



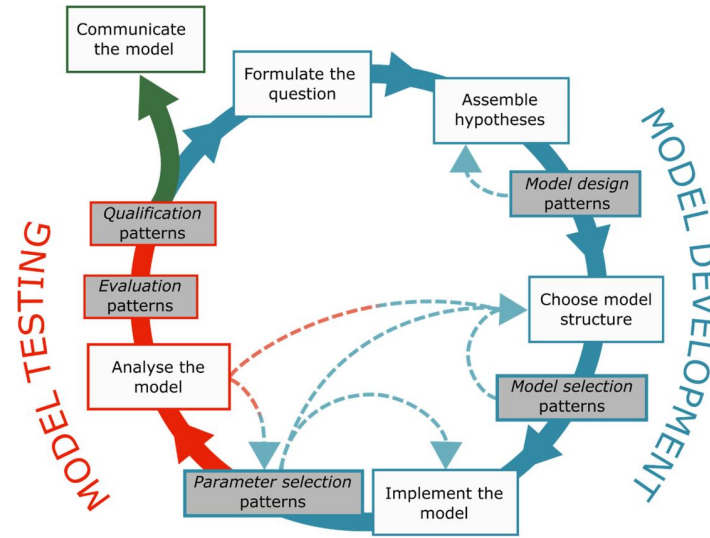
# Emergent phenomena and properties of dynamical models



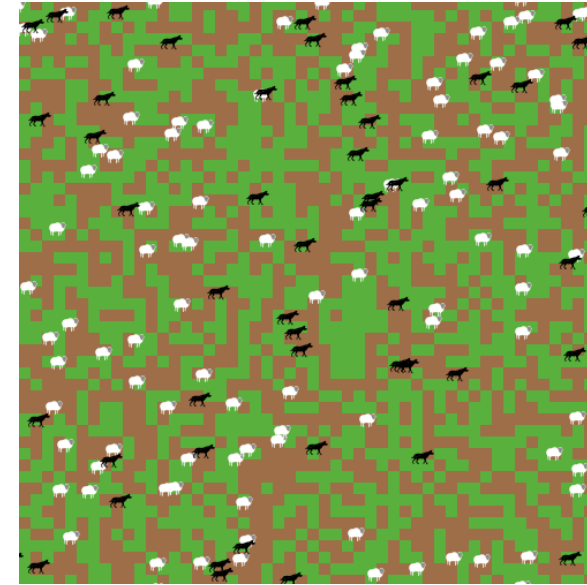
THE UNIVERSITY of EDINBURGH  
**informatics**

**Modelling of Systems for Sustainability**  
INFR10088

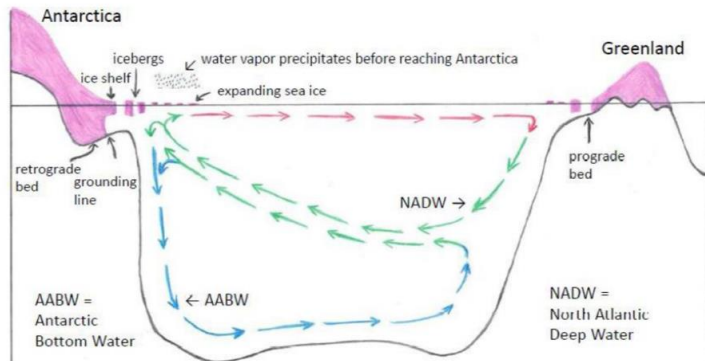
So far...



The modelling cycle

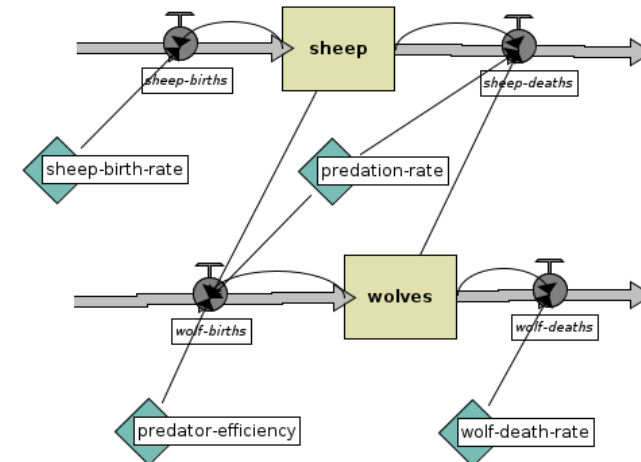


Agent-based modelling...



Real systems, e.g. Atlantic Meridional Overturning Circulation

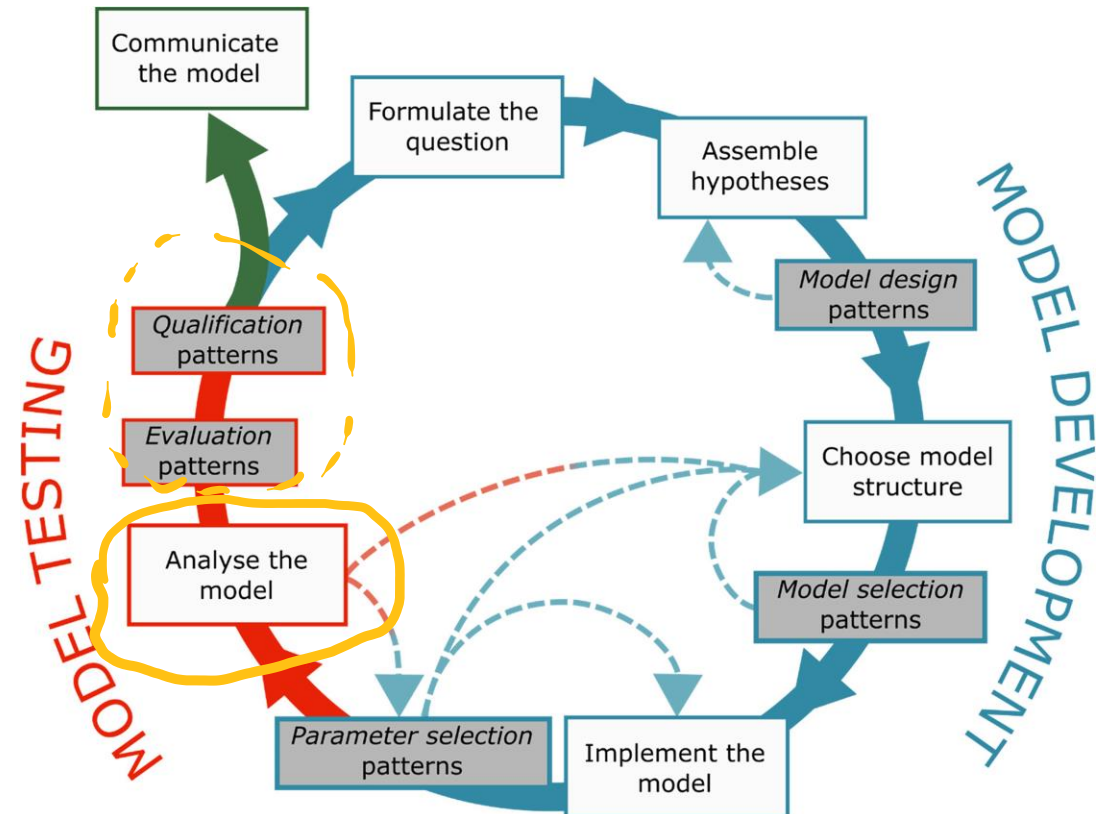
Overview  
Design concepts  
Details



...and system dynamics

# Today...

- **Design concept: Emergence**
- **Model analysis: Properties of dynamical systems**
- **Some:**
  - Quantification of patterns
  - Exploration of parameter space



# Overview - aims

- Overall aim (Course Learning Outcome 1):  
"explain how computational modelling frameworks can be used to understand the behaviours of complex interacting systems involved in sustainability such as social, economic and ecological systems"
- This lecture:
  - Emergent Phenomena
  - Properties of dynamical models
  - Example system

What do we mean by "emergent phenomena"?

# What do we mean by emergence?

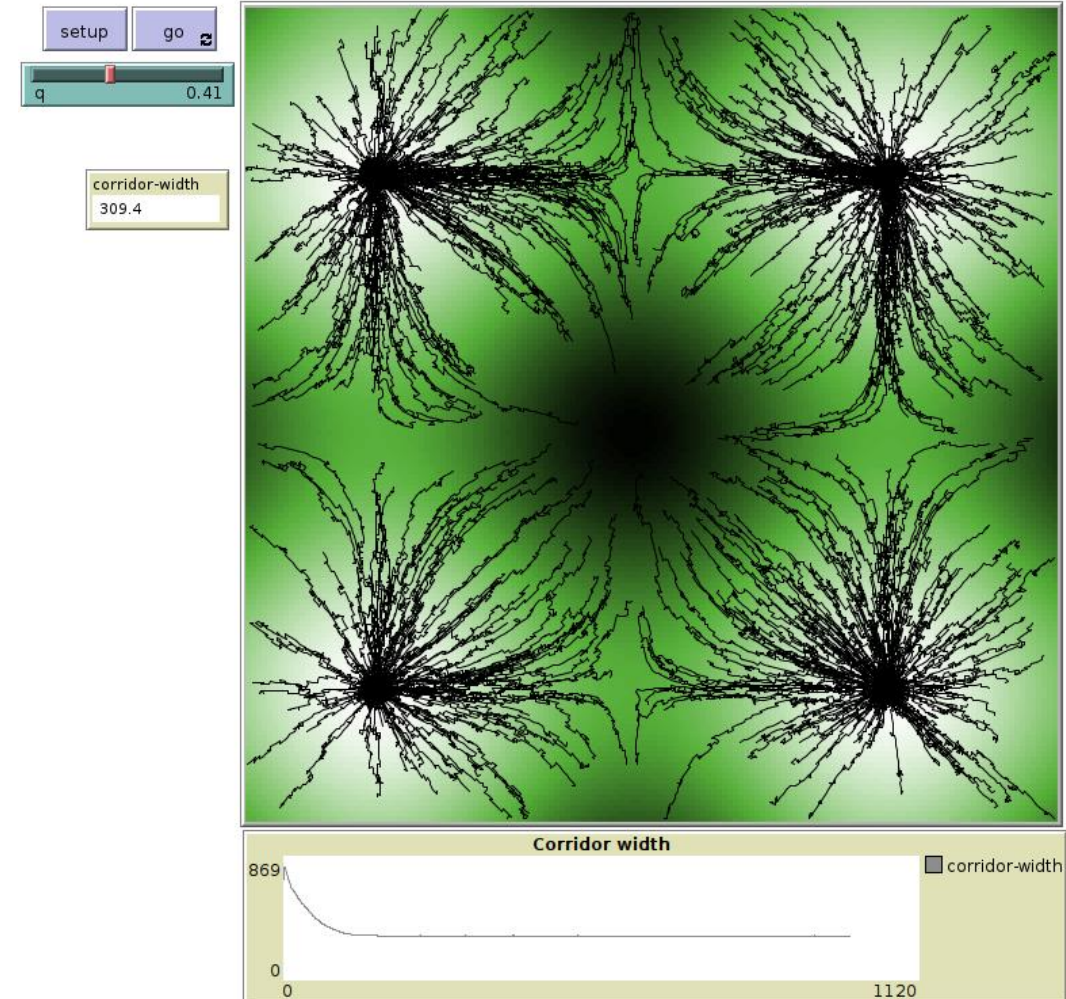
- OED: **emergence**. An unforeseen occurrence.
- OED: **emerge**. Come into being with the passage of events.
- Railsback & Grimm Chapter 8: **emerge**. "Arise in relatively complex and unpredictable ways"
- Cf **impose**: "forced to occur in direct and predictable ways"
- R&G **unpredictable**: "outcomes difficult or impossible to predict just by thinking"
- Explain things by simulation - "can you grow it?"

