

# Viability and resilience

Guillaume Deffuant, Sophie Martin, Justin Calabrese





## Outline

- **Resilience based on attractors**
- **Resilience based on viability**
  - Without management actions and attractors in desired set of states
  - With actions and attractors in desired set of states
  - With actions and attractors not in desired set of states
- **Conclusion**

## Resilience of based on attractors

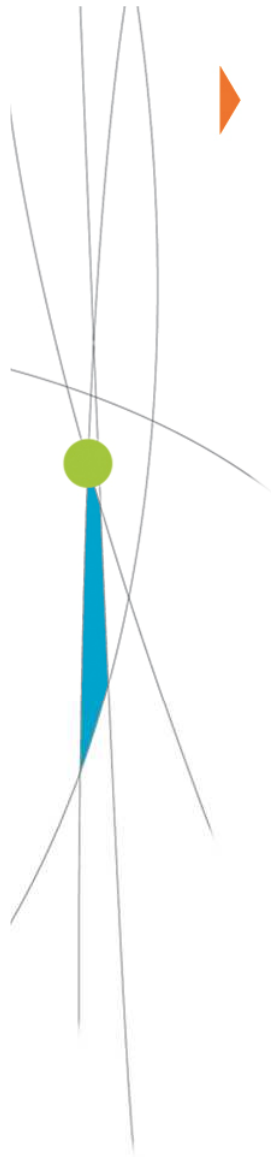
- **Hypothesis:**
  - some attractors provide desired properties of the system (good attractors)
  - some attractors don't (bad attractors).
- **The system is resilient to a perturbation if the perturbation keeps the system in the attraction basin of a « good » attractor, it is not resilient if the perturbation drives the system to the attractor basin of a « bad » attractor.**
- **Resilience value connected with the size of the « good » attractor basins.**

## Simplified example of savanna dynamics

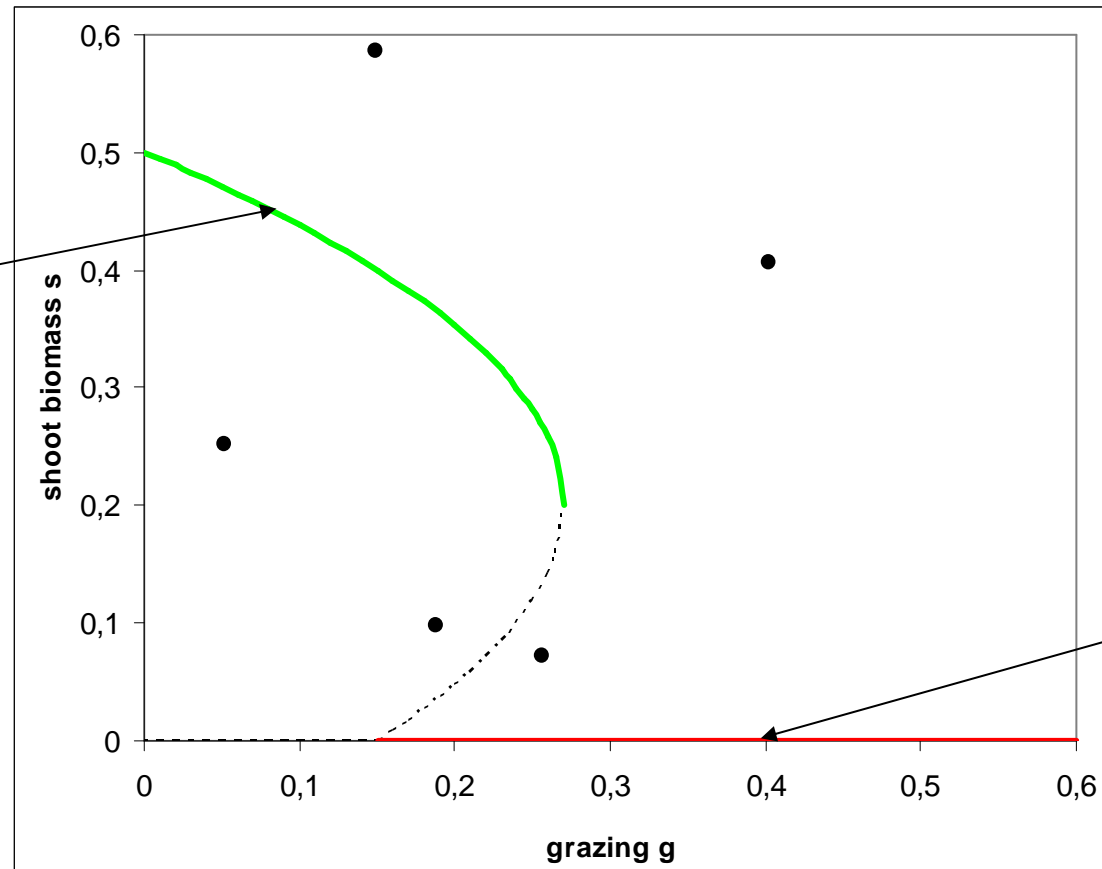
- Taken from Anderies et al. 2003
- Two variables:
  - shoot biomass (grass)  $s$
  - grazing  $g$
- We suppose that once the value of grazing  $g$  is decided, it remains fixed.

$$\frac{ds}{dt} = (\alpha s + \beta s^2)(1 - s - \gamma) - gs$$

# Good and bad attractors



good  
attractors

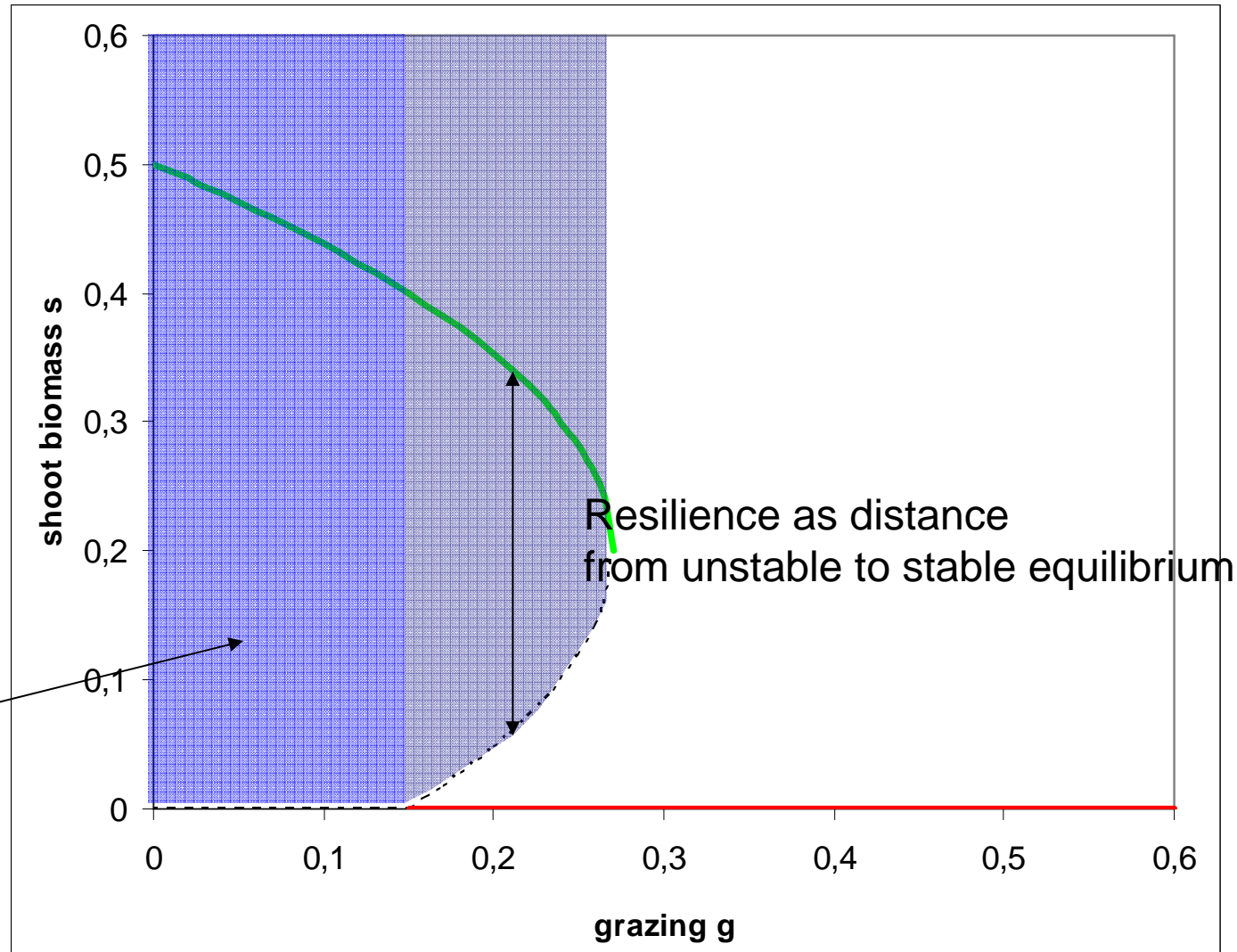
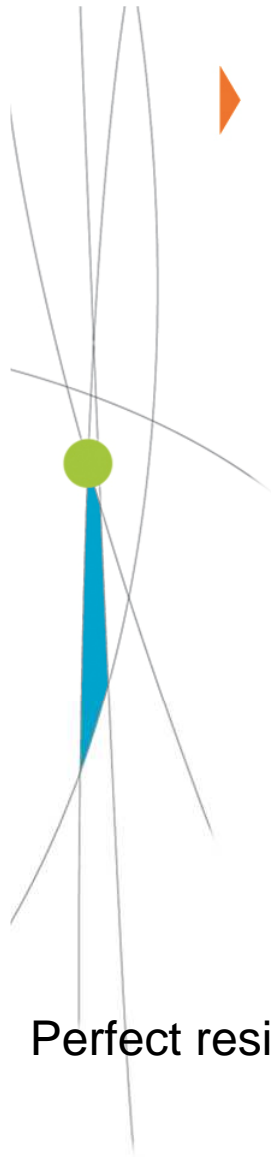


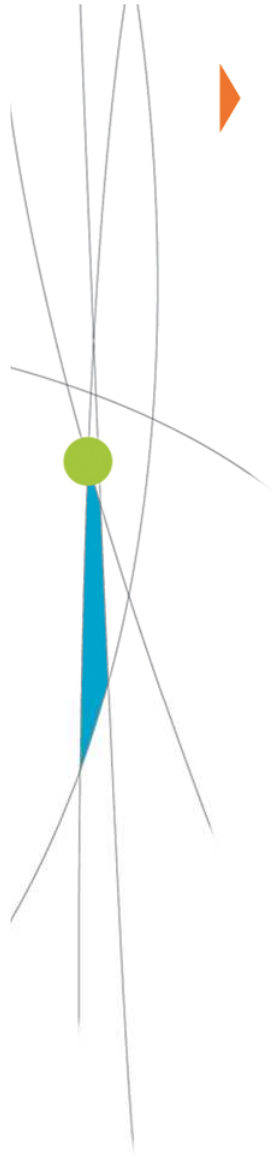
bad  
attractors  
(no grass)

## Resilience based on attractors

- Dynamical system defined for instance in discrete time:
- $\mathbf{x}(t+dt) = \mathbf{x}(t) + \Phi(\mathbf{x}(t))dt$
- A point  $\mathbf{x}_a$  is an attractor if:
  - $\Phi(\mathbf{x}_a) = 0$
  - There exists a subset  $A$  such that, for all  $\mathbf{x}(0)$  in  $A$ :
  - $\mathbf{x}(t) \rightarrow \mathbf{x}_a$  when  $t \rightarrow \infty$
  - The largest set  $A$  is the attraction basin
- **Resilience defined as**
  - the size of the attractor basin
  - the velocity for going to the attractor (based on a linearisation of the dynamics close to the equilibrium)

