

# Estabilización de fluctuaciones en dinámica de poblaciones

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**DDays, Badajoz, 14-11-2014**

# Why the size of certain populations fluctuate?

Intrinsic to the species. Science (1974), Nature (1976).

# review article

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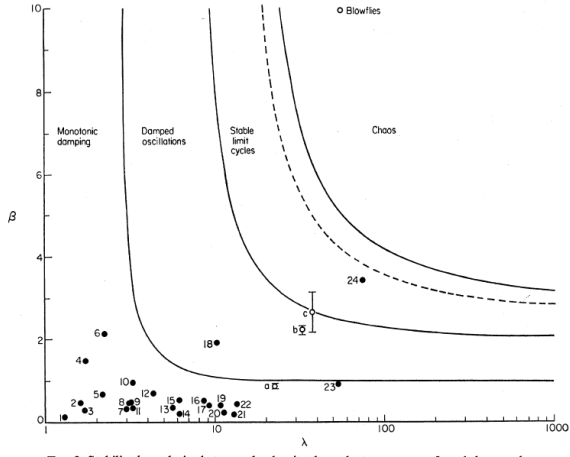
## Simple mathematical models with very complicated dynamics

Robert M. May\*

Is there chaos in **real** ecology?

Hassel and May (1976).

## Hassel and May (1976).

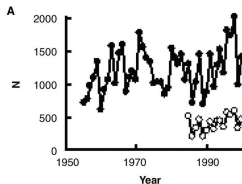


## More examples

Costantino et al. Flour beetle (1997) Science.



Cousin et al. Soay Sheep (2001) Science.

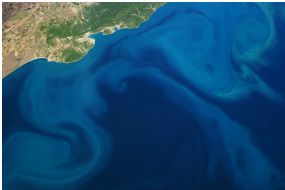


## And even more...

Becks et al. Microbial food web (2005) Nature.



Beninca et al. Plankton community (2008) Nature.





# Why control chaotic fluctuations?



# Outline

- 1 Introduction
  - Models
  - Stabilizing strategies
- 2 Adaptive limiter control
  - Stabilization
  - Targeting
  - Cost
- 3 Global stability

# Evolution of population size

## Discrete population model

$$x_t = f(x_{t-1})$$

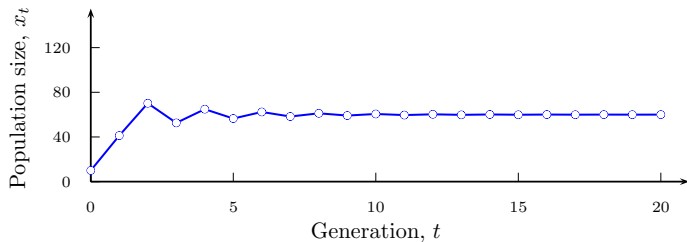
- $x_t$  population size
- $f$  describes the population production

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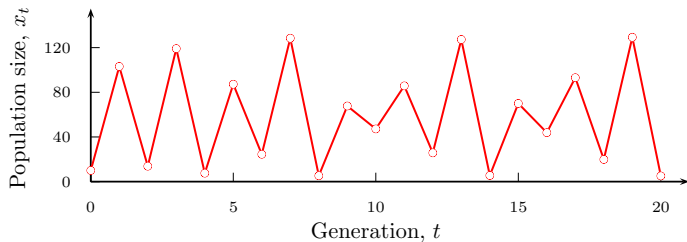


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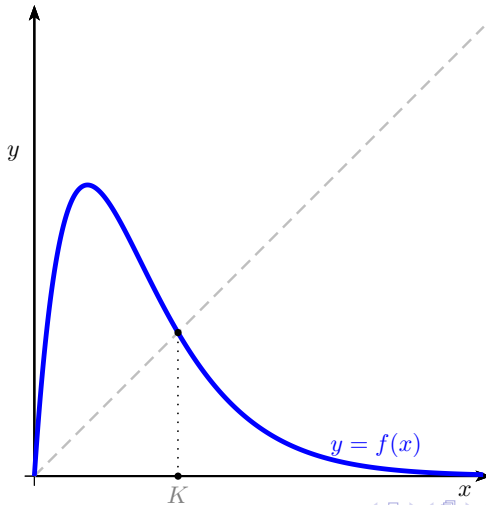
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# How is $f$ in overcompensatory models?

