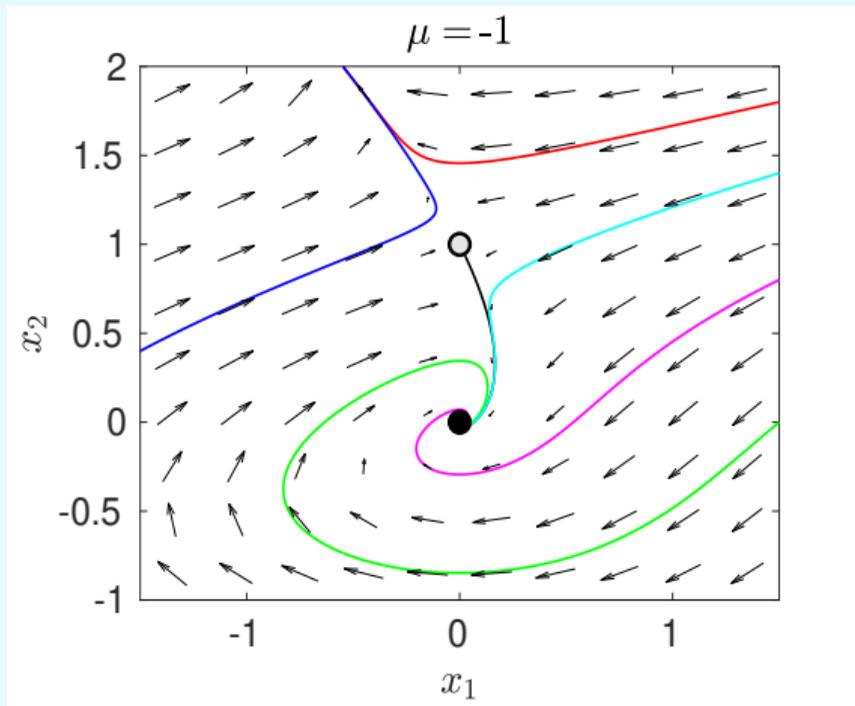
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$\mu < 0$:

fixed point $\mathbf{x}_{c1} = [0, 0]$ is a **stable** spiral

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Example

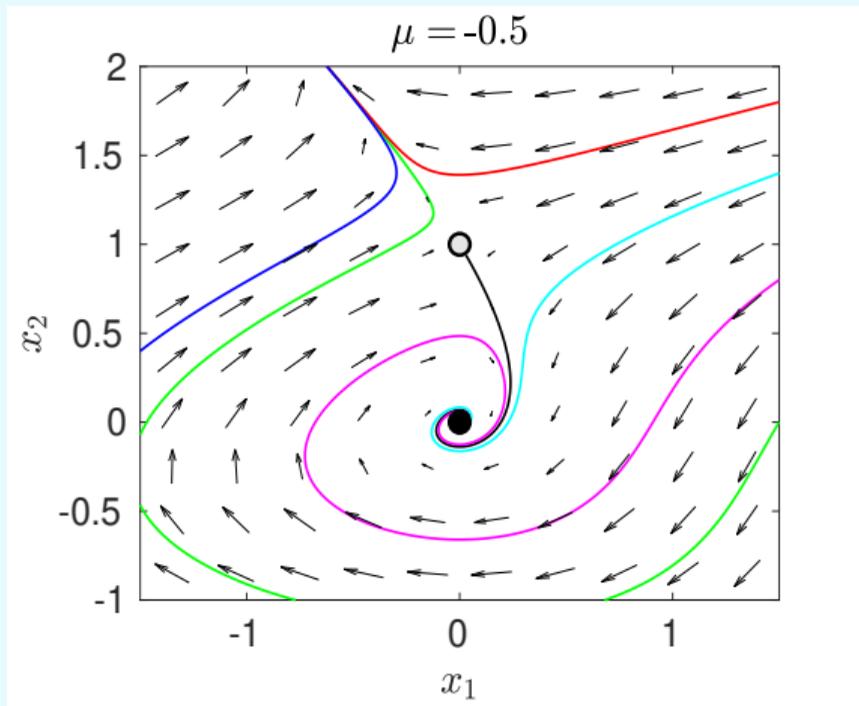
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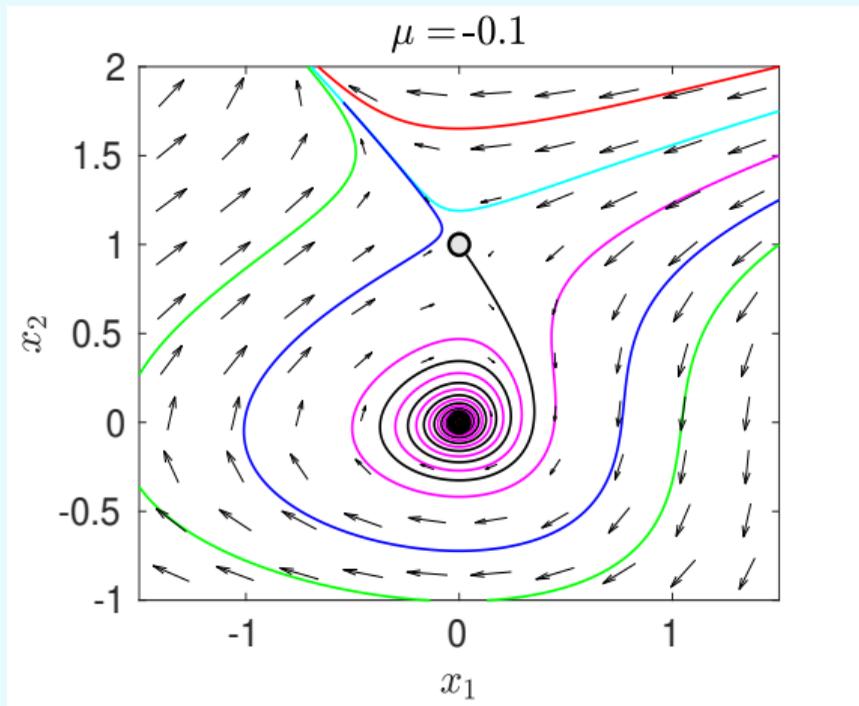
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as μ increases from negative to positive values, eigenvalues cross the imaginary axis from left to right



Example: supercritical Hopf bifurcation

$$\frac{dx_1}{dt} = \mu x_1 + x_2 - x_2^2 - x_1 x_2$$

$$\frac{dx_2}{dt} = -x_1$$

$\mu < 0$:

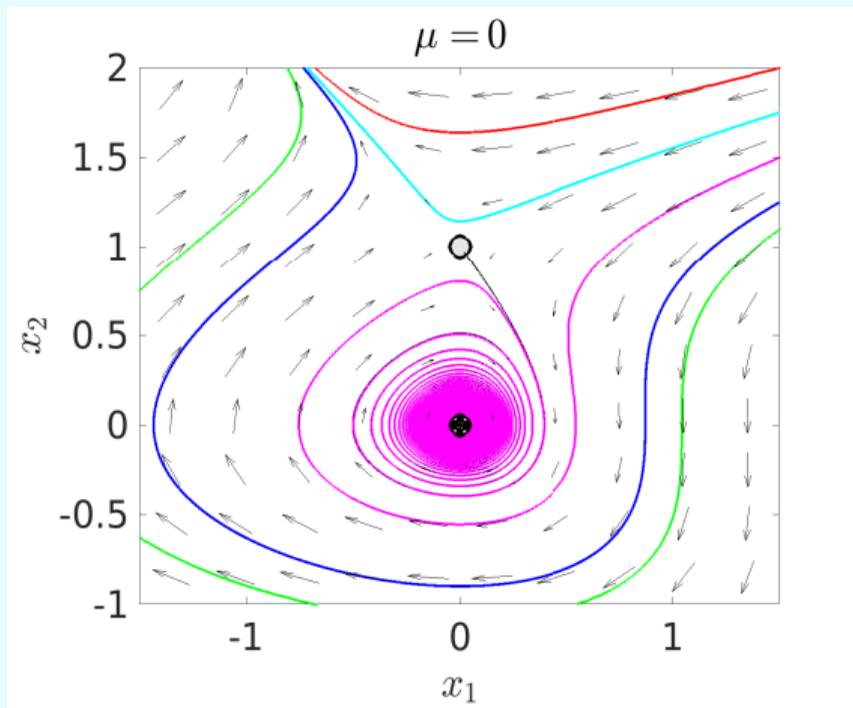
fixed point $\mathbf{x}_{c1} = [0, 0]$ is a **stable** spiral

$$\text{eigenvalues } \lambda_{\pm} = \frac{\mu}{2} \pm \frac{\sqrt{\mu^2 - 4}}{2}$$

as μ increases from negative to positive values, eigenvalues cross the imaginary axis from left to right

$\mu = 0$: fixed point $\mathbf{x}_{c1} = [0, 0]$ is still a **stable** spiral, though a very weak one
supercritical Hopf bifurcation at $\mu = 0$

the limit cycle exists in interval $\mu \in (0, 0.135454802155 \dots)$



Example: homoclinic (saddle-loop) bifurcation

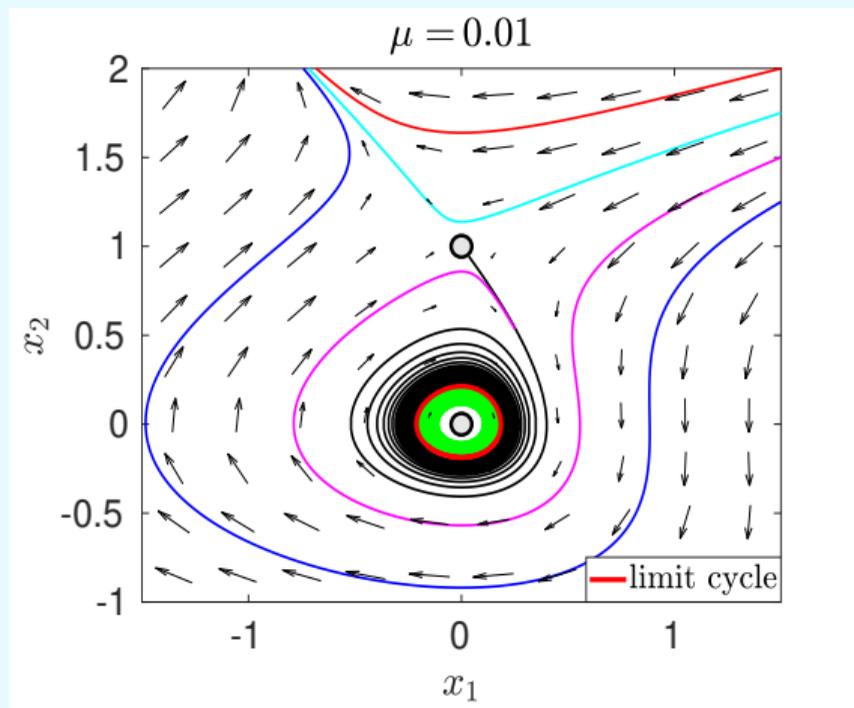
$$\frac{dx_1}{dt} = \mu x_1 + x_2 - x_2^2 - x_1 x_2$$

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$\mu > 0$: $\mathbf{x}_{c1} = [0, 0]$ is an **unstable** spiral

the limit cycle exists in interval

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Example: homoclinic (saddle-loop) bifurcation

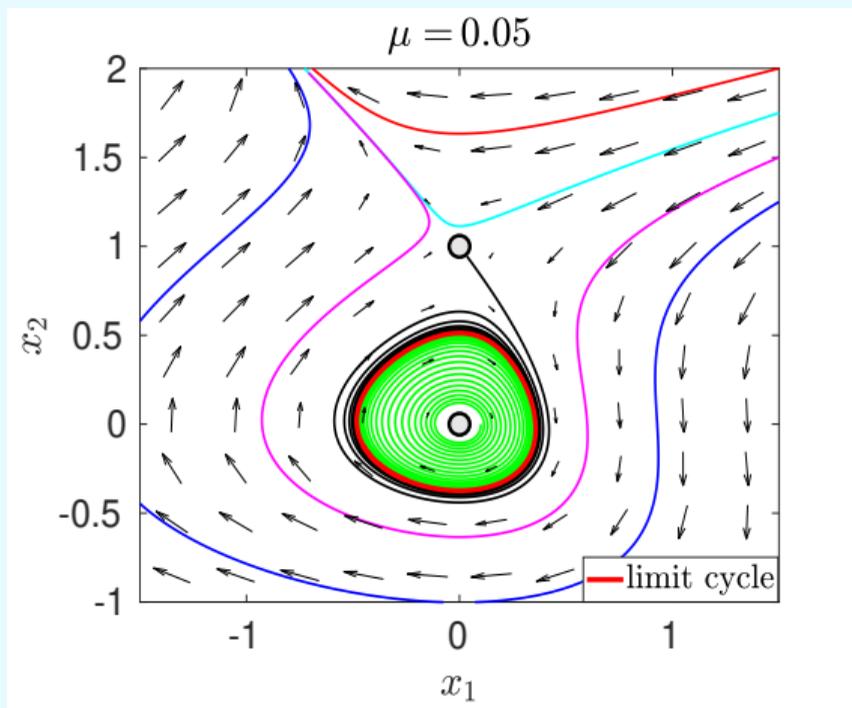
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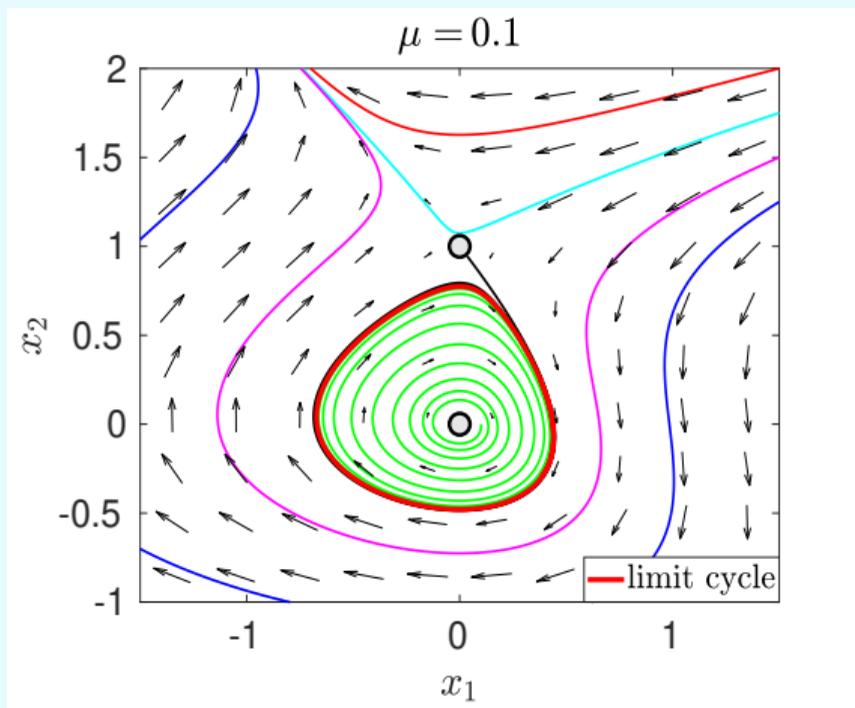
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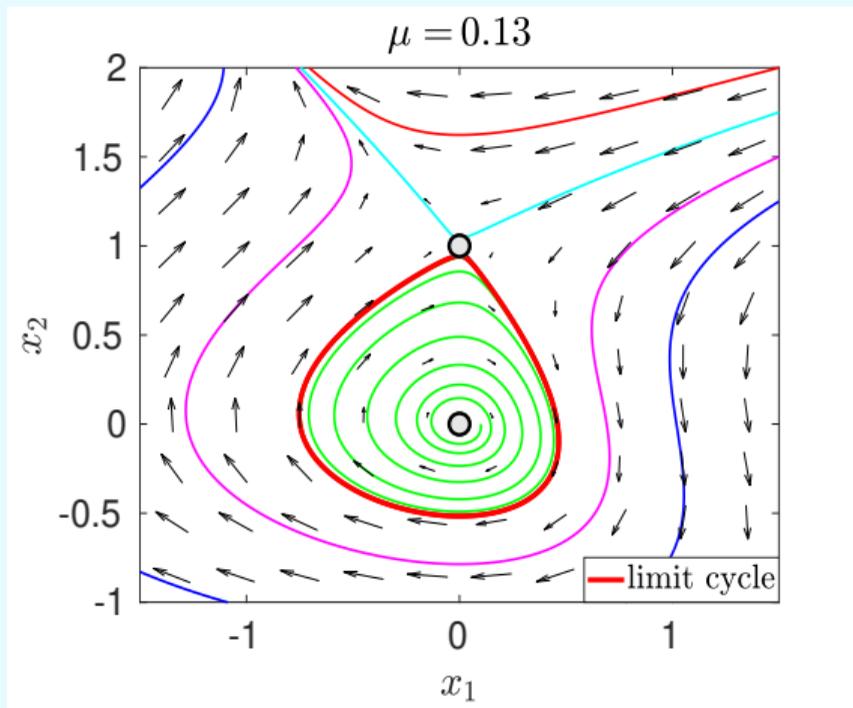
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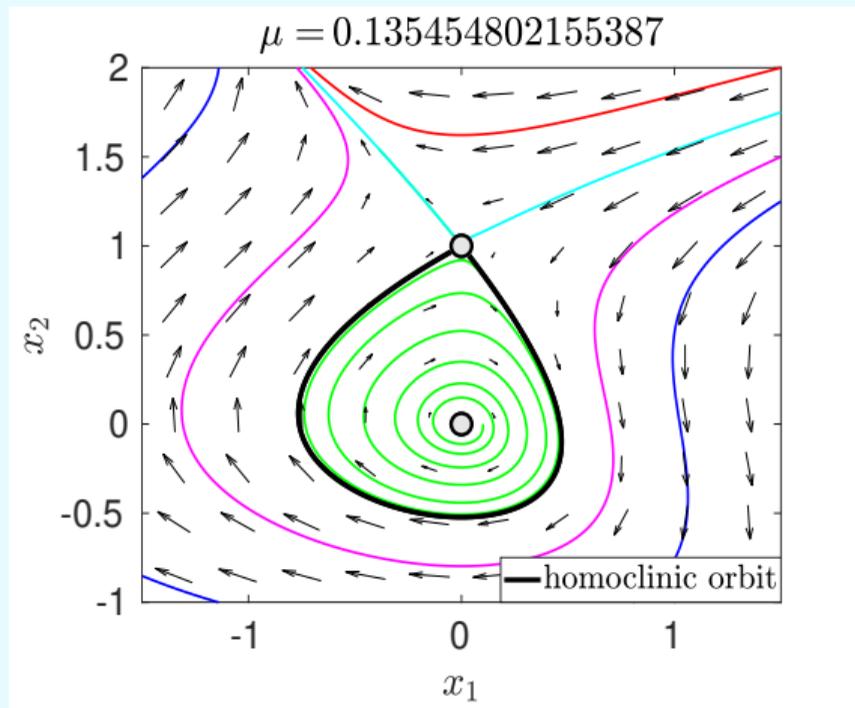
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the limit cycle exists in interval

$$\mu \in (0, 0.135454802155\dots)$$

$\mu = 0.135454802155\dots$: limit cycle collides with the saddle at $\mathbf{x}_{c2} = [0, 1]$ and it becomes a homoclinic orbit

homoclinic (saddle-loop) bifurcation



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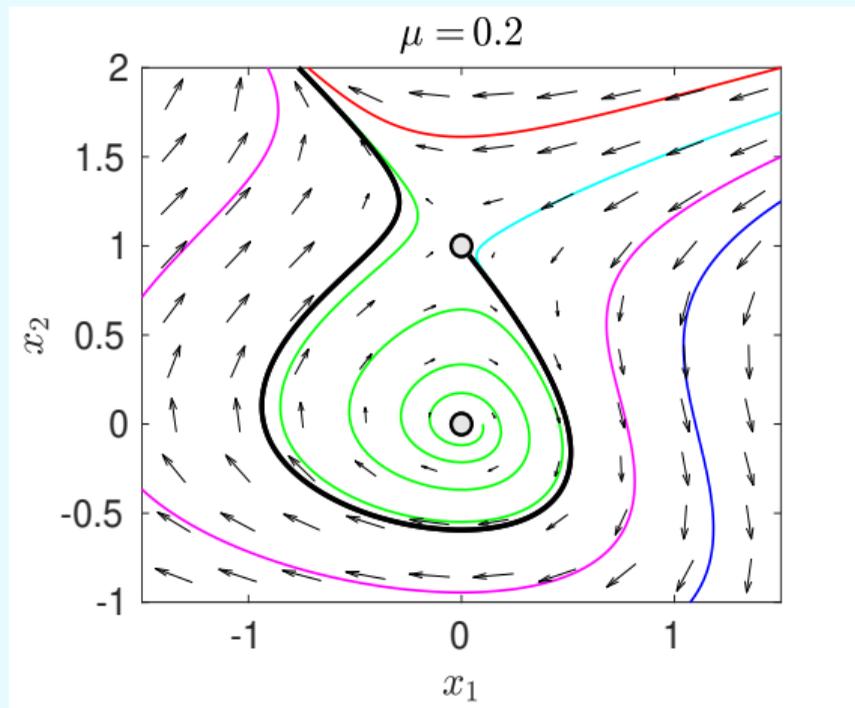
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homoclinic (saddle-loop) bifurcation

$\mu > 0.135454802155\dots$: no limit cycle



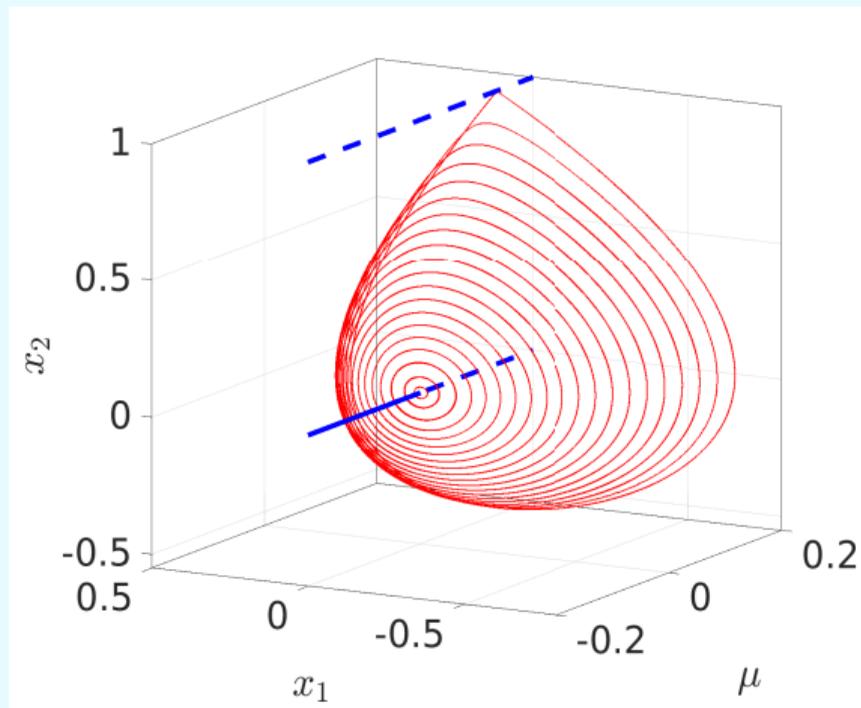
Example: homoclinic bifurcation and supercritical Hopf bifurcation

$$\frac{dx_1}{dt} = \mu x_1 + x_2 - x_2^2 - x_1 x_2$$

$$\frac{dx_2}{dt} = -x_1$$

bifurcation diagram

[show 3D animation]



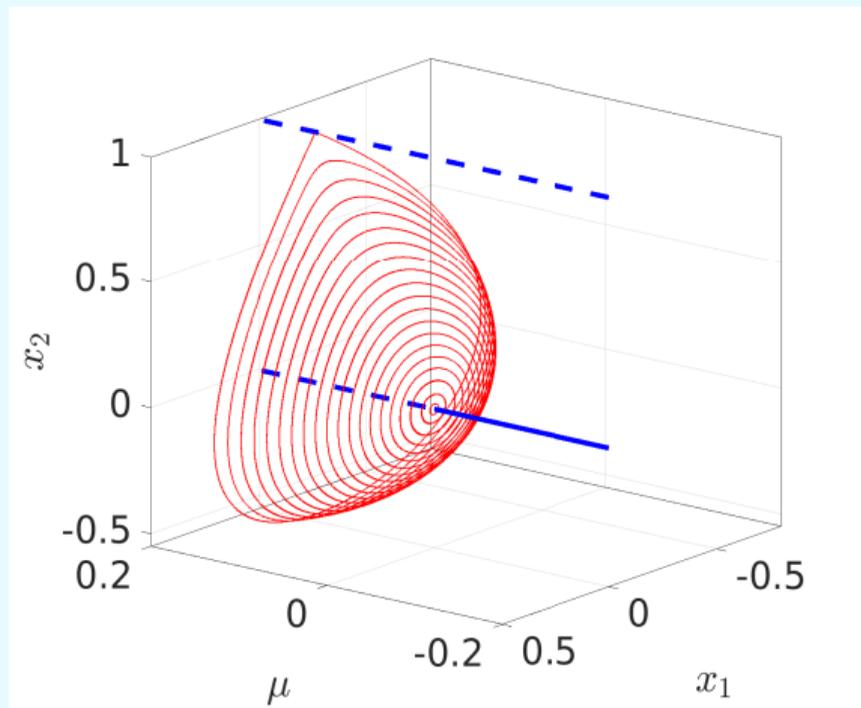
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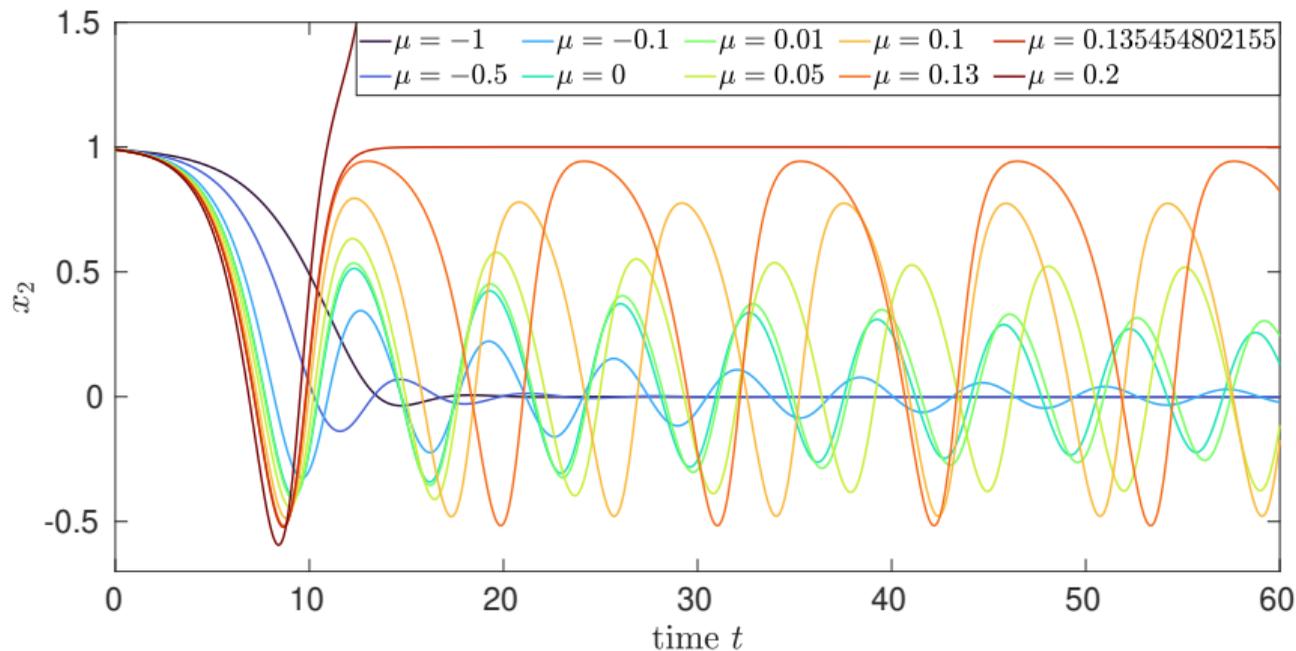
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bifurcation diagram

[show 3D animation]



Example: homoclinic bifurcation and supercritical Hopf bifurcation



bifurcation at $\mu = \mu_c$	amplitude	period
supercritical Hopf bifurcation	$\mathcal{O}(\sqrt{\mu - \mu_c})$	$\mathcal{O}(1)$
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Summary: bifurcations of limit cycles

bifurcation at $\mu = \mu_c$	amplitude	period
supercritical Hopf bifurcation	$\mathcal{O}\left(\sqrt{ \mu - \mu_c }\right)$	$\mathcal{O}(1)$
subcritical Hopf bifurcation	$\mathcal{O}\left(\sqrt{ \mu - \mu_c }\right)$	$\mathcal{O}(1)$
saddle-node bifurcation of cycles	$\mathcal{O}(1)$	$\mathcal{O}(1)$
infinite-period (SNIC, SNIPER)	$\mathcal{O}(1)$	$\mathcal{O}\left(\frac{1}{\sqrt{ \mu - \mu_c }}\right)$
homoclinic (saddle-loop) bifurcation	$\mathcal{O}(1)$	$\mathcal{O}(\log \mu - \mu_c)$

Additional examples: [Questions 1, 4, 5 and 6 on Problem Sheet 3](#).