# Definition of resilience as a function of grazing

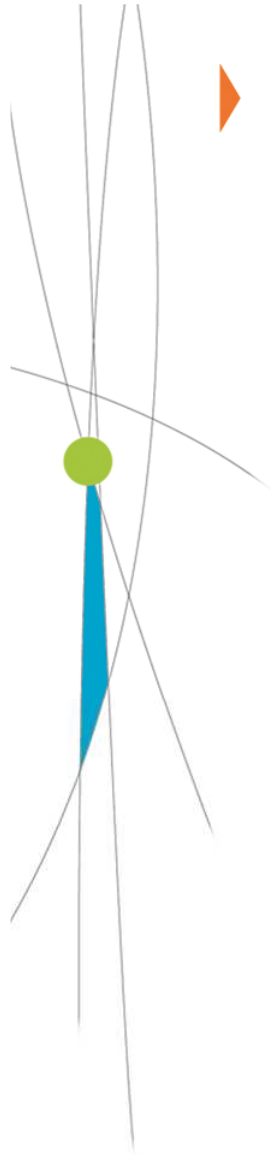




## Limits of the attractor based resilience

- **Need to define the desired functioning of the system as a set of attractors.**
- **Difficult to introduce a management policy (it is supposed included in the system dynamics)**

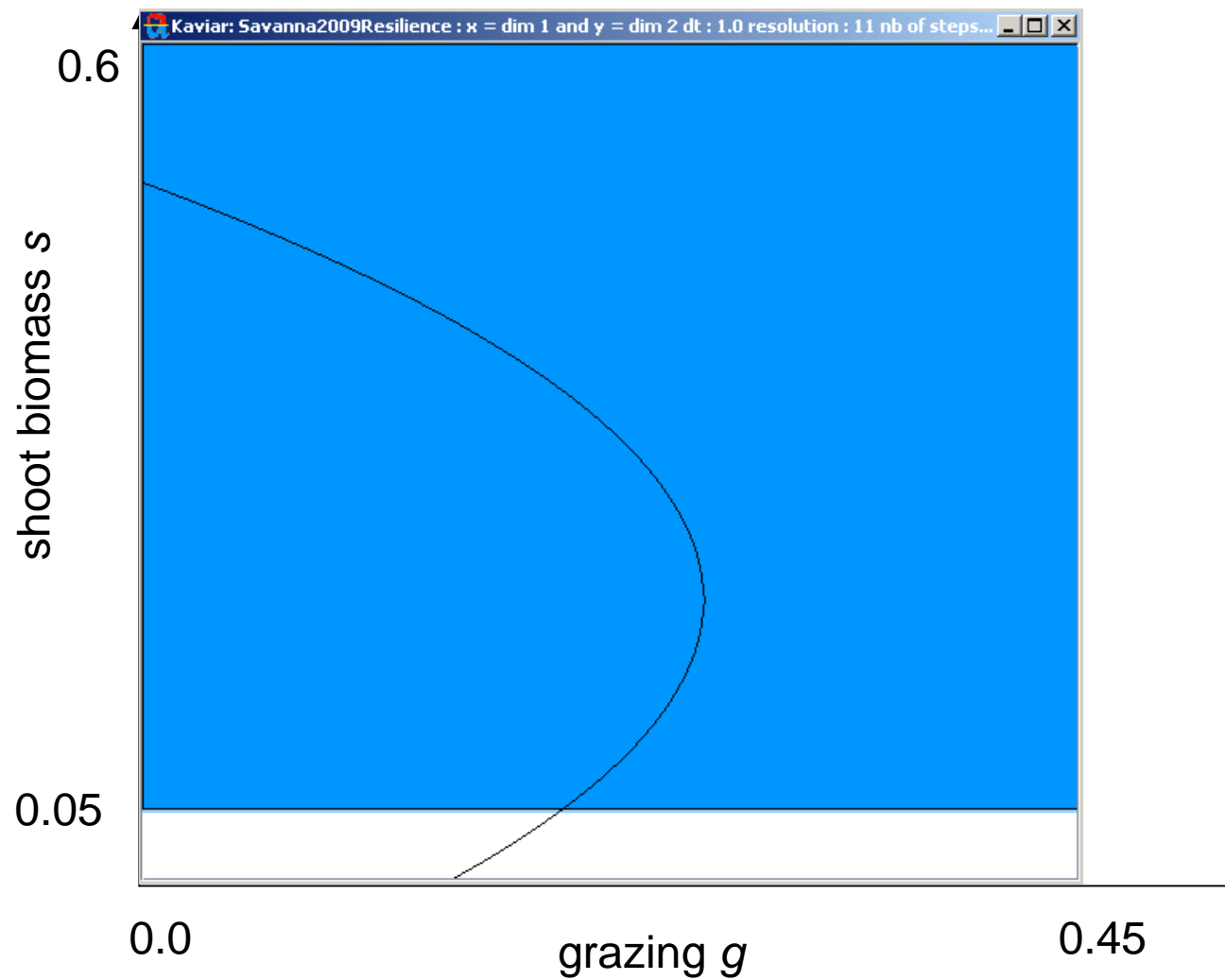
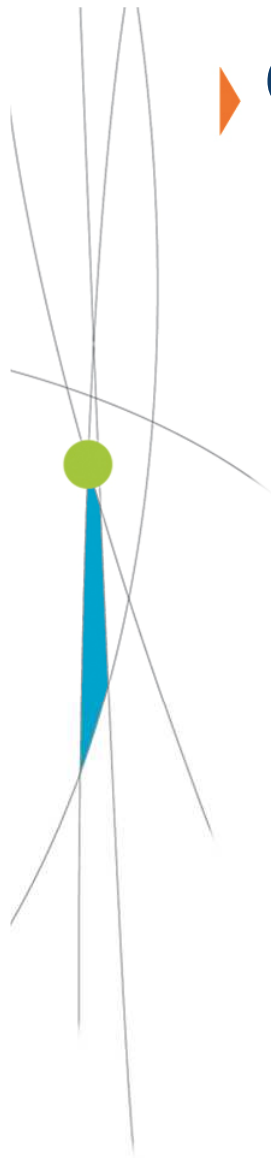


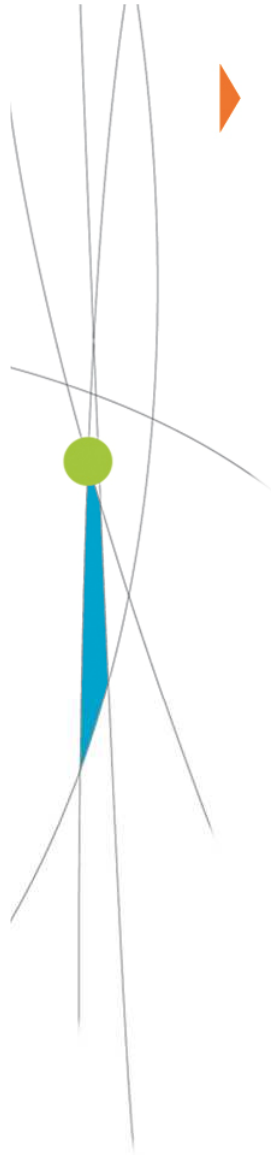


## Different view: Desired property as state set

- Resilience supposes to define a desired property (functioning) of the system
- Main idea: define the desired property as a subset of the state space (independently from the presence of attractors)

▶ **Constraint set:  $s > 0;05$**

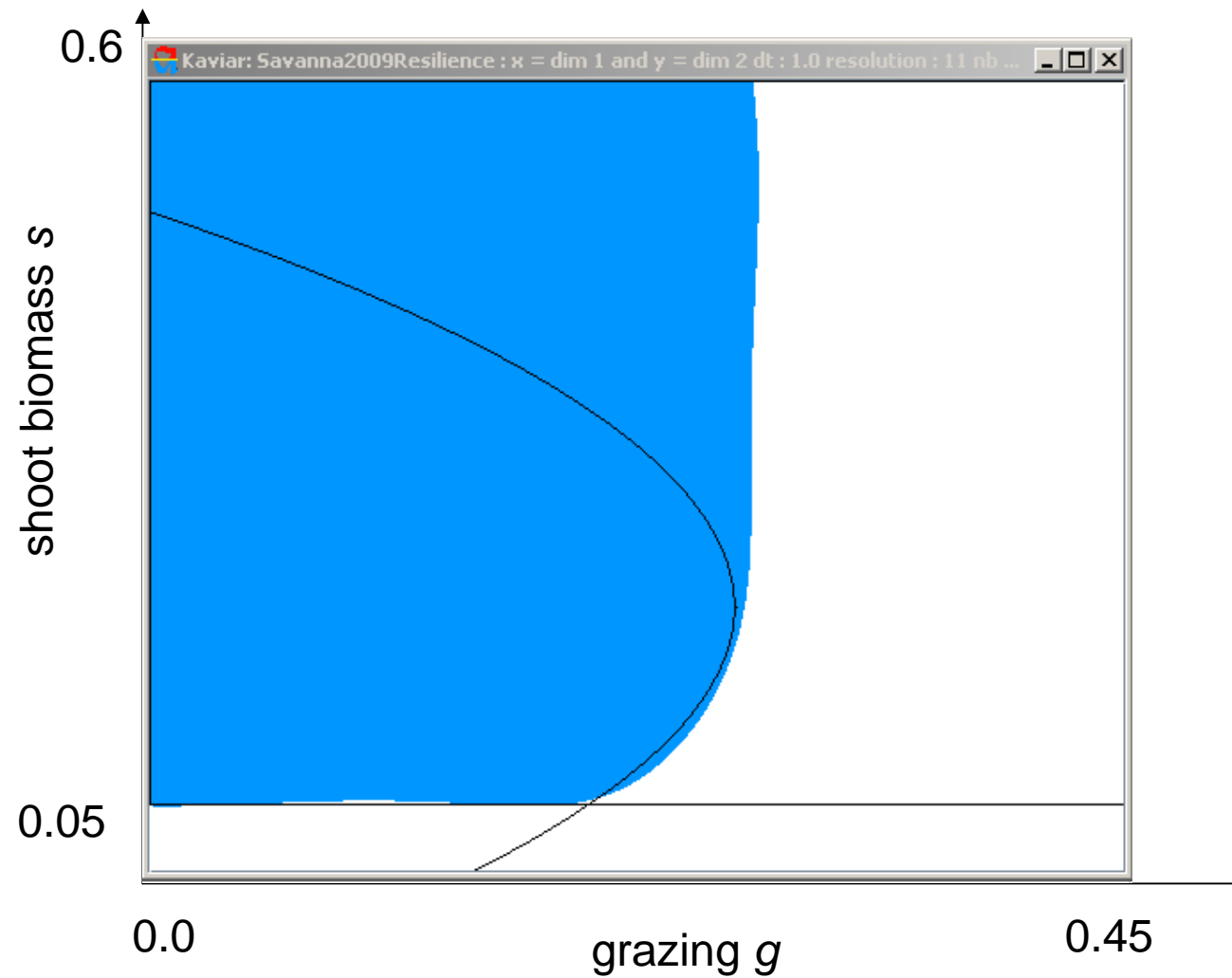
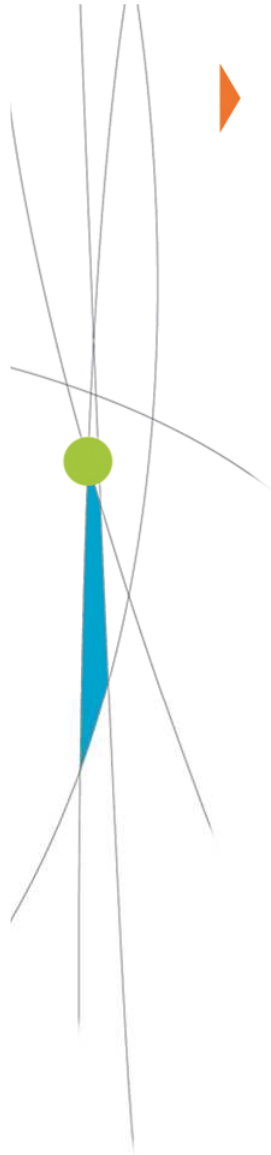




## Link with viability

- We need to determine all points of space from which the desired property is maintained – i.e. which trajectory stays in the desired set
- Viability theory, developed in the 90ies by J.P. Aubin:
  - considers a system that collapses or badly deteriorates if it goes beyond a state subset  $K$ .
  - It needs also to determine the points from which the trajectory remains in  $K$ , that is  $x(0)$  such that, for all value  $t$ , we have  $x(t)$  in  $K$ , is called the **viability kernel**.

