

# Introduction to the FitzHugh-Nagumo model

**Workshop on Dynamics of Excitable Systems**

# FitzHugh-Nagumo Model

The essential features of excitability

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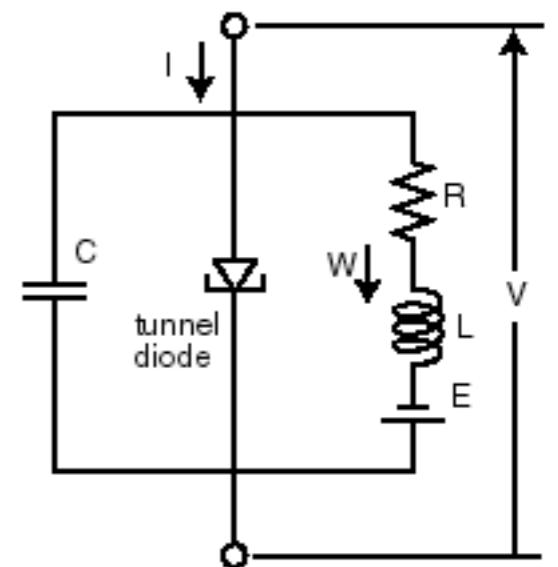
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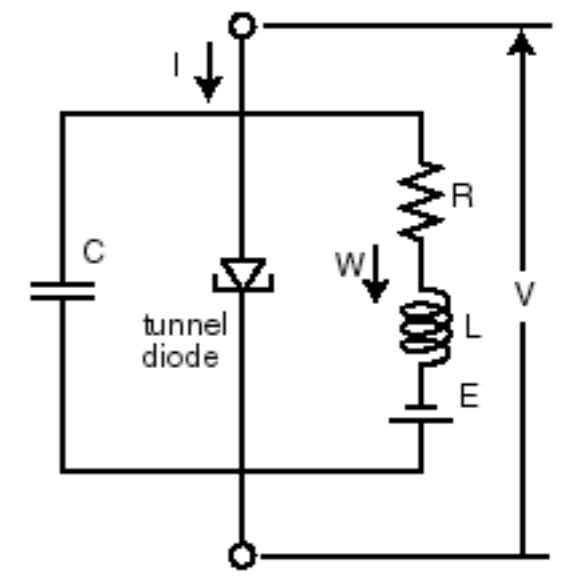


Scholarpedia;  
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- Delineate the essential features of mathematical excitation from physiological details
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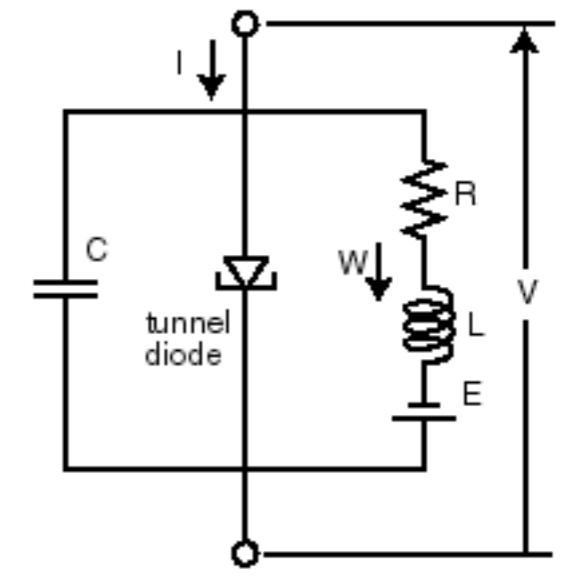


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- $\dot{u} = u - u^3/3 - v + I, \dot{v} = 0.08(u - 0.8v + 0.7)$



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  - Near these curves, the dynamics are slower for smooth systems, so they organize the dynamics in the phase plane

