

# Modeling and Control of Renewable Resources

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Benasque, 25.08.2017

# Introduction: Structured Population Models

$u$             population density

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$u = u(t)$     population density  
 $t$                     time

# Introduction: Structured Population Models

$u = u(t, x)$  population density

$t$  time

$x$  age

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$$\partial_t u + \partial_x u = 0$$

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$x$  biological age/size/(trait)

$$\partial_t u + \partial_x ([\text{growth/aging}] u) = 0$$

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$$\partial_t u + \partial_x (g(t, x, f u) u) = 0$$

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$$\partial_t u + \partial_x (g(t, x, f u) u) = [\text{death}]$$



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$$\partial_t u + \partial_x (g(t, x, fu) u) = -d(t, x, u, fu)$$

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$$u(t, 0) = [\text{birth}]$$

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$$u(t, 0) = b(t, fu)$$

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$$u(t, 0) = b(t, \int u)$$

$$u(0, x) = [\text{initial datum}]$$

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Balance Law with **Boundary**  
**NonLocal** flow  
**NonLocal** source  
**NonLocal** boundary data

# Introduction: Conservation Laws – Keywords

$$\partial_t u + \operatorname{div}_x f(t, x, u) = g(t, x, u)$$

$$x \in \mathbb{R}^N \quad u \in \mathbb{R}^n$$

$$N = 1, n \geq 1$$

$$N \geq 1, n = 1$$

Norm:  $\|\cdot\|_{\mathbf{L}^1}$

$x$  regularity:  $x \rightarrow u(t, x)$   $\mathbf{L}^1, \mathbf{L}^\infty, \mathbf{BV}$

$t$  regularity:  $t \rightarrow u(t)$   $\mathbf{L}^1$  – Lipschitz

$u_o$  dependence:  $u_o \rightarrow u(t)$   $\mathbf{L}^1$  – Lipschitz

# Introduction: NonLocal Conservation Laws

- ▶ NonLocal Conservation Laws:
  - ▶ Elastodynamics
    - ▶ (Chen & Christoforou: PAMS, 2007)
    - ▶ (Christoforou: JHDE, 2007)



# Introduction: NonLocal Conservation Laws

- ▶ NonLocal Conservation Laws:
  - ▶ Elastodynamics
  - ▶ Granular Materials
    - ▶ (Amadori & Shen: Comm.PDE, 2009)
    - ▶ (Guerra & Shen: JDE, 2014)

# Introduction: NonLocal Conservation Laws

- ▶ NonLocal Conservation Laws:
  - ▶ Elastodynamics
  - ▶ Granular Materials
  - ▶ Vehicular Traffic
    - ▶ (Colombo, Corli & Rosini: ZAMM, 2007)
    - ▶ (Colombo, Herty & Mercier: COCV, 2011)
    - ▶ (Dong & Tong: NHM, 2011)
    - ▶ (Blandin & Goatin: Numer.Math., 2015)

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- ▶ NonLocal Conservation Laws:
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  - ▶ Vehicular Traffic
  - ▶ Crowd Dynamics
    - ▶ (Colombo, Garavello & Mercier: M3AS, 2011)
    - ▶ (Colombo & Mercier: Acta Math.Sc., 2011)
    - ▶ (Piccoli & Tosin: ARMA, 2011)
    - ▶ (Hoogendoorn et al.: Physica A, 2014)
    - ▶ (Goatin & Rossi: CMS, 2017)

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  - ▶ Granular Materials
  - ▶ Vehicular Traffic
  - ▶ Crowd Dynamics
  - ▶ Numerical Methods
    - ▶ (Betancourt, Burger, Karlsen & Tory: Nonlin., 2011)
    - ▶ (Amorim, Colombo & Teixeira: M2AN, 2015)
    - ▶ (Aggarwal, Colombo & Goatin: SINUM, 2015)

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  - ▶ Degasperis–Procesi, Camassa–Holm & Ostrovsky–Hunter
    - ▶ (Gesztesy & Holden: Cambridge Studies, 2003)
    - ▶ (Holden & Xavier: CPDE, 2007)
    - ▶ (Coclite, Ridder & Risebro: Preprint, 2016)

# Introduction: Structured Population

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  - ▶ Numerical Methods
  - ▶ Degasperis–Procesi, Camassa–Holm & Ostrovsky–Hunter
  - ▶ Measure Valued Conservation Laws
    - ▶ (Carrillo, Colombo, Gwiazda & Ulikowska: JDE, 2012)
    - ▶ (Gwiazda, Jamróz & Marciniak-Czochra: SIMA, 2012)
    - ▶ (Canizo, Carrillo & Cuadrado: Acta Appl.Math., 2013)
    - ▶ (Piccoli & Rossi: ARMA, 2014)
    - ▶ (Colombo, Gwiazda & Rosinska: COCV, To appear)

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- ▶ NonLocal Conservation Laws:
  - ▶ Elastodynamics
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  - ▶ Crowd Dynamics
  - ▶ Numerical Methods
  - ▶ Degasperis–Procesi, Camassa–Holm & Ostrovsky–Hunter
  - ▶ Measure Valued Conservation Laws
  
- ▶ Structured Population
  - ▶  $\approx$  2000 items in MathSciNet (190 books)
    - ▶ (Webb: Theory of Nonlinear Age-Dependent Population Dynamics, 1985)
    - ▶ (Diekmann & Heesterbeek: Mathematical Epidemiology of Infectious Diseases, 2000)
    - ▶ (Perthame: Transport Equations in Biology, 2008)
    - ▶ (Broom & Richtář: Game-Theor. Models in Biology, 2013)

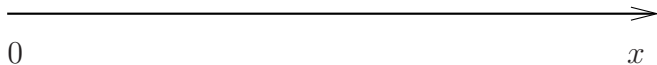
# Structured Population on Graphs

(in collaboration with M.Garavello)