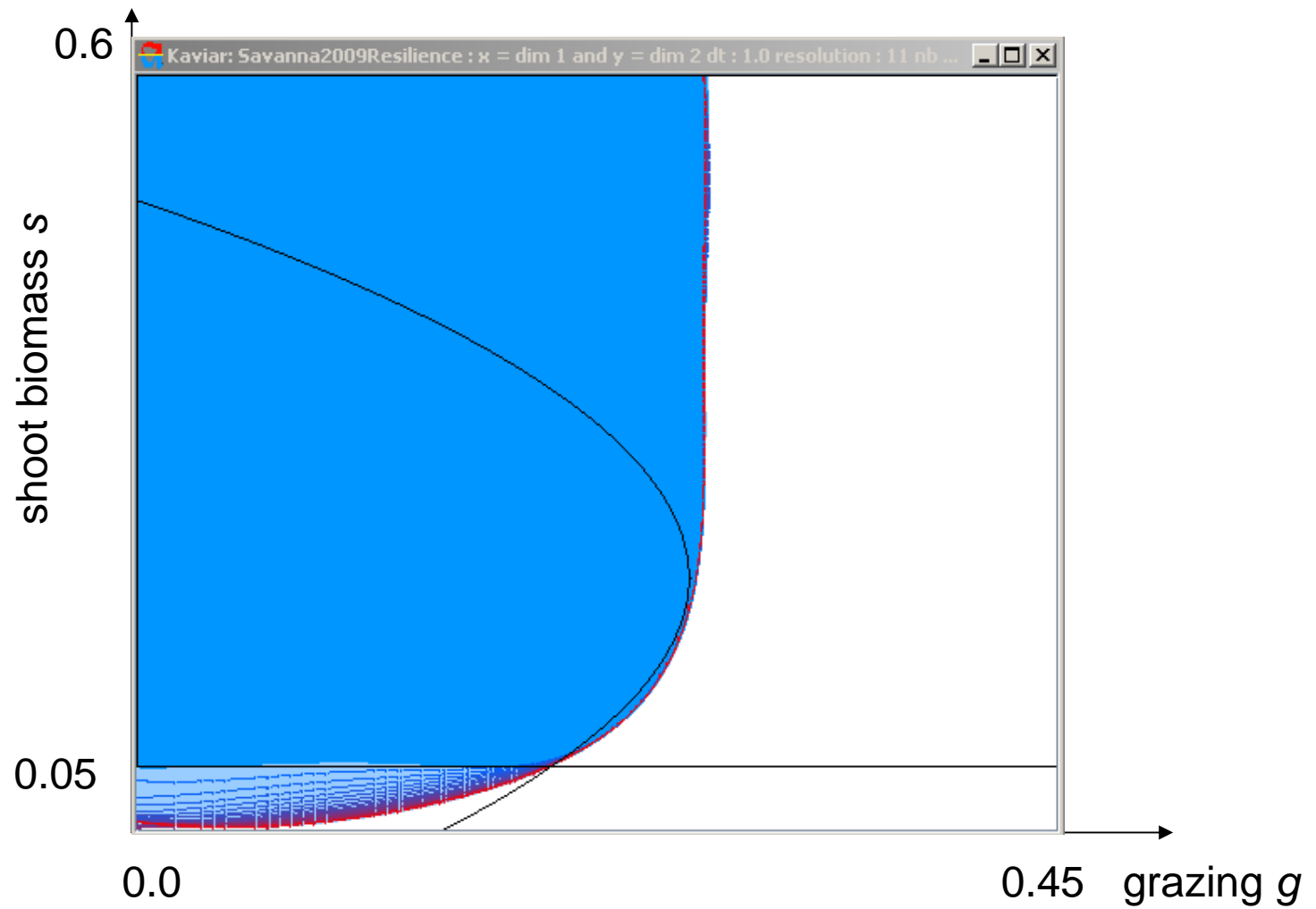
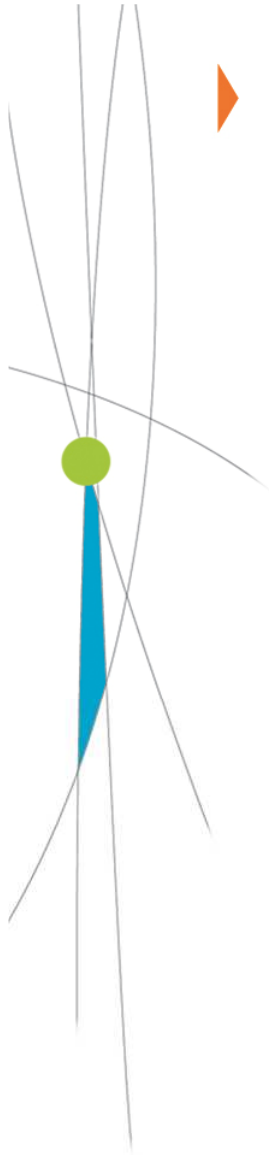
# Viability kernel

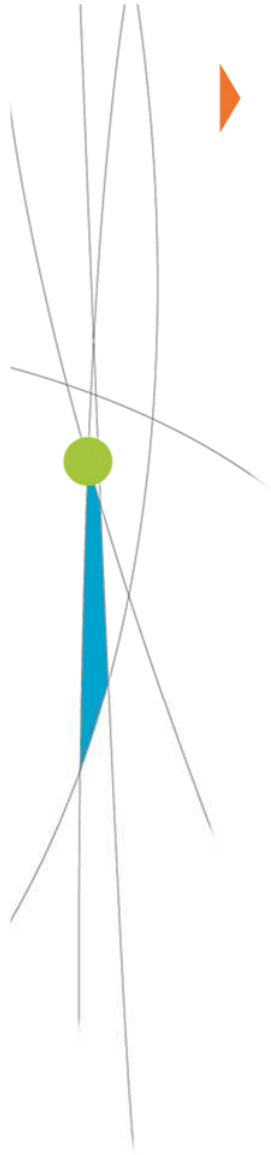


## Resilience based on viability (without action)

- We also need to determine the points from which trajectories come back to the desired set and stay there.
- A resilient state is a state from which the trajectory goes to the viability kernel (because it guarantees it will stay there)
- In viability theory, the set of points from which trajectories go to a target set is the **capture basin** of this target set.
- The **set of resilient states** is the **capture basin of the viability kernel**
- The measure of resilience is the inverse of the integral of a cost per unit of time along the trajectory to the viability kernel.

# Resilience

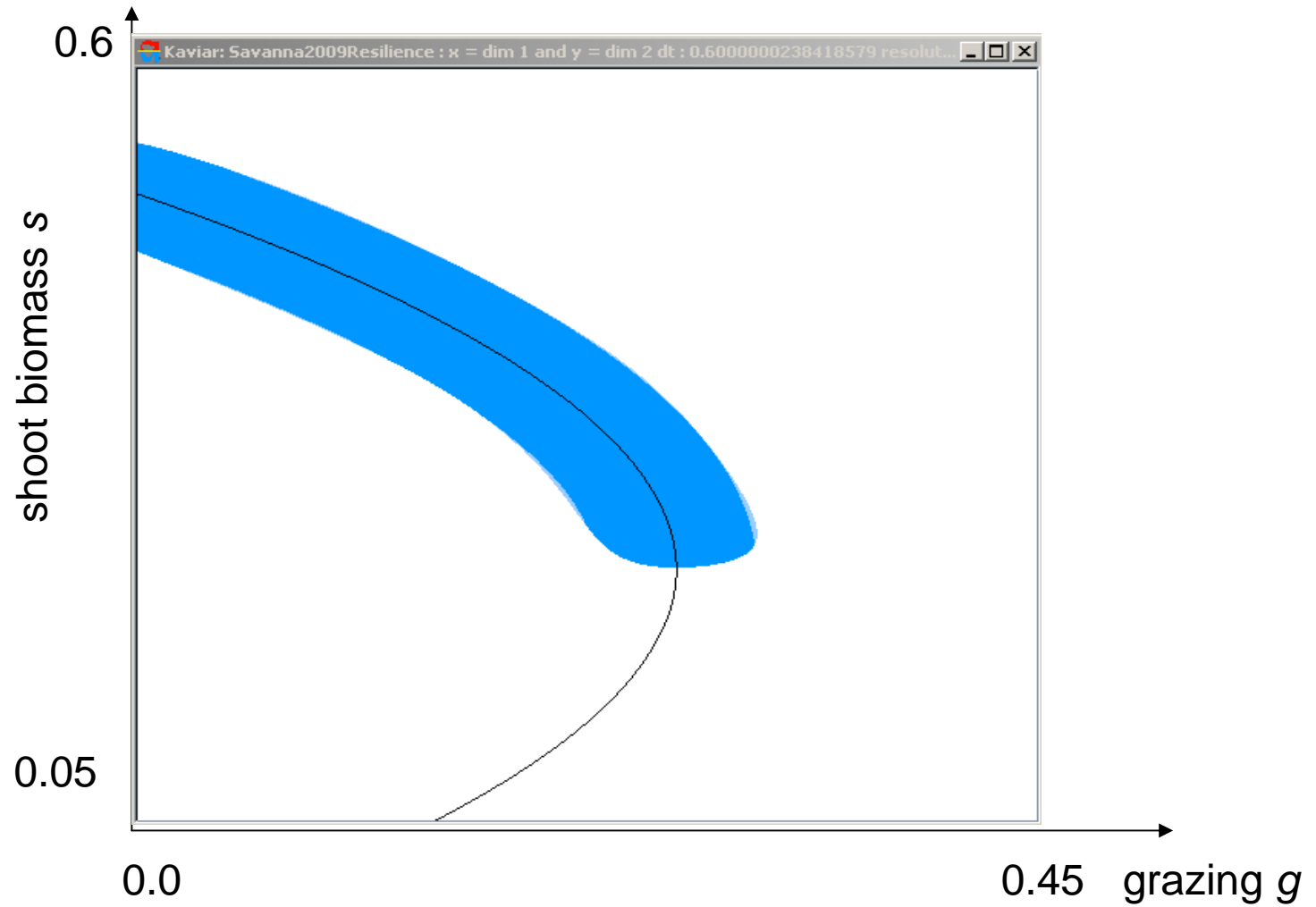
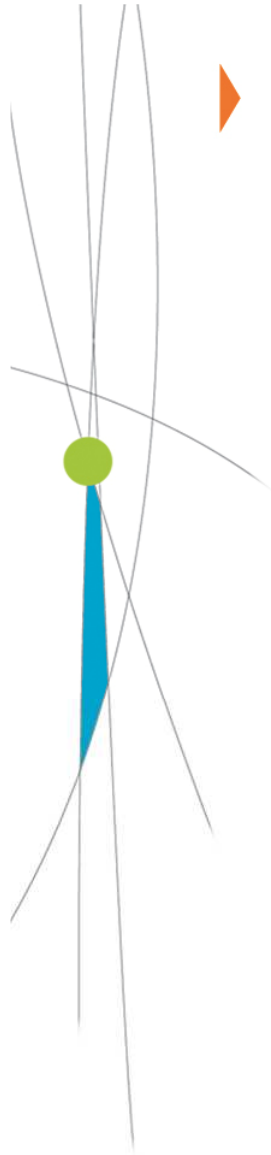




