

# Dynamical Systems Tutorial

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PART 1 OF 2 – THE BASICS

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# What is a Dynamical System?

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**A way to take a state and predict its future**

More formally:

- A state space  $S$  of possible configurations (e.g. location, some vector, ...)
- A range of times  $t$  where it works
- A rule that takes a state  $x$  from  $S$  at time  $t$  and tells us how  $s$  changes
- We call this rule the dynamics (or vector field)

For us, the rule is a differential equation

A time course of a dynamical variable is called a trajectory

# What is a Dynamical System?

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A differential equation gives us the rate of change of a variable in time a function of that variable

For instance: We have a position, we get a velocity

Simplest example:  $\frac{dx}{dt} = -\frac{x}{\tau}$

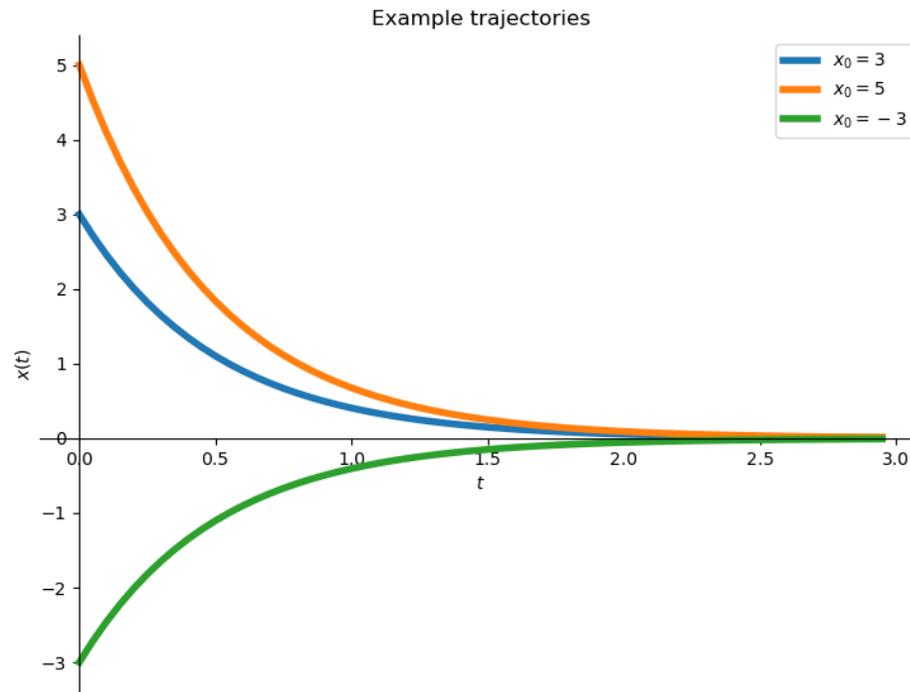
At the heart of the neuron model we use

We often use the abbreviation of the time derivative  $\dot{x}(t) = \frac{dx}{dt}$

# Separation of Variables

Procedure to solve simple differential equations

Not technically quite clean, but works anyway



$$\frac{dx}{dt} = -\frac{x}{\tau}$$

$$\frac{dx}{x} = -\frac{1}{\tau} dt$$

$$\int \frac{dx}{x} = \int -\frac{1}{\tau} dt$$

$$[\log(x)]_{x_0}^x = \left[ -\frac{t}{\tau} \right]_0^t$$

$$\log(x) - \log(x_0) = -\frac{t}{\tau}$$

$$\frac{x}{x_0} = \exp\left(-\frac{t}{\tau}\right)$$

$$x(t) = x_0 \exp\left(-\frac{t}{\tau}\right)$$

# Numerical Methods

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Used to simulate systems

Always have an error

Simplest method: Euler step

Sample time discretely:  $t_i$  with  $i \in \{1, 2, \dots, N\}$

Then  $t_i = i\Delta t$

Approximate change  $\Delta x_i$  during step  $\Delta t$  via derivative

$$\frac{\Delta x(t_i)}{\Delta t} \approx \left. \frac{dx}{dt} \right|_{t=t_i} = f(x(t_i), t_i)$$

$$x(t_{i+1}) = x(t_i) + \Delta x(t_i) \approx x(t_i) + \Delta t f(x(t_i), t_i)$$

# Different form of Dynamical Systems

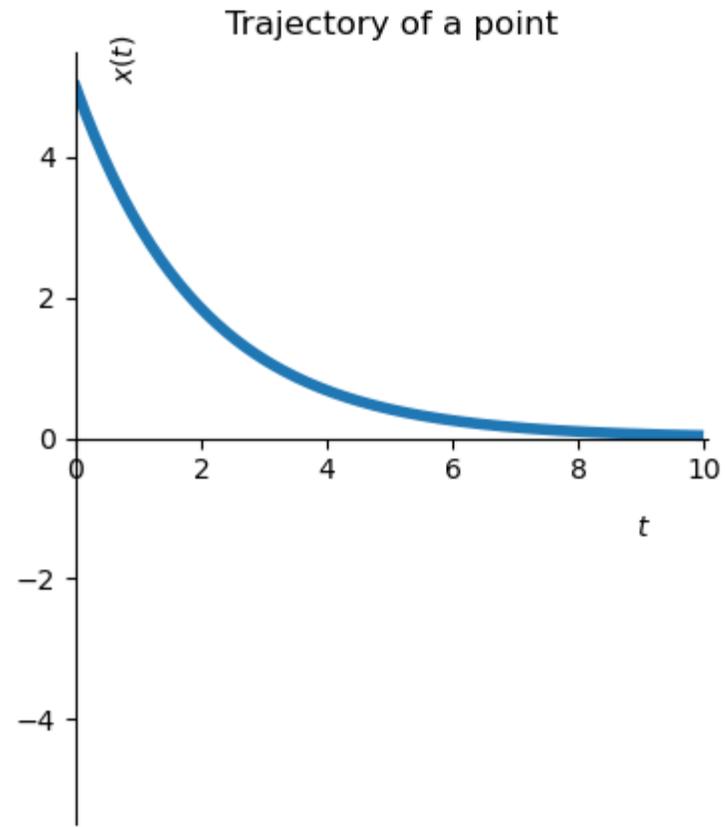
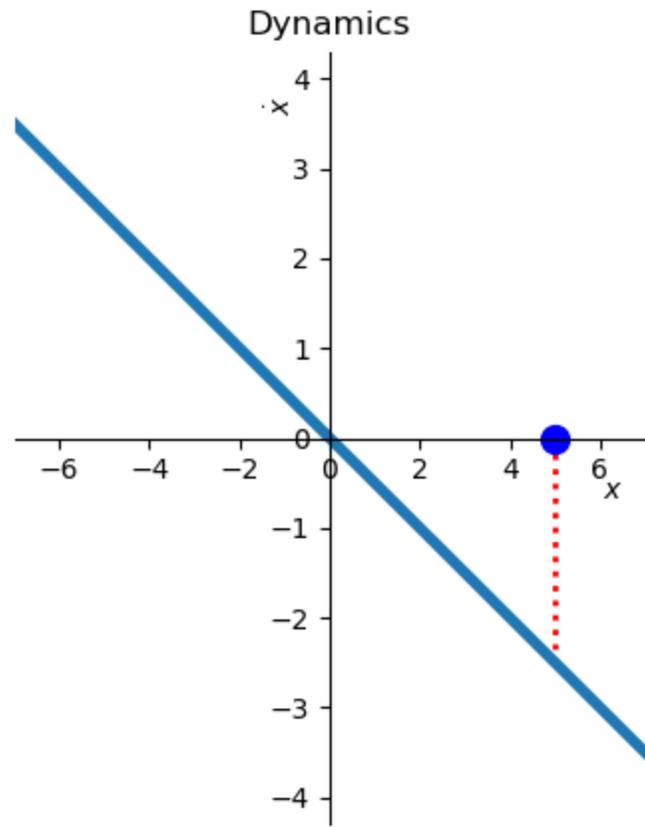
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We saw an example of a one-dimensional differential equation

