

Development of a MIP for the improvement on efficiency in management plans of threatened species through the use of sensitivity curves

Diego Sfeir¹, Virgilio Hermoso², Eduardo Álvarez-Miranda³,
Jordi Garcia-Gonzalo², Andrés Weintraub¹

1 Department of Industrial Engineering, Universidad de Chile and Complex Engineering Systems Institute, Santiago, Chile

2 Forest Sciences Center of Catalonia, Solsona, España

3 Department of Industrial Engineering, Universidad de Talca, Curicó, Chile

March 6, 2019

Outline

- 1 Introduction
 - Motivation
 - Problem description
- 2 Methods
 - Case study
 - Conceptual frame
 - MIP formulation and others explanations
 - Experiment description
- 3 Results
 - First experiment
 - Second experiment
- 4 Conclusions
 - Discussion
 - Closure

Section 1

Introduction

Why to study this?

Why to study this?

- **Limited** available **resources** for ecological conservation

Why to study this?

- **Limited** available **resources** for ecological conservation
- Urgent need to secure the persistence of **species**, with specific **actions** to abate the **threats** that affect them

Why to study this?

- **Limited** available **resources** for ecological conservation
- Urgent need to secure the persistence of **species**, with specific **actions** to abate the **threats** that affect them
- Typically, a **binary** approach is used, where a specific action is taken and assumed to eliminate the threat [Auerbach et al., 2014]



What is the problem?

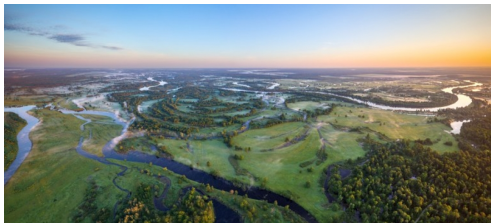
- How to generate a efficient management plan for simultaneous conservation actions?

What is the problem?

- How to generate a efficient management plan for simultaneous conservation actions?
- The fundamental objective is:
 - Determine the most **efficient actions**, in each planning unit, to achieve ecological **targets** at the **minimum cost**

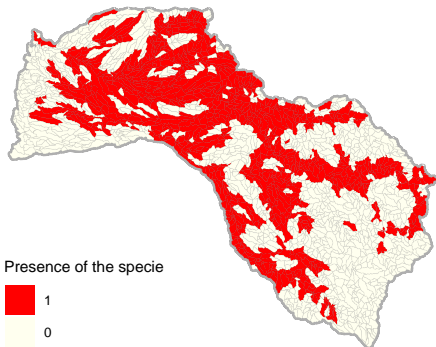
What is the problem?

- How to generate a efficient management plan for simultaneous conservation actions?
- The fundamental objective is:
 - Determine the most **efficient actions**, in each planning unit, to achieve ecological **targets** at the **minimum cost**
- In addition, some minimum **spatial conditions** are required, such as the connection of selected units and actions performed



Preliminary definitions - 1

a. Distribution of a species



b. Distribution of a threat

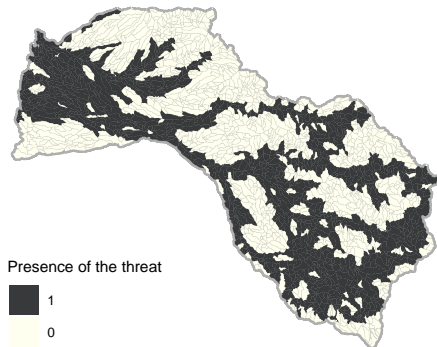


Figure: Graphical representation of the spatial distribution of a fish species (a) and of a major threat (b)

Preliminary definitions - 2

c. Distribution of a threatened species

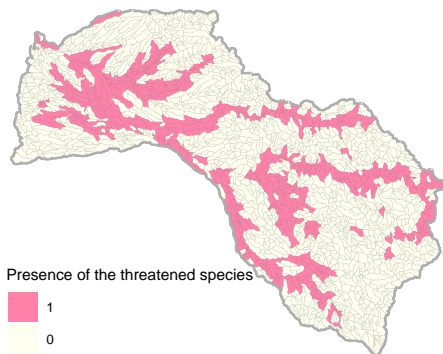
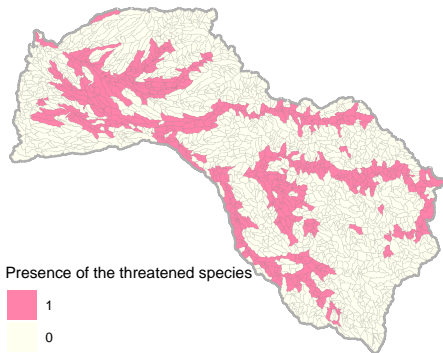


Figure: Graphical representation of the spatial distribution of a threatened fish species.

Preliminary definitions - 2

c. Distribution of a threatened species



- In a multi-action planning for threat management:

Figure: Graphical representation of the spatial distribution of a threatened fish species.

Preliminary definitions - 2

c. Distribution of a threatened species

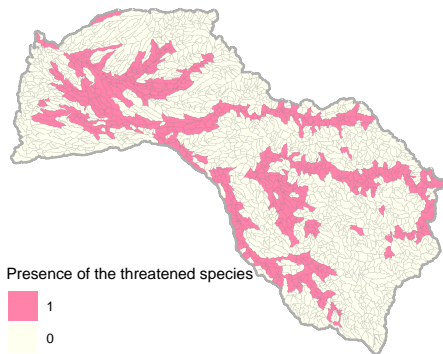


Figure: Graphical representation of the spatial distribution of a threatened fish species.