## Comparing the definitions

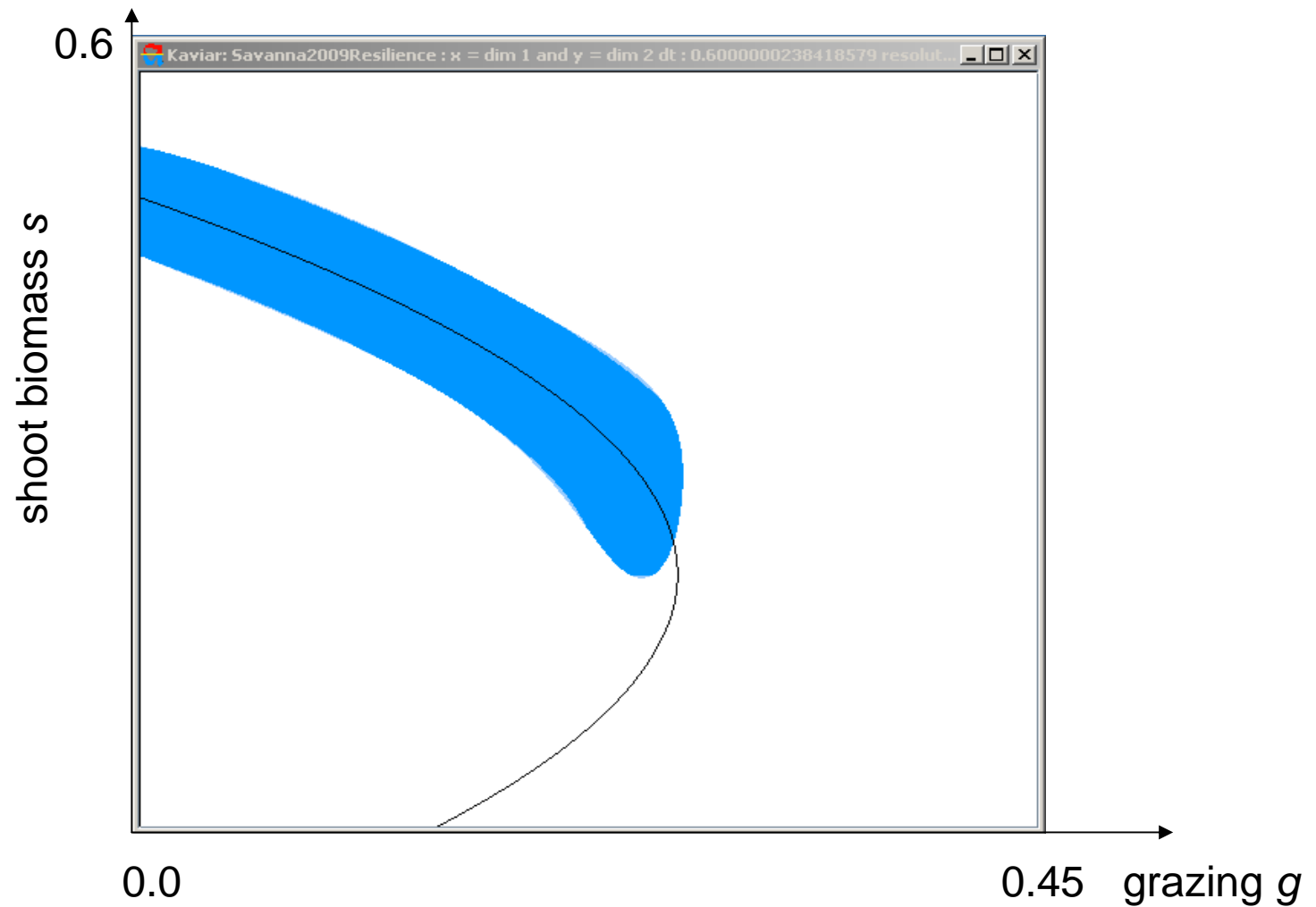
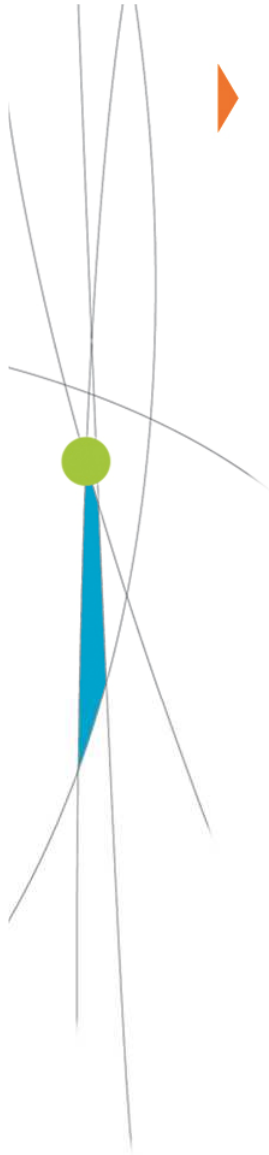
- Most states which are *resilient* in the attractor definition are *viable* in the viability definition
- The measure of resilience is different (depends on the dynamics)
- Difference due to the choice of the constraint set ?

# New constraint set

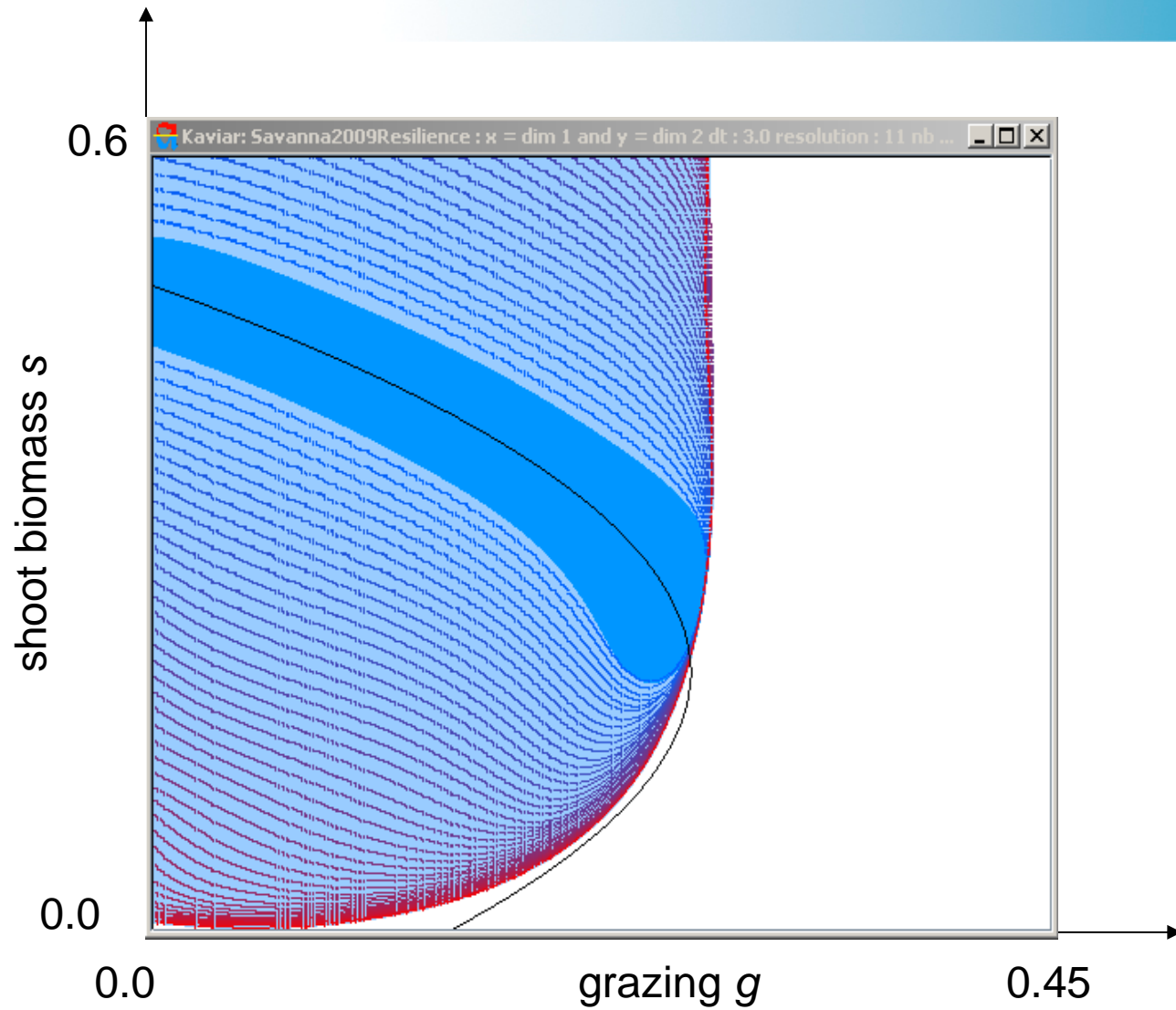
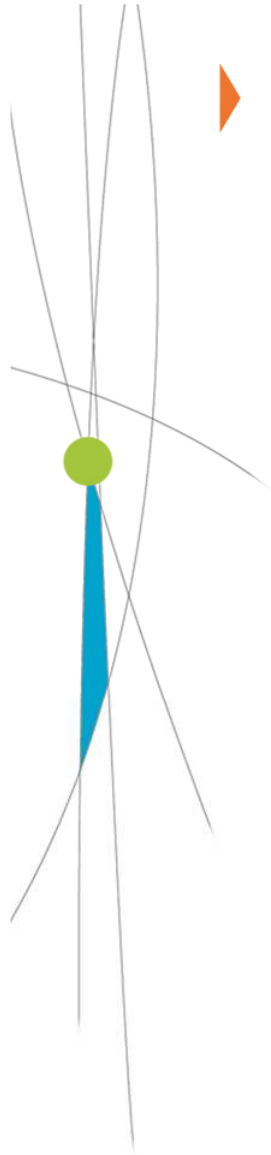