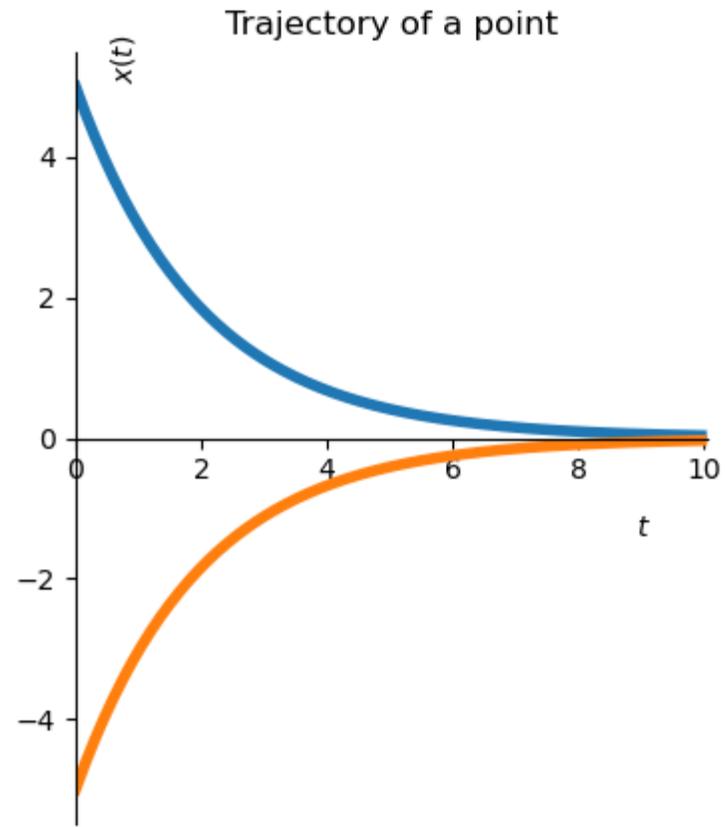
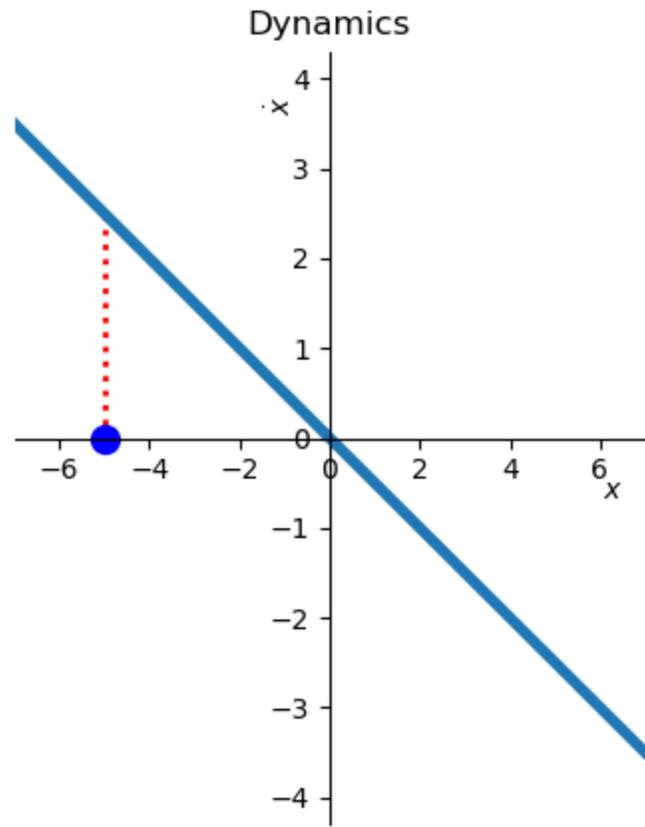
There are also

- Vector valued (N-dimensional) equations
- integro-differential equations
- partial differential equations
- functional differential equations
- delay differential equations
- ...

$$\frac{dx}{dt} = -\frac{x}{\tau}$$



$$\frac{dx}{dt} = -\frac{x}{\tau}$$



# Fixed Points

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When the rate of change is zero we have, of course, no change

Our system is in balance!

But is that balance stable?

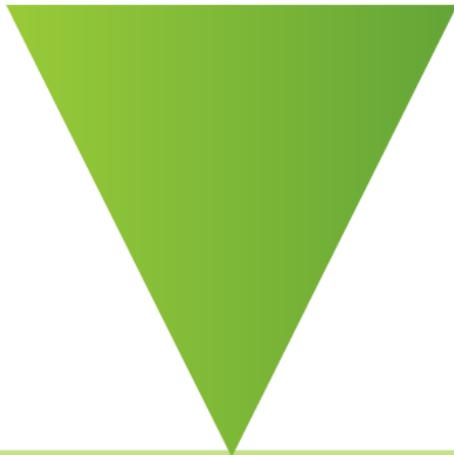
$$\dot{x} = 0$$

# Stability

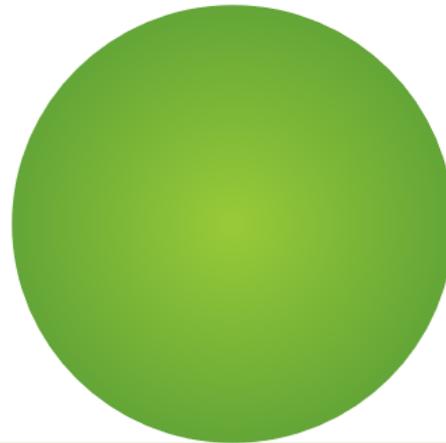
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How does the system react if you disturb it slightly?

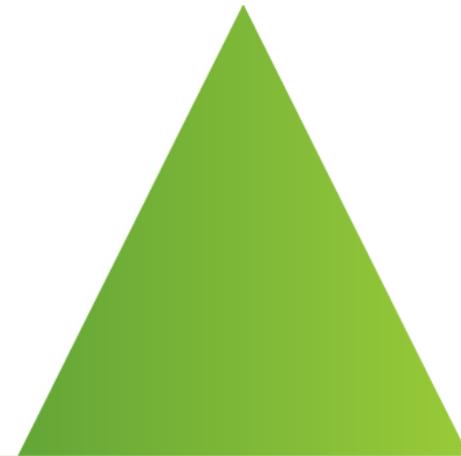
- Moves away -> unstable
- Stays in the vicinity -> stable
- Goes back to fixed point -> asymptotically stable



Unstable

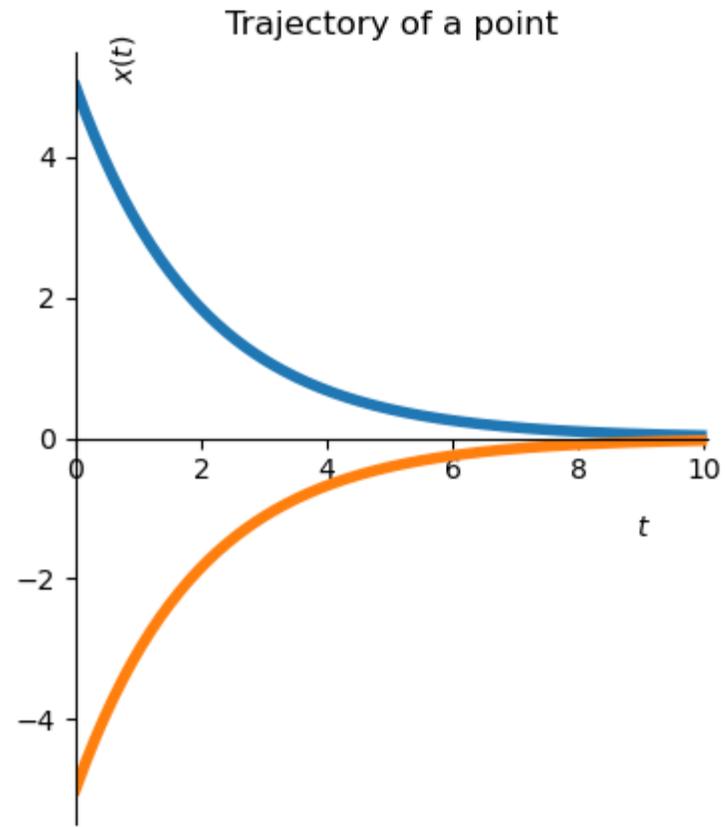
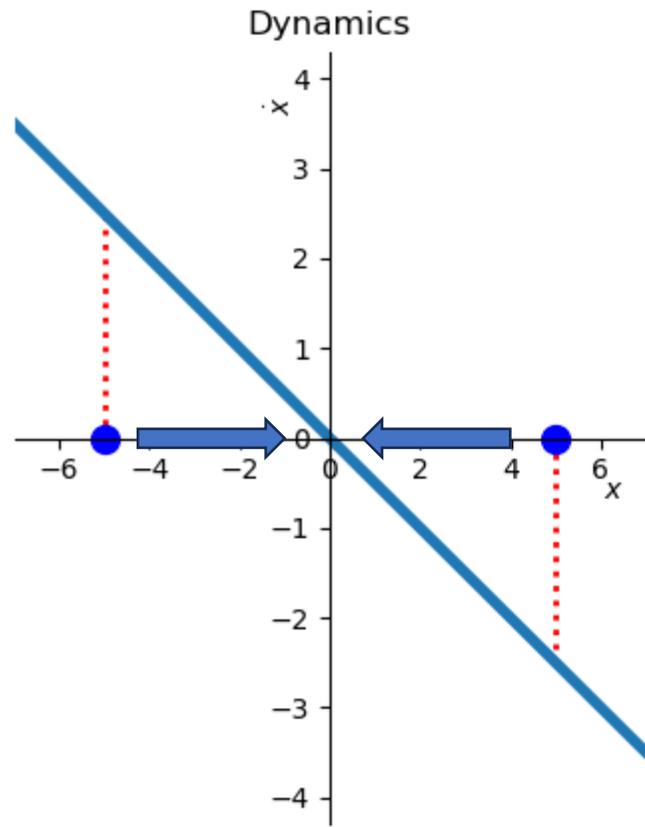


