Reading for this lecture:

NDAC, Sec. 6.0-6.4, 7.0-7.3, & 9.0-9.4

ID Flows: Fixed Points model of static equilibrium

ID Flow: $x \in \mathbb{R}$

 $\dot{x} = F(x)$

Fixed Points: $x^* \in \mathbb{R}$ such that $\dot{x}|_{x^*} = 0$ or $F(x^*) = 0$

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ID Flows: Fixed Points ...

Stability: What is linearized system at x ? Investigate evolution of perturbations: $x'=x+\delta x$ $|\delta x|\ll 1$

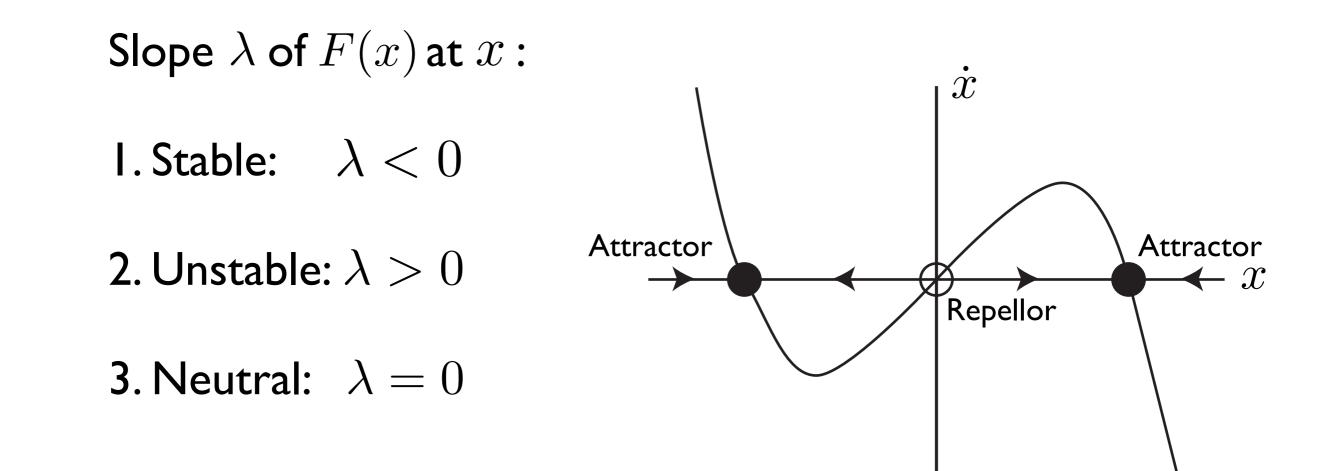
Local Flow:
$$\delta \dot{x} = \frac{dF}{dx}\Big|_{x(t)} \delta x$$

 $\dot{\mathbf{r}} - F(\mathbf{r})$

Local Linear System: $\delta \dot{x} = \lambda \ \delta x$

Solution:
$$\delta x(t) \propto e^{\lambda t} \delta x(0)$$

Example Dynamical Systems ... ID Flows ... Stability Classification of Fixed Points:



Example Dynamical Systems ...

Linear algebra review:

 $n \times n$ matrix: A

Determinant: Det(A)

Trace: Tr(A)

For example:

$$2 \times 2$$
 matrix: $A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$
 $Det(A) = ad - bc$
 $Tr(A) = a + d$

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(See, for example, NDAC, Chapter 5)

Example Dynamical Systems ...

Linear algebra review ...

(Left) eigensytem: $\lambda_i \vec{v}_i = \vec{v}_i \cdot A$

(Left) eigenvalues: $\{\lambda_i \in \mathbb{C} : i = 1, \dots, n\}$

(Left) eigenvectors: $\{ \vec{v}_i \in \mathbb{R}^n : i = 1, \dots, n \}$

Scaling factors Eigen-directions that are invariant

Example Dynamical Systems ...

Linear algebra review ...

(Right) eigensytem: $\lambda_i \vec{v}_i = A \cdot \vec{v}_i$

(Right) eigenvalues: $\{\lambda_i \in \mathbb{C} : i = 1, \dots, n\}$

(Right) eigenvectors: $\{ \vec{v}_i \in \mathbb{R}^n : i = 1, \dots, n \}$

Scaling factors Eigen-directions that are invariant

Example Dynamical Systems ...

Linear algebra review ...

Determinant:
$$Det(A) = \prod_{i=1}^{n} \lambda_i$$

Trace:

$$\operatorname{Tr}(A) = \sum_{i=1}^{n} \lambda_i$$

Example Dynamical Systems ...

2D Flows: Fixed Points model of static equilibrium

2D Flow: $\vec{x} \in \mathbf{R}^2$

$$\dot{\vec{x}} = \vec{F}(\vec{x})$$
 or $\dot{x} = f(x,y)$
 $\vec{x} = (x,y)$ $\dot{y} = g(x,y)$
 $\vec{F} = (f,g)$

•

Fixed Points:

 (x^*, y^*) such that

$$\dot{\vec{x}}|_{(x^*,y^*)} = (0,0)$$

$$0 = f(x^*, y^*) 0 = g(x^*, y^*)$$

Example Dynamical Systems ...

2D Flows: Fixed Points ... model of static equilibrium

Stability: What is linearized system at \vec{x} ? Investigate evolution of perturbations: $\vec{x}' = \vec{x} + \delta \vec{x}$ $|\delta \vec{x}| \ll 1$

$$\dot{\vec{x}} = \vec{F}(\vec{x})$$

Local Flow:
$$\delta \dot{\vec{x}} = \left. \frac{\partial \vec{F}}{\partial \vec{x}} \right|_{\vec{x}(t)} \cdot \delta \vec{x}$$

Initial conditions: $x(0) \ \delta x(0)$

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Example Dynamical Systems ...

2D Flows: Fixed Points ...

Local Linear System: $\delta \dot{\vec{x}} = A \cdot \delta \vec{x}$

Jacobian:
$$A = \frac{\partial \vec{F}}{\partial \vec{x}} = \begin{pmatrix} \frac{\partial f}{\partial x} \frac{\partial f}{\partial y} \\ \frac{\partial g}{\partial x} \frac{\partial g}{\partial y} \end{pmatrix}$$

Solution:

$$\delta \vec{x}(t) \propto e^{At} \delta \vec{x}(0)$$

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Example Dynamical Systems ...

2D Flows: Fixed Points (an aside) ...

Solve linear ODEs: Find x(t) given

 $\begin{aligned} x(0) \\ \dot{x} = Ax \end{aligned}$

Eigenvalues and eigenvectors: $\lambda_j \, v_j$

$$Av_j = \lambda_j, \ j = 1, \dots, n$$

Solution:

$$x(t) = \sum_{j=1}^{\infty} \alpha_j e^{\lambda_j t} v_j$$

where calculate α_j :

$$x(0) = \sum_{j=1}^{\infty} \alpha_j v_j$$

Example Dynamical Systems ... 2D Flows: Fixed Points ...

Linear system:
$$\delta \vec{x} = A \cdot \delta \vec{x}$$

If $\delta \vec{x}$ aligns with an eigen-direction, then $A \sim$

and $\begin{array}{rcl} \delta \dot{x}_1 &=& \lambda_1 \ \delta x_1 \\ \delta \dot{x}_2 &=& \lambda_2 \ \delta x_2 \end{array}$

$$\sim \left(\begin{array}{ccccc} \lambda_1 & 0 & 0 & \dots \\ 0 & \lambda_2 & 0 & \dots \\ 0 & 0 & \lambda_3 & \dots \\ \vdots & \vdots & \vdots & \ddots \end{array}\right)$$

so that

$$\begin{split} \delta x_1(t) &= e^{\lambda_1 t} \cdot \delta x_1(0) & \lambda_i > 0 \quad \text{Growth} \\ \delta x_2(t) &= e^{\lambda_2 t} \cdot \delta x_2(0) & \lambda_i < 0 \quad \text{Decay} \end{split}$$

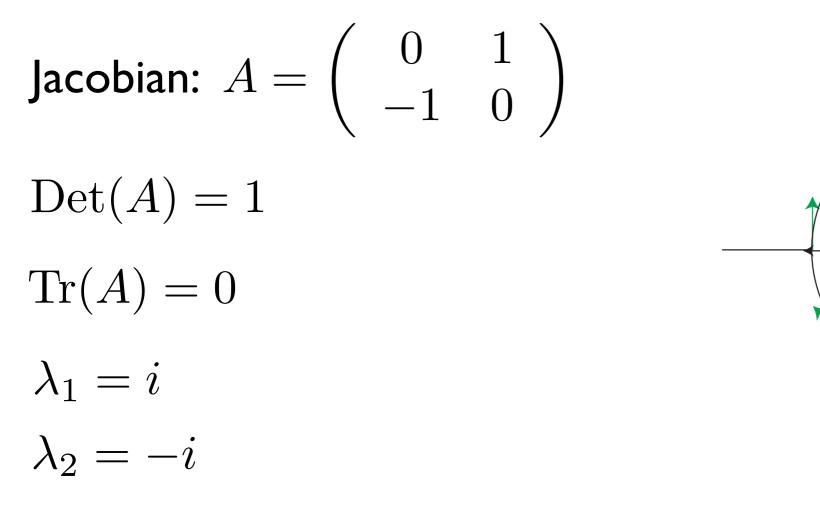
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Example Dynamical Systems ... 2D Flows: Fixed Points ... Simple harmonic oscillator:

$$\dot{x} = y$$
$$\dot{x} = 0$$
$$\dot{y} = -x$$

Fixed point at the origin: (x, y) = (0, 0)



2D Flows: Fixed Points ... Damped harmonic oscillator:

$$\ddot{x} + \gamma \dot{x} + x = 0$$
 or $\dot{x} = y$
 $\dot{y} = -x - \gamma y$

Fixed point at the origin: (x, y) = (0, 0)

Jacobian:
$$A = \begin{pmatrix} 0 & 1 \\ -1 & -\gamma \end{pmatrix}$$

 $\begin{array}{ll} \operatorname{Det}(A) = 1 & \qquad \gamma > 0 & \quad \mathsf{Damped} \\ \operatorname{Tr}(A) = -\gamma & \qquad \gamma < 0 & \quad \mathsf{Unstable} \end{array}$

2D Flows: Fixed Points ...