## Viability kernel 2



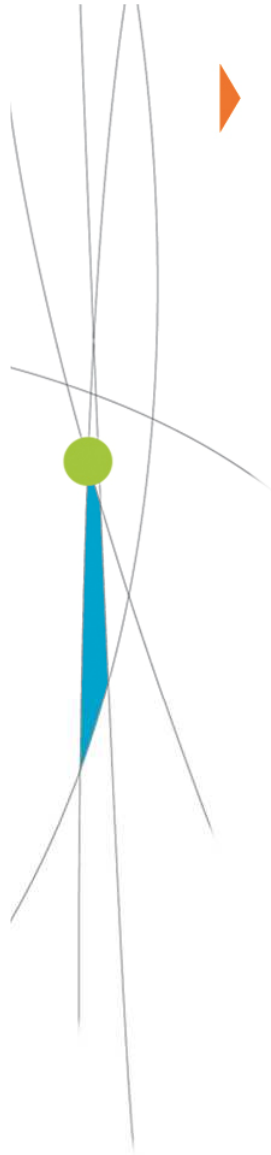
## Resilience 2





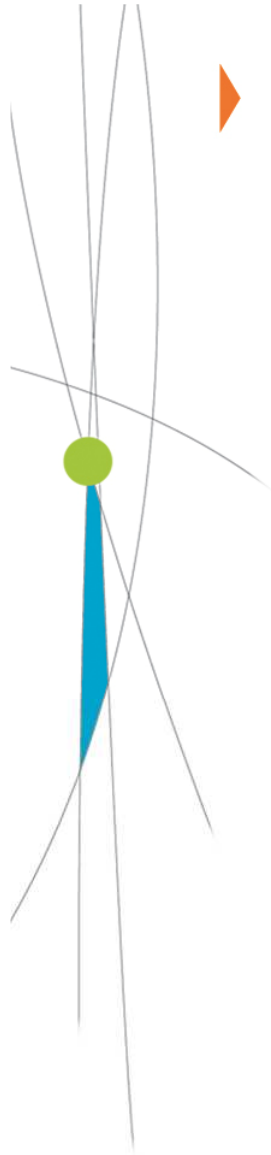
## Comparing the definitions

- **The resilient states almost coincide.**
- **The resilience values are close but in some places depend more on the dynamics in the resilience viability definition.**



## Introducing a possibility to act on the system

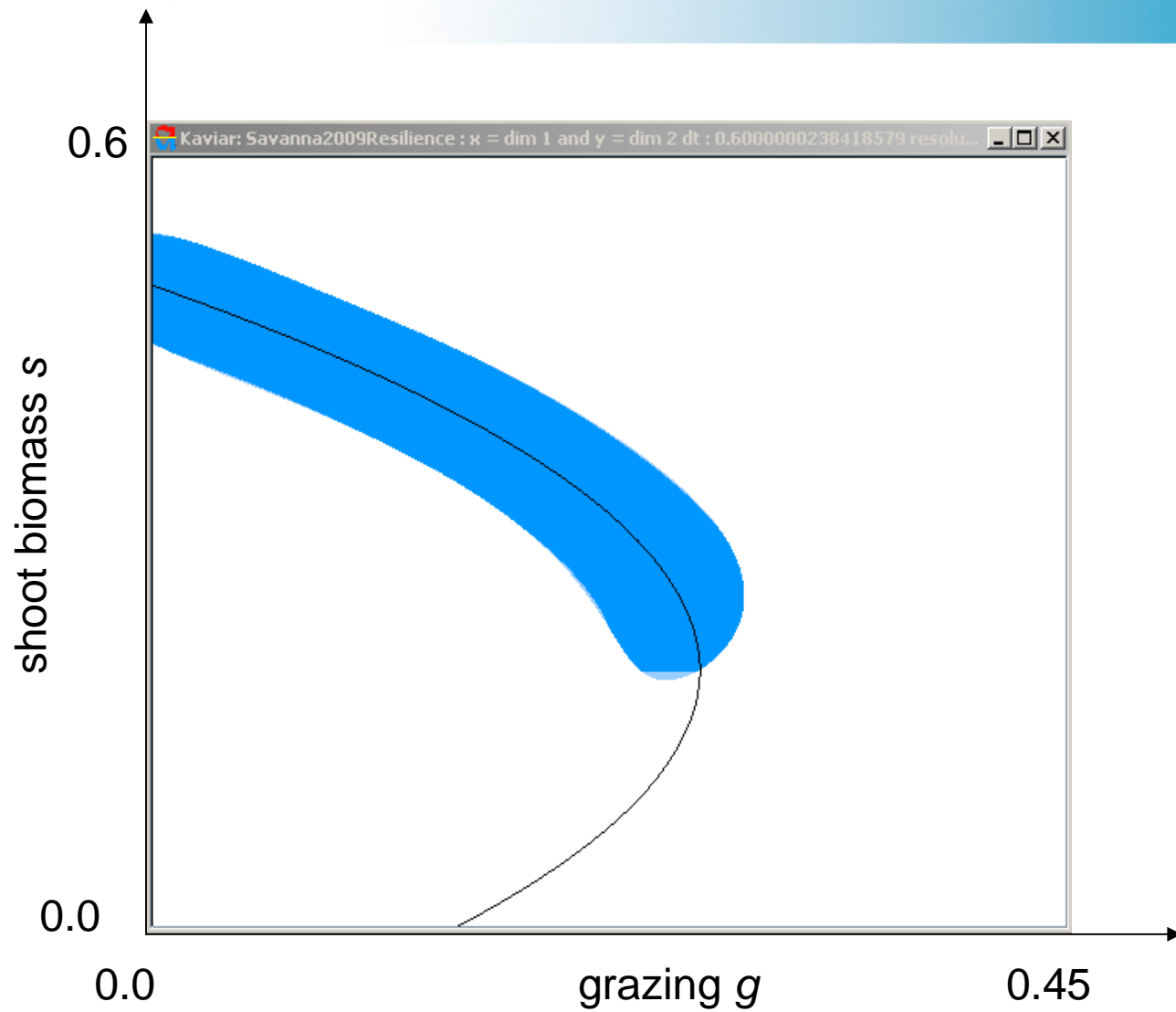
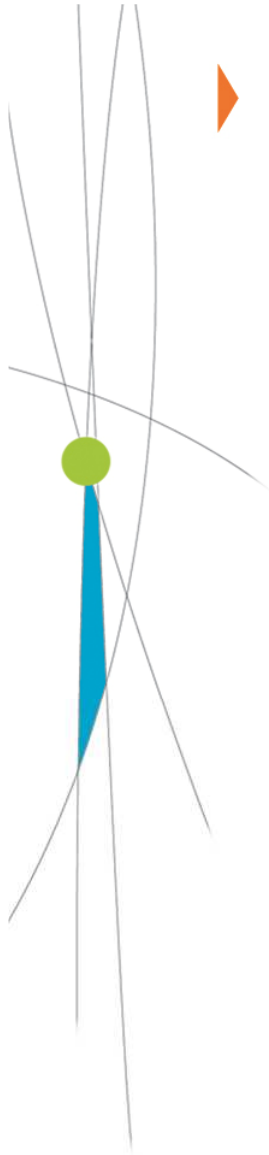
- At each time step, we suppose that it is possible to act on the system.
- For instance, we suppose that the grazing pressure can be modified of a value  $dg$ , with  $-0.02 < dg < 0.02$



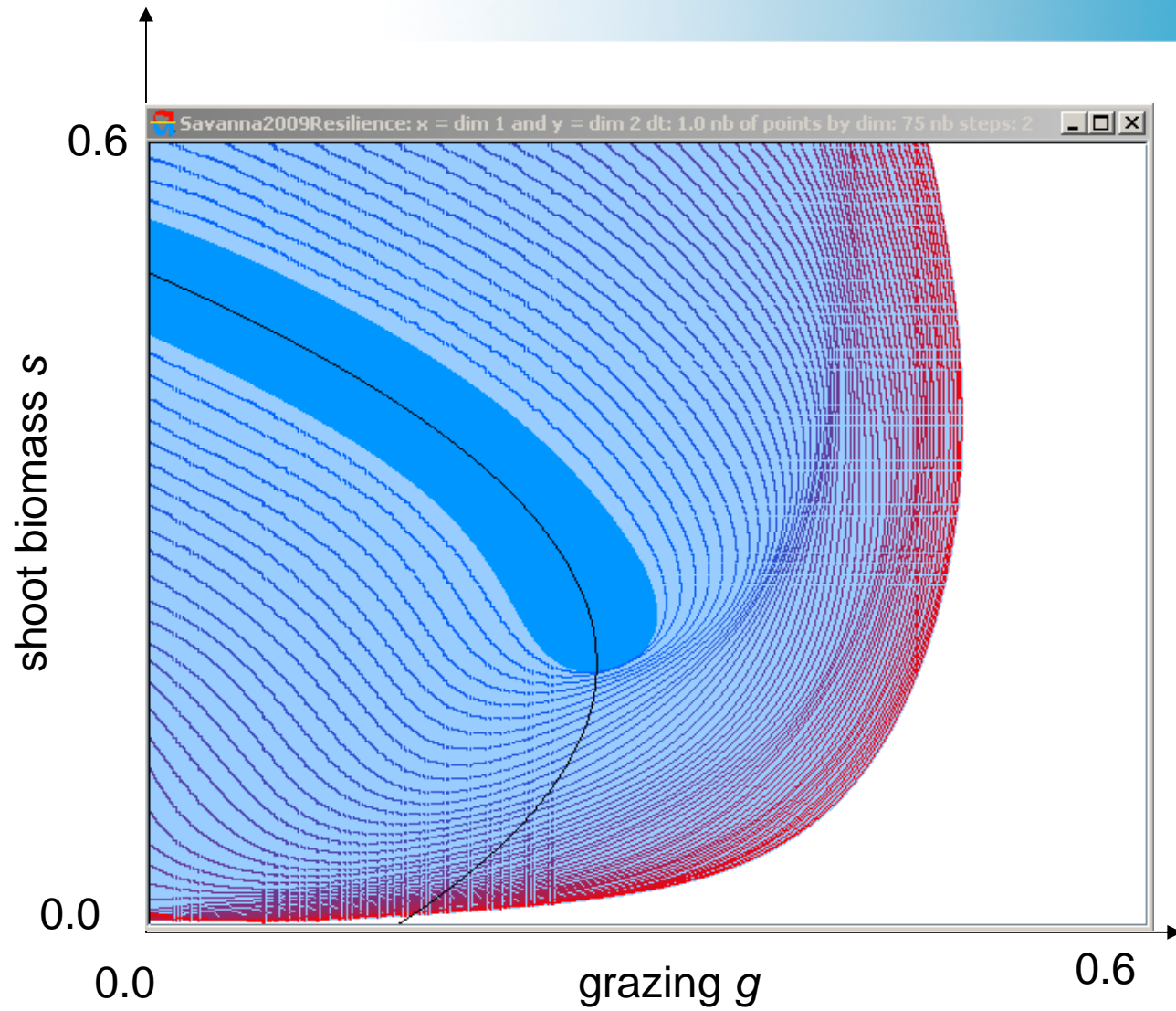
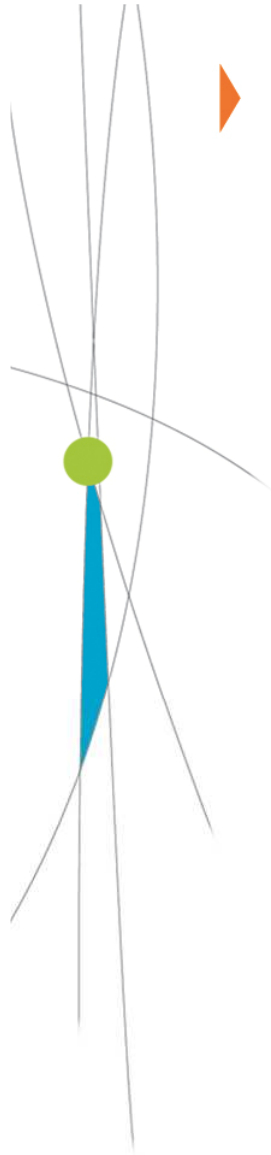
## Viability theory with action