# Qualitative criteria for emergence (Railsback and Grimm)

- It is not simply the sum of the properties of the model's individuals
- It is a different type of result than individual-level properties or decisions
- It cannot easily be predicted from the properties of the individuals

# Butterfly corridors – a good example of emergence?

- It is not simply the sum of the properties of the model's individuals
  - Yes-ish – corridors are perhaps a bit more than a sum of butterfly locations
- It is a different type of result than individual-level properties or decisions
  - Yes
- It cannot easily be predicted from the properties of the individuals
  - We would probably expect more noise to give wider corridors – do they?





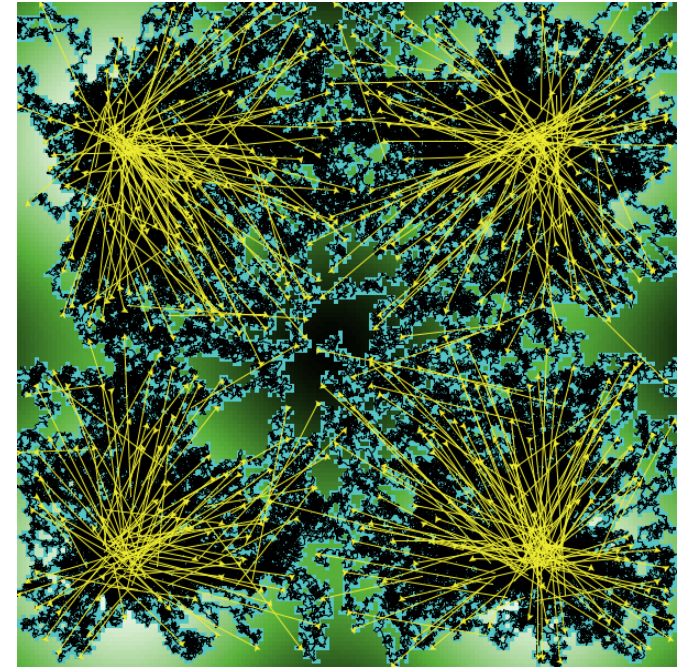
# Quantifying corridor width

Corridor width =

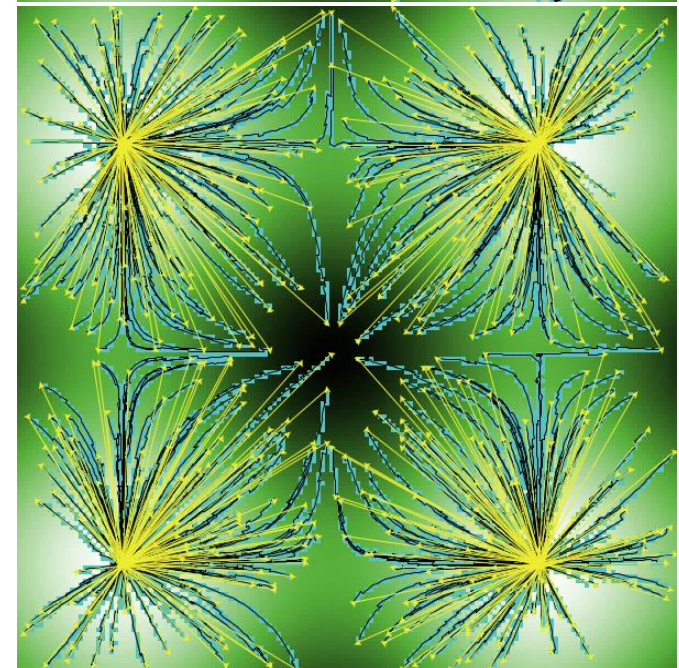
(#patches visited)/

(Mean distance between starting point and end point of each butterfly)

$q=0.04$   
Corridor  
width = 663

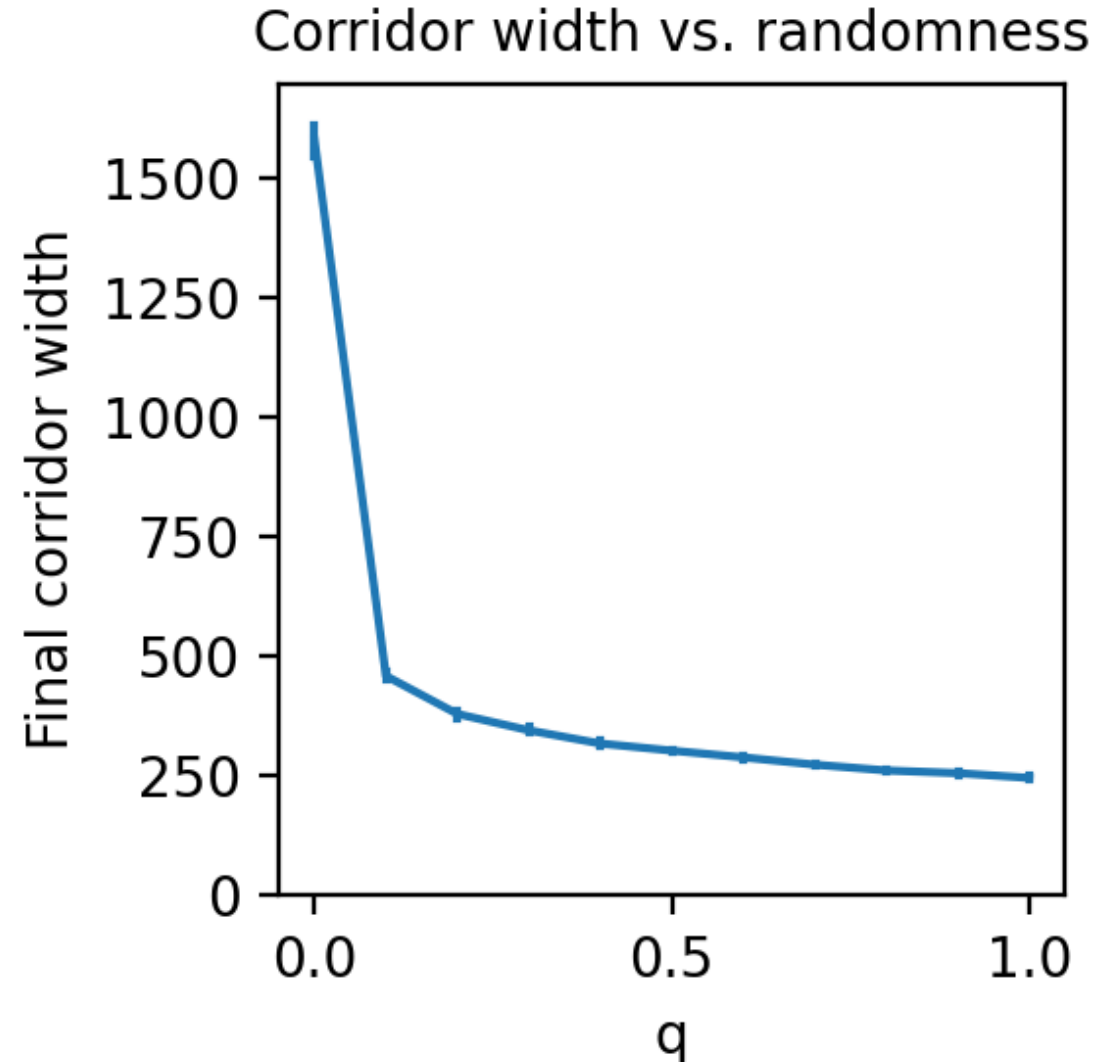


$q=0.79$   
Corridor  
width = 269



# Butterfly corridor width as function of randomness

- Results of running an experiment using NetLOGO BehaviourSpace:
  - Parameter  $q$  varied in steps of 0.1
  - 10 replications for each value of  $q$
  - Data exported and plotted using Python/Matplotlib/Seaborn
  - Mean and S.D. shown
- Corridor width increases with increasing randomness (decreasing  $q$ )
  - as expected, so perhaps not very emergent – but agent behaviour is very simple.