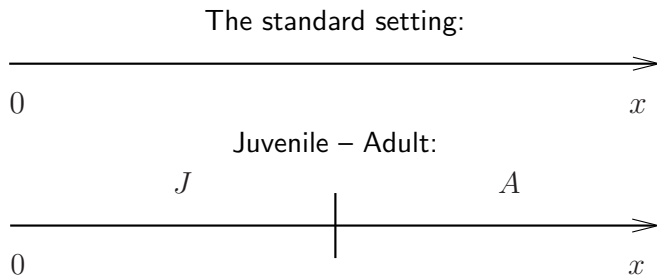


# Role of the Graph

The standard setting:



# Role of the Graph



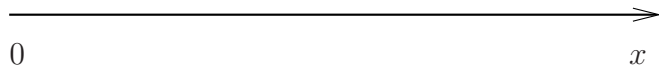
(Carrillo, Cuadrado & Perthame: Math.Biosci., 2007)

(Ackleh & Deng: SIAM Appl.Math., 2009)

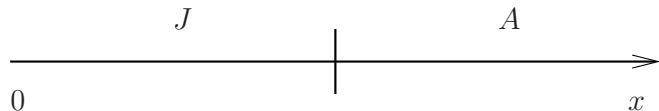
(Ackleh & Ma: Numer.Funct.Anal.Opt., 2013)

# Role of the Graph

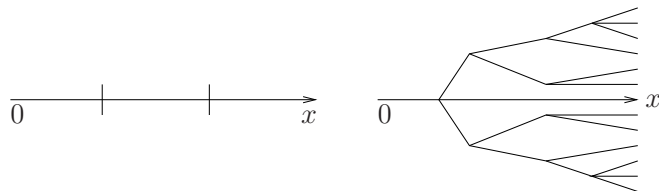
The standard setting:



Juvenile – Adult:



Etc. ...



## A General Theorem

$$\left\{ \begin{array}{l} \partial_t u_i + \partial_x (g_i(t, x) u_i) = d_i(t, x) u_i \\ g_i(t, 0) u_i(t, 0+) = \mathcal{B}_i(t, u_1(t), \dots, u_n(t)) \\ u_i(0, x) = \bar{u}_i(x) \end{array} \right.$$

$t \in \mathbb{R}^+$	time	$g_i$	growth
$x \in \mathbb{R}^+$	age	$-d_i$	death
$u_i \in \mathbb{R}^+$	density	$\mathcal{B}_i$	birth/change

## A General Theorem

$$\begin{cases} \partial_t u_i + \partial_x (g_i(t, x) u_i) = d_i(t, x) u_i \\ g_i(t, 0) u_i(t, 0+) = \mathcal{B}_i(t, u_1(t), \dots, u_n(t)) \\ u_i(0, x) = \bar{u}_i(x) \end{cases}$$

$$\mathcal{B}_i(t, u_1, \dots, u_n) = \alpha_i(t, u_1(\bar{x}_1-), \dots, u_n(\bar{x}_n-)) \\ + \beta_i \left( \int_{I_1} w_1(x) u_1(x) dx, \dots, \int_{I_n} w_n(x) u_n(x) dx \right)$$

$u_i$  lives on  $[0, \bar{x}_i]$   
 $u_i$  is fertile for  $x \in I_i$   
 $w_i$  fertility of  $u_i$

$\alpha_i$  transmission  
 $\beta_i$  natality

## A General Theorem

$$\begin{cases} \partial_t u_i + \partial_x (g_i(t, x) u_i) = d_i(t, x) u_i \\ g_i(t, 0) u_i(t, 0+) = \mathcal{B}_i(t, u_1(t), \dots, u_n(t)) \\ u_i(0, x) = \bar{u}_i(x) \end{cases}$$

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Theorem (Colombo & Garavello: MBE, 2015)

If  $g_i \in \mathbf{C}^1$ ,  $\inf g_i > 0$ ,  $\sup_t [\text{TV}(g_i(t)) + \text{TV}(\partial_x g_i(t))] < +\infty$   
 $d_i \in (\mathbf{C}^1 \cap \mathbf{L}^\infty)$ ,  $\sup_t \text{TV}(d_i(t, \cdot)) < +\infty$   
 $\alpha_i, \beta_i, w_i \in \mathbf{C}^{0,1}$ ,  $\alpha_i(t, 0) = 0$ ,  $\beta_i(0) = 0$ ,  $\inf_x w_i > 0$

## A General Theorem

$$\begin{cases} \partial_t u_i + \partial_x (g_i(t, x) u_i) = d_i(t, x) u_i \\ g_i(t, 0) u_i(t, 0+) = \mathcal{B}_i(t, u_1(t), \dots, u_n(t)) \\ u_i(0, x) = \bar{u}_i(x) \end{cases}$$

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Then: *Existence of a solution*

*Uniqueness of the solution*

*Continuous dependence from the initial data*

*Stability with respect to  $\alpha_i, \beta_i, w_i$*

## A General Theorem

$$\begin{cases} \partial_t u_i + \partial_x (g_i(t, x) u_i) = d_i(t, x) u_i \\ g_i(t, 0) u_i(t, 0+) = \mathcal{B}_i(t, u_1(t), \dots, u_n(t)) \\ u_i(0, x) = \bar{u}_i(x) \end{cases}$$

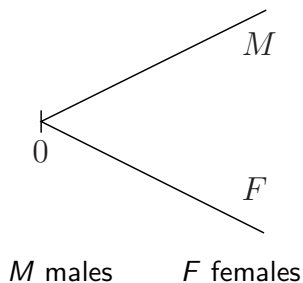
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Theorem (Colombo & Garavello: MBE, 2015)

$$\begin{aligned} \|u_i'(t) - u_i''(t)\|_{\mathbf{L}^1} \leq & \mathcal{K}(t) \sum_{j=1}^n \left[ \|\bar{u}_j' - \bar{u}_j''\|_{\mathbf{L}^1} + t \|\bar{u}_j' - \bar{u}_j''\|_{\mathbf{L}^\infty} \right] \\ & + \mathcal{H}(t) \sum_{j=1}^n \left[ \|\alpha_j' - \alpha_j''\|_{\mathbf{C}^0} + \|\beta_j' - \beta_j''\|_{\mathbf{C}^0} + \|w_j' - w_j''\|_{\mathbf{C}^0} \right] \end{aligned}$$



