

Modeling the biodiversity crisis: the roles of space and market structure in bioeconomic modeling.

Modéliser la crise de la biodiversité : les rôles de l'espace et de la structure de marché dans la modélisation bioéconomique

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Introduction

Humanity is amidst a critical ecological era, whereby the limits of the earth system have been crossed. The notion of *planetary boundaries* (Rockström et al., 2009; Steffen et al., 2015) illustrates how the anthroposphere, the planetary-scale effects of human activities, have become an additional functional component and capable of changing the Earth system (Richardson et al., 2023) alongside the geosphere (energy flow and nonliving materials in Earth and atmosphere) and biosphere (all living organisms/ecosystems). The *planetary boundaries* framework identifies the limits to the impact of the anthroposphere on the Earth system that can safeguard Earth's interglacial state - the only one where civilization is known - by identifying a *safe operating space*. The boundaries concern biosphere integrity, climate change, novel entities (e.g. man made introduction to the Earth system such as chemical and material pollutants), stratospheric ozone depletion, atmospheric aerosol loading, ocean acidification, biogeochemical flows, freshwater change and land system change. Among these 9 boundaries, Richardson et al. (2023) estimate that 6 have been crossed, threatening the stability and resilience of the Earth system.

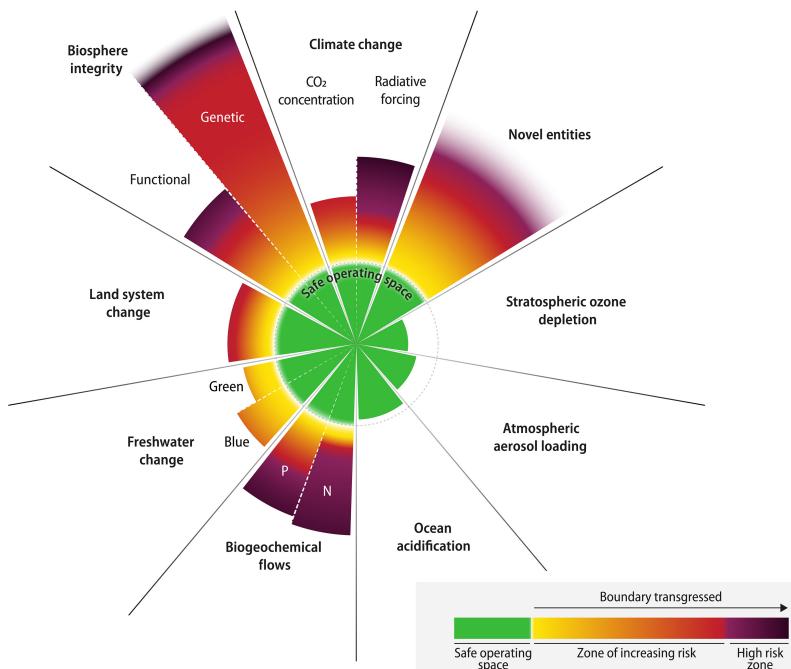


Figure 1: Current status of control variables for all nine planetary boundaries, from Richardson et al. (2023)

Among these planetary limits, the integrity of the biosphere has gradually become of particular interest, along with its interaction with other limits, such as climate change, or novel entities (including pollution). Created in 2012, the Interdisciplinary Panel on Biodiversity and Ecosystem Services (IPBES) has been raising the alarm on the state of "Nature" globally. Its chair, Sir Robert Watson, put it clearly¹:

"The overwhelming evidence of the IPBES Global Assessment² from a wide range of different field of knowledge, presents an ominous picture [...]. The health of ecosystems on which we and other species depend is deteriorating more rapidly than ever. We are eroding the foundations of our economies, livelihoods, food security, health and quality of life worldwide"

The IPBES insists on four key elements about "Nature", a central concept in its framework ([IPBES, 2019](#)):

Nature (also defined as living nature) [is] the nonhuman world, including coproduced features, with particular emphasis on living organisms, their diversity, their interactions among themselves, and with their abiotic environment. Within the framing of natural sciences, nature includes e.g. all dimensions of biodiversity, species, genotypes, populations, ecosystems, the biosphere, ecosystem functioning, communities, biomes, Earth life support's systems and their associated ecological, evolutionary, biogeochemical processes and biocultural diversity. Within the framework of economics, it includes categories such as biotic natural resources, natural capital, and natural assets. Within a wider context of social sciences and humanities and interdisciplinary environmental sciences, it is referred to with categories such as natural heritage, living environment, or the nonhuman. Within the context of other knowledge systems, it includes categories such as Mother Earth [...], Pachamama [...] ([IPBES \(2019\)](#), p.14, see also [Díaz et al. \(2015\)](#))

Nature, as defined in this approach, is a very large object. It is defined across different ontologies (living and non living) and types of interactions, at different scales (genotypes v. ecosystems), at different types of processes (biological v. ecological), and cross different fields of inquiry (natural sciences v. social sciences). In this dissertation, I study more specifically "biodiversity", which focuses on the variability among living organisms. While it is itself an ambiguous concept, biodiversity tends to isolates the functioning of living organisms, in relationship with their material, biotic and abiotic environment. Putting the focal on biodiversity allows to isolate

¹See the [press release](#) address of the 2019 report

²e.g. [IPBES \(2019\)](#)

First, it documents the drastic changes the biosphere is going through, how Nature and more specifically, biodiversity is changing, across scales, time and geographic areas. Second, it mostly considers these changes through an *anthropocentric* lense, e.g. mediating the aforementioned changes through the multiple and diverse contributions that Nature and biodiversity bring to people and how its disruption impacts human lives. Third, it highlights the role of anthropogenic (e.g. of human origin) drivers of the disruption of Nature and biodiversity. Finally, as it underpins "*our economies, livelihoods, food security health and qualify of life worldwide*", the synthesis of the available science calls for collective action and suggests policy pathways to remedy the demise of Nature and biodiversity.

This reports sets different objectives to scientific research. The first objective is to explain the feedback mechanisms : how do human livelihoods impact biodiversity? In response, how does biodiversity impact human livelihoods?. This objective involves understanding the causes and measuring the direct and indirect anthropogenic drivers of change in Nature and biodiversity on the one hand, and understanding the channels and scales through which Nature and biodiversity contribute to human livelihoods, as well as measuring these contributions. Hence, studying the demise of nature, and the potential to remedy it, calls for an integrated perspective, that joins natural sciences to social sciences, through frameworks such as *social-ecological systems* ([Ostrom, 2009](#)).

The second objective is to provide a framework to assess the desirability, the feasibility and means of implementation of collective pathways that would remedy the crisis Nature is facing. In a way, it involves designing and implementing *policy pathways* towards *sustainable futures*, e.g. finding definite courses or methods of action selected from alternatives, at the individual, collective or governmental levels, to achieve future states of the world which remain in a safe operating space regarding planetary bounds ([Rockström et al., 2009; Steffen et al., 2015](#)).

In this dissertation, I take on these two objectives using a framework stemming from economics and ecology. I study the feedback relationship between biodiversity and anthropogenic drivers of its decline, through their causes and consequences, and I analyze policy pathways to remedy this demise.

The decline of Nature

"Biodiversity" is also an ambiguous concept, which nonetheless has measurable features, which all point towards a marked decline attributable to mankind. Existing policy frameworks at different levels are geared to halt this decline.

Emergence and definition of biodiversity as a concept

Biodiversity emerged as a concept in the 1980s, along with the emergence of "conservation biology", a branch of biology concerned with the protection of "biological diversity" (Soulé, 1985), as a response to an acceleration in the loss of species that have intrinsic value, e.g. should be protected for their own sake (Soulé, 1986). As highlighted by Mouysset (2023) and Dyke and Lamb (2008), the conceptual definition of biodiversity is difficult, as it recovers a number of different dimensions. As a matter of fact, it recovers ethical and measurement concerns. Biodiversity can be viewed as "an intrinsic, value-ladden quality of natural systems that should be preserved for its own sake" (Dyke and Lamb, 2008; Mouysset, 2023), but it also refers to measurable features relevant to understanding community structure, environmental processes, and ecosystem functions. Although measurable, these features are epistemologically distinct as they measure different types of phenomena.

In the wake of the 1992 Rio United Conference on Environment and Development, the Convention on Biological Diversity emerged as an international treaty to safeguard biodiversity. In doing so, it provided an internationally agreed upon definition:

“Biological diversity” means the variability among living organisms from all sources including, inter alia, terrestrial, marine and other aquatic ecosystems and the ecological complexes of which they are part; this includes diversity within species, between species and of ecosystems.”