They are formulated in a way that the questions do not specify what bifurcations of limit cycles are there.

There is also Question 2 on Problem Sheet 3 which asks you to look for a Hopf bifurcation.

B5.6: Nonlinear Dynamics, Bifurcations and Chaos (Lecture 12)

- summary of Lecture 11: we discussed Bifurcations of limit cycles, covering saddle-node bifurcation of cycles, infinite-period (SNIC) bifurcation and homoclinic (saddle-loop) bifurcation.
- today: we will conclude our discussion of Problem Sheet 3

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- course synopsis of **Lectures 9-16**:
Bifurcations of limit cycles, covering supercritical and subcritical Hopf bifurcations, saddle-node bifurcation of cycles, infinite-period (SNIC) bifurcation and homoclinic (saddle-loop) bifurcation. Oscillations in chemical reaction networks. **Weakly nonlinear oscillators. Poincaré-Lindstedt method. Conservative and non-conservative systems. Liénard systems, van der Pol oscillator.** Hilbert's 16th problem. Lorenz equations. Lorenz map. Poincaré section. Poincaré map. Converse of Sharkovsky's theorem. Bernoulli shift map, symbolic dynamics. Tent map. Dynamics on metric spaces, sensitive dependence on initial conditions, transitivity, conjugate maps, chaotic dynamics.

Weakly nonlinear-oscillators, Poincaré-Lindstedt method

- Weakly nonlinear-oscillators: $\frac{d^2x}{dt^2} = -x + \varepsilon g\left(x, \frac{dx}{dt}\right)$ where $0 < \varepsilon \ll 1$
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- we will apply the Poincaré-Lindstedt method to examples of both conservative and non-conservative systems
- conservative systems:
 - derivation on the whiteboard: $\frac{d^2x}{dt^2} = -x + \varepsilon x^3$
 - additional example $\frac{d^2x}{dt^2} = -x + \varepsilon x^2$ is analyzed in [Question 3 on Problem Sheet 3](#) (solutions are available on the course website)

- non-conservative systems: we will consider the van der Pol oscillator

$$\frac{d^2x}{dt^2} = -x + \mu(1 - x^2) \frac{dx}{dt}$$

which can be analyzed using the Poincaré-Lindstedt method for $\mu = \varepsilon \ll 1$

van der Pol oscillator

$$\frac{d^2x}{dt^2} = -x + \mu(1 - x^2) \frac{dx}{dt}$$

van der Pol oscillator

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Denoting $y_1 = x$ and $y_2 = \frac{dx}{dt}$, we can rewrite the van der Pol equation as

$$\frac{dy_1}{dt} = y_2$$

$$\frac{dy_2}{dt} = -y_1 + \mu(1 - y_1^2)y_2$$

van der Pol oscillator

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van der Pol oscillator

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van der Pol oscillator

$$\omega^2(\varepsilon) \frac{d^2x}{d\tau^2} = -x + \varepsilon \omega(\varepsilon)(1 - x^2) \frac{dx}{d\tau}$$

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- The origin $\mathbf{0} = [0, 0]$ is an unstable spiral for $0 < \mu = \varepsilon \ll 1$.
- To apply the Poincaré-Lindstedt method for $\mu = \varepsilon$, we transform the time variable as $\tau = \omega(\varepsilon)t$ where $2\pi/\omega(\varepsilon)$ is the period of the periodic solution.

van der Pol oscillator

$$\omega^2(\varepsilon) \frac{d^2x}{d\tau^2} = -x + \varepsilon \omega(\varepsilon)(1 - x^2) \frac{dx}{d\tau}$$

Substituting

$$x(\tau; \varepsilon) = x_0(\tau) + \varepsilon x_1(\tau) + \varepsilon^2 x_2(\tau) + \dots \quad \text{and} \quad \omega(\varepsilon) = \omega_0 + \varepsilon \omega_1 + \varepsilon^2 \omega_2 + \dots$$

and equating coefficients of ε^0 and ε^1 , we obtain $\omega_0 = 1$, $x_0(\tau) = A \cos(\tau)$ and

$$\frac{d^2x_1}{d\tau^2} + x_1 = -2\omega_1 \frac{d^2x_0}{d\tau^2} + (1 - x_0^2) \frac{dx_0}{d\tau} = 2\omega_1 A \cos(\tau) + \left(\frac{A^3}{4} - A\right) \sin(\tau) + \frac{A^3}{4} \sin(3\tau)$$

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Eliminating the secular terms gives $\omega_1 = 0$ and $A = 2$.

van der Pol oscillator

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Eliminating the secular terms gives $\omega_1 = 0$ and $A = 2$.

We have $x(\tau; \varepsilon) = 2 \cos(\omega t) + \varepsilon \sin^3(\omega t) + \dots$ with $\omega = 1 - \varepsilon^2/16 + \dots$

\Rightarrow the limit cycle is approximately circular with radius 2 for $\mu = \varepsilon \ll 1$

van der Pol oscillator

$$\frac{d^2x}{dt^2} = -x + \mu(1 - x^2) \frac{dx}{dt}$$

analysis for $\mu \ll 1$:

Poincaré-Lindstedt method implies that the limit cycle is approximately circular

with radius 2 and period $\frac{2\pi}{1 - \varepsilon^2/16 + \dots}$

van der Pol oscillator

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intermediate values of μ : we can computationally investigate limit cycles

analysis for $\mu \gg 1$: the limit cycle has period $\mu(3 - 3 \log(2))$ as $\mu \rightarrow \infty$

van der Pol oscillator

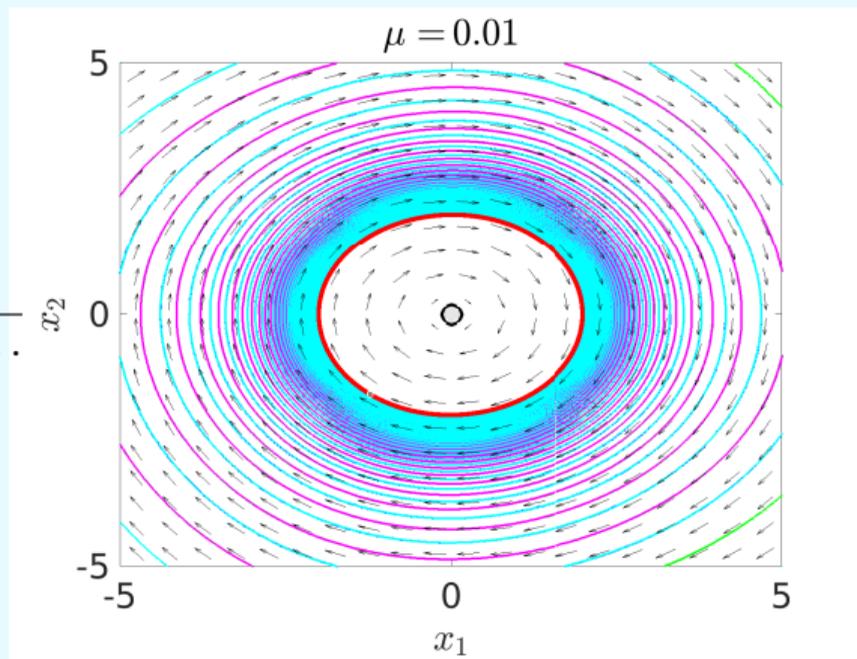
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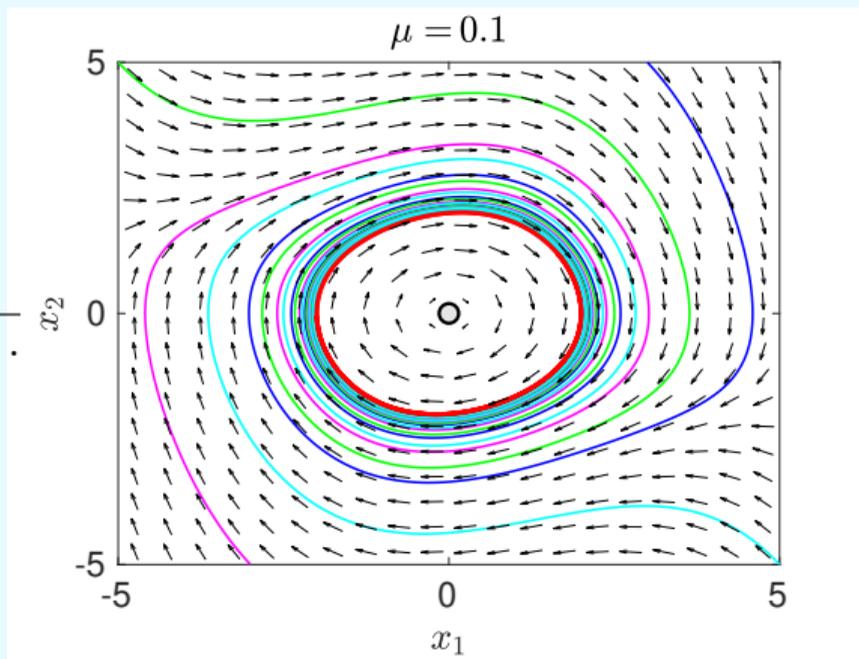
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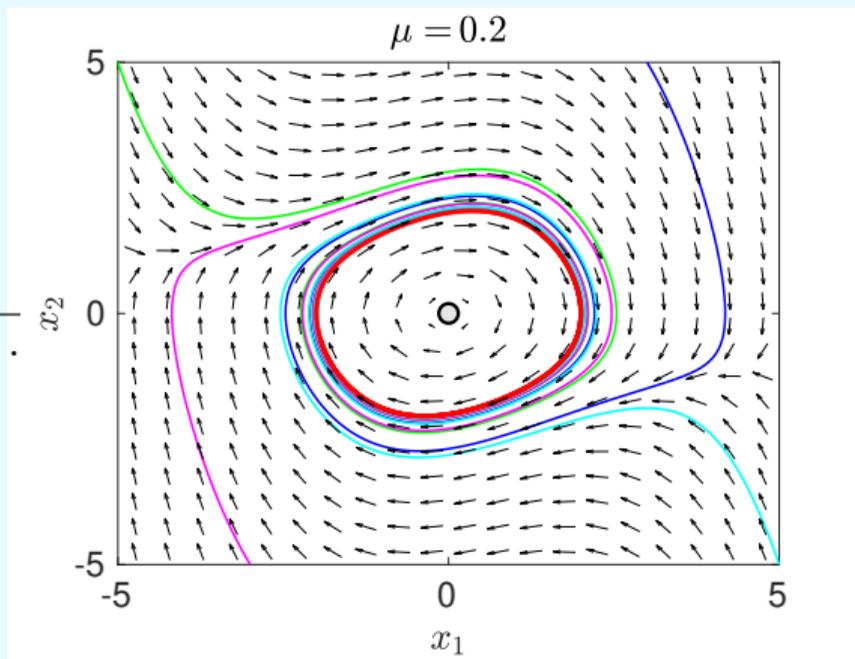
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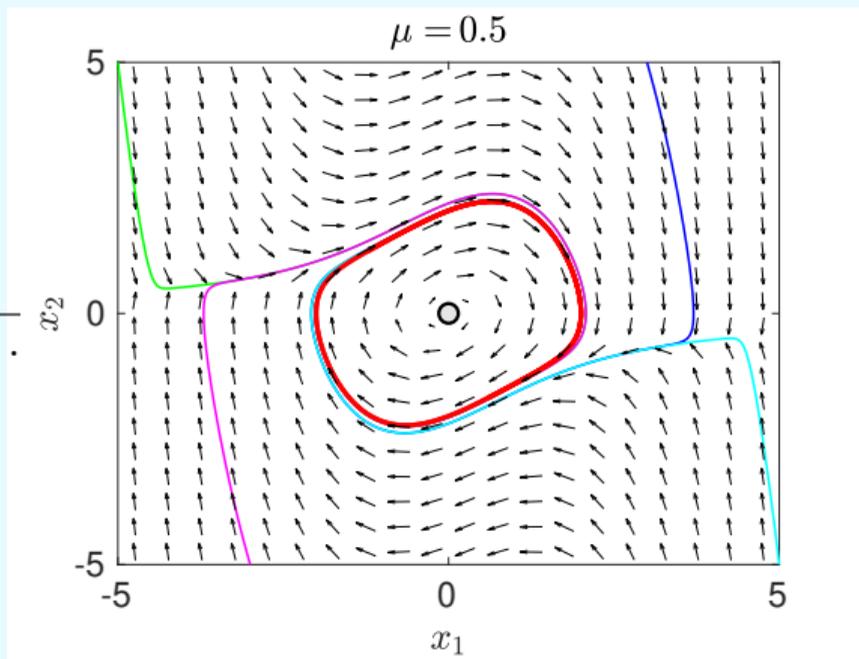
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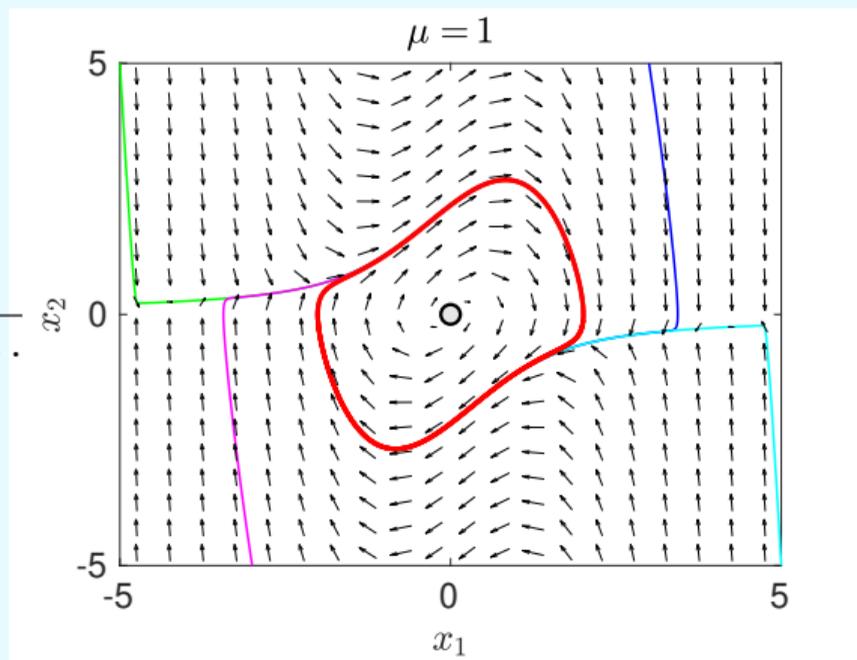
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van der Pol oscillator

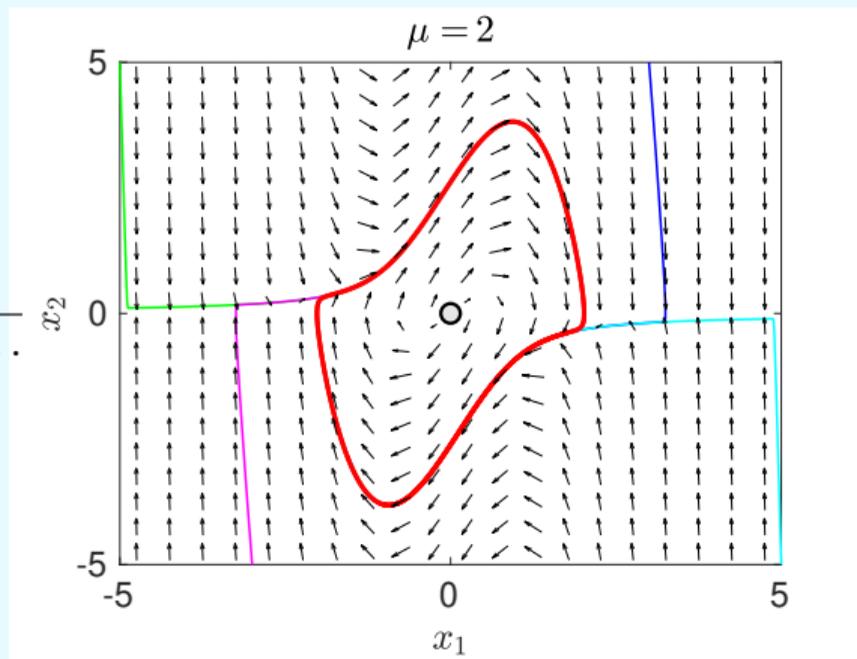
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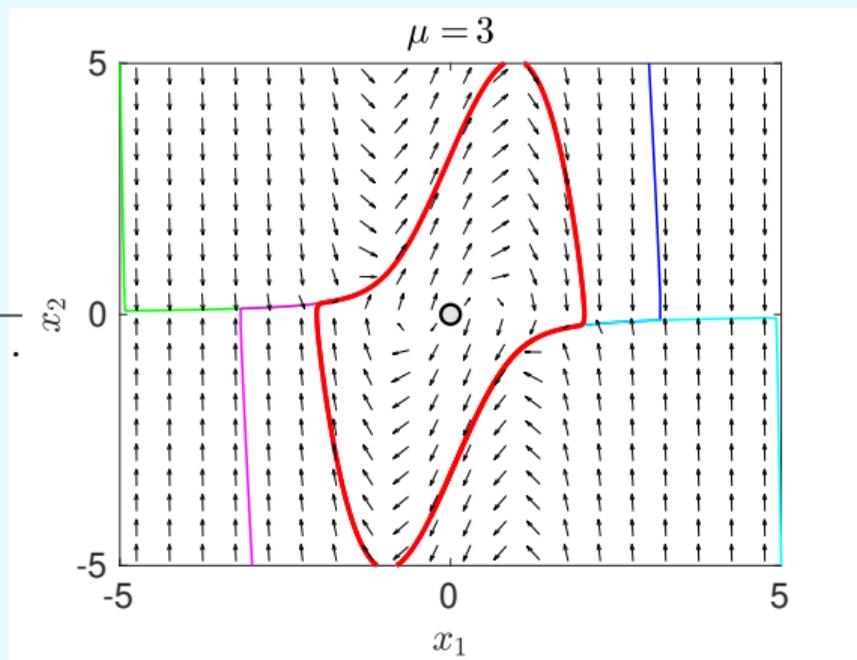
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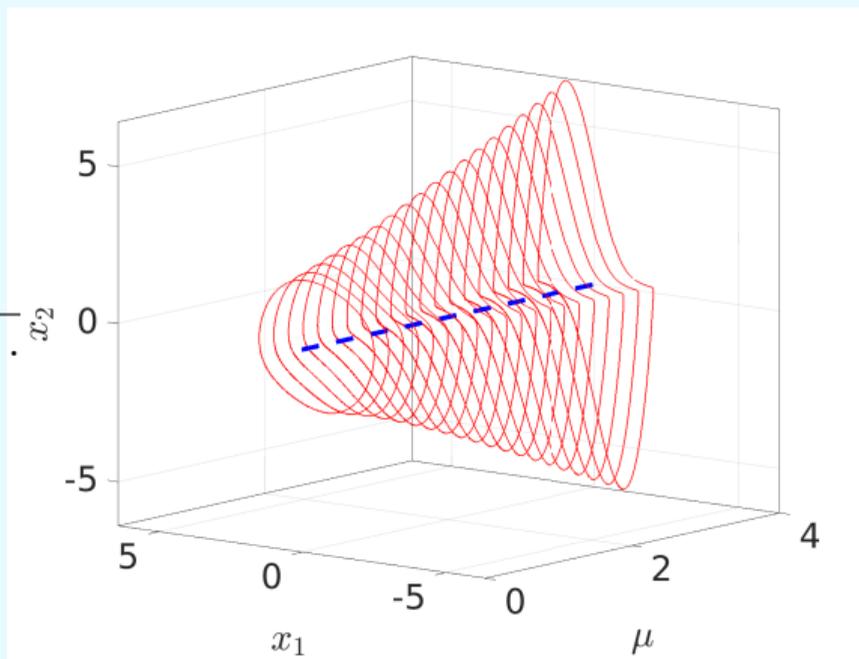
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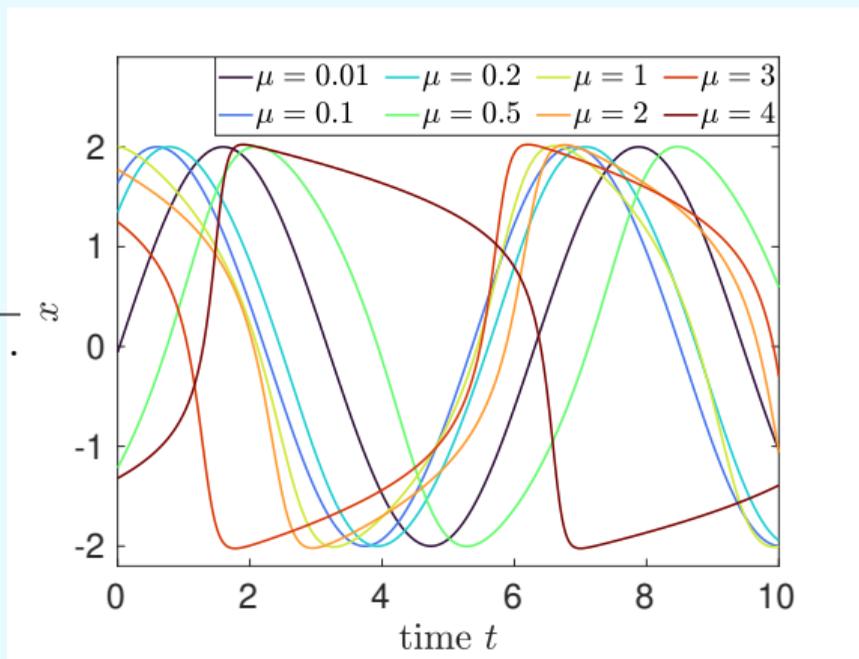
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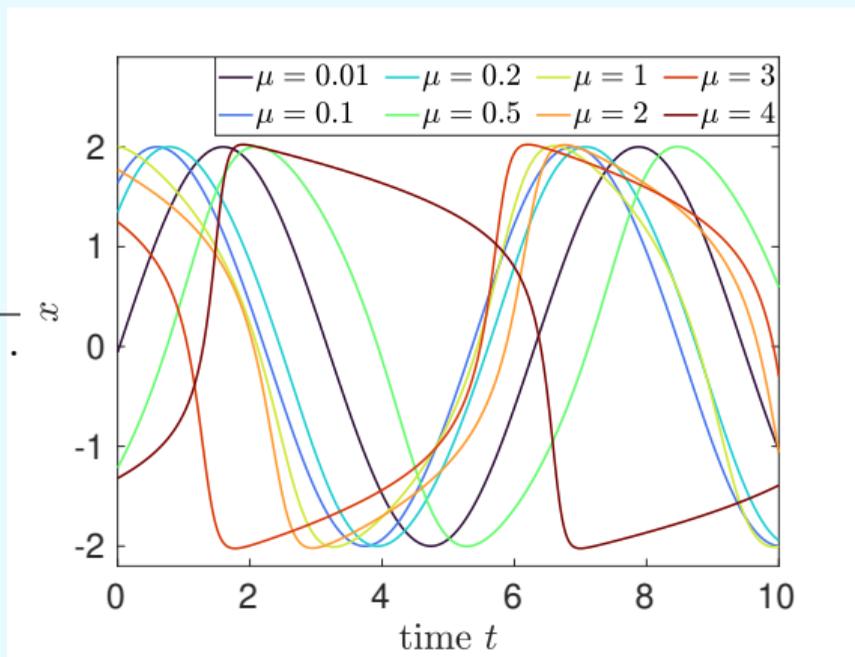
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the van der Pol equation is a special case of the Liénard equation

$$\frac{d^2x}{dt^2} = -g(x) - f(x) \frac{dx}{dt}$$

for $g(x) = x$ and $f(x) = \mu(x^2 - 1)$

B5.6: Nonlinear Dynamics, Bifurcations and Chaos (Lecture 13)

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- course synopsis of **Lectures 9-16**:
Bifurcations of limit cycles, covering supercritical and subcritical Hopf bifurcations, saddle-node bifurcation of cycles, infinite-period (SNIC) bifurcation and homoclinic (saddle-loop) bifurcation. Oscillations in chemical reaction networks. Weakly nonlinear oscillators. Poincaré-Lindstedt method. Conservative and non-conservative systems. Liénard systems, van der Pol oscillator. Hilbert's 16th problem. Lorenz equations. Lorenz map. Poincaré section. Poincaré map. Converse of Sharkovsky's theorem. Bernoulli shift map, symbolic dynamics. Tent map. Dynamics on metric spaces, sensitive dependence on initial conditions, transitivity, conjugate maps, chaotic dynamics.