- 1  $f: \mathbb{R}_+ \rightarrow \mathbb{R}_+$  continuous,  $f(0) = 0$  and  $f(x) > 0$  for  $x > 0$ .
- 2  $f$  has a unique positive equilibrium  $K$ ,  $f(x) > x$  for  $x \in (0, K)$ , and  $f(x) < x$  for  $x > K$ .
- 3 There exists  $d < K$  such that  $f$  is increasing in  $(0, d)$  and decreasing elsewhere.

# How is $f$ in overcompensatory models?





## Chaos control problem (stabilization)

- Fluctuating population
- Transform it in a (more) stable population
- How?
- Small perturbations of  $f$  are not considered suitable, e.g. OGY method

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Adding / Stocking



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Removing / Harvesting



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Mixing both



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Constant harvesting,  
Constant enrichment, Constant  
effort harvesting, Constant  
effort enrichment, Limiter  
controls, Target oriented  
controls, Pulse strategies,...

Stabilization, Global stability,  
Hydra effects, Bubbles,  
Generated risks (Allee effects),  
Cost, Initial transient,...

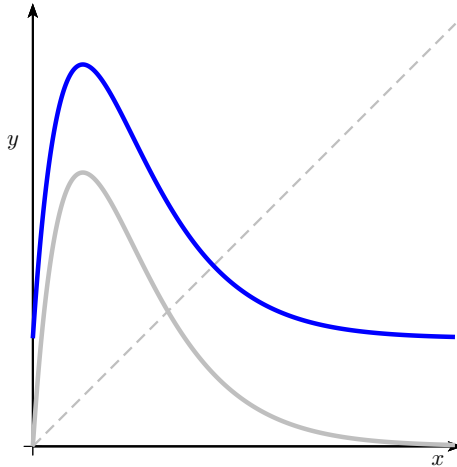
# Constant control

## Constant immigration

$$x_{n+1} = f(x_n) + C, \quad C > 0$$

McCallum (1992) J. Theor. Biol.; Parthasarathy & Sinha (1995) Phys. Rev. E; Stone & Hart (1999) Theor. Pop. Biol.; Wieland (2002) Phys. Rev. E.

# Constant control



# Population maps and GAS

Global stability?



# Population maps and GAS

Local dynamics drive global dynamics for  $S$ -unimodal maps!  
D. Singer (1978) *SIAM J. Appl. Math.*

④  $f$  satisfies  $Sf < 0$  with

$$(Sf)(x) := \frac{f'''(x)}{f'(x)} - \frac{3}{2} \left( \frac{f''(x)}{f'(x)} \right)^2.$$

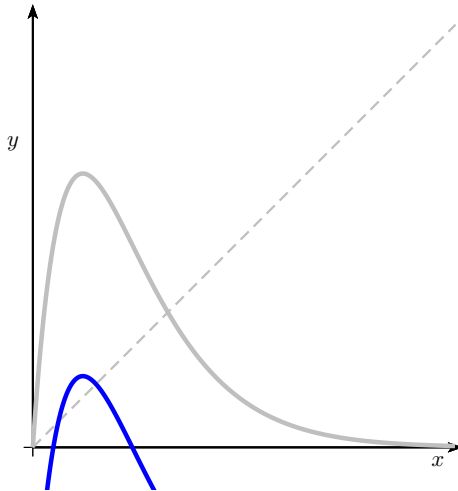
# Constant control

## Harvesting a constant quota

$$x_{n+1} = f(x_n) - C, \quad C > 0$$

Guéron (1998) Phys. Rev. E ; Schreiber (2001) J. Math. Biol.; Liz (2010) Theor. Ecol.

# Constant control



# Constant control

$$x_{n+1} = \max\{0, f(x_n) - C\}, \quad C > 0$$

- Extinction
- Essential extinction
- Bistability

Schreiber (2001) J. Math. Biol.