- **Viability kernel** : the set of states from which a policy of action keeps the system within the constraint set.
- **Capture basin of a target set**: set of states from which there exists a policy of action leading to the target.
- **Resilient states**: states belonging to the capture basin of the viability kernel.
- **Resilience value**: inverse of the time to go back to the viability kernel.
- **There exist general algorithms to compute viability kernels and capture basins**

## Viability kernel 3

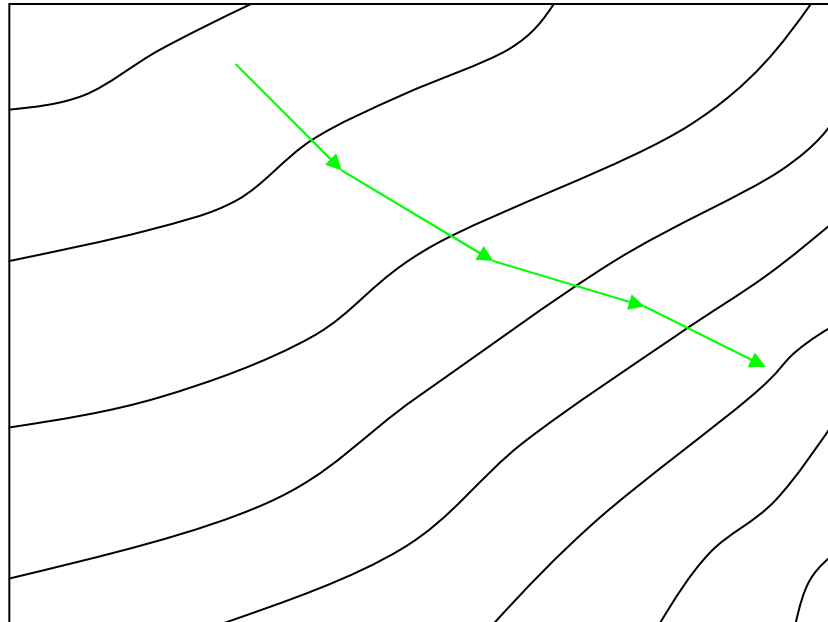


# Resilience 3



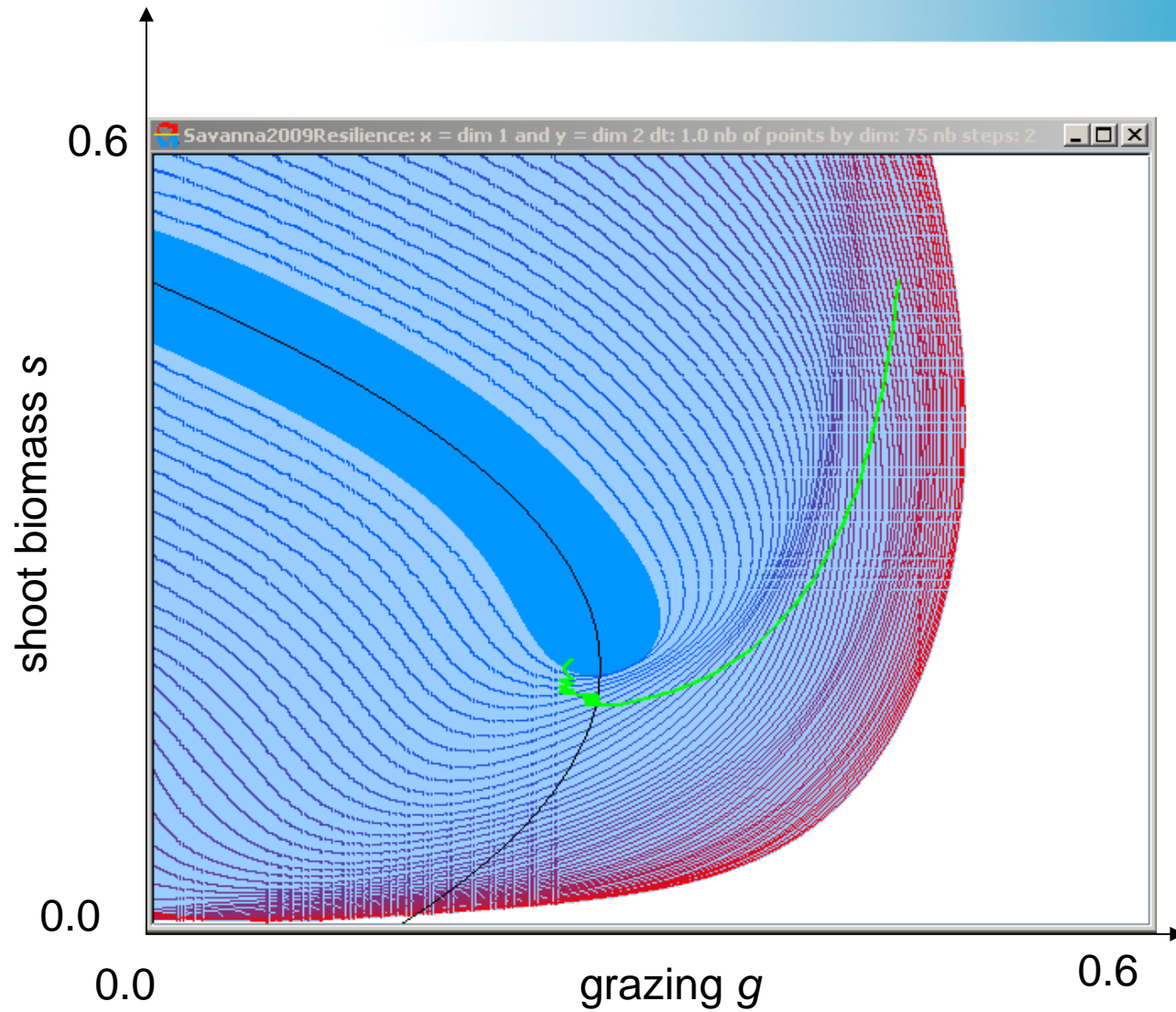
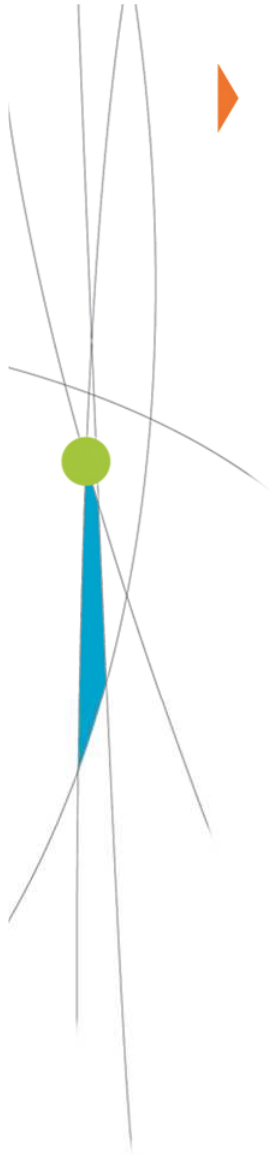
## Defining actions

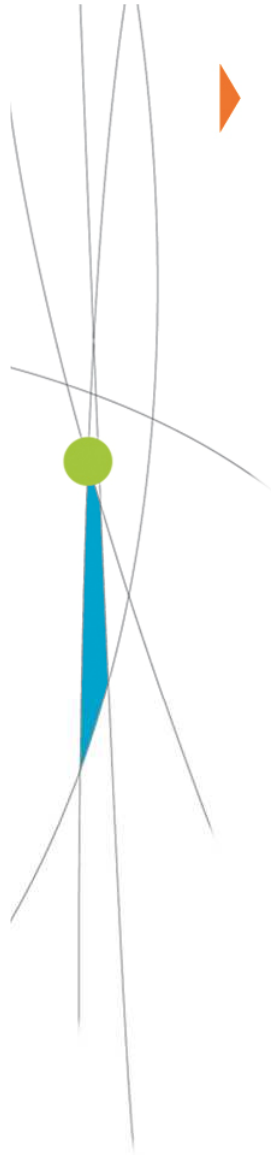
The action is the one that drives the system as close as possible in the normal direction of the next level line.





# Resilient trajectory

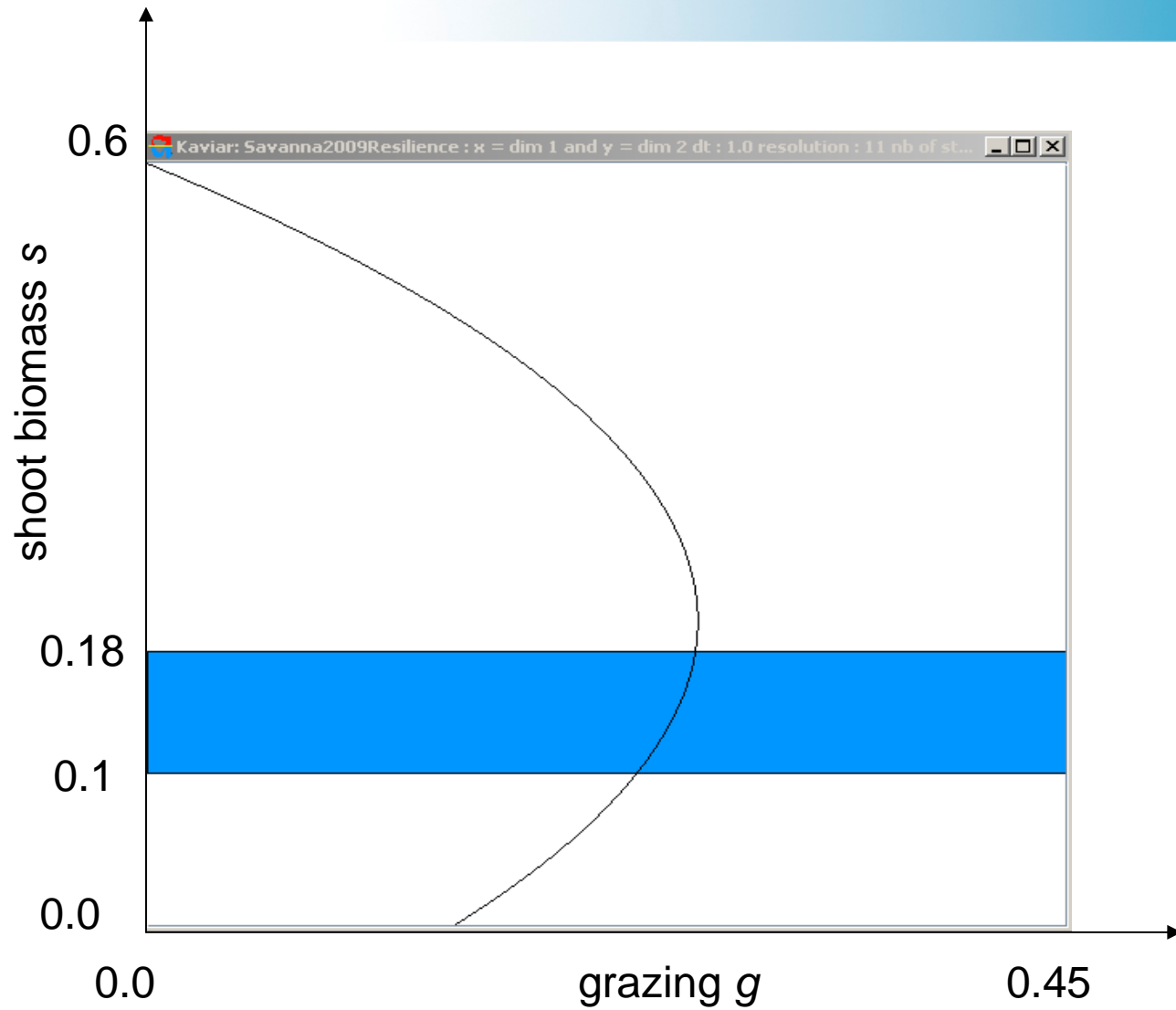
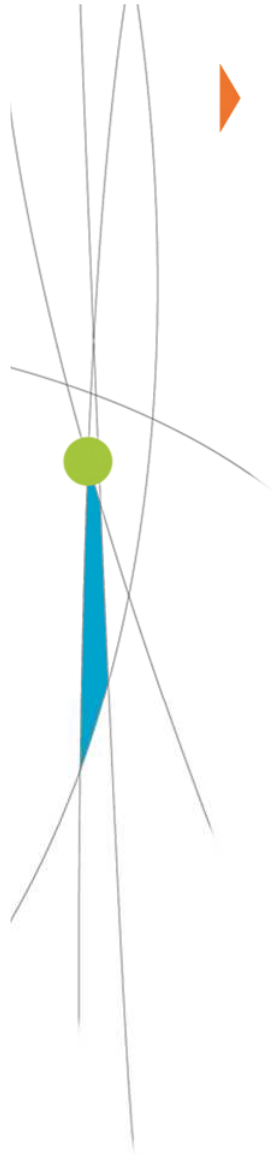