Planar ODEs with polynomial right-hand sides ($n = 2$)

Consider the planar autonomous system of ODEs:

$$\frac{dx_1}{dt} = f_1(x_1, x_2)$$

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Suppose that R is compact (*i.e.* closed and bounded) and it does not contain any fixed points. Suppose that there exists $\mathbf{x}_0 \in R$ such that $\phi_t(\mathbf{x}_0) \in R$ for all $t \geq 0$, *i.e.* the trajectory is confined in R for $t \geq 0$.

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There is no chaotic behaviour of planar ($n = 2$) polynomial ODE systems, but there could still be relatively complicated dynamics (multiple limit cycles) and there are a number of unsolved problems.

Hilbert's 16th problem

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[Question 7 on Problem Sheet 4](#): Show that $H(2) \geq 2$.

This is not the best known lower bound on $H(2)$: one can find quadratic systems with four limit cycles, giving $H(2) \geq 4$.

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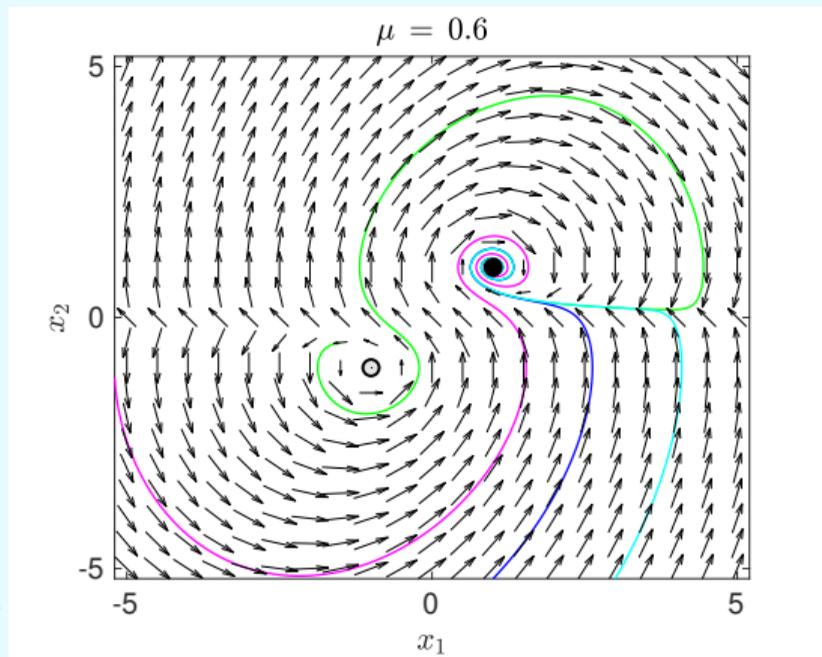
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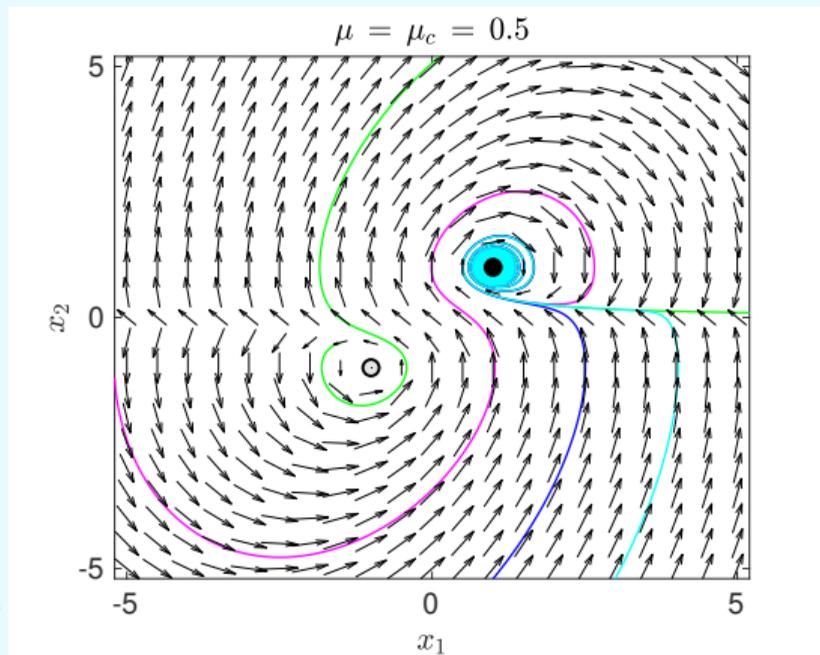
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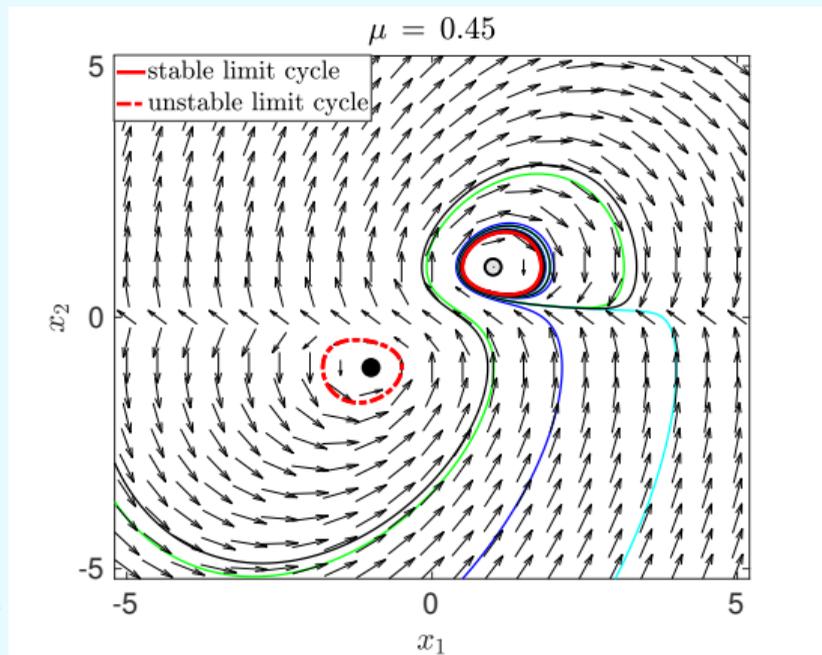
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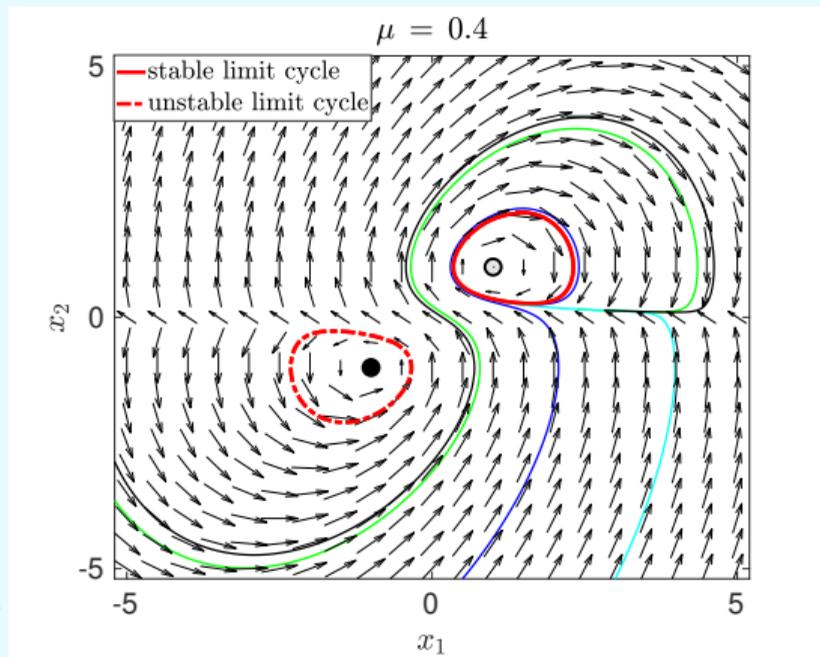
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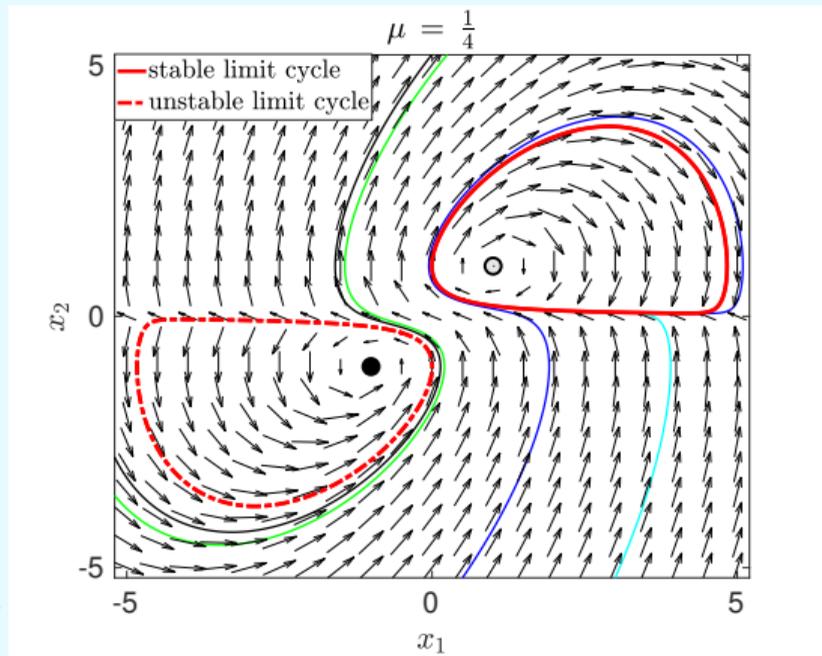
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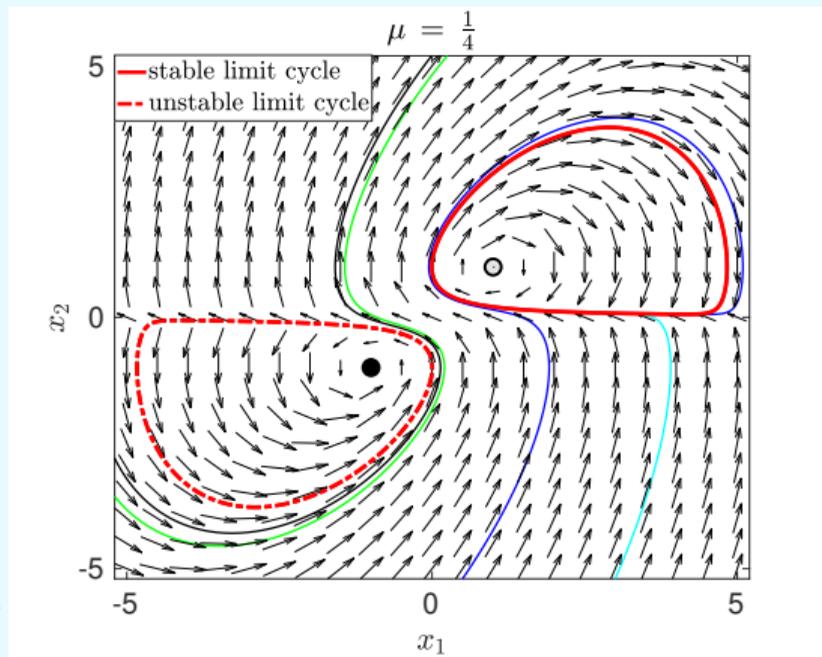
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Question 2 on Problem Sheet 4: another question to find a planar quadratic system undergoing a specific bifurcation (SNIC, SNIPER)



Designing chemical systems with prescribed dynamical behaviour

applications: synthetic biology, DNA computing, engineering artificial networks

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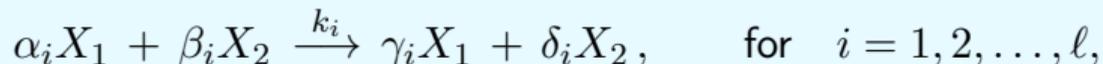
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where each reaction has at most two reactants and at most two products, *i.e.*

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Such chemical systems are sometimes called bimolecular. In this question, you will show that there is no bimolecular chemical system which would have a limit cycle.

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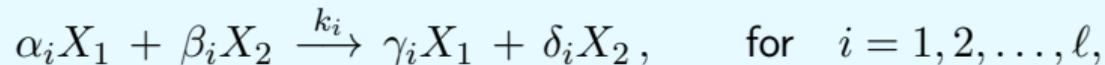
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\implies we need more complex reaction networks to get one, two, three, ... limit cycles

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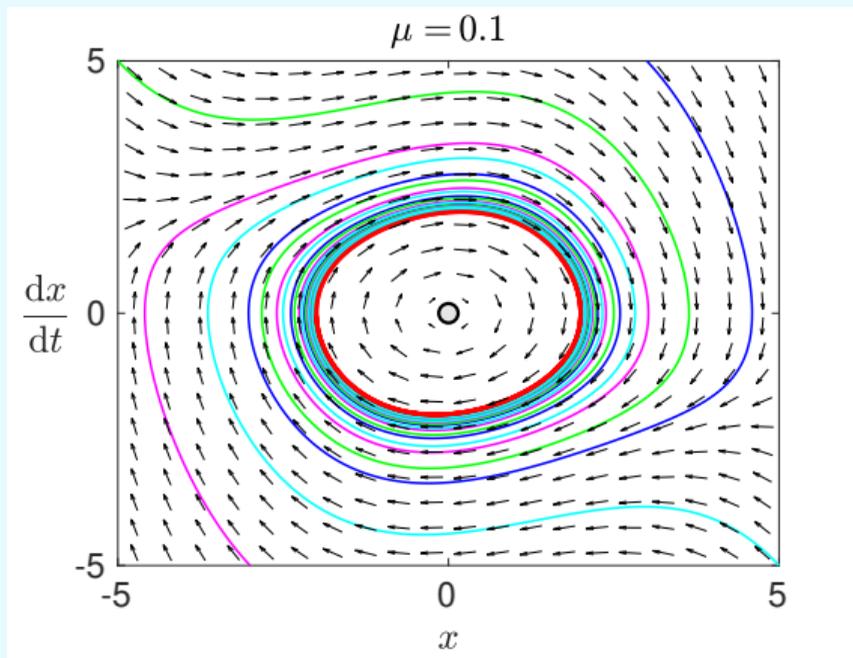
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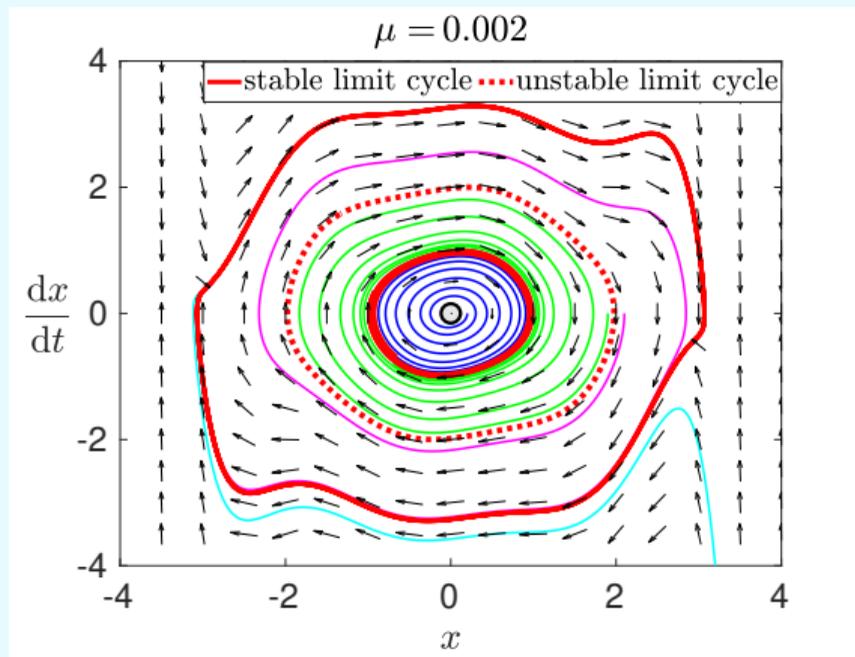
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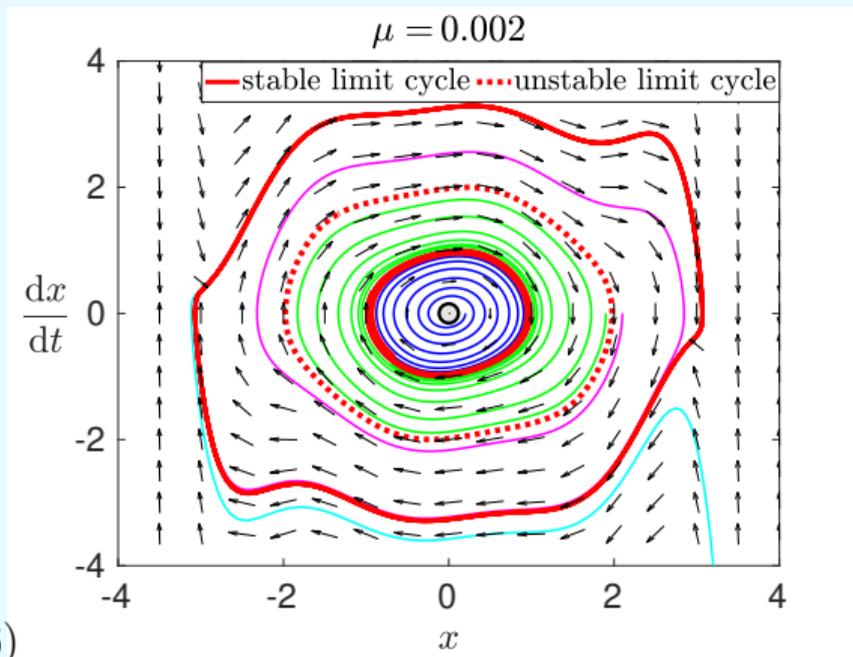
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we could also start in polar coordinates

$$\frac{dr}{dt} = r(r^2 - 1)(r^2 - 4)(r^2 - 9)(r^2 - 16)$$

$$\frac{d\theta}{dt} = 1$$

giving us a polynomial system of degree $d = 9$ with four limit cycles at $r = 1, 2, 3$ and 4 (this is not the most optimal approach: one can find quadratic systems with four limit cycles, giving $H(2) \geq 4$)



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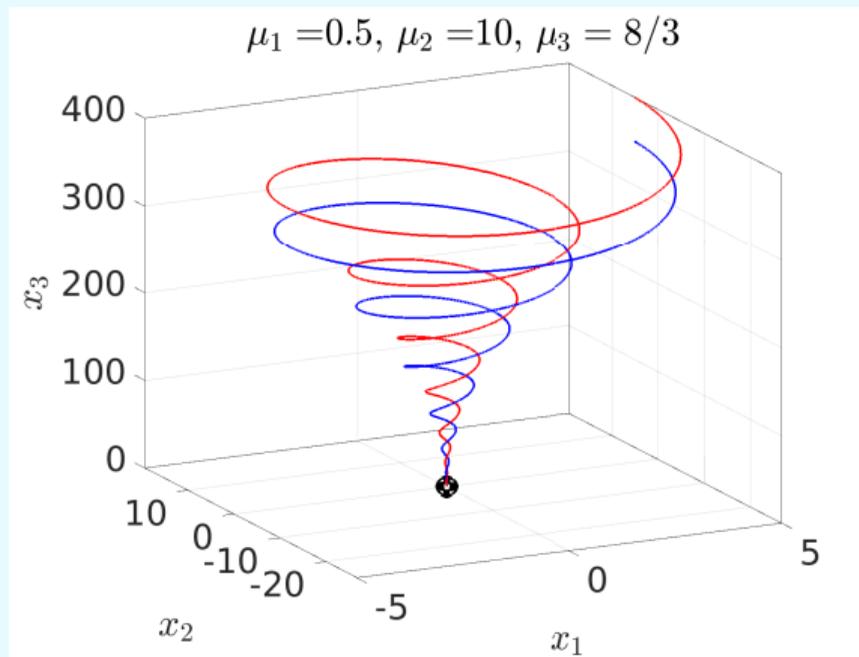
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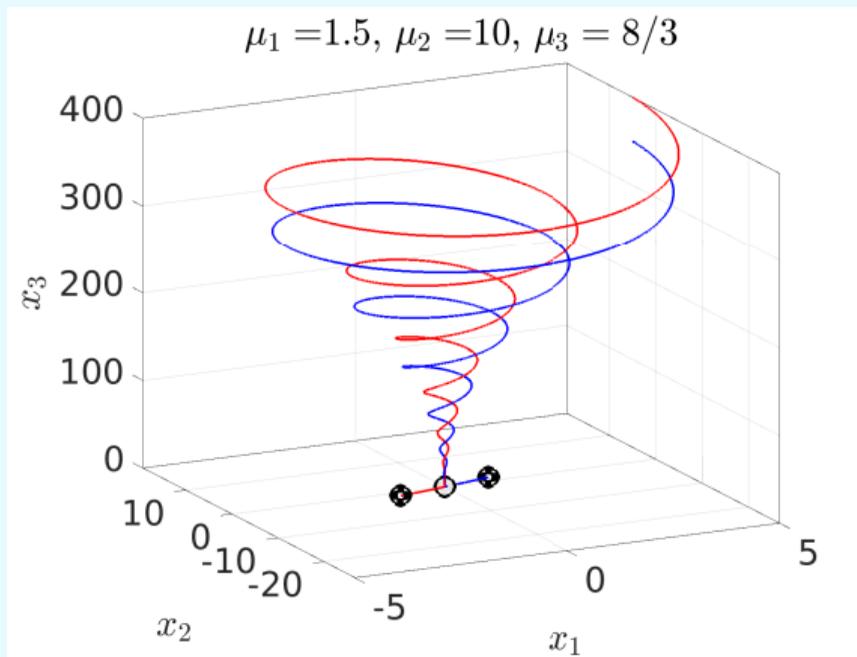
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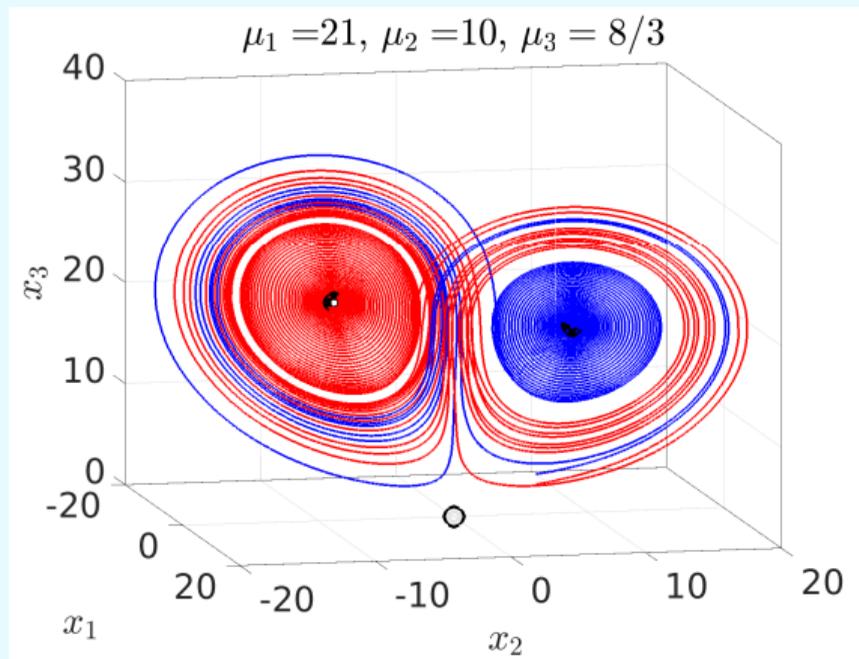
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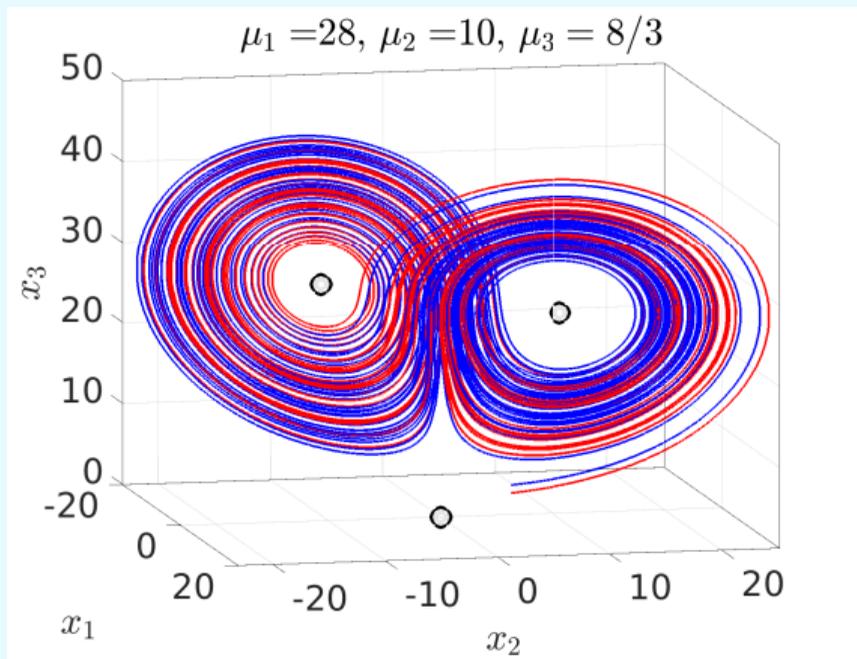
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Lecture 8: we started with a 3D demonstration viewing trajectories in the phase space for different values of parameters μ_1 , μ_2 , μ_3 and illustrating the convergence to fixed points, limit cycles, chaos, strange attractor and transient chaos

we varied μ_1 , while we fixed the values of parameters μ_2 and μ_3 :

$$\mu_2 = 10 \text{ and } \mu_3 = \frac{8}{3} \quad (\text{Lorenz used } \mu_1 = 28 \text{ to get chaos})$$