Stable

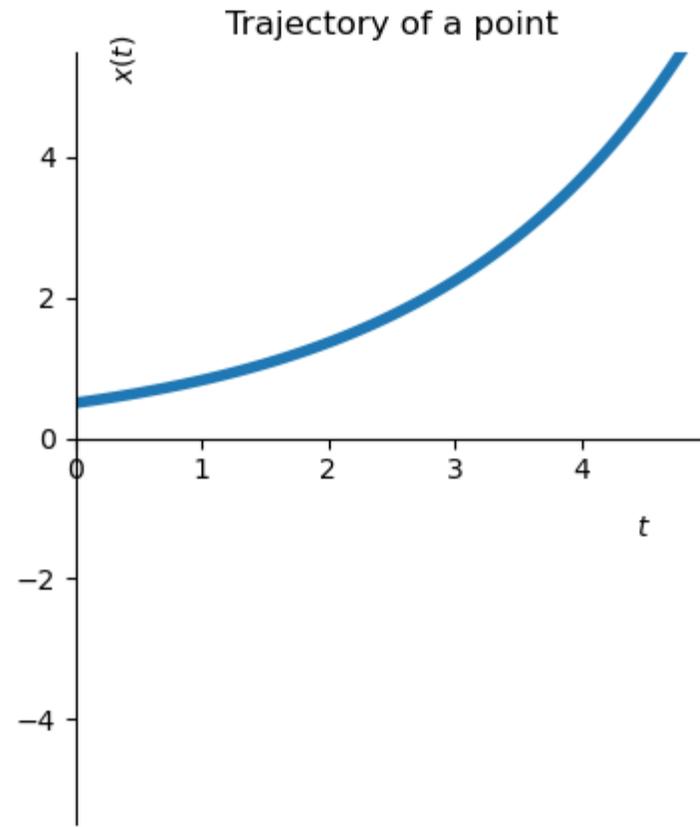
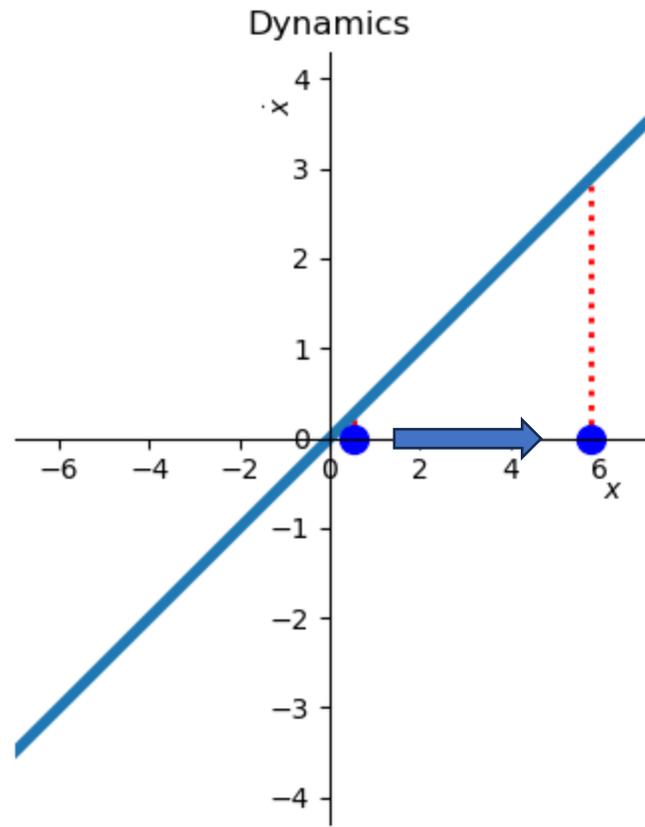


Asymptotically stable

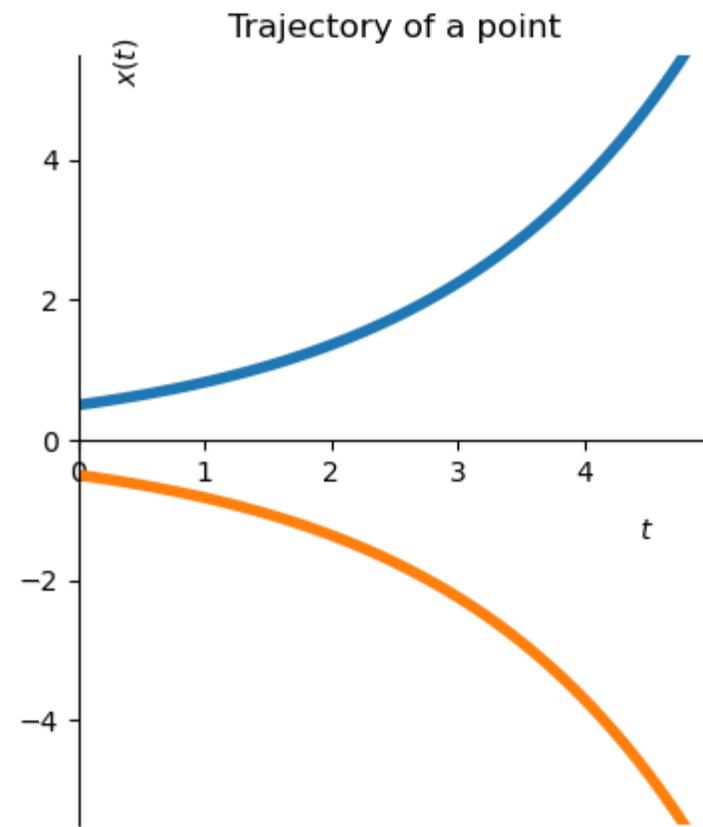
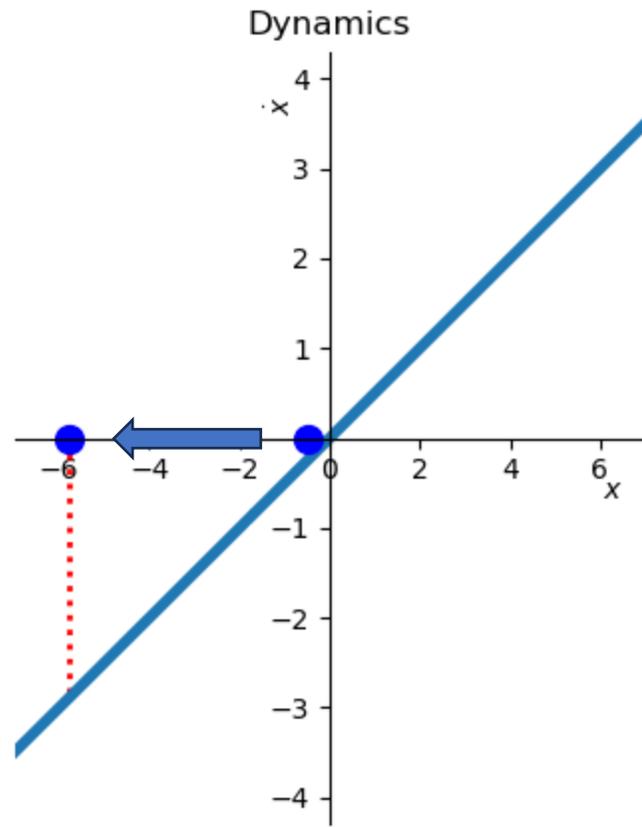
$$\frac{dx}{dt} = -\frac{x}{\tau}$$