## Example: Age & Sex Structured Population



(N. Keyfitz: VI Berkeley Symp. Math. Stat. Prob., 1972)

## Example: Age & Sex Structured Population

$$\left\{ \begin{array}{l} \partial_t M + \partial_a M = -\kappa \mu M \\ \partial_t F + \partial_a F = -(1 - \kappa) \mu F \\ M(t, 0) + F(t, 0) = \nu \min \left\{ \vartheta \int_{m_1}^{m_2} M(t, a) da, (1 - \vartheta) \int_{f_1}^{f_2} F(t, a) da \right\} \\ \eta M(t, 0) = (1 - \eta) F(t, 0) \end{array} \right.$$

$\eta$  relative natality in  $[0, 1]$

$\kappa$  relative mortality in  $[0, 1]$

$\vartheta$  coupling habits in  $[0, 1]$

$\mu$  mortality in  $\mathbb{R}^+$

$\nu$  natality in  $\mathbb{R}^+$

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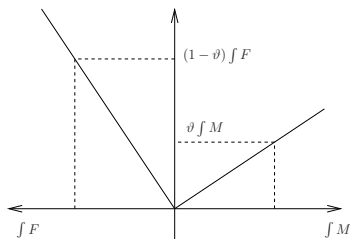
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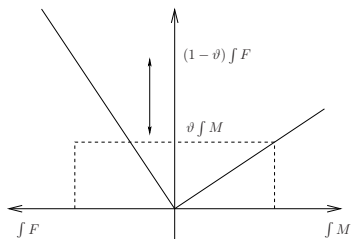
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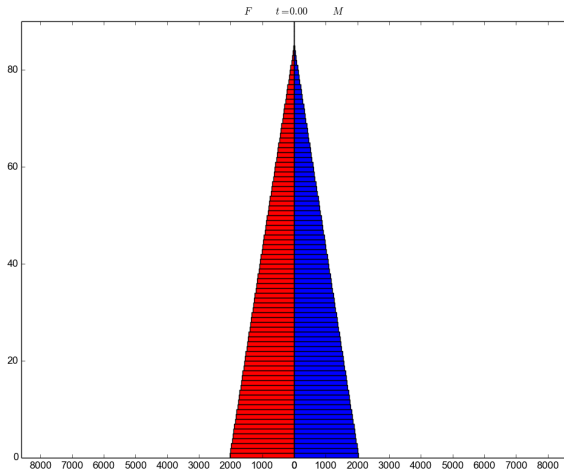


# Example: Age & Sex Structured Population

$$\kappa = 0.600$$

$$\eta = 0.485$$

$$\vartheta = 0.7067$$



Age pyramid - "slowly" growing

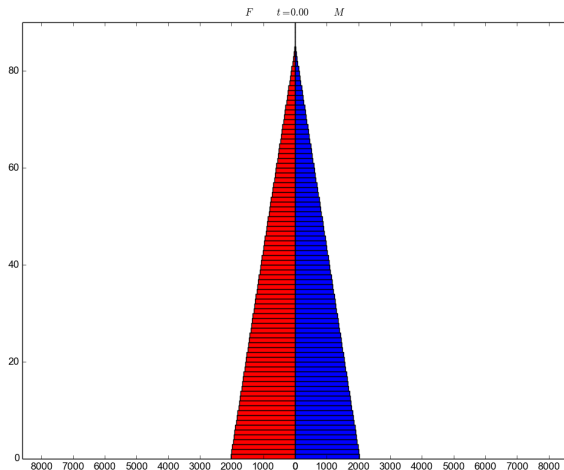


# Example: Age & Sex Structured Population

$$\kappa = 0.600$$

$$\eta = 0.300$$

$$\vartheta = 0.7067$$



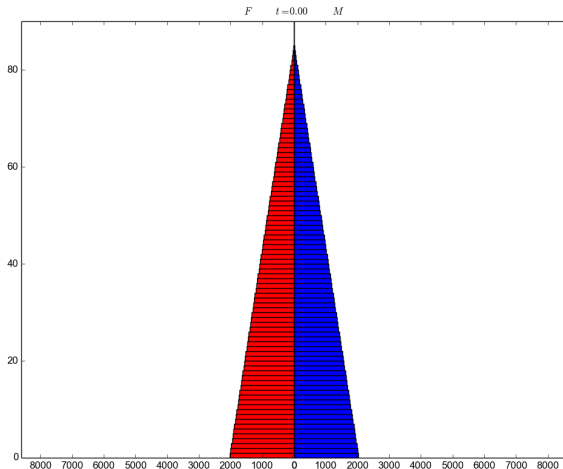
Lower  $\eta \Rightarrow$  too many  $M$  are born  $\Rightarrow$  extinction

# Example: Age & Sex Structured Population

$$\kappa = 0.400$$

$$\eta = 0.485$$

$$\vartheta = 0.7067$$



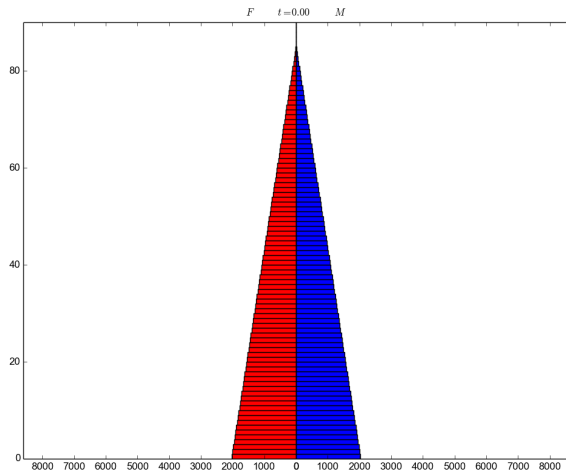
Lower  $\kappa \Rightarrow$  too many  $F$  die  $\Rightarrow$  extinction