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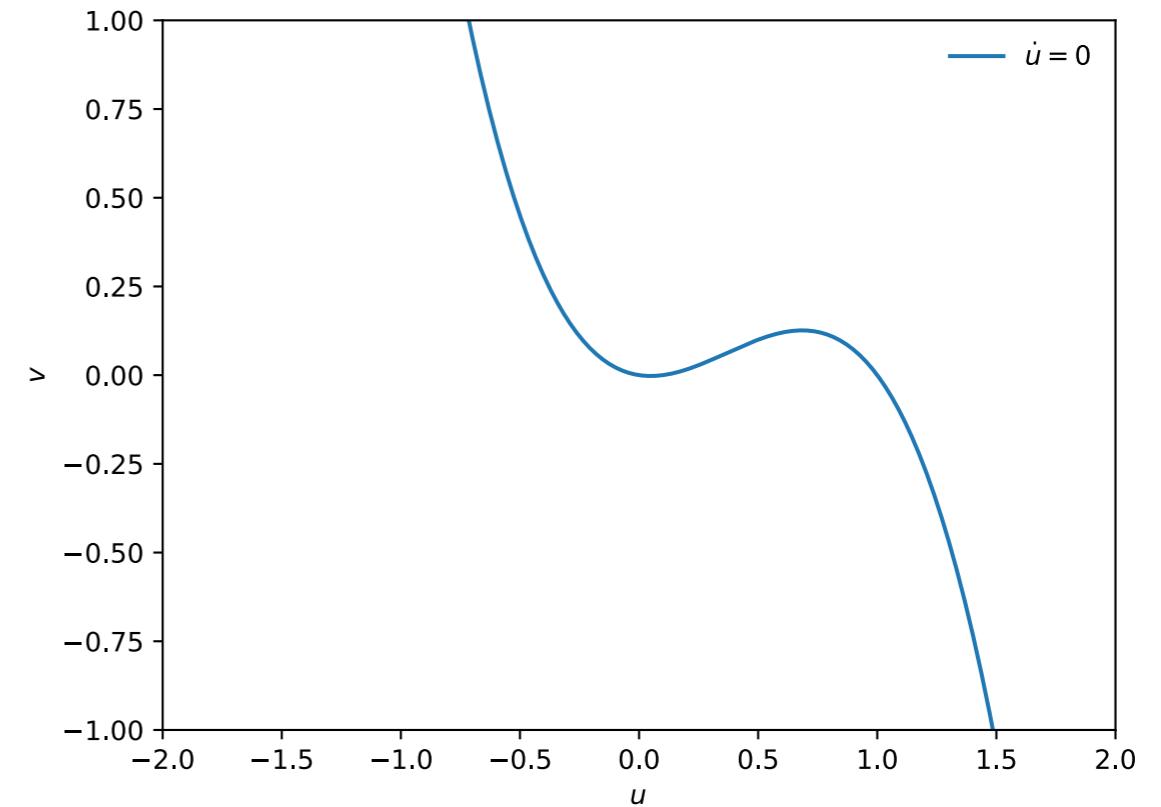
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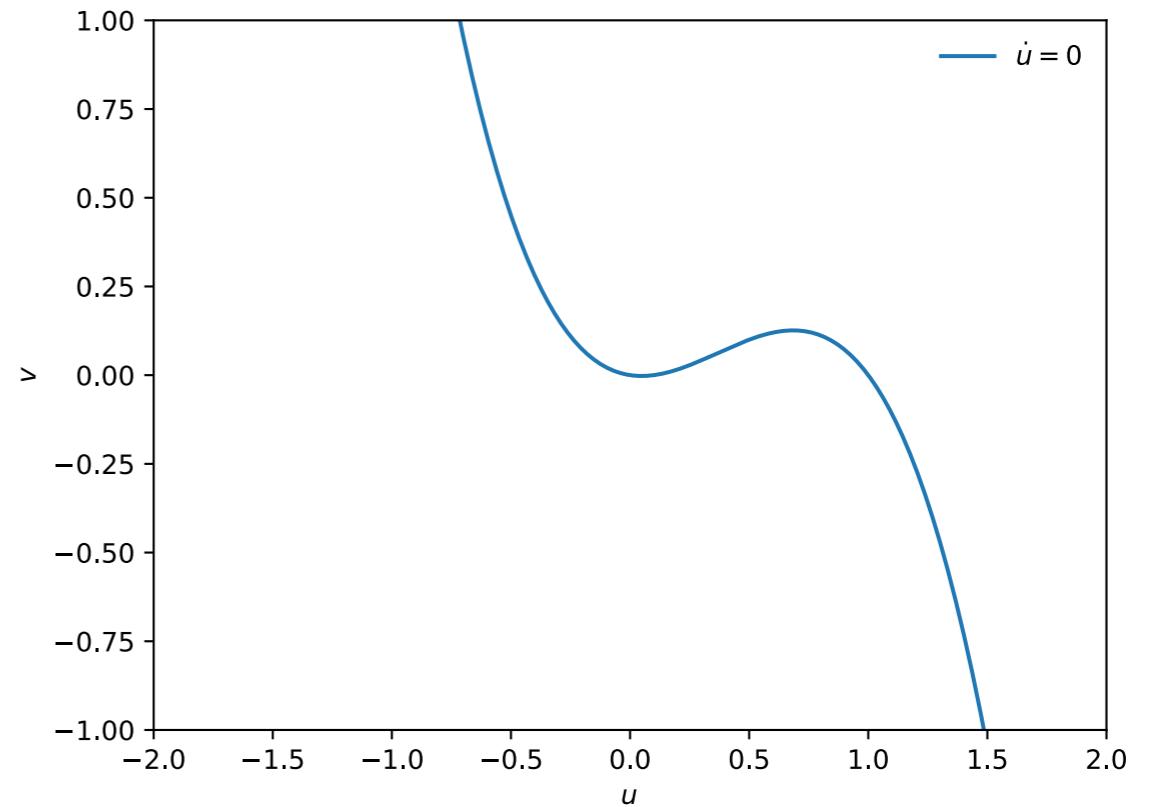
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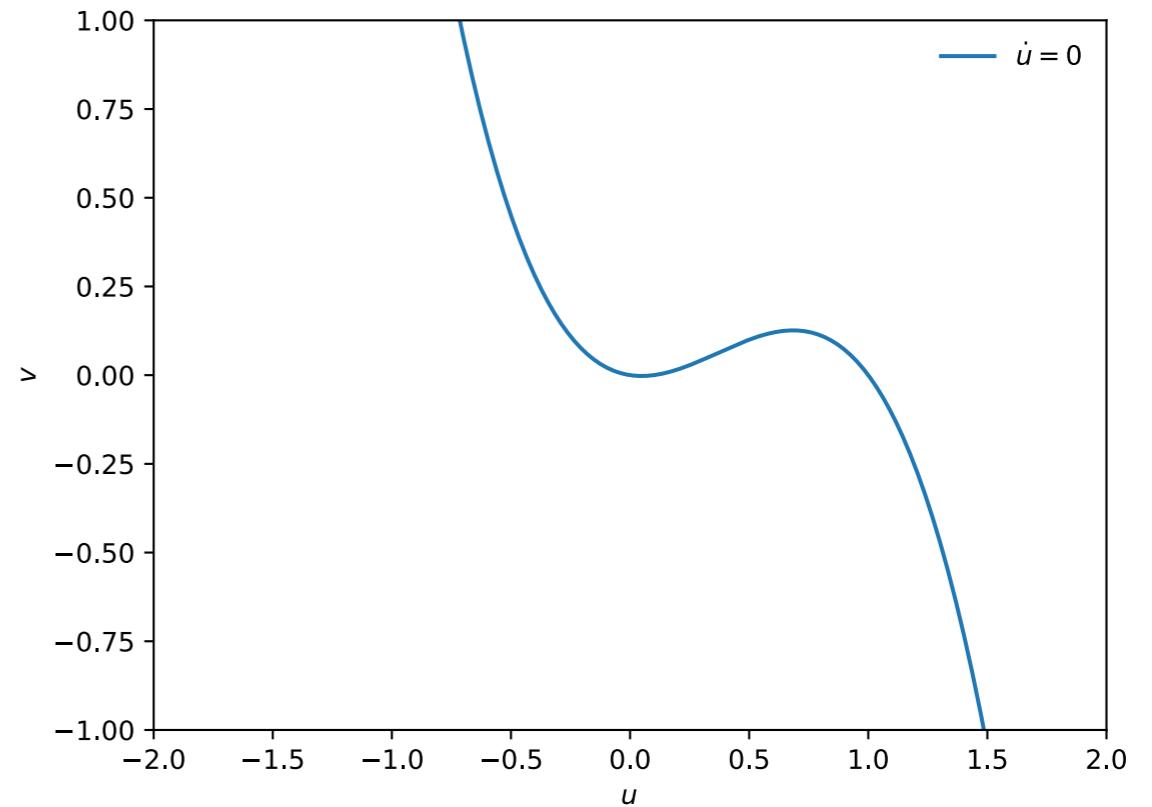
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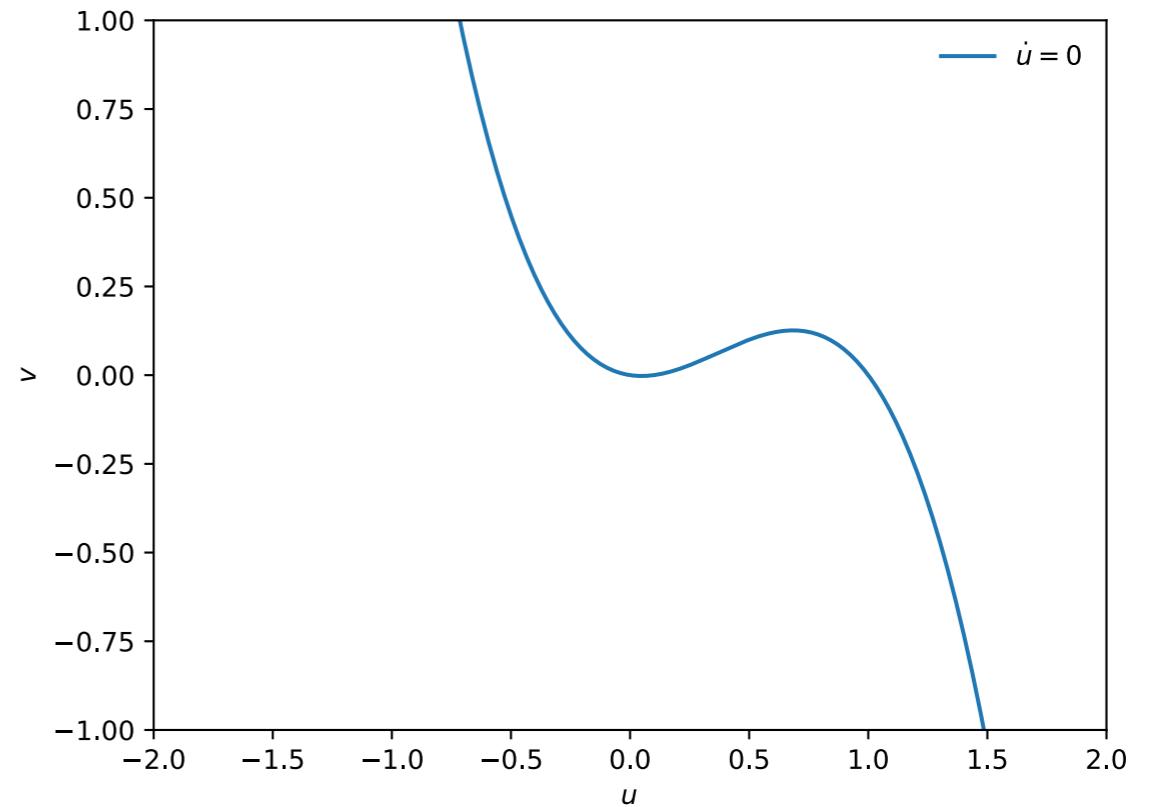
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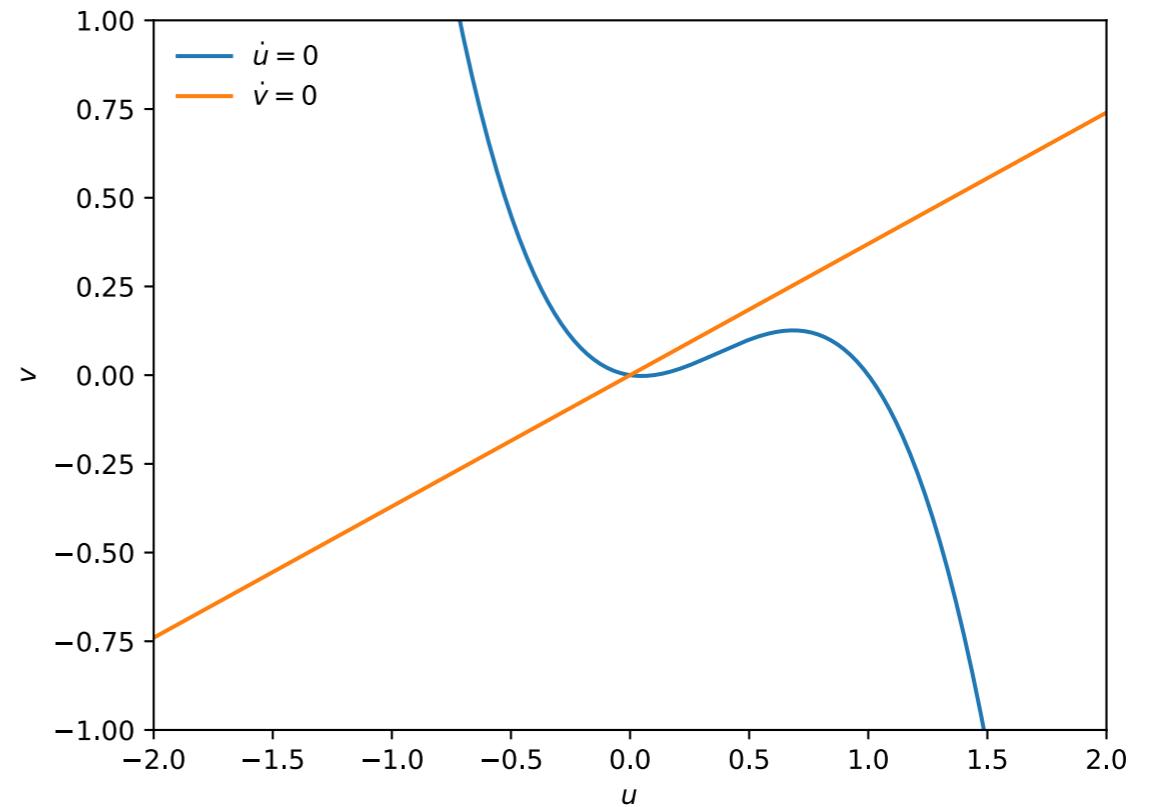
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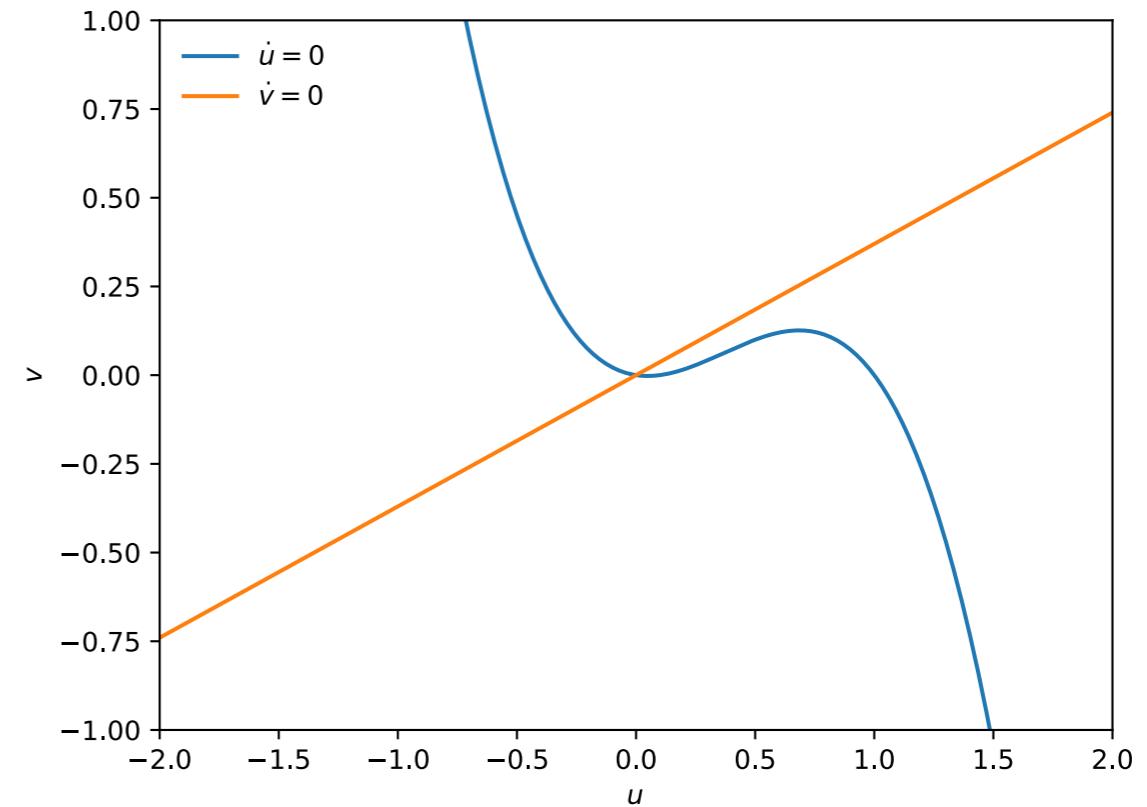
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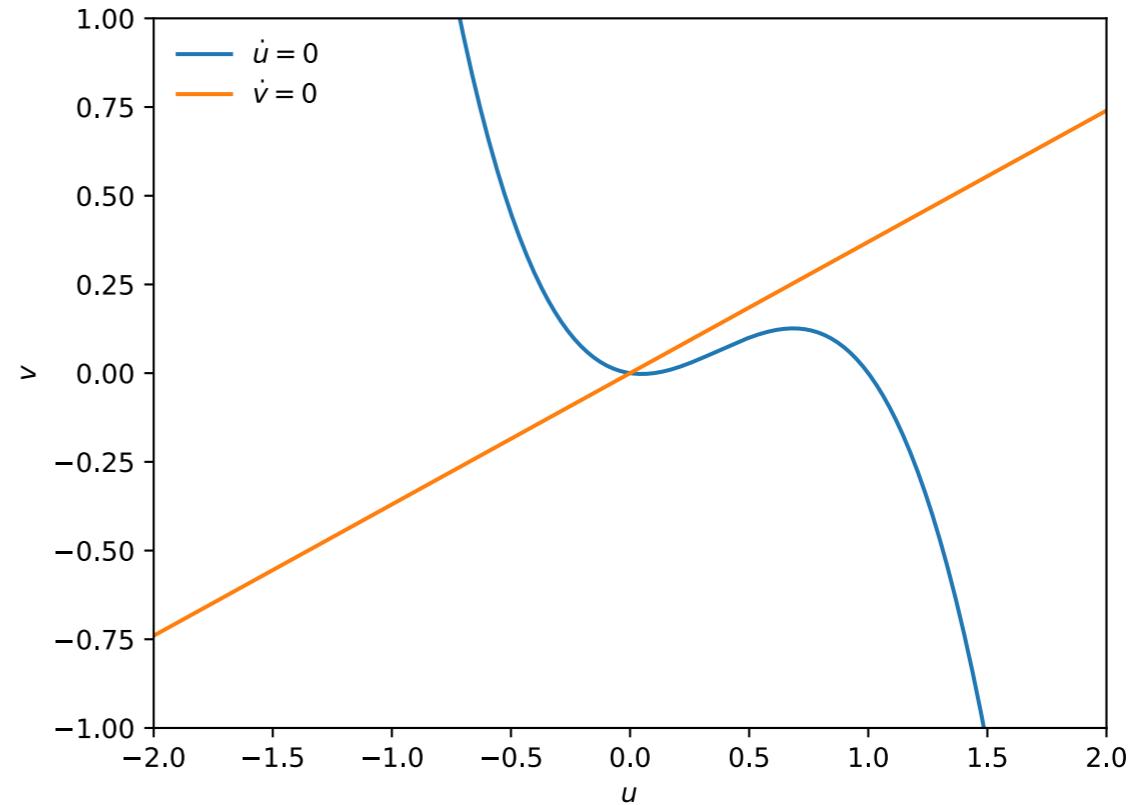
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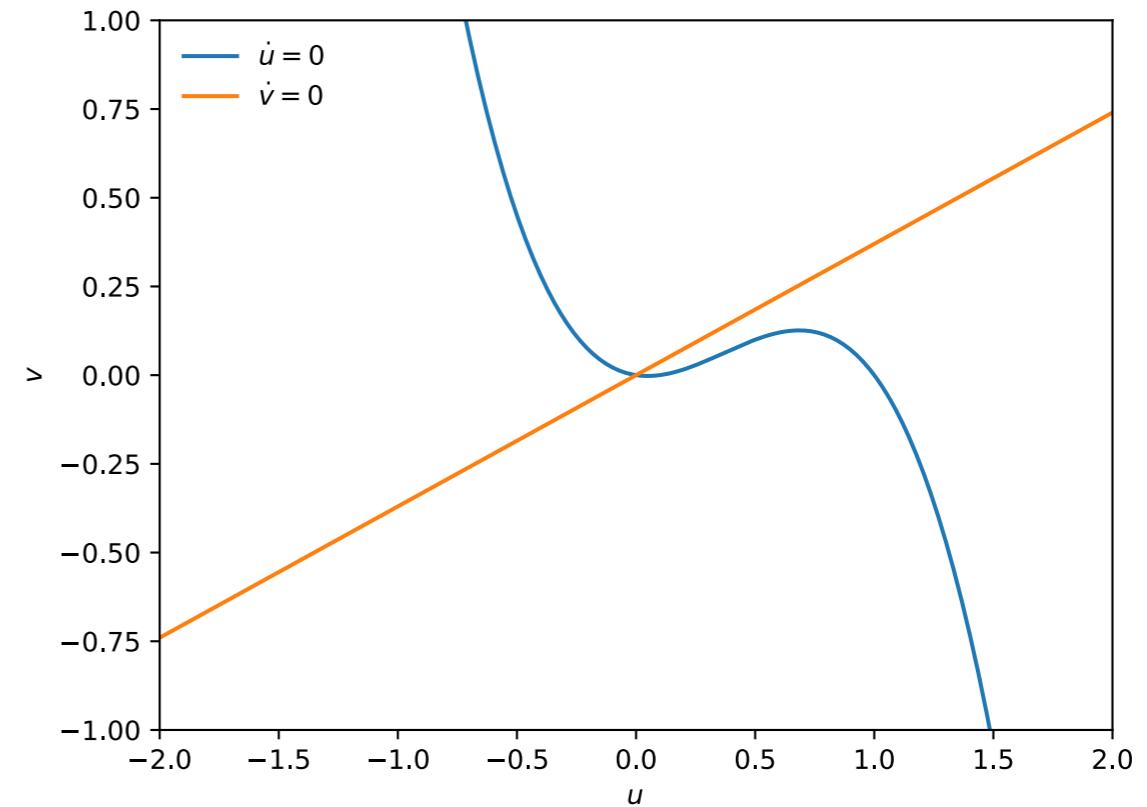
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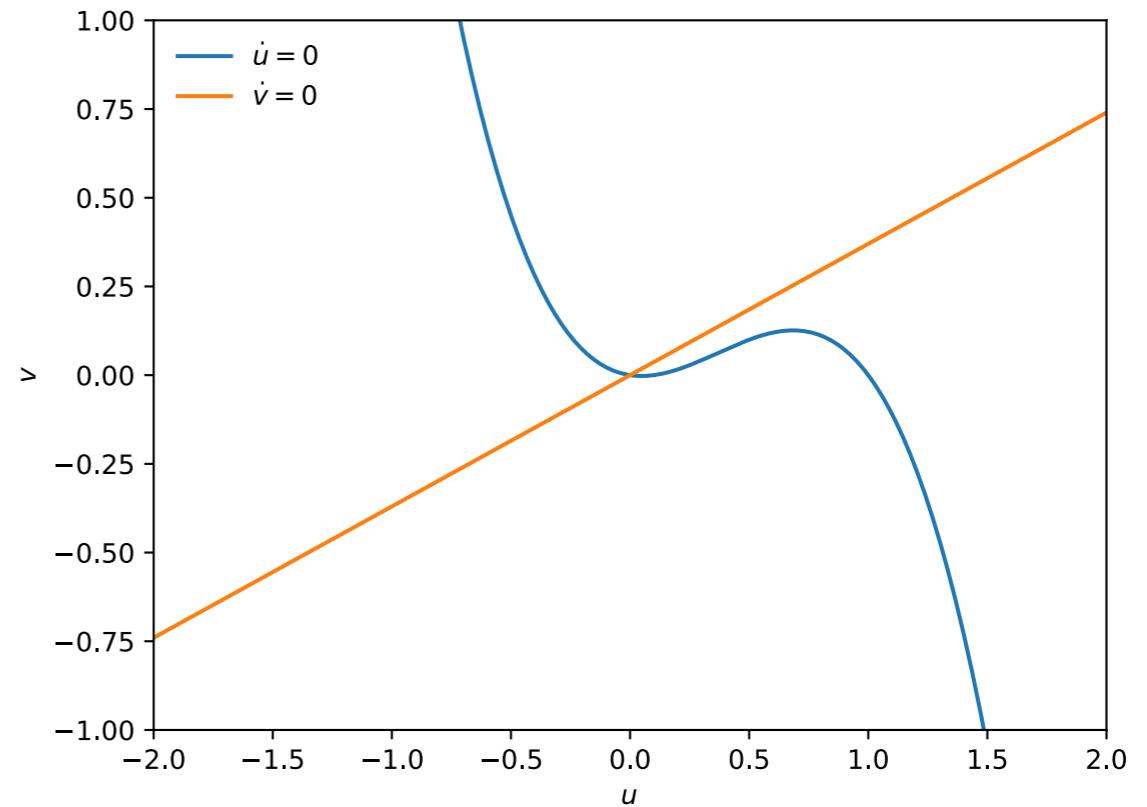
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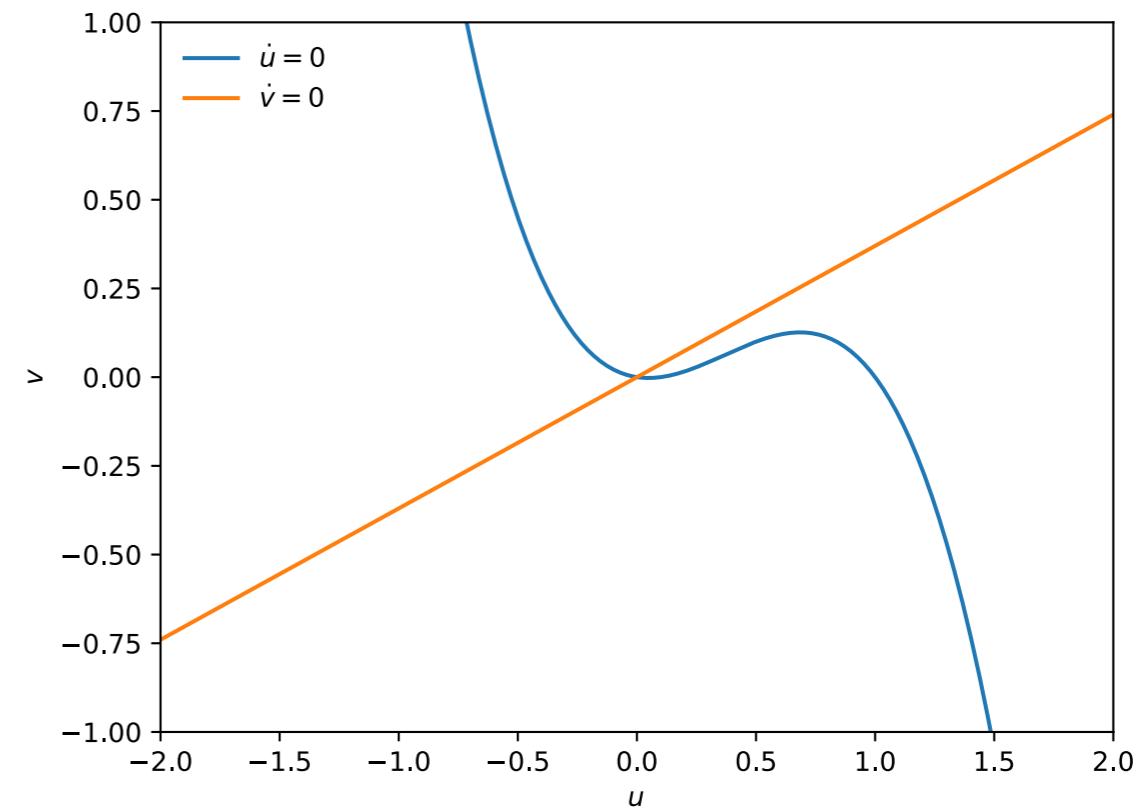
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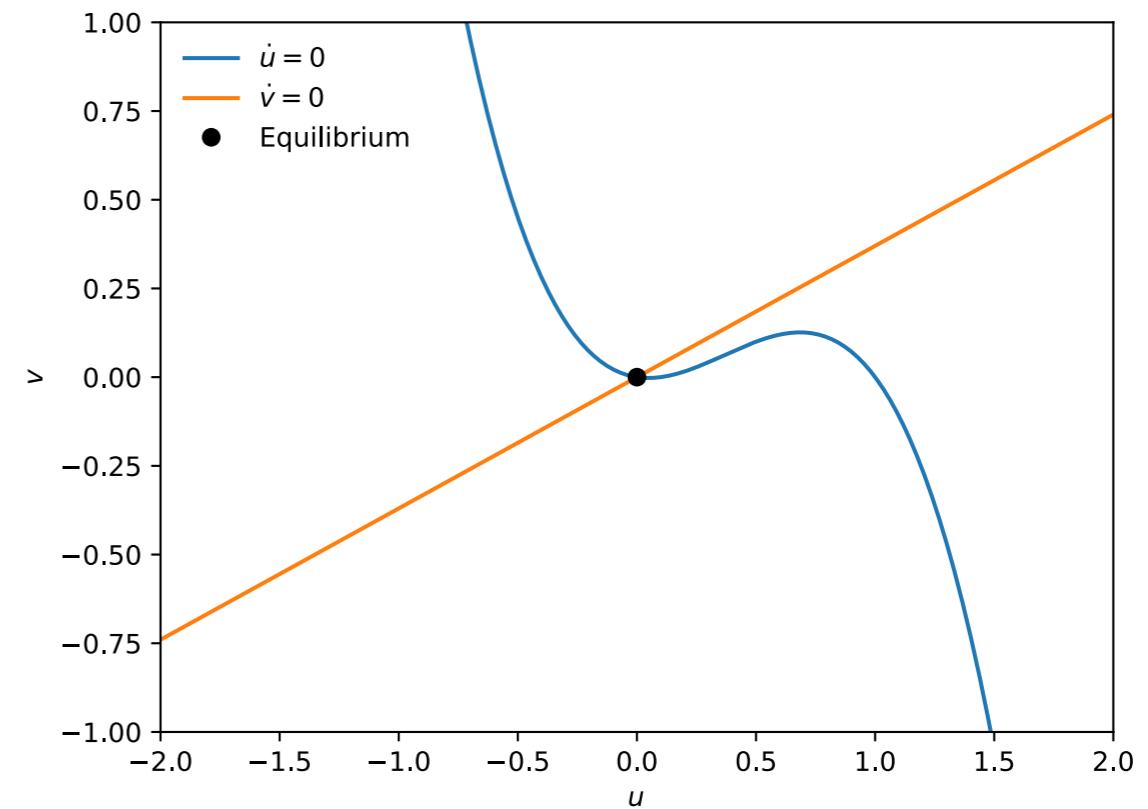
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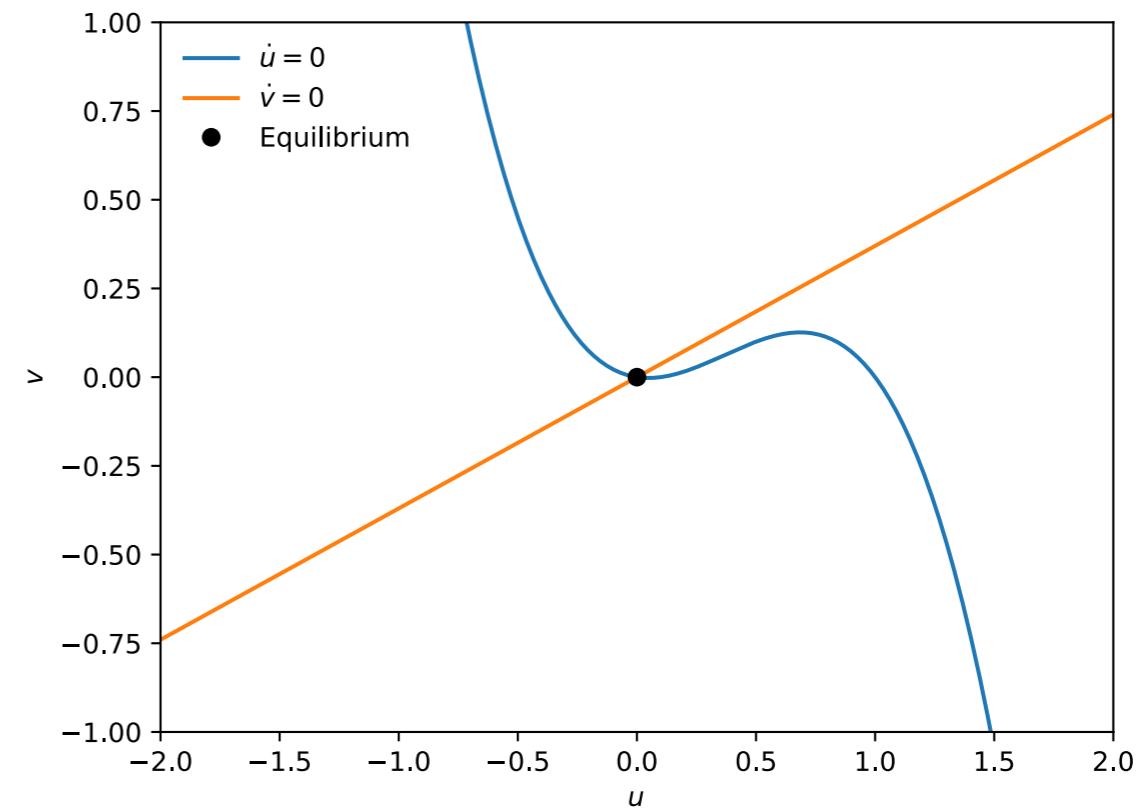