$$\frac{dx}{dt} = \frac{x}{\tau}$$



$$\frac{dx}{dt} = \frac{x}{\tau}$$



# Attractor

# Repellor

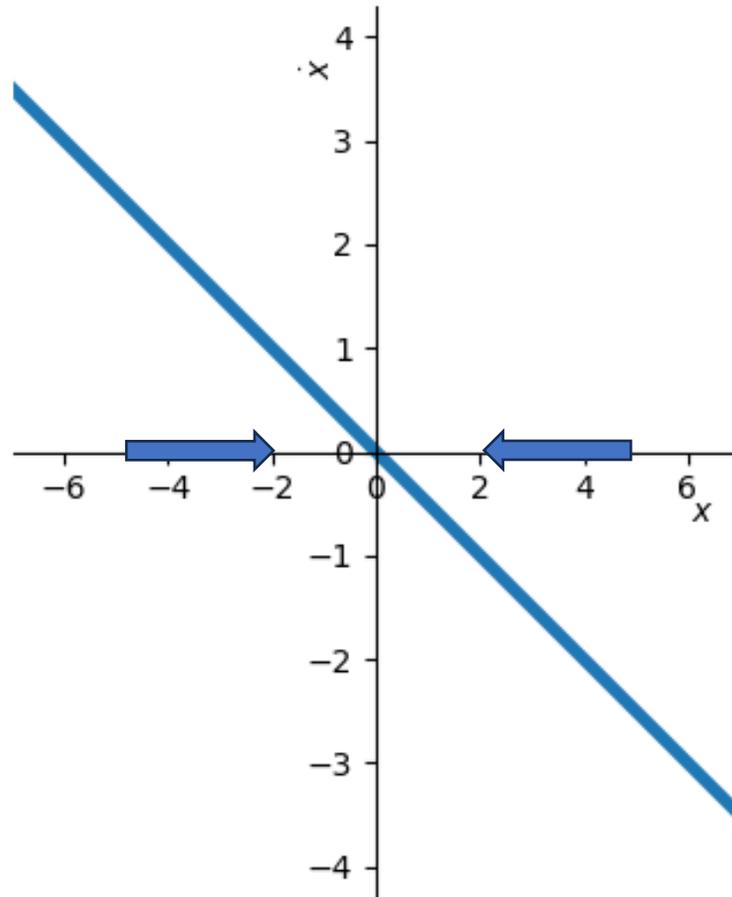
## Stability

For linear system: Look at slope at fixed point

- Negative slope -> stable
- Positive slope -> unstable

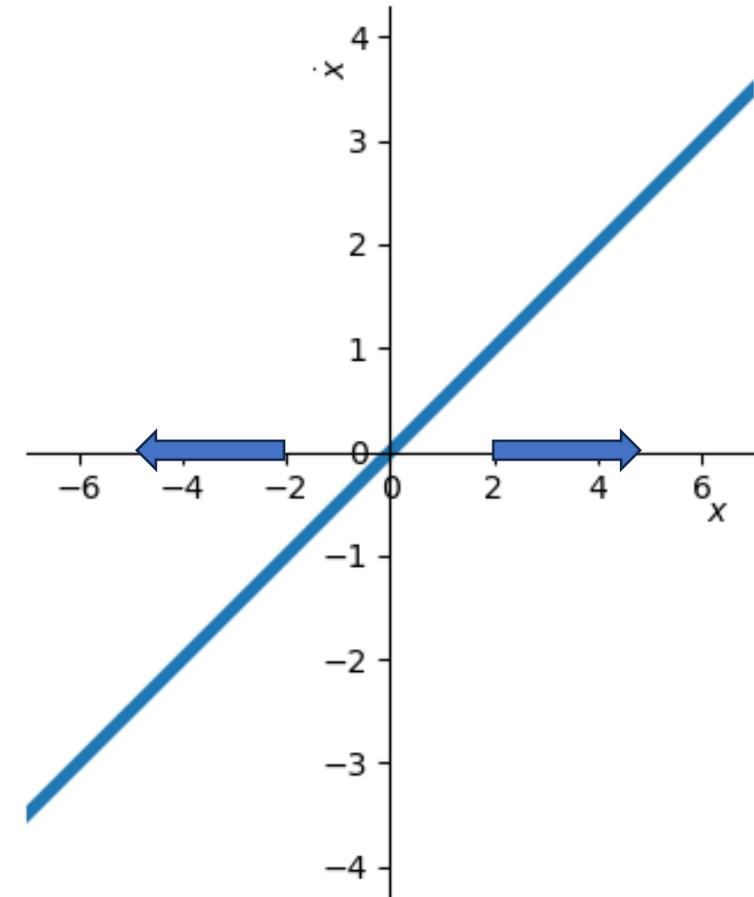
In practice, there is always noise pushing us away from reppelors

Stable fixed point



$$\frac{dx}{dt} = -\frac{x}{\tau}$$

Unstable fixed point



$$\frac{dx}{dt} = \frac{x}{\tau}$$

# Linear Approximation

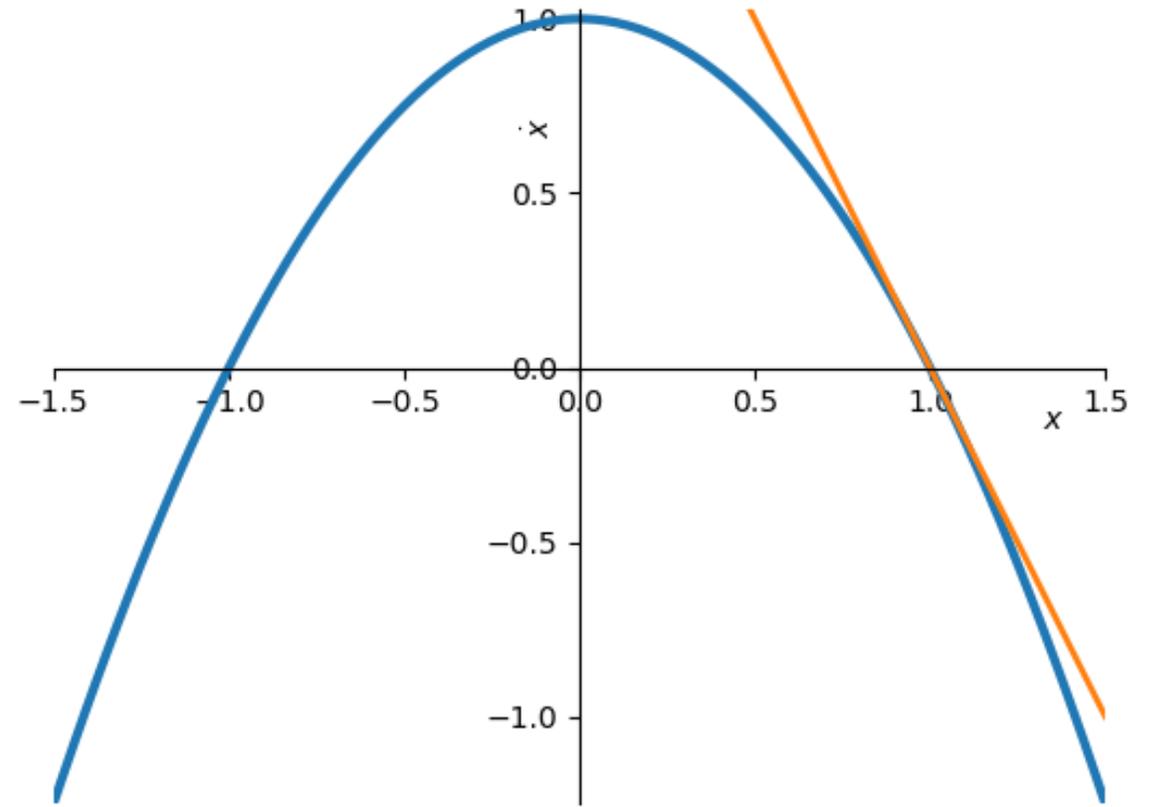
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We mostly only understand linear systems well

What to do with non-linear problems?