- In a multi-action planning for threat management:
- Each planning unit, could have a set of species and an other one of threats [Cattarino et al., 2015]

Preliminary definitions - 2

c. Distribution of a threatened species

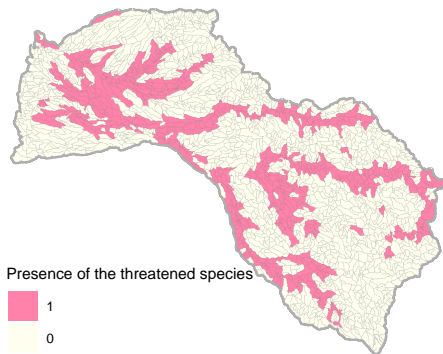


Figure: Graphical representation of the spatial distribution of a threatened fish species.

- In a multi-action planning for threat management:
- Each planning unit, could have a set of species and an other one of threats [Cattarino et al., 2015]
- Where and what conservation actions should be done to minimize the cost of management?

Section 2

Methods

Case study: Mitchell River catchment, northern Australia

- Total area: **71,630 km²**, divided into **2316 planning units** (hydrologically defined sub-catchment)

Case study: Mitchell River catchment, northern Australia

- Total area: **71,630 km²**, divided into **2316 planning units** (hydrologically defined sub-catchment)
- **45 freshwater fish species** in the study area

Case study: Mitchell River catchment, northern Australia

- Total area: **71,630 km²**, divided into **2316 planning units** (hydrologically defined sub-catchment)
- **45 freshwater fish species** in the study area
- **4 important threats** in the catchment are studied



Case study: Mitchell River catchment, northern Australia

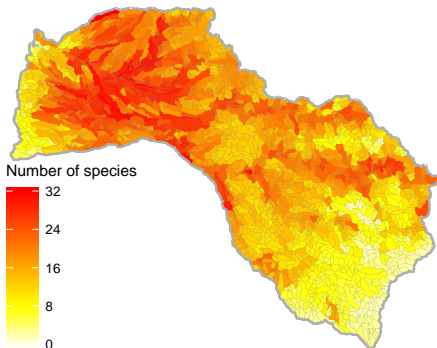
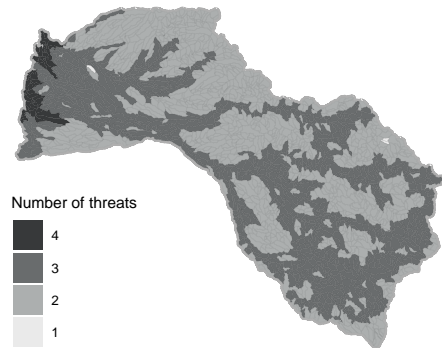
a. Species distribution**b. Threats distribution**

Figure: Graphical representation of the spatial distributions of the 45 species of fish (a) and of the four major threats (b) analyzed in the case study

Threats

- **4 major threats:**

- Water buffalo (*Bubalis bubalis*)
- Cane toad (*Bufo marinus*)
- River flow alteration (caused by impoundments, channels for water extractions and levee banks)
- Grazing land use



Actions

- **4 conservations actions for remediating each specific threat:**
 - Shooting for buffalo control
 - Chemical or biological treatment for cane toad control
 - Removal of dams or redesign of levee banks for flow-regime restoration
 - Pasture fencing and stewardship programs for grazing management



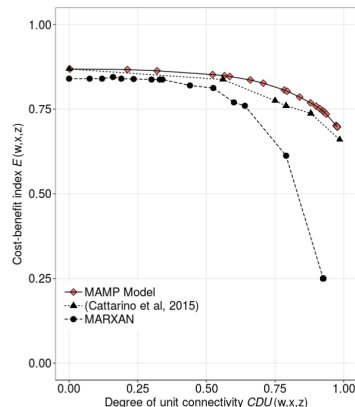
Monitoring and land adquisition

- **Before implementing each specific action:**
 - We need to buy the planning unit
 - We need to monitor if the threat is present in that unit.



What has been done so far

- Marxan [Watts et al., 2009] and Marxan with zones [Levin et al., 2013]
- An algorithm based on Simulated Annealing [Cattarino et al., 2015].
- A MIP approach for multi-action planning for threat management [Salgado & Alvarez-Miranda et al., 2018, submitted to Elsevier]



Adding another dimension

- Normally, a **binary** approach is used
 - Whether we implement an **action** or not
- **Continuous responses** by species against their corresponding threats
- Different forms of curves to represent species with higher or lower levels of **sensitivity** to their threats

Levels of effort for the actions

- It is beneficial to use levels of **effort**, and we use **4 degrees**
 - i Zero level effort
 - ii 33% effort (low)
 - iii 66% effort (intermediate)
 - iv 100% effort (high)
- A lower number of levels requires a **smaller amount of data**, and therefore the implementation of the model becomes **simpler** and more attractive.

Response curves

Two dimensions:

Response curves

Two dimensions:

- Species' probability of persistence
- **Threat intensity:**
 - i For the **binary** scenario, it is whether the threat is **present** or not
 - ii For the **continuous** responses, it is the **degree** of the remaining threat

Response curves

Two dimensions:

- Species' probability of persistence
- **Threat intensity:**
 - i For the **binary** scenario, it is whether the threat is **present** or not
 - ii For the **continuous** responses, it is the **degree** of the remaining threat

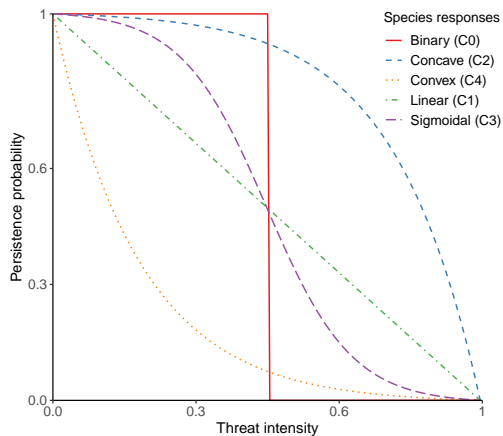


Figure: Graphical representation of the species persistence probability according to the threat intensity