[Article 2 of the Convention on Biological Diversity](#)

Following Mouysset (2023), this definition implies different scales from a hierarchical perspective, at the genetic level, at the species, the community (e.g assembly of interacting species in a given area), and the ecosystem levels (defined as the interaction of communities and their abiotic environment). These levels imply different forms of measurement, including the distribution of genes, species abundance (e.g. the number of individuals), species richness (e.g. the number of different species) within communities, among communities, and across larger scales (e.g. alpha, beta and gamma diversities.), as well as variations in the abiotic factors that form ecosystems, such as temperature, humidity, water quality, soil quality etc.

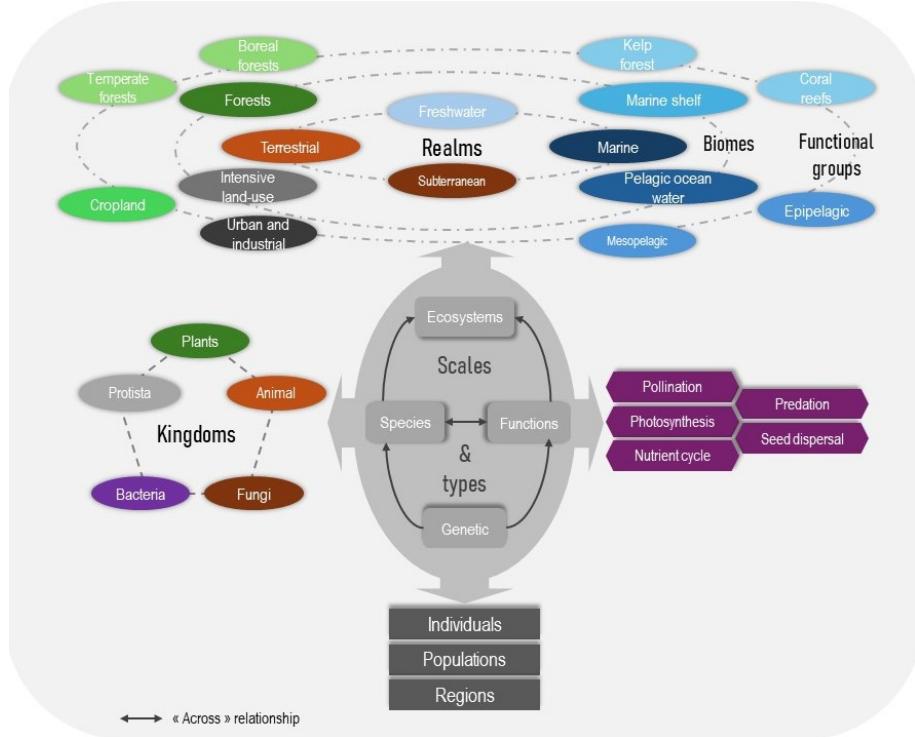


Figure 2: Biodiversity : a multiform concept across scales and types

It also comprises different types of diversity : structural diversity (for example, the layers of canopy in forests, the sex-ratio in animal populations), compositional diversity (the variety and abundance of species within a community), and functional diversity (variety of environmental processes performed by living organisms in a given area e.g. carbon sequestration, nutrient cycling or seed dispersal). Second, structural diversity refers to the way elements are arranged within a group e.g. the layers of canopy in forests, the sex-ratio in animal populations. Finally, functional diversity is the variation of processes performed by elements within a group, e.g. carbon sequestration, nutrient cycling or seed dispersal.

In this dissertation, I focus on the evolution of single species populations and on the evolution of communities within a given ecosystem. Moreover, I study the importance of structural diversity at the species and community levels, and focus on animal species.

Documenting the decline through different scales and types

There has been an ongoing decline in metrics surrounding biodiversity. Across all the scales of analysis, the state of nature is critical. The structural conditions of ecosystems (e.g. interactions between biotic and abiotic elements that cause different structural phenomena and lead ecosystems to provide various resources, including wildlife habitat, carbon sequestration etc), the compositions

(e.g. species richness³) of ecological communities and populations (e.g. species abundance⁴) of species have experienced dramatic changes.

Ecosystem structure e.g. the arrangement of biotic and abiotic elements through time and space, forms the basis of natural and social-ecological processes. Ecosystems can be classified among realms (terrestrial, subterranean, marine, freshwater etc), biomes (e.g. forests, intensive land uses, marine shelf, pelagic ocean water) and functional groups (temperate forests, cropland, coral reefs, etc) (see figure ??). Ecosystem characteristics are severely degraded: only 13% of oceans and 23% of land remains sufficiently unimpacted by humanity to be classified as *wilderness* (Watson et al., 2016; Jones et al., 2018). e.g. areas with "biologically and ecologically largely intact landscapes that are mostly free of human disturbance". On land, at a more refined scale, while deforestation has slowed since the 1990s, vegetation biomass (including trees) has dropped to below 50% of the level expected absent human land-use, suggesting that a planetary boundary has been crossed (Steffen et al., 2015). Additionally, anthropogenic climate change drives ecosystem disruptions on land (Burrell et al., 2020; Conradi et al., 2024) and at sea (Gomes et al., 2024), through changes in various channels including ecological suitability and foodweb disturbances.

Community compositions are changing rapidly, but the average effect remains unclear, as the introduction of alien, disturbance tolerant, climate-migrant species faces local extinctions is spatially heterogeneous (Cardinale et al., 2018). Analysis from long series of spatial data show contrasted results, because of geographic biases, on land and at sea (Dornelas et al., 2014; Gonzalez et al., 2016). When taking a historical reference point, however, the fraction of originally present biodiversity falls well below 90% across all biomes (Hill et al., 2018). Nonetheless, it appears that local communities are becoming more and more similar (McKinney and Lockwood, 1999), driven by the increased extent of animal and plant non-alien invasive species, rising by 13% per decade (Seebens et al., 2017). On aggregate, species richness per grid cell is difficult to measure, as slight decreases measured since 1970 (Kim et al., 2018) do not account for species introduction, nor potentially unredeemed extinction debts (see below) (Jackson and Sax, 2010).

Finally, at the species level, global species richness is threatened by a mass extinction, as the global rate of species extinction is at least ten times higher than the average rate over the past 10 million years and is accelerating (Barnosky et al., 2011; Ceballos et al., 2015). On average, 25% of species are currently threatened with global extinction (Figure 3, International Union for the Conservation of Nature - IUCN (2024)) across a wide range of plant and animal species, on land and at sea. Using different methods⁵, Hoskins et al. (2020) find that hundreds of

³Species richness measures the *number of species* in a given period and location

⁴Species abundance measures the *population of species* in a given period and location

⁵The IUCN Redlist uses detailed accounts for species, in a bottom-up approach, to analyze the

thousands of plant and animal species are threatened, and will repay the *extinction debt* caused by anthropogenic changes to their habitats : only 92.1% of terrestrial vertebrate species, 91.6% of terrestrial invertebrates and 90.7% of terrestrial plants have enough habitat to persist. These results suggest that around half a million terrestrial animal and plant species - including over 3000 vertebrates and over 40,000 plants - *dead species walking*, doomed to become extinct, unless their habitats improve in time to prevent it ([IPBES, 2019](#)).

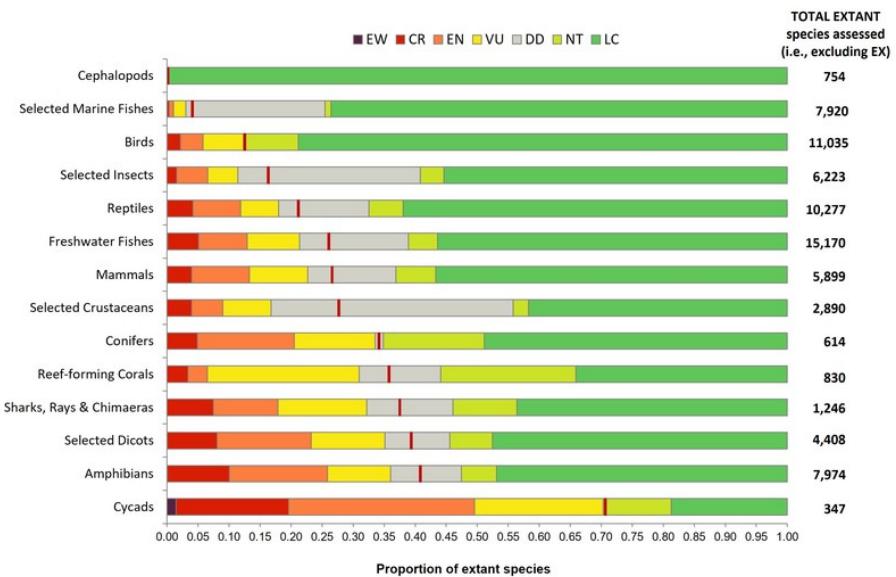


Figure 3: The proportion of extant (i.e., excluding Extinct) species in [International Union for the Conservation of Nature - IUCN \(2024\)](#)

Assessed in each category for the more comprehensively assessed (i.e., at least 80% of the group has been assessed) groups containing ≥ 150 species. Species are grouped into classes. The numbers to the right of each bar represent the total number of extant species assessed for each group. EW - Extinct in the Wild, CR - Critically Endangered, EN - Endangered, VU - Vulnerable, NT - Near Threatened, DD - Data Deficient, LC - Least Concern.