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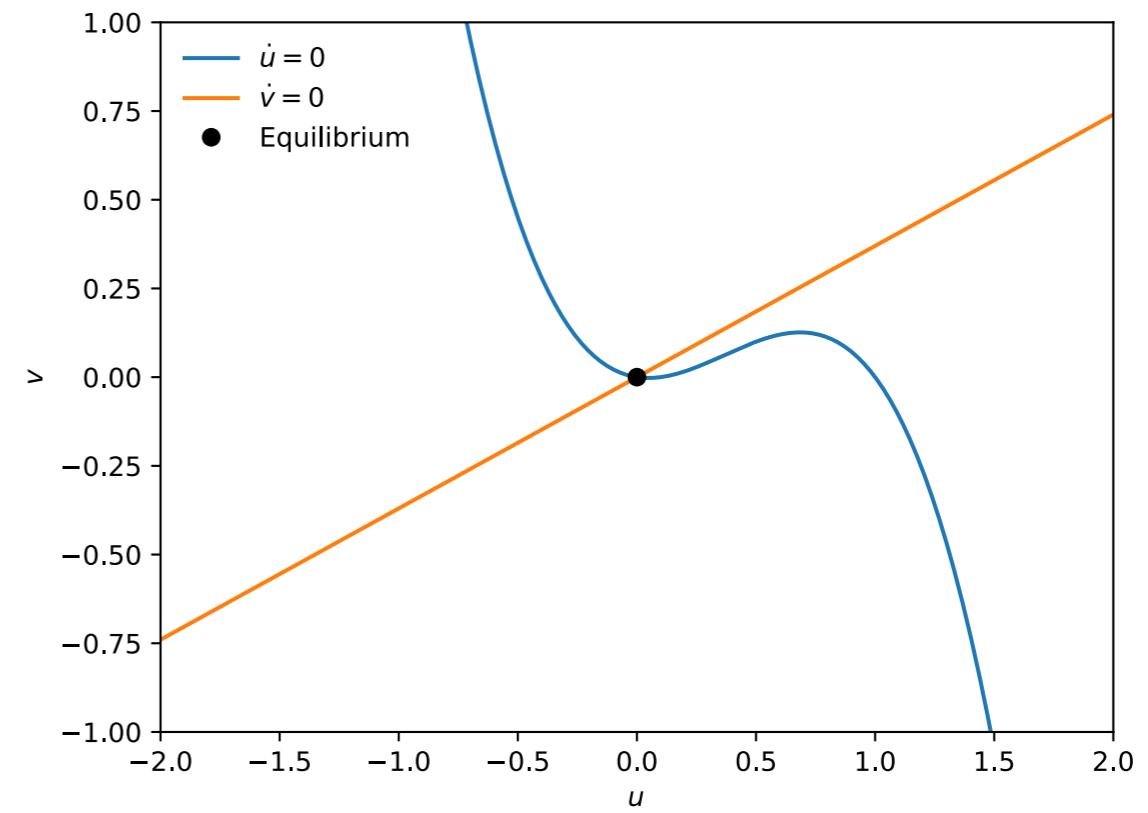


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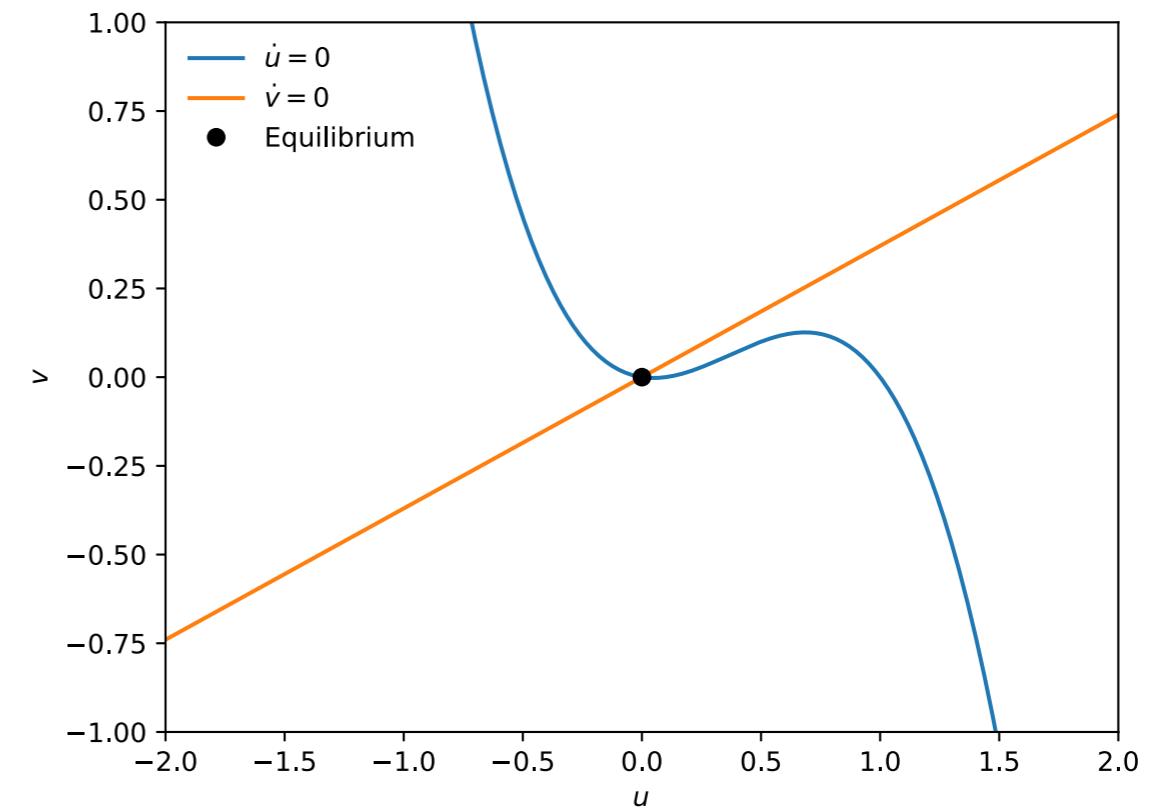
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  - Plug it back into the equations and verify that  $f(0,0) = 0$  and  $\gamma(\alpha 0 - 0) = 0$