# Example: Age & Sex Structured Population

$$\kappa = 0.600$$

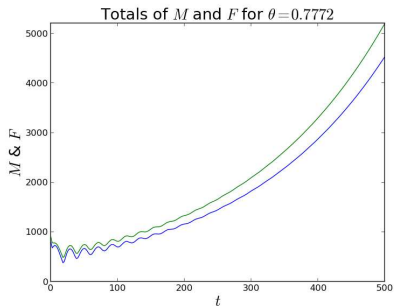
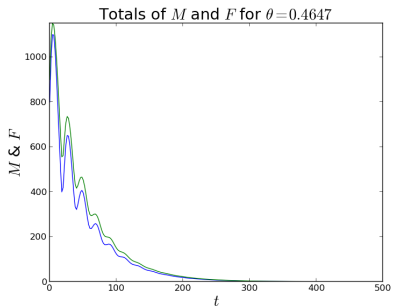
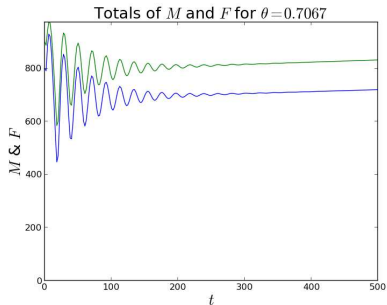
$$\eta = 0.485$$

$$\vartheta = 0.5000$$



Lower  $\vartheta \Rightarrow$  extinction

# Example: Age & Sex Structured Population



## Example: Age & Sex Structured Population

Optimal Mating Ratio: 
$$\vartheta(t) = \frac{\int_{f_1}^{f_2} F(t, a) da}{\int_{m_1}^{m_2} M(t, a) da + \int_{f_1}^{f_2} F(t, a) da}$$

Optimal Fertility Rate: 
$$\frac{\nu}{\frac{1}{\int_{m_1}^{m_2} M(t, a) da} + \frac{1}{\int_{f_1}^{f_2} F(t, a) da}}$$

(Colombo & Garavello: MBE, 2015)

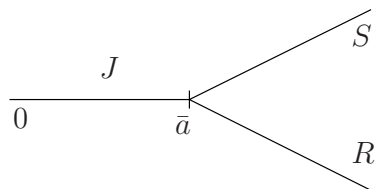
Control



Management of a Biological Resource

(in collaboration with M. Garavello)

# Management of a Biological Resource



$J$	Juveniles	$a \in [0, \bar{a}]$
$S$	Sold	$a \in [\bar{a}, +\infty[$
$R$	Reproduction	$a \in [\bar{a}, a_{\max}]$

## Management of a Biological Resource

$$\begin{array}{l}
 J \\
 0 \qquad \qquad \qquad \bar{a}
 \end{array}
 \left\{
 \begin{array}{l}
 S \quad \left\{ \begin{array}{l}
 \partial_t J + \partial_a (g_J(t, a) J) = d_J(t, a) J \\
 \partial_t S + \partial_a (g_S(t, a) S) = d_S(t, a) S \\
 \partial_t R + \partial_a (g_R(t, a) R) = d_R(t, a) R
 \end{array} \right. \\
 R \quad \left\{ \begin{array}{l}
 g_J(t, 0) J(t, 0) = \int w(\alpha) R(t, \alpha) d\alpha \\
 g_S(t, \bar{a}) S(t, \bar{a}) = \eta g_J(t, \bar{a}) J(t, \bar{a}) \\
 g_R(t, \bar{a}) R(t, \bar{a}) = (1 - \eta) g_J(t, \bar{a}) J(t, \bar{a})
 \end{array} \right.
 \end{array}
 \right.$$

$J$	Juveniles	$a \in [0, \bar{a}]$
$S$	Sold	$a \in [\bar{a}, +\infty[$
$R$	Reproduction	$a \in [\bar{a}, a_{\max}]$



## Management of a Biological Resource

A diagram showing a horizontal line starting at 0 and ending at  $\bar{a}$ . A vertical tick mark is at  $\bar{a}$ . The label  $J$  is placed above the line between 0 and  $\bar{a}$ . From the tick mark at  $\bar{a}$ , two lines branch out: one upwards to the label  $S$  and one downwards to the label  $R$ .

$$\left\{ \begin{array}{l} \partial_t J + \partial_a (g_J(t, a) J) = d_J(t, a) J \\ \partial_t S + \partial_a (g_S(t, a) S) = d_S(t, a) S \\ \partial_t R + \partial_a (g_R(t, a) R) = d_R(t, a) R \end{array} \right.$$
$$\left\{ \begin{array}{l} g_J(t, 0) J(t, 0) = \int w(\alpha) R(t, \alpha) d\alpha \\ g_S(t, \bar{a}) S(t, \bar{a}) = \eta g_J(t, \bar{a}) J(t, \bar{a}) \\ g_R(t, \bar{a}) R(t, \bar{a}) = (1 - \eta) g_J(t, \bar{a}) J(t, \bar{a}) \end{array} \right.$$