Damped harmonic oscillator ...

Eigenvalues:
$$\lambda_1 = \frac{-\gamma + \sqrt{\gamma^2 - 4}}{2}$$

 $\lambda_2 = \frac{-\gamma - \sqrt{\gamma^2 - 4}}{2}$

Real when: $|\gamma| \ge 2$

Critically damped

Complex when: $|\gamma| < 2$

Under-damped: spiral

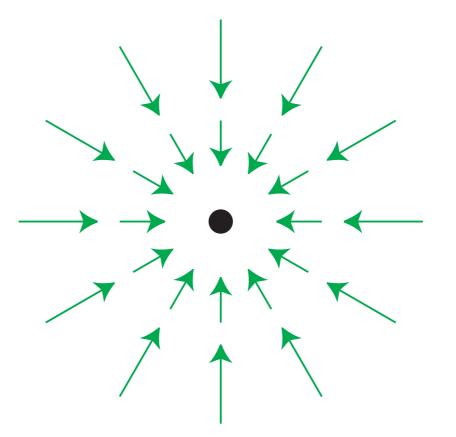
Example Dynamical Systems ... 2D Flows ... Stability Classification of Fixed Points:

Eigenvalues of Jacobian A at $\vec{x}: \lambda_1 \& \lambda_2 \in \mathbf{C}$

(Review: NDAC, Chapter 5)

Stable fixed point (aka sink, attractor):

 $\Re(\lambda_1), \Re(\lambda_2) < 0$

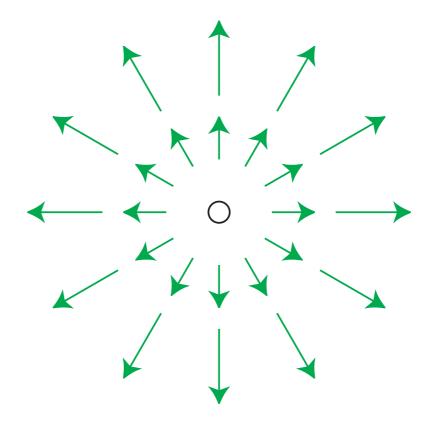


Example Dynamical Systems ... 2D Flows ... Stability Classification of Fixed Points ...

Eigenvalues of Jacobian A at $\vec{x}: \lambda_1 \& \lambda_2 \in \mathbf{C}$

Unstable fixed point (aka source, repellor):

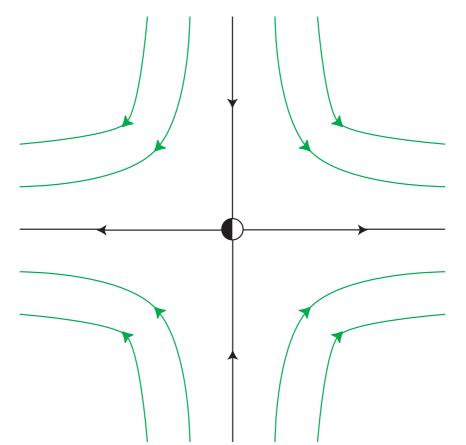
 $\Re(\lambda_1), \Re(\lambda_2) > 0$



Example Dynamical Systems ... 2D Flows ... Stability Classification of Fixed Points:

Eigenvalues of Jacobian at $\vec{x} : \lambda_1 \& \lambda_2 \in \mathbf{C}$

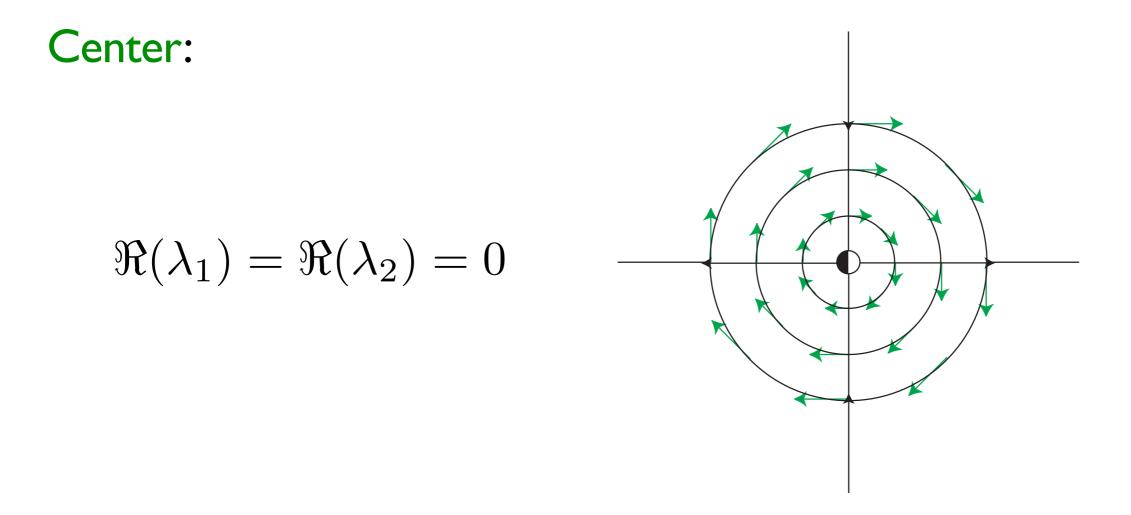
Saddle fixed point (mixed stability):



 $\Re(\lambda_1) > 0 \& \Re(\lambda_2) < 0$

Example Dynamical Systems ... 2D Flows ... Stability Classification of Fixed Points:

Eigenvalues of Jacobian at $\vec{x} : \lambda_1 \ \& \ \lambda_2 \ \in \ \mathbf{C}$



Example Dynamical Systems ...

2D Flows ...

Stability Classification of Fixed Points ...

Class of submanifolds: $Det(A) = \lambda_1 \cdot \lambda_2$

 $Det(A) < 0 : \lambda_1, \lambda_2 \in \mathbf{R}, \ \lambda_1 > 0 \Rightarrow \lambda_2 < 0$ Saddles

Det(A) > 0: Stability $Tr(A) = \lambda_1 + \lambda_2$

Stable: Tr(A) < 0

Unstable: Tr(A) > 0

Marginal: Tr(A) = 0

Example Dynamical Systems ...

2D Flows ... Stability Classification of Fixed Points ... $\operatorname{Tr}(A)$ $\mathrm{Tr}^2(A) - 4\mathrm{Det}(A) = 0$ Unstable **Saddles Unstable Spirals** Det(A)Stable Spirals **Saddles** Stable