The concept of biodiversity was forged to highlight the need to safeguard the diversity of living organisms across scales from an extinction crisis, acknowledging the threat is first and foremost human.

Anthropogenic drivers of biodiversity decline

The drivers of biodiversity decline are of anthropogenic source. They can be classified between *direct* drivers, e.g. that directly flow from human actions, such as land use change, anthropogenic climate change, overexploitation, and *indirect* drivers, that can be viewed as the root cause for direct drivers, such as , changes in the value systems that underpin nature uses ([IPBES \(2019\) p.55](#)), demography

extinction risk of species. A top-down approach, relying on the evolution of available habitat and the species-area relationship, uses changes in land use to forecast the extinction of species in a more aggregate manner ([Diamond, 1972](#))

(urbanization and migration), technology, economy (sectoral transitions, trade expansion) and governance (including risht systems for access to resources).

The impact of direct drivers can be differentiated among terrestrial and marine species, as well among types of biomes. Synthesis of the available science (IPBES, 2019) (section 2.2.6.2) show that land and sea use, reefering to the loss, fragmentation and degradation of wildlife habitat are responsible for 30% of the impacts on biodiversity. The direct exploitation of wildlife, wild plants and trees represents 23% of impacts. Climate change, through shifts in biogeographic conditions and changes in habit, impacts on species traits and genetic evolution represents 14%, and pollution represents 14% of impacts. Finally invasive alien species represent 11%. These drivers have differentiated impacts across ecosystems and biomes (IPBES, 2019).

For terrestrial species, land use change is the most important driver (30.5%), driven by deforestation and agriculture, and direct exploitation follows next (21%). Tropical and subtropical dry and humid forest host the greatest biological diversity. For example, they host the 10 hotspots with the greatest total number of vertebrates (?). In such forests, habitat loss and degradation are the main drivers of reductions in species abundance and richness (?). Legal and illegal selective logging destroy habitat (??) and are combined with hunting and poaching of wildlife (?), Mediterranean forests, woodlands and scrubs, covering 4 million km², are areas of exceptionally high diversity too (??). However, they are faced with a conjunction of threats, including climate change, land-use transformations (?) and wildfires (Dupuy et al., 2019). Indeed, wildfire frequency and severity are expected to increase with global warming, causing important direct and indirect costs to society including destruction of infrastructure and perturbations to economic activity (Wang et al., 2021), smoke related health conditions (Burke et al., 2023; Heft-Neal et al., 2023), disrupting structural features of ecosystems (Ayars et al., 2023) and threatening biological diversity (Wintle et al., 2020).

For marine species, overexploitation is the main driver (29%) (IPBES, 2019). With 90 million tons of capture in 2020 (?), fisheries stock within biologically sustainable levels have decreased to 64.6% in 2019, from 90% in 1974⁶, driven by overfishing in the Southeast Pacific and the Mediterranean and Black seas. Assessment of fisheries stock and catch management have been proved to improve livelihoods as well as fish stocks globally (??). Nonetheless, illegal, unreported and unregulated (IUU) fishing is a threat to fisheries. Estimates from 15 years ago (?) estimated it represented between 11 and 26 million tonnes of fish. It typically arises in weak governance contexts, with high economic incentives and barriers to enforcement (?).

⁶In this calculation, all fishery stocks are equally counted, irrespective of their abundance or catch

The Importance of Biodiversity and the Challenges of Conservation

Nature's Contributions to People should be protected

NCPs Need for action Weak substitutability and call for action

Addressing the drivers of decline

Pas oublier la conjonction des effets e.g. land and climate change and effects aggregated

Policies for remedying the decline

From the economy to the economics of biodiversity

Biodiversity as an economic object

How to do the economics of biodiversity? Choice of method

Specific bioeconomic modeling challenges in the face of anthropogenic drivers

Dissertation outline

Chapter 1

Bioeconomic models for terrestrial social ecological system management: a review

This article [was published](#) in the International Review of Environmental and Resource Economics with Lauriane Mouysset. [Data](#) and [code](#) are publicly available - DOI 10.1561/101.00000131
It is slightly modified to account for minor errors and add several references.

Abstract

We present a cartography of 319 bioeconomic models applied to terrestrial habitats, combining quantitative analysis of methodological criteria and the narratives behind the equations. Using Multiple Correspondence Analysis and clustering, we identify four groups. Two adopt a conservation focus: the first emphasizes cost-effectiveness in preserving species without monetizing biodiversity, while the second focuses on habitat-based conservation, particularly in agriculture and forestry. The other two groups focus on harvesting, monetizing biodiversity to maximize agent utility and raising cost-benefit issues. One group focuses on endangered and invasive species, while the other highlights forestry. Temporal analysis reveals a recent decline in bioeconomic models for terrestrial social-ecological systems. We discuss this in relation to correlative and data-driven models and propose future challenges for mathematically-based bioeconomic models to reduce uncertainty and incorporate diverse frameworks.

Keywords : Biodiversity; land use change; maximum economic yield; mathematical model; ecological economics; environmental and resource economics; natural capital; ecosystem services; multiple correspondence analysis; K-modes clustering

1 Introduction

Implementing sustainable development constitutes one of the main challenges of the 21st century, given the current ecological crisis. In the last fifty years, two successive trends have paved the way for ongoing studies in sustainability issues. Beginning in the 1970s, large-scale pollution betrayed many of the pressures exerted on the environment by anthropogenic activities. This was followed in the 1990s by a new trend that highlighted the impact of the ecosystem on human development and economic activities (Costanza et al., 1997). The idea that an ecosystem could affect economics yielded new concepts such as the well-known concept of *ecosystem services* (Daily et al., 1997; Hassan et al., 2005; Bateman et al., 2013). The current understanding of sustainability combines these two perspectives, as reflected by the concept of *sustainable development*, which is defined as the management of a complex system, namely, a *social-ecological system* (Ostrom, 2009), and which articulates human society and the ecosystem (Dasgupta, 2008). This dual concern notably led to the creation of the International Panel for Biodiversity and Ecosystem Services (IPBES¹).

Managing these social-ecological systems, therefore, requires understanding the co-evolution of society and ecosystems. On a more technical note, designing sustainable development paths in the context of the ecological crisis requires identifying sustainable dynamics or equilibria, defined as the long-term states needed to maintain viable both socioeconomic and ecological systems. To characterize such sustainable states and their underlying drivers, an adequate understanding and representation of the relationships between society and ecosystems are required. In this respect, we are forced to deal simultaneously with considerations of economic and ecological dynamics as well as their mutual interactions in interdisciplinary-opened scientific researches. Different modeling frameworks that probe the relationships between ecosystems and economics have already been developed in the literature with the economics of natural resources (for an overview, see Halvorsen and Layton (2015))

The integration of natural resources in economics models started with the management of exhaustible resources (Hotelling, 1931; Dasgupta and Heal, 1974). Typically, economic models have been developed to study the extraction of fossil energy. In these settings, natural resources are characterized by a regeneration rate negligible in comparison with its extraction rate. The central economic question about such an exhaustible natural resource regards the investment of the

¹<http://www.ipbes.net/> - "The Intergovernmental Science-Policy Platform on Biodiversity and Ecosystem Services (IPBES) is an independent intergovernmental body established by States to strengthen the science-policy interface for biodiversity and ecosystem services for the conservation and sustainable use of biodiversity, long-term human well-being and sustainable development" see <https://ipbes.net/about>

rent emerging from extraction into a non-natural asset. The extraction rate thus depends on the interest rate: the larger the interest rate, the faster the extraction. Besides these models, other economic models have been dedicated to exploring the management of renewable resources (Smith, 1968; Plourde, 1970; ?). Contrary to exhaustible resources interpreted as a stock, renewable resources are modeled as a flow. Indeed, renewable resources are characterized by commensurable rates of regeneration and extraction. Economic models thus investigate how to maintain the balance between the regeneration and extraction rates, and how to avoid large extraction rates, which would unbalance ecological dynamics and yield to resource erosion. Because biodiversity is a typical example of renewable resources, such resource models are usually designated as bioeconomic models (Gordon, 1954; Scott, 1955).