Response curves - Transitivity

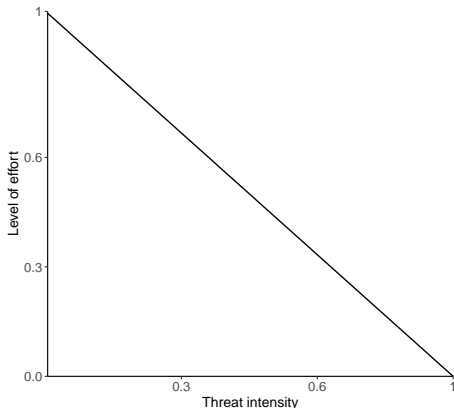


Figure: Graphical representation of the level of effort made according to the remaining threat intensity

Response curves - Transitivity

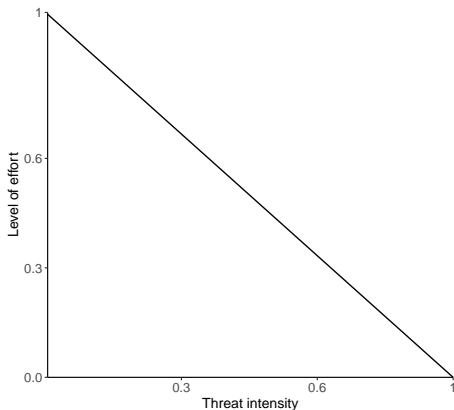


Figure: Graphical representation of the level of effort made according to the remaining threat intensity

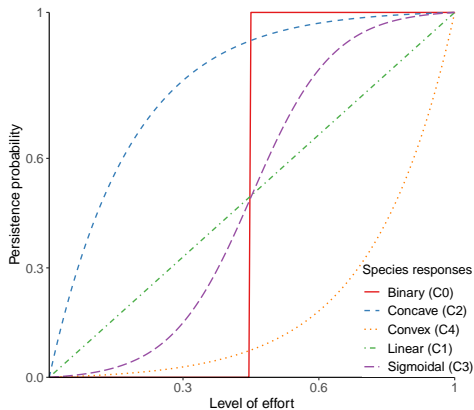


Figure: Graphical representation of the species persistence probability according to the level of effort made

Some parameters

- **Monitoring cost** of a planning unit (c_i)

Some parameters

- **Monitoring cost** of a planning unit (c_i)
- **Action Cost** at a specific level of effort in a planning unit ($c_{i,k}^n$)

Some parameters

- **Monitoring cost** of a planning unit (c_i)
- **Action Cost** at a specific level of effort in a planning unit ($c_{i,k}^n$)
- **Surface area** α_i in square kilometers

Some parameters

- **Monitoring cost** of a planning unit (c_i)
- **Action Cost** at a specific level of effort in a planning unit ($c_{i,k}^n$)
- **Surface area** α_i in square kilometers
- **Ecological targets** of preservation for each species, measured in square kilometers (T_s)

Some parameters

- **Monitoring cost** of a planning unit (c_i)
- **Action Cost** at a specific level of effort in a planning unit ($c_{i,k}^n$)
- **Surface area** α_i in square kilometers
- **Ecological targets** of preservation for each species, measured in square kilometers (T_s)
- **Penalty factor** for the selected **units fragmentation** (β_u) and for the **actions fragmentation** (β_a)

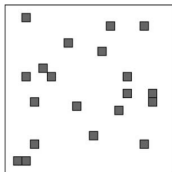


Figure: Fragmentated units

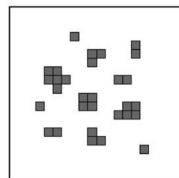


Figure: Connected units

Some parameters

- For the **two costs** involved, we used that the greater the area to be monitored or treated, the greater the associated cost.

Some parameters

- For the **two costs** involved, we used that the greater the area to be monitored or treated, the greater the associated cost.
- **Actions costs** $c_{i,k}^n$:
 - i For a **null** effort level, a cost equal to **zero** is associated
 - ii For the **low** level, a 33% from the area α_i
 - iii For the **intermediate** level, a 66% of the area α_i
 - iv For the **high** level, the total surface area α_i
- To adjust to the real costs, the **monitoring cost** c_j was reduced to a value of 20% of the surface area α_j

MIP formulation of the effort management planning problem, oriented to multi-conservation actions

Two **decision** variables are defined to solve the problem:

MIP formulation of the effort management planning problem, oriented to multi-conservation actions

Two **decision** variables are defined to solve the problem:

- | w_i : **binary** variable that indicates if the planning **unit** i is selected to be **monitored**, and to be part of the reserve

MIP formulation of the effort management planning problem, oriented to multi-conservation actions

Two **decision** variables are defined to solve the problem:

- I w_i : **binary** variable that indicates if the planning **unit i** is selected to be **monitored**, and to be part of the reserve
- II $x_{i,k}^n$: **binary** variable that determines whether or not an **action k** is performed, in planning **unit i** and in intensity **level n**