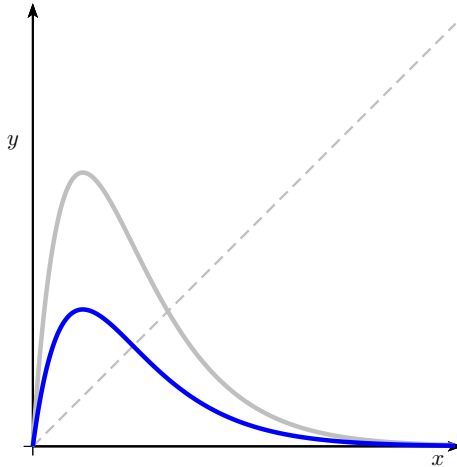
# Proportional Control

## Harvesting with constant effort

$$x_{n+1} = (1 - \gamma)f(x_n), \quad \gamma \in (0, 1).$$

Güémez & Matías (1993) Phys. Lett. A; Liz (2010) Phys. Lett. A

# Proportional Control



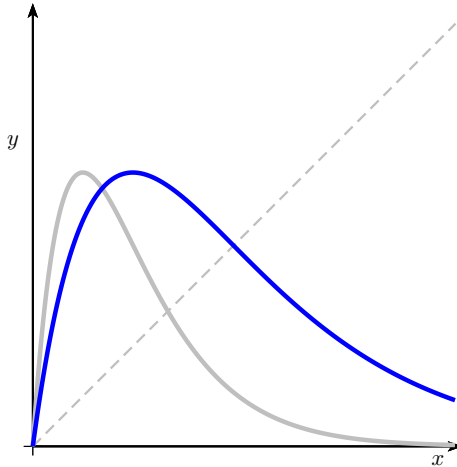
# Proportional Control

Harvesting before reproduction,

$$x_{n+1} = f((1 - \gamma)x_n), \quad \gamma \in (0, 1),$$

discloses an interesting effect.

# Proportional Control





# Proportional Control



Hydra effect

# Proportional Control

## Proportional immigration

$$x_{n+1} = f((1 + \gamma)x_n), \quad \gamma \in (0, \infty).$$

Carmona & Franco. (2011) Nonlinear Anal. R.W.A.

## Target oriented control (focusing on the size)

$$x_{n+1} = f(\gamma T + (1 - \gamma)x_n), \quad \gamma \in (0, 1).$$

Dattani et al. (2011) Phys. Lett. A; Franco & Liz (2013) Int. J. Bifurcations and Chaos

- There is not danger of extinction.
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- Defines certain threshold (the limiter)
- Acts when the population surpasses the limiter
- Sends the population back to the limiter

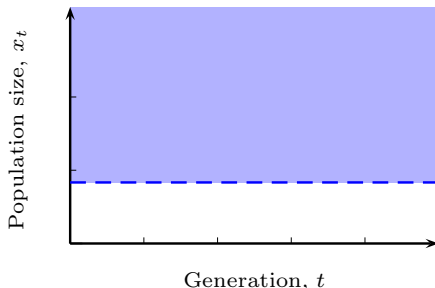
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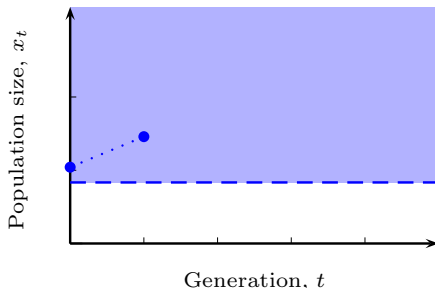


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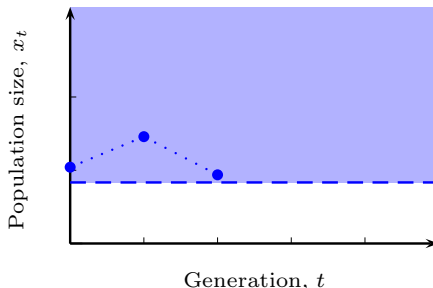


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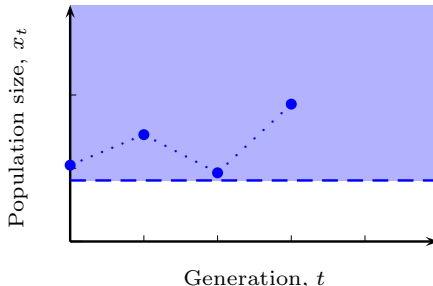


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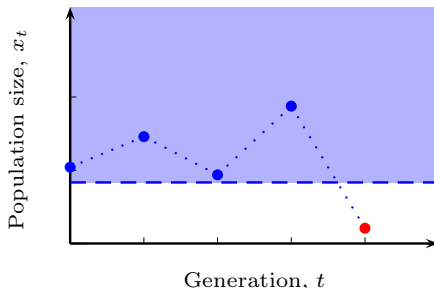


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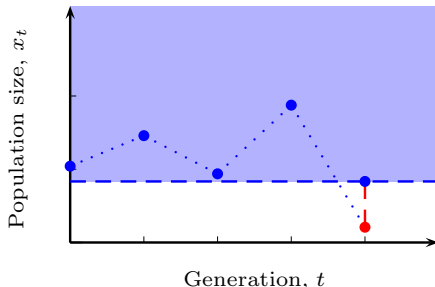