Lorenz equations: summary of Lecture 8

$$\frac{dx_1}{dt} = 10(x_2 - x_1)$$

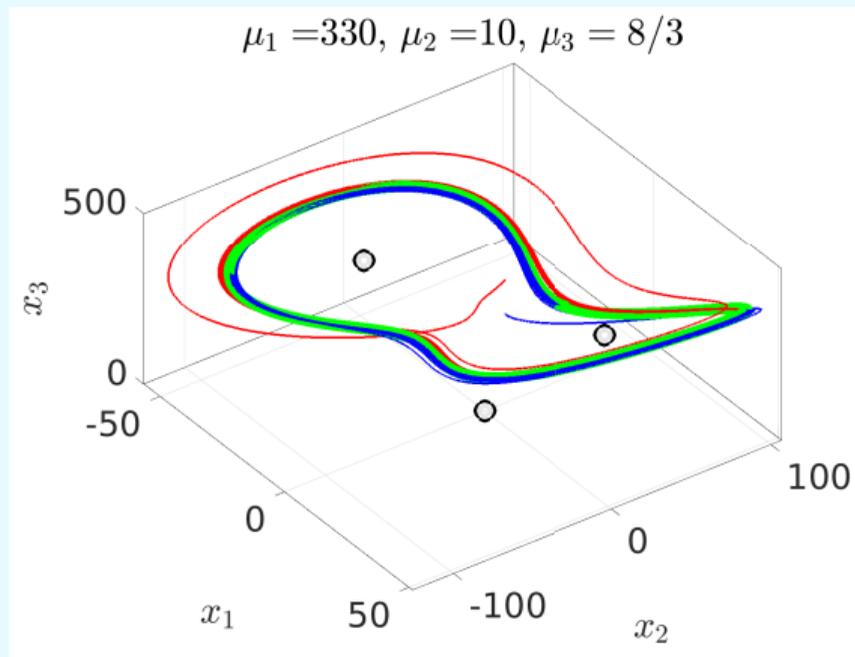
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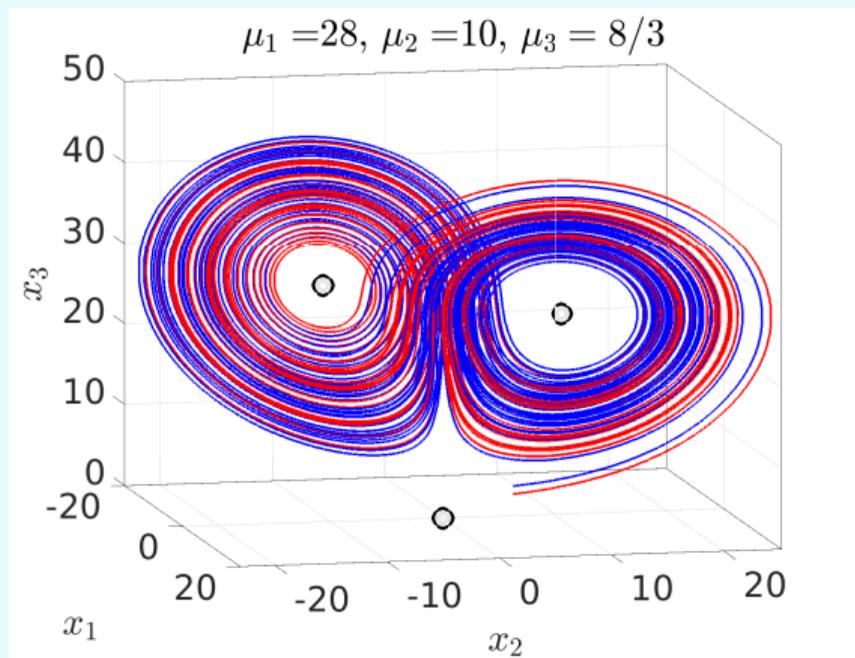
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$$\frac{dx_1}{dt} = \mu_2 (x_2 - x_1)$$

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Problem Sheet 2: we used the Lorenz system to further practice techniques studied in Lectures 1–8



Lorenz equations

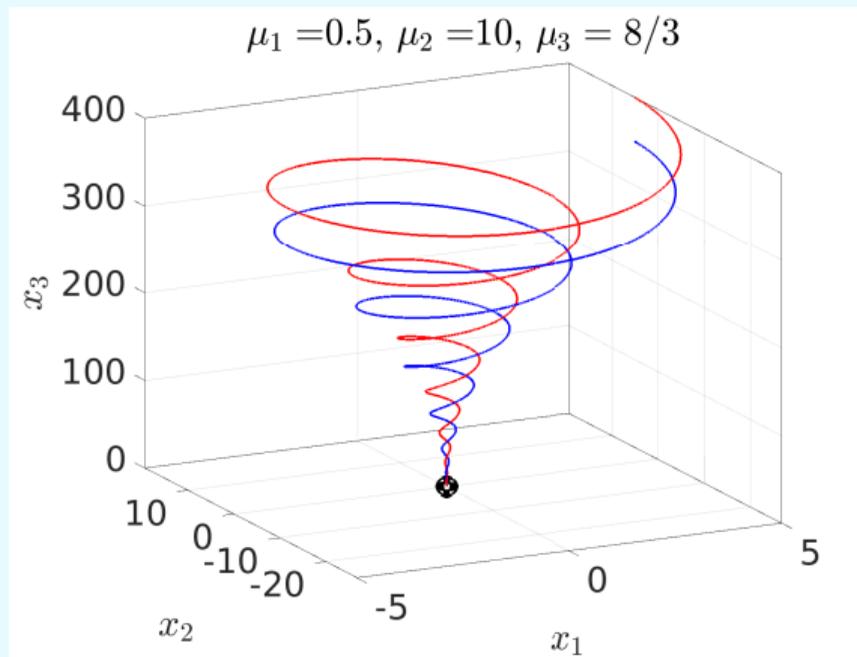
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Problem Sheet 2: we used the Lorenz system to further practice techniques studied in Lectures 1–8 including:

- finding the Lyapunov function to prove the global stability of the fixed point at the origin $\mathbf{0} = [0, 0, 0]$ for $\mu_1 < 1$



Lorenz equations

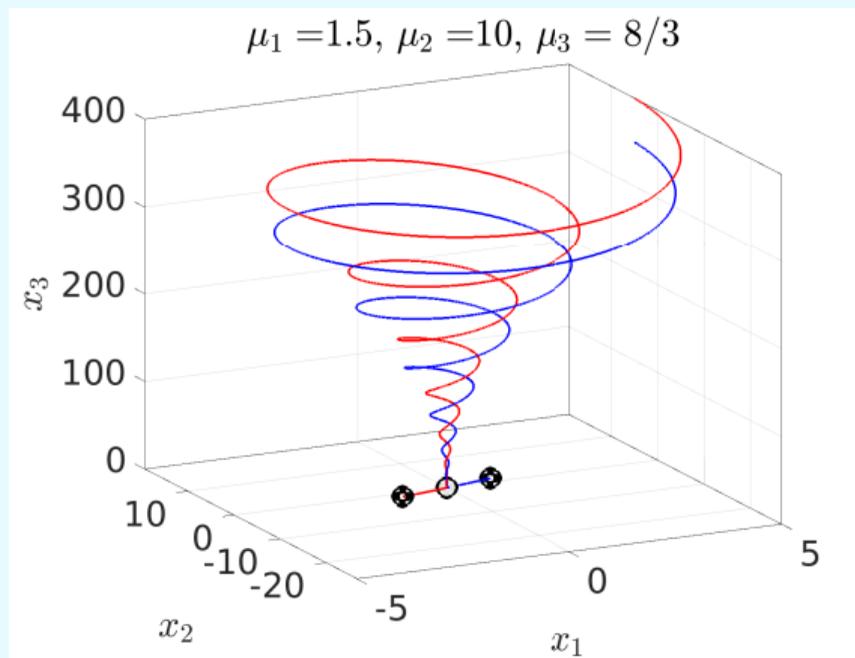
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Problem Sheet 2: we used the Lorenz system to further practice techniques studied in Lectures 1–8 including:

- finding the Lyapunov function to prove the global stability of the fixed point at the origin $\mathbf{0} = [0, 0, 0]$ for $\mu_1 < 1$
- using the extended center manifold theory to analyze the supercritical pitchfork bifurcation at $\mu_1 = 1$, calculating the center manifold and the dynamics on it



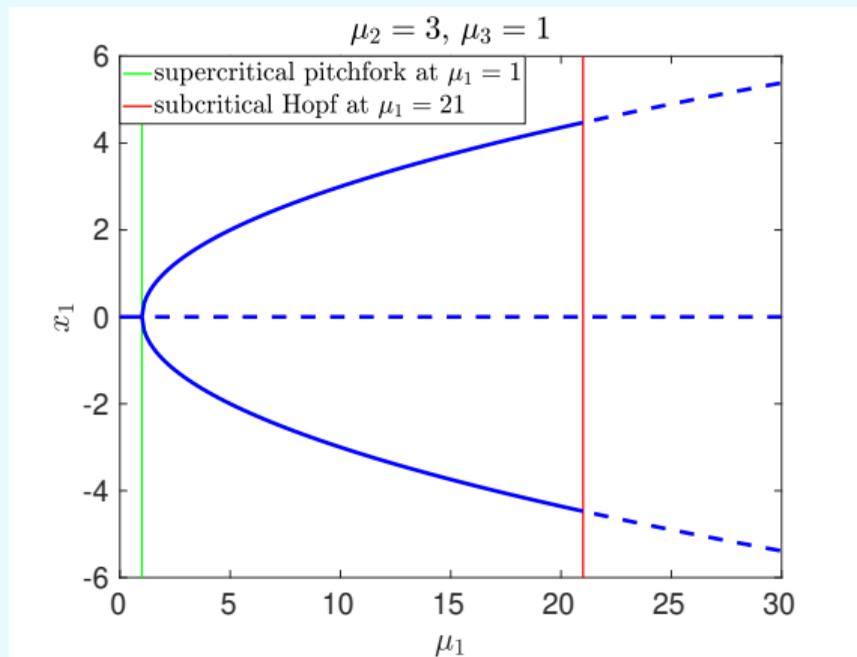
Lorenz equations: Question 6 on Problem Sheet 2

$$\frac{dx_1}{dt} = 3(x_2 - x_1)$$

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- fixed points $\mathbf{x}_{c1} = \mathbf{0} = [0, 0, 0]$
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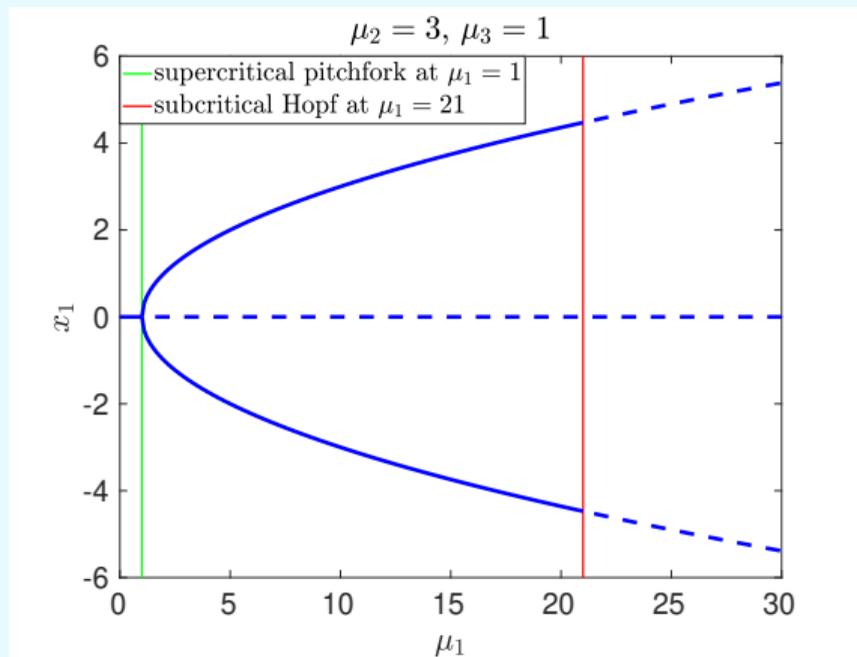
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at $\mu_1 = 1$ ($\mathbf{x}_{c1} = \mathbf{0}$ is stable for $\mu_1 < 1$ and unstable for $\mu_1 > 1$)



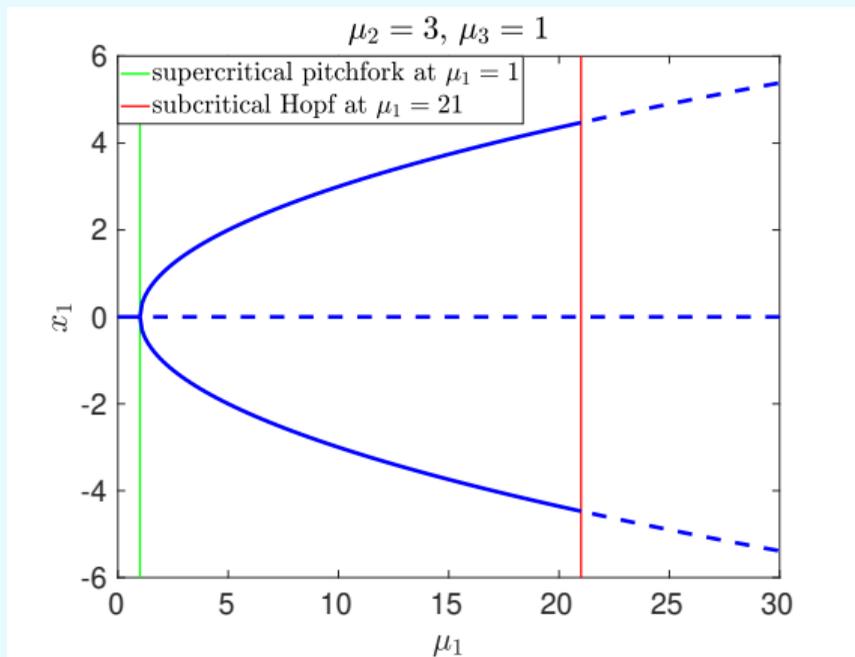
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- supercritical pitchfork bifurcation at $\mu_1 = 1$ ($\mathbf{x}_{c1} = \mathbf{0}$ is stable for $\mu_1 < 1$ and unstable for $\mu_1 > 1$)
- subcritical Hopf bifurcation at $\mu_1 = 21$
 \mathbf{x}_{c2} and \mathbf{x}_{c3} are stable for $\mu_1 < 21$ and unstable for $\mu_1 > 21$



Lorenz equations: general case

$$\frac{dx_1}{dt} = \mu_2 (x_2 - x_1)$$

$$\frac{dx_2}{dt} = \mu_1 x_1 - x_2 - x_1 x_3$$

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- fixed points $\mathbf{x}_{c1} = \mathbf{0} = [0, 0, 0]$

$$\mathbf{x}_{c2} = \left[\sqrt{\mu_3(\mu_1 - 1)}, \sqrt{\mu_3(\mu_1 - 1)}, \mu_1 - 1 \right]$$

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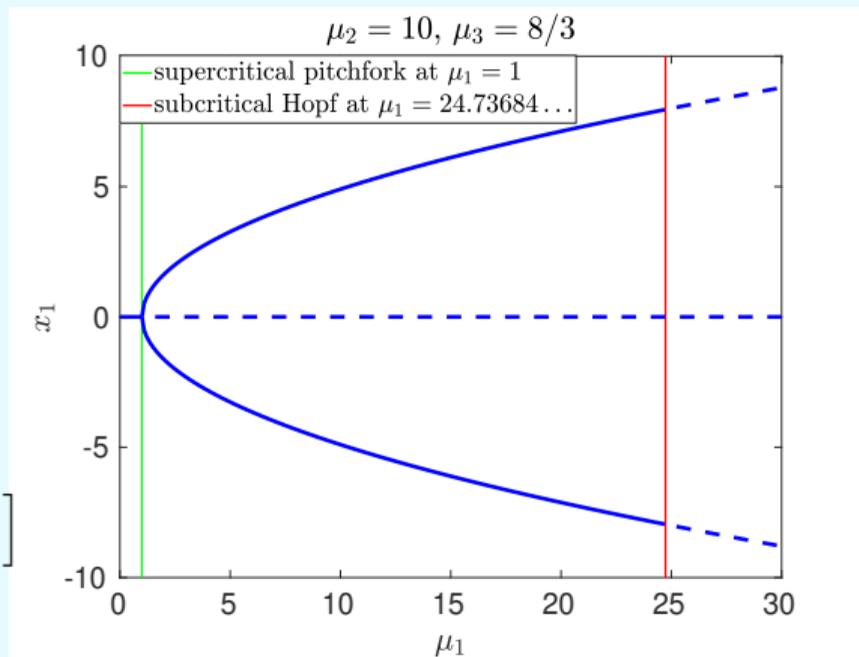
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- supercritical pitchfork bifurcation

at $\mu_1 = 1$ ($\mathbf{x}_{c1} = \mathbf{0}$ is stable for $\mu_1 < 1$ and unstable for $\mu_1 > 1$)

- subcritical Hopf bifurcation at $\mu_1 = \mu_c = \mu_2(\mu_2 + \mu_3 + 3)/(\mu_2 - \mu_3 - 1)$

\mathbf{x}_{c2} and \mathbf{x}_{c3} are stable for $\mu_1 < \mu_c$ and unstable for $\mu_1 > \mu_c$



Lorenz equations: trapping region

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$$\frac{dx_2}{dt} = \mu_1 x_1 - x_2 - x_1 x_3$$

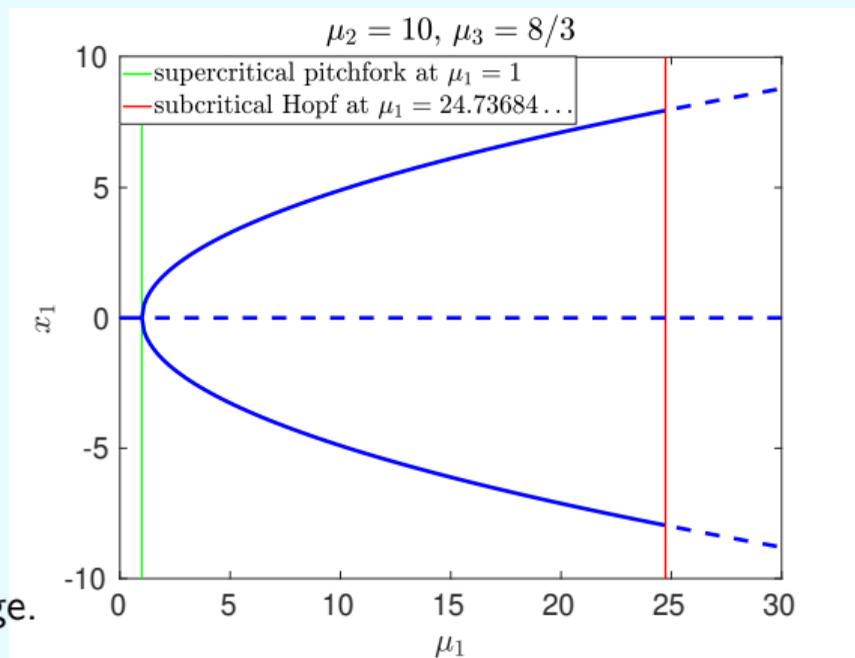
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Questions 6 on Problem Sheet 2:

All trajectories eventually enter and remain inside a large sphere of the form

$$x_1^2 + x_2^2 + (x_3 - \mu_1 - 3)^2 = C(\mu_1)$$

where constant $C(\mu_1)$ is sufficiently large.



Lorenz equations: trapping region and volume contraction

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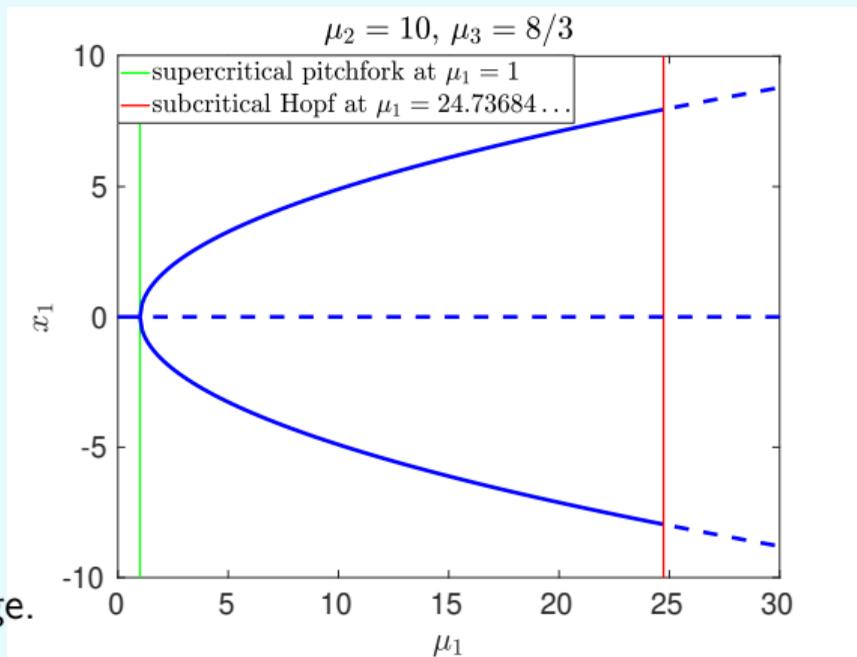
$$x_1^2 + x_2^2 + (x_3 - \mu_1 - 3)^2 = C(\mu_1)$$

where constant $C(\mu_1)$ is sufficiently large.

Let $U \equiv U(0) \subset \mathbb{R}^3$ be a compact connected subset of initial conditions.

Let $U(t) = \phi_t(U)$ and $v(t) = |U(t)| = |\phi_t(U)|$ be the volume of $U(t)$. Then

$$\lim_{t \rightarrow \infty} v(t) = 0$$



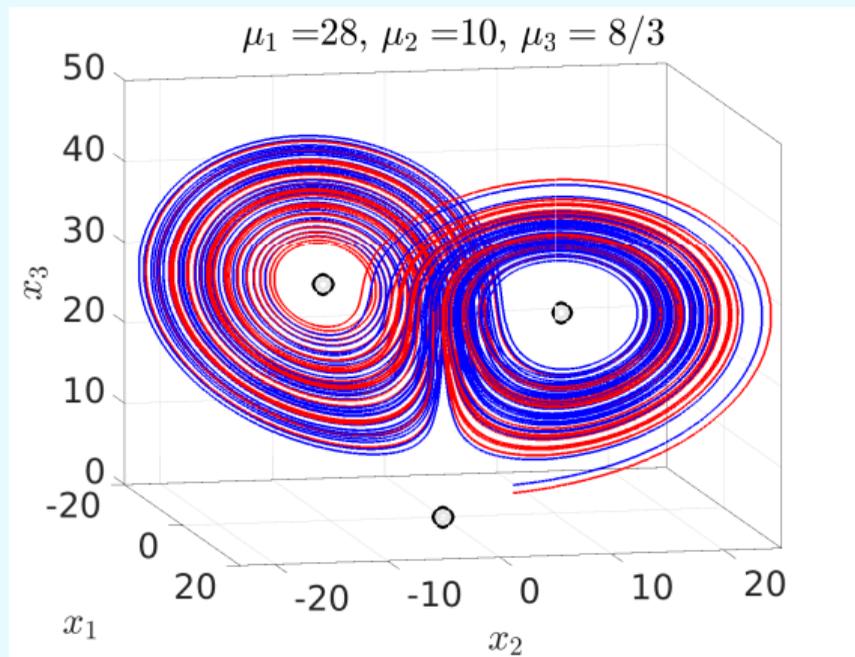
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Lorenz map: we investigate chaos using a discrete-time dynamical system



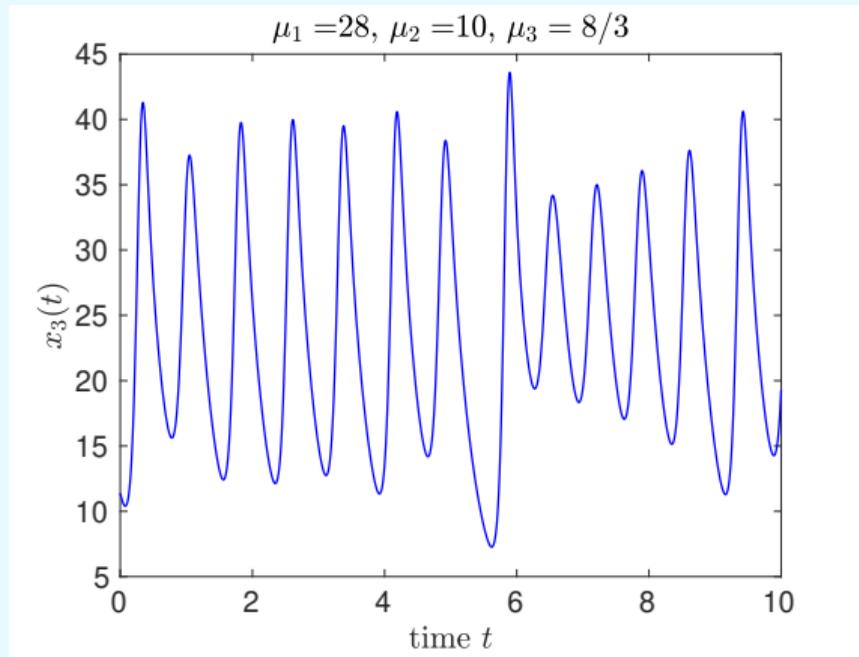
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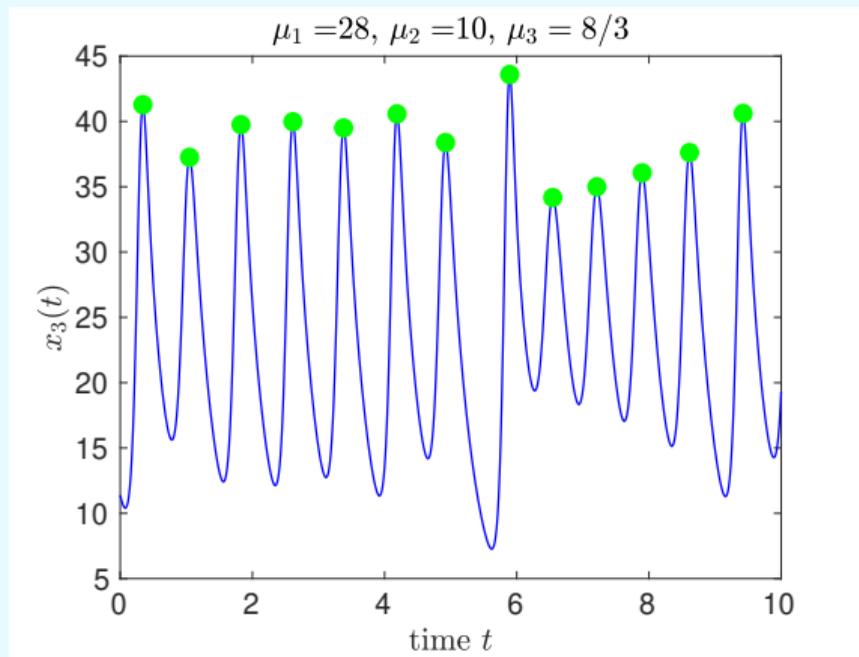
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Lorenz map: we investigate chaos using a discrete-time dynamical system

Consider local maxima z_n , $n = 1, 2, \dots$ of $x_3(t)$ and define Lorenz map by:

$$z_{n+1} = F(z_n)$$

Then a closed orbit corresponds to an N -cycle $\{z_0, z_1, z_2, \dots, z_{N-1}\}$ of the Lorenz map.