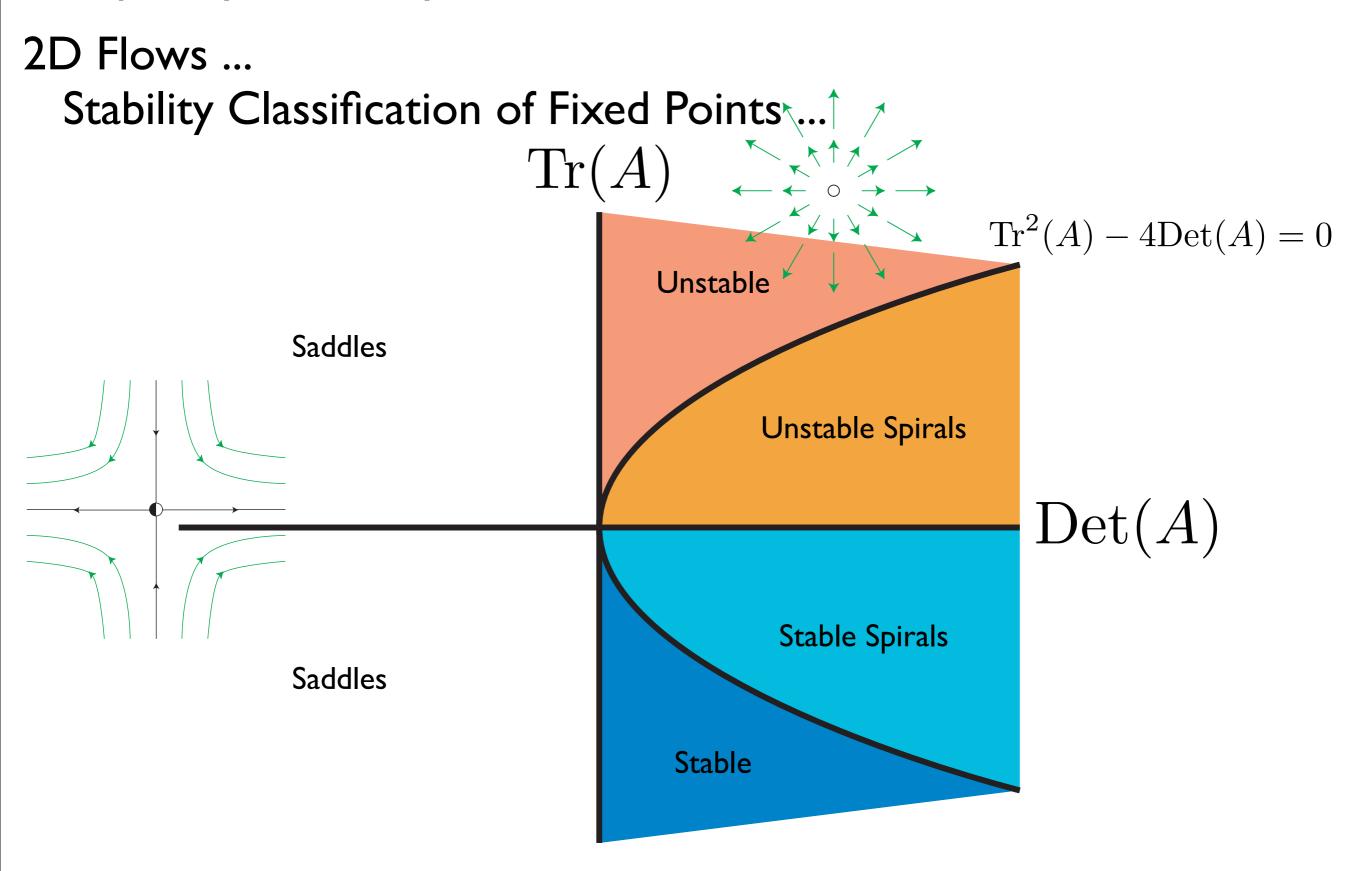
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Example Dynamical Systems ...

2D Flows ... Stability Classification of Fixed Points ... $\operatorname{Tr}(A)$ $\mathrm{Tr}^2(A) - 4\mathrm{Det}(A) = 0$ Unstable **Saddles Unstable Spirals** Det(A)Stable Spirals **Saddles** Stable

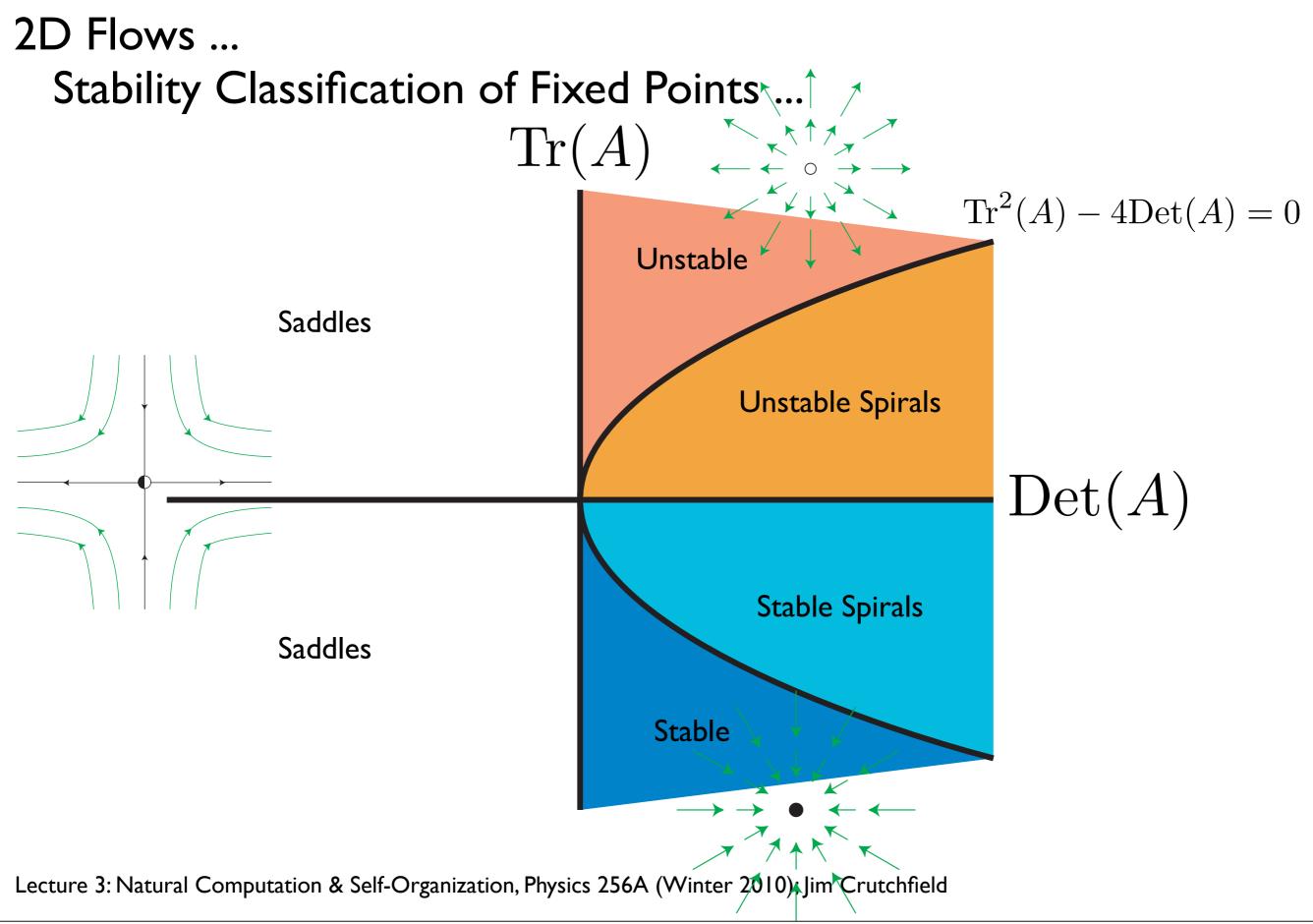
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Example Dynamical Systems ...