Historically, bioeconomic modeling for renewable resources (Clark, 1973b; Kontoleon et al., 2007) has extensively been developed for fisheries. Mathematical models of species extinction have been developed on Gordon's and Schaeffer's fisheries models (Gordon, 1954; Schaefer, 1954) to examine the conditions under which the eradication of a given species might appear to be the most attractive policy for a resource owner. Clark's work, which has popularized the concepts of Maximum Sustainable Yield (MSY) and Maximum Economic Yield (MEY), provided a crucial framework for policy-making in regards to exploited marine resources. Typically, these equilibria show that economic decisions that account for interactions between ecosystems and economics reduce the fishing effort compared to decisions taken in ignorance of these interactions. Many extensions of these fishery models have been specifically developed to introduce complexity into the ecological and economic processes (see Petrakis et al. (2012) for a review), towards ecosystem-based fishery management. The development of bioeconomic mathematical modeling for renewable resources in the case of fisheries can probably be explained by the fact that marine biodiversity has been one of the first ecosystems to be strongly damaged by anthropogenic actions. For example, the North Sea herring population collapsed from more than 2 million tons to less than 50 000 tons in the 70s due to overfishing, (Nash and Dickey-Collas, 2005). This marine decline clearly affected economic activities: in the UK alone, the value of the herring fishery dropped from 14 to 2 million pounds between 1977 and 1979, before a slow recovery (Wood and Hopper, 1984).

However, the intensification of anthropogenic pressures over all the ecosystems for the last 50 years, combined with a substantial improvement in the knowledge about ecosystems, has called for bioeconomic studies on other types of biodiversity and habitats (such as estuarine, aquatic or terrestrial habitats). Among them, terrestrial biodiversity is of special interest due to its competition for land with humans. Indeed, urbanization (McDonald et al., 2008; McKinney, 2008) and

agricultural land-use changes ([Dudley and Alexander, 2017](#); [Reidsma et al., 2006](#)) over the last decades have been identified as major drivers of the erosion of terrestrial biodiversity. Such land uses are responsible for the degradation of habitat quality, thus altering species nesting success and survival.

In spite of some early models focusing on pest management in agricultural settings ([Hueth and Regev, 1974](#); [Feder and Regev, 1975](#)), bioeconomic models have been widely developed for the management of non-marine social-ecological systems 20 years after their application to marine resources. Considering such a development of the literature, several reviews have tried to summarize its findings. Some of them adopted an explicit public policy perspective: for example [Boyd et al. \(2015\)](#) focus on bioeconomic model-based articles which investigate conservation planning and the use of return on investment measures or [Epanchin-Niell \(2017\)](#) who reviewed bioeconomic models about the management of terrestrial invasive species. While these studies review the policy issues and the solutions brought by bioeconomic models, they lack methodological consistency since they use a variety of elements, such as narratives, methodological traits, and mixing methodological and statistical approaches. These reviews thus fail at giving an overview of a single methodological framework applied to the management of terrestrial social-ecological systems. On the opposite, other reviews consider a methodological perspective about the bioeconomic modeling framework. We can notably cite [Eppink and van den Bergh \(2007\)](#) which study the biodiversity indicators and theories underlying bioeconomic modeling, as well as [Castro et al. \(2018\)](#) who explore the methodological advances in bioeconomic models applied to agriculture (and mostly abiotic elements) and [Drechsler \(2020\)](#) who explores the integration of spatiality, dynamics and uncertainty in "ecological economic models" for the management of biodiversity and ecosystem services. If these reviews bring valuable insights on the bioeconomic modeling fields, they usually fail in providing a quantitative assessment of the field, with a notable exception in [Drechsler \(2020\)](#). Moreover, these studies often disregard the analysis of the narratives deployed with the mathematical specifications.

In this article, we aim at providing a cartography of the bioeconomic models applied to terrestrial biodiversity based on quantitative methods by combining mathematical and narrative elements of the modeling frameworks. To do so, we performed a review of 319 articles fitting with our specific focus on mathematical and process-based bioeconomic models as popularized for fisheries, but applied to wild and weakly managed terrestrial biodiversity (agro-biodiversity which is strongly managed by humans has been excluded since it has been widely reviewed by agricultural economics). We then studied our database through a methodological perspective by combining an analysis of the methodological criteria included in the economic model, the ecological one and their linkage, and

an analysis of the narratives underlying the equations. In this way, we adopted Gibbard and Varian's standpoint ([Gibbard and Varian, 1978](#)), on stories as an integral part of the model in economics. We provide a cartography of our database using a quantitative analysis relying on Multiple Correspondence Analysis and clustering techniques. Our cartography is organized in 4 groups that we depict in terms of methodological and narrative specifications. More precisely, two of them adopt a conservation perspective: while the first one focuses on how to efficiently preserve species given a limited budget through a cost-effectiveness approach without any biodiversity monetarization, the second one stands for a second generation of models tackling habitat-based conservation measures with specific applications in agriculture and forestry. The last two groups are concerned with the notion of harvesting. Biodiversity is monetized and the problem is framed as the maximization of the utility (or profit) of agents, derived from the flow of the biodiversity variable raising thus a cost-benefit problem. While the notion of harvesting is mostly applied to endangered species and invasive species in one group, a specific interest for forestry characterizes the second one. Surprisingly, the method exhibits a recent and on-going decline over the last years. In regards with this result, some elements of discussion regarding the competition with neighbouring methods, especially the correlative and data-driven models, are in investigated. Since the IPBES methodological report ([IPBES, 2016](#)) highlights the need to maintain a diversity of modeling frameworks to investigate the management of social-ecological systems, especially to embrace different understandings and decrease uncertainty, a discussion on the future of the mathematically-based bioeconomic models is therefore of special interest. In this perspective, we conclude by providing some challenges for its development.

2 Review method

1 Article selection

We performed bibliographic searches on SCOPUS using a wide array of keywords regarding bioeconomic models for renewable terrestrial resources (see annex A for the specified query). Based on these, we ruled out all the articles applied to marine ecosystems. Furthermore, as [Eppink and van den Bergh \(2007\)](#), we tracked the references of the selected articles by hand, and used the website [Connected Papers](#), which provides a map of the earlier and derivative papers from an article. This first screening provided approximately 1000 articles. Then we refined our article selection by precising the concepts of *model* and *bioeconomic*. Figure 1.4 illustrates the process.

First, we need to precise the definition of *model* we used for this review. Indeed

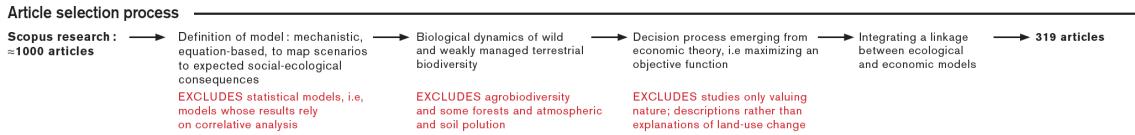


Figure 1.4: Article selection process for inclusion in the systematic review

the modeling literature usually mixes scenarios and models which are both used to provide information to support policy and decision making. However they refer to two different modeling components: scenarios describe plausible futures for drivers of change and options for altering the course of these drivers through policy and management interventions while models enable scenarios of change in drivers to be translated into expected consequences for social-ecological systems (IPBES, 2016). Adopting a methodological perspective of the field instead of a public policy one, we will focus here on models only. Nevertheless, different types of models coexist in the literature. They can rely on quantitative relationships between the components of the social-ecological systems, or on qualitative relationships between them. While the first ones usually have mathematical foundations, the second ones are expert-based. In those models, the experience of experts and stakeholders, including local and indigenous knowledge holders, is used to describe relationships. In consistency with the bioeconomic models popularized by Clark about fisheries, we restrict our attention to quantitative models. Eventually, the scientific literature distinguishes 2 types of quantitative models. On the one hand, the correlative models which rely on empirical data and estimate values for parameters through statistical relationships. In these models, processes are rather implicit. Second, the process-based models which describe explicitly-stated processes or mechanisms based on established scientific understanding. In these models, model parameters therefore have a clear and predefined interpretation. The scope of our review focuses on process-based bioeconomic models. This is of special interest since correlative modelling is probably the best known to manage social-ecological systems, especially due to the popularity of correlative species distribution modelling (Elith and Leathwick, 2009). In such a context, we postulate that a specific focus on the alternative method might bring new insights about social-ecological system management.

Second, bioeconomics is a polysemous term which is used in different strands of literature: a first group is related to N. Georgescu-Roegen and develops a thermodynamics understanding of social-ecological systems (Georgescu-Roegen, 1971); a second group, led by Clark on Gordon's and Schaeffer's foundations, develops mathematical models integrating ecological and economic processes (Clark, 1973a; Cla; Schaefer, 1957)²; finally a third group is related to biomimetism

²See Parent et al. (2024) for a history of the 'standard' fishery model, notably the gradual inclusion of dynamics

where technological innovations are inspired by living systems (Van Lancker et al., 2016). In this review, we focus on the second group of literature, related to Clark's bioeconomic mathematic modeling.

To do so, we define bioeconomic models at the intersection of 3 conditions:

1. integrating an explicit biological dynamic
2. integrating a decision process emerging from economic theory,
3. integrating a linkage between ecological and economic models.