➤ Make it linear!

We can still use the sign of the derivative

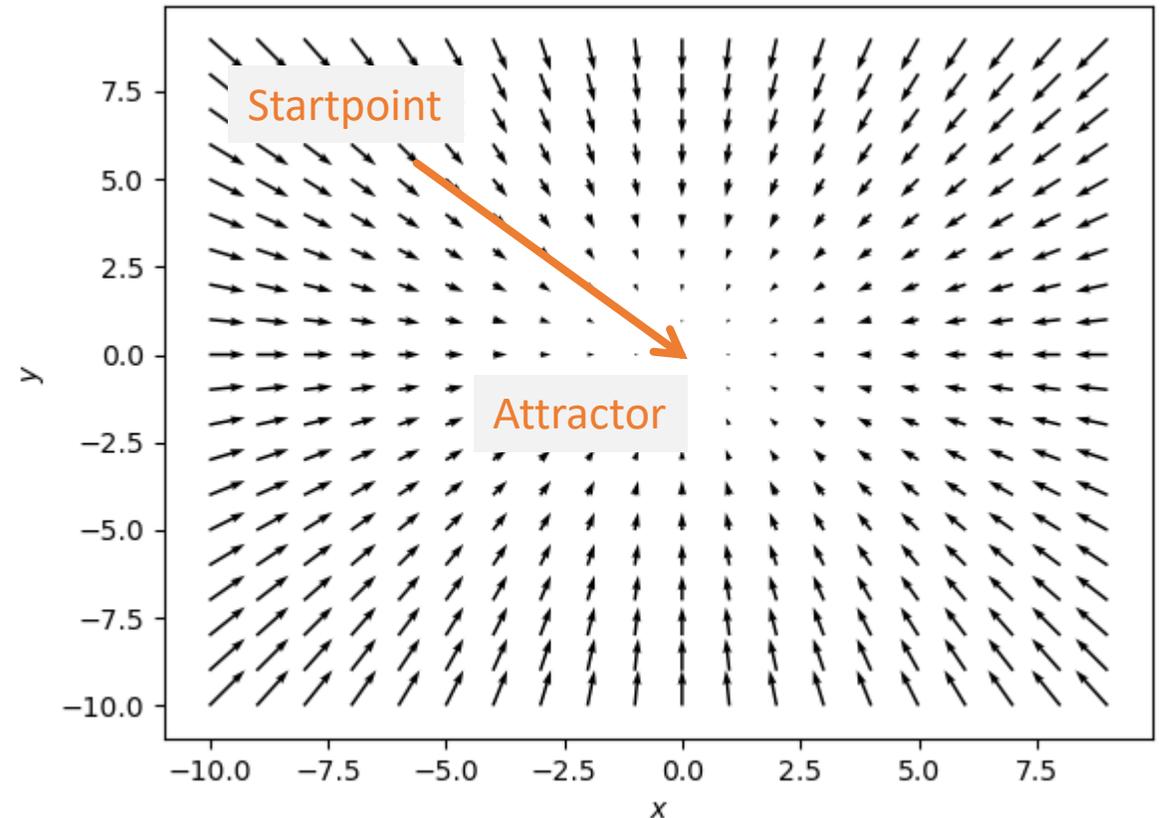


# Linear Multidimensional Systems

$$\dot{\vec{x}} = M\vec{x}$$

Stability depends on eigenvalues of M

- if the real-parts of all Eigenvalues are negative: stable
- if the real-part of any Eigenvalue is positive: unstable
- if the real-part of any Eigenvalue is zero: marginally stable in that direction (stability depends on other eigenvalues)



# Dynamical Systems Tutorial

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PART 2 OF 2 - BIFURCATIONS

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# What is a bifurcation

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We have a dynamical system with a parameter  $c$

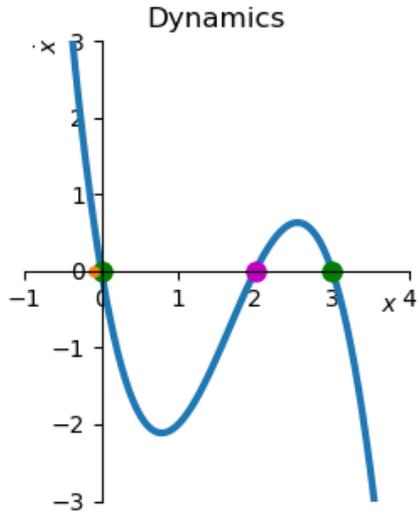
As  $c$  changes smoothly, the behavior of the system as an abrupt change

Technically: Infinitesimal parameter change make for topologically inequivalent systems

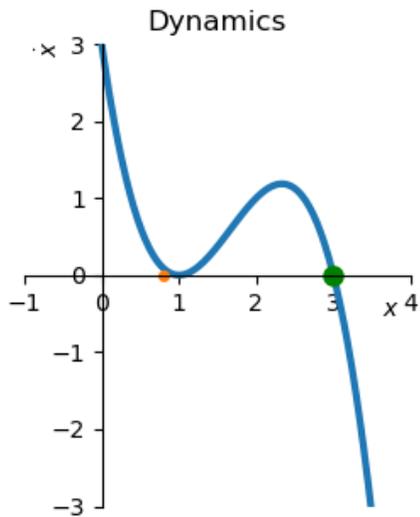


# Tangent Bifurcation

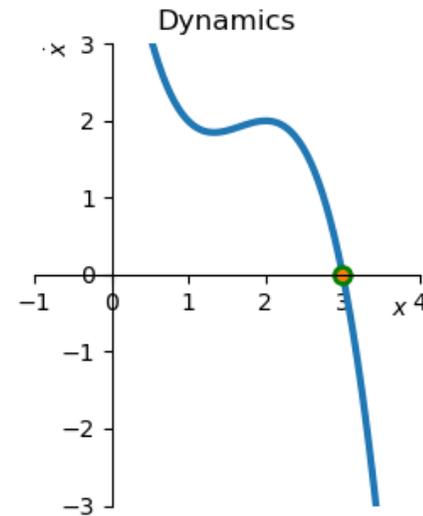
$$\dot{x} = (-(x - 1)^2 - c)(x - 3)$$



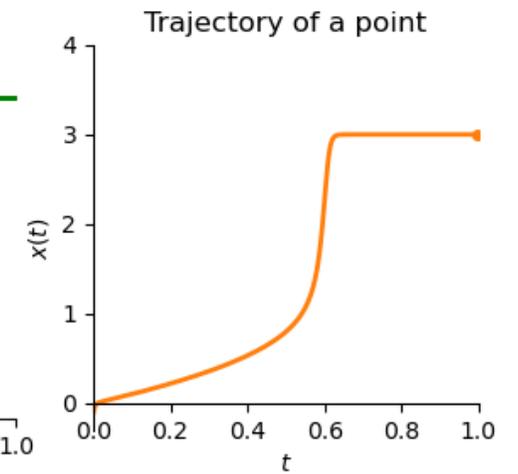
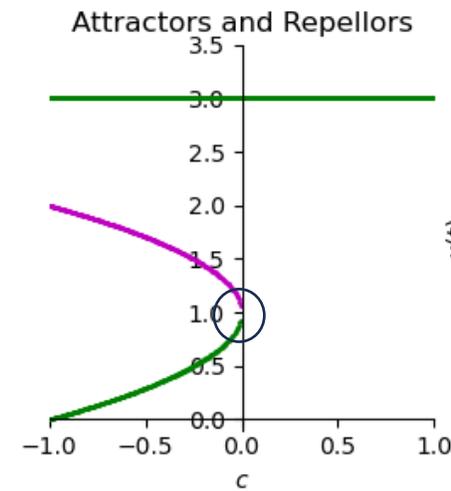
$t=0$   
 $c=-1$



$t=0.5$   
 $c=0$



$t=1$   
 $c=1$



# Hopf Theorem

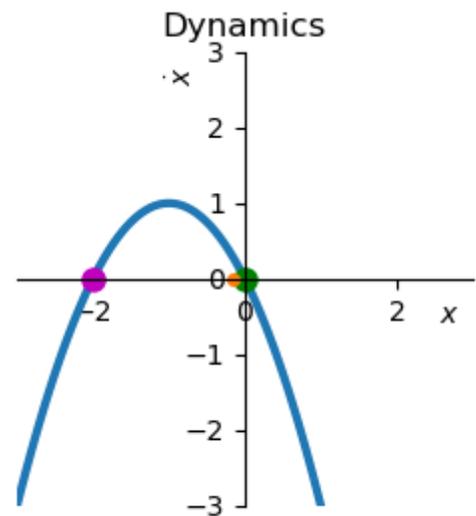
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When a single (or pair of complex conjugate) eigenvalue crosses the imaginary axis, one of four bifurcations occur