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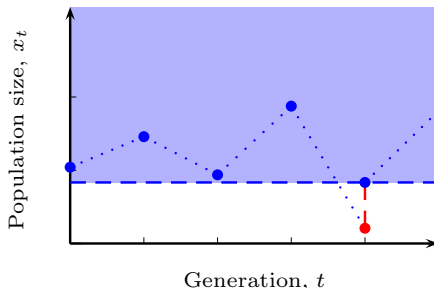


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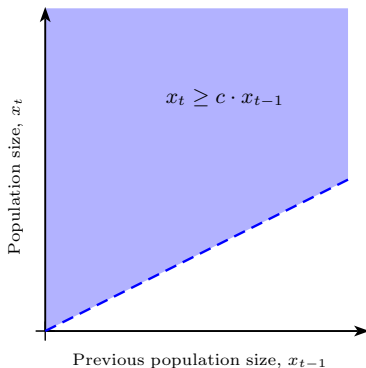
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### Limiter variable over time

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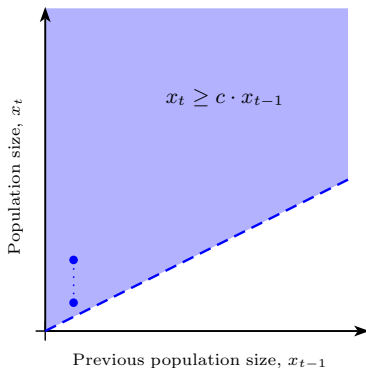
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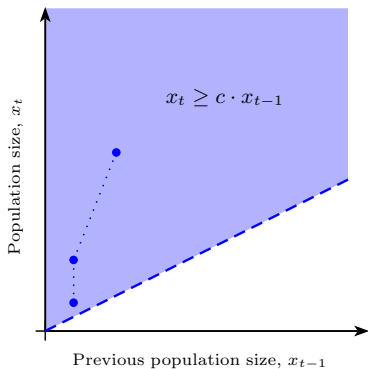
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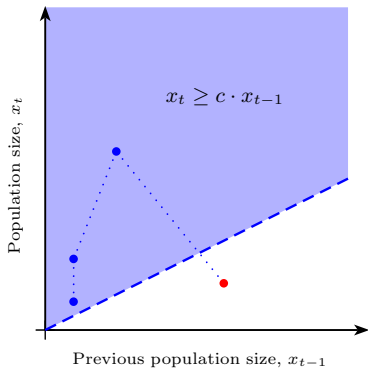




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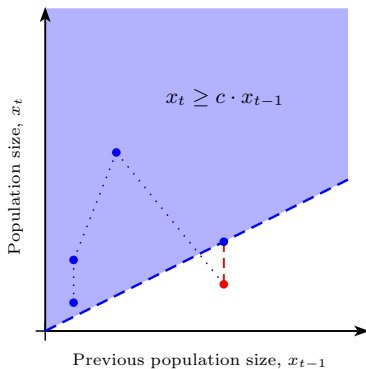
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# Atypical control

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Sah, Salve and Dey (J. Theor. Biol. 2013)

Mathematical model

Numerical simulations

Empirically tested

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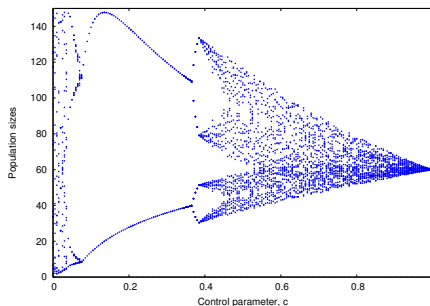
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# Analytical results

Effect of increasing the ALC intensity

For a **general family** population maps (Franco & Hilker (2013) *J. Theor. Biol.* ):

- 1 ALC is not able to stabilise an equilibrium
- 2 ALC **globally** confines the population in a **trapping region** with size tending to 0 as  $c$  increases
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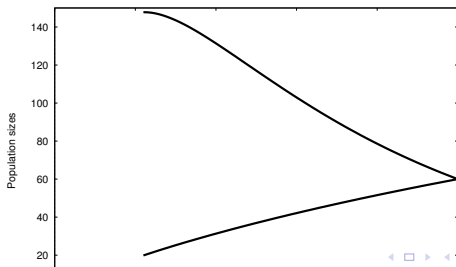