# Phase Plane

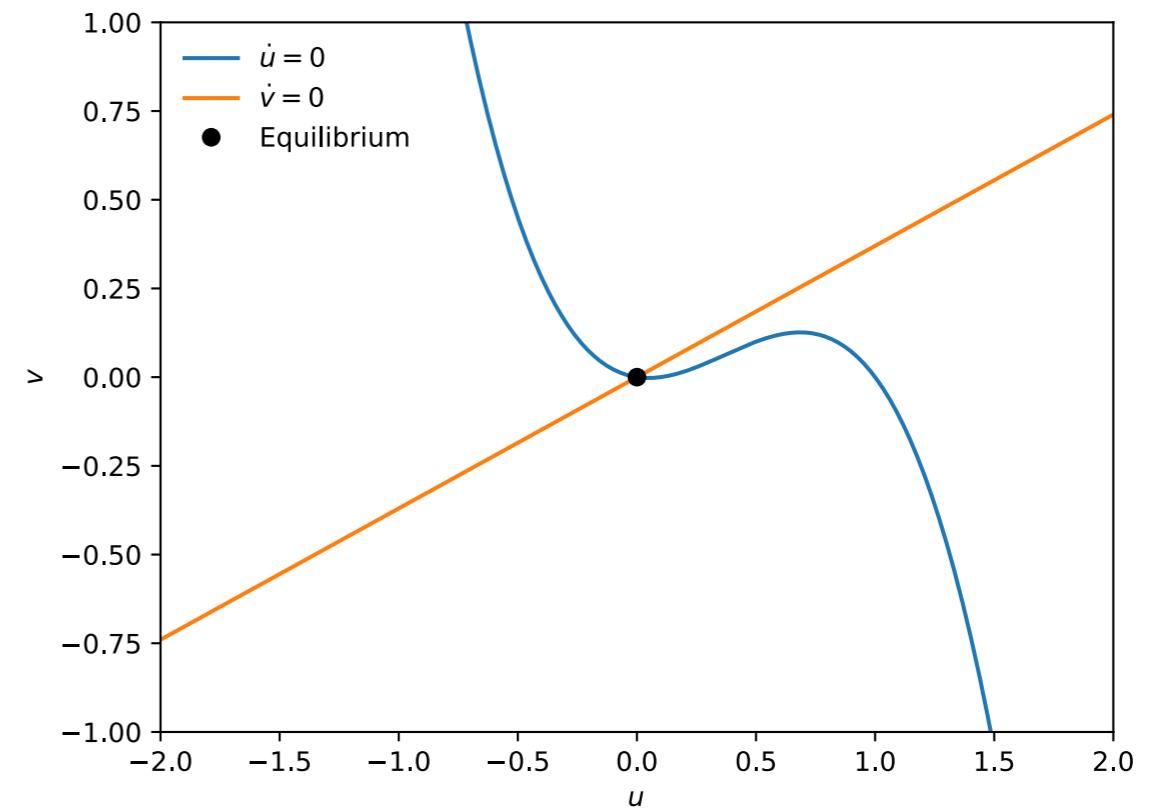
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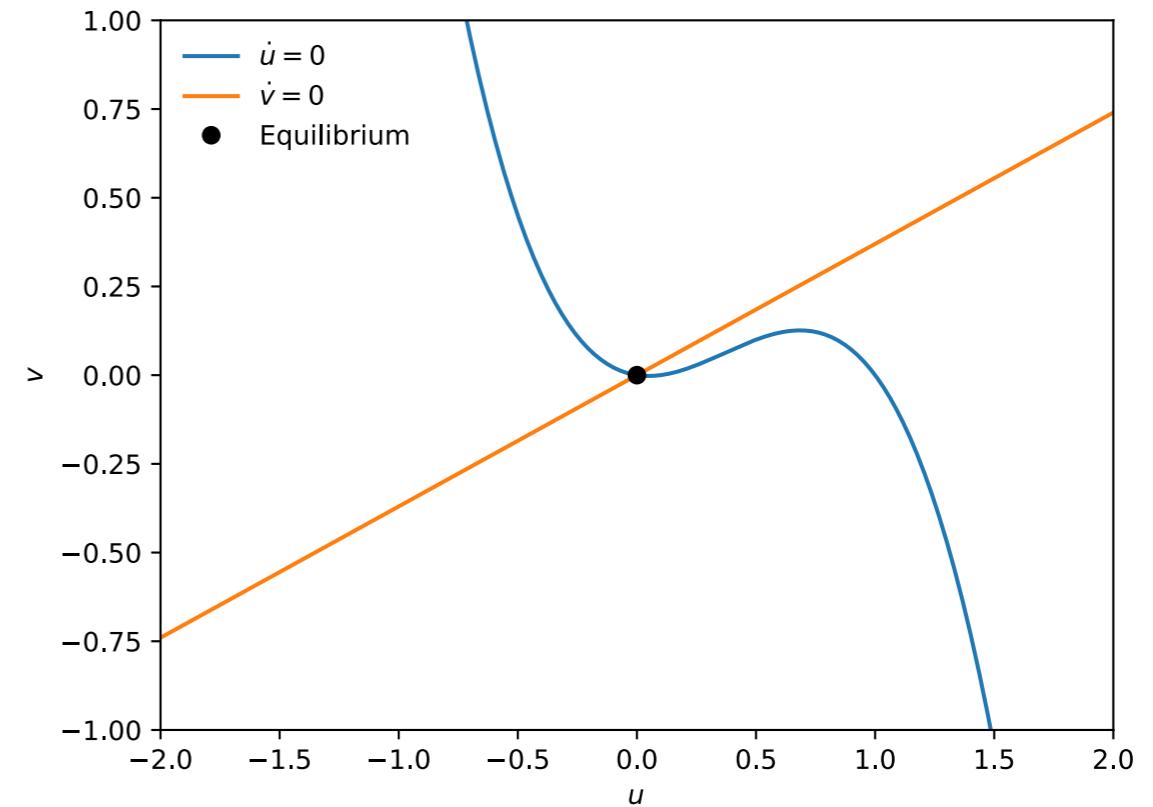
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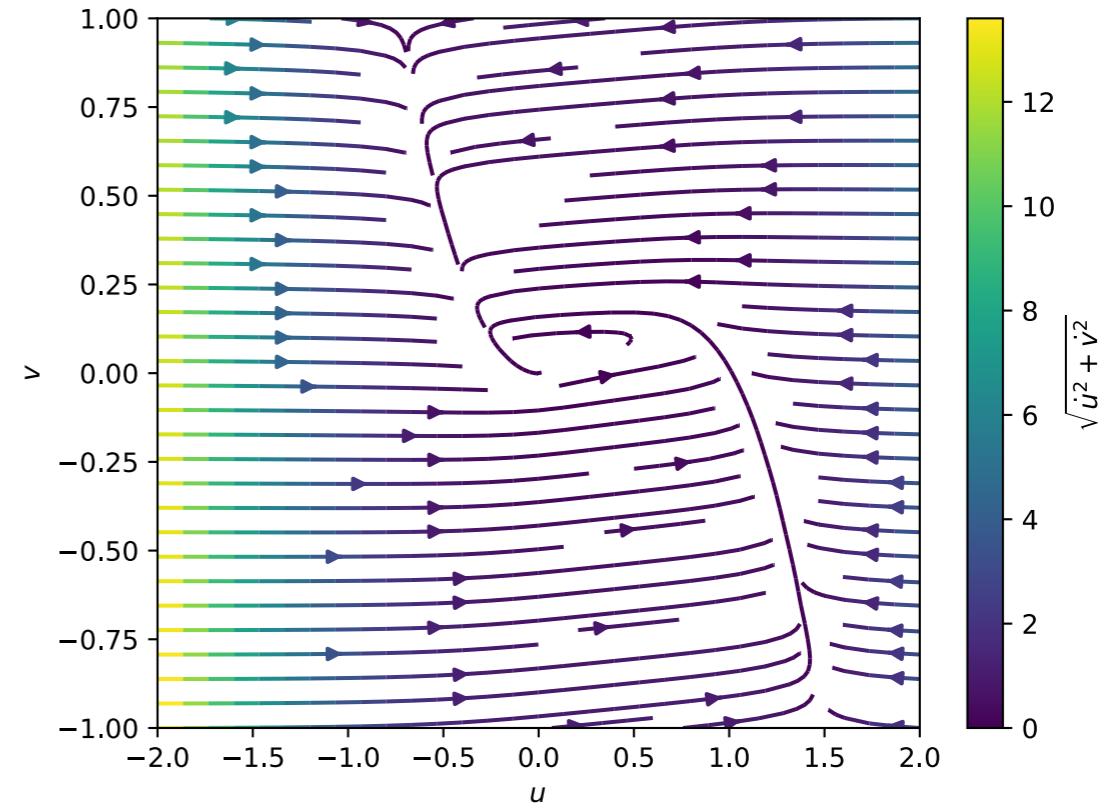
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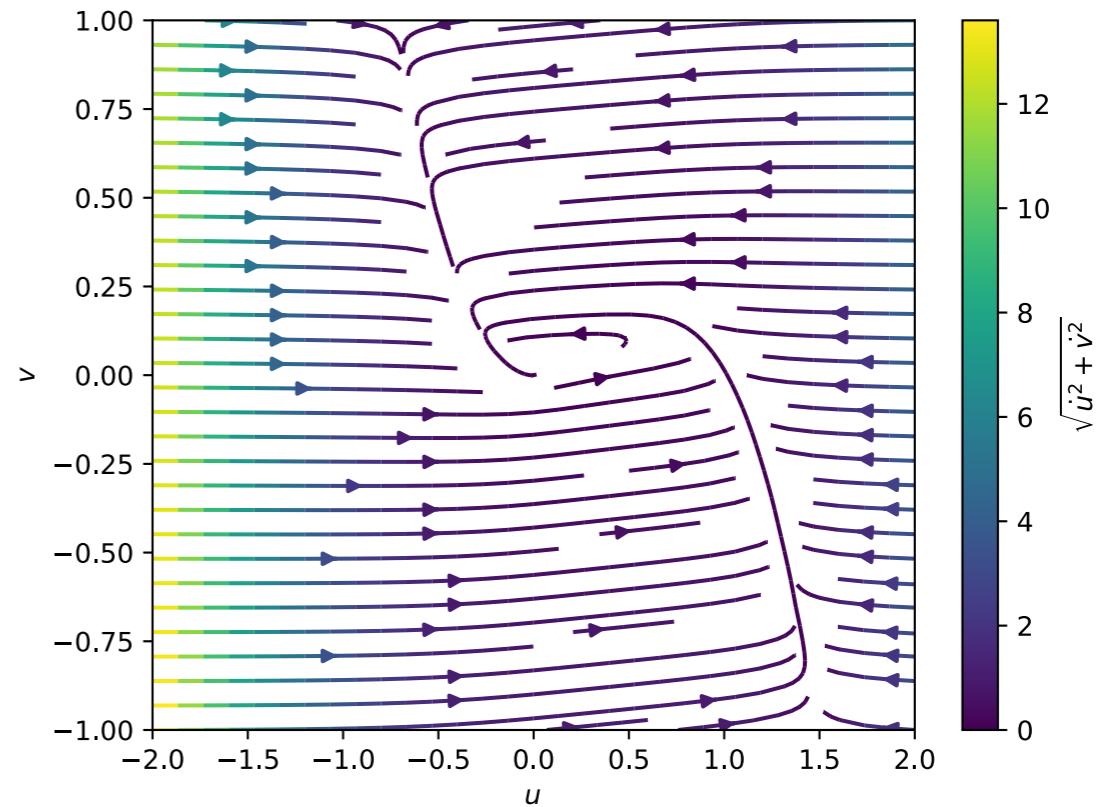
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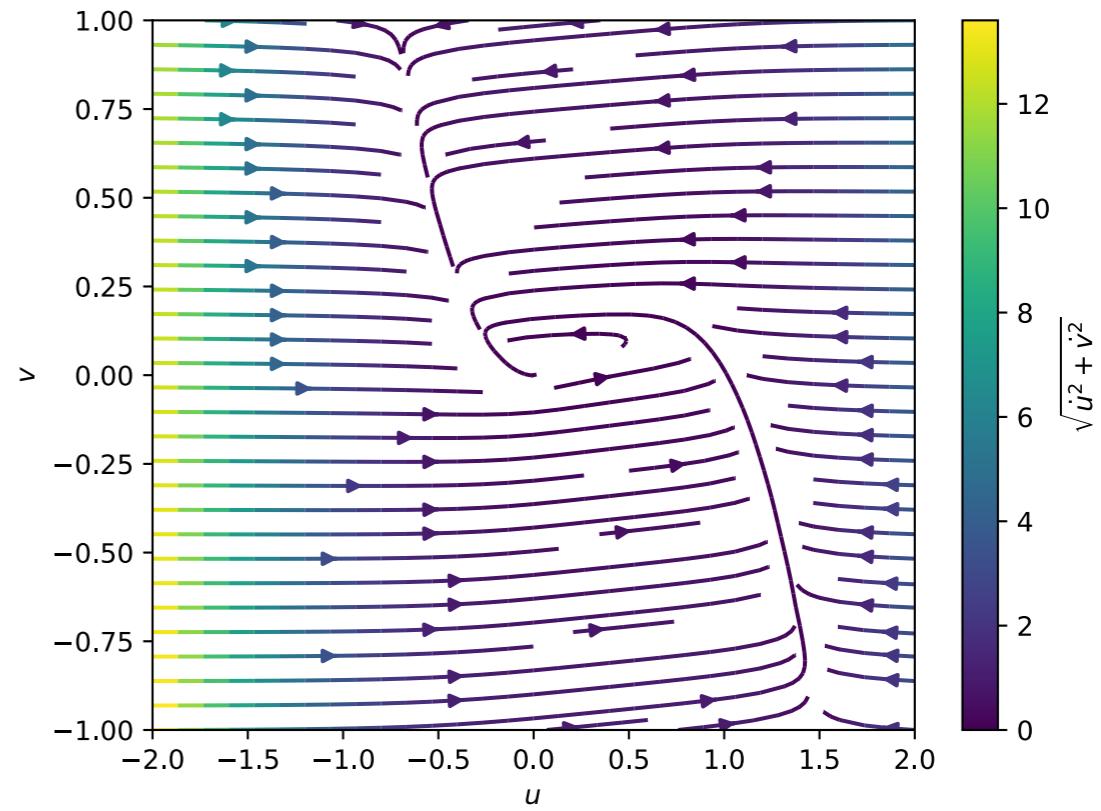
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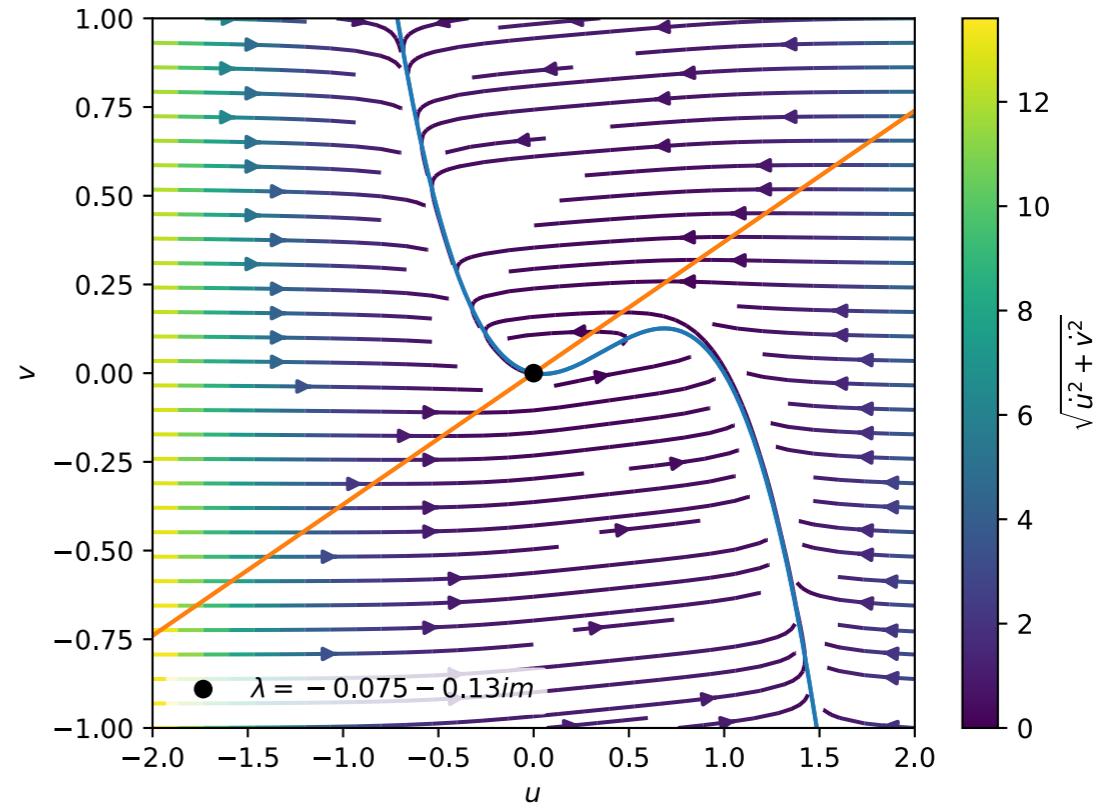
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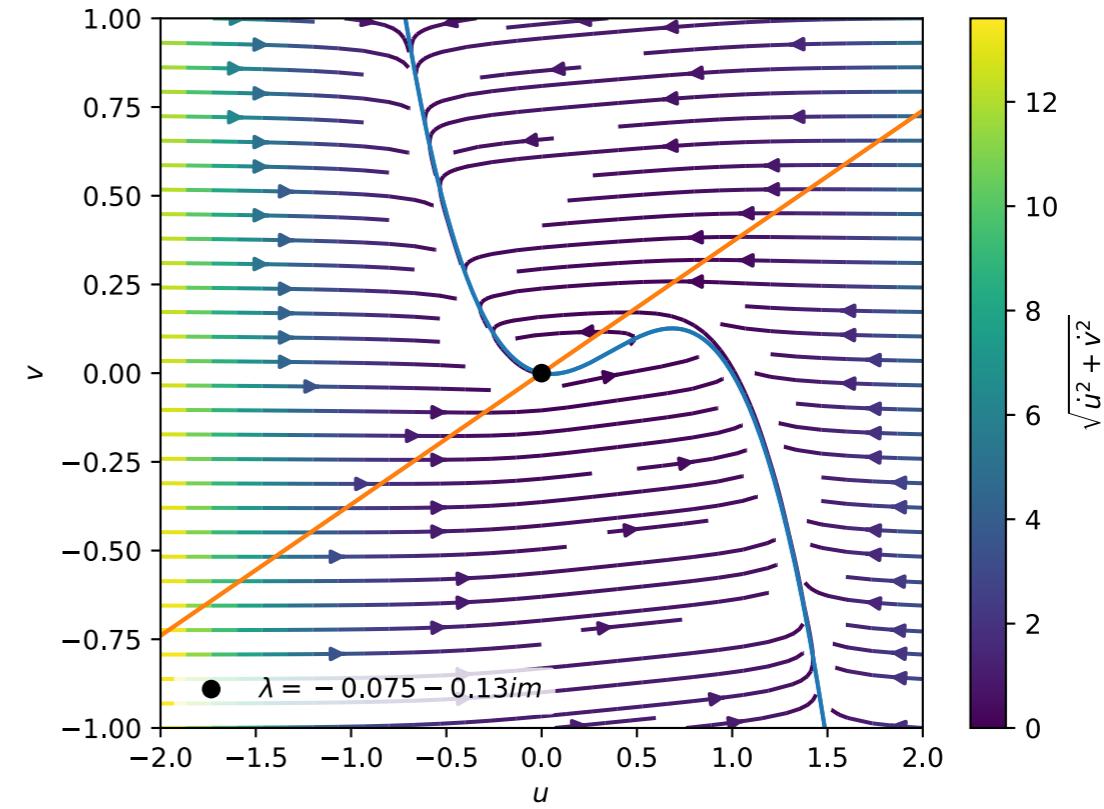
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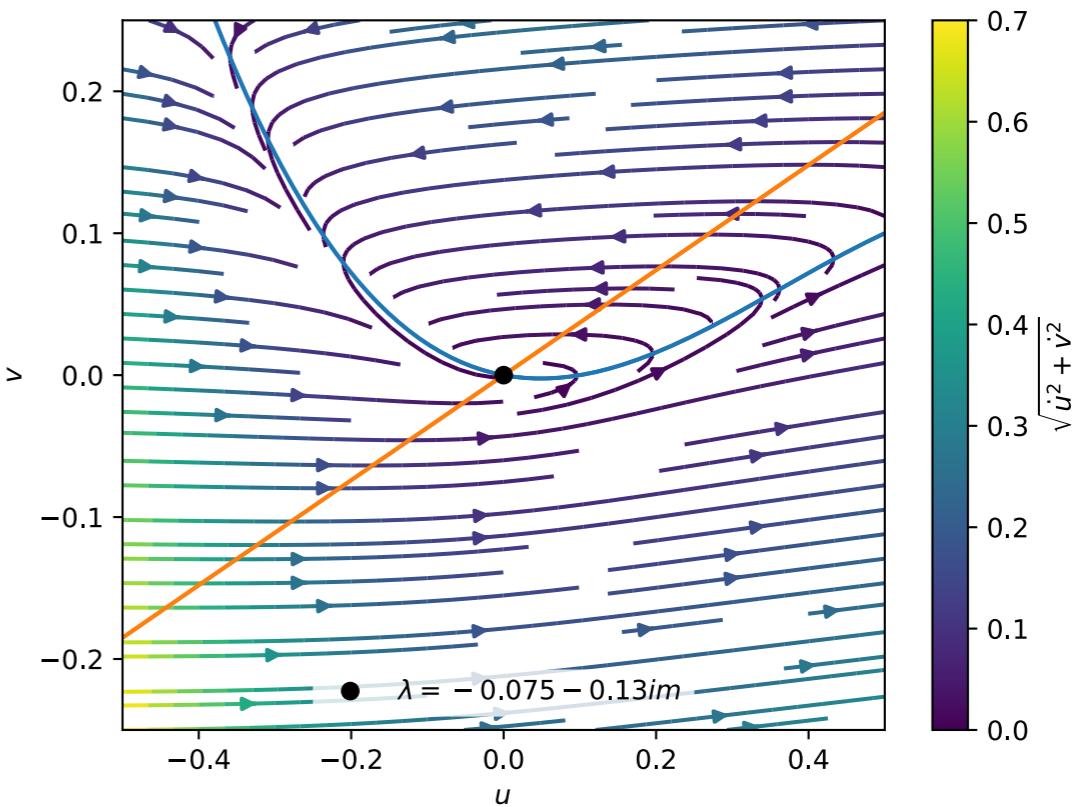
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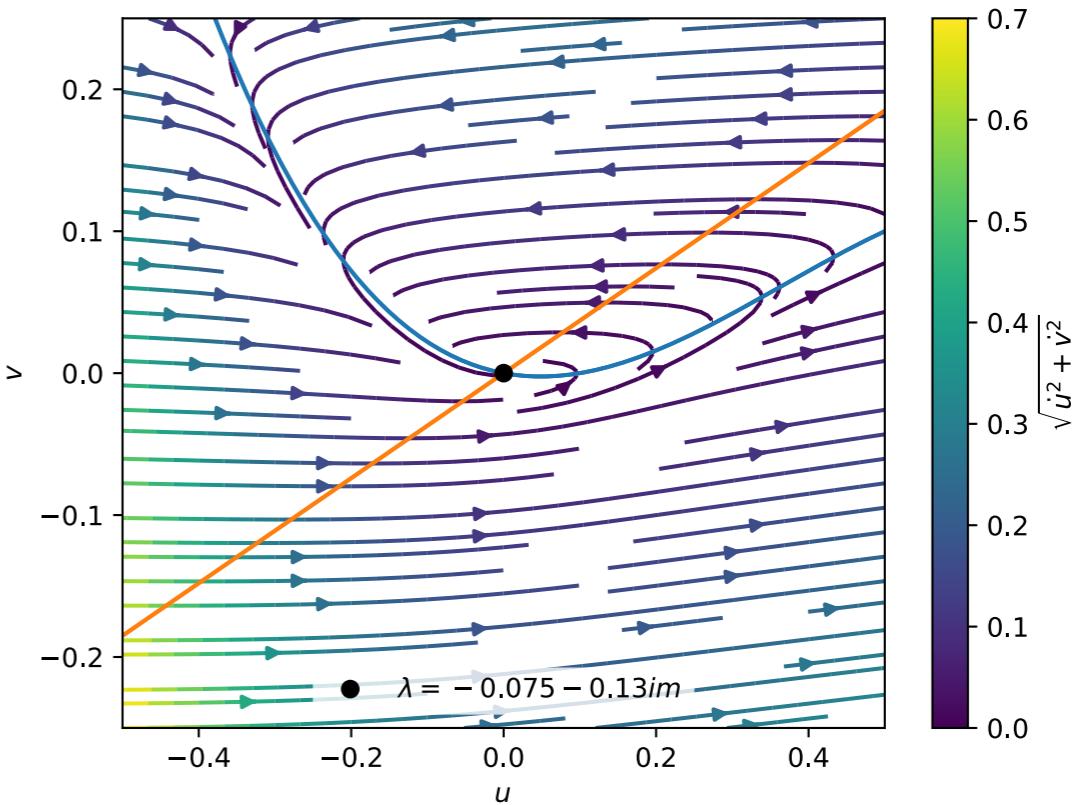


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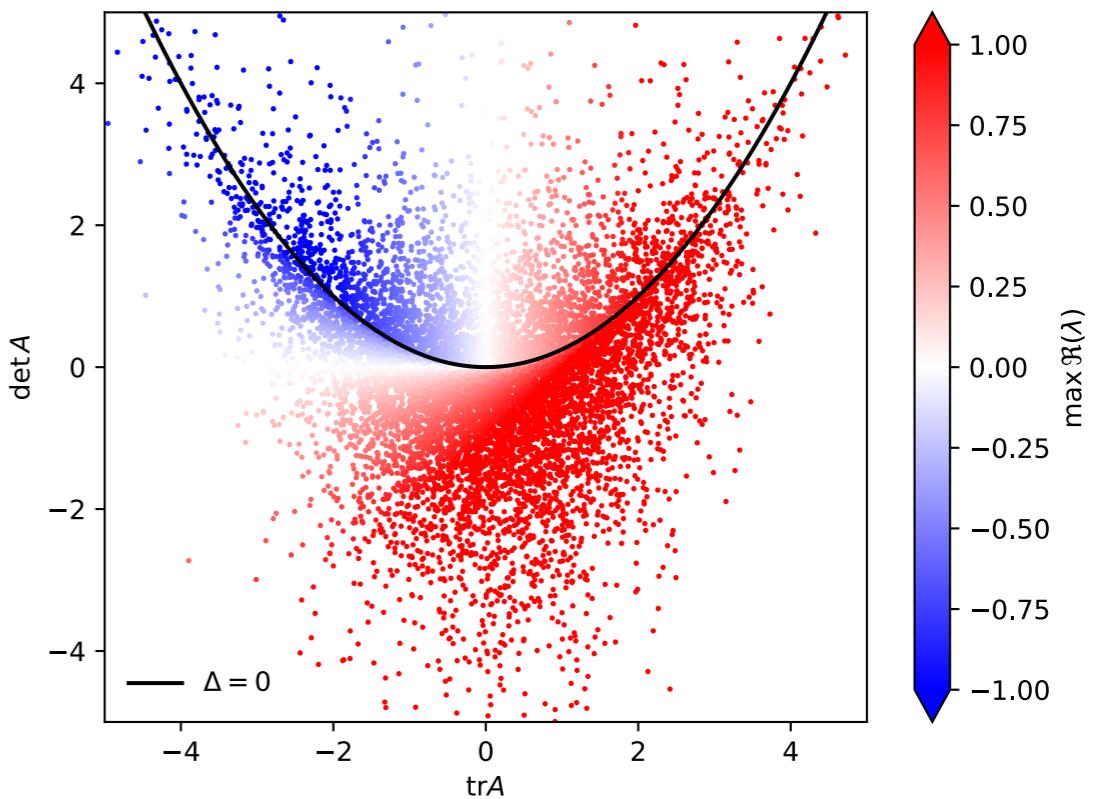
## Just a quick review