The first item characterizes the ecological dynamics of a renewable resource where the rates of regeneration and extraction are commensurable. Except for this condition, no specific requirement of the ecological process at play is needed. Different ecological processes such as population dynamics or niche distribution are thus eligible. By biological, we mean that the dynamics have to be related to living organisms. In other terms, the stake of the model has to be related to biotic elements. This condition aims at excluding pollution models or carbon and nitrogen models (Nordhaus, 1994; Lemoine and Traeger, 2014). Eventually, by explicit we mean mathematically formalized. This condition is necessary to exclude exclusively declarative bioeconomic models (*i.e* bioeconomic frameworks without any mathematical formulation). Indeed, the objective of this study is focused on changes in a specific method (*i.e* the mathematical process-based bioeconomic model) rather than in a problem (*i.e* the bioeconomic one). Because they adopt a different methodological framework, correlative or declarative bioeconomic studies need to be excluded from our corpus.

The second item precises the economic side of bioeconomic models. By considering an economic decision process, we aim at excluding articles performing an economic valuation of biodiversity such as empirical studies giving the monetary values of species, like owls or bats (Montgomery et al., 1994; Penn et al., 2019). Although such studies are highly valuable to deal with the ecological crisis, they stem from a very different methodological tradition (choice experiments and monetary valuation). By explicitly requiring a decision process from economic theory, we ensure to avoid agro-ecological models. Indeed, many agro-ecological models address the question of sustainable management of terrestrial social-ecological systems and bring valuable knowledge to this question. However, they combine ecological dynamics with land-use change models without specifying the economic determinants of these land-use changes (some costs are sometimes associated with these land-use changes but without being driven by economic processes) (Sabatier et al., 2010). The methodological corpus we are interested in in this article is rooted in economic theory. Thus, we only consider economic decision models, in which agents allocate scarce resources to fulfill their

objectives. Agents can, for example, maximize their utility or profit, or act as cost-minimizers to achieve specific goals.

The third condition is that of an integrated ecological-economic system, *i.e.*, how the ecological and economic models are coupled. This bioeconomic linkage is not specified and can take different forms: for example, it can be mutual (by considering simultaneously the anthropogenic effects on ecosystems and the economic valuation of biodiversity in an economic problem) or unidirectional (one of the two effects mentioned above), and it can be done by prices or by physical variables. If the bioeconomic coupling is done by prices, economic value will be granted to the biodiversity elements to make them commensurable with other economic determinants. However, this monetary quantification has to be incorporated into a decision model (cf previous item).

Eventually, these bioeconomic modelling specifications have been applied to terrestrial social-ecological systems. Since many studies take place in an agricultural context, it is necessary to specify here the distinction between *agro-biodiversity* and *agricultural biodiversity*. *Agro-biodiversity* stands for species which are directly managed by farmers (for examples the crops species, the battle species etc) while *agricultural biodiversity* stands for wild biodiversity living into agricultural habitats (such as birds, bats etc). Because the economic aspects of agro-biodiversity have been broadly studied by agricultural economics, we focus here on wild terrestrial biodiversity. In this perspective, bioeconomic models applied to only managed forests, such as the seminal article of [Faustmann \(1849a\)](#)³, are excluded. Models with natural forest ingrowth are however in our scope. To finish, we formally exclude articles with marine case studies. For example, articles with both marine and terrestrial case studies have been excluded in this review. In doing so, we aim at providing a restricting view about terrestrial social-ecological system management. However the integration of such excluded articles appears as a natural perspective of future extensions of this work.

Based on these criteria, we individually screened all the papers selected in the first look to refine our database. Among 1000 articles identified after the first literature screening, we selected 319 articles developing bioeconomic models applied to terrestrial social-ecological systems.

2 Analytical framework

To analyze mathematical tools such as bioeconomic models, we adopted here a methodological perspective. However as mentioned by Gibbard and Varian ([Gibbard and Varian, 1978](#)) at an early stage, stories are an integral part of the model in economics. More precisely, the authors explain that a model is a *story* with a

³Moreover, Faustmann's work focusing on tree values does not feature any biological dynamics.

*specified structure*⁴. In that perspective, methodological specifications are not sufficient to characterize the model since the questions the authors want to explore and the stories they can tell with it are at the core of the model identity. Such narrative elements are more than chronicles, they are essential to connect economic modeling research with the specifics of the world (Morgan, 2001). Without these narrative elements, it is impossible to apply model-structures directly onto the facts of the economic world. Since we are interested in models that are motivated by concrete stakes such as resource management, the biodiversity crisis and sustainable development, exploring narratives associated with the methodological specifications of the mathematical model is crucial to characterize the outline of such a bioeconomic modeling.

In this context, we developed an analytical framework based on two dimensions: the first one is based on a set of methodological criteria related to mathematical equations, while the second one is related to the narratives associated with the mathematical tool. Based on the combination of these two dimensions, we aim at providing an overall cartography of bioeconomic modeling as a tool to investigate the management of terrestrial biodiversity.

3 Cartography method

For our methodological analysis, we first investigated a set of 18 criteria related to the ecological model, the bioeconomic linkage, and the economic model.

1 Ecological criteria

The ecological criteria aim at precising how biodiversity is captured by the ecological model. To do so, we mobilize 8 criteria split into 2 groups. The first group of criteria helps to understand the paradigm of biodiversity while the second group is related to the technical specifications of the ecological model.

Within the first group, the first criterion is related to the measure of biodiversity. Indeed, biodiversity can either be modeled *per se* (for example based on population dynamics models) or be deducted from a proxy (typically, the habitat suitable for biodiversity or economic activity). The second criterion precises the proxy measure: this proxy can be habitat, economic activity, a conservation budget, or not be specified. The third criterion precises the ecological state variable in the ecological model. More precisely, the biodiversity variable can be related

⁴"A model [...] is a story with a specified structure. The structure is given by the logical and mathematical form of a set of postulates, the assumptions of the model. The structure forms an uninterpreted system [...] Although the term 'model' is often applied to a structure alone, we shall use it in another sense. In economists' use of models, there is always an element of interpretation: the models always tells a story." (Gibbard and Varian, 1978), p.666)

to the individuals (such as in population or metapopulation models), the species (when focusing on species richness), or the community, when both species abundance and richness are taken into account. The fourth criterion focuses on the type of biological diversity, *i.e.* we distinguish functional and genetic diversity definitions. Finally, the fifth criterion characterizes the biodiversity level at which the model intends to contribute. Some articles are focused on a single species (for example, articles based on a population model developed for one species) while some others adopt a community perspective by integrating a pool of species. In some cases, when species interact, models display two species. However, many of the articles we reviewed did not focus on species interactions and therefore encompassed a larger number of species. This community perspective can be either explicit, as in articles modeling populations of different species, or implicit, in studies using a habitat proxy as a biodiversity measure and informing about the community living in this habitat. It is interesting to note that there is no systematic implications between habitat, proxy-based models and community level contribution since the habitat might be related to one single species.

Besides this first group of criteria for the characterization of biodiversity, we mobilize a second group of criteria related to the ecological technical specifications. The first criterion is related to the category of biological dynamics: we distinguish population dynamics models (such as in the seminal model developed by Clark, or articles implementing age-structured modeling) and other ecological dynamics. These other dynamics can be for example either a niche distribution model or Brownian motion models. The second criterion characterizes the spatial dimension of the ecological process. Spatial considerations can be explicit when the ecological process implies spatial exchanges (typically a metapopulation model) or implicit, when the ecological process at play takes place in a heterogeneous context (for example when heterogeneous patches are taken into consideration for an aggregated analysis without any exchange between the patches). Eventually, the spatial dimension can be absent. Then the third criterion is related to the integration of stochasticity in the ecological modeling. Stochastic components may include dispersal probabilities of species across land patches as well as probabilities of species extinction.

Table 1.A sums up the ecological criteria with their related items.

2 Bioeconomic linkage criteria

Bioeconomic linkage criteria characterize how biodiversity is taken into account in the economic model and the economic decision. To do so, we mobilize 3 criteria. The first criterion indicates whether the biological element has been monetarized or not. In order to make biodiversity commensurable with other economic

variables in the decision problem, some articles rely on an economic valuation of biodiversity (in other words, biodiversity is expressed in monetary units, such as dollars). A monetary bioeconomic linkage occurs in two situations: either if the study is directly driven in monetary terms (for example when biodiversity is measured through a proxy in economic units) or if the ecological model is developed in non-monetary terms (with a biodiversity measure *per se* or based on a habitat-based proxy) but the biodiversity is monetarized thanks to a monetarization method to be integrated into the economic decision.