$$\text{Profit} = [\text{income from } S] - [\text{costs of } J, S, R]$$

# Management of a Biological Resource

$$\begin{array}{l} J \\ 0 \qquad \qquad \bar{a} \end{array} \left\{ \begin{array}{l} S \\ R \end{array} \right. \begin{cases} \partial_t J + \partial_a (g_J(t, a) J) = d_J(t, a) J \\ \partial_t S + \partial_a (g_S(t, a) S) = d_S(t, a) S \\ \partial_t R + \partial_a (g_R(t, a) R) = d_R(t, a) R \\ g_J(t, 0) J(t, 0) = \int w(\alpha) R(t, \alpha) d\alpha \\ g_S(t, \bar{a}) S(t, \bar{a}) = \eta g_J(t, \bar{a}) J(t, \bar{a}) \\ g_R(t, \bar{a}) R(t, \bar{a}) = (1 - \eta) g_J(t, \bar{a}) J(t, \bar{a}) \end{cases}$$

$$\text{Profit} = [\text{income from } S] - [\text{costs of } J, S, R]$$

Find  $\eta$  to maximize the profit

Stability estimates  $\Rightarrow$  existence of an optimal  $\eta$

# Management of a Biological Resource

$J$

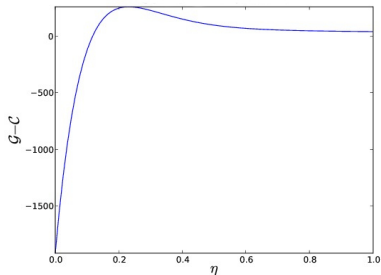
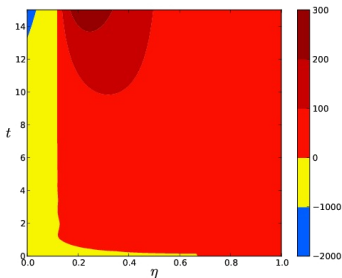
$0$

$\bar{a}$

$S$

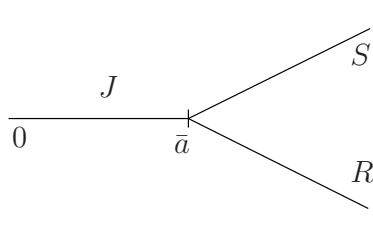
$R$

$$\begin{cases} \partial_t J + \partial_a (g_J(t, a) J) = d_J(t, a) J \\ \partial_t S + \partial_a (g_S(t, a) S) = d_S(t, a) S \\ \partial_t R + \partial_a (g_R(t, a) R) = d_R(t, a) R \\ g_J(t, 0) J(t, 0) = \int w(\alpha) R(t, \alpha) d\alpha \\ g_S(t, \bar{a}) S(t, \bar{a}) = \eta g_J(t, \bar{a}) J(t, \bar{a}) \\ g_R(t, \bar{a}) R(t, \bar{a}) = (1 - \eta) g_J(t, \bar{a}) J(t, \bar{a}) \end{cases}$$



(Colombo & Garavello: MBE, 2015)

# Management of a Biological Resource


$$\left\{ \begin{array}{l} \partial_t J + \partial_a (g_J(t, a) J) = d_J(t, a) J \\ \partial_t S + \partial_a (g_S(t, a) S) = d_S(t, a) S \\ \partial_t R + \partial_a (g_R(t, a) R) = d_R(t, a) R \\ \\ g_J(t, 0) J(t, 0) = \int w(\alpha) R(t, \alpha) d\alpha \\ g_S(t, \bar{a}) S(t, \bar{a}) = \eta g_J(t, \bar{a}) J(t, \bar{a}) \\ g_R(t, \bar{a}) R(t, \bar{a}) = (1 - \eta) g_J(t, \bar{a}) J(t, \bar{a}) \end{array} \right.$$

Profit = [income from  $S$ ] – [costs of  $J, S, R$ ]

Find  $\eta$  to maximize the profit

Stability estimates  $\Rightarrow$  existence of an optimal  $\eta$

The Profit is Differentiable w.r.t  $\eta$ !

# Management of a Biological Resource

Diagram illustrating the management of a biological resource  $J$  over time  $t$  and space  $a$ . The resource is initially at  $0$  and is managed up to  $\bar{a}$ . The system is governed by the following equations:

$$\left\{ \begin{array}{l} S \\ R \end{array} \right. \left\{ \begin{array}{l} \partial_t J + \partial_a (g_J(t, a) J) = d_J(t, a) J \\ \partial_t S + \partial_a (g_S(t, a) S) = d_S(t, a) S \\ \partial_t R + \partial_a (g_R(t, a) R) = d_R(t, a) R \\ \\ g_J(t, 0) J(t, 0) = \int w(\alpha) R(t, \alpha) d\alpha \\ g_S(t, \bar{a}) S(t, \bar{a}) = \eta(\mathbf{t}) g_J(t, \bar{a}) J(t, \bar{a}) \\ g_R(t, \bar{a}) R(t, \bar{a}) = (1 - \eta(\mathbf{t})) g_J(t, \bar{a}) J(t, \bar{a}) \end{array} \right.$$

## Management of a Biological Resource

$$\begin{cases}
 \partial_t J + \partial_a (g_J(t, a) J) = d_J(t, a) J \\
 \partial_t S + \partial_a (g_S(t, a) S) = d_S(t, a) S \\
 \partial_t R + \partial_a (g_R(t, a) R) = d_R(t, a) R \\
 \\
 g_J(t, 0) J(t, 0) = \int w(\alpha) R(t, \alpha) d\alpha \\
 g_S(t, \bar{a}) S(t, \bar{a}) = \eta(t) g_J(t, \bar{a}) J(t, \bar{a}) \\
 g_R(t, \bar{a}) R(t, \bar{a}) = (1 - \eta(t)) g_J(t, \bar{a}) J(t, \bar{a})
 \end{cases}$$