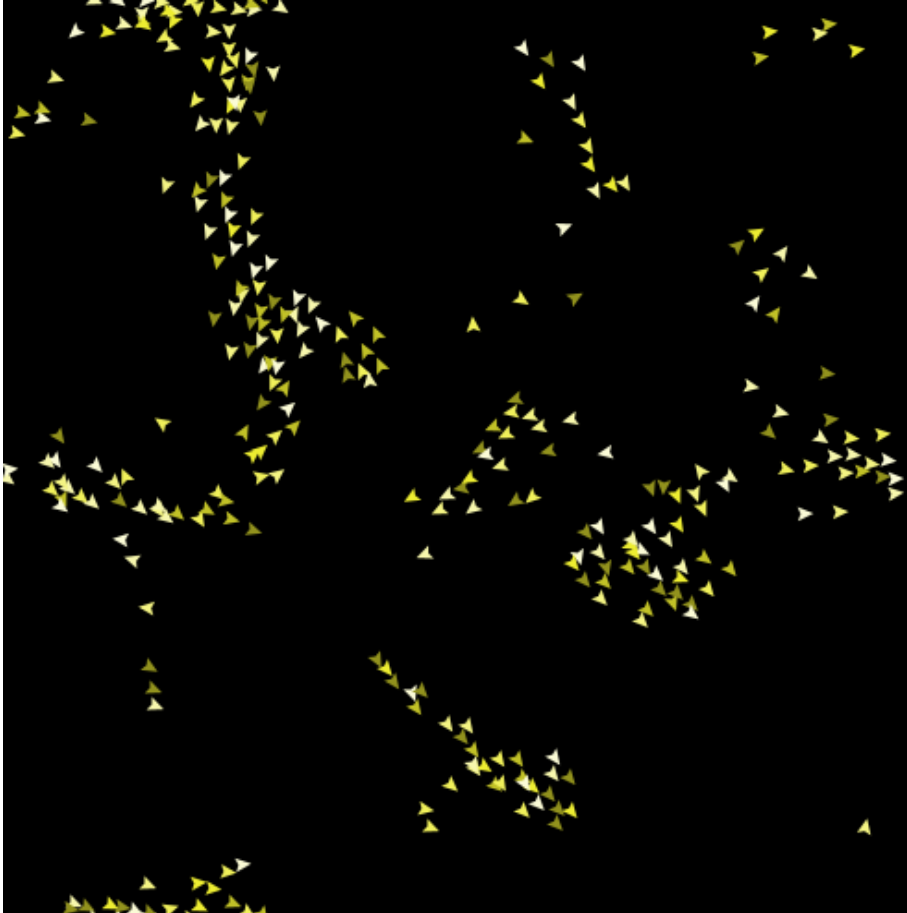
# The ideal amount of emergence?

- Highly imposed results, not enough emergence
  - Too predictable
- Very complex model, complex emergent behaviours
  - Too difficult to understand and learn from
- Complex behaviours emerge from apparently simple rules
  - Just right!
- Crucial that the model is appropriate for question



The Three Bears, pictured by John R Neill (1908).  
Chicago: Reilly & Britton

# A good example of emergence? - The NetLOGO flocking model



- It is not simply the sum of the properties of the model's individuals?
- It is a different type of result than individual-level properties or decisions?
- It cannot easily be predicted from the properties of the individuals?

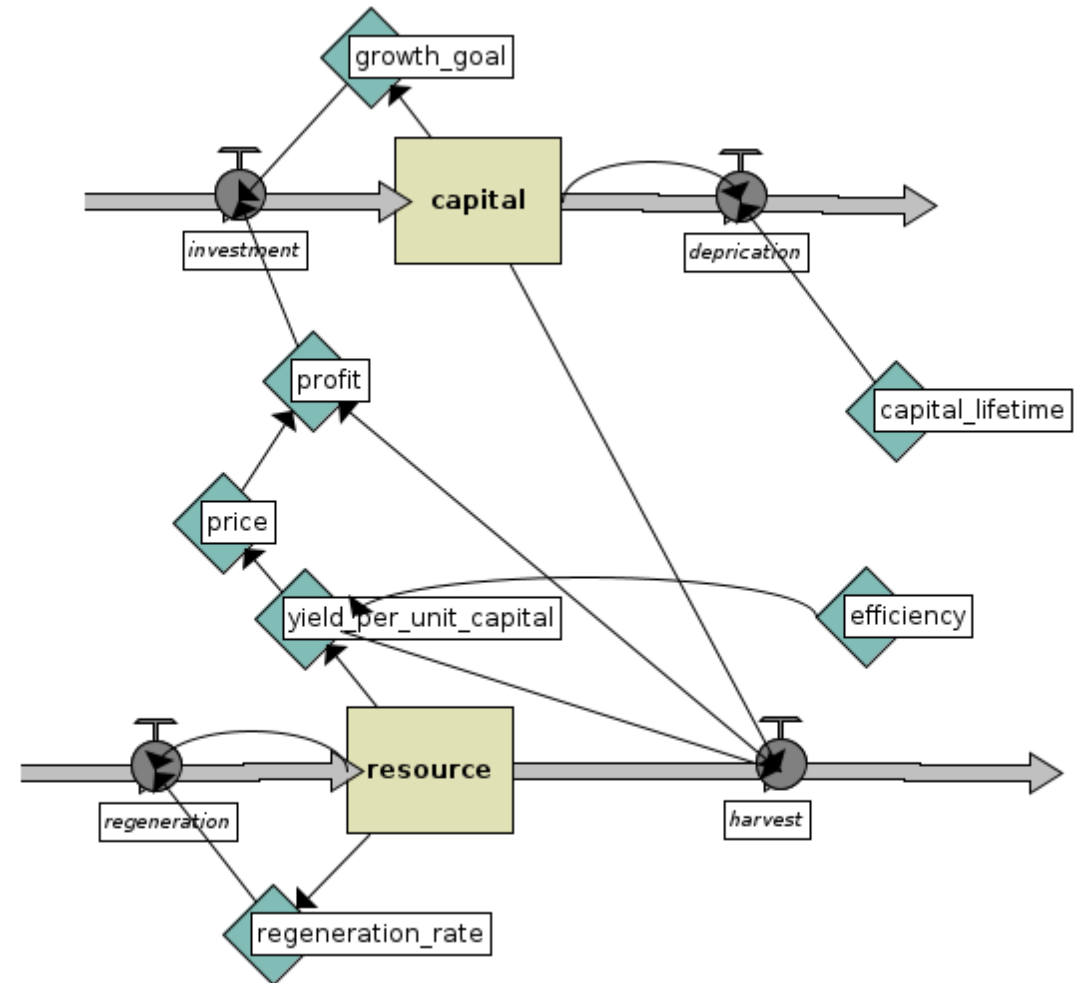
Properties of dynamical systems

# Deterministic dynamical system (system dynamics)

- System dynamics models are a special case of a mathematical dynamical system composed of differential equations
- Such systems have characteristic behaviours:
  - Steady state behaviour
  - Oscillating behaviour
  - For 3 stocks (state variables) or more, chaotic behaviour
- The **same model** can exhibit more than one behaviour depending on **its parameter** setting

# Example 1: Fishing economy (Meadows Chapter 2, Fig 42)

- Renewable **resource** stock: fish
- Renewable **capital** stock: fishing boats
- Fish **harvest** depends on **capital**, **resource** and **efficiency** (a parameter)
- **Price** gets higher when there is scarcity
- More **profit** => more **investment** => more **capital** => more **harvest** => more **profit**
  - But more **harvest** => less **resource**

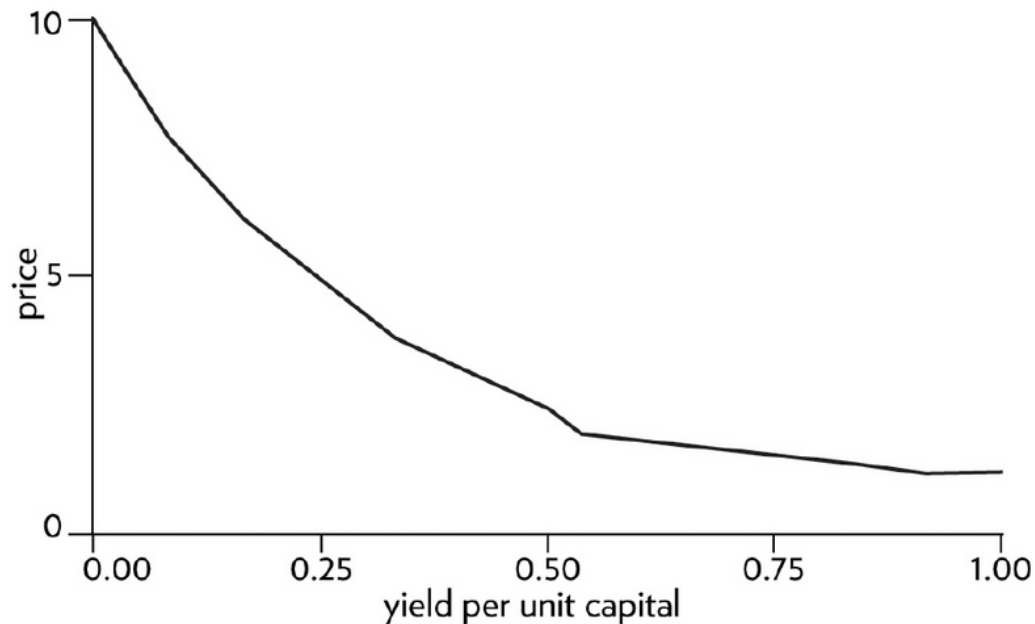


# Aside: replication

## Meadows, *Thinking in Systems*, p. 227

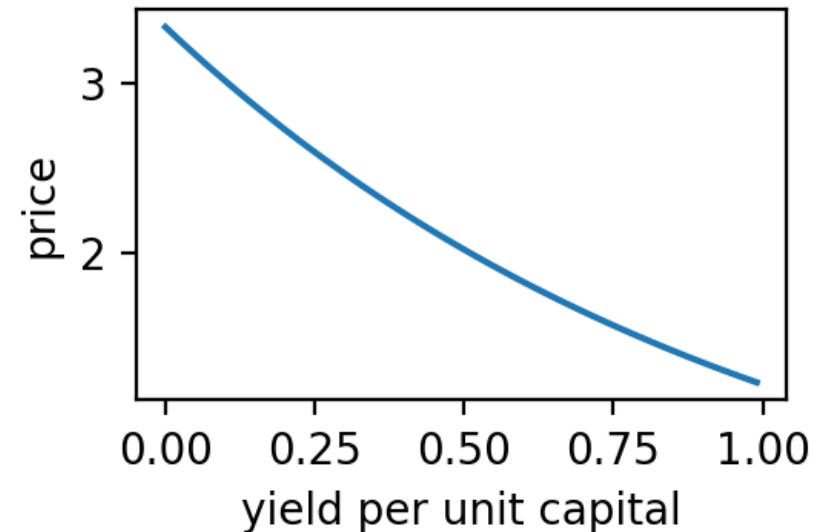
$$\text{profit} = (\text{price} \times \text{harvest}) - \text{capital}$$

*price* starts at 1.2 when yield per unit capital is high and rises to 10 as yield per unit capital falls. This is the same nonlinear relationship for price and yield as in the previous model.