MIP formulation of the effort management planning problem, oriented to multi-conservation actions

$$\min \sum_{i \in I} w_i c_i + \sum_{i \in I} \sum_{k \in K_i} \sum_{n \in N} x_{i,k}^n c_{i,k}^n + \beta_u \sum_{i \in I} \sum_{\substack{j \in I: \\ \exists d_{i,j} > 0}} w_i (1 - w_j) \frac{1}{d_{i,j}^2} \quad (1)$$

$$\text{s.t. } \sum_{n \in N} x_{i,k}^n = w_i, \quad \forall k \in K_i, \forall i \in I \quad (2)$$

$$\sum_{n \in N} x_{i,k}^n = 0, \quad \forall k \notin K_i, \forall i \in I \quad (3)$$

$$\sum_{\substack{i \in I_s: \\ |K_i| > 0}} B_{i,s} \alpha_i + \sum_{\substack{i \in I_s: \\ |K_i| = 0}} F_{i,s} \alpha_i \geq T_s, \quad \forall s \in S \quad (4)$$

$$w_i, x_{i,k}^n \in \{0, 1\}, \quad \forall i \in I, \forall k \in K, \forall n \in N \quad (5)$$

$$F_{i,s} \in [0, 1], \quad \forall i \in I_s, \forall s \in S \quad (6)$$

Extended model

- It is practical to think about **gathering the actions** implemented within the selected planning units
- A new **penalty** associated with fragmentated **action** decisions

Extended model

- It is practical to think about **gathering the actions** implemented within the selected planning units
- A new **penalty** associated with fragmentated **action** decisions

$$\beta_a \sum_{i \in I} \sum_{\substack{j \in I: \\ \exists d_{i,j} > 0}} \sum_{k \in K_i \cap K_j} \sum_{n \in N \setminus \{0\}} x_{i,k}^n (1 - \sum_{m \in N \setminus \{0\}} x_{j,k}^m) \frac{1}{d_{i,j}^2}$$

Description of the experiment

| Name of the parameters | Assigned values |
|---|--|
| Contribution target to the ecological benefit (T_s) | {110, 220, 440, 880, 1760, 3520, 7040, 14080, 24160} km ² |
| Unit fragmentation penalty factor (β_u) | {15} |
| Action fragmentation penalty factor (β_a) | {0, 3, 7, 11, 15, 19, 23} |

Table: Summary of the parameters used to solve the experiments for the case study of the Mitchell River catchment

Composition of the functions used for the responses

| Curve type | Associated function $r_{k,s}^n(x)$ |
|------------|---|
| Binary | 0 if $x < 1$, 1 if $x = 1$ |
| Concave | $\frac{1-e^{ax}}{1-e^a}$ |
| Convex | $\frac{e^{ax}-1}{e^a-1}$ |
| Linear | x |
| Sigmoidal | $\frac{(1+e^{-b(x-0.5)})^{-1}-(1+e^{0.5b})^{-1}}{(1+e^{-0.5b})^{-1}-(1+e^{0.5b})^{-1}}$ |

Nota 1: x is used instead of $\sum_{n \in N} x_{i,k}^n e_n$.

Nota 2: Parameters a and b are constant.

Table: Summary of the functions used to represent the response curves, which relate the level of effort chosen to a probability of persistence associated with each species

Definition of indicators

- The indicators calculated for the two parts of the experiment:

Definition of indicators

- The indicators calculated for the two parts of the experiment:
 - i "Cost-benefit ratio": ratio of total **protected area** and real **cost** of the management plan. ($\frac{km^2}{\$}$)

Definition of indicators

- The indicators calculated for the two parts of the experiment:
 - i "Cost-benefit ratio": ratio of total **protected area** and real **cost** of the management plan. ($\frac{km^2}{\$}$)
 - ii "Actions connectivity degree": division between the **average fragmentation** of each action k and the theoretical **maximum fragmentation** for each action k.

Section 3

Results

Resolution time

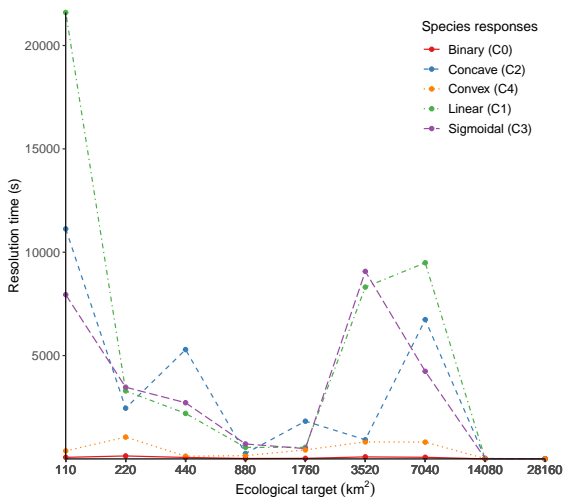


Figure: Graphical representation of the resolution times obtained by solving the instances related to the five response curves

GAP

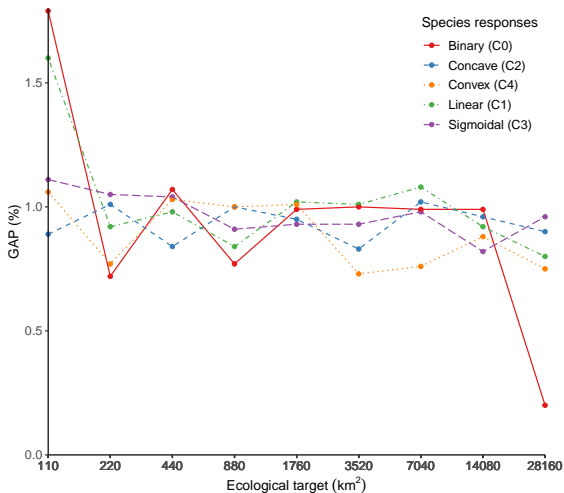


Figure: Graphical representation of the GAPs obtained by solving the instances related to the five response curves