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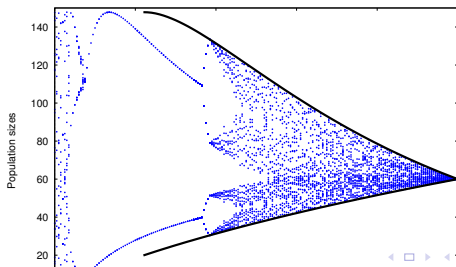


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# Application

## How to choose the ALC intensity

### Recipes for:

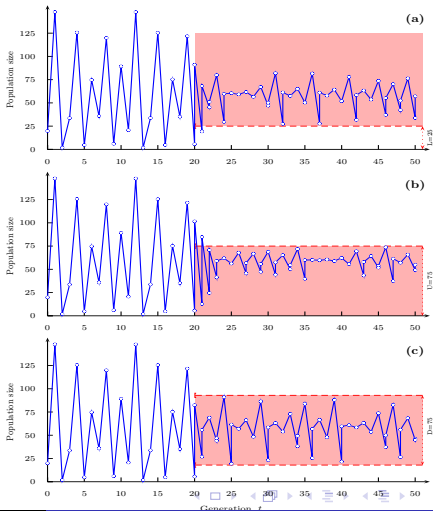
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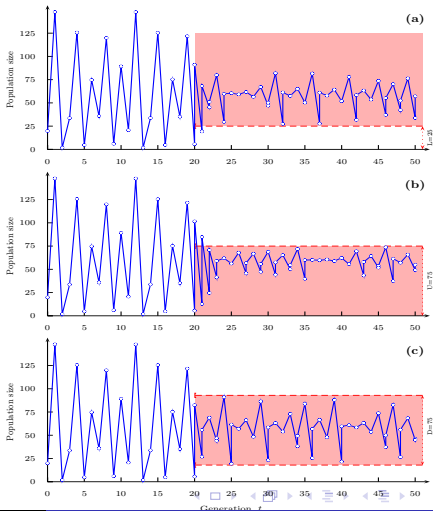


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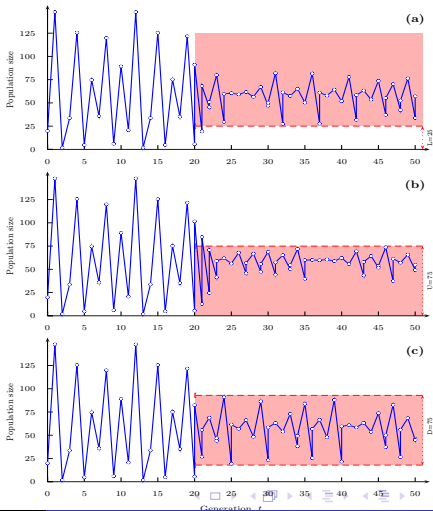


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# Main tool

## Importance of intra-generation variations

- **Correct** mathematical model  $a_{t+1} = \max\{f(a_t), ca_t\}$
- When ALC acts we have two population sizes  $b_t$  and  $a_t$
- We must choose one to define the limiter in  $t + 1$
- In experiments and numeric simulations  $a_t$  was selected
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$$b_{t+1} = f(a_t) \quad \text{and} \quad a_{t+1} = \begin{cases} b_{t+1}, & b_{t+1} \geq c \cdot a_t \\ c \cdot a_t, & b_{t+1} < c \cdot a_t \end{cases}$$

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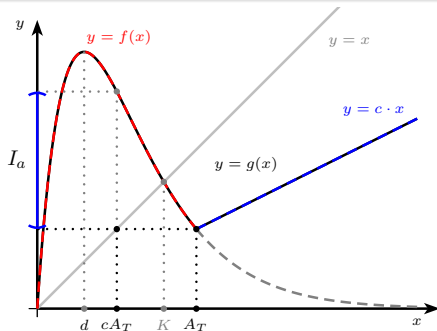
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## Activation threshold

There exists a unique  $A_T$  such that  $a_{t+1} = \begin{cases} f(a_t), & a_t \leq A_T, \\ c \cdot a_t, & a_t > A_T, \end{cases}$