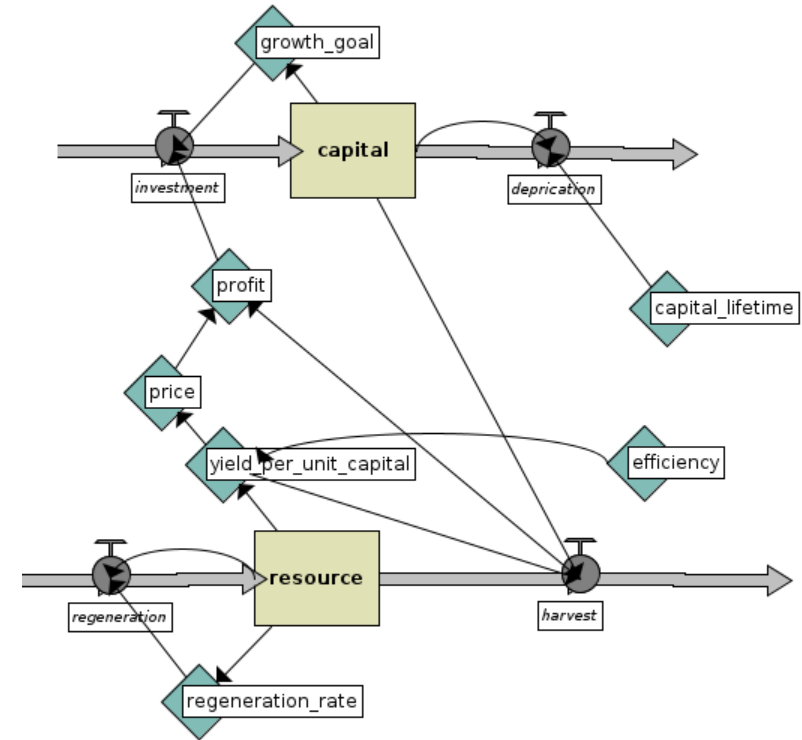
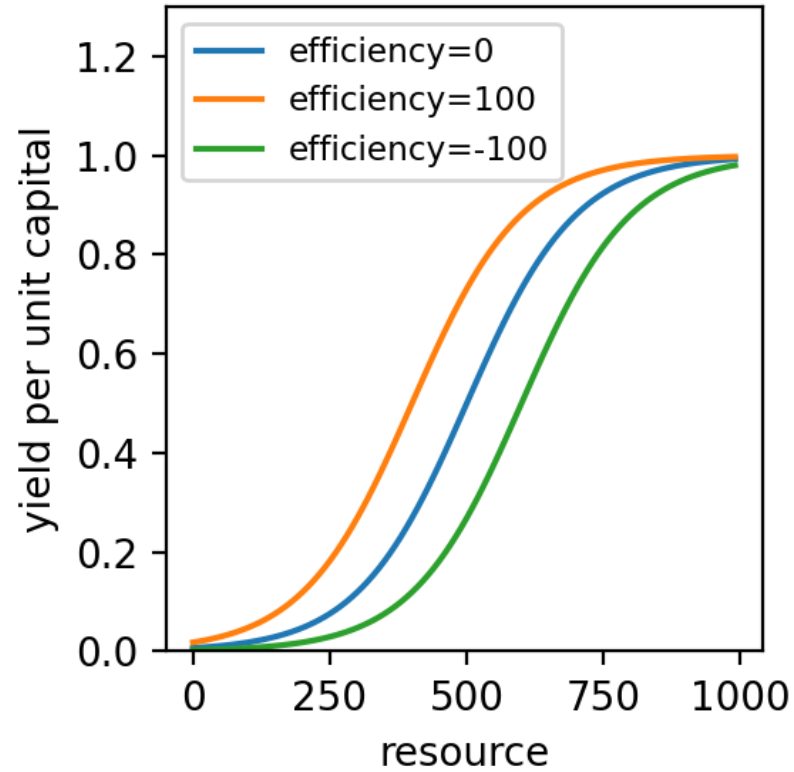
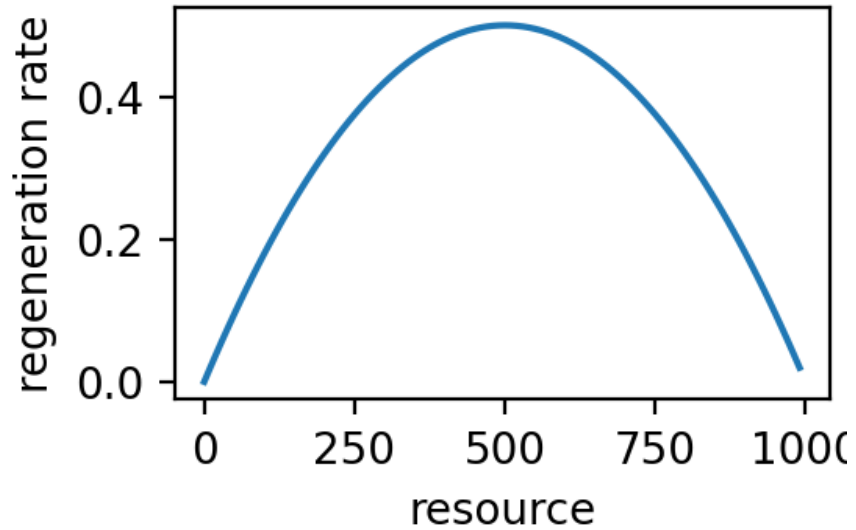
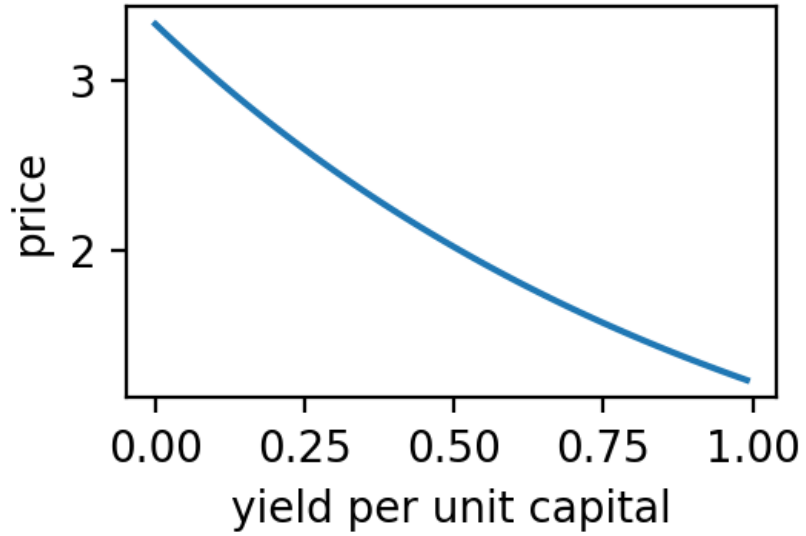
## Me

- Created a function that I *thought* was the same, but wasn't!
- Lesson: plot your functions before simulating