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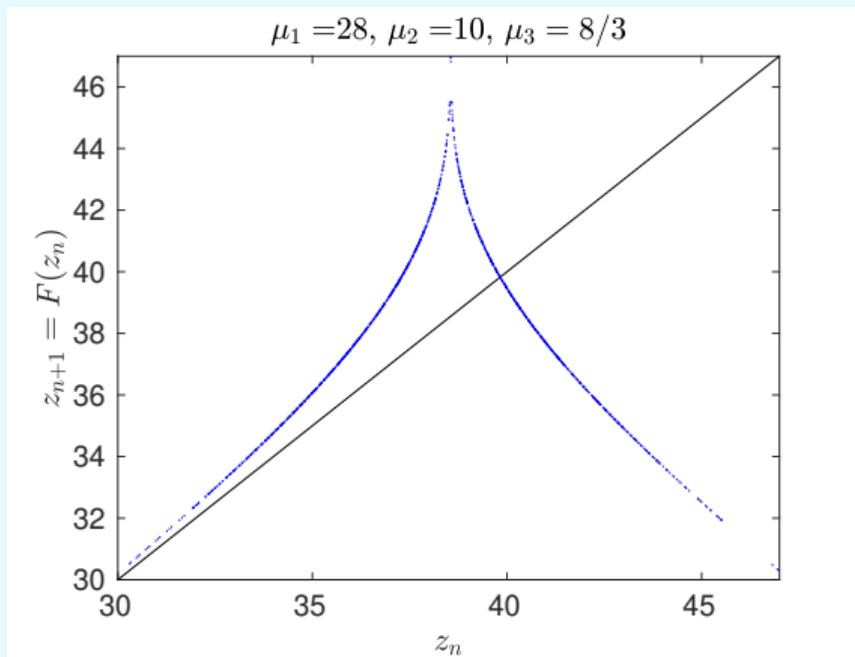
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Lecture 7: N -cycle is *unstable* if $|F'(z_0) F'(z_1) \dots F'(z_{N-1})| > 1$

There are no stable fixed points or limit cycles for: $\mu > \mu_c = \frac{\mu_2(\mu_2 + \mu_3 + 3)}{\mu_2 - \mu_3 - 1}$



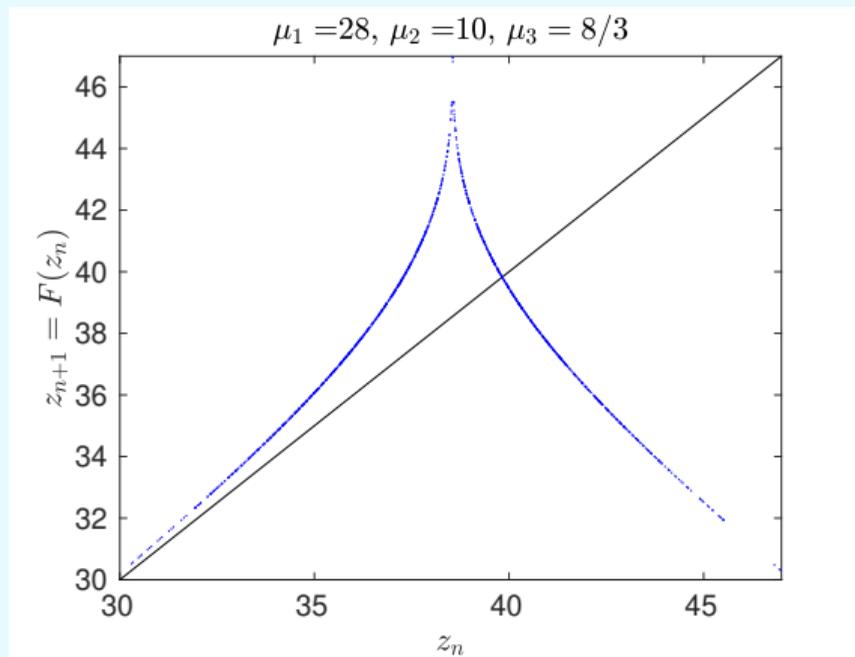
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Lorenz map: we investigate chaos using a discrete-time dynamical system



NEXT LECTURE:

Poincaré map: we investigate ODEs using a discrete-time dynamical system

B5.6: Nonlinear Dynamics, Bifurcations and Chaos (Lecture 14)

- summary of Lecture 13: we discussed Hilbert's 16th problem. Oscillations in chemical reaction networks. Lorenz equations. Lorenz map. [Questions 2, 6 and 7 on Problem Sheet 4](#).
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- course synopsis of **Lectures 9-16**:
Bifurcations of limit cycles, covering supercritical and subcritical Hopf bifurcations, saddle-node bifurcation of cycles, infinite-period (SNIC) bifurcation and homoclinic (saddle-loop) bifurcation. Oscillations in chemical reaction networks. Weakly nonlinear oscillators. Poincaré-Lindstedt method. Conservative and non-conservative systems. Liénard systems, van der Pol oscillator. [Hilbert's 16th problem](#). [Lorenz equations](#). [Lorenz map](#). [Poincaré section](#). [Poincaré map](#). [Converse of Sharkovsky's theorem](#). [Bernoulli shift map](#), symbolic dynamics. [Tent map](#). [Dynamics on metric spaces](#), sensitive dependence on initial conditions, transitivity, conjugate maps, chaotic dynamics.

Poincaré section and Poincaré map

consider an illustrative example:

$$\frac{dx_1}{dt} = -x_2 + \frac{x_1}{10} (1 - x_1^2 - x_2^2)$$

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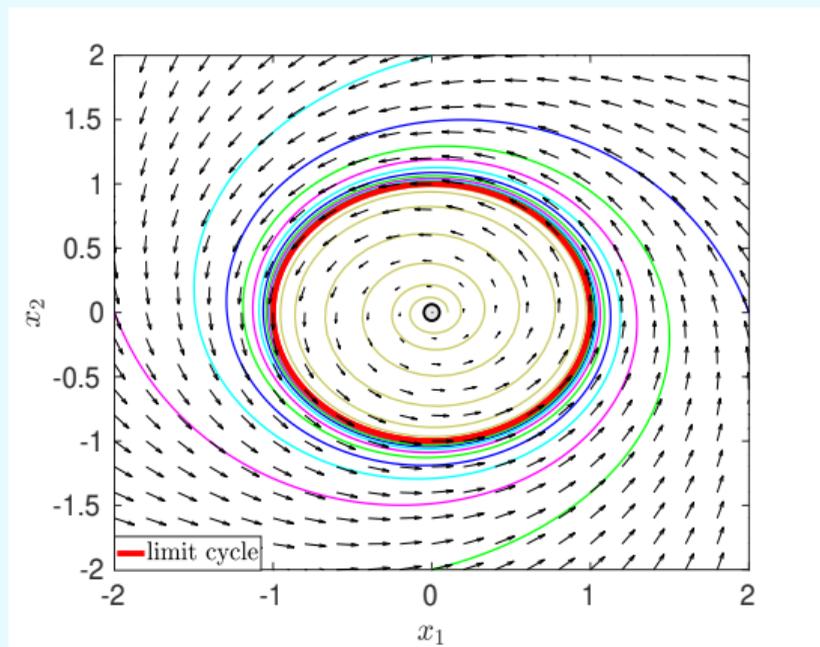
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transform to polar coordinates to get:

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\implies stable limit cycle at $r = 1$



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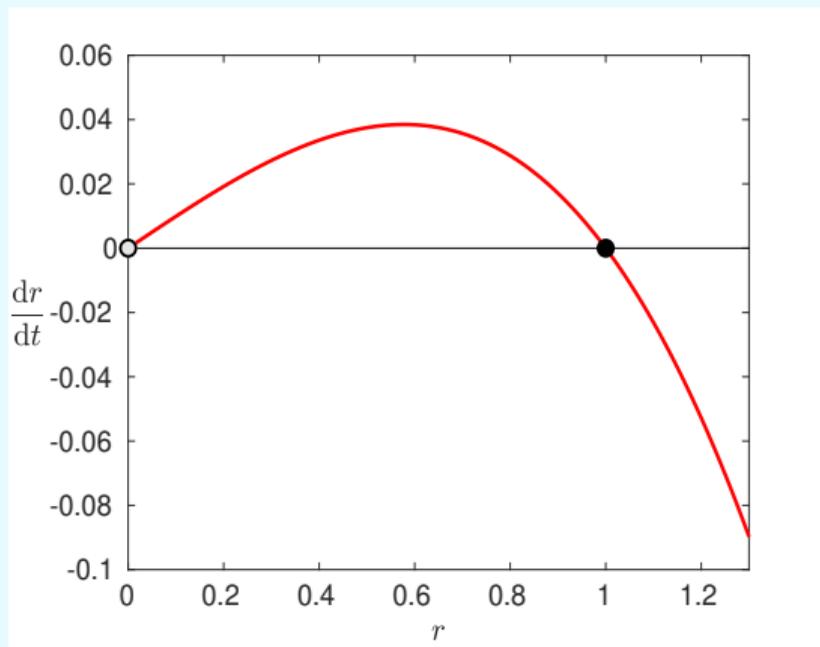
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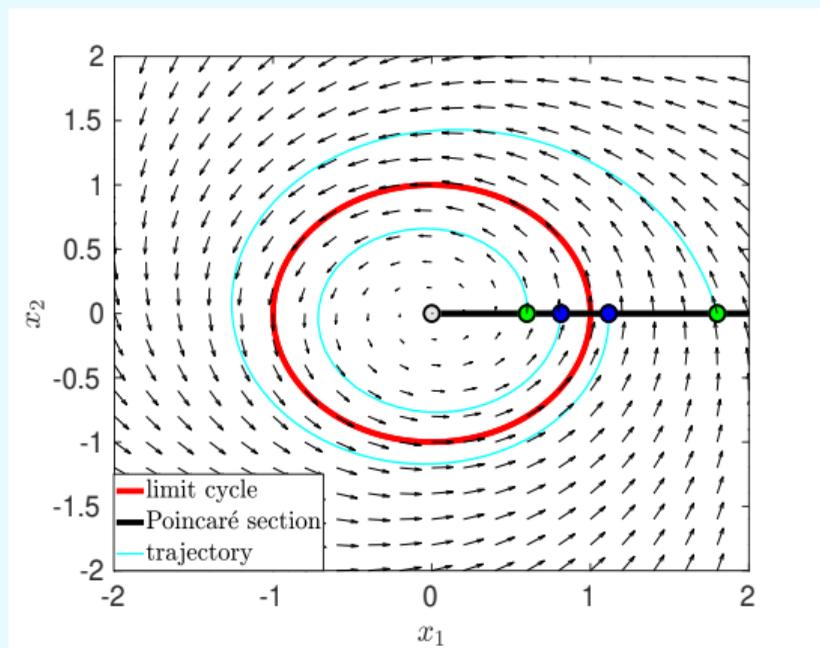
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$$\Sigma = \{[x_1, 0] \in \mathbb{R}^2 \mid x_1 > 0\}$$

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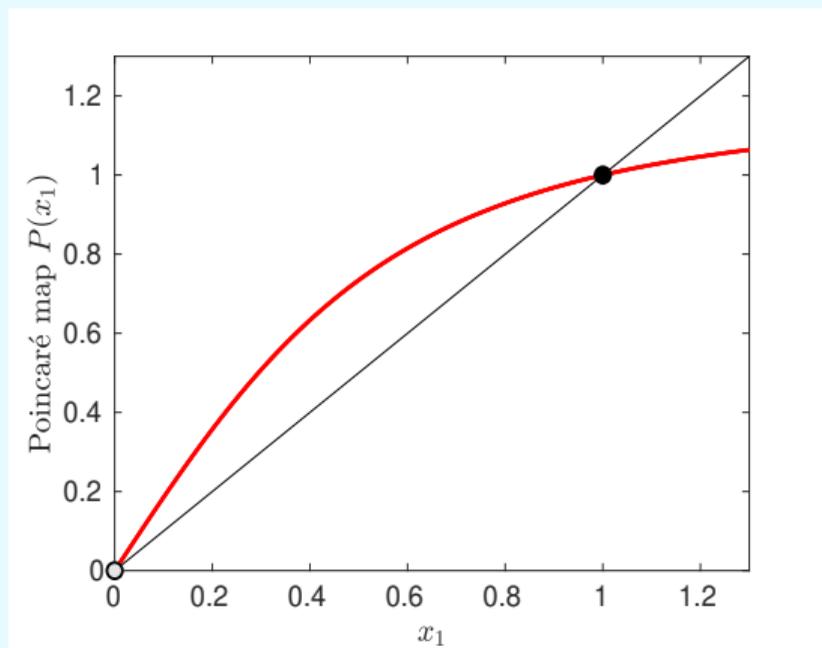
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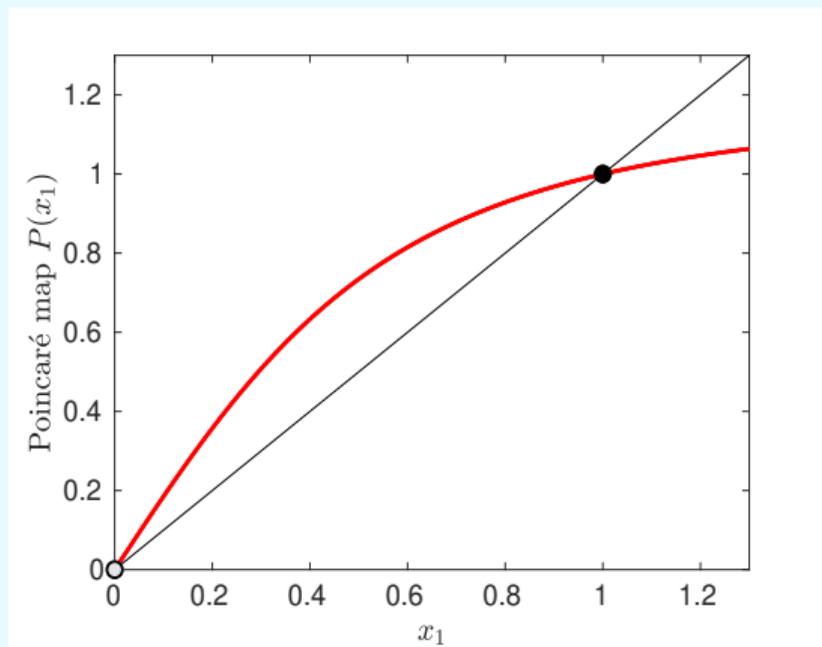
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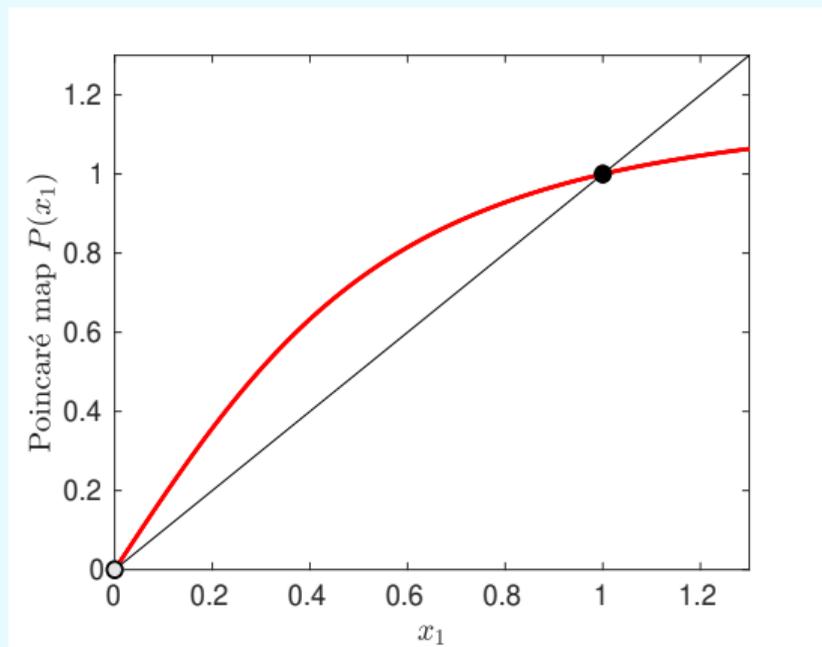
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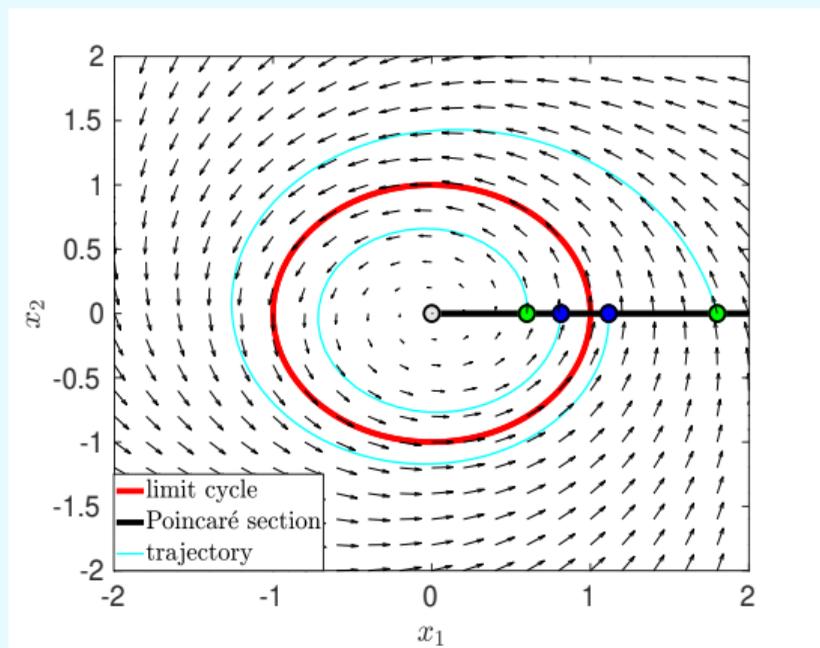
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Poincaré map : saddle-node bifurcation of cycles example

saddle-node bifurcation of cycles at $\mu = -1/4$: a half-stable cycle appears, it splits into a pair of limit cycles for $\mu > -1/4$, one stable, one unstable

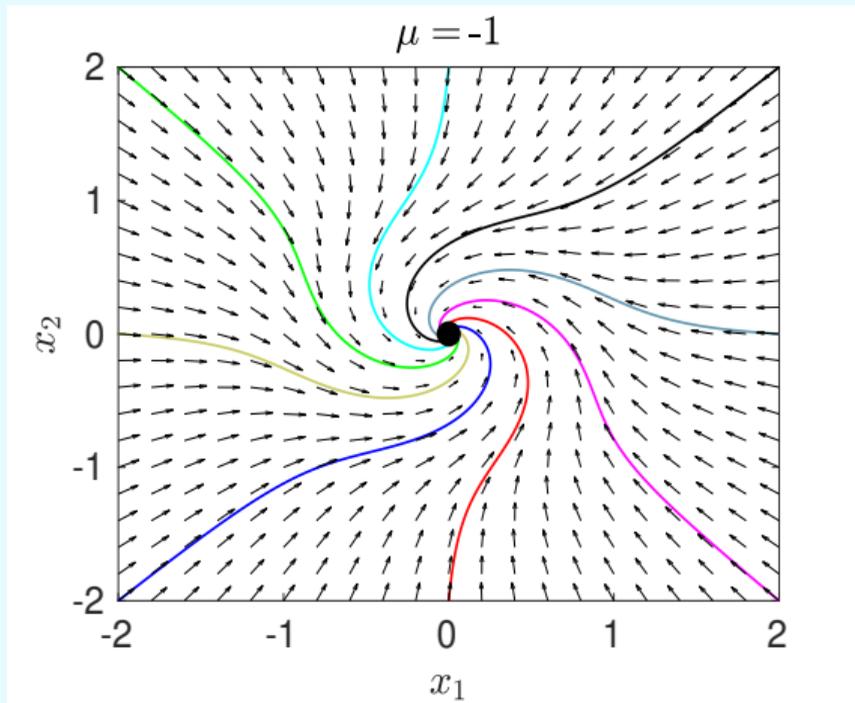
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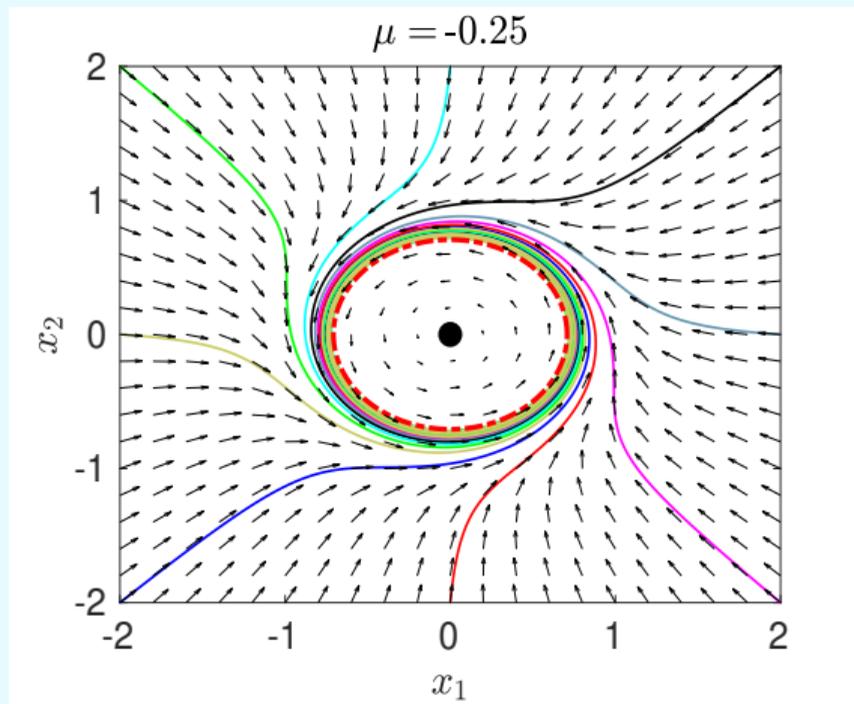


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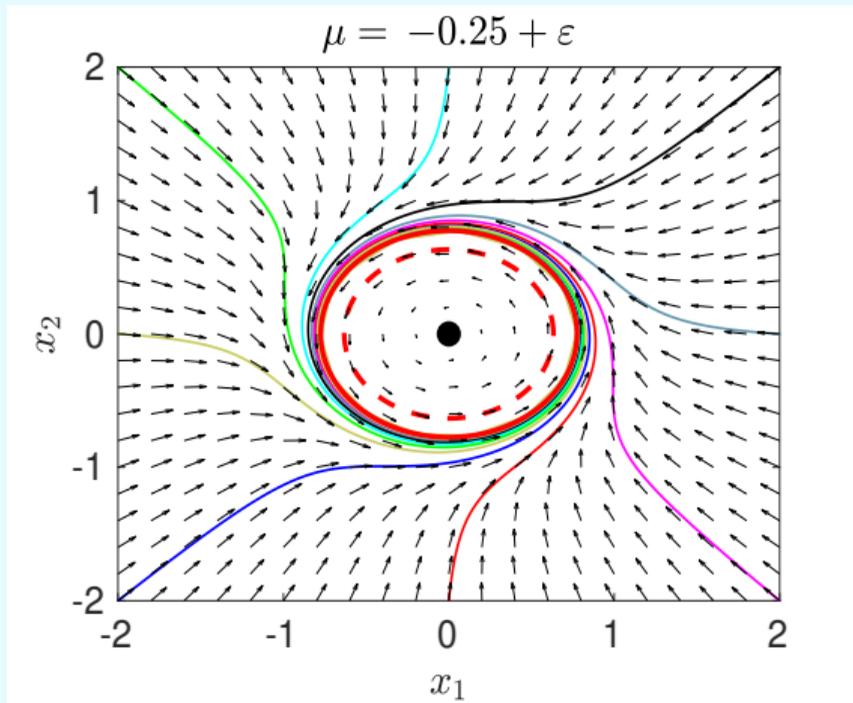