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$$\det(A - \lambda I) = 0$$

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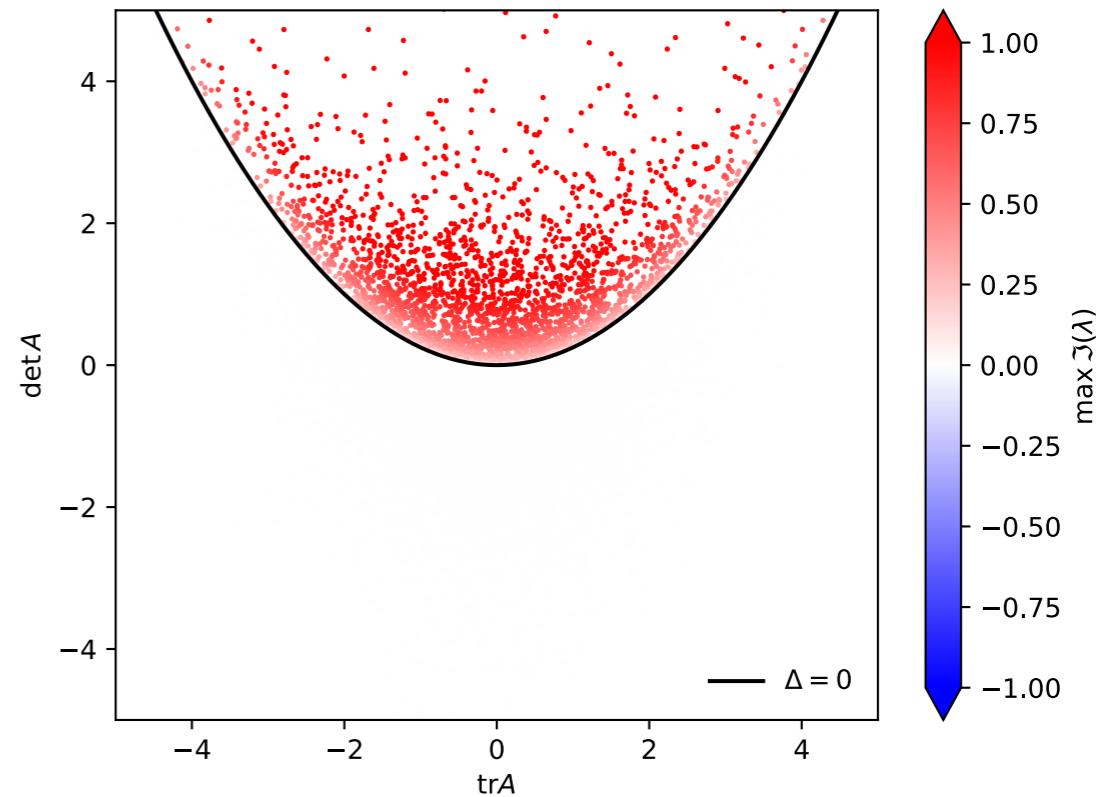
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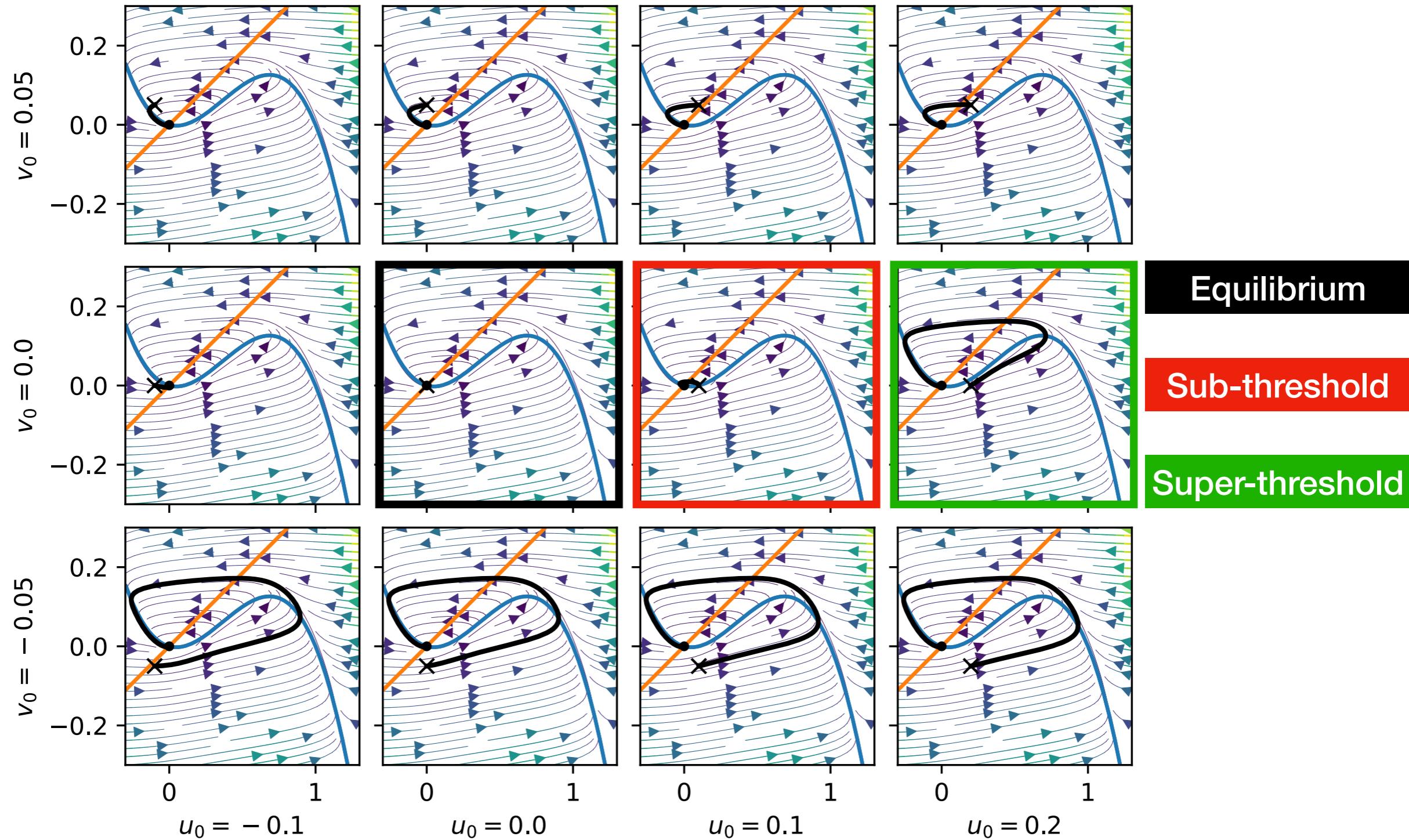
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- $\Delta(A) = \text{tr}(A)^2 - 4 \det(A)$ 
  - Oscillations:  $\Delta(A) < 0$
- For general systems,  $A = d\dot{\mathbf{x}}/d\mathbf{x}$  evaluated at  $\mathbf{x} : \dot{\mathbf{x}} = \mathbf{0}$

# Where is the excitability?

What makes FHN an archetypal model



# Bifurcations

Qualitative changes of the dynamics

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## Qualitative changes of the dynamics

- Consider decreasing  $\alpha$

# Bifurcations

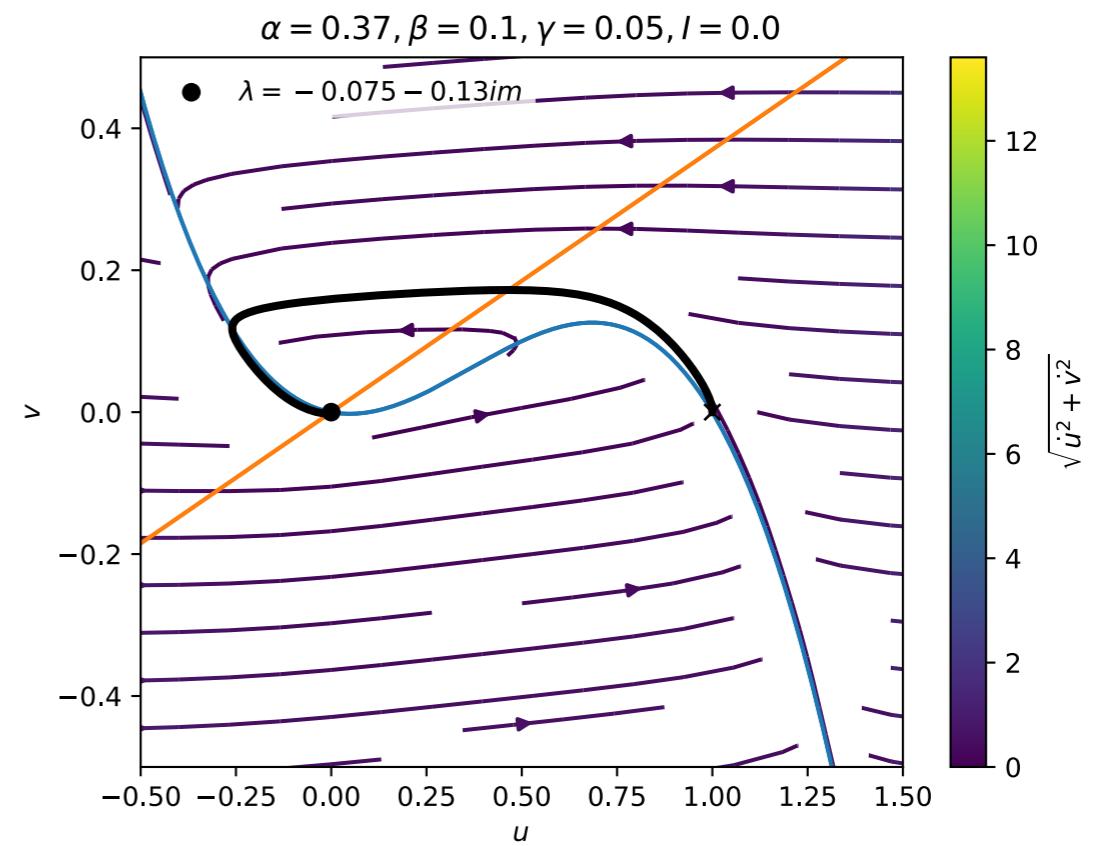
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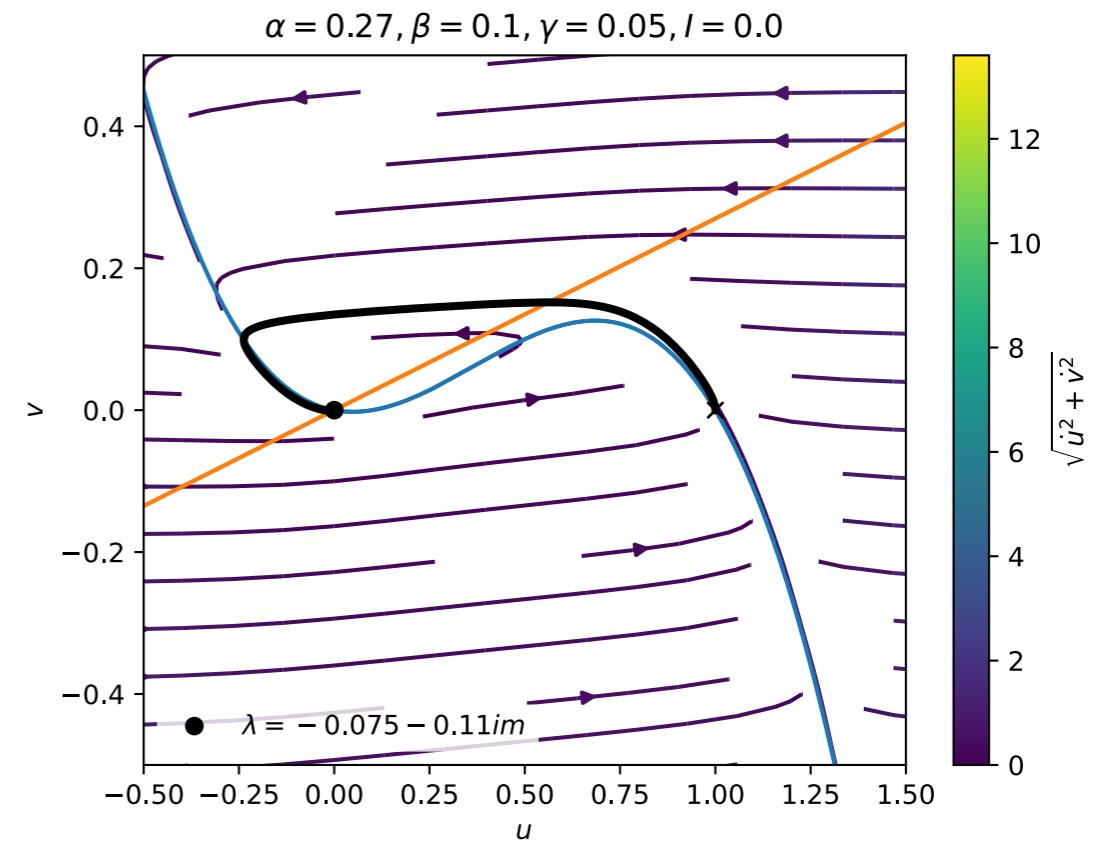
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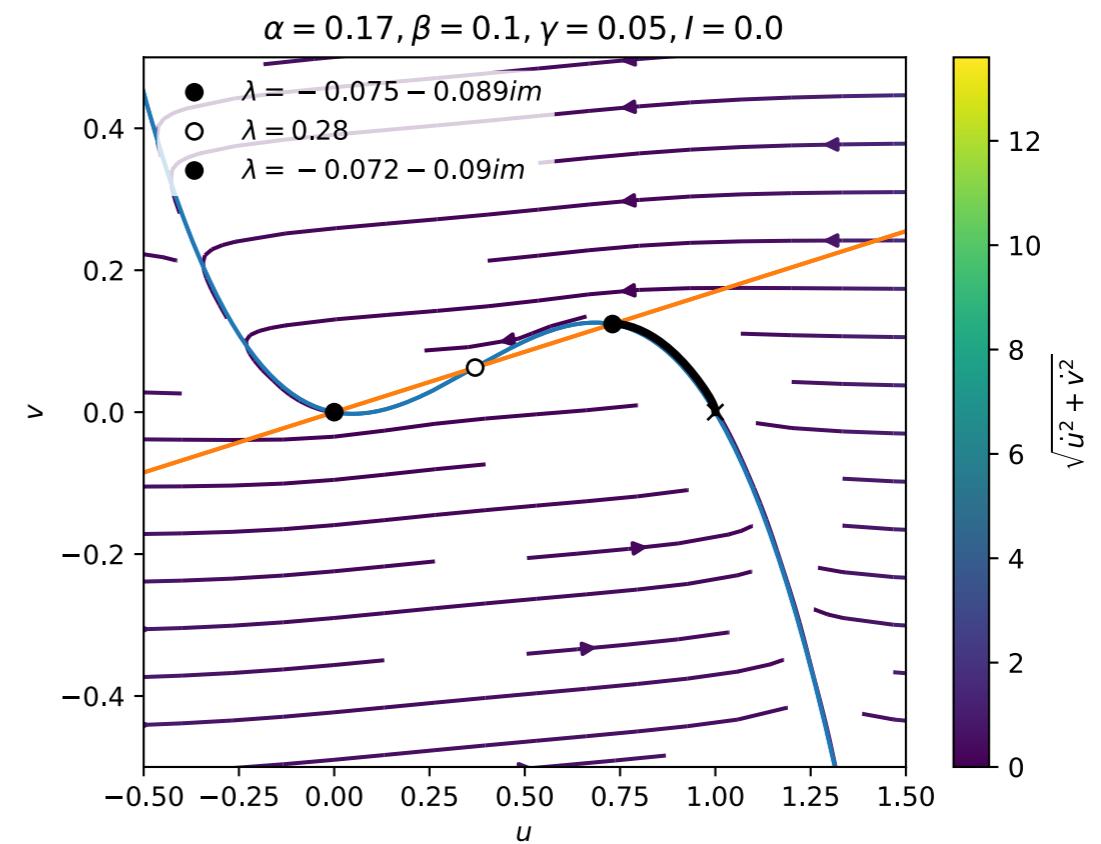
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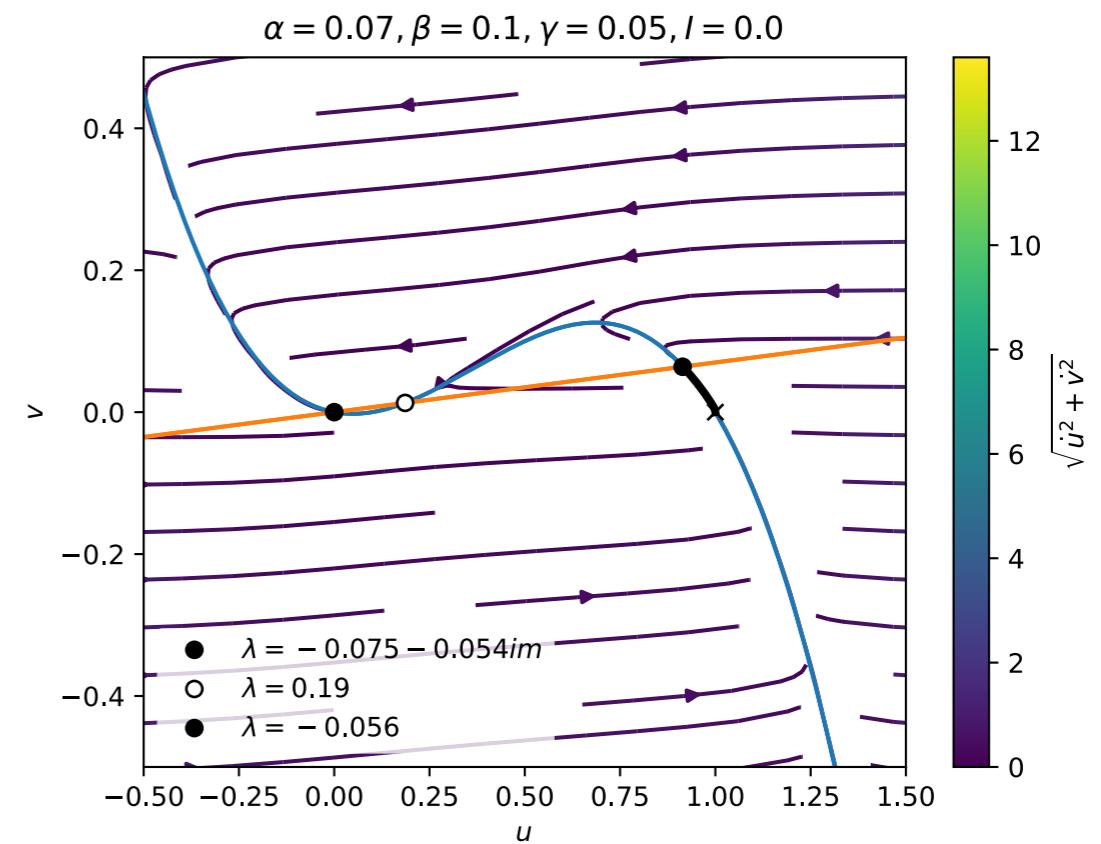
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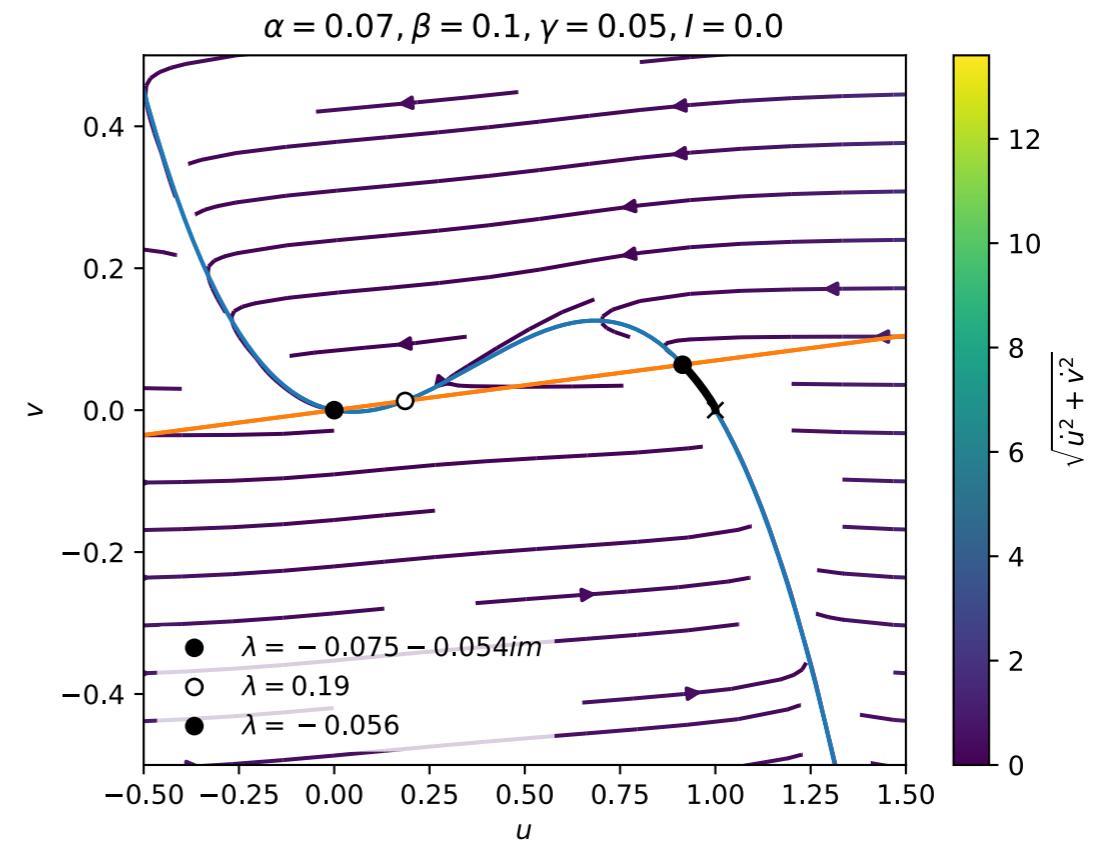
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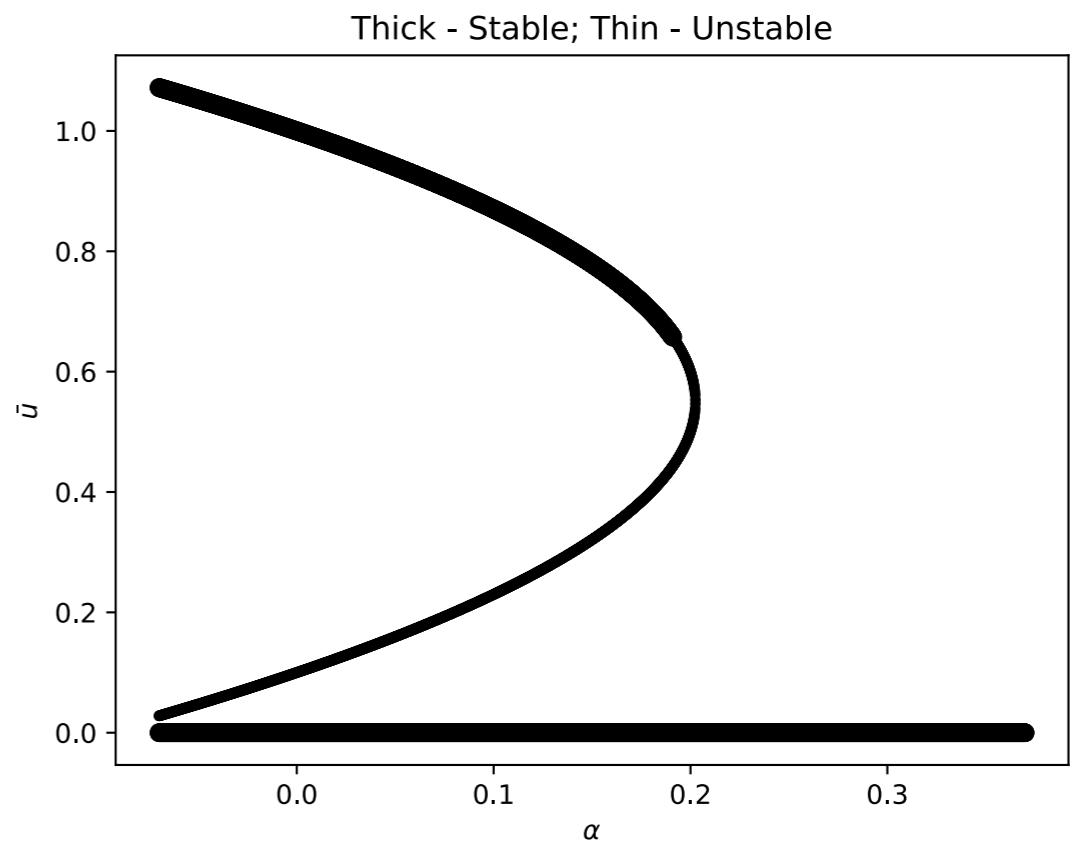
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  - Controls the slope of orange null cline –  $\alpha$  sufficiently small creates more intersections / equilibria
  - Bifurcation near  $\alpha \approx 0.2$
  - Bifurcation diagrams are an important part of understanding dynamical systems with parameters