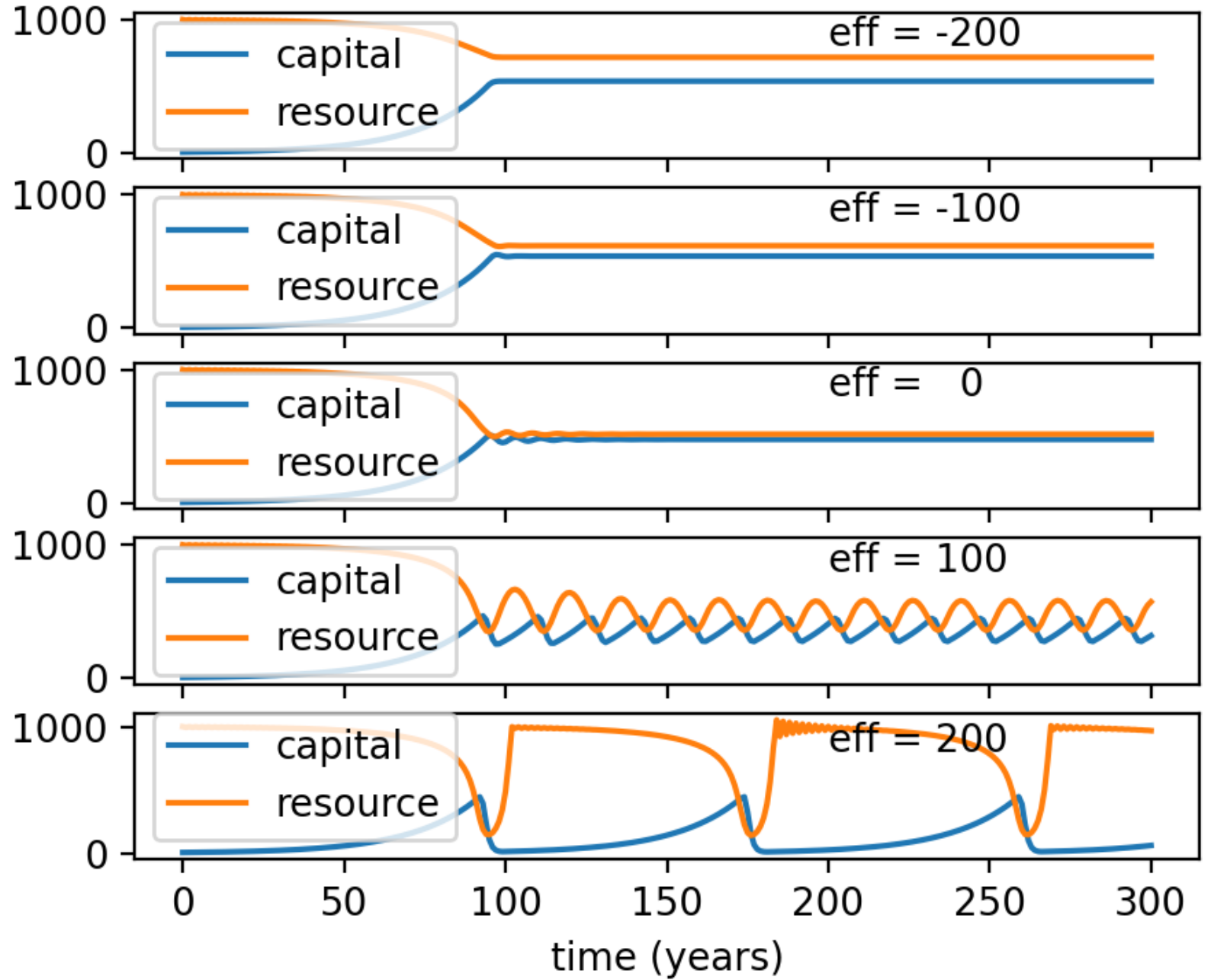


# Some key functions



# Parameter Search

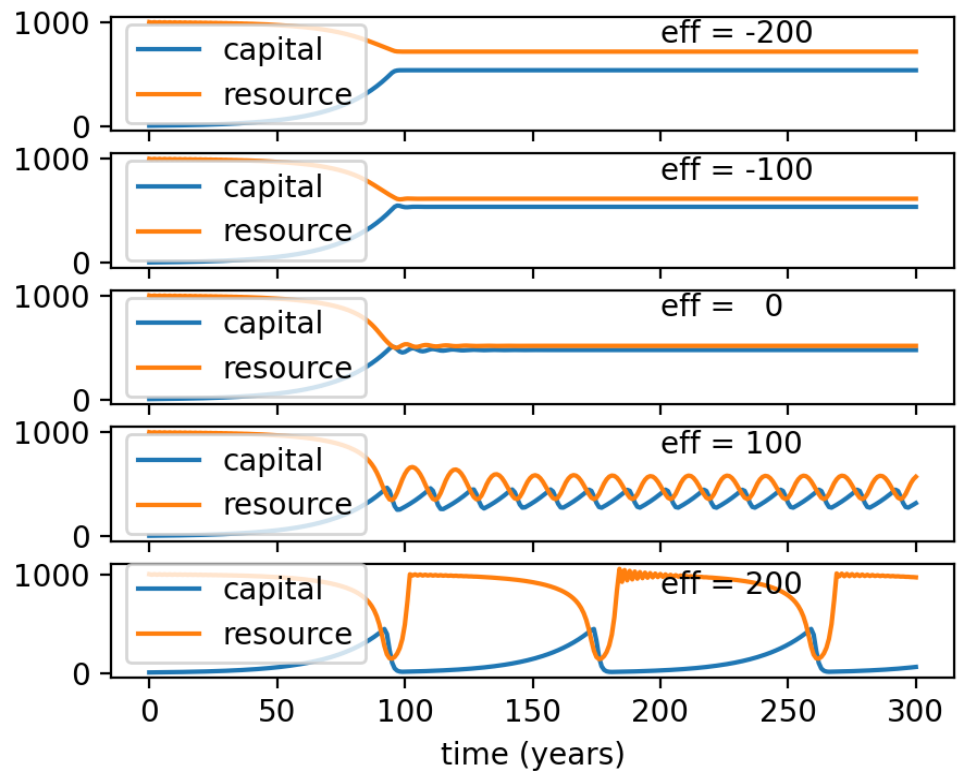
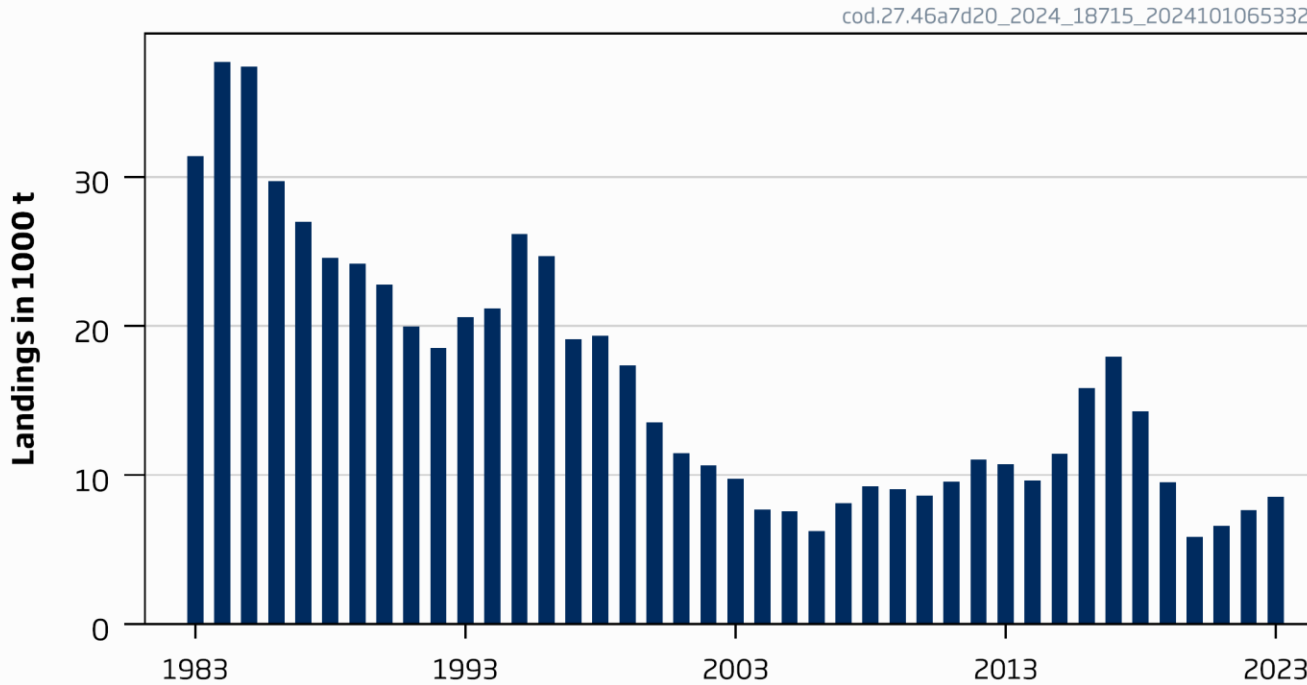
- Increase the efficiency parameter from low to high using BehaviourSpace
- What happens?
- Note **dynamic stability**



# Data versus model

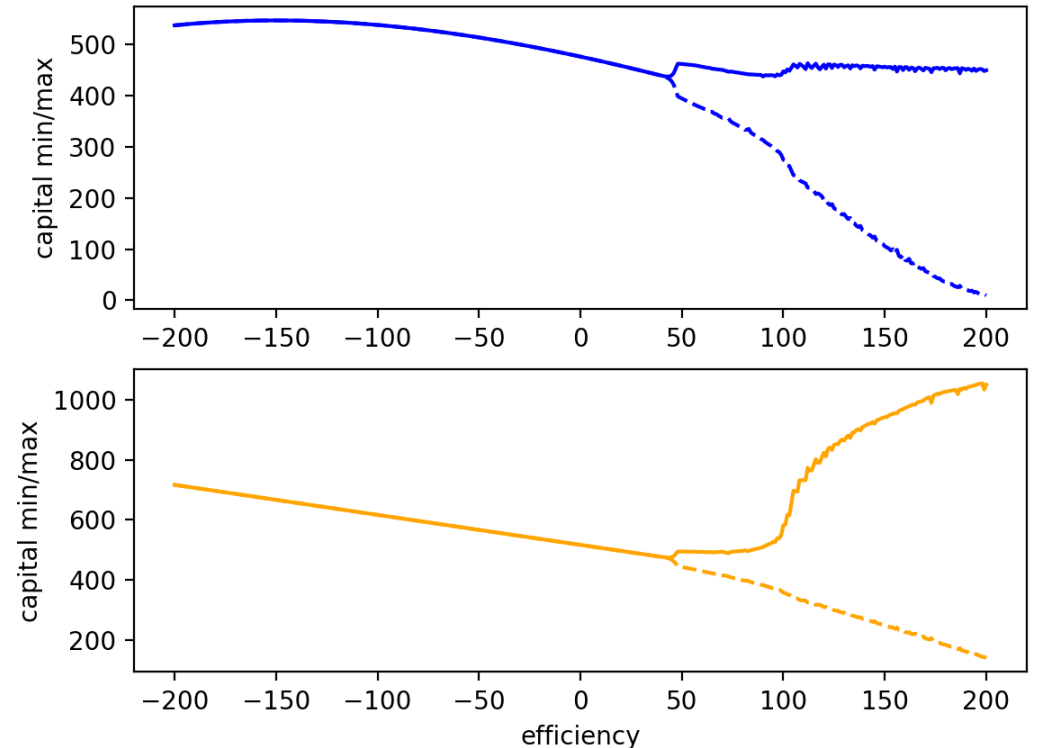
- Cod in Subarea 4, divisions 6.a and 7.d, and Subdivision 20 (North Sea, West of Scotland, eastern English Channel and Skagerrak)