The second criteria precises how the bioeconomic problem is raised. We distinguish two problems: the cost-benefit problem and the cost-effective problem. Cost-benefit analysis integrates costs and benefits related to classical economic factors and ecological factors, then selects the decision which maximizes the overall utility. Due to criticism on monetarization methods ([Diamond and Hausman, 1994](#)), some authors favor cost-effectiveness analysis which separates classical economic factors and the ecological ones. The economic decision is thus taken according to a maximization under constraints. Typically, the optimal decision maximizes the profit or the utility under an ecological constraint. By isolating ecological and economic objectives, this cost-effectiveness method aims at limiting the substitutability between natural and non-natural capital. Interestingly some studies consider the ecological value in economic terms (for example when biodiversity is measured through an economic proxy) but keep separated the benefits or costs emerging from the ecosystem and the ones emerging from classical economic factors. In other terms, an economic value for biodiversity does not necessarily imply a cost-benefit problem. It is the reason why it is informative to keep in our review the two criteria, relative to biodiversity monetarization and the bioeconomic problem respectively.

The third criterion captures the position of the biodiversity stake in the bioeconomic model. The biodiversity stake can be within the objective of the maximization such as in cost-benefit problems but also in a cost-effectiveness problem which maximizes the ecological output while satisfying a cost constraint. Then, the biodiversity stake can be a constraint (in a cost-effectiveness problem which maximizes profit under ecological constraint for example). Eventually, other stakes occur either when the biodiversity stake emerges in both the maximization and constraint or when the biodiversity stake is an output. The simultaneous consideration is possible when are considered different taxonomic groups (one being in constraint while the other is included in the maximization) or when non-human well-being is taken into consideration in the objective function while biological dynamics constitute mechanistic constraints. The biodiversity stake can be assessed as an output computed after the economic decision.

Table 1.A sums up the bioeconomic linkage criteria with their related items.

3 Economic criteria

Economic criteria specify the economic side of bioeconomic models. More precisely, we explore a set of criteria related to the technical economic specifications. They are related to dynamics⁵ and spatial dimensions, and to uncertainty. Bioeconomic models are economically either static or dynamic. Similarly to ecological technical specifications, we explore the spatial and uncertain dimensions. Economic spatiality can be investigated explicitly through a spatial process such as trade between regions or implicitly by spatial heterogeneity of economic variables, or eventually absent of bioeconomic models. Eventually, economic models can be either deterministic or stochastic if an economic variable is primarily subject to a source of uncertainty.

To finish, we explore four last criteria regarding the general characteristics of bioeconomic models. The first one is related to the solving method used to explore the bioeconomic question. We distinguish 3 forms of solving method in the articles of our corpus: closed form resolution, numerical resolution and the combination of both. The second criterion informs whether the study is empirical or theoretical or whether its combines empirical and theoretical perspectives. Additionally, we explore how the model is used to highlight the economic question. If the solution emerging from the bioeconomic model characterizes a judgement on the best behavioral options or policy instruments, the model use is normative. On the other hand, if a paper investigates some behaviors of the system without any recommendation, the model use is descriptive. Eventually, we characterize whether the model is framed in terms of general equilibrium or partial equilibrium.

Table 1.A sums up the economic criteria with their related items.

4 Methodology-based cartography

These methodological criteria have been analysed through a Multiple Correspondence Analysis (MCA). MCA allows to uncover the underlying structure of categorical data by performing a recombination of the data into a two-dimensional space formed by orthogonal vectors which maximize the variance (inertia) explained by the data (see [Benzécri \(1976\)](#) for seminal works and [Roux and Rouanet \(2010\)](#) for a modern presentation).

As MCA can be sensitive to unbalanced variables (*i.e.* variables whose distribution are highly skewed towards one value), we performed a sensitivity analysis to select the optimal combination of variables to use in the MCA, based on the ex-

⁵Indeed whereas our definition of bioeconomic models requires a condition of dynamics in the ecological model (see section 1), we do not impose any dynamics specifications on the economic side to be included in the database.

plained variance. To do so, we performed an MCA analysis with all the possible combinations of our sample. The graph 1.C in appendix C exhibits the explained variance as a function of the number of criteria. Among the 18 methodological criteria, we selected the set of 14 methodological criteria (see tab. 1.A) which keeps a large set of criteria while reaching 31% of the explained variance. The rationale for variable selection was to avoid redundancies as well as excluding variables that are too skewed and would impair MCA analysis.

Based on these selected criteria, we performed a classification with a K-modes algorithm ([Huang \(1998\)](#)). The K-modes algorithm generalizes the K-means method⁶ to categorical data, and uses a dissimilarity measure to assign observations to clusters. One of the inconvenients of K-modes algorithms is the need to specify the number of clusters. Therefore, the number of groups used in the classification was determined using a cost function, namely the sum of the within variance of each clusters. We used the so-called 'elbow-method' ([Ketchen and Shook \(1998\)](#)), stating that the optimal number of clusters is located at an elbow of the curve relating the sum of the within cluster variances and the number of clusters. Indeed, after this point, the reduction in the sum of the within cluster variances becomes less important, suggesting additional clusters do not significantly improve results. Figure 1.D in appendix C depicts this cost function. For the following analysis, we will consider the optimal cluster of 4. However since 9 clusters might also be considered as optimal clustering, we also present an MCA classification with 9 clusters as robustness test (see figure 1.E in appendix C).

5 Narrative-based analysis

In order to perform a narrative analysis, we used the titles, keywords and abstracts of the papers in our database. Then we pre-processed the data by removing stopwords and grouped similar words together (for example, *farming*, *farmers* and *farms* were all grouped under *farm* with this procedure). Moreover, because our analysis relies on single words, typical nominal groups were recoded (for example, *endangered species* was recoded into *endangeredspecies*)⁷. For our analysis, we kept words which occurred at least 5 times in our database⁸. Since we have 319 articles, we thus kept a significant portion of the words database, which

⁶The K-means algorithm ([MacQueen, 1967](#); [Lloyd, \(1957\)](#)) is a standard classification algorithm in Natural Language Processing. Documents are mapped to a vector space featuring as many dimensions as there are distinct words in the document, and are thus coded in a binary fashion. The algorithm picks random initial centroids, computes the Euclidean distance to other observations, which are assigned to the closest clusters. Centroids are thus actualized, and the procedure is repeated. If no observations changes cluster upon a new iteration, it converges.

⁷The following expressions were recoded and grouped together : endangered species, bio-economic, invasive species, ecosystem service, optimal-control, dynamic-programming, integer-programming, cost-effective, cost-benefit, reserve-design, optimal-management, land use, property rights, conservation-planning

⁸These 5 occurrences can come from a single article or at most 5 articles.

displays the most information. More precisely, a total of 1202 words out of 4355 relevant words (27.6%) were kept, accounting for 81% of word occurrences.

Eventually, we classified words according to semantic fields. Based on the 1202 words words kept for the analysis, we designed 8 lexical groups, pertaining to two habitats (agricultural and forest), two species status (invasive species and endangered species), two management semantic fields (policy and risk) and two human-nature paradigms (conservation and harvesting). The list of words in each semantic field is depicted in appendix B. In order to characterize the narratives underlying the different groups resulting from the methodology-based classification, we investigated the bias of each semantic fields into them. More precisely, for each methodology-based group, we assess the ratio between the frequency of the semantic fields and the number of papers included in this group. This ratio avoids size effects between groups.

4 Database overview

1 Temporal and geographical distributions

Figure 1.5 presents the distribution of the articles in the database. Most articles range from the 90's, testifying the recentness of the use of such methodology for terrestrial social-ecological systems. Except some early-bird articles published in the 70's related to the management of agricultural pests and pesticides use ([Hueth and Regev, 1974](#); [Feder and Regev, 1975](#)), the distribution of the articles follows a Gaussian function with a 20-years spike between 1995 and 2015. This indicates that the use of such bioeconomic models to investigate the sustainable management of terrestrial social-ecological systems has decreased recently. This decline is of special interest as the question of sustainably managing terrestrial social-ecological system is far from solved. This situation is quite unusual for a methodology associated with such a burning issue which calls for a strong research effort and generates a huge amount of literature.

To complete this temporal distribution, we investigate the geographical origins of the authorship of the 319 articles (fig. 1.6). We observe that our database in majority emerges from North American and European research even if the part from Oceania is not negligible. Eventually a small part comes from Asia and Africa. This relative dominance of North American research could be explained by the original diffusion of Schaeffer's (American) and Clark's (Canadian) seminal models. From a more naturalist stand, the magnitude of the resources and the early conservation movement in North America could have paved the way for this trend.