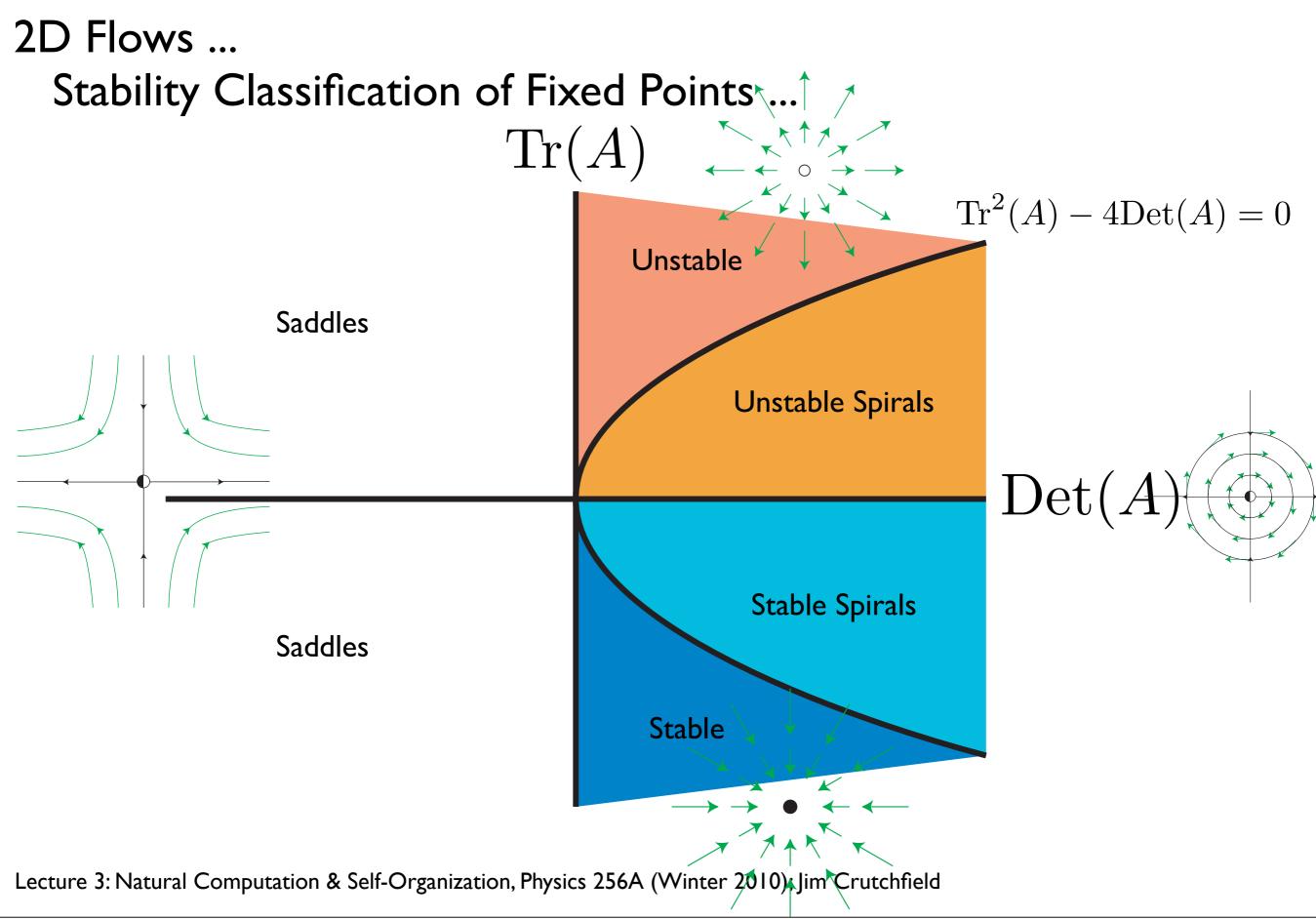


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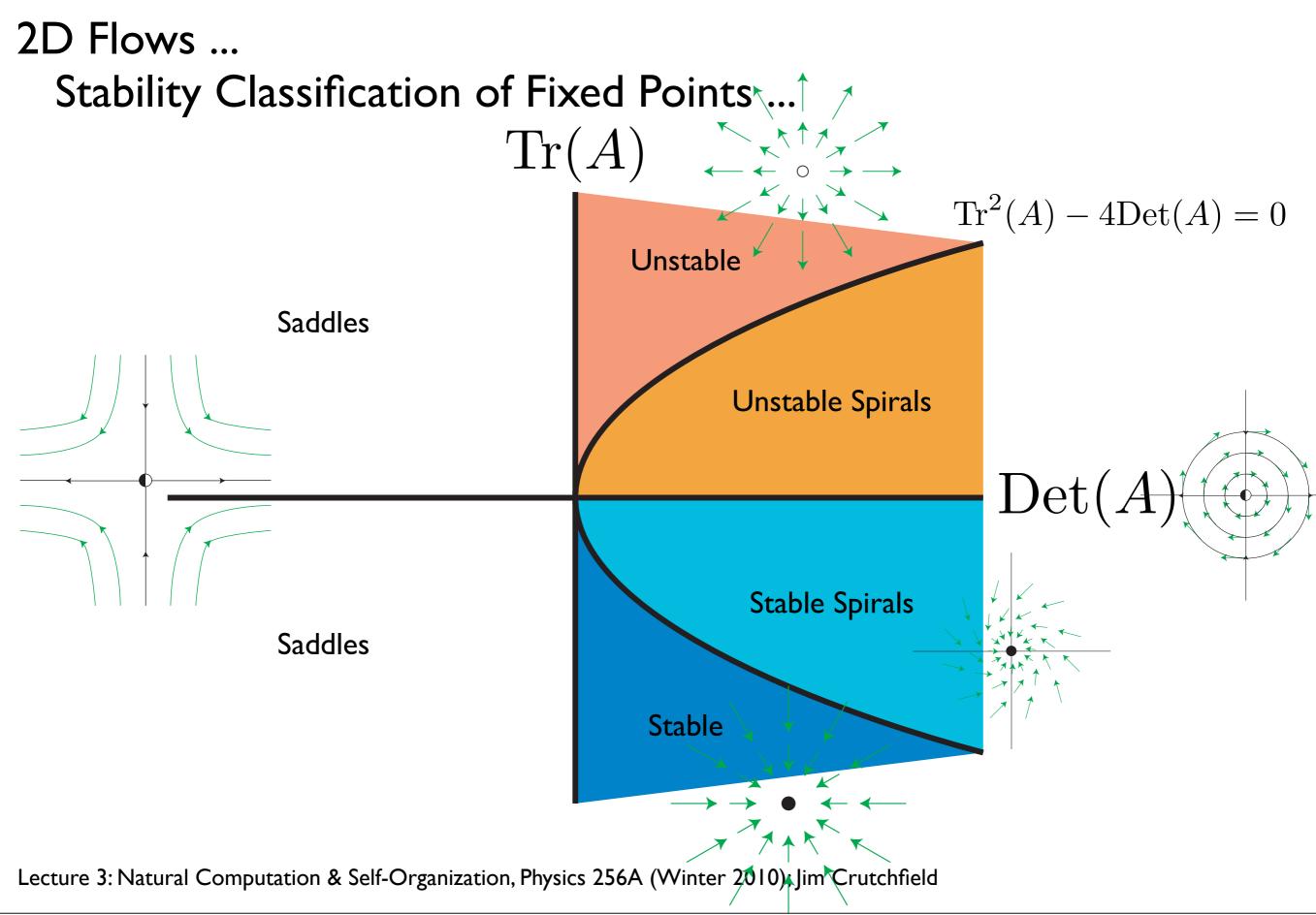
Example Dynamical Systems ...



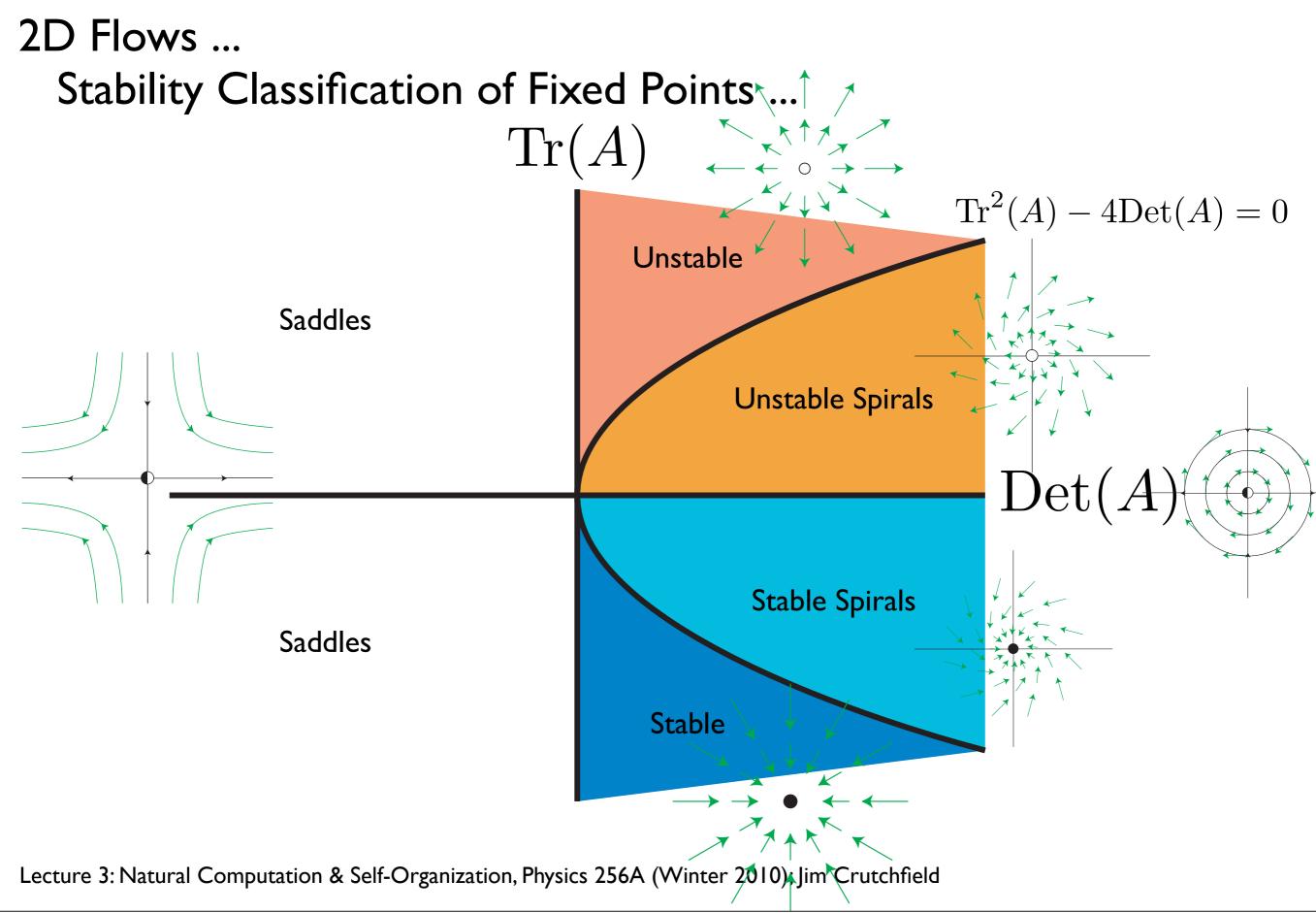
Example Dynamical Systems ...