- “Saddle-node” bifurcation
  - New solutions appear out of collision of a saddle (unstable) and node (stable)
  - Normal form:  $\dot{x} = r + x^2$  in 1D, harder in 2D
  - Verify using Taylor expansion around bif. pt.

# **Changing Parameter values**

## **Bifurcations and qualitative changes to the dynamics**

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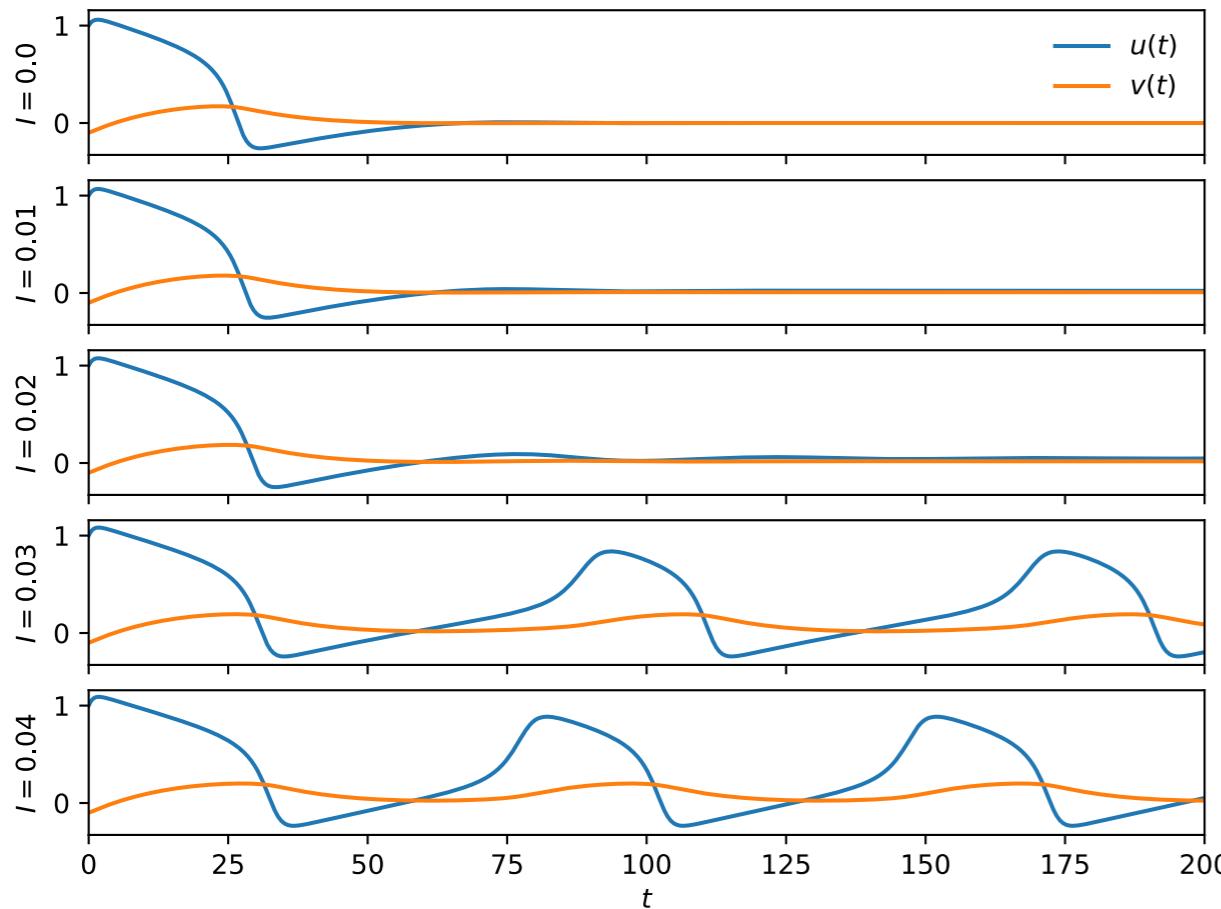
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- But for sufficiently large  $I > 0$ , something interesting happens

# Excitable to Oscillatory

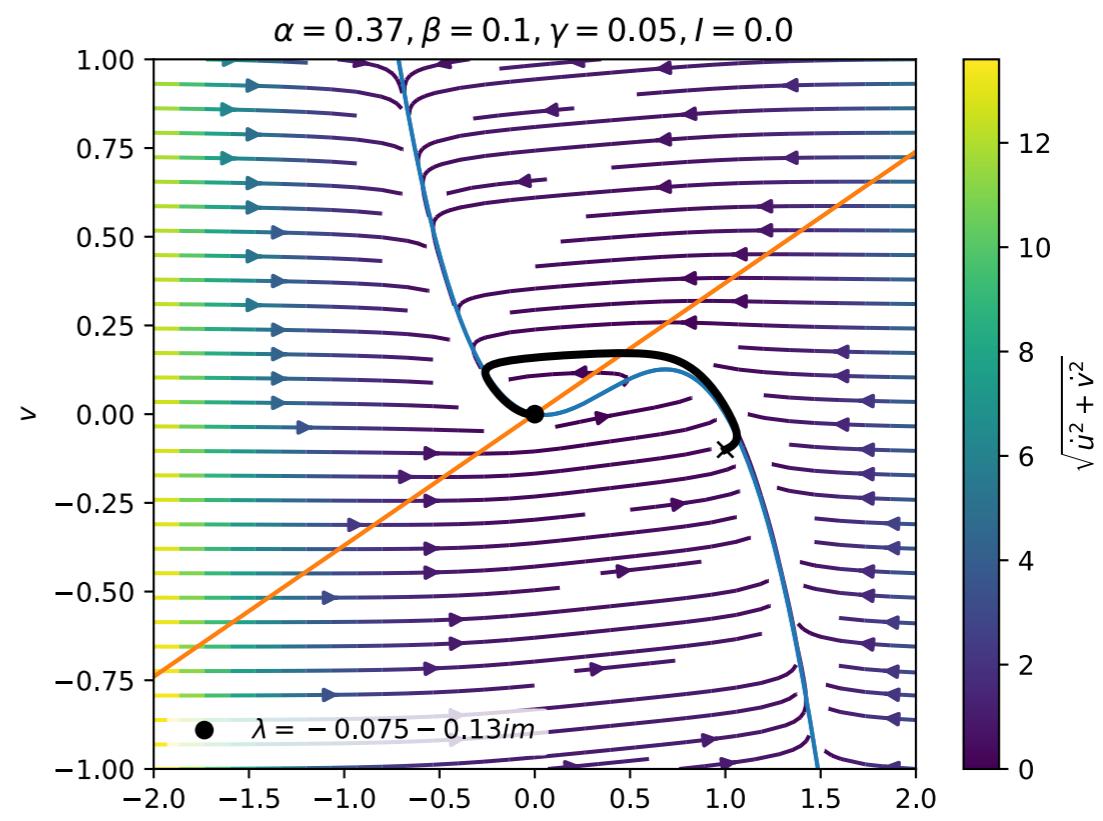
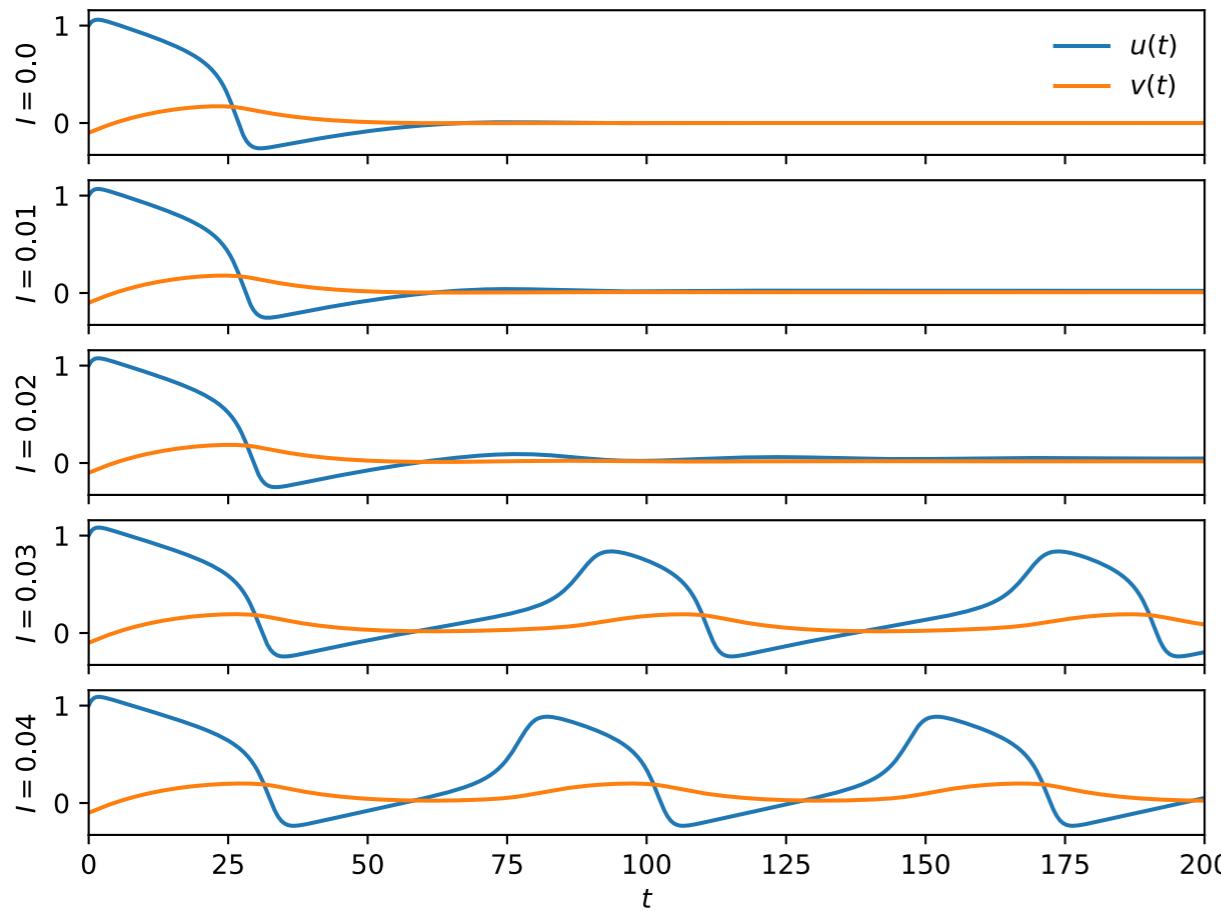
## Increasing stimulation current



- For sufficiently high input current  $I \gtrsim 0.03$ , solution spontaneously oscillates
  - Why do the dynamics change?
- The equilibrium stability changes at  $I \approx 0.0298$ 
  - This is called a Hopf bifurcation, and occurs with the creation of a periodic orbit or limit cycle