First experiment

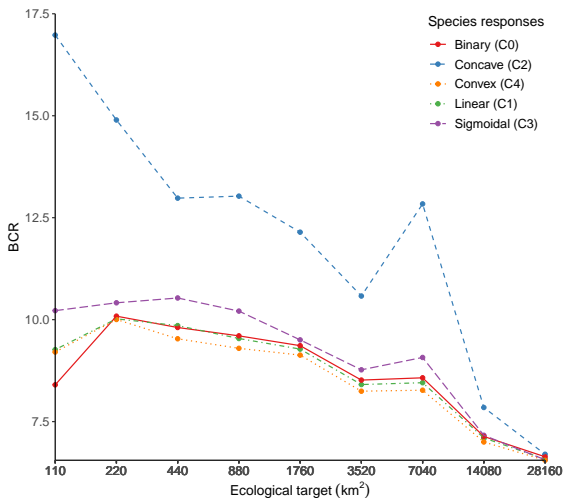
Benefit-Cost Ratio ($\frac{km^2}{\$}$)

Figure: Graphical representation of the BCR obtained by solving the instances related to the five response curves

Differences through the illustration of maps - 1

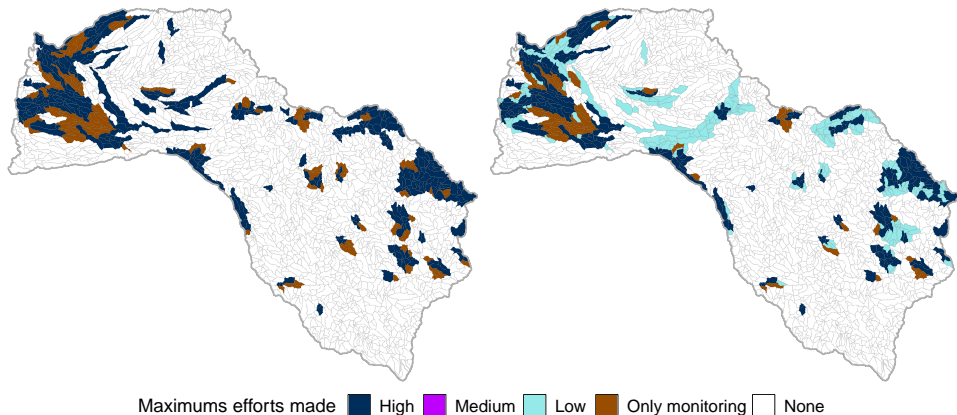


Figure: Graphical representation of the maximum efforts spatial distribution, for an objective of 3520 km², using a degree of global connectivity equal to 15, a degree of connectivity of actions equal to 11 and evaluated for the **binary** and **concave** responses

Differences through the illustration of maps - 2

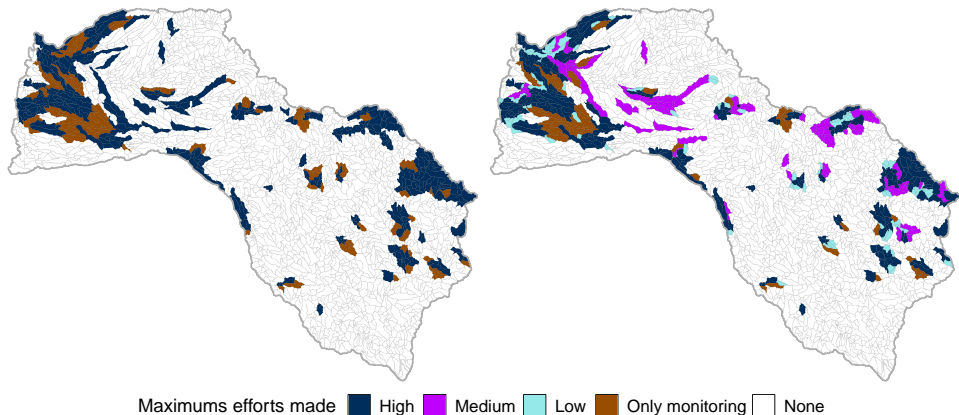


Figure: Graphical representation of the maximum efforts spatial distribution, for an objective of 3520 km², using a degree of global connectivity equal to 15, a degree of connectivity of actions equal to 11 and evaluated for the **binary** and **sigmoidal** responses

Second experiment

- To analyze the behavior of the costs associated with each response curve when the degree of penalty is increased (Actions fragmentation)
- Seven values for the degrees of penalty between 0 and 23
- Ecological target of the species equal to 3520 km²

Resolution time

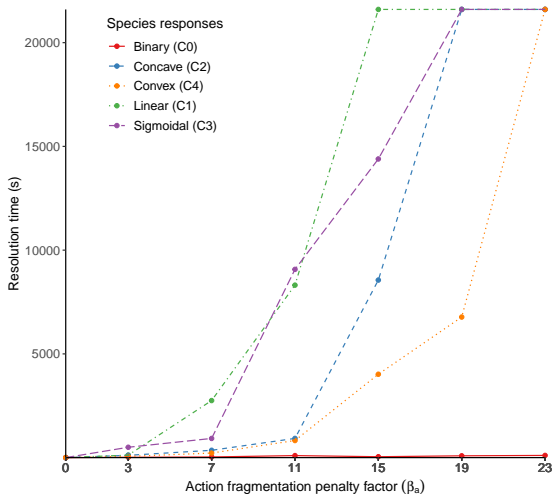


Figure: Graphical representation of the resolution times obtained when solving the instances for the actions fragmentation penalty factor β_a