$$C_J = \iint C_J(t, a, J(t, a)) da dt$$

$$C_S = \iint C_S(t, a, S(t, a)) da dt$$

$$C_R = \iint C_R(t, a, R(t, a)) da dt$$

## Management of a Biological Resource

$$\left\{ \begin{array}{l} \partial_t J + \partial_a (g_J(t, a) J) = d_J(t, a) J \\ \partial_t S + \partial_a (g_S(t, a) S) = d_S(t, a) S \\ \partial_t R + \partial_a (g_R(t, a) R) = d_R(t, a) R \\ \\ g_J(t, 0) J(t, 0) = \int w(\alpha) R(t, \alpha) d\alpha \\ g_S(t, \bar{a}) S(t, \bar{a}) = \eta(\mathbf{t}) g_J(t, \bar{a}) J(t, \bar{a}) \\ g_R(t, \bar{a}) R(t, \bar{a}) = (1 - \eta(\mathbf{t})) g_J(t, \bar{a}) J(t, \bar{a}) \\ \\ S(t, \bar{a}_i+) = (1 - \vartheta_i(\mathbf{t})) S(t, \bar{a}_i-) \end{array} \right.$$

$$C_J = \iint C_J(t, a, J(t, a)) da dt$$

$$C_S = \iint C_S(t, a, S(t, a)) da dt$$

$$C_R = \iint C_R(t, a, R(t, a)) da dt$$

$$\mathcal{I} = \int \sum_i \vartheta_i(\mathbf{t}) P_i(t) S(t, \bar{a}_i-) dt$$

## Management of a Biological Resource

$$\left\{ \begin{array}{l} \partial_t J + \partial_a (g_J(t, a) J) = d_J(t, a) J \\ \partial_t S + \partial_a (g_S(t, a) S) = d_S(t, a) S \\ \partial_t R + \partial_a (g_R(t, a) R) = d_R(t, a) R \\ \\ g_J(t, 0) J(t, 0) = \int w(\alpha) R(t, \alpha) d\alpha \\ g_S(t, \bar{a}) S(t, \bar{a}) = \boldsymbol{\eta(t)} g_J(t, \bar{a}) J(t, \bar{a}) \\ g_R(t, \bar{a}) R(t, \bar{a}) = (1 - \boldsymbol{\eta(t)}) g_J(t, \bar{a}) J(t, \bar{a}) \\ \\ S(t, \bar{a}_i+) = (1 - \boldsymbol{\vartheta}_i(t)) S(t, \bar{a}_i-) \end{array} \right.$$

$$\mathcal{C}_J = \iint \mathcal{C}_J(t, a, J(t, a)) da dt$$

$$\mathcal{C}_S = \iint \mathcal{C}_S(t, a, S(t, a)) da dt$$

$$\mathcal{C}_R = \iint \mathcal{C}_R(t, a, R(t, a)) da dt$$

$$\mathcal{I} = \int \sum_i \boldsymbol{\vartheta}_i(t) P_i(t) S(t, \bar{a}_i-) dt$$

$$\mathcal{P} = \mathcal{I} - (\mathcal{C}_J + \mathcal{C}_S + \mathcal{C}_R)$$



# Management of a Biological Resource

$$\left\{ \begin{array}{l} \partial_t J + \partial_a (g_J(t, a) J) = d_J(t, a) J \\ \partial_t S + \partial_a (g_S(t, a) S) = d_S(t, a) S \\ \partial_t R + \partial_a (g_R(t, a) R) = d_R(t, a) R \\ \\ g_J(t, 0) J(t, 0) = \int w(\alpha) R(t, \alpha) d\alpha \\ g_S(t, \bar{a}) S(t, \bar{a}) = \eta(\mathbf{t}) g_J(t, \bar{a}) J(t, \bar{a}) \\ g_R(t, \bar{a}) R(t, \bar{a}) = (1 - \eta(\mathbf{t})) g_J(t, \bar{a}) J(t, \bar{a}) \\ \\ S(t, \bar{a}_i+) = (1 - \vartheta_i(\mathbf{t})) S(t, \bar{a}_i-) \end{array} \right.$$

$$\mathcal{C}_J = \iint \mathcal{C}_J(t, a, J(t, a)) da dt$$

$$\mathcal{C}_S = \iint \mathcal{C}_S(t, a, S(t, a)) da dt$$

$$\mathcal{C}_R = \iint \mathcal{C}_R(t, a, R(t, a)) da dt$$

$$\mathcal{I} = \int \sum_i \vartheta_i(\mathbf{t}) P_i(t) S(t, \bar{a}_i-) dt$$

$$\mathcal{P} = \mathcal{I} - (\mathcal{C}_J + \mathcal{C}_S + \mathcal{C}_R)$$

$\mathcal{P}$  is Gateaux  
differentiable  
w.r.t.  $\eta$  and  $\vartheta$

(Colombo & Garavello: ESAIM – COCV, 2017)

# Management of a Biological Resource

How to find the optimal control?  
maximal profit?

# Management of a Biological Resource

How to find the optimal control?  
maximal profit?

Gâteaux differentiability  $\Rightarrow$  gradient methods

# Management of a Biological Resource

How to find the optimal control?  
maximal profit?

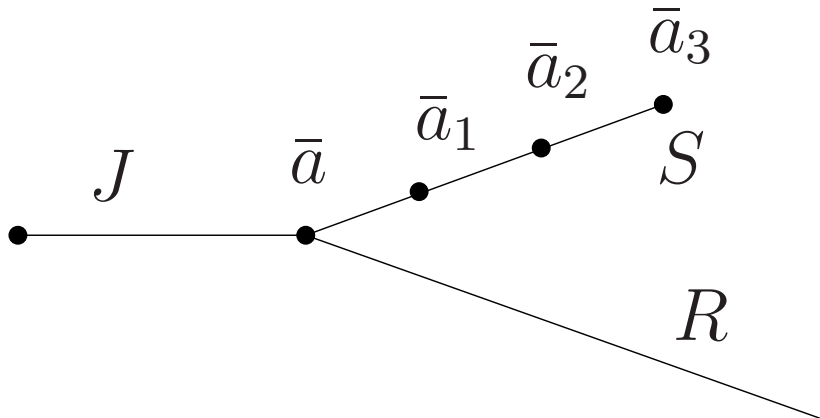
Bang–Bang controls

Theorem (Colombo & Garavello: ESAIM – COCV, 2017)