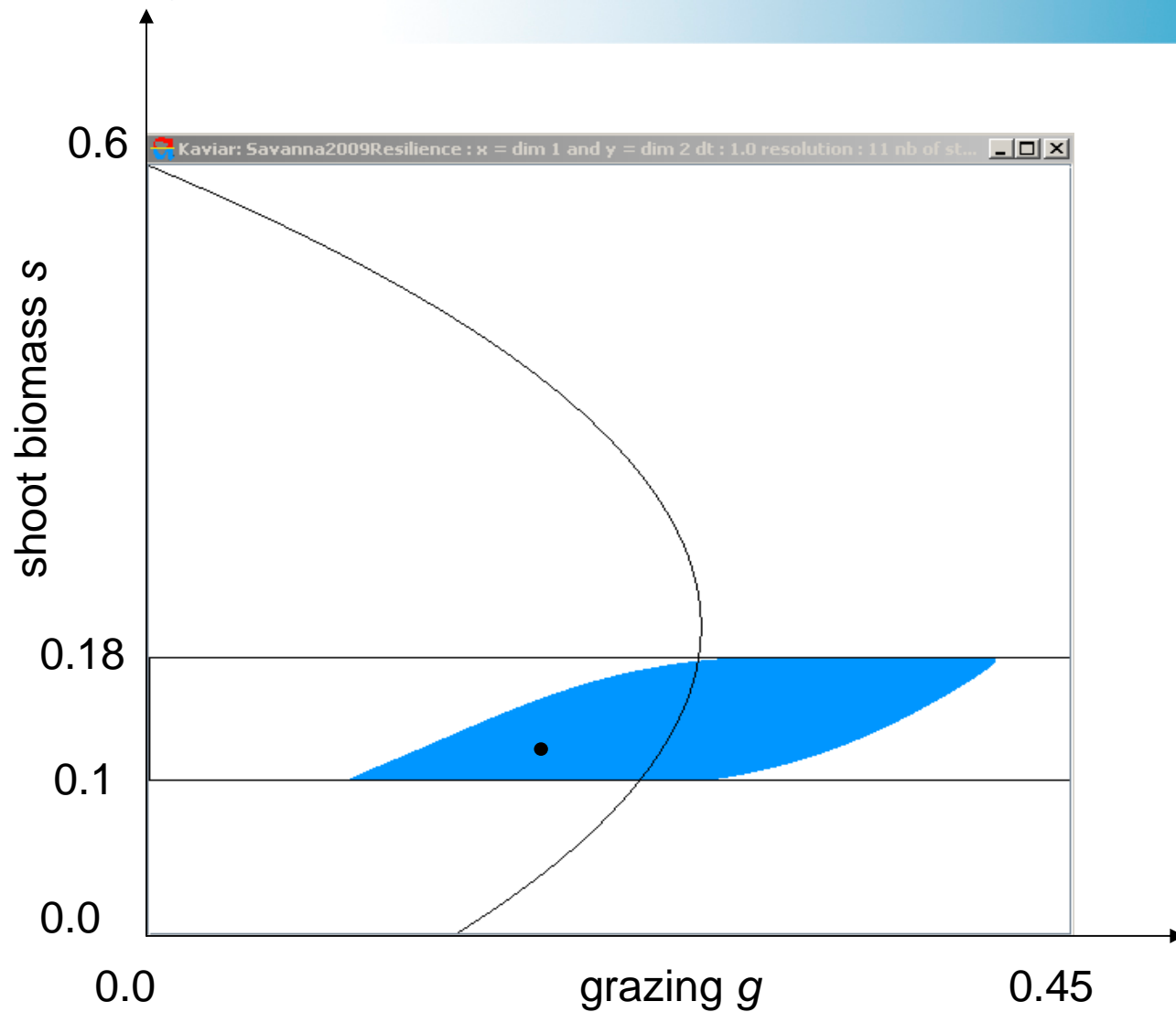
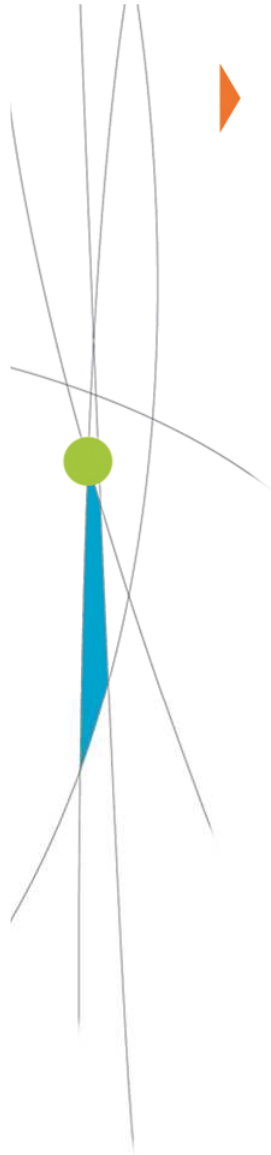
## No attractor in the constraint set

- The dynamics do not necessary lead to an attractor.
- Suppose now that we want to keep the level of grass between 0.05 and 0.18
- We still can change the grazing of at most 0.02 (positive or negative)

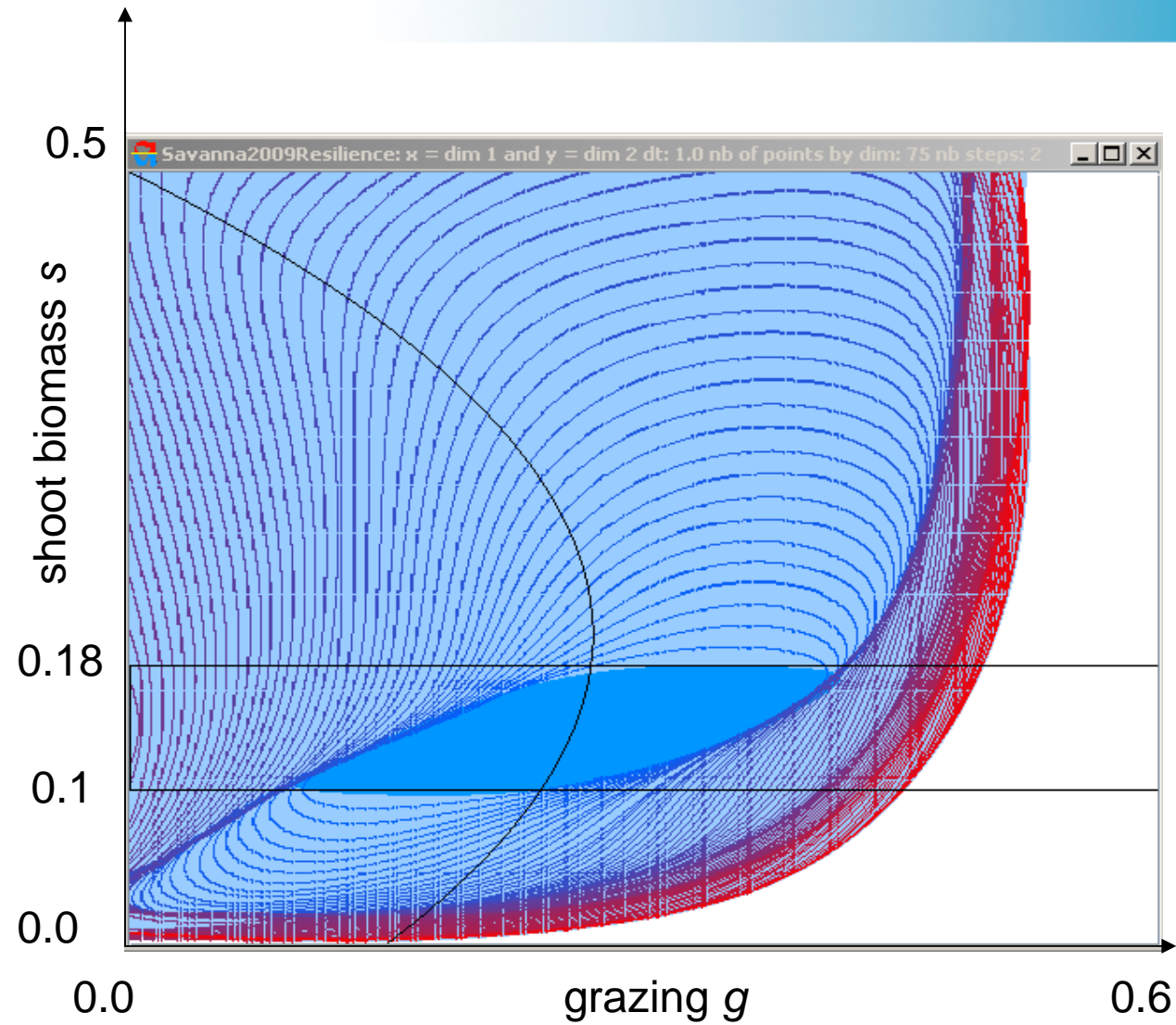
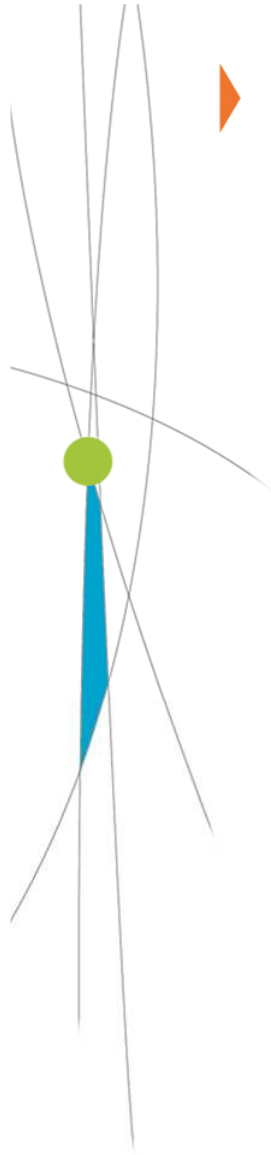
# Constraint set : $0.1 < s < 0.18$