## Model estimated catches



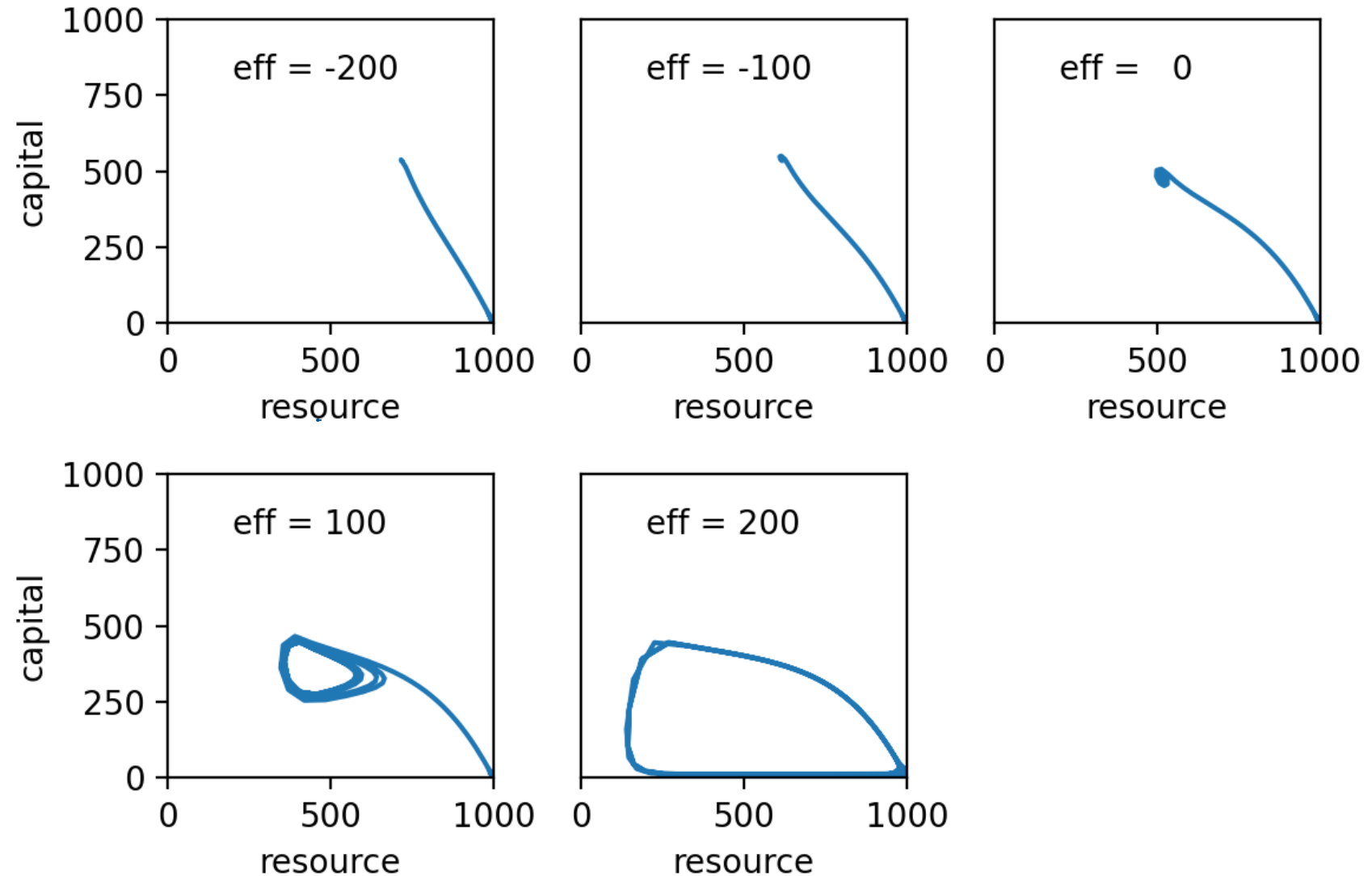
# Tipping point

- There is a **control parameter** that is gradually adjusted or changes
  - E.g. increasing efficiency
- At some value of the parameter the behaviour suddenly changes
  - E.g. an equilibrium point changes to a new value
  - Or the dynamical behaviour of the system changes
- In **System Dynamics** models this is a "**Tipping point**"
- Known as a "**Bifurcation**" in the mathematical field of **dynamical systems**

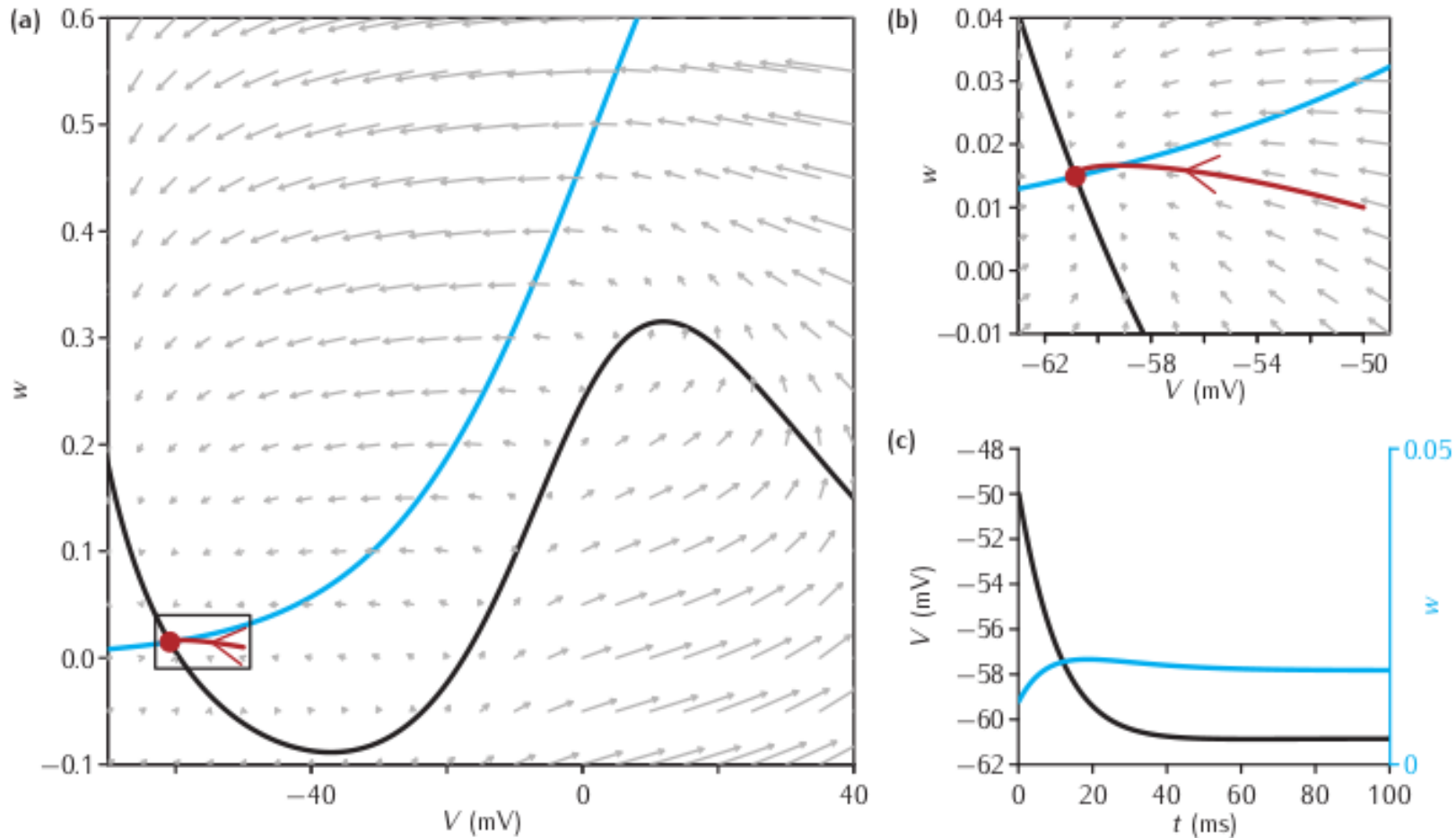


# Phase planes

- Equilibrium points, or **fixed points**, corresponding to **steady state** solutions
- These equilibrium points can be **stable** or **unstable**
- **Limit cycles**, i.e. oscillations