Resolution time and GAP

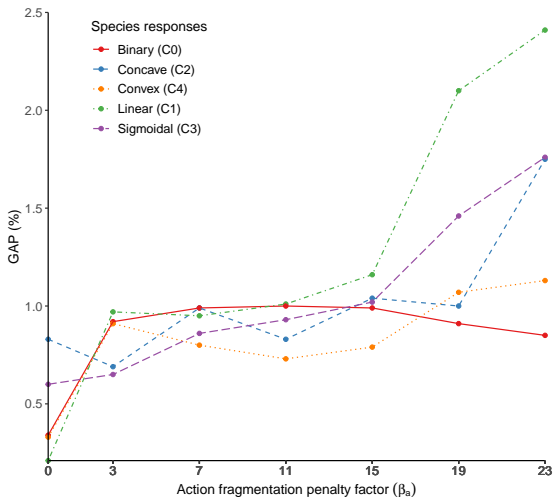


Figure: Graphical representation of the GAPs obtained when solving the instances for the actions fragmentation penalty factor β_a

Benefit-Cost Ratio

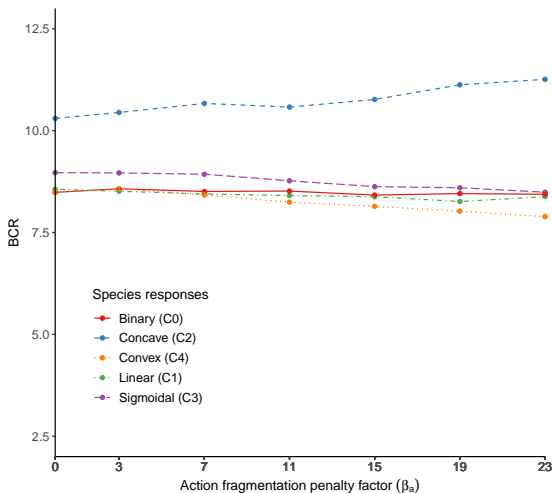


Figure: Graphical representation of the BCR obtained when solving the instances for the actions fragmentation penalty factor β_a

Actions connectivity degree

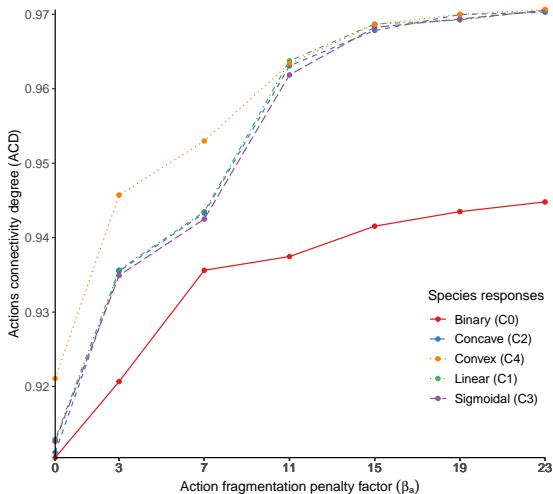
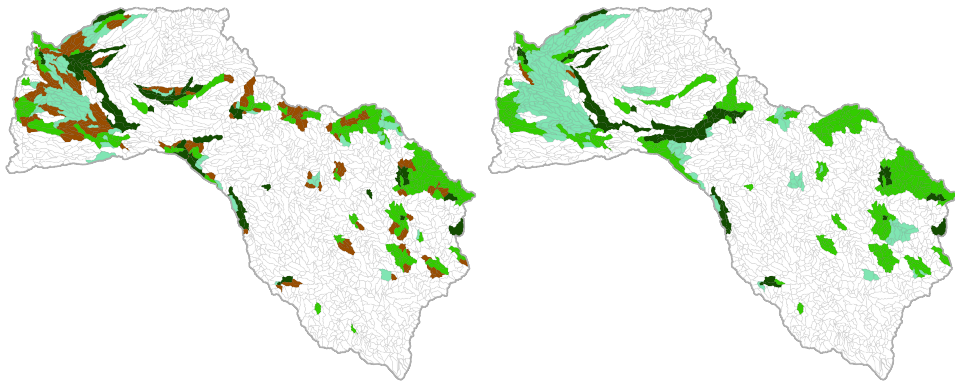


Figure: Graphical representation of the ACD obtained when solving the instances for the actions fragmentation penalty factor β_a

Second experiment

Connectivity differences through the illustration of maps - 1

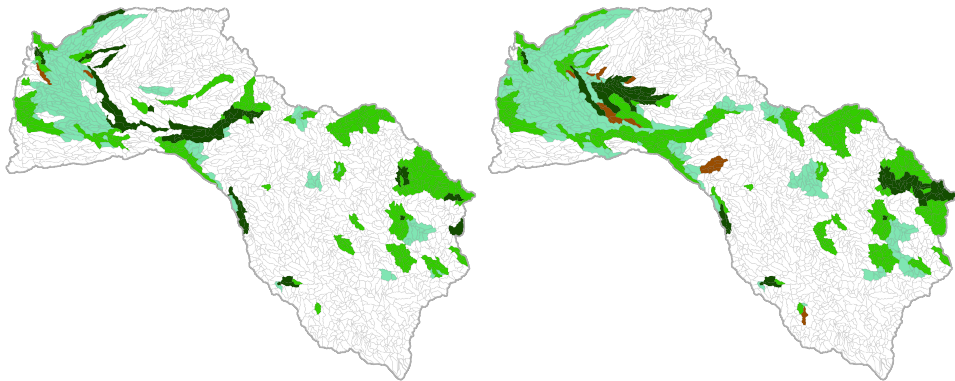
a. Case with $\beta_a = 0$.b. Case with $\beta_a = 23$.

Interventions made 3 actions 2 actions 1 action Only monitoring None

Figure: Graphical representation of the interventions spatial distribution, for an objective of 3520 km² and a degree of global connectivity of 15

Second experiment

Connectivity differences through the illustration of maps - 2

b. Case with $\beta_a = 23$.c. Case with $\beta_a = 50$.