Example Dynamical Systems ...



Example Dynamical Systems ...



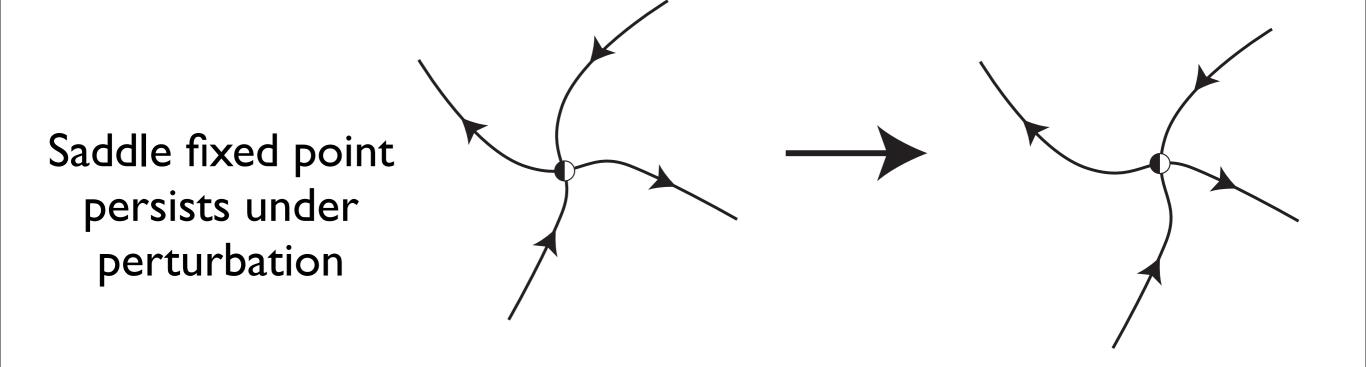
Example Dynamical Systems ...

2D Flows ...

Stability Classification of Fixed Points ...

Hyperbolic intersection of W^s and W^u :

Robust, if $\Re(\lambda_i) \neq 0, \forall i$



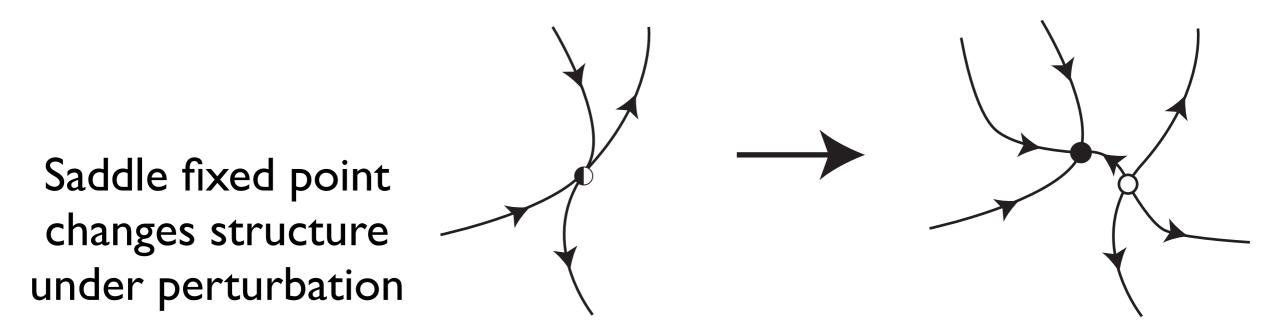
Example Dynamical Systems ...

2D Flows ...

Stability Classification of Fixed Points ...

Non-hyperbolic intersection of W^s and W^u :

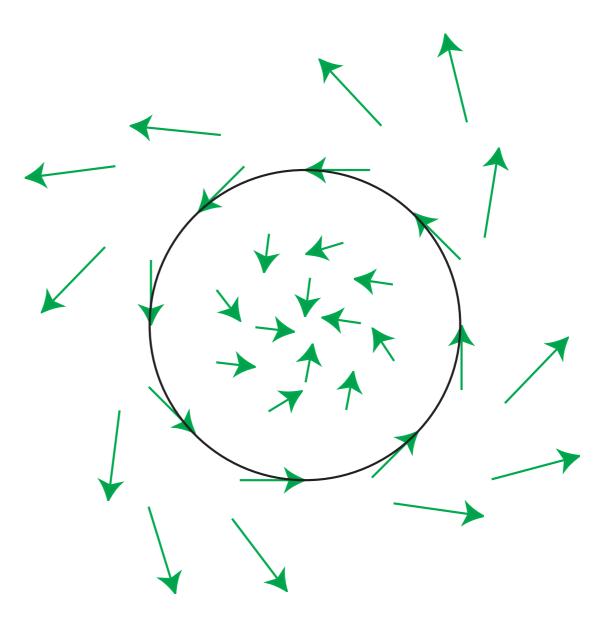
Fragile



2D Flows: Limit Cycles isolated, closed trajectory: a periodic orbit: $\vec{x}(t) = \vec{x}(t+p)$, for all t (p is the period) model of stable oscillation this is a new behavior type not possible in ID flows Stable limit cycle

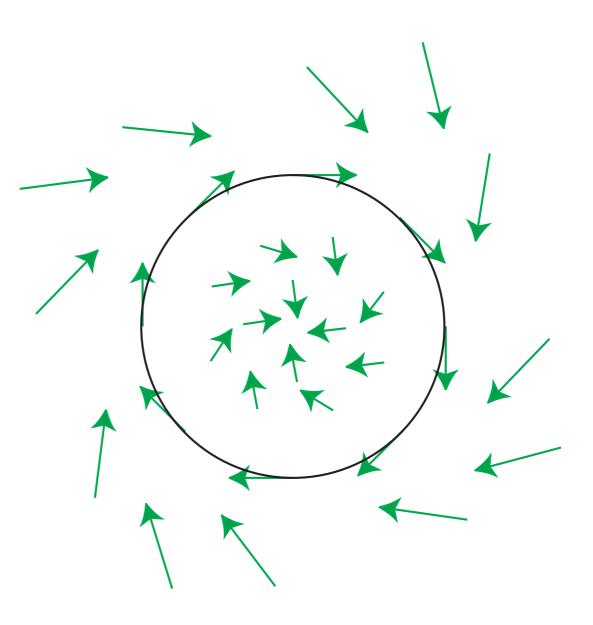
2D Flows: Limit Cycles ...

Unstable cycle



2D Flows: Limit Cycles ...

Saddle cycle



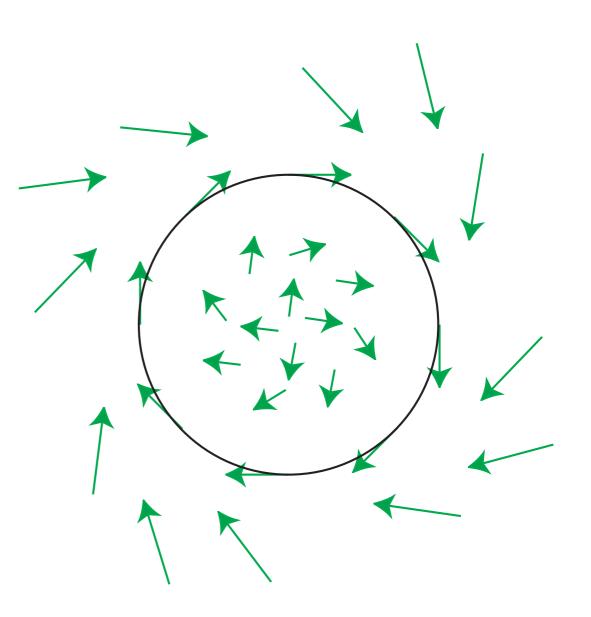
Example Dynamical Systems ...

2D Flows ... Limit Cycle Examples

Easy in polar coordinates:

$$\dot{r} = r(1 - r^2)$$

 $\dot{\theta} = 1$



Example Dynamical Systems ...

2D Flows ... Limit Cycle Examples ...