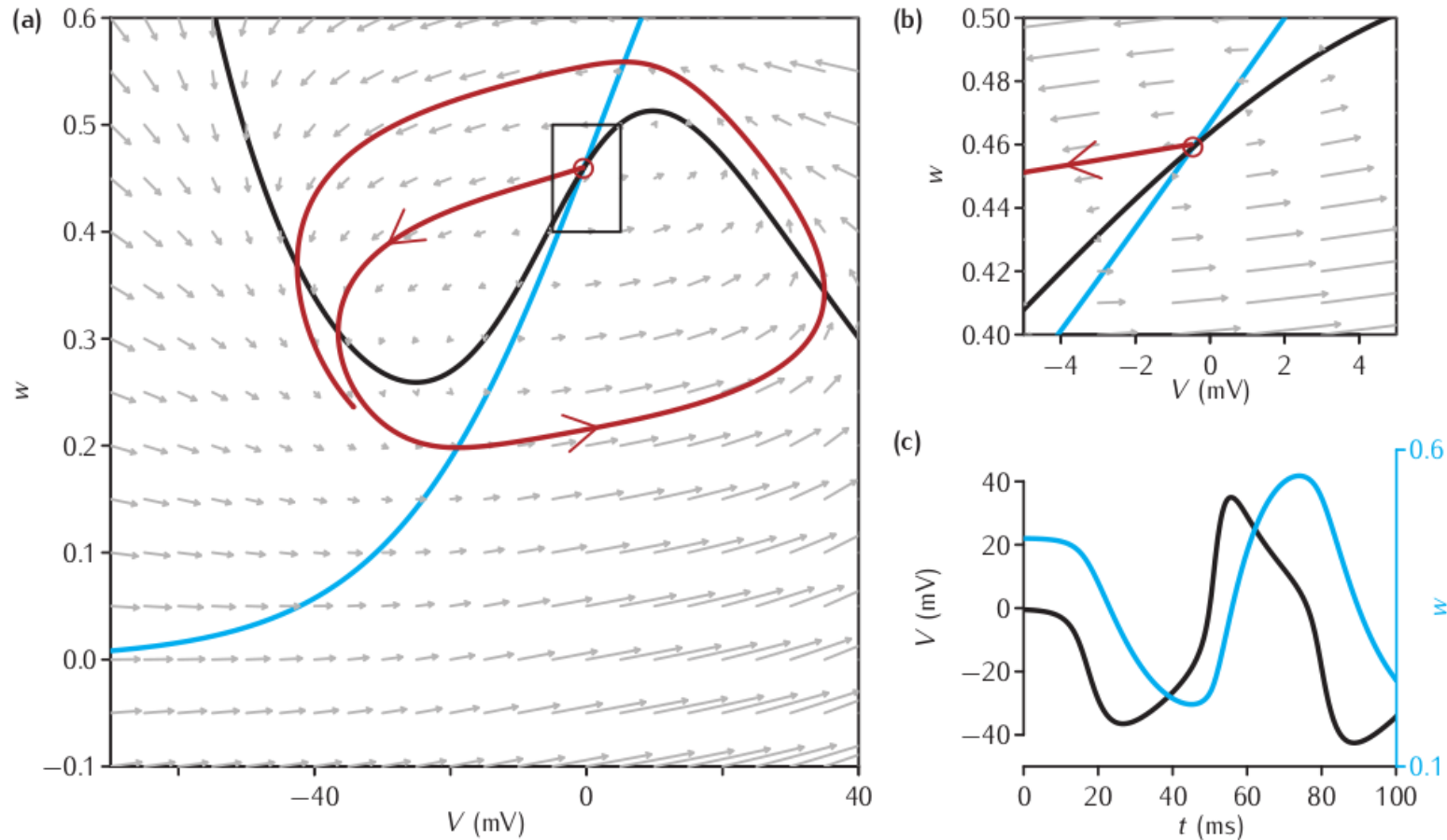


# Stable fixed point



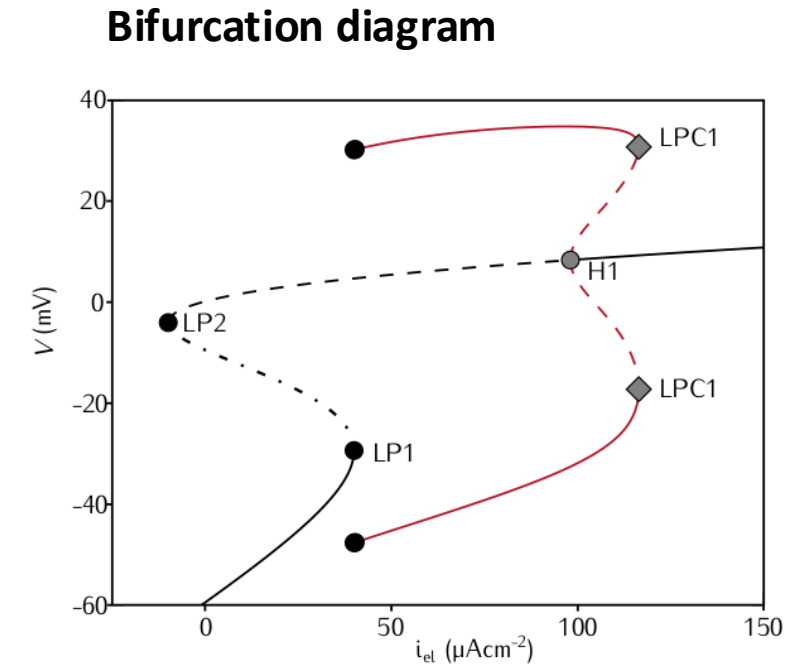
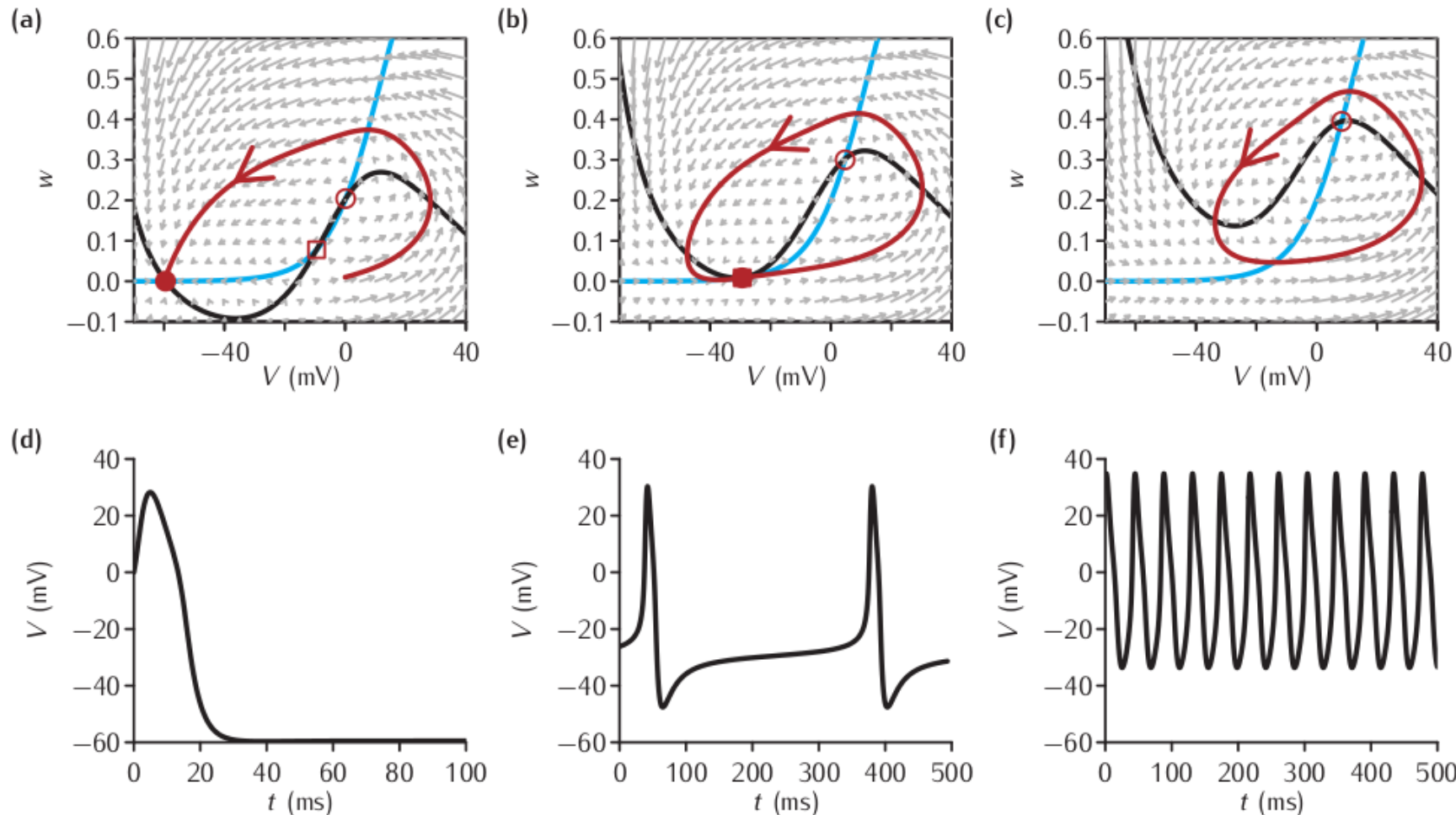
Sterratt, Graham, Gillies, Einevoll & Willshaw (2023), *Principles of Computational Modelling in Neuroscience*, CUP.

# Starting from an unstable fixed point and going into a limit cycle



Sterratt, Graham, Gillies, Einevoll & Willshaw (2023), *Principles of Computational Modelling in Neuroscience*, CUP.

# A saddle-node bifurcation



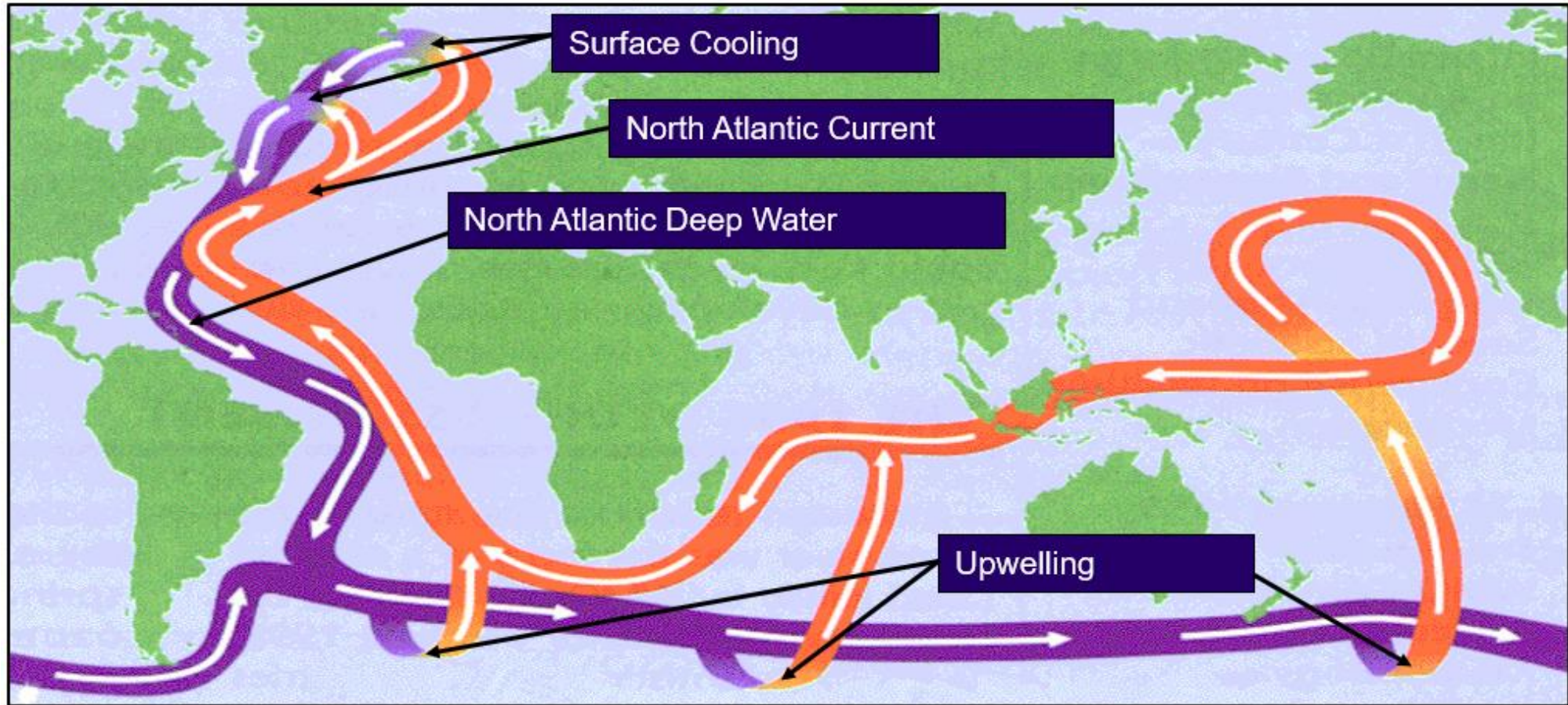
Sterratt, Graham, Gillies, Einevoll & Willshaw (2023), *Principles of Computational Modelling in Neuroscience*, CUP.