However with a higher pressure on land use and land use change in Europe

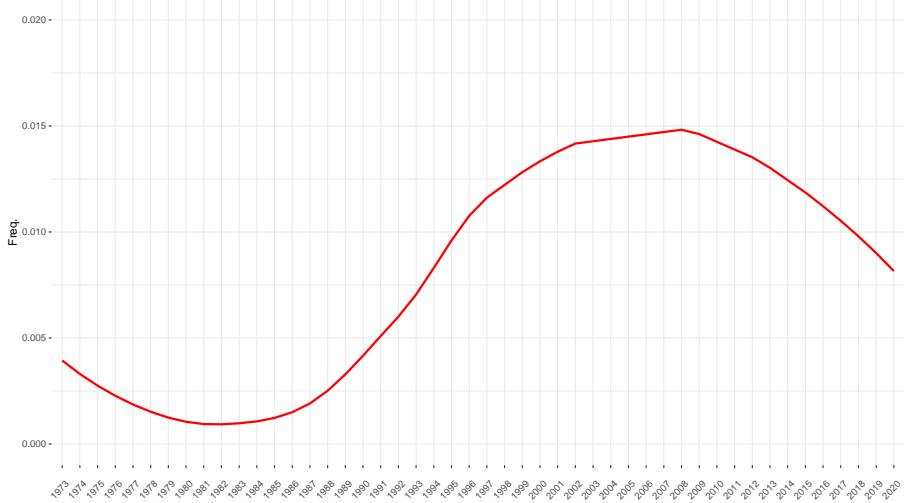


Figure 1.5: Temporal distribution of articles in the database.

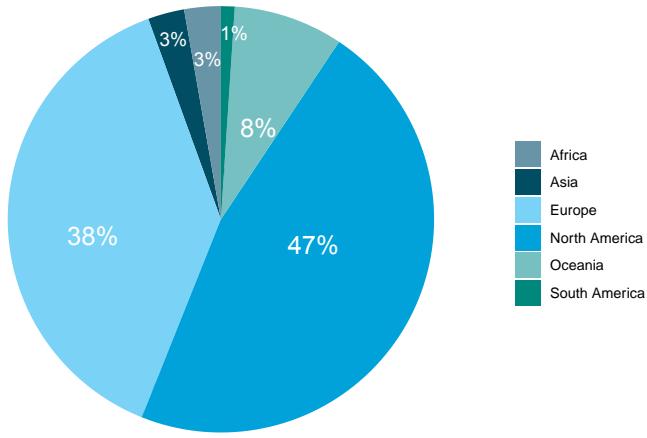


Figure 1.6: Geographical distribution of articles in the database.

than in the USA, European researchers look also very active to face the urgent and concrete stake of managing terrestrial ecosystems while reconciling socioeconomic goals and ecological requirements.

2 Journal and discipline distributions

The articles emerge from 97 journals which are related to different disciplines such as applied mathematics, economics, ecology and sustainability sciences (see tab. ??). Based on the journal affectation, table 1.1 sums up the frequencies of these 3 disciplines among our corpus of 319 articles.

We observe that most of the articles have been published in journals related to economics (60%) confirming the anchorage of bioeconomic modeling as an eco-

Table 1.1: Distribution of journal across fields

Field	Count	Percentage
Economics	190	60%
Ecology	83	26%
Sustainability science	34	10%
Applied Mathematics	12	4%

nomic approach. Among the journals, one of them captures a substantial part of the papers: 44 papers (*ie* 14% of the overall database and 23% of the papers published in economic journals) are indeed published in *Ecological Economics*. This dominance was expected since the methodology brought by bioeconomic modeling fits perfectly with the scope of the journal. Indeed, this journal focuses on the articulation of ecological and economic issues in the perspective of sustainable development⁹. Beside this journal, bioeconomic models contribute to classical resource management questions (with environmental and energy journals such as *Environmental and Resource Economics*, *Journal of Environmental Economics and Management*), applied questions and notably agricultural economic journals (such as in the *American Journal of Agricultural Economics* or the *Agricultural Economics*), theoretical economic questions (with classical theoretical journals such as *Econometrica*, *American Economic Review*).

The proportion of articles published in non-economic journals (40%) testifies an interest for bioeconomic models beyond its economic expected arena. The non-negligible part of articles published in Ecology journals (such as *Ecological Modeling*, *Conservation Biology*, *Ecology Letters* or *Journal of Theoretical Biology*), sustainability sciences journals (such as *Natural Resource Modeling*, *Agricultural Systems* or *Environmental Modeling and Software*) and applied mathematics (*Journal of Mathematical Analysis and Applications* and *Journal of Mathematical Biology* for example) emphasizes an acceptance of bioeconomic models outside the field of economics. And more specifically, it confirms a certain legitimacy of bioeconomic models regarding ecological theory and knowledge. In this perspective, bioeconomic modeling embraces a genuine interdisciplinary aspiration.

⁹The scope of *Ecological Economics* mentions that "The journal is concerned with extending and integrating the understanding of the interfaces and interplay between "nature's household" (ecosystems) and "humanity's household" (the economy). Ecological economics is an interdisciplinary field defined by a set of concrete problems or challenges related to governing economic activity in a way that promotes human well-being, sustainability, and justice."

5 Database cartography

1 Methodology-based classification

Figure 1.7 presents the results of the Multiple Correspondence Analysis (MCA) running on 14 methodological criteria. It also displays the classification of the articles resulting from our K-modes algorithm into 4 groups : group 1, in green, with 47 articles; group 2, in purple, with 48 articles; group 3, in yellow, with 162 articles; group 4, in black, with 62 articles.

We observe on figure 1.7 that the MCA is well structured on both sides of the y -axis with groups 1 and 2 on the right side, and groups 3 and 4 on the left side. The x -axis offers a split between groups 1 and 2 while it does not strongly play on the groups 3 and 4 even if the main part of group 3 tends to be below the x -axis while the main part of group 4 tends to be above. Figure 1.E in appendix C exhibits the MCA based on 9 groups. While being more fragmented, the cartography exhibits a similar structure to the 4 groups classification.

The interpretation of the 4 groups comes with figure 1.8 which depicts the distribution of the items of the selected 14 methodological criteria. The colors stand for the contribution of the items to the structuration of the axes. We observe that the x -axis is strongly driven by the criteria related to the bioeconomic problem e.g. biodiversity monetarization, data anchorage, solving method and spatiality. More precisely the left side is characterized by a cost-benefit problem where biodiversity is monetized. The problems are theoretical and solved with closed-form solutions. Eventually, the problems do not integrate spatiality. On the contrary, the right side is characterized by the cost-effective problem where biodiversity is not monetized. The problems are mainly empirical and solved with numerical tools, and take into account spatiality. The y -axis is mainly driven by the integration of spatiality in the economic model and the framing of the economic problem as a general equilibrium.

Combining figures 1.7 and 1.8, we understand that the MCA classifies the articles in our database in a first group (in green) specified by a cost-effective problem, an absence of biodiversity monetarization, empirical and theoretical studies and stochasticities being present both in ecological and economic models ; a second group (in purple in figure 1.7) specified by a cost-effective problem, an absence of biodiversity monetarization, empirical studies and spatiality being explicit on the economic side and implicit on the ecological side ; a third group (in yellow in figure 1.7) specified by a cost-benefit analysis, biodiversity monetarization, numerical and theoretical solving and the absence of spatiality ; a fourth group (in black in figure 1.7) specified by a cost-benefit analysis, biodiversity monetarization, theoretical solving and explicit spatiality.

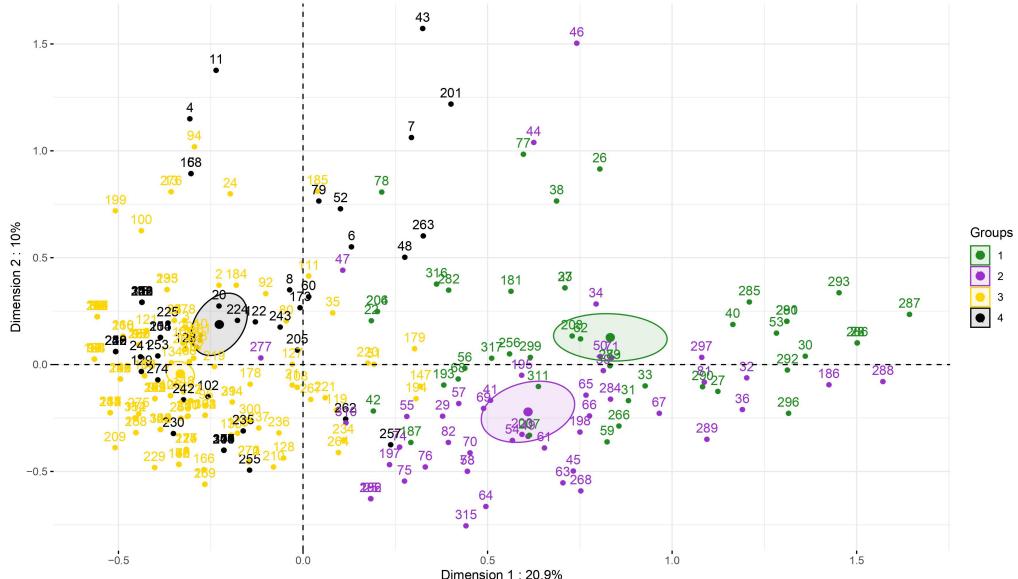


Figure 1.7: Multiple Correspondence Analysis (MCA) running on 12 methodological criteria and 4 clusters (K-modes)