Poincaré map : saddle-node bifurcation of cycles example

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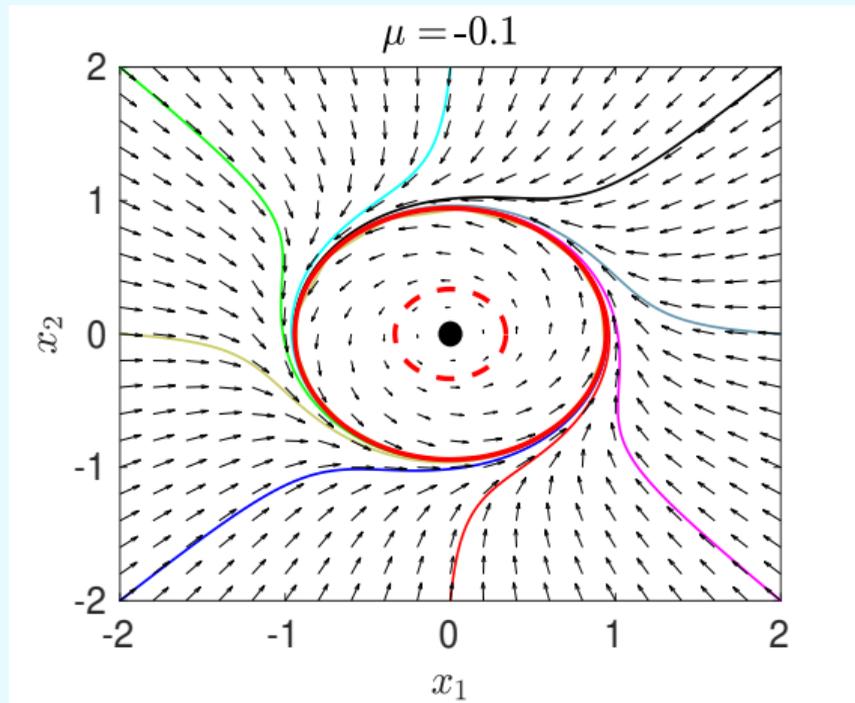
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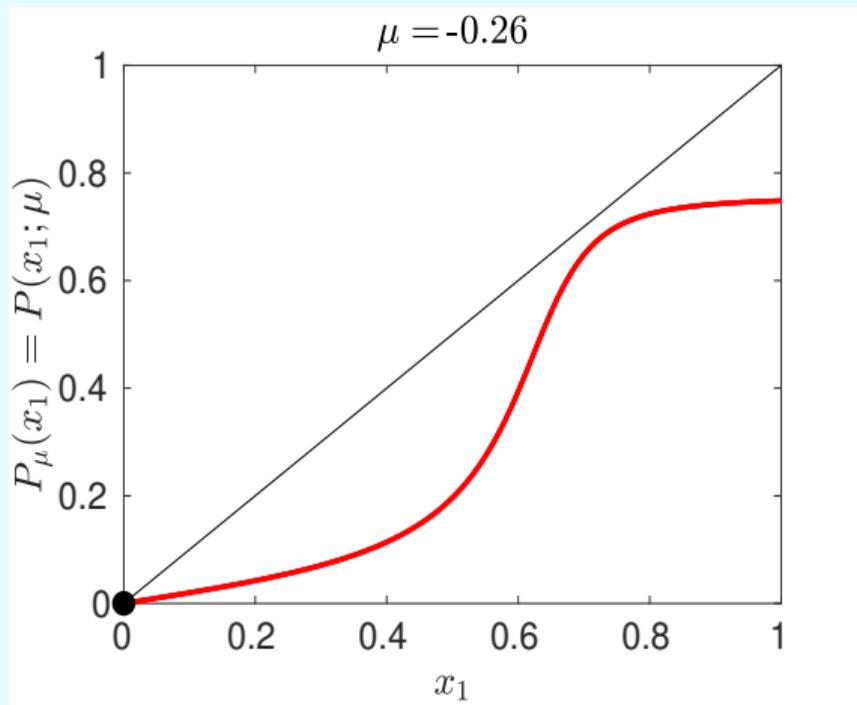
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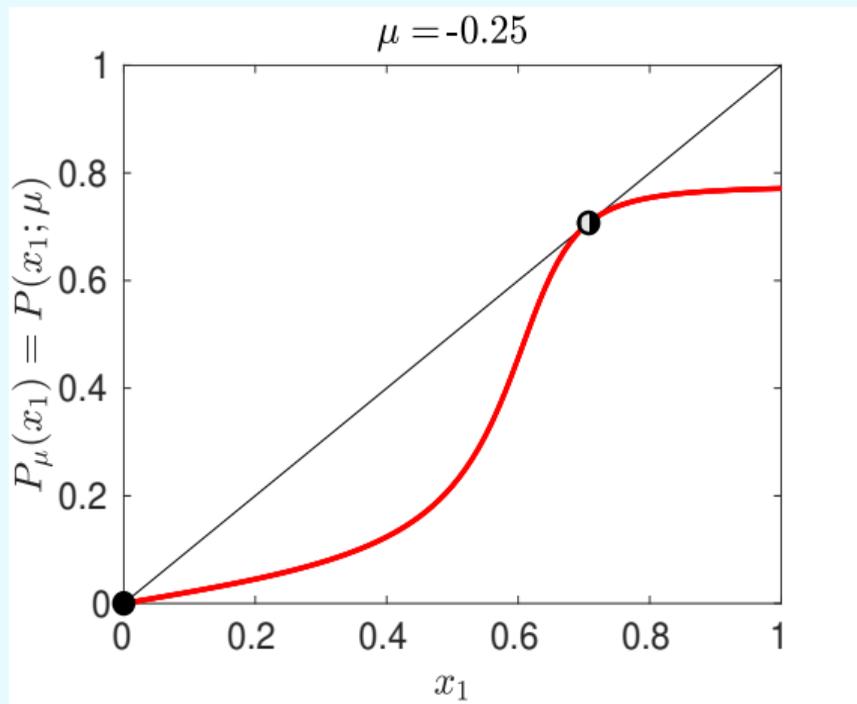
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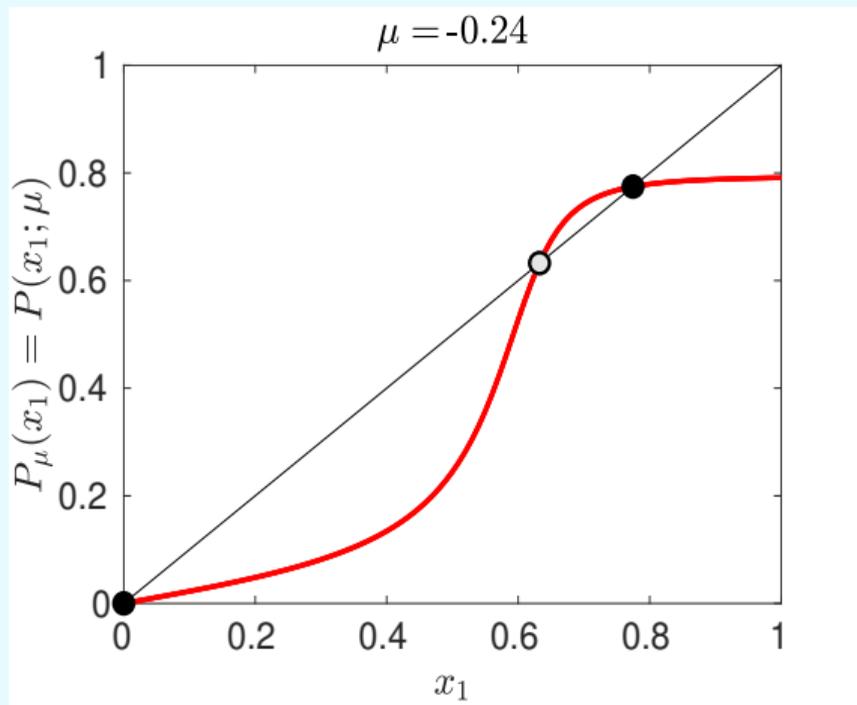
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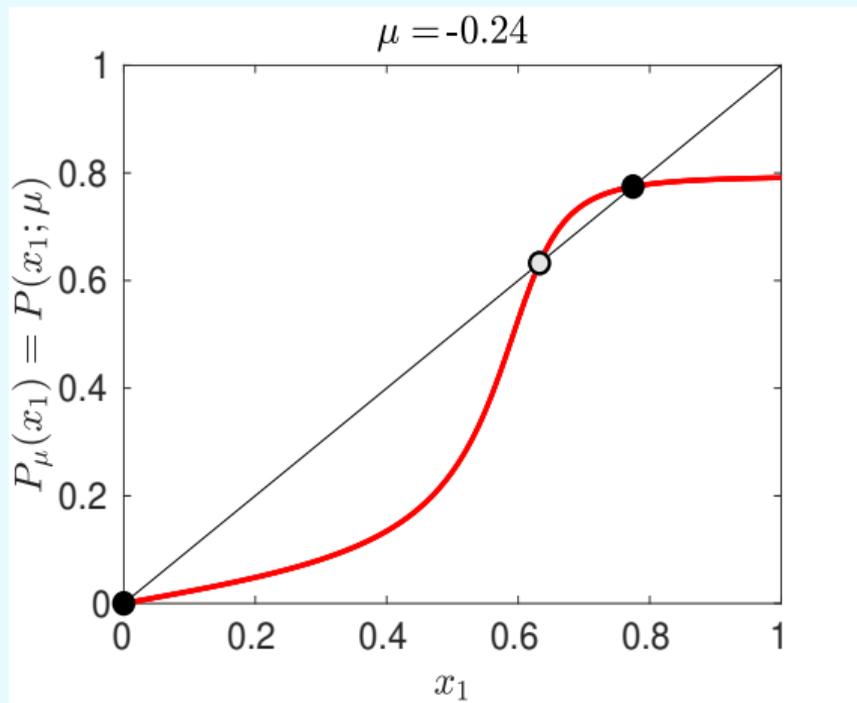
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Another example: [Question 1 on Problem Sheet 4](#)

Summary of our whiteboard derivations

- we discussed chaos, symbolic dynamics and the Bernoulli shift map
- we studied dynamical systems associated with function $F : \mathbb{M} \rightarrow \mathbb{M}$, where \mathbb{M} is a metric space, *i.e.* a set with metric (distance) $d : \mathbb{M} \times \mathbb{M} \rightarrow [0, \infty)$

Summary – general theory

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- $F : \mathbb{M} \rightarrow \mathbb{M}$ has **sensitive dependence on initial conditions** if there exists $\delta > 0$ such that for all $x \in \mathbb{M}$ and any open set $U \subset \mathbb{M}$ satisfying $x \in U$, there is a point $y \in U$ and $j \in \mathbb{N}$ with $d(F^{(j)}(x), F^{(j)}(y)) > \delta$
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 - (i) the set of all periodic points is dense in \mathbb{M}
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- if $F : \mathbb{M} \rightarrow \mathbb{M}$ is continuous and \mathbb{M} is not a finite set, then (i) and (ii) imply (iii)

Summary – Bernoulli shift map

$$\mathbb{M}_{01} = \{(a_1, a_2, a_3, a_4, \dots) \mid \text{such that } a_j = 0 \text{ or } a_j = 1 \text{ for } j = 1, 2, 3, 4, \dots\}$$

\mathbb{M}_{01} is a metric space with metric defined by

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Bernoulli shift map: $\sigma : \mathbb{M}_{01} \rightarrow \mathbb{M}_{01}$ where $\sigma((a_1, a_2, a_3, a_4, \dots)) = (a_2, a_3, a_4, a_5, \dots)$

we stated and proved some of properties of the shift map, namely:

- fixed points are $(0, 0, 0, 0, \dots)$ and $(1, 1, 1, 1, \dots)$
2-cycle is $\{(0, 1, 0, 1, 0, 1, 0, 1, \dots), (1, 0, 1, 0, 1, 0, 1, 0, \dots)\}$
- shift map $\sigma : \mathbb{M}_{01} \rightarrow \mathbb{M}_{01}$ is **continuous**
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NEXT LECTURE: we will prove that shift map $\sigma : \mathbb{M}_{01} \rightarrow \mathbb{M}_{01}$ is **chaotic**, *i.e.*

- (i) the set of all periodic points is dense in \mathbb{M}_{01}
- (ii) σ is transitive
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B5.6: Nonlinear Dynamics, Bifurcations and Chaos (Lecture 15)

- summary of Lecture 14: we discussed Poincaré section. Poincaré map. Bernoulli shift map, symbolic dynamics. Dynamics on metric spaces, sensitive dependence on initial conditions, transitivity, chaotic dynamics. [Questions 1, 3, 4 and 5 on Problem Sheet 4.](#)
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- course synopsis of **Lectures 9-16:**
Bifurcations of limit cycles, covering supercritical and subcritical Hopf bifurcations, saddle-node bifurcation of cycles, infinite-period (SNIC) bifurcation and homoclinic (saddle-loop) bifurcation. Oscillations in chemical reaction networks. Weakly nonlinear oscillators. Poincaré-Lindstedt method. Conservative and non-conservative systems. Liénard systems, van der Pol oscillator. [Hilbert's 16th problem.](#) [Lorenz equations.](#) [Lorenz map.](#) [Poincaré section.](#) [Poincaré map.](#) [Converse of Sharkovsky's theorem.](#) [Bernoulli shift map,](#) [symbolic dynamics.](#) [Tent map.](#) [Dynamics on metric spaces,](#) [sensitive dependence on initial conditions,](#) [transitivity,](#) [conjugate maps,](#) [chaotic dynamics.](#)

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Lemma 1: If $x = (a_1, a_2, a_3, a_4, \dots) \in \mathbb{M}_{01}$ and $y = (b_1, b_2, b_3, b_4, \dots) \in \mathbb{M}_{01}$ with $a_1 = b_1, a_2 = b_2, \dots, a_n = b_n$. Then $d(x, y) \leq 2^{-n}$.

Lemma 2: If $d(x, y) < 2^{-n}$. Then $a_1 = b_1, a_2 = b_2, \dots, a_n = b_n$.

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Lemma 3: The Bernoulli shift map $\sigma : \mathbb{M}_{01} \rightarrow \mathbb{M}_{01}$ is **continuous**.

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Remark: we could obtain the same properties if we worked with the metric space of bi-infinite sequences of 0's and 1's, *i.e.* where

$x = (\dots, a_{-j}, \dots, a_{-2}, a_{-1} \mid a_0, a_1, a_2, \dots, a_j, \dots)$ and
 $y = (\dots, b_{-j}, \dots, b_{-2}, b_{-1} \mid b_0, b_1, b_2, \dots, b_j, \dots)$ have distance $d(x, y) = \sum_{j=-\infty}^{\infty} \frac{|a_j - b_j|}{2^{|j|}}$

Question 3 on Problem Sheet 4

Let $x_0 \in [0, 1)$ and $F : [0, 1) \rightarrow [0, 1)$. Define sequence $x_k \in [0, 1)$, $k = 0, 1, 2, \dots$, iteratively by $x_{k+1} = F(x_k)$, where

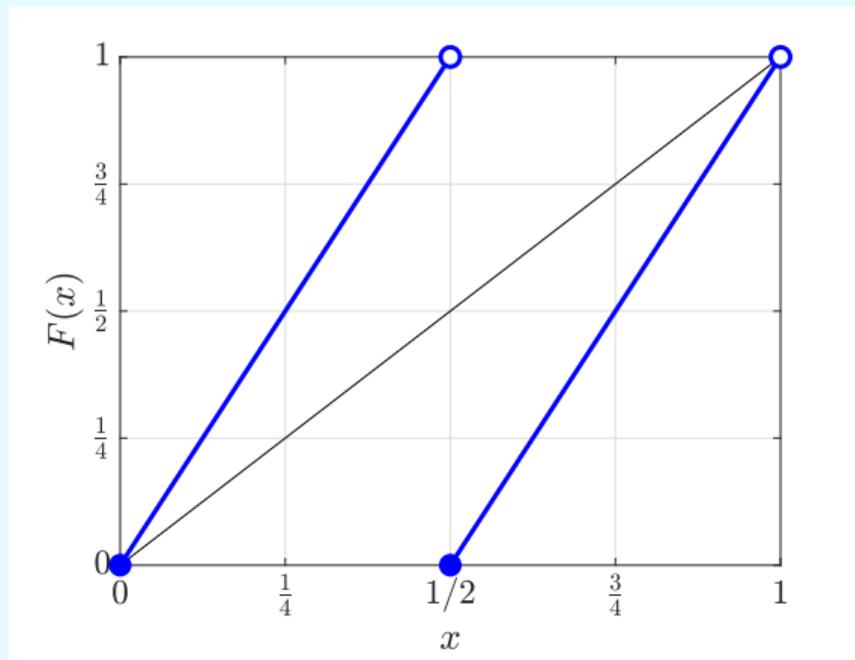
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$\mathbb{M} = [0, 1)$ with $d(x, y) = |x - y|$



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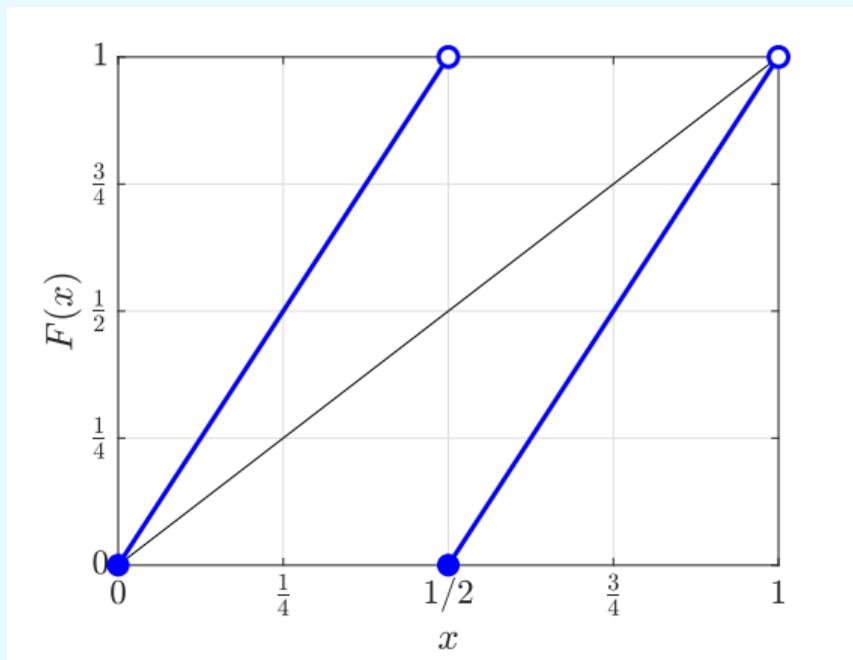
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If $x_0 \in [0, 1/2)$ has a binary expansion

$$x_0 = 0.0a_2a_3a_4 \dots = \sum_{j=2}^{\infty} \frac{a_j}{2^j}$$

where $a_j \in \{0, 1\}$ for $j = 2, 3, 4, \dots$,
then $2x_0 = 0.a_2a_3a_4a_5 \dots$



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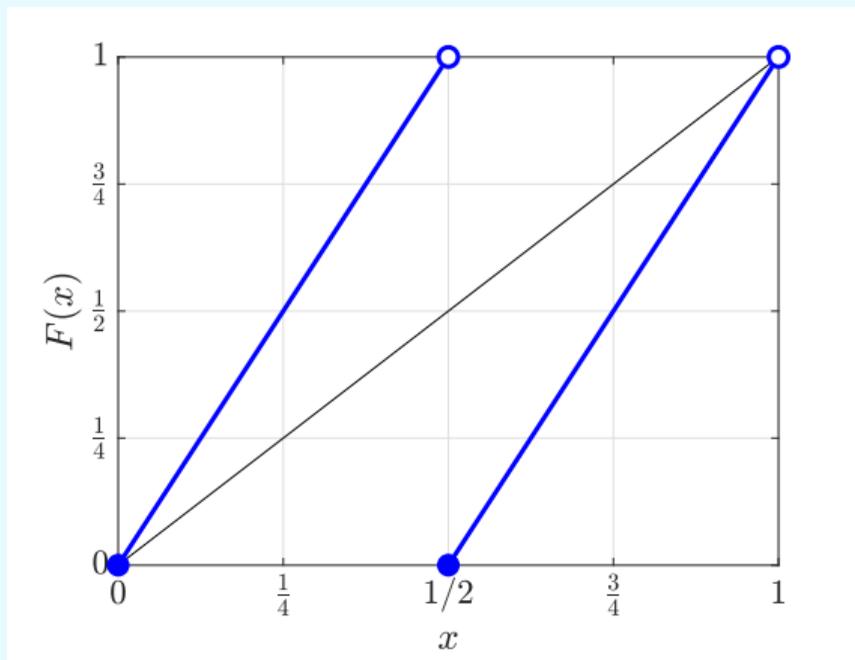
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Q3(a): if $x_0 \in [0, 1)$ is not a dyadic rational, then $F^{(k)}(x) = 0.a_{k+1}a_{k+2}a_{k+3}a_{k+4} \dots$

F satisfies properties (i)–(iii) in our definition of chaotic maps $\implies F$ is chaotic

Question 5 on Problem Sheet 4

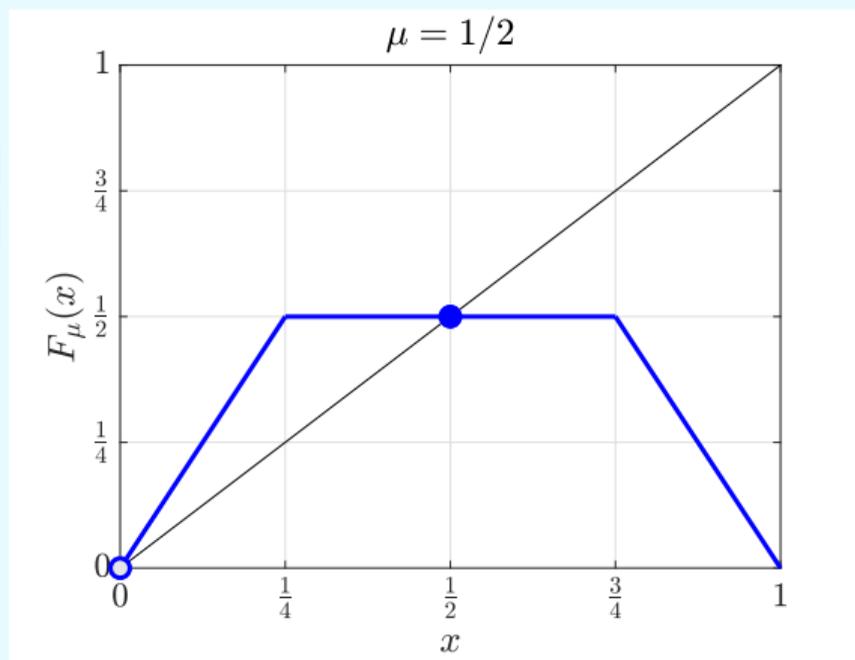
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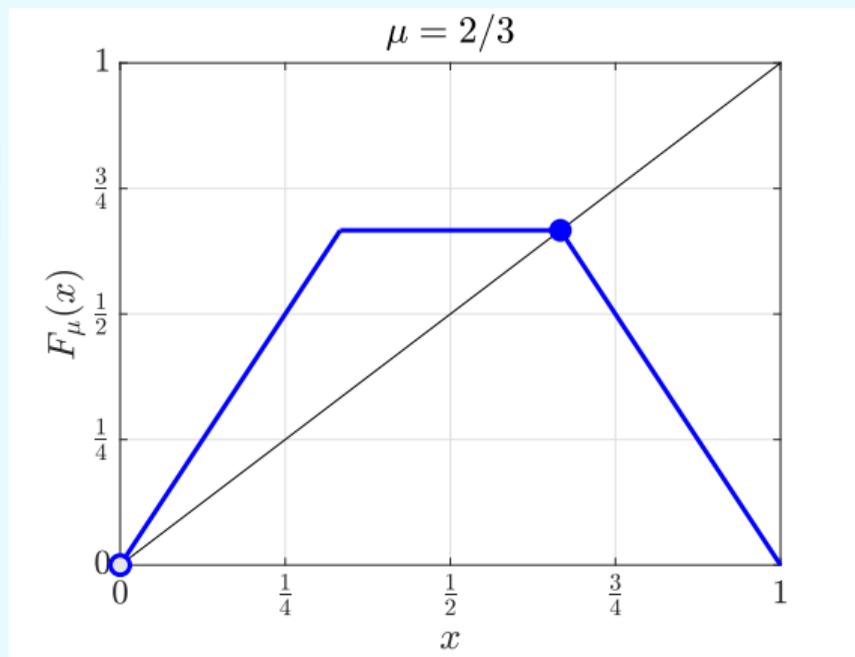
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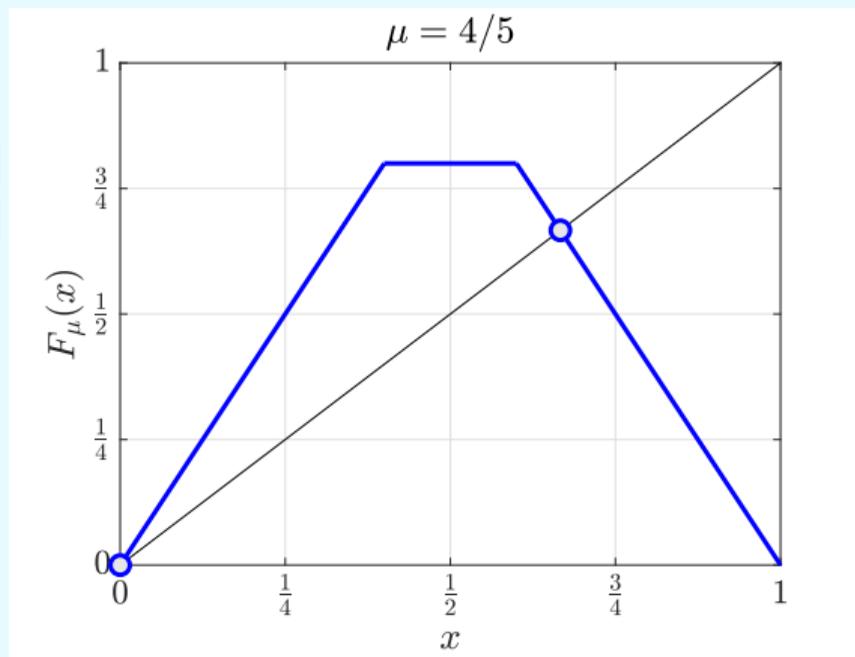
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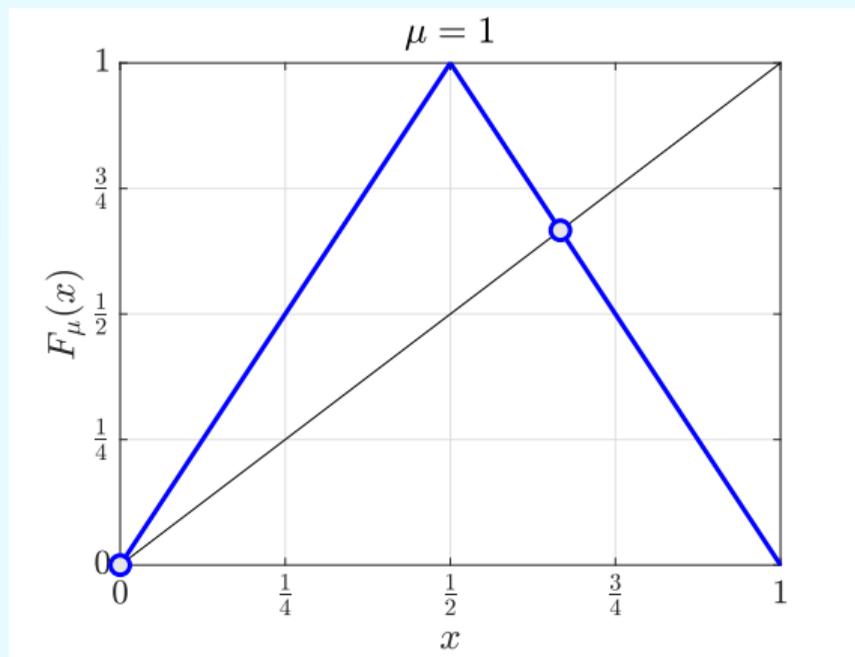
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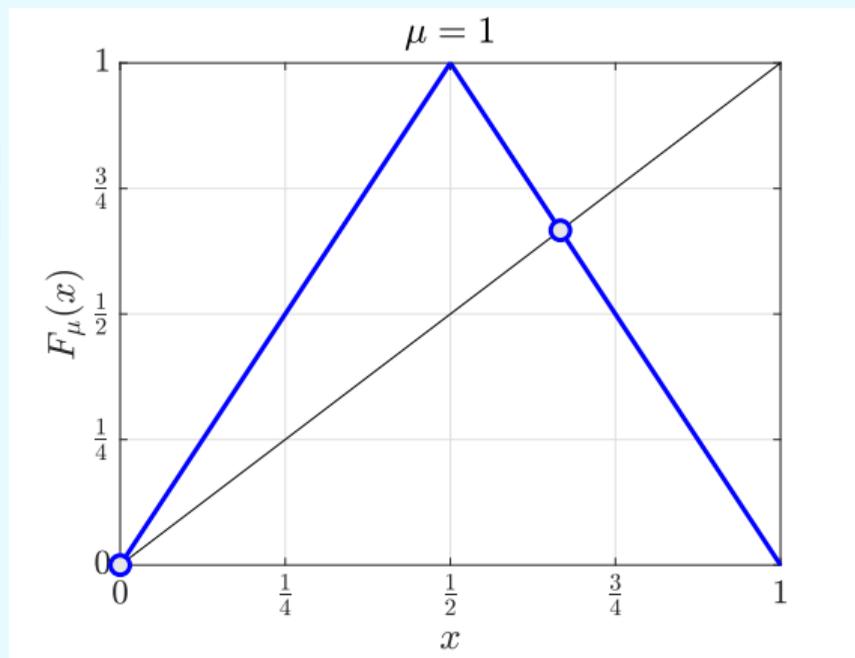
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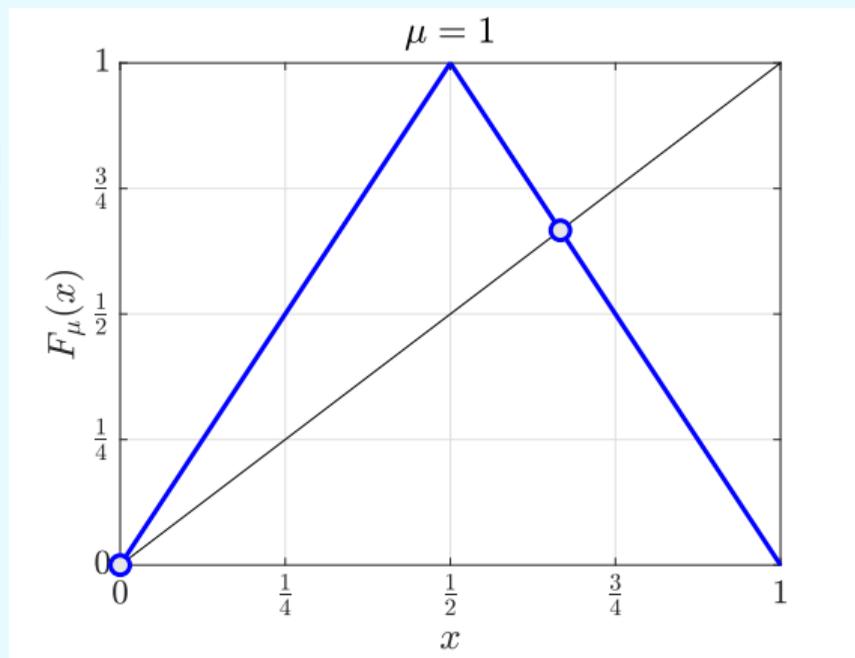
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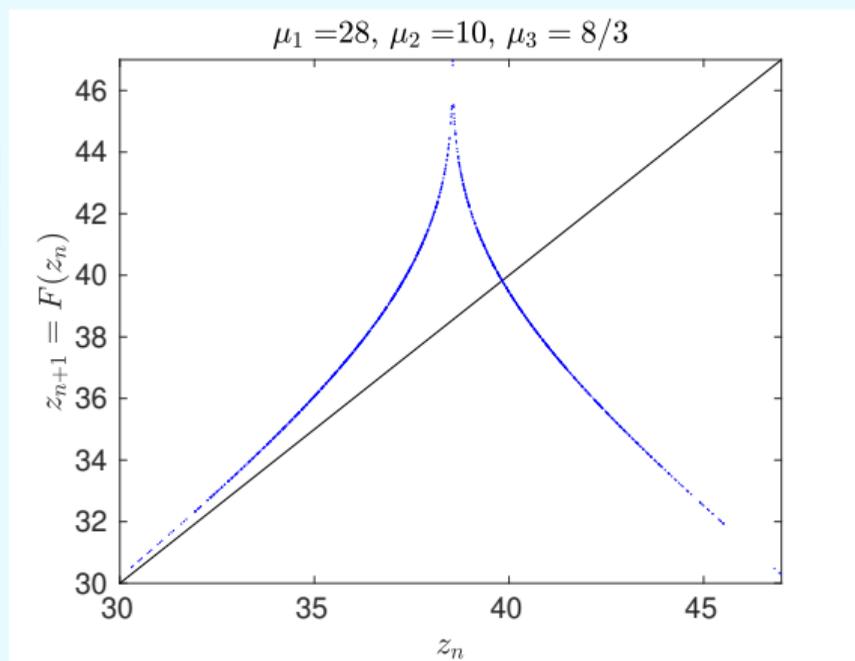
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some maps look 'similar' to the tent map F_1 :