Van der Pol Equations:

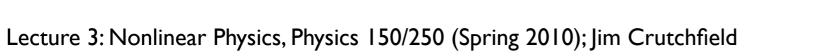
$$\ddot{x} + \mu(x^2 - a)\dot{x} + x = 0$$

or

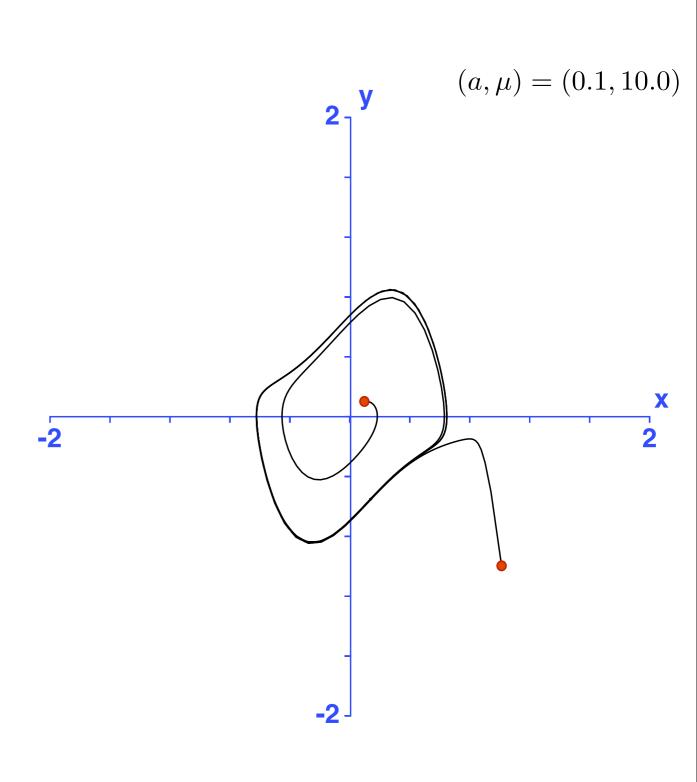
$$\dot{x} = y$$

$$\dot{y} = -x + \mu y (a - x^2)$$

Nonlinear damping changes sign: Small oscillation ($x < \sqrt{a}$): Growth Large oscillation ($x > \sqrt{a}$): Damped







2D Flows ... Limit cycle existence (requires real work to show!)

Systems that can't have stable oscillations:

- I. Simple harmonic oscillator
- 2. Gradient systems: $\dot{\vec{x}} = -\nabla V(\vec{x})$
- 3. Lyapunov systems

2D Flows ... Limit cycle existence (requires real work to show!) How to find limit cycles?

Poincaré-Bendixson Theorem: (a) trajectory confined to trapping region (b) no fixed points then have limit cycle Csomewhere inside R.

Example Dynamical Systems ...

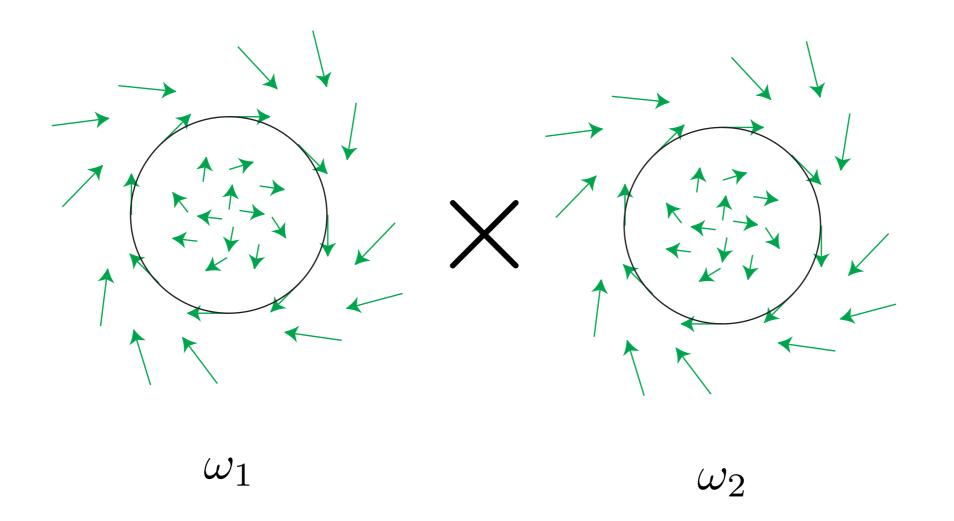
3D Flows: Fixed points

Limit cycles

and ... ?

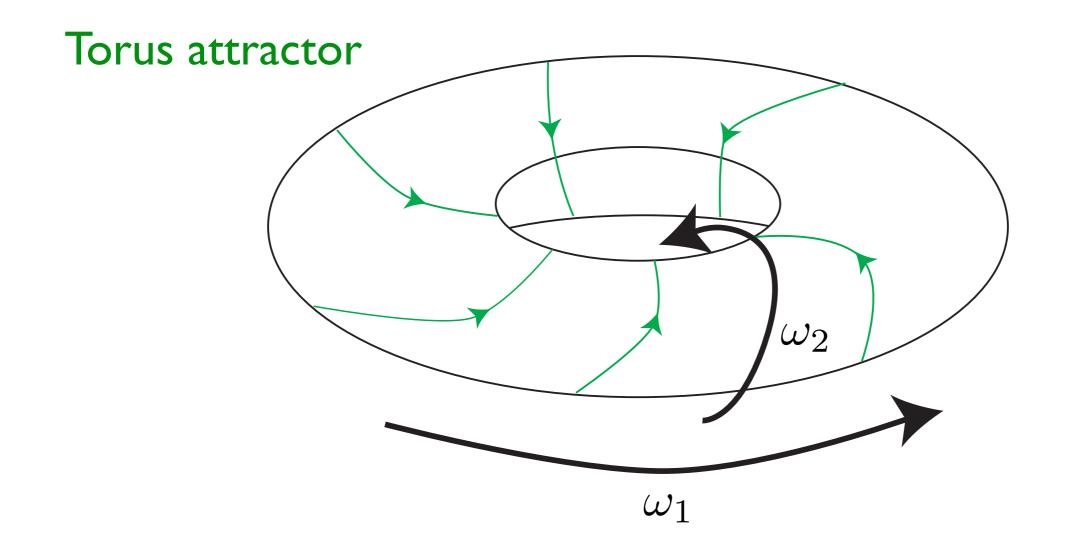
Example Dynamical Systems ...

3D Flows: Quasiperiodicity product of two limit cycles: two irrational frequencies $\omega_1 \neq \omega_2$



3D Flows: Quasiperiodicity

product of two limit cycles: two frequencies, $n \cdot w_1 \neq m \cdot w_2$ a new kind of behavior: *not possible* in ID or 2D



The Big, Big Picture (Bifurcations II) ...

Torus attractor in 3D Flow: Driven van der Pol:

$$\ddot{x} + \mu(x^2 - 1)\dot{x} + x = A\sin(\omega t)$$

One way to verify 3D state space: Write out equivalent first-order ODEs: (Goal: RHS is time independent)

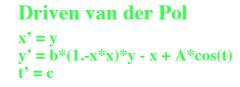
$$\dot{x} = y \dot{y} = \mu(1 - x^2)y - x + A\sin(\phi) \dot{\phi} = \omega$$

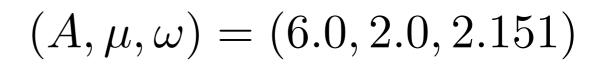
 $(x, y, \phi) \in \mathbb{R}^2 \times S^1$

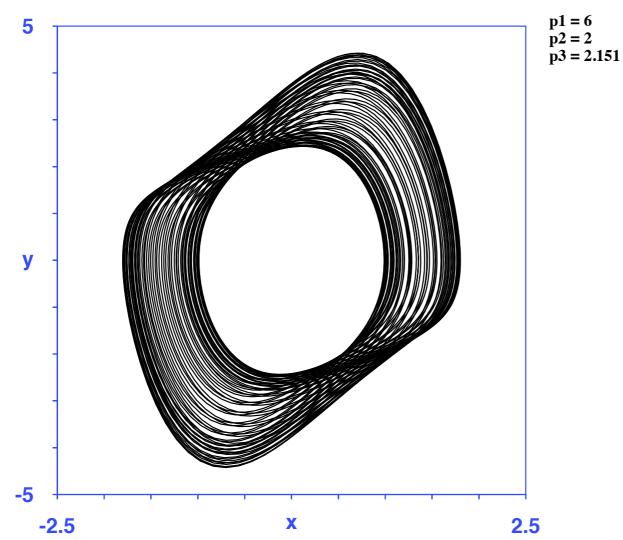
The Big, Big Picture (Bifurcations II) ...

Torus attractor in 3D Flow: Driven van der Pol:

$$\ddot{x} + \mu (x^2 - 1)\dot{x} + x = A\sin(\omega t)$$







3D Flows: Chaos

recurrent instability

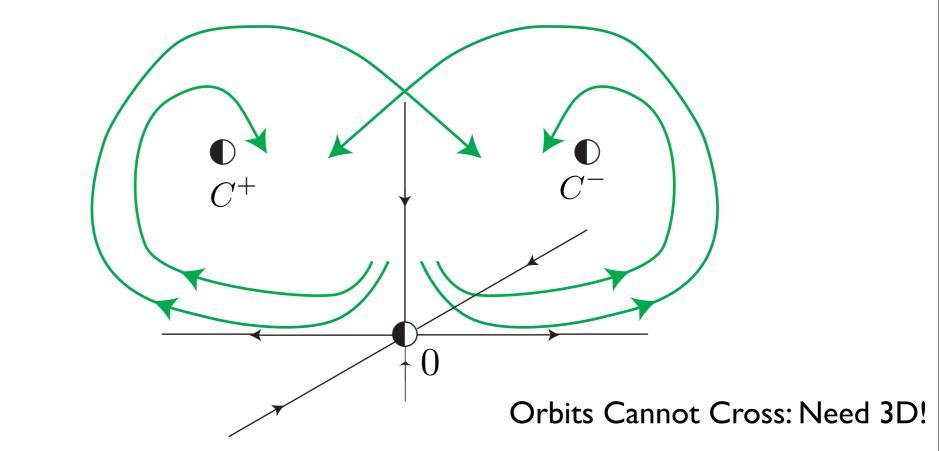
one way to do this: Orbit reinjection near unstable fixed point

not possible in lower D flows

a new behavior type

3D Flows: Chaos ...