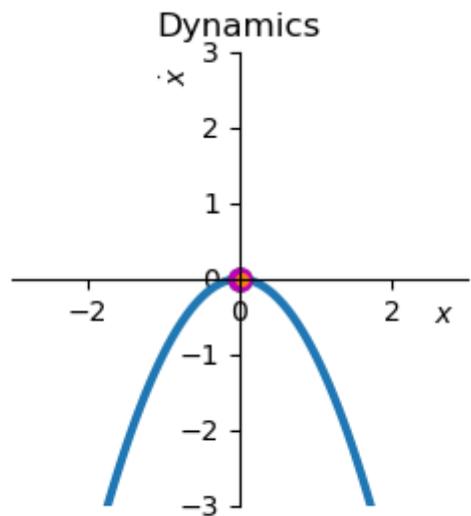
- tangent bifurcation
- transcritical bifurcation
- pitchfork bifurcation
- Hopf bifurcation

# Transcritical Bifurcation

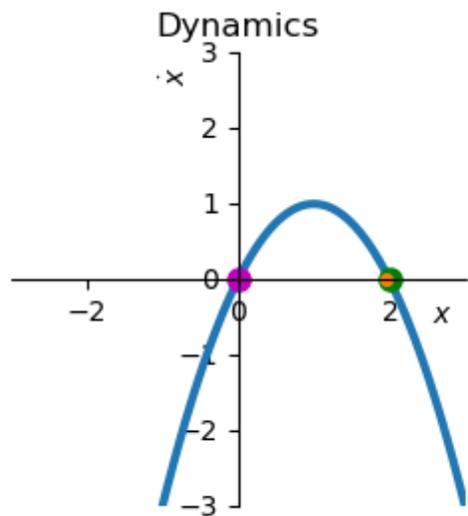
$$\dot{x} = \alpha x - x^2$$



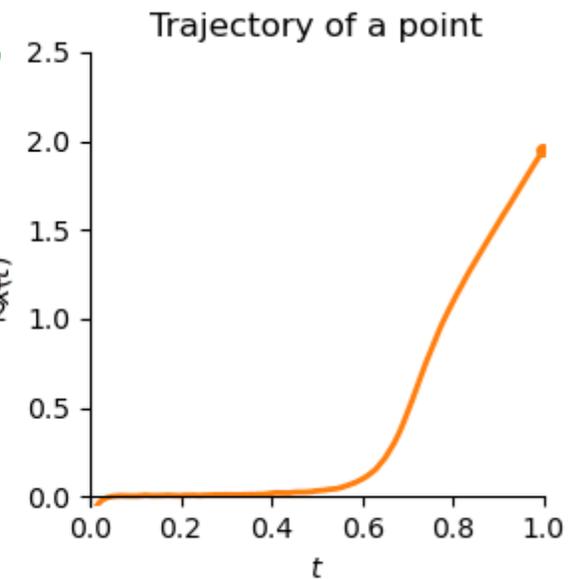
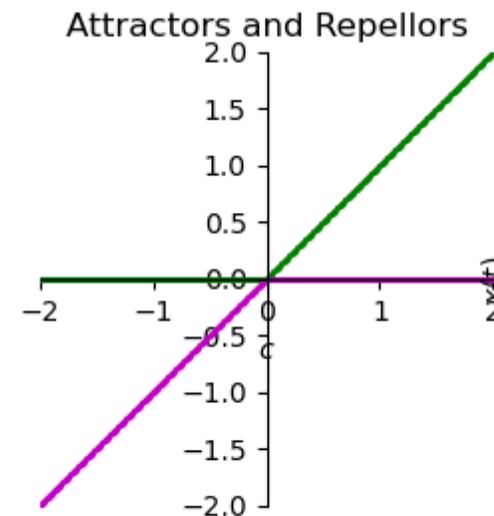
$t=0$   
 $c=-1$



$t=0.5$   
 $c=0$

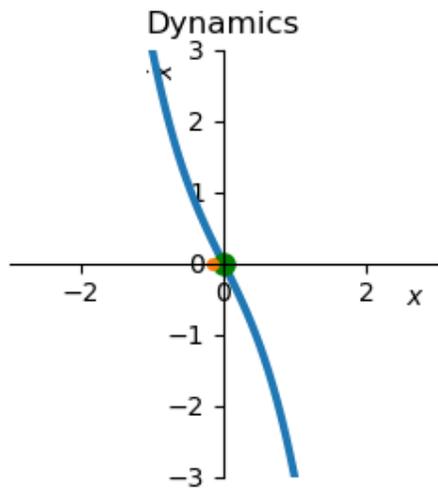


$t=1$   
 $c=1$

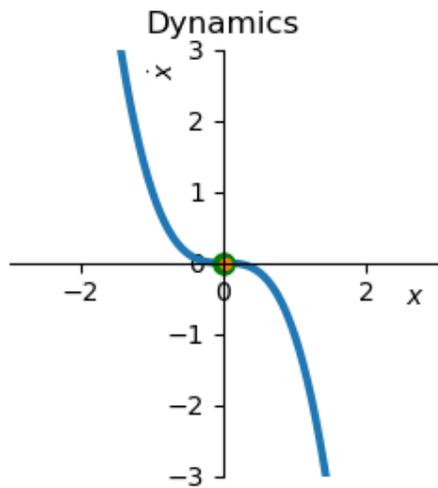


# Pitchfork Bifurcation

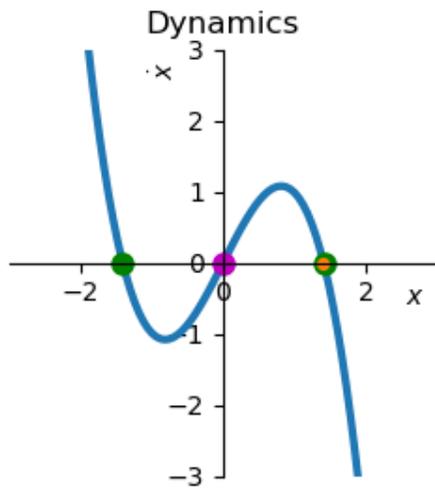
$$\dot{x} = cx - x^3$$



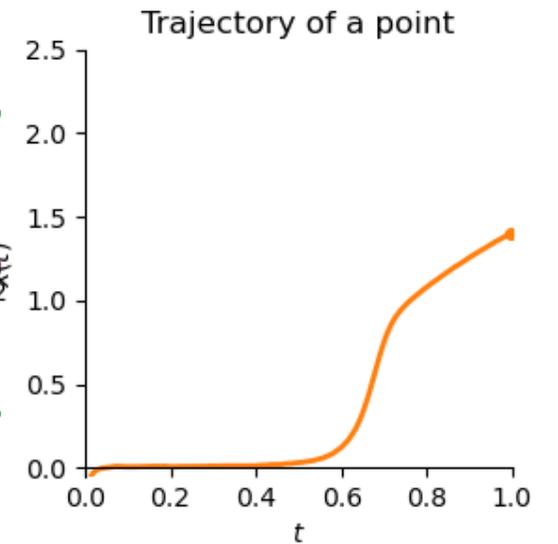
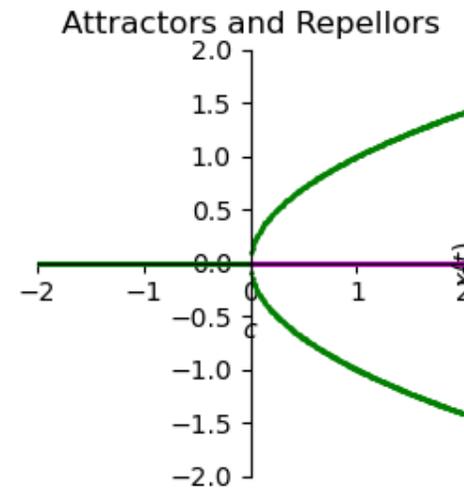
$t=0$   
 $c=-1$