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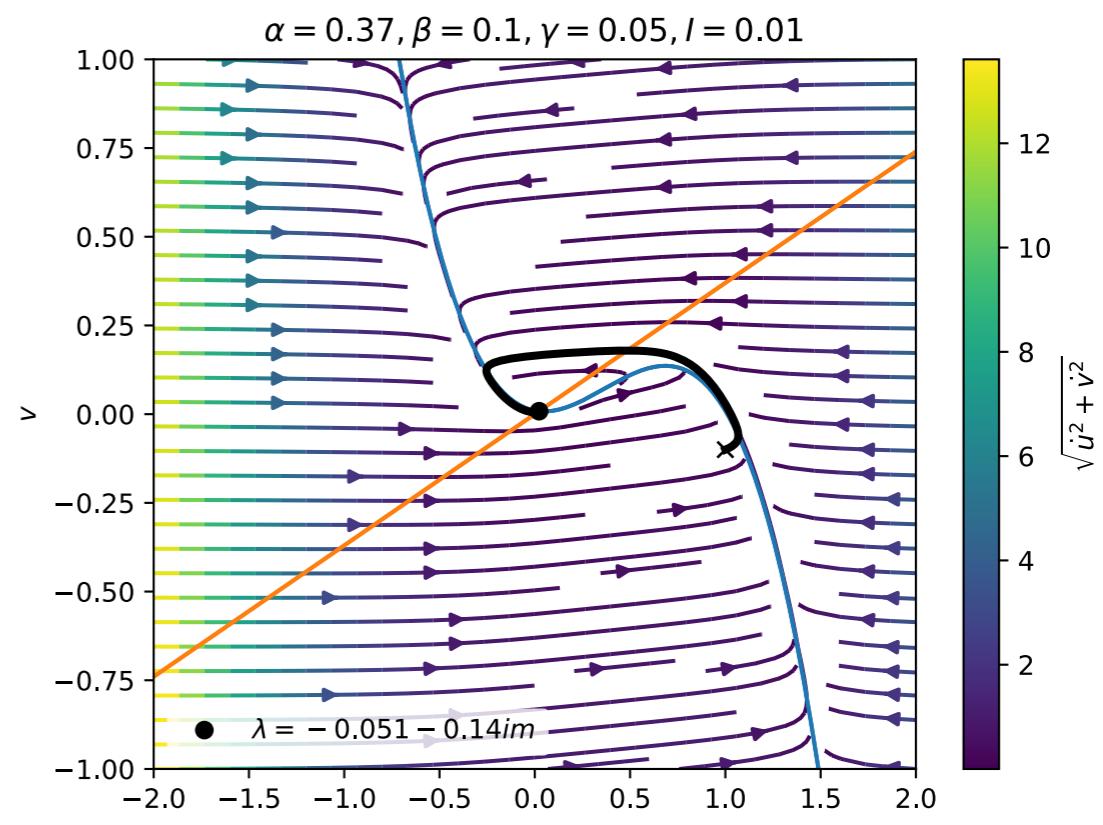
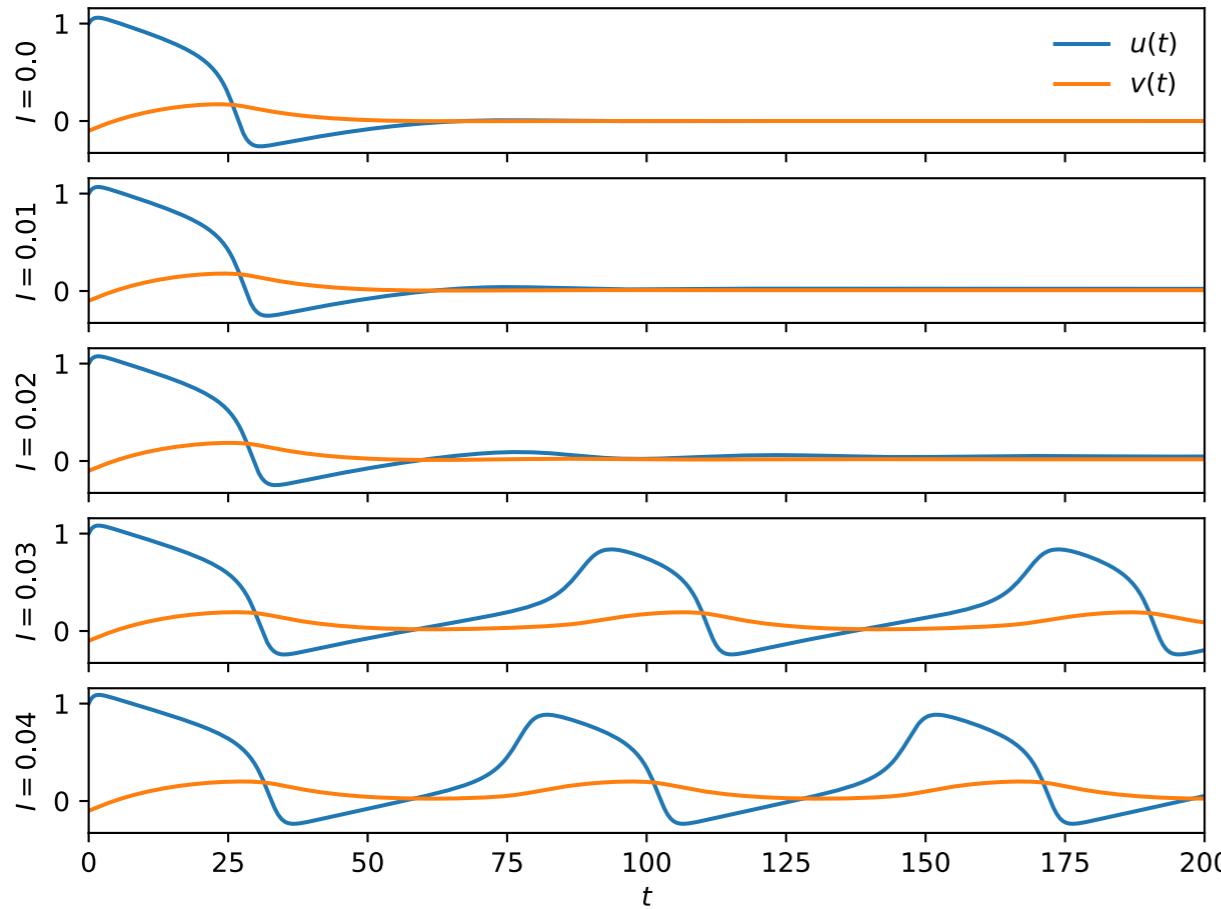


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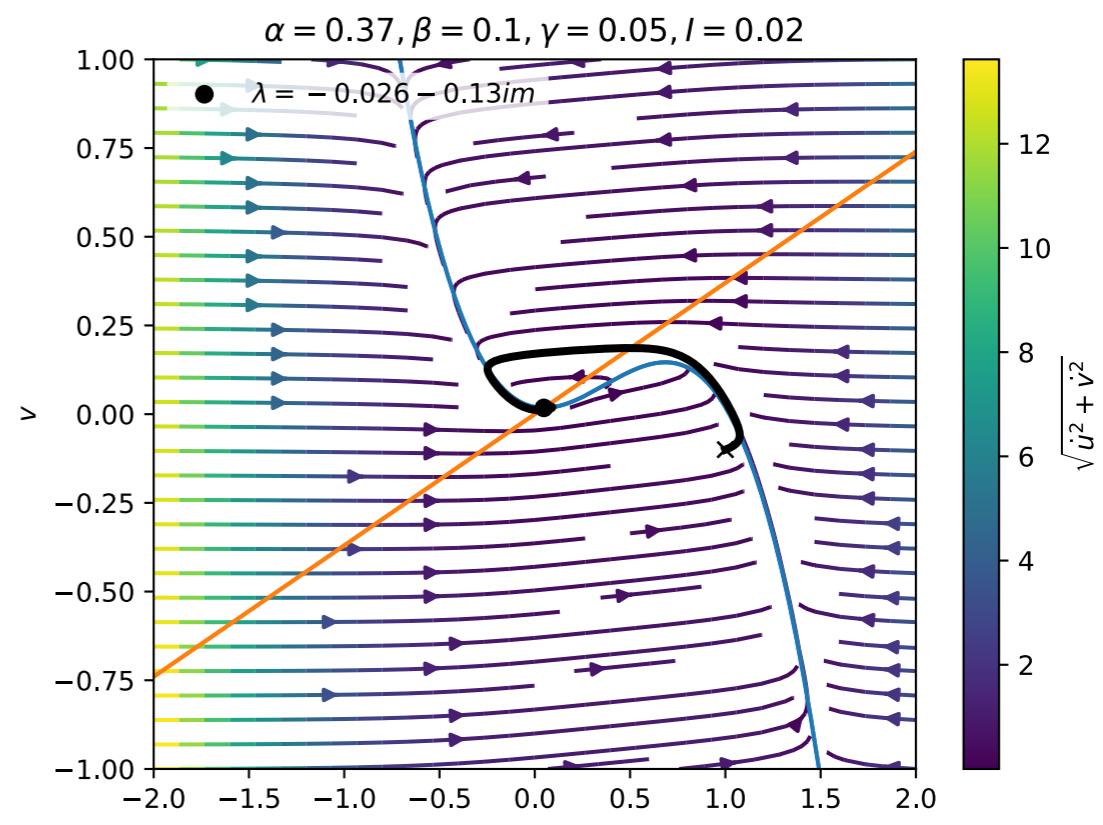
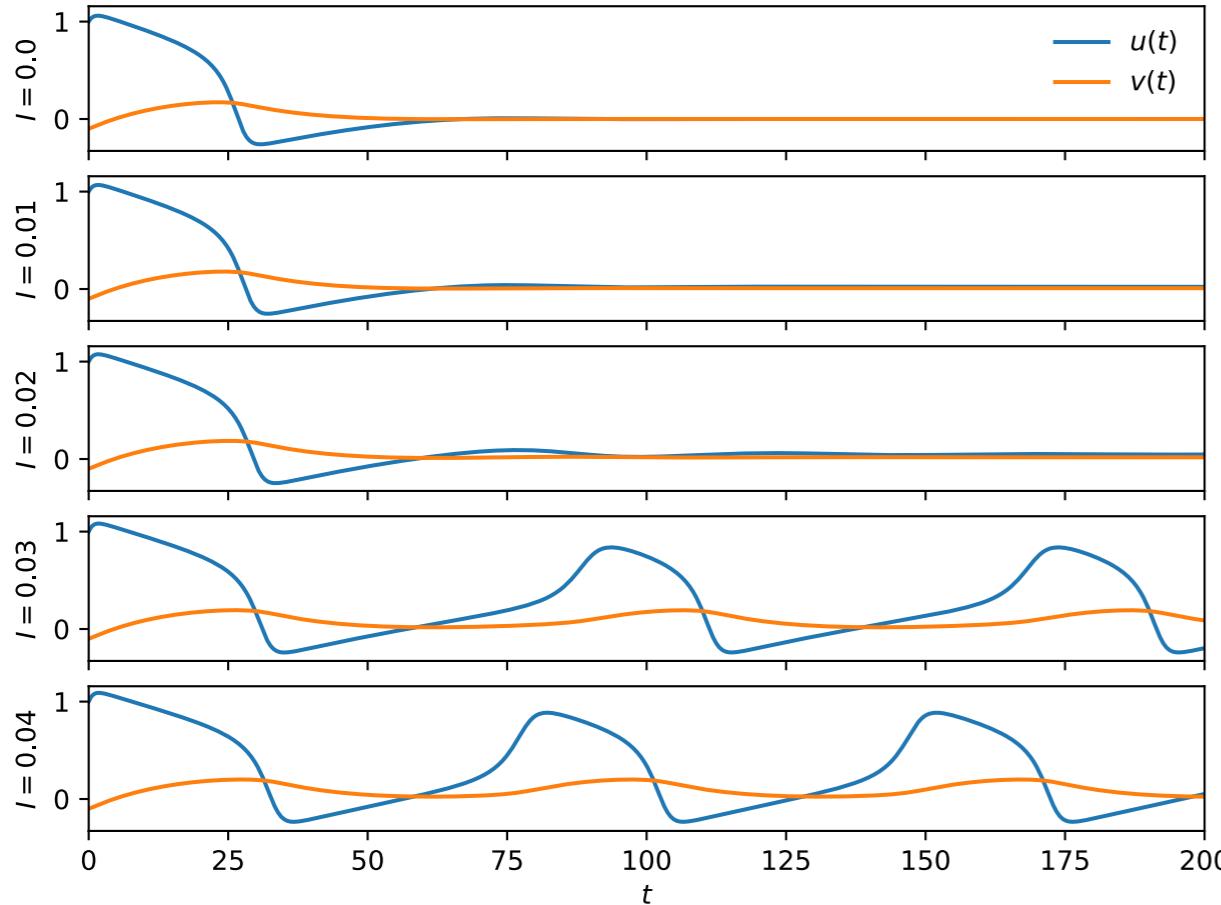


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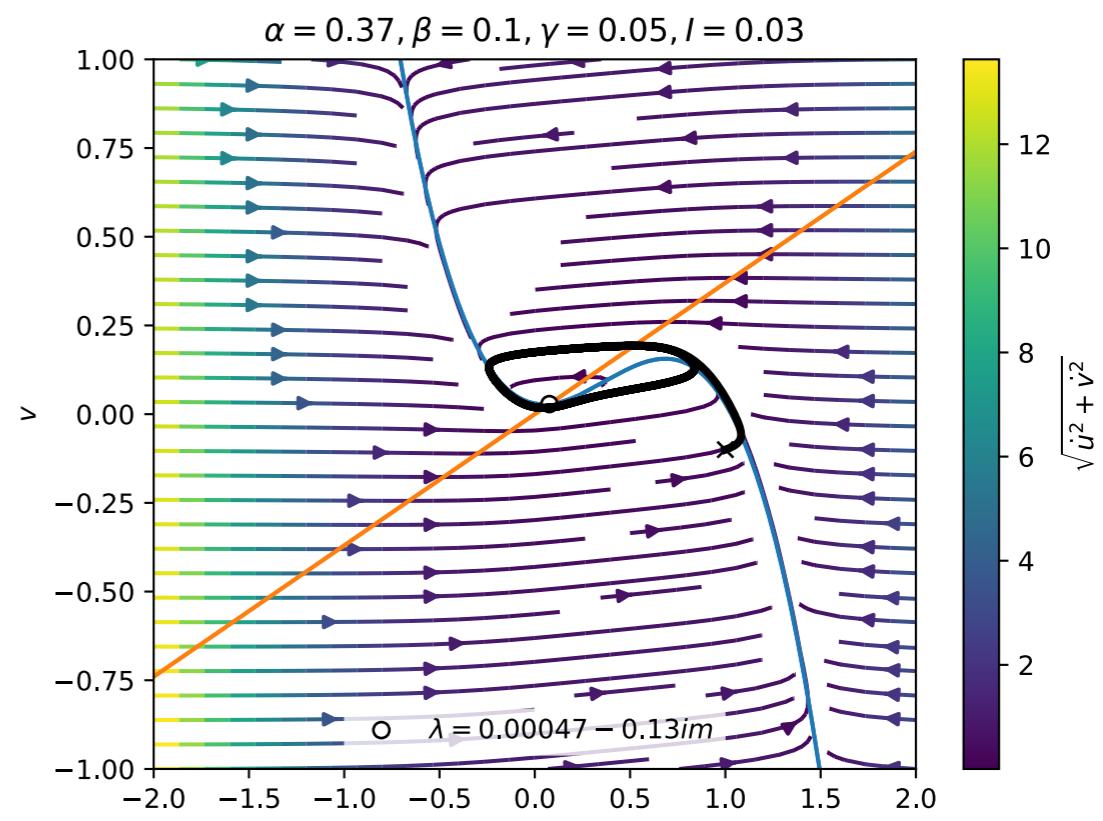
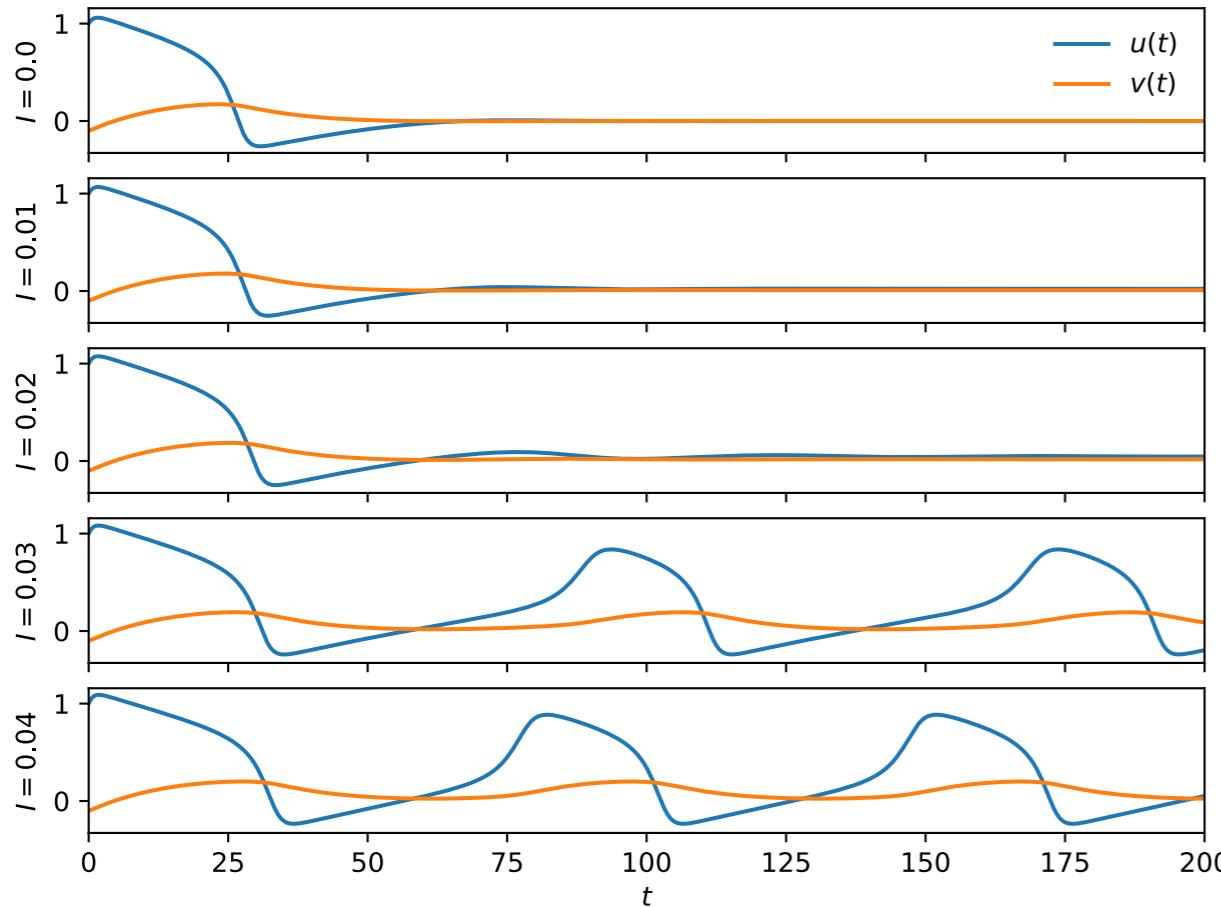