A topological construction: saddle fixed point at origin: 0ID unstable manifold: $\dim(W^u(\mathbf{0})) = 1$ 2D stable manifold: $\dim(W^s(\mathbf{0})) = 2$ two fixed points: $C^+ \& C^-$



Does any ODE implement this flow design?

Example Dynamical Systems ...

3D Flows: Chaos ... Does any ODE implement this design? Yes, the Lorenz equations:

$$\dot{x} = \sigma(y - x)$$

 $\dot{y} = rx - y - xz$
 $\dot{z} = xy - bz$

Parameters: $\sigma, r, b > 0$

Exercise: Show fixed point at the origin can be a saddle, with 2 stable and 1 unstable directions Exercise: Show there is a symmetry $(x, y) \rightarrow (-x, -y)$

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Example Dynamical Systems ...

3D Flows: Chaos ...

Lorenz ODE properties:

Trajectories stay in a bounded region near origin No stable fixed points or stable limit cycles inside region Volume shrinks to zero (everywhere inside):

$$\begin{split} \dot{V} &= \int dV \ \nabla \cdot \vec{F}(\vec{x}) & \nabla &= (\partial/\partial x, \partial/\partial y, \partial/\partial z) \\ \nabla \cdot \vec{F}(\vec{x}) &= \operatorname{Tr}(A) = -\sigma - 1 - b \\ \dot{V} &= -(\sigma + 1 + b)V \\ V(t) &= e^{-(\sigma + 1 + b)t} & \underset{\text{exponentially fast!}}{\text{Region volume shrinks}} \quad \text{e.g.} : \sigma + 1 + b \approx 10 \end{split}$$

What does the invariant set look like?

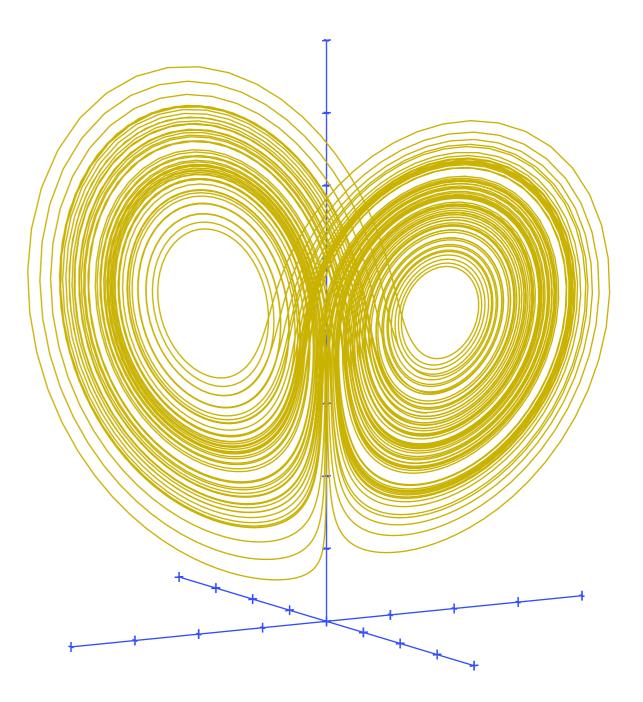
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3D Flows: Chaos ... Lorenz simulation demo: Chaotic attractor: $(\sigma, r, b) = (10, 28, 8/3)$

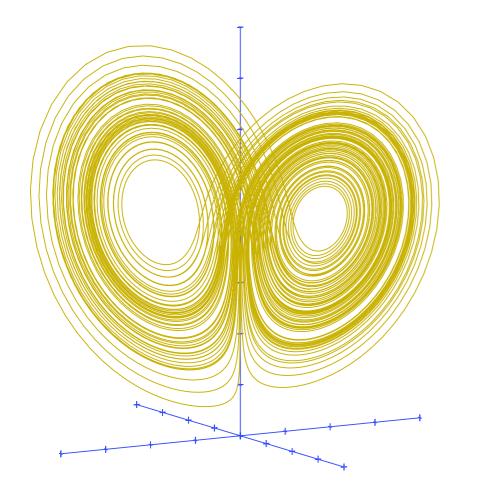
Demo:

Time step = 0.001Parameters IC = (3,3,3)3D view, orientation len = 100K

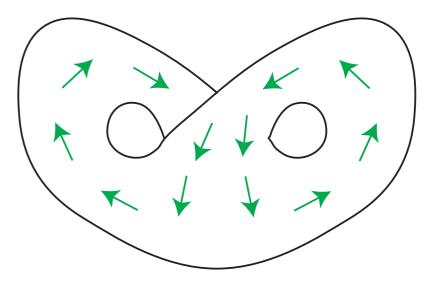


3D Flows: Chaos ...

Lorenz attractor structure



Branched manifold

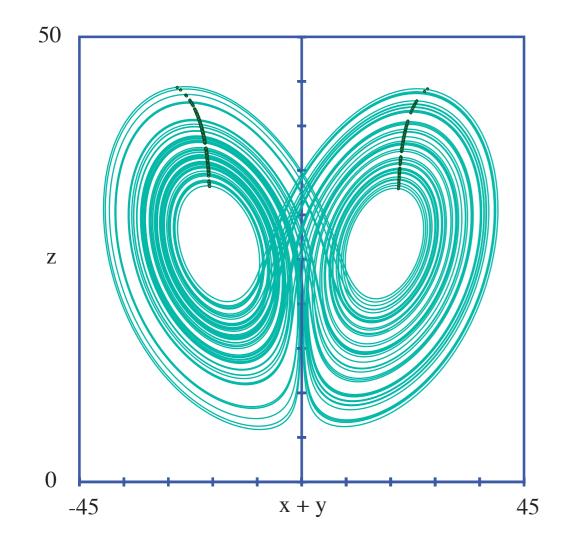


3D Flows: Chaos ... Lorenz simulation demo: fixed point: $(\sigma, r, b) = (10, 15, 8/3)$

> Demo: Time step = 0.0001Parameters IC = (3,3,3)3D view, orientation len = 100K

Example Dynamical Systems ...

From Continuous-Time Flows to Discrete-Time Maps:

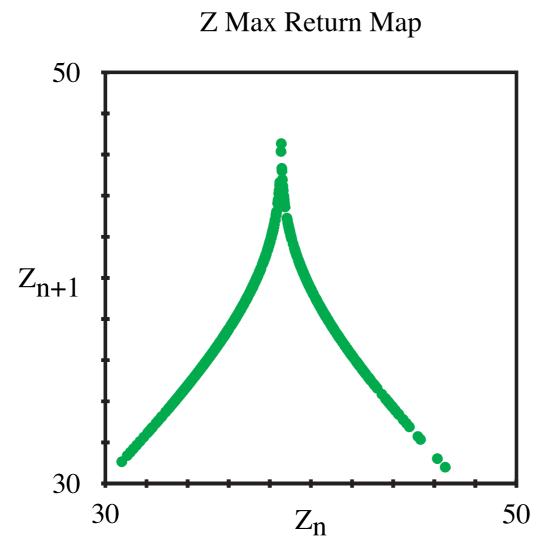


Series of z-maxima: $\hat{z}_1, \hat{z}_2, \hat{z}_3, \dots$ What happens if you plot \hat{z}_{n+1} versus \hat{z}_n ?

Example Dynamical Systems ...

From Continuous-Time Flows to Discrete-Time Maps:

Max-z Return Map: $z_{n+1} = f(z_n)$

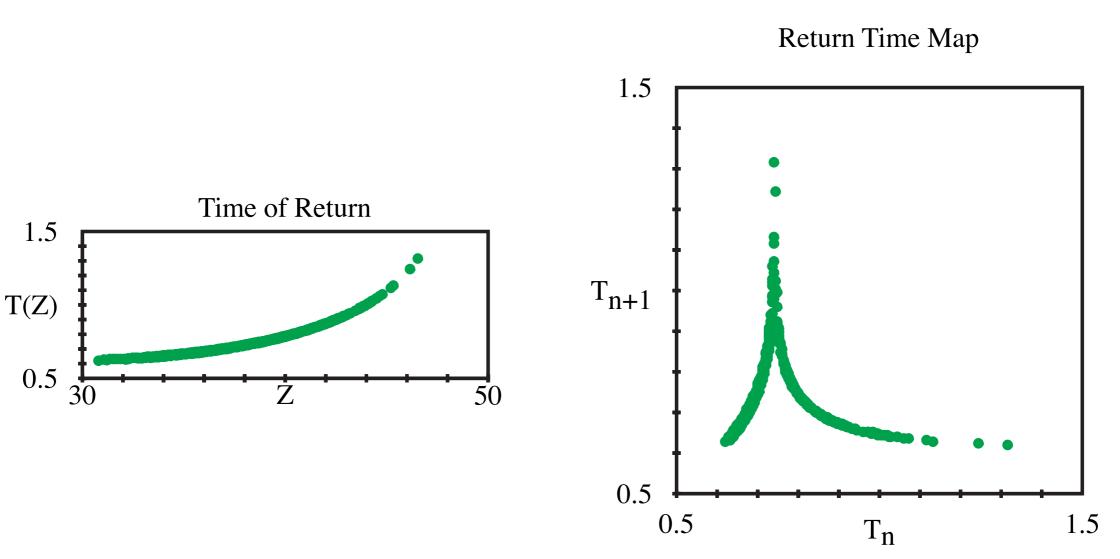


Example Dynamical Systems ...