$t=0.5$   
 $c=0$

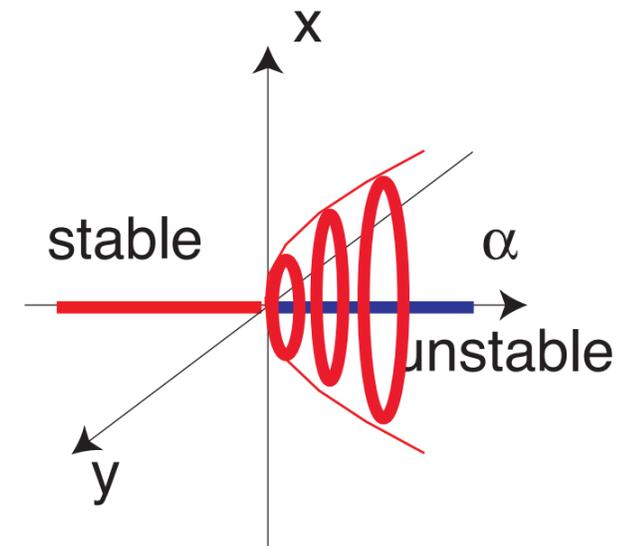
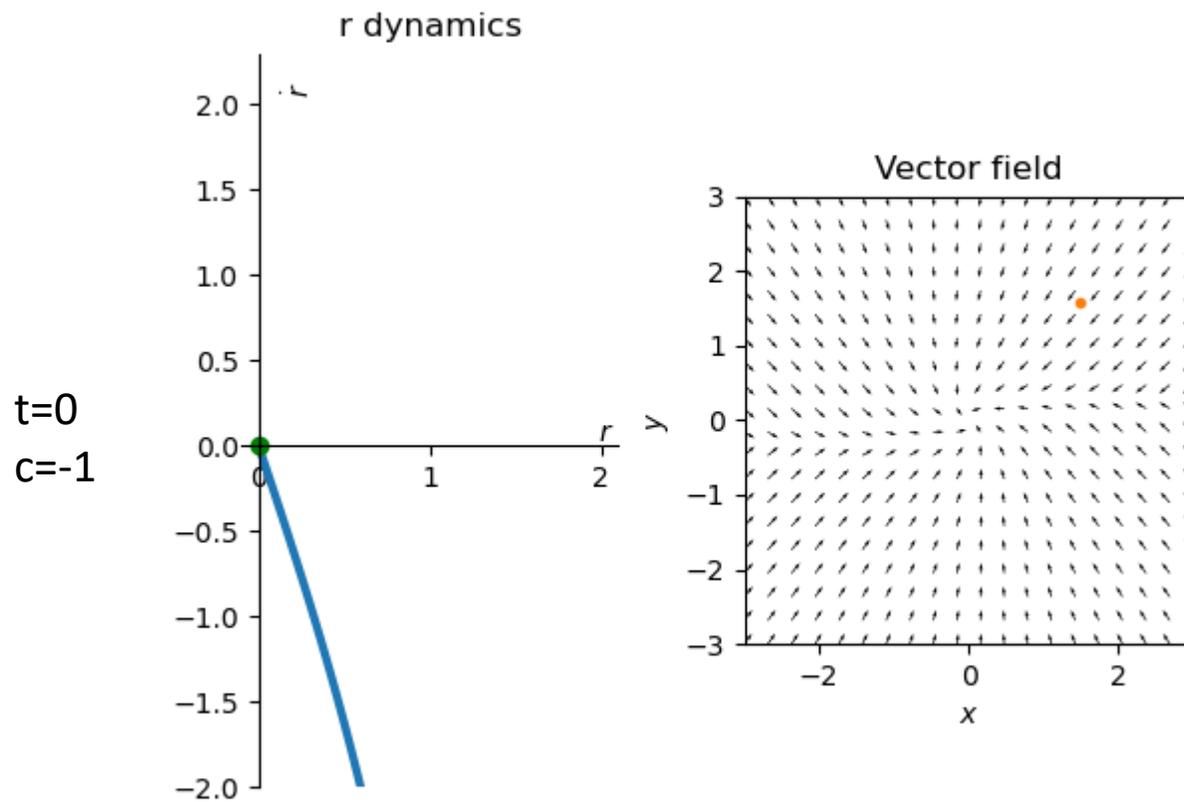


$t=1$   
 $c=1$



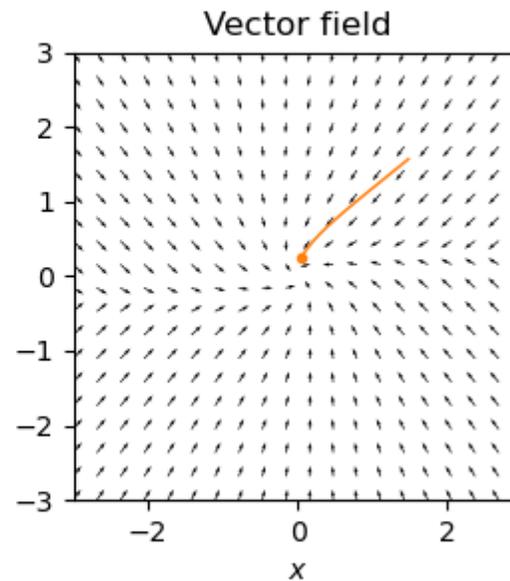
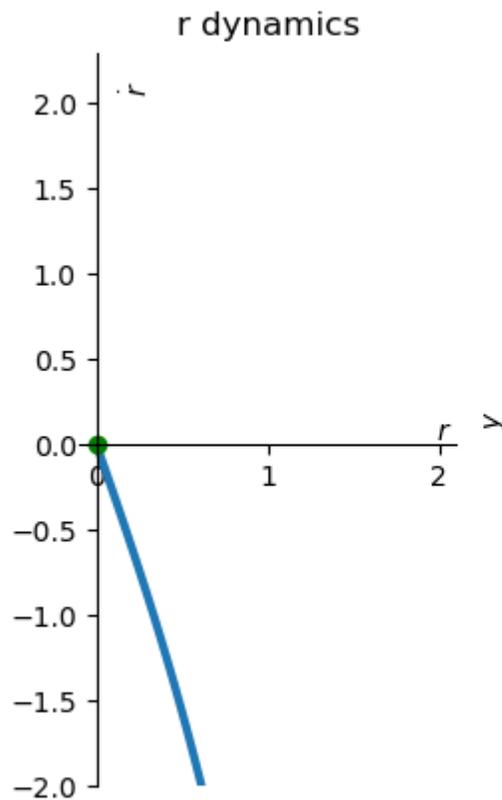
# Hopf Bifurcation

$$\begin{aligned}\dot{r} &= \alpha r - r^3 \\ \dot{\phi} &= \omega\end{aligned}$$