# Hysteresis

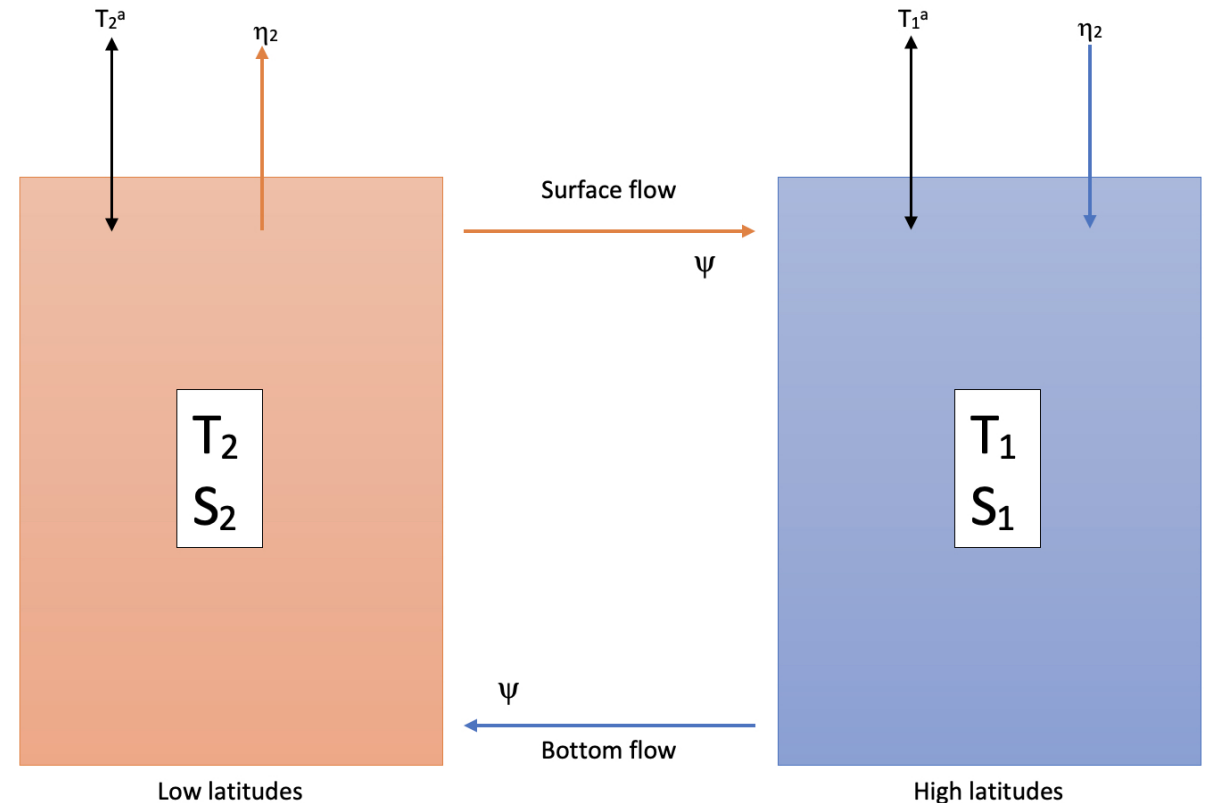
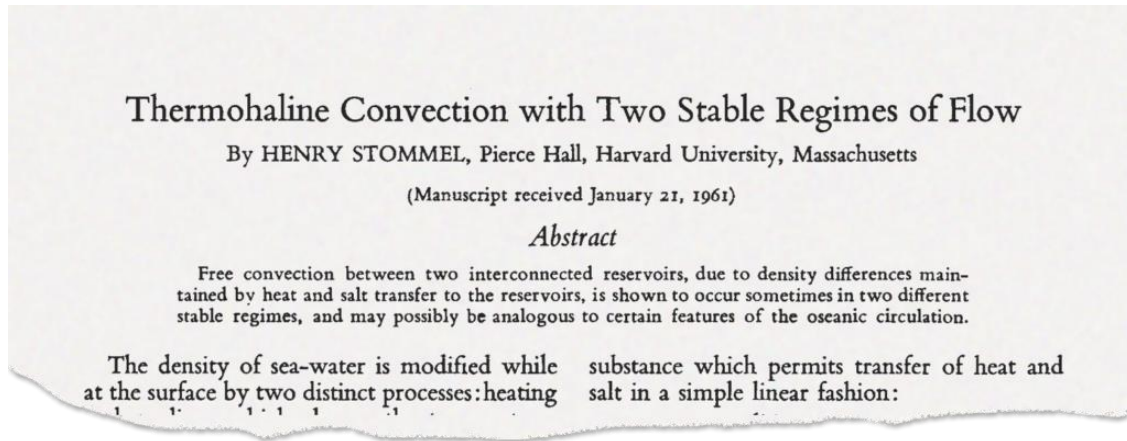
- Just taking the control parameter back past the tipping point value doesn't put the system back into the old state
  - E.g. lowering current (in the previous case) might not immediately restore the old behaviour
- This behaviour is called **hysteresis loop**

# Example 2: Atlantic Meridional Overturning Circulation



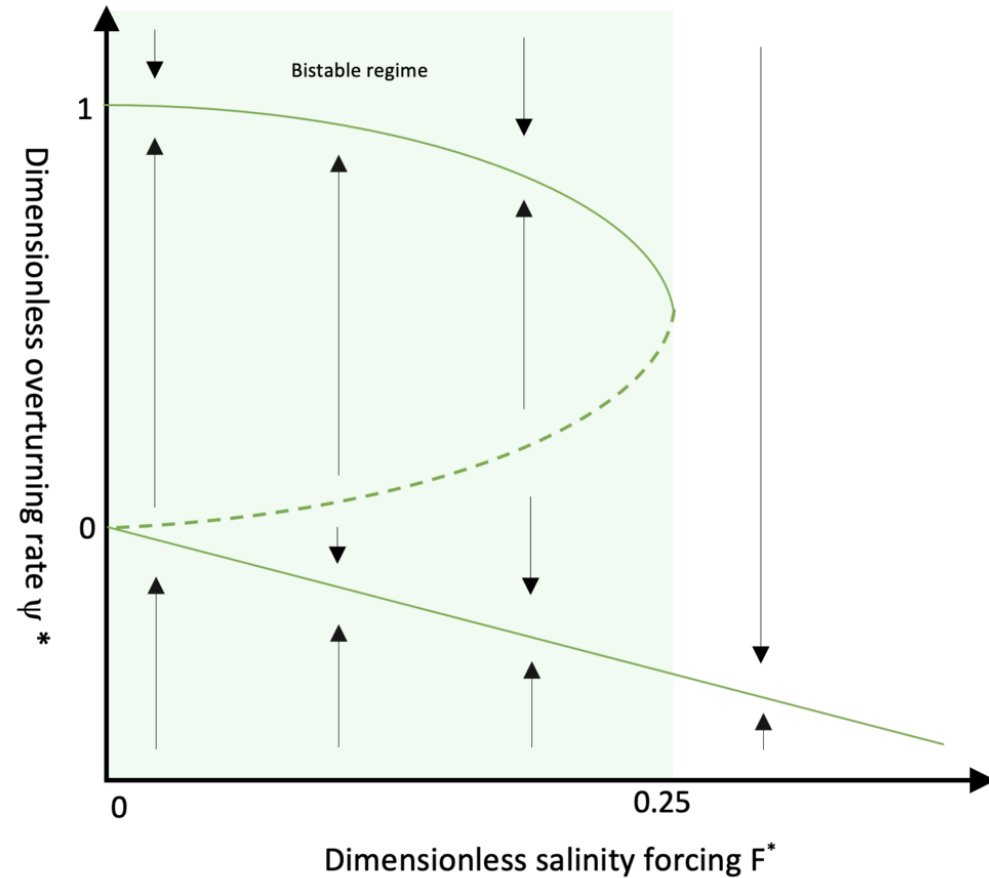
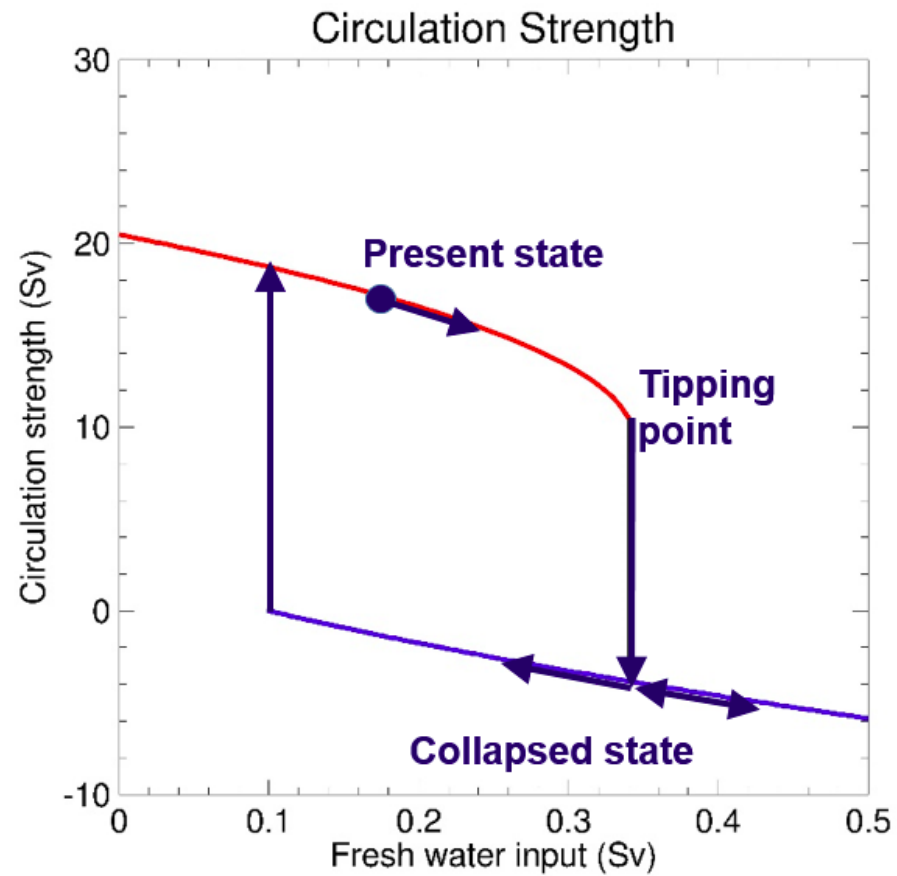
# Example 2: Atlantic Meridional Overturning Circulation

- Stommel (1961) two-box model



[Wikipedia article on Multiple equilibria in the Atlantic meridional overturning circulation](#)

# Bifurcation diagrams



Met office

<https://www.metoffice.gov.uk/binaries/content/assets/metofficegovuk/pdf/weather/learn-about/climate/deliverables/thresholds-and-feedbacks-introduction.pdf>

[Wikipedia article on Multiple equilibria in the Atlantic meridional overturning circulation](#)

# Summary

- **Design concept:** Emergence
- **Model analysis:** Properties of dynamical systems
  - Effect of changing parameter
  - Dynamical states
  - Bifurcations
  - Phase plots
  - Application to ecological-economic model and earth model
- Some quantification of patterns and exploration of parameter space
- **Next:**
  - Tutorial (Thursday): Paper mini-presentations
  - Lab (Thursday): Butterfly hilltopping model