Interventions made 3 actions 2 actions 1 action Only monitoring None

Figure: Graphical representation of the interventions spatial distribution, for an objective of 3520 km² and a degree of global connectivity of 15

Section 4

Conclusions

Scope and limitations of the study

- **Mixed integer optimization** model, which finds sites and **levels** of efficient **efforts** of conservation actions, considering **multiple threats** and relying on the different **response curves** of the species

Scope and limitations of the study

- **Mixed integer optimization** model, which finds sites and **levels** of efficient **efforts** of conservation actions, considering **multiple threats** and relying on the different **response curves** of the species
- MIP gives **flexibility** to eventually modify the objective function, incorporate new restrictions and perform sensitivity analysis to measure the impacts of each parameter on the indicators of the management plan

Scope and limitations of the study

- **Mixed integer optimization** model, which finds sites and **levels** of efficient **efforts** of conservation actions, considering **multiple threats** and relying on the different **response curves** of the species
- MIP gives **flexibility** to eventually modify the objective function, incorporate new restrictions and perform sensitivity analysis to measure the impacts of each parameter on the indicators of the management plan
- Due to the type of problem faced, each of these improvements requires the **incorporation** of a large number of **variables** and **equations** that lead to **exponentially** resolution times

Analysis of how response curves affect efficiency

- Gains in **efficiency** (RCB) depending on the **shape** of the response curve. Sigmoidal and concave response curves can achieve targets using **lower effort** levels

Analysis of how response curves affect efficiency

- Gains in **efficiency** (RCB) depending on the **shape** of the response curve. Sigmoidal and concave response curves can achieve targets using **lower effort** levels
- When considering a convex curve, the model loses a bit of efficiency by virtue of **connecting** more the conservation **actions** exercised

Analysis of how response curves affect efficiency

- Gains in **efficiency** (RCB) depending on the **shape** of the response curve. Sigmoidal and concave response curves can achieve targets using **lower effort** levels
- When considering a convex curve, the model loses a bit of efficiency by virtue of **connecting** more the conservation **actions** exercised
- The binary case generate **high values** of penalty for actions **fragmentation**, and the value of the objective function is greater than for the convex case

Principal conclusions

- Not significant differences on **efficiency** when the responses to actions are **convex** and **linear**, as can be expected

Principal conclusions

- Not significant differences on **efficiency** when the responses to actions are **convex** and **linear**, as can be expected
- **Benefits** on efficiency when we considered a **sigmoidal** curve and a **concave** curve, making this proposal more attractive than the traditional methods

Principal conclusions

- Not significant differences on **efficiency** when the responses to actions are **convex** and **linear**, as can be expected
- **Benefits** on efficiency when we considered a **sigmoidal** curve and a **concave** curve, making this proposal more attractive than the traditional methods
- When demanding a **higher connectivity** level for the **actions**, efficiency levels remained almost the **same** and **continuous** responses achieved more **connected** solutions than the **binary**

Principal conclusions

- Not significant differences on **efficiency** when the responses to actions are **convex** and **linear**, as can be expected
- **Benefits** on efficiency when we considered a **sigmoidal** curve and a **concave** curve, making this proposal more attractive than the traditional methods
- When demanding a **higher connectivity** level for the **actions**, efficiency levels remained almost the **same** and **continuous** responses achieved more **connected** solutions than the **binary**
- We can **efficiently** identify the optimum **effort** to be allocated to **multiple interventions**, and **where** to perform them, to combat the existing **threats**

Future work

- It would be of great interest to study the **impact** of adding **uncertainty** to the responses curves
- This work can be extended by considering a new dimension of **time**, making a dinamic approach
- Finally, a future research could be oriented to analize the variation of the species **density** within each planning unit as the remaining threats are not eradicated

Development of a MIP for the improvement on efficiency in management plans of threatened species through the use of sensitivity curves

Diego Sfeir¹, Virgilio Hermoso², Eduardo Álvarez-Miranda³,
Jordi Garcia-Gonzalo², Andrés Weintraub¹

1 Department of Industrial Engineering, Universidad de Chile and Complex Engineering Systems Institute, Santiago, Chile

2 Forest Sciences Center of Catalonia, Solsona, España

3 Department of Industrial Engineering, Universidad de Talca, Curicó, Chile

March 6, 2019

References and Further Reading I

Anni Arponen, Risto K Heikkinen, Chris D Thomas, and Atte Moilanen. The value of biodiversity in reserve selection: representation, species weighting, and benefit functions. *Conservation Biology*, 19(6):2009–2014, 2005.

Nancy A Auerbach, Ayesha IT Tulloch, and Hugh P Possingham. Informed actions: where to cost effectively manage multiple threats to species to maximize return on investment. *Ecological Applications*, 24(6):1357–1373, 2014.

Hawthorne L Beyer, Yann Dujardin, Matthew E Watts, and Hugh P Possingham. Solving conservation planning problems with integer linear programming. *Ecological Modelling*, 328:14–22, 2016.

References and Further Reading II

Alain Billionnet. *Optimisation discrète, De la modélisation à la résolution par des logiciels de programmation mathématique*. Dunod, 2007.

Alain Billionnet. Mathematical optimization ideas for biodiversity conservation. *European Journal of Operational Research*, 231(3): 514–534, 2013.

Stuart HM Butchart, Matt Walpole, Ben Collen, Arco Van Strien, Jörn PW Scharlemann, Rosamunde EA Almond, Jonathan EM Baillie, Bastian Bomhard, Claire Brown, John Bruno, et al. Global biodiversity: indicators of recent declines. *Science*, 328 (5982):1164–1168, 2010.