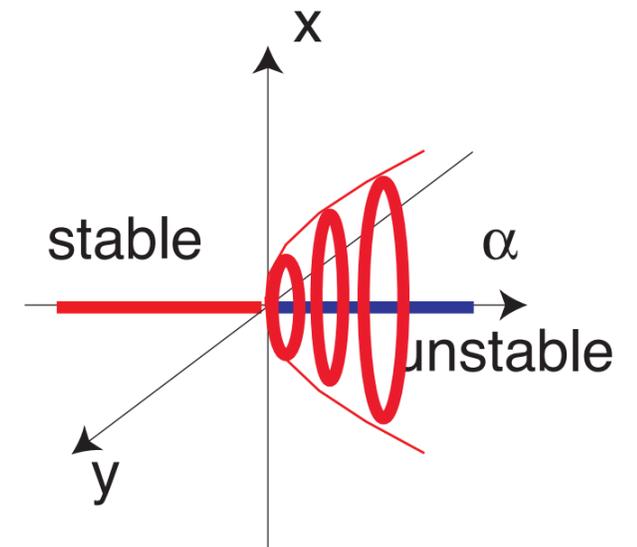


# Hopf Bifurcation

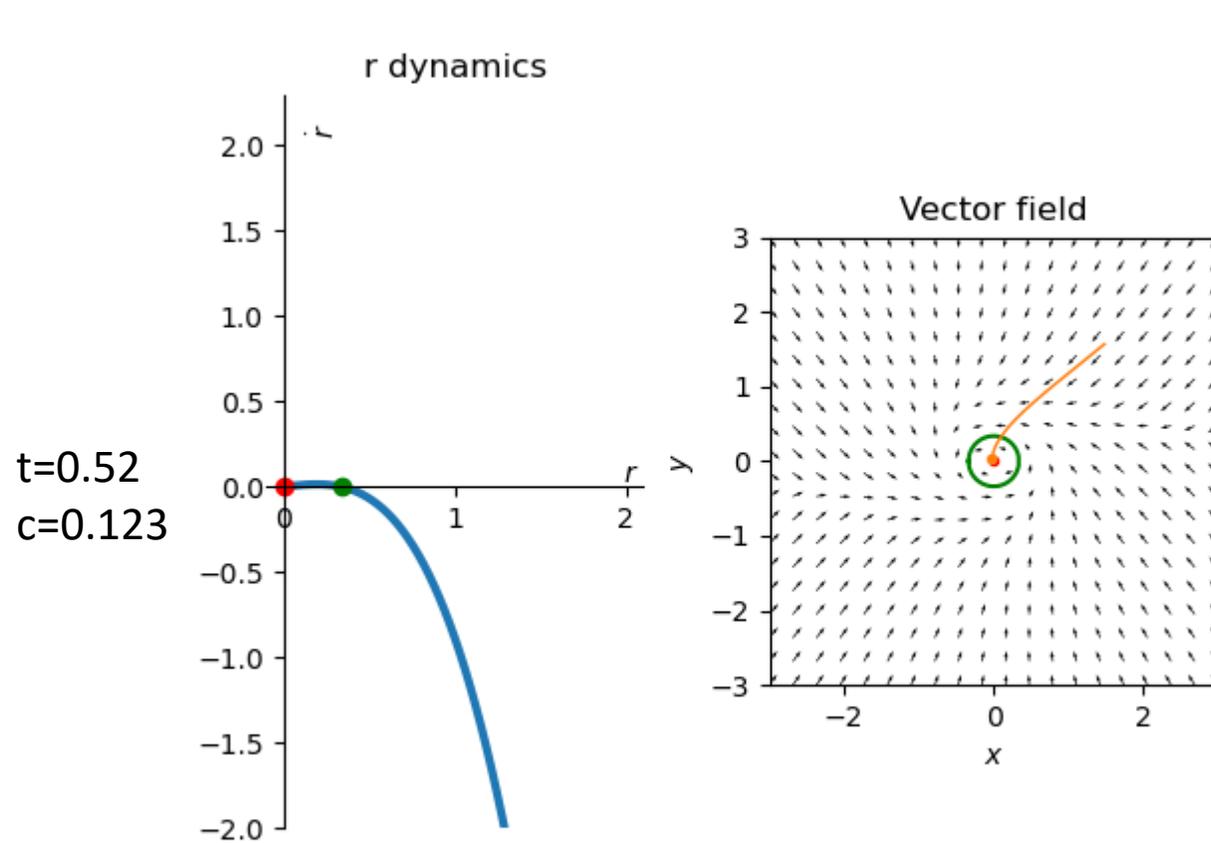
$t=0.02$   
 $c=-2.8$



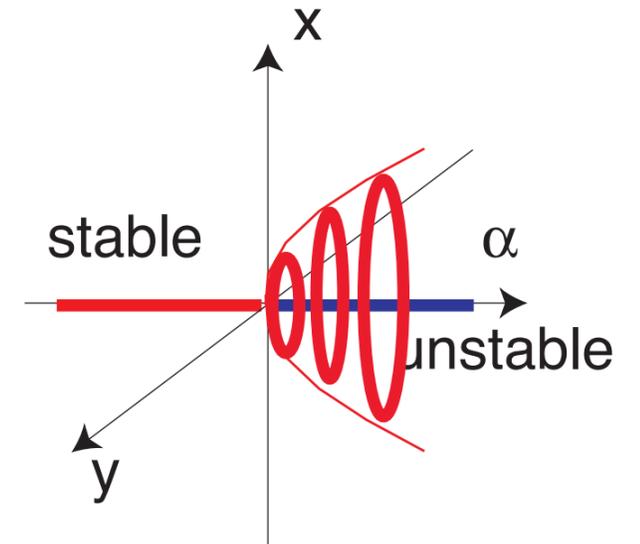
$$\dot{r} = \alpha r - r^3$$
$$\dot{\phi} = \omega$$



# Hopf Bifurcation

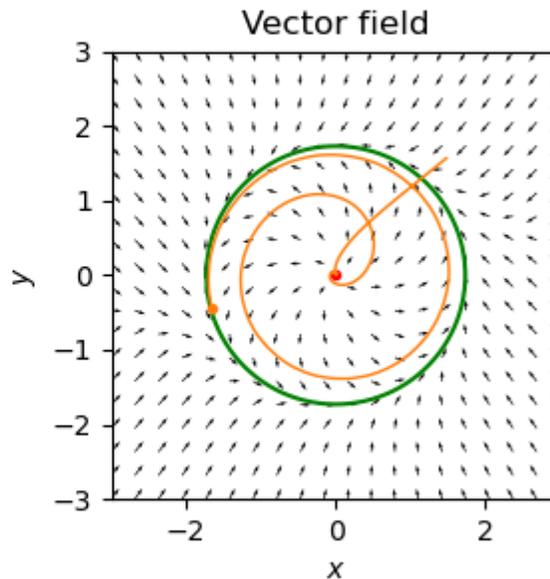
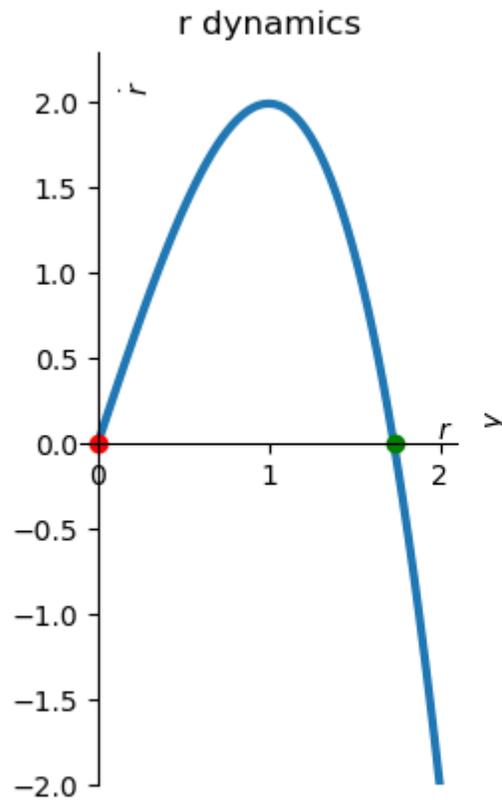


$$\dot{r} = \alpha r - r^3$$
$$\dot{\phi} = \omega$$

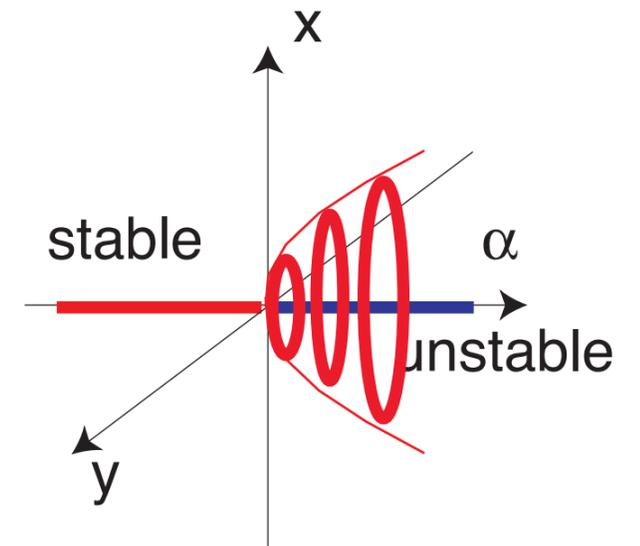


# Hopf Bifurcation

$t=1$   
 $c=3$



$$\begin{aligned}\dot{r} &= \alpha r - r^3 \\ \dot{\phi} &= \omega\end{aligned}$$



# Forward dynamics

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given known equation, determined fixed points / limit cycles and their stability

more generally: determine invariant solutions (stable, unstable and center manifolds)

Basically, what we covered here

# Inverse Dynamics

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Given a desired behavior, construct a system that fits

Given for instance:

- Stable states
- Attractors
- Time courses
- ...