Lorenz map

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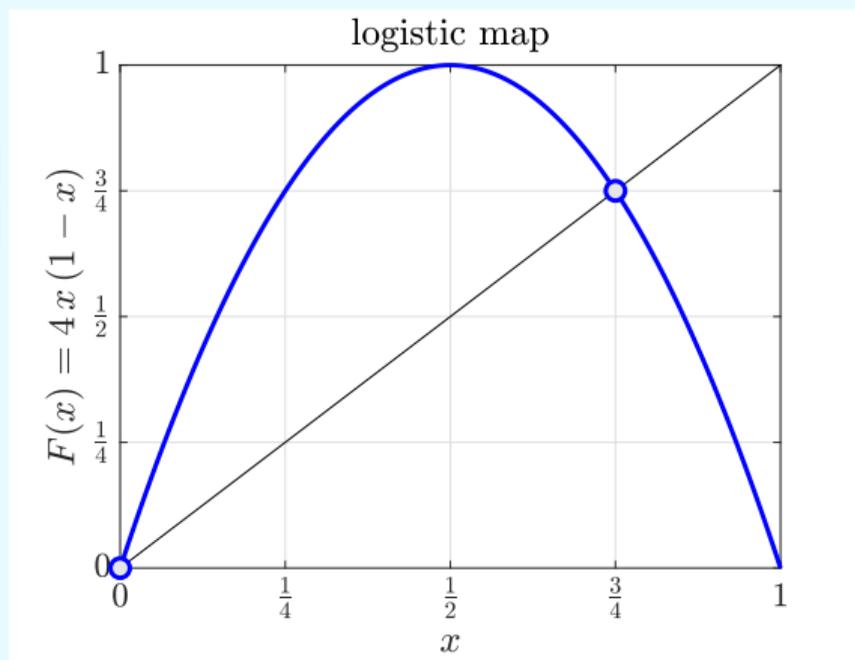
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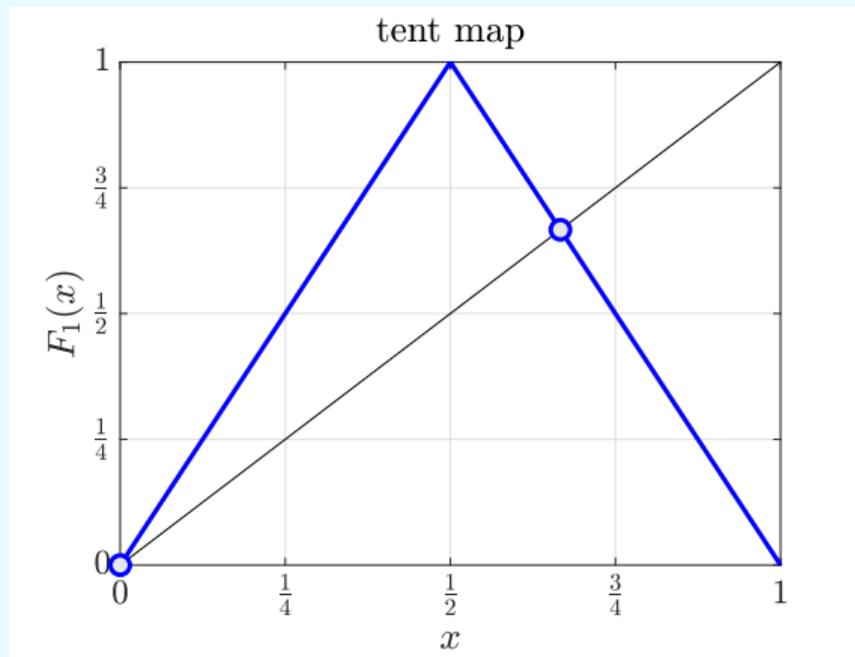
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some maps look 'similar' to the tent map F_1 and **tent map $F_1(x)$ is chaotic**



Question 5 on Problem Sheet 4

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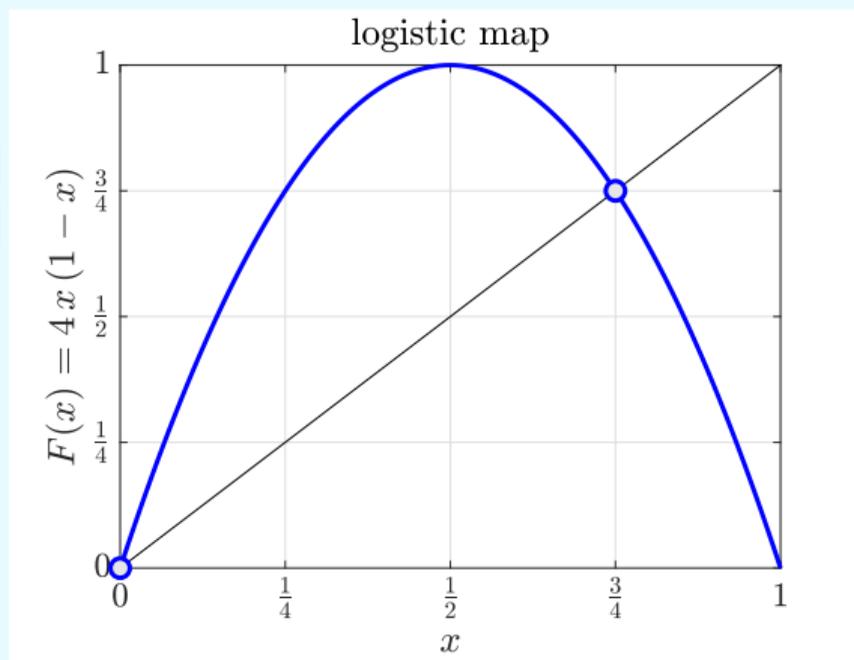
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NEXT LECTURE: we will show that maps 'similar' to chaotic maps are chaotic (in particular, we will show that the logistic map $F(x) = 4x(1-x)$ is chaotic)



B5.6: Nonlinear Dynamics, Bifurcations and Chaos (Lecture 16)

- summary of Lecture 15: we discussed Bernoulli shift map, symbolic dynamics. Tent map. Dynamics on metric spaces, sensitive dependence on initial conditions, transitivity, chaotic dynamics.
[Questions 3, 4 and 5 on Problem Sheet 4](#)
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- course synopsis of **Lectures 9-16**:
Bifurcations of limit cycles, covering supercritical and subcritical Hopf bifurcations, saddle-node bifurcation of cycles, infinite-period (SNIC) bifurcation and homoclinic (saddle-loop) bifurcation. Oscillations in chemical reaction networks. Weakly nonlinear oscillators. Poincaré-Lindstedt method. Conservative and non-conservative systems. Liénard systems, van der Pol oscillator. [Hilbert's 16th problem](#). [Lorenz equations](#). [Lorenz map](#). [Poincaré section](#). [Poincaré map](#). [Converse of Sharkovsky's theorem](#). [Bernoulli shift map](#), symbolic dynamics. [Tent map](#). [Dynamics on metric spaces](#), sensitive dependence on initial conditions, transitivity, conjugate maps, chaotic dynamics.

Summary of Lecture 15 – general theory

- we studied dynamical systems associated with function $F : \mathbb{M} \rightarrow \mathbb{M}$, where \mathbb{M} is a metric space, *i.e.* a set with metric (distance) $d : \mathbb{M} \times \mathbb{M} \rightarrow [0, \infty)$
- $F : \mathbb{M} \rightarrow \mathbb{M}$ is called **transitive** if there exists $x_0 \in \mathbb{M}$ such that orbit $O(x_0) = \{x_0, F(x_0), F^{(2)}(x_0), F^{(3)}(x_0), \dots\}$ is a dense subset of \mathbb{M} (a **transitive point** of F is a point $x_0 \in \mathbb{M}$ which has a dense orbit $O(x_0)$ under F)
- $F : \mathbb{M} \rightarrow \mathbb{M}$ has **sensitive dependence on initial conditions** if there exists $\delta > 0$ such that for all $x \in \mathbb{M}$ and any open set $U \subset \mathbb{M}$ satisfying $x \in U$, there is a point $y \in U$ and $j \in \mathbb{N}$ with $d(F^{(j)}(x), F^{(j)}(y)) > \delta$
- $F : \mathbb{M} \rightarrow \mathbb{M}$ is said to be **chaotic** if:
 - (i) the set of all periodic points is dense in \mathbb{M}
 - (ii) F is transitive
 - (iii) F has sensitive dependence on initial conditions
- if $F : \mathbb{M} \rightarrow \mathbb{M}$ is continuous and \mathbb{M} is not a finite set, then (i) and (ii) imply (iii)

Summary of Lecture 15 – chaotic maps

Bernoulli shift map: $\sigma : \mathbb{M}_{01} \rightarrow \mathbb{M}_{01}$ where $\sigma((a_1, a_2, a_3, a_4, \dots)) = (a_2, a_3, a_4, a_5, \dots)$ and $\mathbb{M}_{01} = \{(a_1, a_2, a_3, a_4, \dots) \mid \text{such that } a_j = 0 \text{ or } a_j = 1 \text{ for } j = 1, 2, 3, 4, \dots\}$

\mathbb{M}_{01} is a metric space with metric defined by

$$d(x, y) = \sum_{j=1}^{\infty} \frac{|a_j - b_j|}{2^j} \text{ for } x = (a_1, a_2, a_3, a_4, \dots) \text{ and } y = (b_1, b_2, b_3, b_4, \dots)$$

Lemma 6: The Bernoulli shift map $\sigma : \mathbb{M}_{01} \rightarrow \mathbb{M}_{01}$ is **chaotic**.

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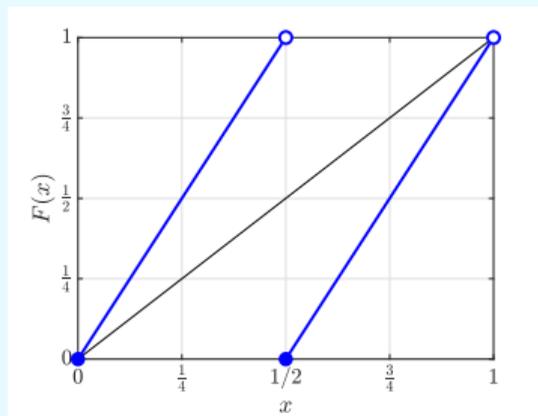
Doubling map: (Question 3 on Problem Sheet 4)

$F : [0, 1) \rightarrow [0, 1)$ where

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$\mathbb{M} = [0, 1)$ with $d(x, y) = |x - y|$

F satisfies properties (i)–(iii) in our definition of chaotic maps \implies **F is chaotic**



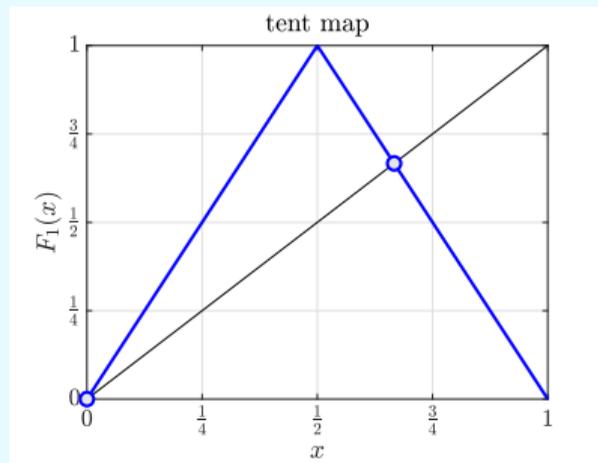
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Tent map F_1 is chaotic

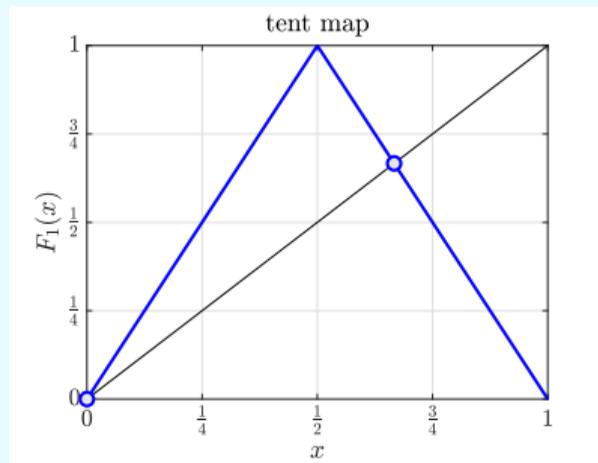
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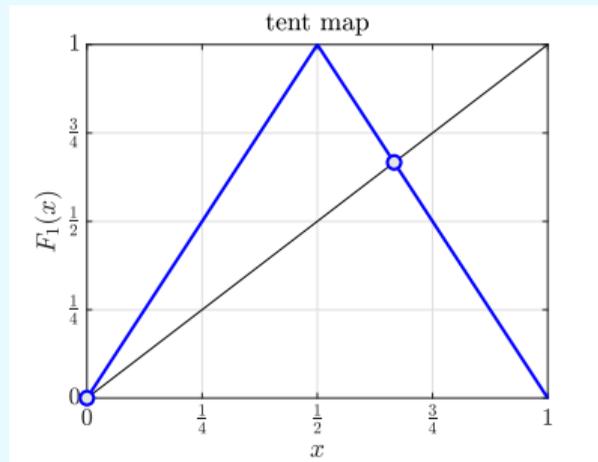
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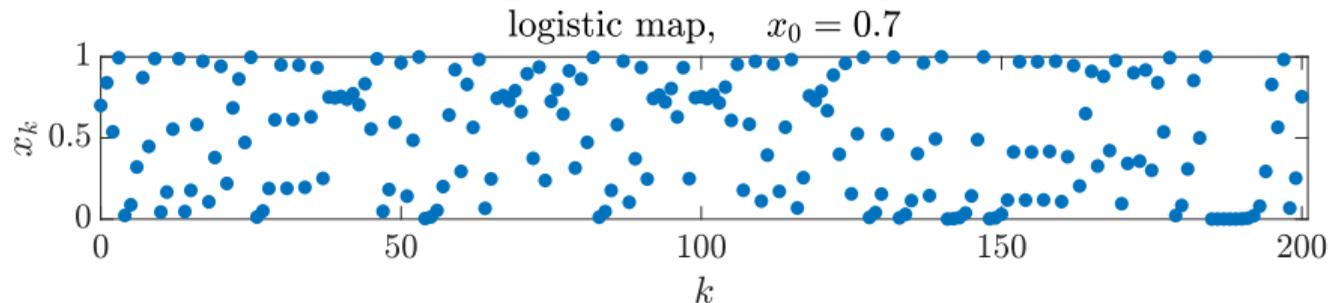
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Another 'intuitive definition' of chaos: [Question 3 on Problem Sheet 0](#)

Starting with $x_0 = 0.7$, we plot $x_{k+1} = F(x_k)$, for the **logistic map** $F(x) = 4x(1-x)$



Tent map F_1 is chaotic

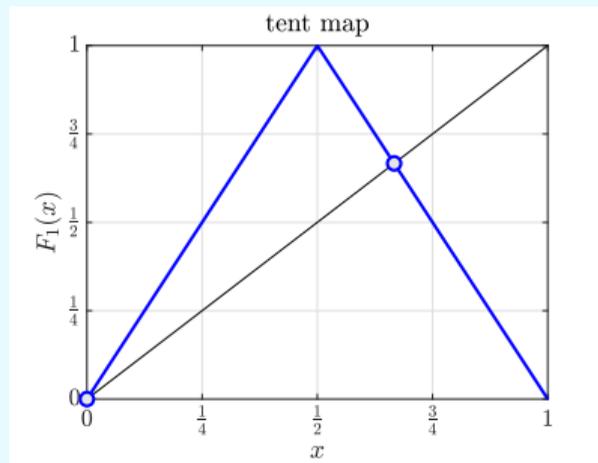
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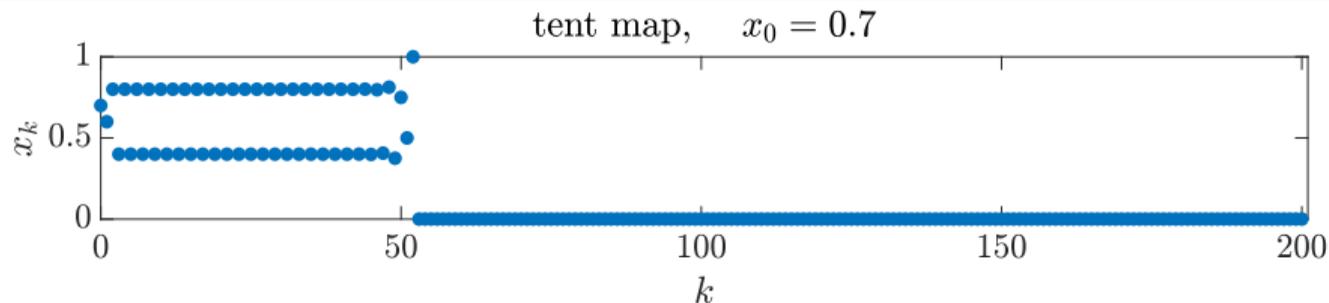
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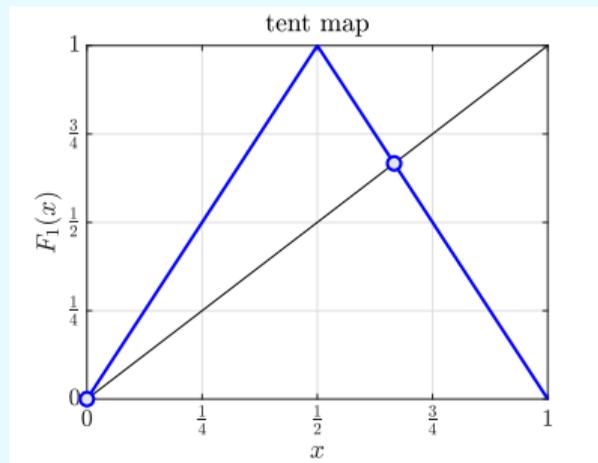
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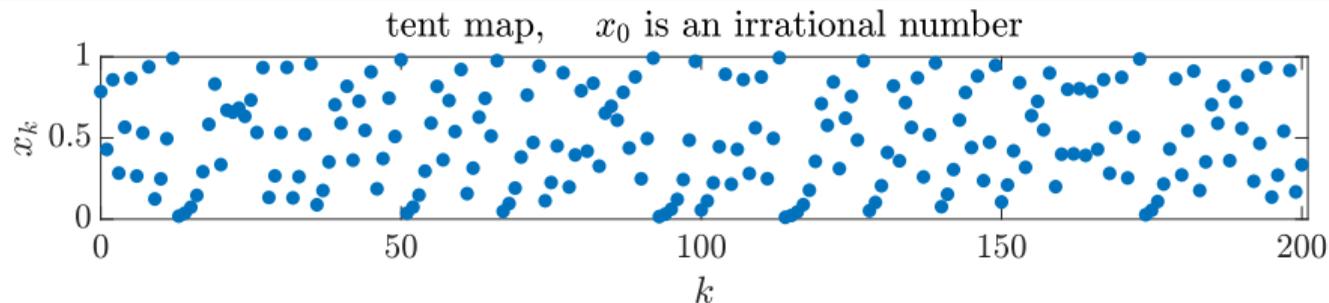
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Tent map F_1 is chaotic

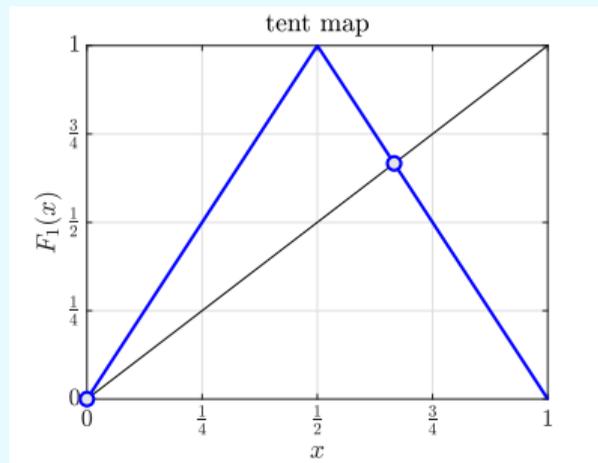
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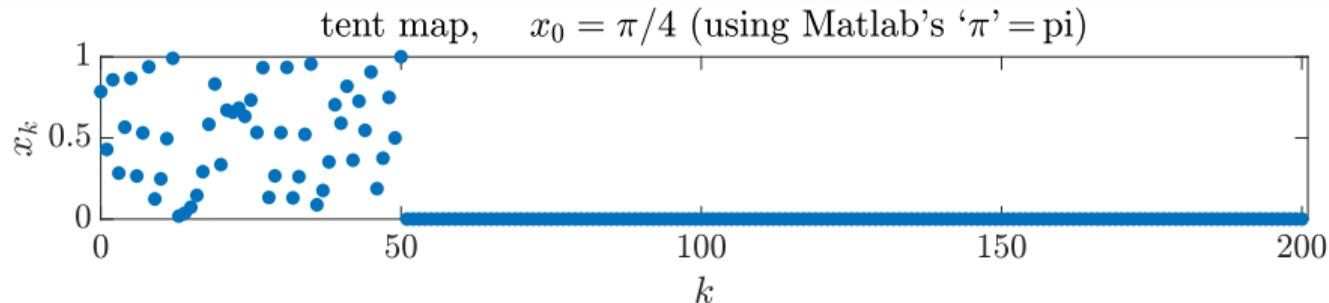
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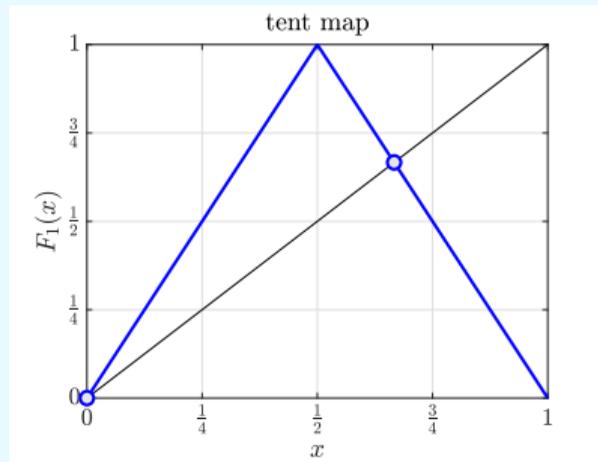
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Conclusions:

- the tent map F_1 is chaotic using our definition of chaos – we can prove it
 - invariant distribution is the uniform distribution on $[0, 1]$ – we can prove it
- but the tent map F_1 is not a good random number generator:

- very small numbers $x_k \approx 0$ are followed by very small numbers $2x_k \approx 0$
- x_k becomes 0 after 50 iterations for $x_0 = \pi/4$ in Matlab
(in computers, irrational numbers are represented as rational numbers, $2^{50} \approx 10^{15}$)