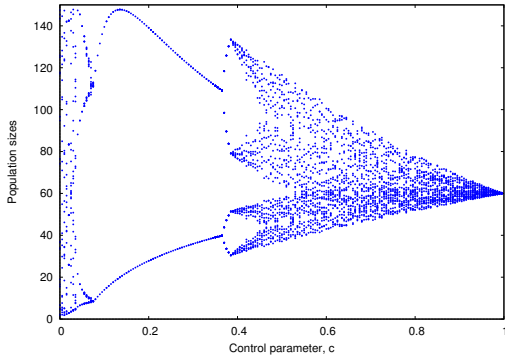
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# Effort

Measured as the number of individuals added

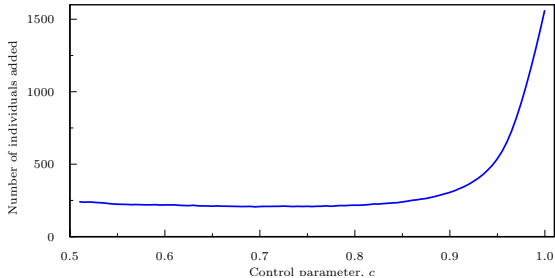




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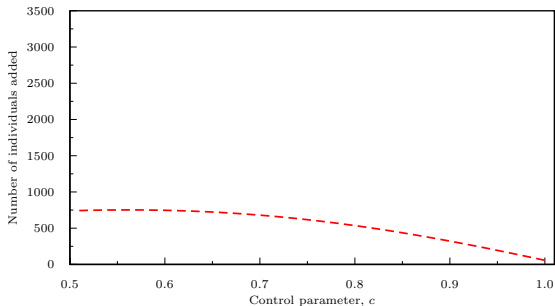
## ALC



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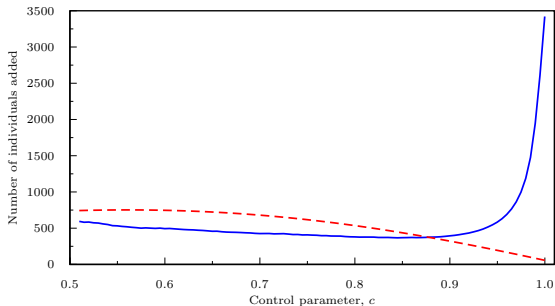
## Classic Limiter Control



# Effort

Measured as the number of individuals added

## LC vs ALC



Importance of **initial transients**

# Effort

Measured as the number of interventions

**Number or consecutive interventions**

ALC never acts consecutively inside the trapping region.

## What happens if we use $b_t$ ?

ALCa

$$a_{t+1} = \begin{cases} b_{t+1}, & b_{t+1} \geq c \cdot a_t, \\ c \cdot a_t, & b_{t+1} < c \cdot a_t, \end{cases}$$

Franco & Hilker (2014) *SIAM J. Appl. Dyn. Syst.*

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ALCb

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Together with  $b_{t+2} = f(a_{t+1})$

Franco & Hilker (2014) *SIAM J. Appl. Dyn. Syst.*

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Model suggested for ALCa!

Franco & Hilker (2014) *SIAM J. Appl. Dyn. Syst.*

# Properties of ALCb

## Similarities with ALCa

### Proposition equilibria

ALCb it is not able to stabilize equilibria.

### Proposition activating threshold

ALCb adds individuals in generation  $t$  with  $t \geq 1$  if and only if  $b_{t-1} > A_T$ .

### Proposition frequency

ALCb never acts in two consecutive generations.

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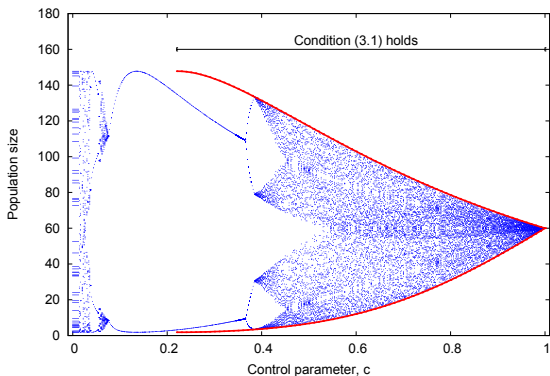
## Similarities with ALCa

### Proposition trapping region

ALCb confines the population sizes before intervention in an interval around  $K$  which shrinks as  $c$  grows.