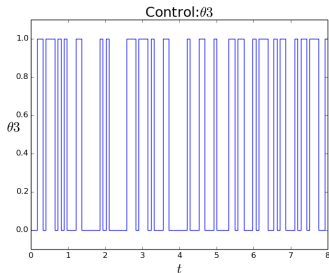
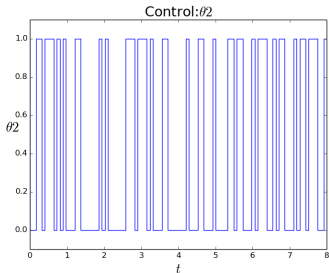
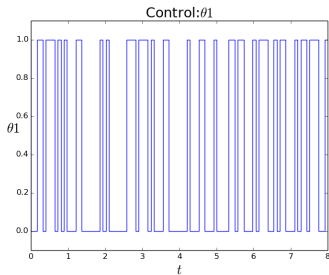
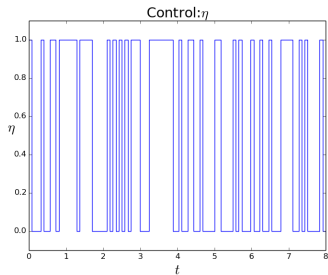
*For any  $\varepsilon > 0$  there exists a bang–bang control  $(\eta_\varepsilon, \vartheta_\varepsilon)$  such that*

$$\mathcal{P}(\eta_\varepsilon, \vartheta_\varepsilon) \geq \sup \mathcal{P}(\eta, \vartheta) - \varepsilon$$

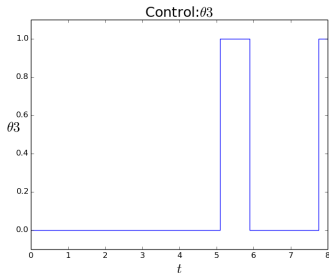
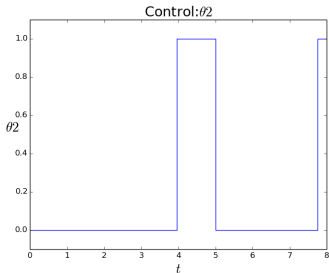
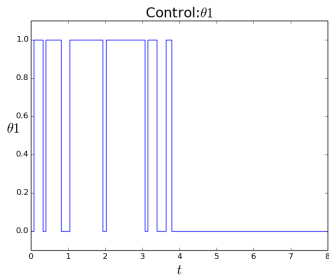
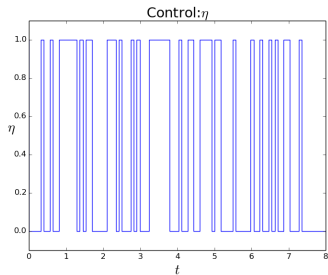
# Management of a Biological Resource



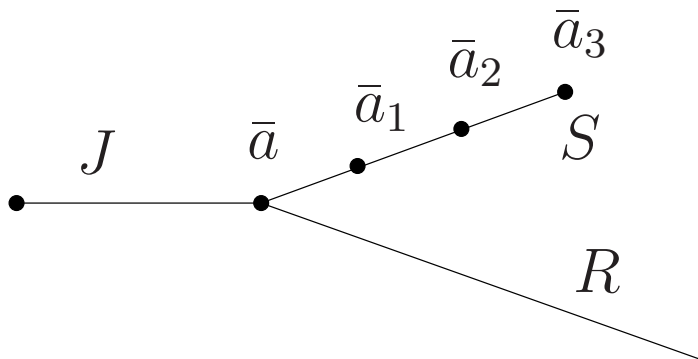
# Management of a Biological Resource



# Management of a Biological Resource



# Management of a Biological Resource





# Management of a Biological Resource

$$\left\{ \begin{array}{l} \partial_t J + \partial_a (g_J(t, a) J) = d_J(t, a) J \\ \partial_t S + \partial_a (g_S(t, a) S) = d_S(t, a) S \\ \partial_t R + \partial_a (g_R(t, a) R) = d_R(t, a) R \end{array} \right.$$

$$\left\{ \begin{array}{l} g_J(t, 0) J(t, 0) = \int w(\alpha) R(t, \alpha) d\alpha \\ g_S(t, \bar{a}) S(t, \bar{a}) = \eta(\mathbf{t}) g_J(t, \bar{a}) J(t, \bar{a}) \\ g_R(t, \bar{a}) R(t, \bar{a}) = (1 - \eta(\mathbf{t})) g_J(t, \bar{a}) J(t, \bar{a}) \end{array} \right.$$

$$C_J = \iint c_J(t, a) J(t, a) da dt$$

$$C_S = \iint c_S(t, a) S(t, a) da dt$$

$$C_R = \iint c_R(t, a) R(t, a) da dt$$

$$\mathcal{I} = \int \sum_i \vartheta_i(\mathbf{t}) P_i(t) S(t, \bar{a}_i -) dt$$

$$\mathcal{P} = \mathcal{I} - (C_J + C_S + C_R)$$

Theorem (Colombo & Garavello: Nonlinear An. RWA, 2017)

If  $\eta(\mathbf{t}) = \sum_i \eta_i \chi_{[\tau_{i-1}, \tau_i]}$  and  $\vartheta(\mathbf{t}) = \sum_i \vartheta_i \chi_{[\tau_{i-1}, \tau_i]}$ , then