References and Further Reading III

Josie Carwardine, Carissa J Klein, Kerrie A Wilson, Robert L Pressey, and Hugh P Possingham. Hitting the target and missing the point: target-based conservation planning in context. *Conservation Letters*, 2(1):4–11, 2009.

Josie Carwardine, Trudy O'Connor, Sarah Legge, Brendan Mackey, Hugh P Possingham, and Tara G Martin. Prioritizing threat management for biodiversity conservation. *Conservation Letters*, 5(3):196–204, 2012.

Lorenzo Cattarino, Virgilio Hermoso, Josie Carwardine, Mark J Kennard, and Simon Linke. Multi-action planning for threat management: a novel approach for the spatial prioritization of conservation actions. *PloS one*, 10(5):e0128027, 2015.

References and Further Reading IV

- Lorenzo Cattarino, Virgilio Hermoso, Lindsay W Bradford, Josie Carwardine, Kerrie A Wilson, Mark J Kennard, and Simon Linke. Accounting for continuous species' responses to management effort enhances cost-effectiveness of conservation decisions. *Biological conservation*, 197:116–123, 2016.
- Edward T Game, Michael Bode, Eve McDonald-Madden, Hedley S Grantham, and Hugh P Possingham. Dynamic marine protected areas can improve the resilience of coral reef systems. *Ecology Letters*, 12(12):1336–1346, 2009.
- Virgilio Hermoso, Doug P Ward, and Mark J Kennard. Prioritizing refugia for freshwater biodiversity conservation in highly seasonal ecosystems. *Diversity and Distributions*, 19(8):1031–1042, 2013.

References and Further Reading V

Virgilio Hermoso, Stephanie R Januchowski-Hartley, and Simon Linke. Systematic planning of disconnection to enhance conservation success in a modified world. *Science of the Total Environment*, 536:1038–1044, 2015.

John G Hof and Linda A Joyce. A mixed integer linear programming approach for spatially optimizing wildlife and timber in managed forest ecosystems. *Forest science*, 39(4):816–834, 1993.

Stephanie R Januchowski-Hartley, Vanessa M Adams, and Virgilio Hermoso. The need for spatially explicit quantification of benefits in invasive-species management. *Conservation Biology*, 32(2):287–293, 2018.

References and Further Reading VI

Noam Levin, James EM Watson, Liana N Joseph, Hedley S Grantham, Liat Hadar, Naomi Apel, Avi Perevolotsky, Niv DeMalach, Hugh P Possingham, and Salit Kark. A framework for systematic conservation planning and management of mediterranean landscapes. *Biological Conservation*, 158:371–383, 2013.

Jie Liang, Xinyue He, Guangming Zeng, Minzhou Zhong, Xiang Gao, Xin Li, Xiaodong Li, Haipeng Wu, Chunting Feng, Wenle Xing, et al. Integrating priority areas and ecological corridors into national network for conservation planning in china. *Science of The Total Environment*, 626:22–29, 2018.

Chris R Margules and Robert L Pressey. Systematic conservation planning. *Nature*, 405(6783):243, 2000.

References and Further Reading VII

Shana M McDermott, Rebecca E Irwin, and Brad W Taylor. Using economic instruments to develop effective management of invasive species: insights from a bioeconomic model. *Ecological applications*, 23(5):1086–1100, 2013.

Jeanne L Nel, Dirk J Roux, Gillian Maree, Cornelius J Kleynhans, Juanita Moolman, Belinda Reyers, Mathieu Rouget, and Richard M Cowling. Rivers in peril inside and outside protected areas: a systematic approach to conservation assessment of river ecosystems. *Diversity and Distributions*, 13(3):341–352, 2007.

Stuart L Pimm, Clinton N Jenkins, Robin Abell, Thomas M Brooks, John L Gittleman, Lucas N Joppa, Peter H Raven, Callum M Roberts, and Joseph O Sexton. The biodiversity of species and their rates of extinction, distribution, and protection. *Science*, 344(6187):1246752, 2014.

References and Further Reading VIII

- Federico Montesino Pouzols and Atte Moilanen. Roboff: software for analysis of alternative land-use options and conservation actions. *Methods in Ecology and Evolution*, 4(5):426–432, 2013.
- Brad Pusey, Danielle Warfe, Simon Townsend, Michael Douglas, Damien Burrows, Mark Kennard, and Paul Close. Condition, impacts and threats to aquatic biodiversity. In *Aquatic Biodiversity in Northern Australia: patterns, threats and future*, pages 151–172. CDU Press, 2011.
- Truly Santika, Clive A McAlpine, Daniel Lunney, Kerrie A Wilson, and Jonathan R Rhodes. Assessing spatio-temporal priorities for species' recovery in broad-scale dynamic landscapes. *Journal of Applied Ecology*, 52(4):832–840, 2015.

References and Further Reading IX

- David L Strayer and David Dudgeon. Freshwater biodiversity conservation: recent progress and future challenges. *Journal of the North American Benthological Society*, 29(1):344–358, 2010.
- Dean Urban and Timothy Keitt. Landscape connectivity: a graph-theoretic perspective. *Ecology*, 82(5):1205–1218, 2001.
- Piero Visconti and Lucas Joppa. Building robust conservation plans. *Conservation biology*, 29(2):503–512, 2015.
- Matthew E Watts, Ian R Ball, Romola S Stewart, Carissa J Klein, Kerrie Wilson, Charles Steinback, Reinaldo Lourival, Lindsay Kircher, and Hugh P Possingham. Marxan with zones: software for optimal conservation based land-and sea-use zoning. *Environmental Modelling & Software*, 24(12):1513–1521, 2009.