From Continuous-Time Flows to Discrete-Time Maps:

Time of Return Function: $T(z_n)$

Return Time Map: $T_{n+1} = h(T_n)$



Example Dynamical Systems ...

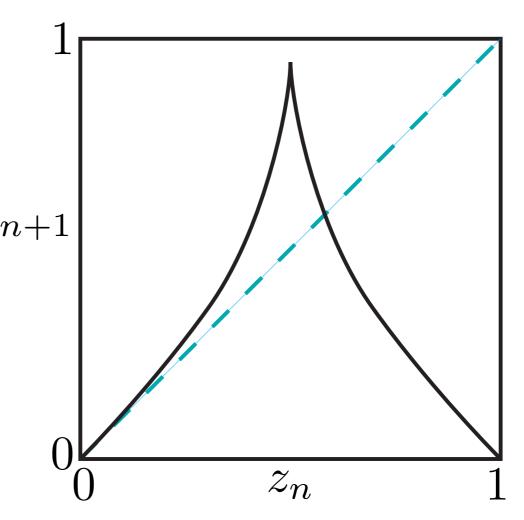
3D Flows ...

Lorenz reduces to a cusp ID map: normalize to $z_n \in [0, 1]$

$$z_{n+1} = a(1 - |1 - 2z_n|^b)$$

Parameters:

height: a > 0peak sharpness: 0 < b < 1 z_{n+1}



Example Dynamical Systems ...

3D Flows ... Rössler equations

$$\dot{x} = -y - z$$
$$\dot{y} = x + ay$$
$$\dot{z} = b + z(x - c)$$

Parameters: a, b, c > 0

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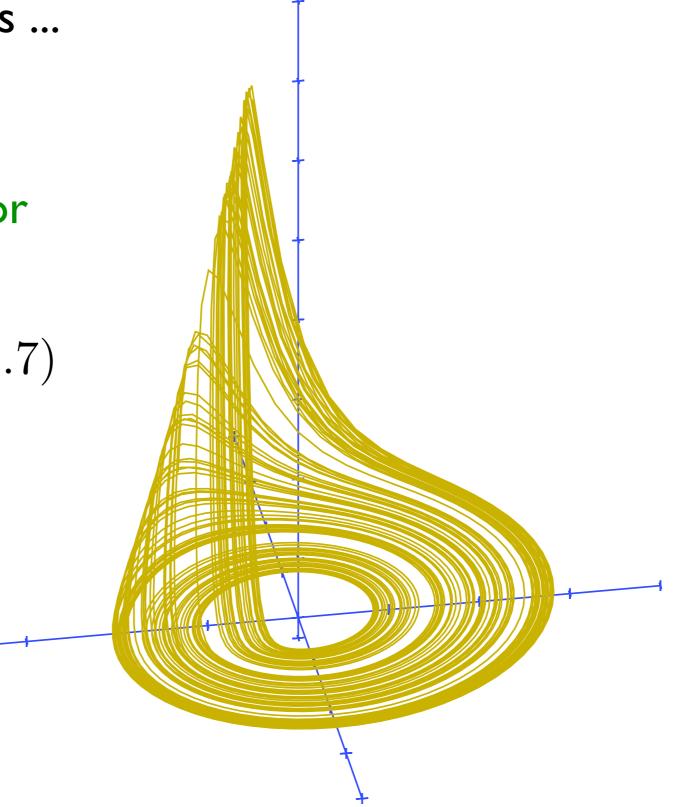
3D Flows ... Rössler chaotic attractor

Parameters:

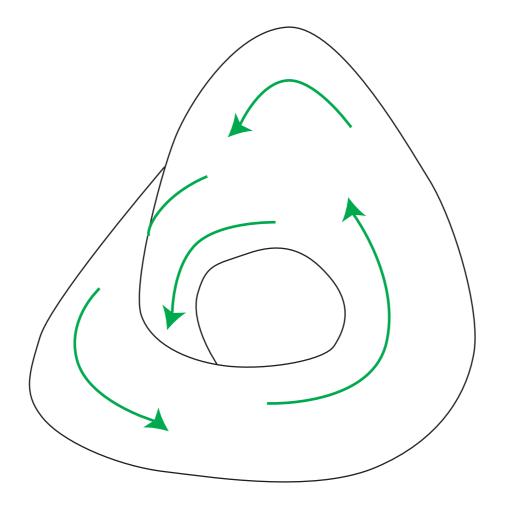
$$(a, b, c) = (0.2, 0.2, 5.7)$$

Demo:

Time step = 0.01Parameters IC = (3,3,3)3D view, orientation len = 100K



3D Flows ... Rössler branched manifold



3D Flows ... Rössler limit cycle attractor

Parameters:

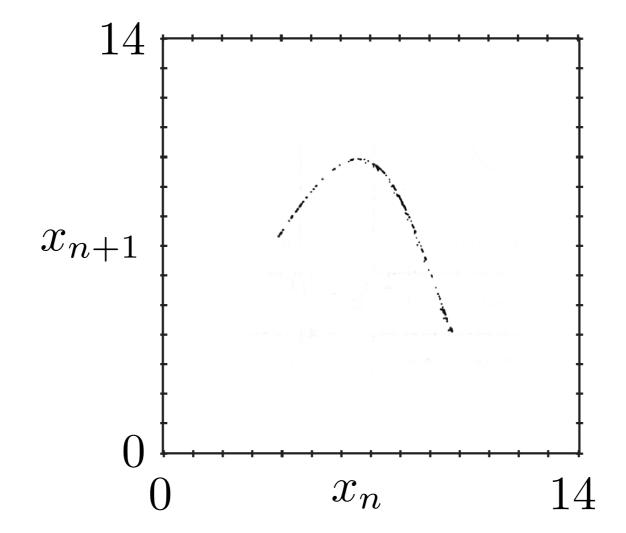
$$(a, b, c) = (0.2, 0.2, 2.0)$$

Demo:

Time step = 0.01Parameters IC = (3,3,3)3D view, orientation len = 100K

Example Dynamical Systems ...

3D Flows ... Rössler maximum-x return map: $x_{n+1} = f(x_n)$



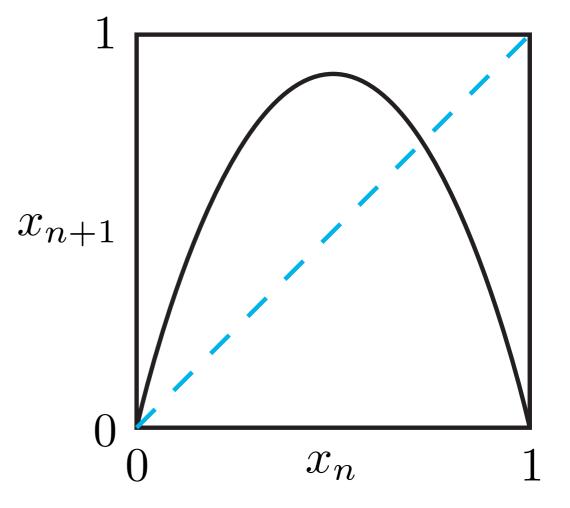
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Example Dynamical Systems ...

3D Flows ... When normalized to $x_n \in [0, 1]$ get the Logistic Map:

$$x_{n+1} = rx_n(1 - x_n)$$





Example Dynamical Systems ...

Classification of Possible Behaviors

Dimension	Attractor
	Fixed point
2	Fixed point, Limit cycle
3	Fixed Point, Limit Cycle, Torus, Chaotic
4	Above + Hyperchaos
5	?

Lorenz:
$$\dot{x} = \sigma(y - x)$$
 $\sigma, r, b > 0$
 $\dot{y} = rx - y - xz$
 $\dot{z} = xy - bz$

Rössler: $\dot{x} = -y - z$ $\dot{y} = x + ay$ $\dot{z} = b + z(x - c)$

Cusp Map: $z_n \in [0,1]$ a > 0, 0 < b < 1 $z_{n+1} = a(1 - |1 - 2z_n|^b)$

Logistic map:

$$x_{n+1} = rx_n(1 - x_n) \qquad x_n \in [0, 1] \qquad r \in [0, 4]$$

Play with these!

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The Big Picture

Global view of the state space structures: The attractor-basin portrait

Reading for next lecture:

NDAC, Chapter 3.