Tent map F_1 is chaotic

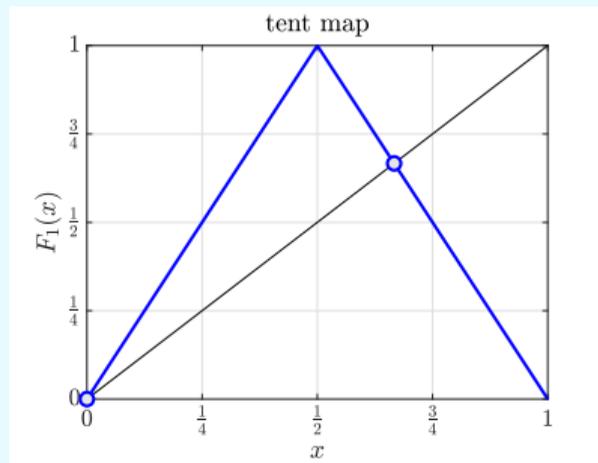
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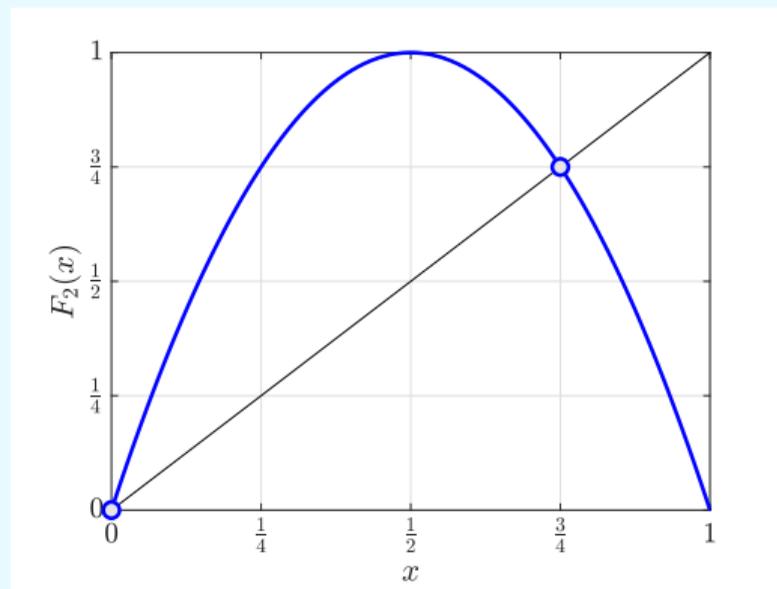
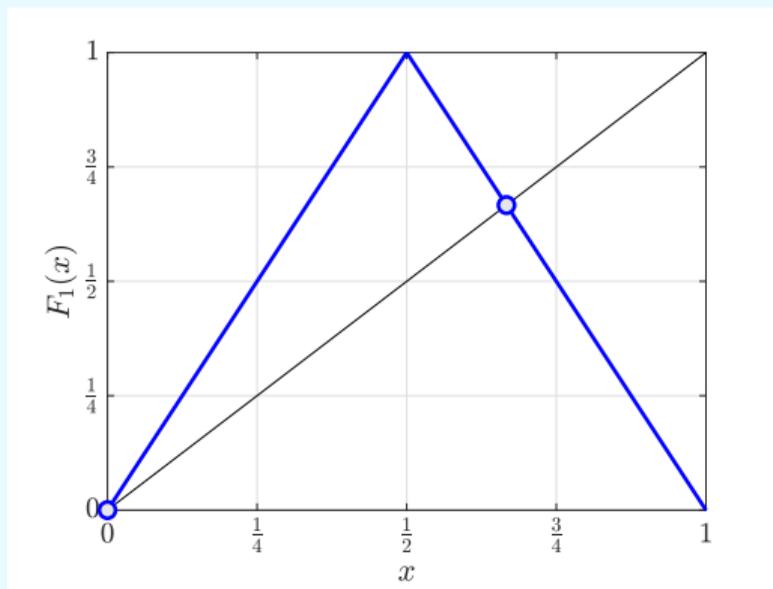


NEXT: the property that the tent map is chaotic can be used to prove the chaotic behaviour of other dynamical systems

Tent map vs. logistic map

$$\text{tent map } F_1(x) = \begin{cases} 2x & \text{for } x \in [0, 1/2] \\ 2 - 2x & \text{for } x \in [1/2, 1] \end{cases}$$

$$\text{logistic map } F_2(x) = 4x(1-x)$$

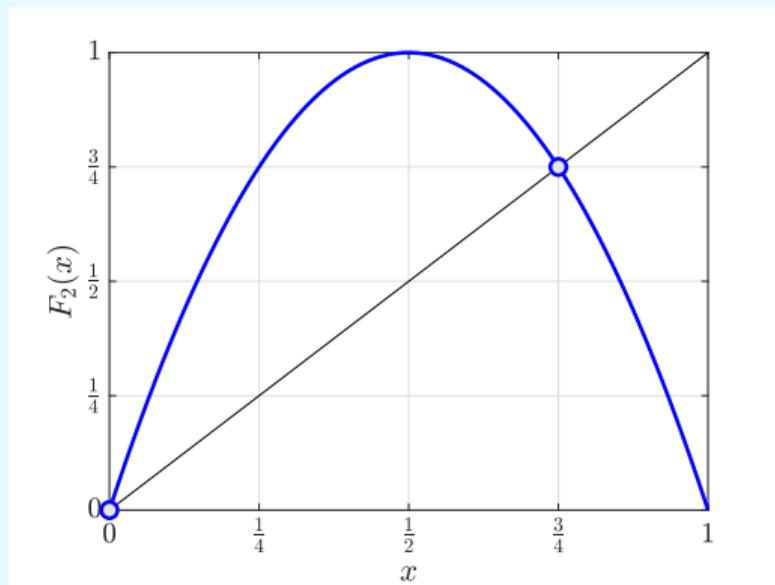
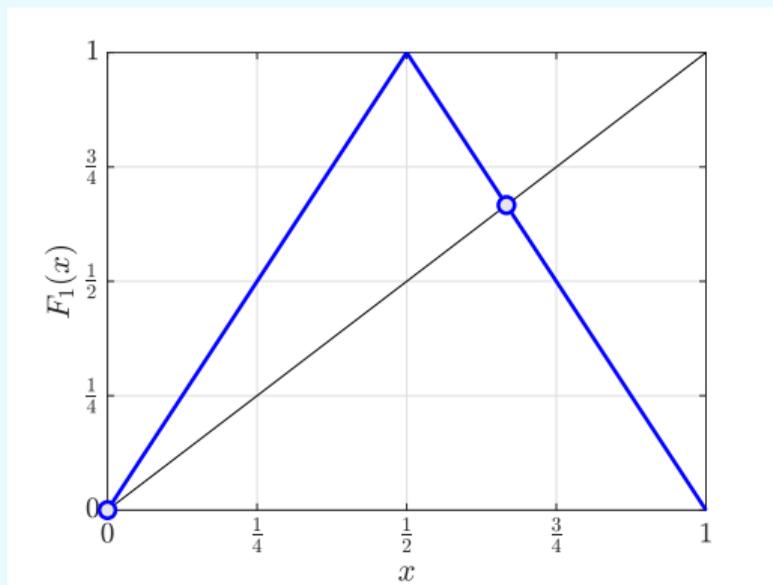


Lemma: We have $h \circ F_1 = F_2 \circ h$ where $h : [0, 1] \rightarrow [0, 1]$ is $h(x) = \sin^2(\pi x/2)$.

Tent map vs. logistic map

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$$\text{logistic map } F_2(x) = 4x(1-x)$$



Lemma: We have $h \circ F_1 = F_2 \circ h$ where $h : [0, 1] \rightarrow [0, 1]$ is $h(x) = \sin^2(\pi x/2)$.

Since h has inverse $h^{-1} = \frac{2}{\pi} \arcsin \sqrt{y}$, we can rewrite this as $F_1 = h^{-1} \circ F_2 \circ h$.

General definition: homeomorphism

Definition: Let \mathbb{M}_1 and \mathbb{M}_2 be two metric spaces.

A function $h : \mathbb{M}_1 \rightarrow \mathbb{M}_2$ is a *homeomorphism* if:

- (i) h is continuous;
- (ii) h is one-to-one, *i.e.* if $h(x) = h(y)$, then $x = y$;
- (iii) h is onto, *i.e.* $\forall y \in \mathbb{M}_2$ there exists $x \in \mathbb{M}_1$ such that $h(x) = y$;
- (iv) the inverse mapping $h^{-1} : \mathbb{M}_2 \rightarrow \mathbb{M}_1$ is continuous.

General definition: conjugate maps

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Definition: Let $F_1 : \mathbb{M}_1 \rightarrow \mathbb{M}_1$ and $F_2 : \mathbb{M}_2 \rightarrow \mathbb{M}_2$ be maps of metric spaces \mathbb{M}_1 and \mathbb{M}_2 , respectively. Then F_1 and F_2 are said to be *conjugate* if there is a homeomorphism $h : \mathbb{M}_1 \rightarrow \mathbb{M}_2$ such that $h \circ F_1 = F_2 \circ h$.

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Example: tent map $F_1(x) = \begin{cases} 2x & \text{for } x \in [0, 1/2] \\ 2 - 2x & \text{for } x \in [1/2, 1] \end{cases}$ is conjugate to the logistic map $F_2(x) = 4x(1-x)$ with conjugacy $h : [0, 1] \rightarrow [0, 1]$ given as $h(x) = \sin^2(\pi x/2)$

Conjugate maps and chaos

Theorem: Let $F_1 : \mathbb{M}_1 \rightarrow \mathbb{M}_1$ and $F_2 : \mathbb{M}_2 \rightarrow \mathbb{M}_2$ be continuous maps of metric spaces \mathbb{M}_1 and \mathbb{M}_2 , respectively, and assume that there is a conjugacy $h : \mathbb{M}_1 \rightarrow \mathbb{M}_2$ with $h \circ F_1 = F_2 \circ h$. Then F_1 is chaotic if and only if F_2 is chaotic.

Conjugate maps and chaos

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Corollary: The logistic map $F_2(x) = 4x(1-x)$ is chaotic,

because the tent map $F_1(x) = \begin{cases} 2x & \text{for } x \in [0, 1/2] \\ 2 - 2x & \text{for } x \in [1/2, 1] \end{cases}$ is conjugate to the logistic map

$F_2(x) = 4x(1-x)$ with conjugacy $h : [0, 1] \rightarrow [0, 1]$ given as $h(x) = \sin^2(\pi x/2)$.

Conjugate maps and chaos

Theorem: Let $F_1 : \mathbb{M}_1 \rightarrow \mathbb{M}_1$ and $F_2 : \mathbb{M}_2 \rightarrow \mathbb{M}_2$ be continuous maps of metric spaces \mathbb{M}_1 and \mathbb{M}_2 , respectively, and assume that there is a conjugacy $h : \mathbb{M}_1 \rightarrow \mathbb{M}_2$ with $h \circ F_1 = F_2 \circ h$. Then F_1 is chaotic if and only if F_2 is chaotic.

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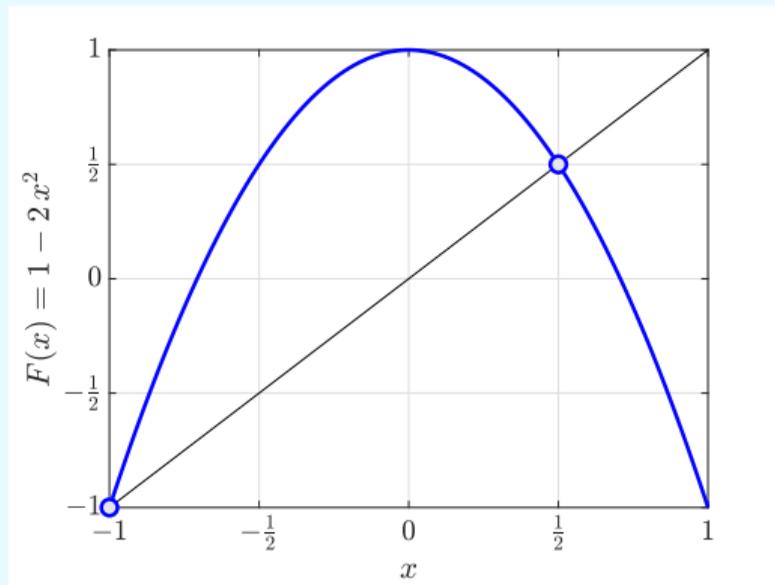
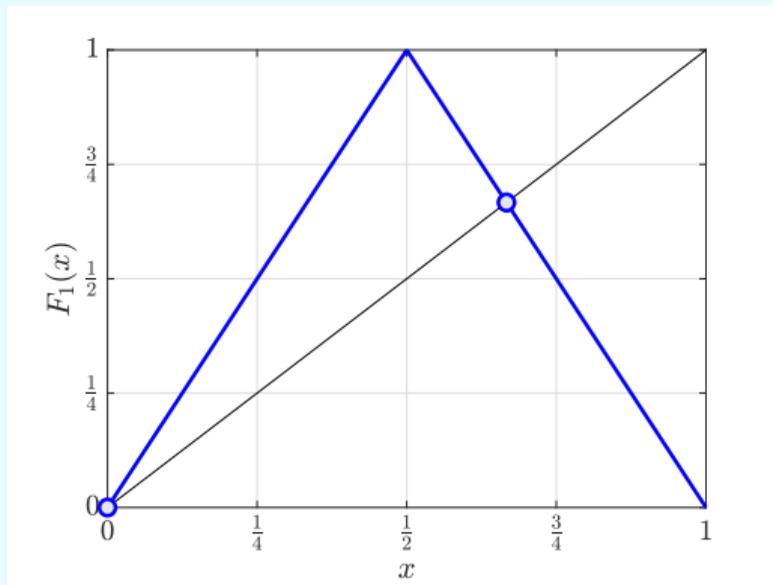
$F_2(x) = 4x(1-x)$ with conjugacy $h : [0, 1] \rightarrow [0, 1]$ given as $h(x) = \sin^2(\pi x/2)$.

Remark: Different variants of our Theorem also hold under weaker assumptions. For example, if h is not a homeomorphism, but h is only continuous and onto satisfying $h \circ F_1 = F_2 \circ h$. Then, assuming that both $F_1 : \mathbb{M}_1 \rightarrow \mathbb{M}_1$ and $F_2 : \mathbb{M}_2 \rightarrow \mathbb{M}_2$ are continuous maps of metric spaces \mathbb{M}_1 and \mathbb{M}_2 , respectively, we have:
If F_1 is chaotic, then F_2 is chaotic.

Map $F(x) = 1 - 2x^2$ in Question 1 on the 2024 Exam Paper

$$\text{tent map } F_1(x) = \begin{cases} 2x & \text{for } x \in [0, 1/2] \\ 2 - 2x & \text{for } x \in [1/2, 1] \end{cases}$$

$$\text{map } F(x) = 1 - 2x^2 \text{ for } x \in [-1, 1]$$

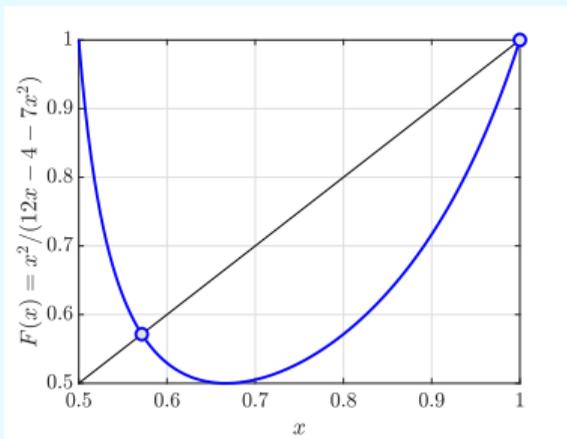


Map $F(x) = 1 - 2x^2$ is chaotic because we have $h \circ F_1 = F \circ h$ where $h : [0, 1] \rightarrow [-1, 1]$ is $h(x) = -\cos(\pi x)$ and $F_1(x)$ is chaotic.

Map $F(x) = \frac{x^2}{12x-4-7x^2}$ in Question 1 on the 2025 Exam Paper

$$F(x) = \frac{x^2}{12x - 4 - 7x^2}$$

for $x \in [\frac{1}{2}, 1]$



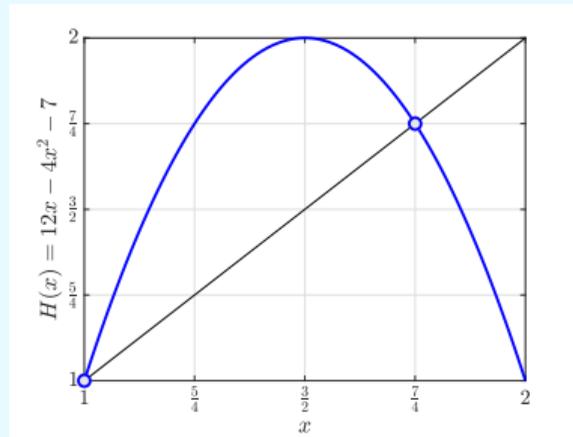
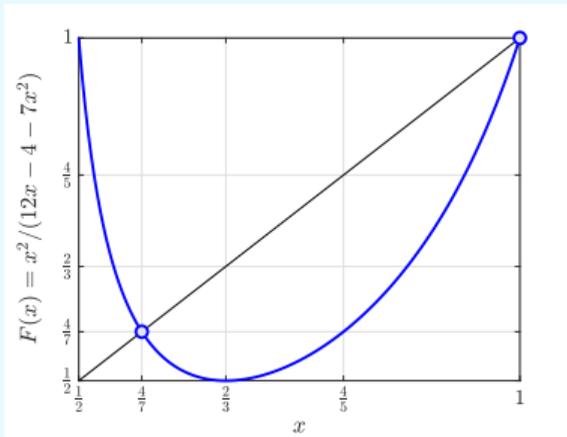
Map $F(x) = \frac{x^2}{12x-4-7x^2}$ in Question 1 on the 2025 Exam Paper

$$F(x) = \frac{x^2}{12x - 4 - 7x^2}$$

for $x \in [\frac{1}{2}, 1]$

map $H(x) = 12x - 4x^2 - 7$

for $x \in [1, 2]$



Map $F(x) = \frac{x^2}{12x-4-7x^2}$ is chaotic because we have $h \circ H = F \circ h$ where $h : [1, 2] \rightarrow [1/2, 1]$ is $h(x) = \frac{1}{x}$ and map $H(x)$ is chaotic.

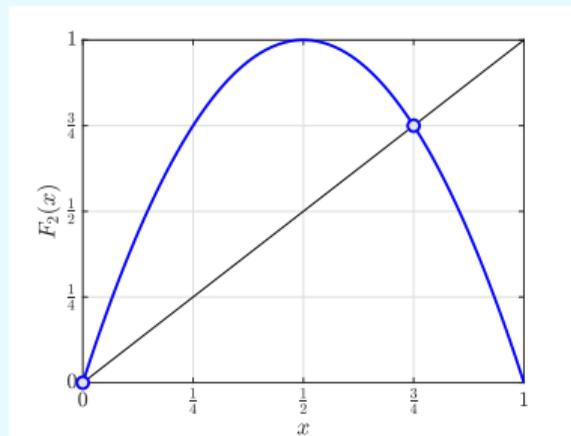
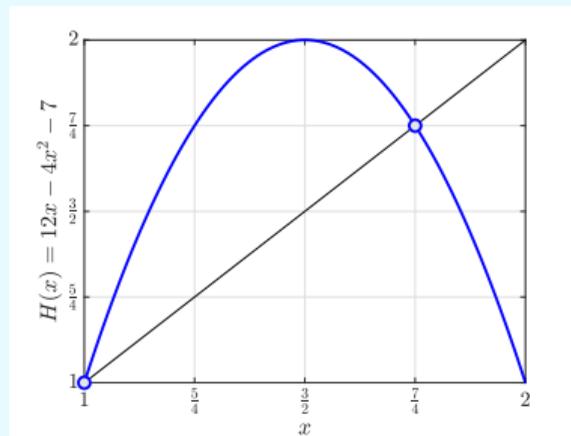
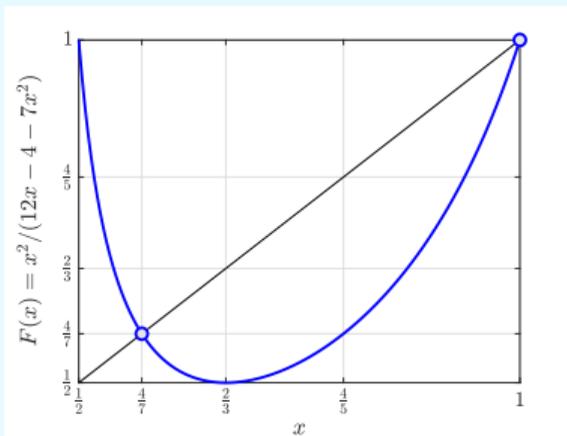
Map $F(x) = \frac{x^2}{12x-4-7x^2}$ in Question 1 on the 2025 Exam Paper

$$F(x) = \frac{x^2}{12x - 4 - 7x^2}$$

for $x \in [\frac{1}{2}, 1]$

map $H(x) = 12x - 4x^2 - 7$ for $x \in [1, 2]$

logistic map $F_2(x) = 4x(1-x)$ for $x \in [0, 1]$



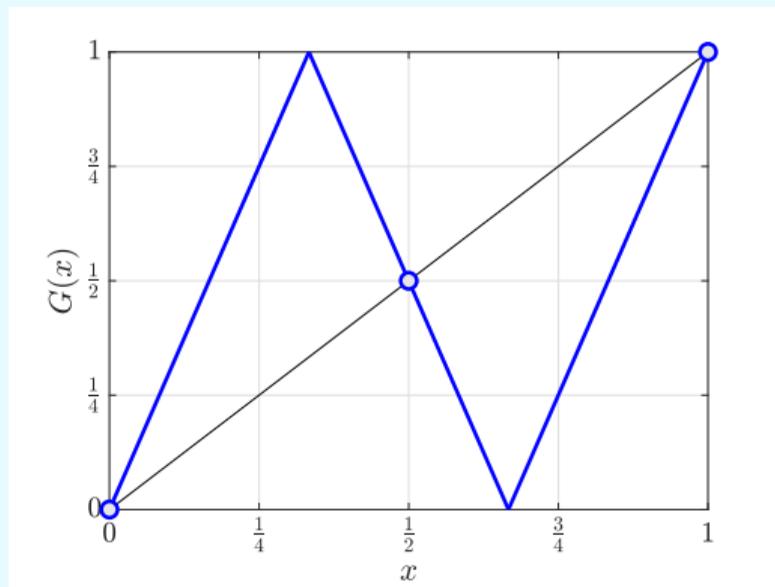
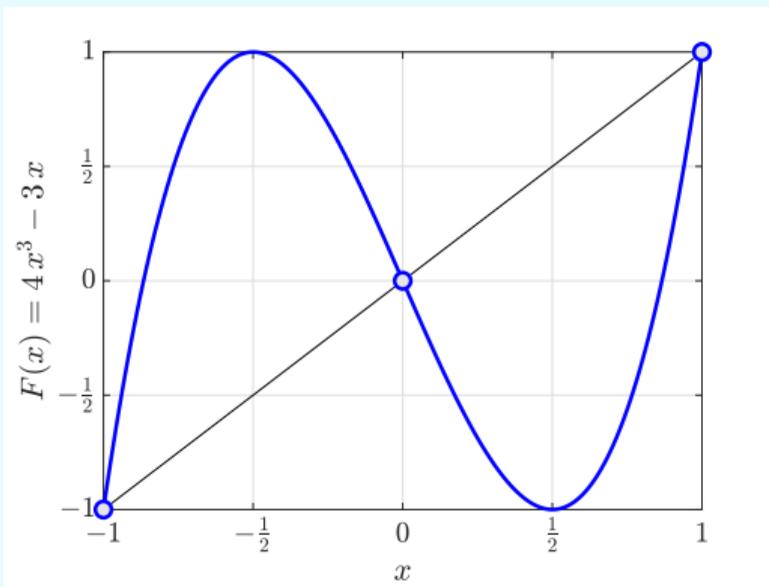
Map $F(x) = \frac{x^2}{12x-4-7x^2}$ is chaotic because we have $h \circ H = F \circ h$ where $h : [1, 2] \rightarrow [1/2, 1]$ is $h(x) = \frac{1}{x}$ and map $H(x)$ is chaotic.

Map $H(x) = 12x - 4x^2 - 7$ is chaotic because we have $g \circ F_2 = H \circ g$ where $g : [0, 1] \rightarrow [1, 2]$ is $g(x) = x + 1$ and logistic map $F_2(x)$ is chaotic.

Map $F(x) = 4x^3 - 3x$ in Question 7 on Problem Sheet 2

$$\text{map } F(x) = 4x^3 - 3x \text{ for } x \in [-1, 1]$$

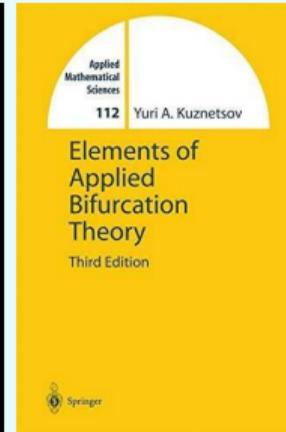
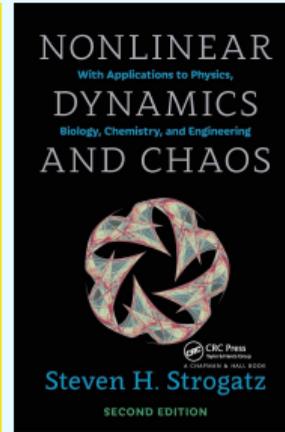
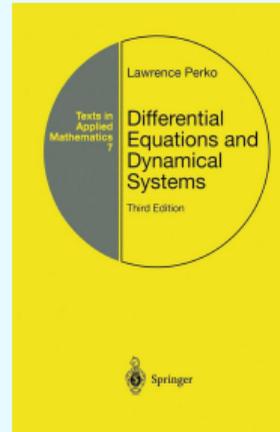
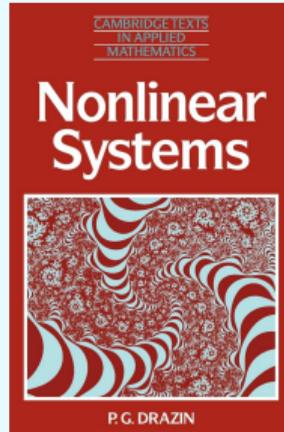
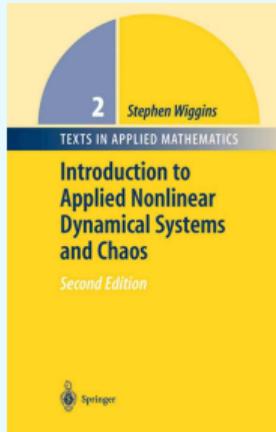
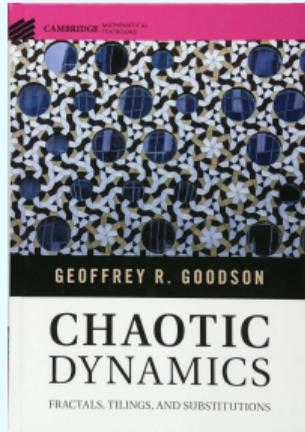
$$\text{map } G(x) = \begin{cases} 3x & \text{for } x \in [0, 1/3] \\ 2 - 3x & \text{for } x \in [1/3, 2/3] \\ 3x - 2 & \text{for } x \in [2/3, 1] \end{cases}$$



Map $F(x) = 4x^3 - 3x$ is chaotic, because we have $h \circ G = F \circ h$ with conjugacy $h : [0, 1] \rightarrow [-1, 1]$ given by $h(x) = \cos(\pi x)$ and $G(x)$ is chaotic.

End of course. This is our last slide.

Further examples and additional discussions can be found in the reading list:



Exams: see slide 3 (discussed in Lecture 1)

Thank you for your attention!