# Properties of ALCb

## Similarities with ALCa



# Properties of ALCb

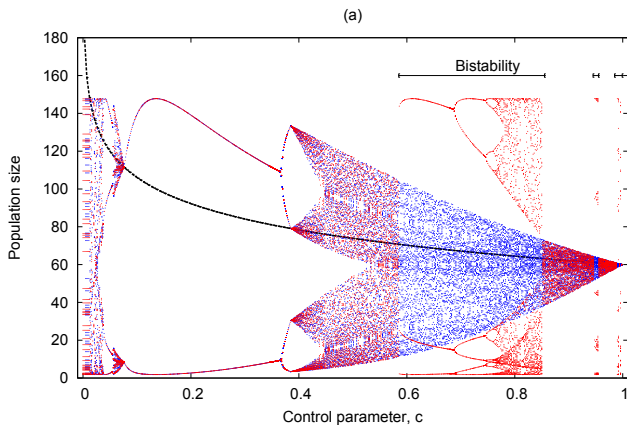
## Difference with ALCa

### Proposition trapping region

ALCb confines the population sizes before intervention in an interval around  $K$  which shrinks as  $c$  grows, **provided that the initial population size  $b_0$  belongs to certain subinterval of the trapping region.**

# Properties of ALCb

## Difference with ALCa





# Global stability in higher order systems

1-D Ricker  $x_{n+1} = x_n e^{r-x_n}$  (40 years ago)

2-D Ricker  $x_{n+1} = x_n e^{r-x_{n-1}}$  (1 year ago, computer aided proof)

# Global stability in higher order systems

1-D Ricker  $x_{n+1} = x_n e^{r-x_n}$  (40 years ago)

2-D Ricker  $x_{n+1} = x_n e^{r-x_{n-1}}$  (1 year ago, computer aided proof)

# Global stability in higher order systems

## Juveniles and adults

$$\begin{cases} x_{t+1,1} = f((1 - h_2)x_{t,2}), \\ x_{t+1,2} = (1 - h_1)s_1x_{t,1} + (1 - h_2)s_2x_{t,2}, \end{cases} \quad (1)$$

$x_{t,1}$  juveniles;  $x_{t,2}$  adults;  $h_1, h_2 \in [0, 1)$  harvest rates;  
 $s_1, s_2 \in (0, 1]$  survivorship rates;  $f(y) = \alpha ye^{-\beta y}$  Ricker map  
 with  $\alpha > 1, \beta > 0$ .

E. Liz & P. Pilarczyk. *J. Theoret. Biol.* (2012)

Using a result of I. Györi & S. Trofimchuk, system (1) has a positive global attractor for

$$\ln \alpha \in (r_0, r_0 + 1], \quad \text{where } r_0 := \ln \left( \frac{1 - (1 - h_2)s_2}{(1 - h_1)(1 - h_2)s_1} \right).$$

# General setting

## Lure system

$$x_{t+1} = Ax_t + bf(c^T x_t), \quad (2)$$

with  $A$  a non-negative matrix in  $\mathbb{R}^{n \times n}$ ;  $b, c \in \mathbb{R}_+^n \setminus \{0\}$ ; and  $f: \mathbb{R}_+ \rightarrow \mathbb{R}_+$  continuous,  $f(0) = 0$ , and  $f(y) > 0, \forall y \in \mathbb{R}_+ \setminus \{0\}$ .

# Absolute stability

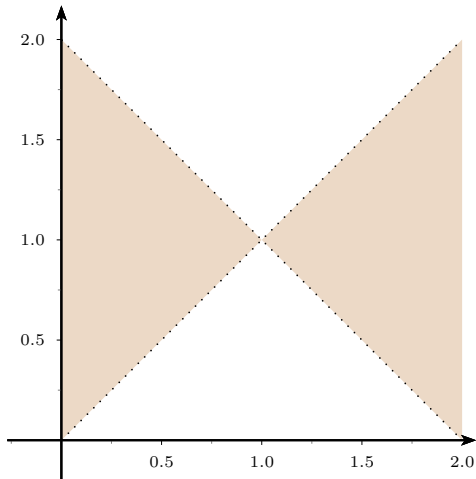
Sector bound condition (Townley et. al. *Systems & Control Letters* 2012)

There exists a unique  $y^* > 0$  so that  $f(y^*) = py^*$  and

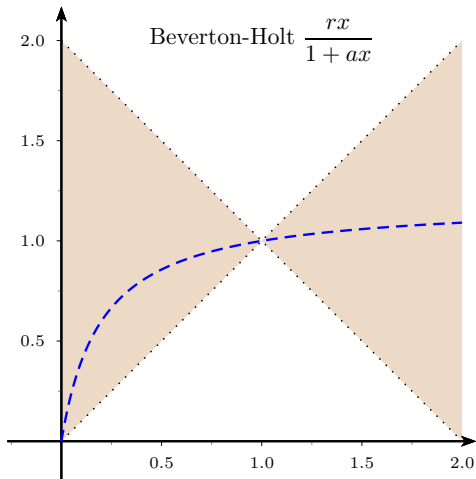
$$|f(y) - py^*| < p|y - y^*|, \quad y \in \mathbb{R}_+ \setminus \{0, y^*\}$$

with  $p := \frac{1}{c^T(I-A)^{-1}b}$ .