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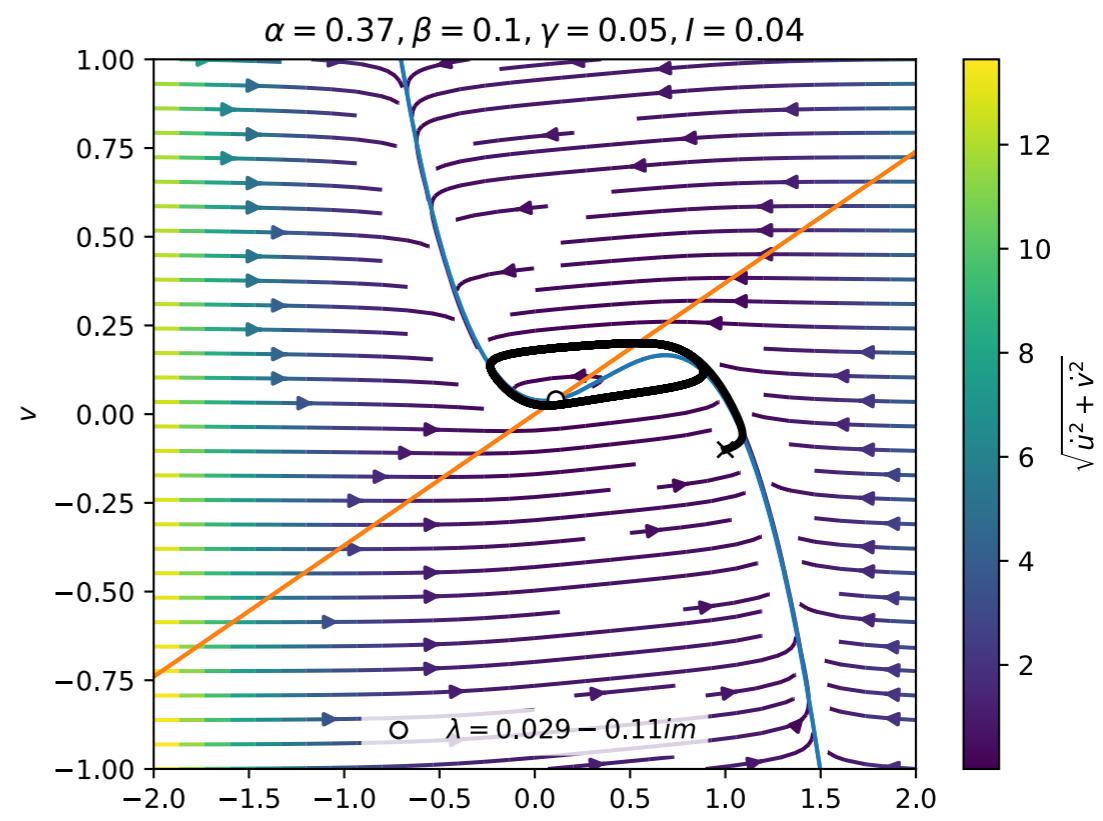
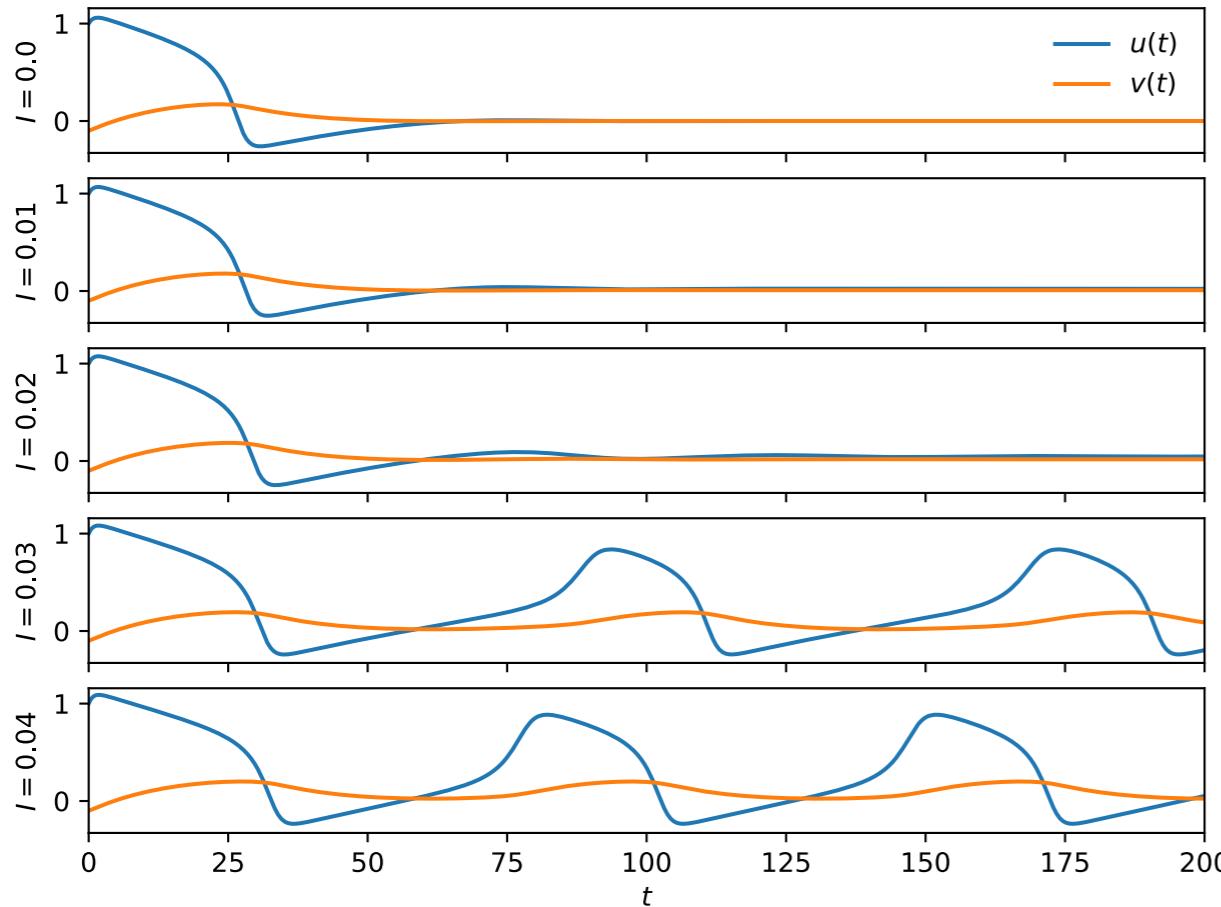
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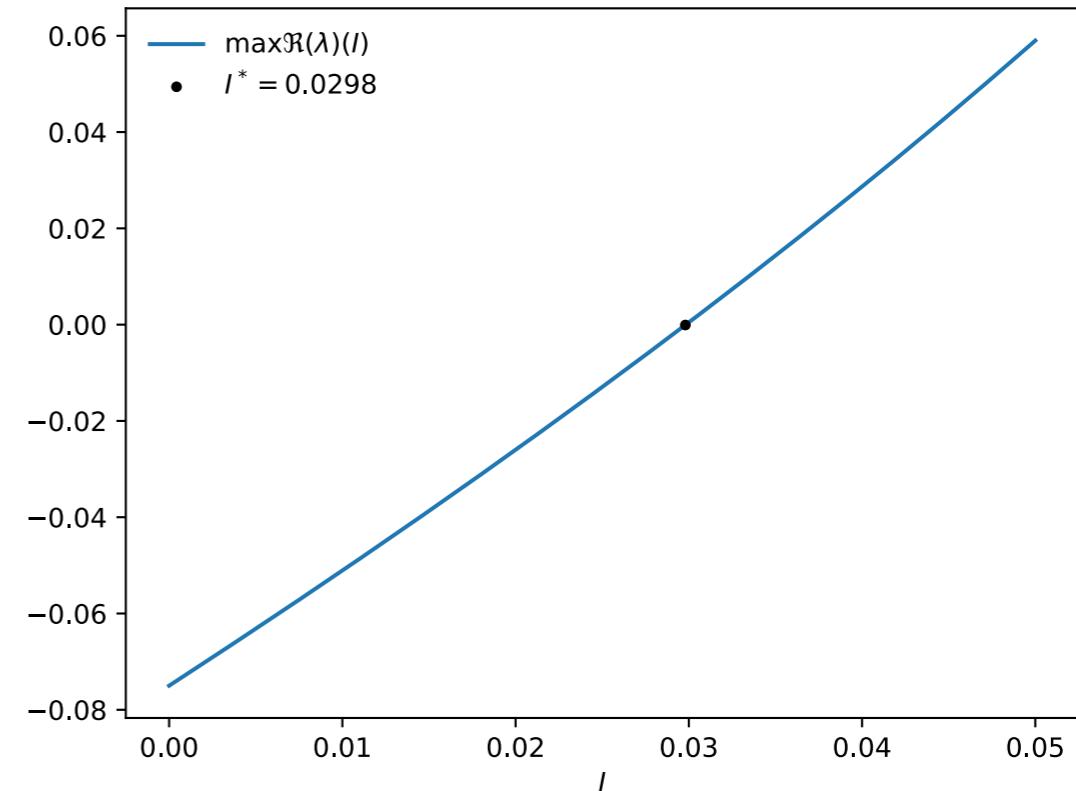
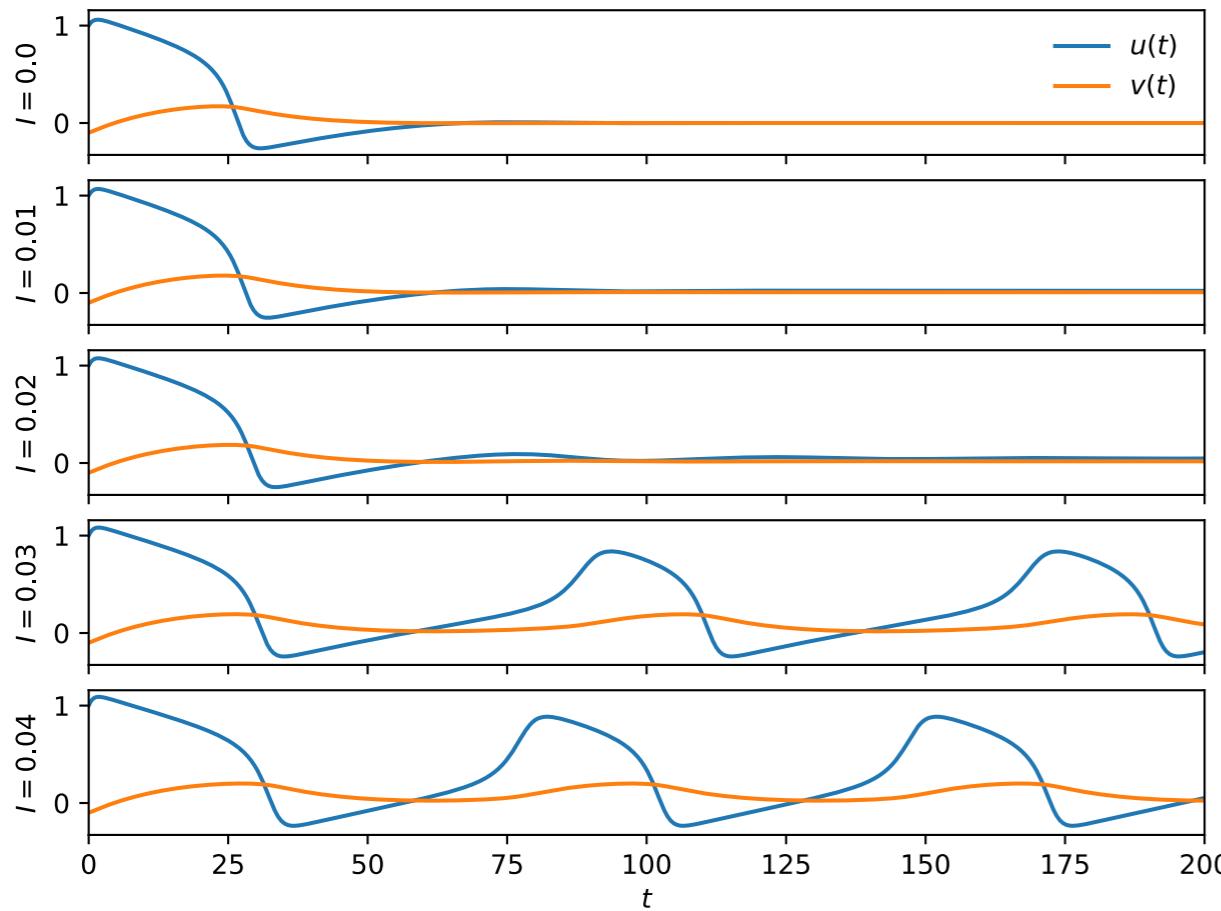
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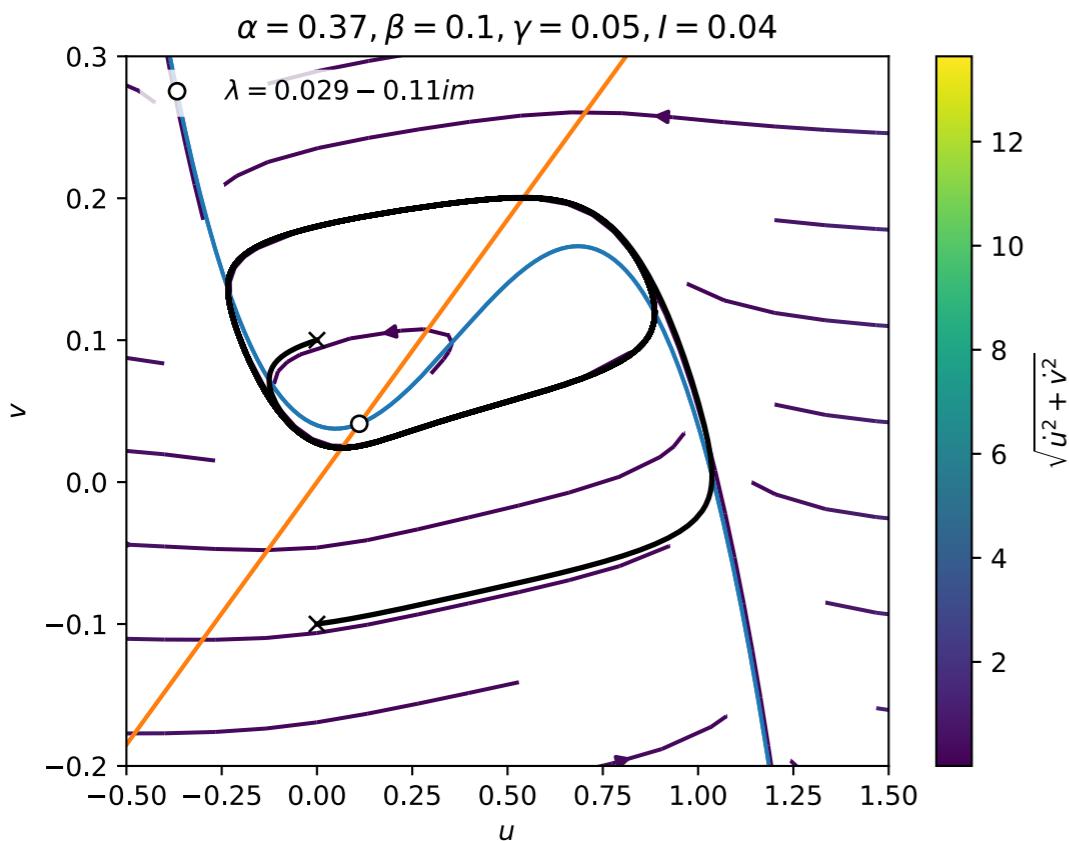


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# Periodic Orbits

## Limit Cycles and Poincaré-Bendixson

**“The Poincaré-Bendixson theorem is one of the central results of nonlinear dynamics. It says that the dynamical possibilities in the phase plane are very limited: if a trajectory is confined to a closed, bounded region that contains no fixed points, then the trajectory must eventually approach a closed orbit. Nothing more complicated is possible.” — *Nonlinear Dynamics and Chaos, with Applications to Physics, Biology, Chemistry, and Engineering*, S. H. Strogatz, §7.3**

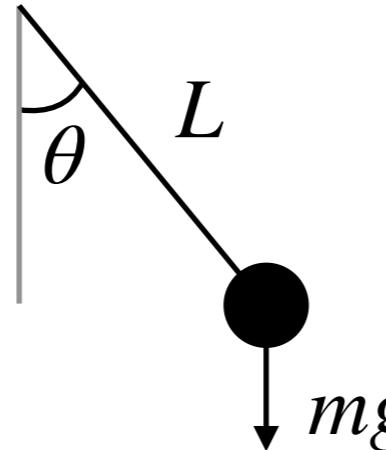


- A corollary to this is that we can not exhibit chaotic dynamics with two or less state variables related by ordinary differential equations
- An example of a chaotic 3D system will be looked at in detail later!

# FHN Oscillations

What makes them different from the pendulum?

- $\frac{d^2\theta}{dt^2} + \frac{g}{L} \sin \theta = 0$



- Rescale time by natural frequency:  $\ddot{\theta} + \sin \theta = 0$ 
  - $\dot{\theta}(\ddot{\theta} + \sin \theta) = 0$  implies energy  $E(\theta, \dot{\theta}) = \dot{\theta}^2/2 - \cos \theta$  is constant!
- $\dot{\theta} = \omega, \dot{\omega} = -\sin \theta$

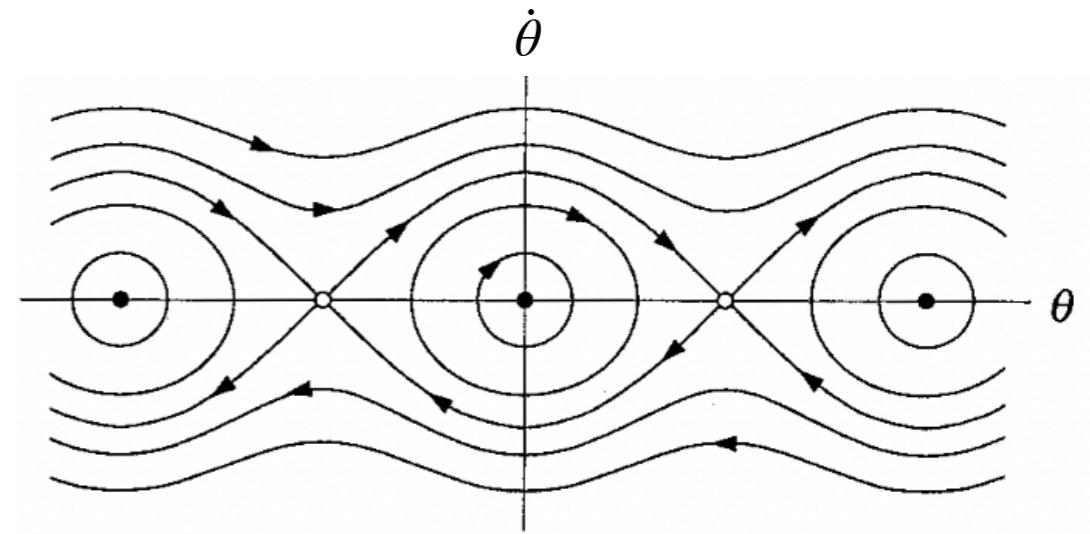


Figure 6.7.3 Strogatz

- The trajectories are just the isocontours of  $E(\theta, \dot{\theta})$
- No attracting or repelling orbits – all separated by energy

# Other slow-fast systems

Keener's listing of slow-fast systems