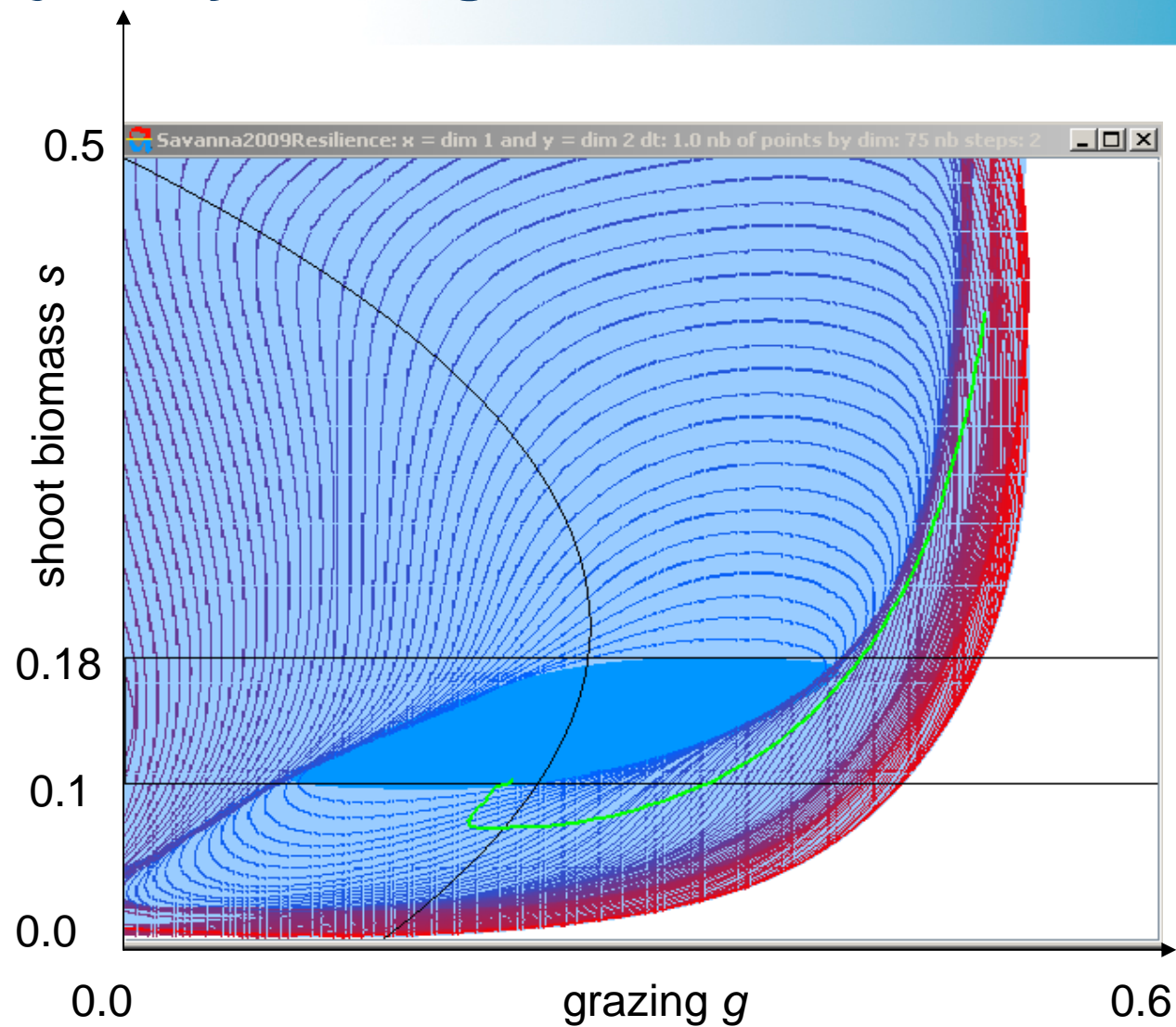
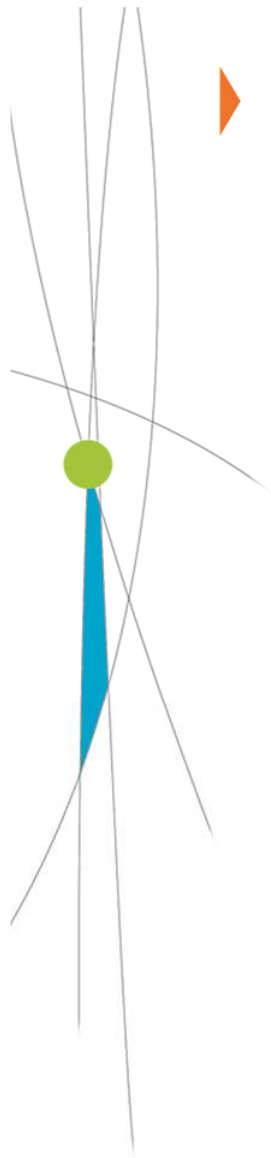
# Viability kernel, with no stable equilibrium



# Resilience

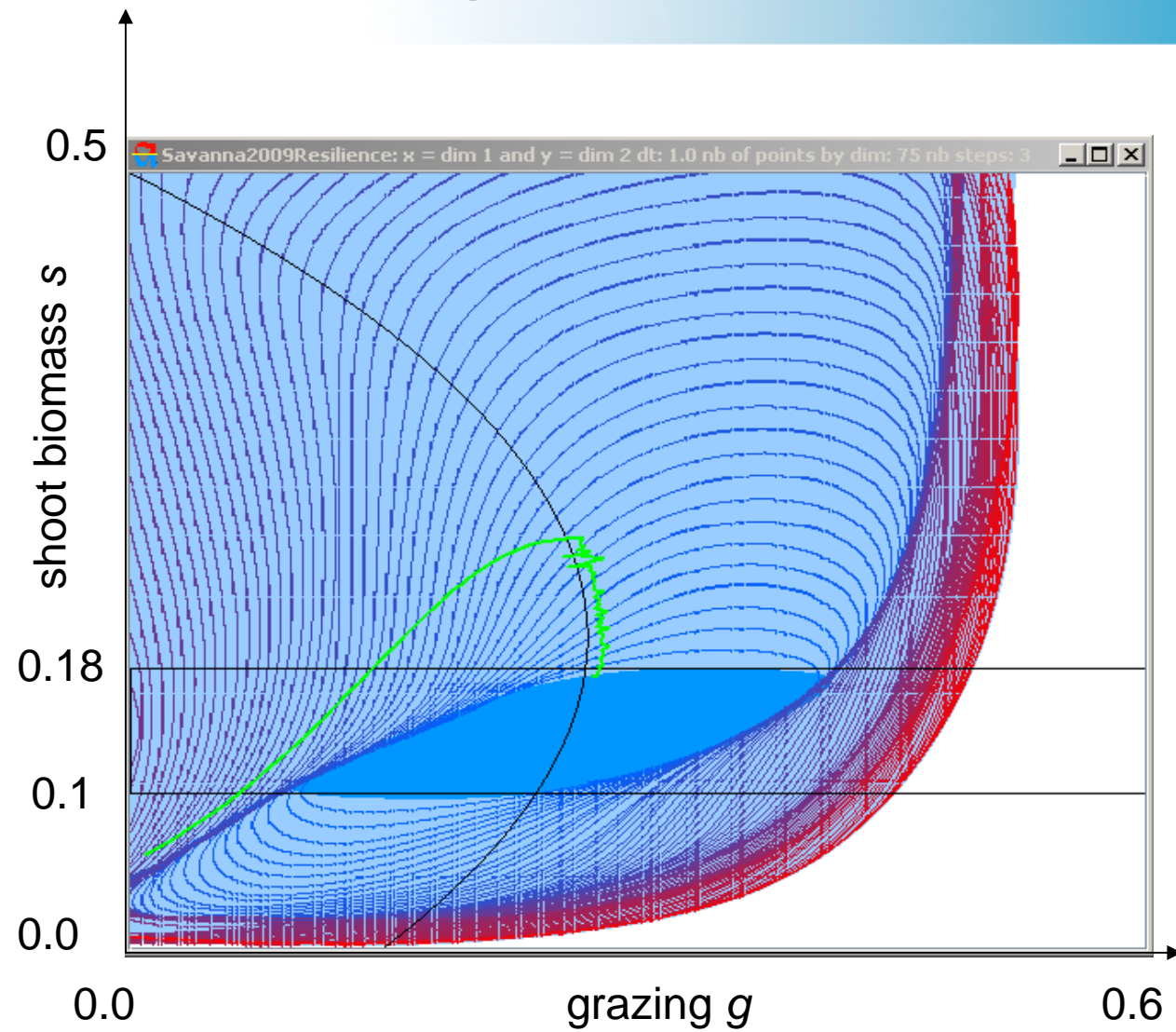
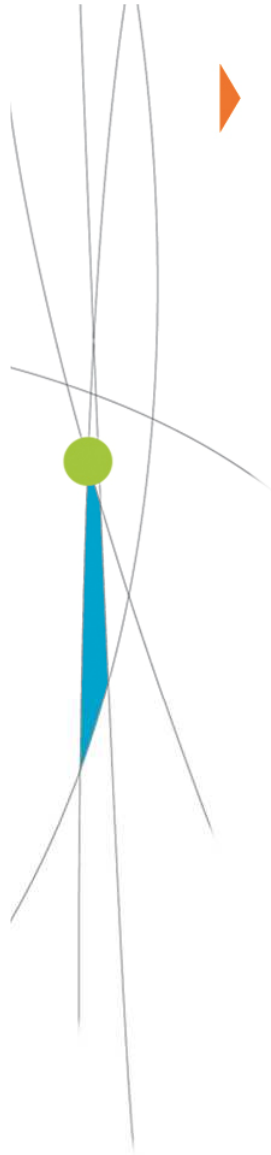


## Trajectory coming back to kernel





# Trajectory coming back to kernel

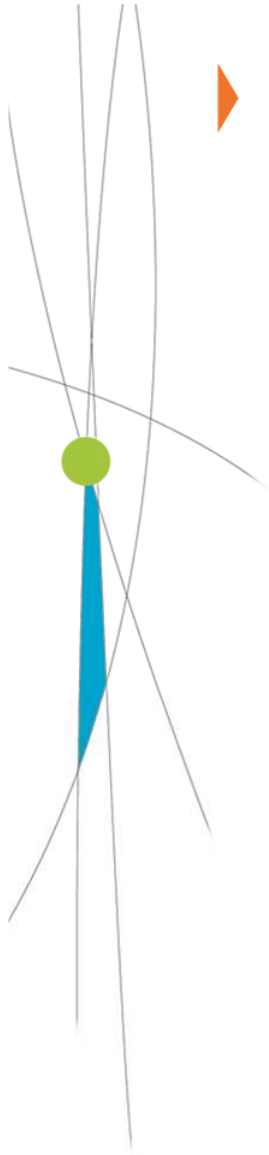




## Conclusion

- **The resilience based on viability can be seen as an extension of resilience based on attractors**
  - The « good » attractor is replaced by the viability kernel defined on the constraint set of the desired property
  - The attraction basin is replaced by the « capture basin » of the viability kernel (i.e. the points for which there exists a policy of action leading to the viability kernel)
- **Advantages**
  - Can include naturally an action in the approach and provides actions to make
  - Does not necessitate equilibrium in the dynamics.





## Problem

- To compute viability kernels and resilience values, one must discretise the space
- When the dimension of the space grows, the number of points of the grid grows exponentially.
- The method cannot be applied on dynamical system with a state of many dimensions.