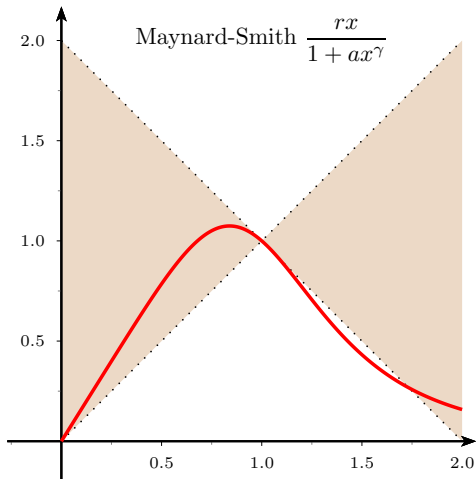
# Absolute stability



# Absolute stability

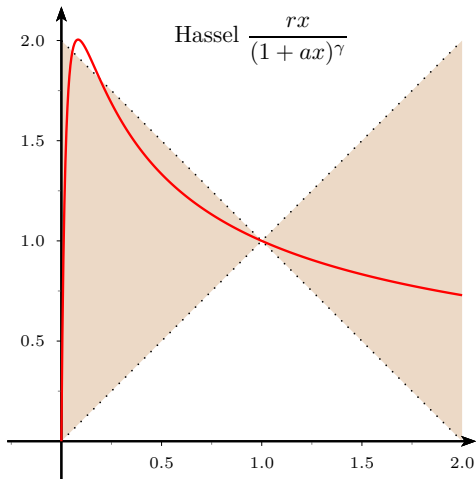


# Absolute stability





# Absolute stability



## Particular case

$$A = \begin{pmatrix} 1 - \delta & 0 & 0 & \cdots & 0 \\ a_1 & 0 & 0 & \cdots & 0 \\ 0 & a_2 & 0 & \cdots & 0 \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & \cdots & 0 & a_{n-1} & 0 \end{pmatrix}, \quad b = \begin{pmatrix} b_1 \\ 0 \\ 0 \\ 0 \\ 0 \end{pmatrix},$$

where  $0 < \delta \leq 1$ , and  $a_i, b_1 > 0$ ,

Results from El-Morshedy and Jiménez-López (2008) can be adapted to show that the GAS of the map implies GAS for the system.

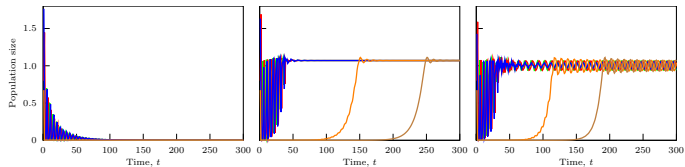
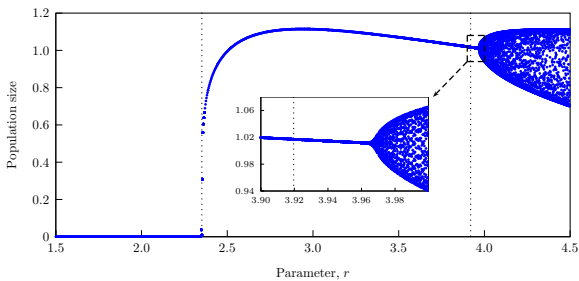
# Consequence

System (1) has a positive global attractor for

$$\ln \alpha \in (r_0, r_0 + 2], \quad \text{where } r_0 := \ln \left( \frac{1 - (1 - h_2)s_2}{(1 - h_1)(1 - h_2)s_1} \right).$$

Franco, Logemann & Perán (2014) *System & Control Letters*

# Sharp?



# Summary

- 1 **Analytical support** for ALC and TOC.
- 2 **New interesting properties** of ALC and TOC.
- 3 Recent global stability results improved.

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Liz



Hilker



Logemann



Perán