# Management of a Biological Resource

$$\left\{ \begin{array}{l} \partial_t J + \partial_a (g_J(t, a) J) = d_J(t, a) J \\ \partial_t S + \partial_a (g_S(t, a) S) = d_S(t, a) S \\ \partial_t R + \partial_a (g_R(t, a) R) = d_R(t, a) R \end{array} \right.$$

$$\left\{ \begin{array}{l} g_J(t, 0) J(t, 0) = \int w(\alpha) R(t, \alpha) d\alpha \\ g_S(t, \bar{a}) S(t, \bar{a}) = \eta(\mathbf{t}) g_J(t, \bar{a}) J(t, \bar{a}) \\ g_R(t, \bar{a}) R(t, \bar{a}) = (1 - \eta(\mathbf{t})) g_J(t, \bar{a}) J(t, \bar{a}) \end{array} \right.$$

$$\mathcal{C}_J = \iint c_J(t, a) J(t, a) da dt$$

$$\mathcal{C}_S = \iint c_S(t, a) S(t, a) da dt$$

$$\mathcal{C}_R = \iint c_R(t, a) R(t, a) da dt$$

$$\mathcal{I} = \int \sum_i \vartheta_i(\mathbf{t}) P_i(t) S(t, \bar{a}_i -) dt$$

$$\mathcal{P} = \mathcal{I} - (\mathcal{C}_J + \mathcal{C}_S + \mathcal{C}_R)$$

**Theorem** (Colombo & Garavello: Nonlinear An. RWA, 2017)

If  $\eta(t) = \sum_i \eta_i \chi_{[\tau_{i-1}, \tau_i]}$  and  $\vartheta(t) = \sum_i \vartheta_i \chi_{[\tau_{i-1}, \tau_i]}$ , then

$\mathcal{P}$  is a **polynomial** in  $\eta_1, \eta_2, \dots, \vartheta_1, \vartheta_2, \dots$

(+ bounds on the degree)

## Management of a Biological Resource – Example

$$g_J(t, a) = 1.00 \quad d_J(t, a) = 1.50 \quad c_J(t, a) = 0.25 \quad J_o(a) = 1.00$$

$$g_S(t, a) = 1.00 \quad d_S(t, a) = 0.50 \quad c_S(t, a) = 0.00 \quad S_o(a) = 0.00$$

$$g_R(t, a) = 1.00 \quad d_R(t, a) = 0.75 \quad c_R(t, a) = 0.00 \quad R_o(a) = 0.00$$

$$\bar{a} = 1.00 \quad \bar{a}_1 = 1.50 \quad N = 1$$

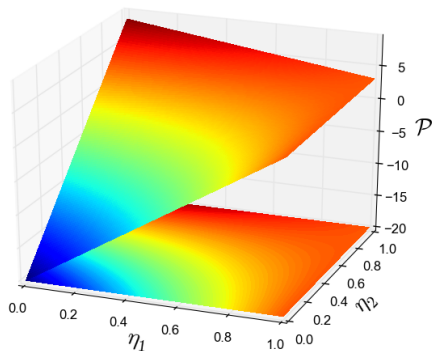
$$p(a) = 0.00 \quad p_1(t) = 8.00 \quad w(a) = 120.00 \chi_{[1.00, 4.00]}(a)$$

## Management of a Biological Resource – Example

$$\begin{array}{llll} g_J(t, a) = 1.00 & d_J(t, a) = 1.50 & c_J(t, a) = 0.25 & J_o(a) = 1.00 \\ g_S(t, a) = 1.00 & d_S(t, a) = 0.50 & c_S(t, a) = 0.00 & S_o(a) = 0.00 \\ g_R(t, a) = 1.00 & d_R(t, a) = 0.75 & c_R(t, a) = 0.00 & R_o(a) = 0.00 \\ \bar{a} = 1.00 & \bar{a}_1 = 1.50 & N = 1 & \\ p(a) = 0.00 & p_1(t) = 8.00 & w(a) = 120.00 \chi_{[1.00, 4.00]}(a) & \end{array}$$

$$\eta = \eta_1 \chi_{[0,1]} + \eta_2 \chi_{[1,2]}$$

$$\begin{aligned} \mathcal{P} = & -19.97 + 23.10 \eta_1 \\ & + 28.18 \eta_2 - 28.18 \eta_1 \eta_2 \end{aligned}$$

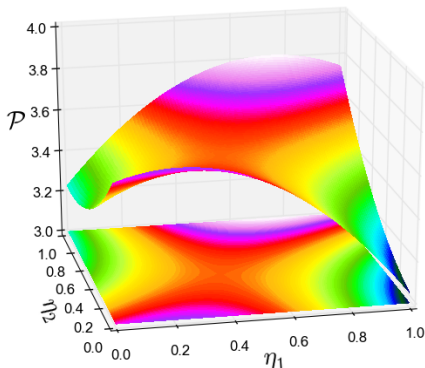


## Management of a Biological Resource – Example

$$\begin{array}{lllll} \bar{a} = 1.00 & d_J(t, a) = 0.50 & c_J(t, a) = 0.25 & p(a) = 1.00 & J_o(a) = 1.00 \\ N = 1 & d_S(t, a) = 1.00 & c_S(t, a) = 0.25 & p_1(t) = 8.20 & S_o(a) = 0.00 \\ \bar{a}_1 = 1.50 & d_R(t, a) = 1.50 & c_R(t, a) = 0.25 & w(a) = 10.00 & R_o(a) = 0.00 \end{array}$$

$$\eta = \eta_1 \chi_{[0,0.5]} + \eta_2 \chi_{[0.5,1]} + \eta_1 \chi_{[1,1.5]} + \eta_2 \chi_{[1.5,2]}$$

$$\mathcal{P}(\eta_1, \eta_2) = 3.65 + 0.46 \eta_1 - 0.88 \eta_2 + 1.11 \eta_1 \eta_2 - 1.06 \eta_1^2 + 0.46 \eta_2^2$$



R.M. Colombo, M. Garavello

*Polynomial Profit in Renewable Resources Management*

Nonlinear Analysis RWA, 37, 374-386, 2017

R.M. Colombo, M. Garavello

*Control of Biological Resources on Graphs*

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*Stability and Optimization in Structured Population Models on Graphs*

Mathematical Biosciences and Engineering, 12, 2, 311-335, 2015