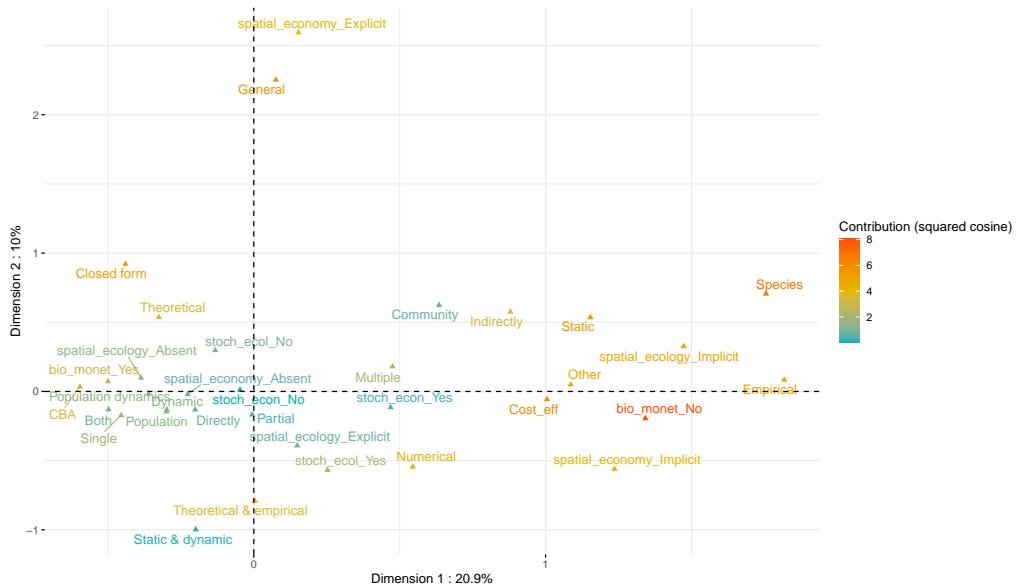


Figure 1.8: Distribution of the values of 14 methodological criteria among the MCA axes.

2 Narrative-based specifications

In order to interpret the narratives of the 4 methodology-based groups, we depicted on figures 1.A and 1.B the distribution profiles of the 50 most frequent words for each group.

First of all, we observe that for all profiles the most common words are those which are in common with the 4 groups. We identified the following keywords: (i) economic, cost, (ii) management, policy, strategy, conservation (iii) biodiversity, resource, species, population, ecological, biological, (iv) model, optimal, dy-

namic, (v) land use, forest, habitat. This observation indicates that the 4 methodology-based groups are driven by a common narrative which regards an economic problem of management of biodiversity and natural resource and land use change based on models, mostly relying on optimal control theory. This result confirms the consistency of our database regarding the research question investigated in the selected papers within the database.

Since the specific words are too disparate to be easily understandable, we completed these profiles by a semantic fields analysis in order to characterize the 4 methodology-based groups. The figure 1.9 depicts the frequency of the different semantic fields in each group¹⁰. We observe that groups 1 and 2 are related to conservation issues. Among conservation-related articles, the ones with specific applications into agricultural landscapes, especially related to public policy issues, are preferentially located in group 2. On the opposite, groups 3 and 4 are related to harvesting issues. A specific focus on endangered and invasive species characterizes group 3. Eventually, group 4 looks dedicated to the risk problematic with forestry applications.

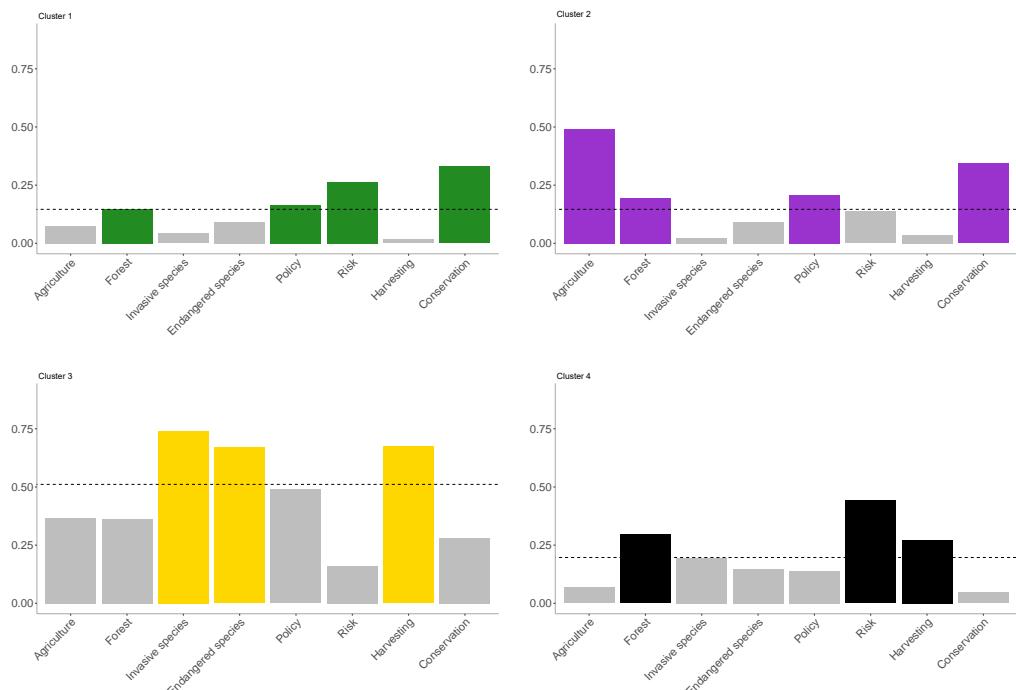


Figure 1.9: Frequency of the different semantic fields in the 4 methodology-based groups.

On the *y* axis is the proportion of mentions of lexical groups in a given cluster across all the mentions. The dashed line represents the size of each cluster as the share of articles of the database they represent. Each color corresponds to the cluster color in figure 1.7

¹⁰see appendix B for the listing of words within each semantic field

3 Overall cartography

Combining methodological and narrative specifications draws thus a 4-groups cartography where each group can be described as follows.

The first group is polarized towards conservation issues broadly rather than specifically applied to a type of habitat. Spanning from 1992 to 2019, with a median year in 2006, it can be viewed as a first generation of models applied to conservation, i.e, focusing on the optimal ways to conserve species rather than harvesting them. This corpus focuses on how to efficiently preserve species given a limited budget for land acquisition through a cost-effectiveness approach without any biodiversity monetarization. It can be viewed as a generalization of the so-called "Noah's Ark" problem ([Weitzman \(1998\)](#)). In this paper, Weitzman considers the genetic diversity of an array of species to maximize the amount of biological diversity one can fit into an Ark, e.g, given a limited budget and the cost of conserving a species. An array of papers, such as [Courtois et al. \(2014\)](#) revisit Weitzman's definition of diversity, including species interactions, to refine the criterion to be maximized in conservation planning.

Considering not only one Ark, but a variety of habitat patches for species conservation broadens again the issue. Moreover, including costs in the decision process is required to design efficient conservation strategy. This concern yields optimal reserve site selection problems. Whereas the seminal Weitzman's Ark framework was a theoretical one, this extension usually comes with a theoretical enquiry and an empirical case study. For example, [Costello and Polasky \(2004\)](#) focus on the optimal combination of sites suitable for an array of species that need to be set aside from development, permanently or temporarily. Using dynamic integer programming, the authors showed that the timing of decisions, the quality of habitat in patches as well as their costs is key to designing optimal reserve sites for a large set of Southern California vertebrates. Using the same approach, a wide array of papers focus on the static problem of optimal reserve site selection at a very large scale, in order to prioritize conservation projects. For example, [Moore et al. \(2004\)](#) focus on the minimization of the costs to operate a network of reserves in Africa that covers 10% of its 118 ecoregions. Using species-area relationships and considering that land costs are correlated with high endemism or threat, focusing only on cheap areas was unlikely to yield the desired conservation outcome. Moreover, factoring in land prices in the reserve site decision problems was shown to increase the cost-effectiveness of the prioritization scheme.

The second group spans from 1993 to 2021, with a median publication year in 2010. Focusing on conservation, it can be viewed as a second generation of models tackling specific habitat-based conservation measures. The typical re-

search question is how to conserve biodiversity in a working landscape, i.e, when land-use is devoted to agriculture, and to a lesser extent, to forestry. Considering biodiversity, mostly in the form of multiple species, as a separate entity, a cost-effective problem is framed in order to find optimal solutions to reconcile the economic and ecological objectives. In this context, a wide array of solutions are considered. For example [Polasky et al. \(2005\)](#) develop a spatially explicit framework in which a large set of vertebrates from Oregon can stochastically migrate across land patches as they compete for habitat with agriculture and forestry. Due to the analytical complexity of the problem, [Polasky et al. \(2005\)](#) use a variety of algorithms to gradually increase the biodiversity objective and find the least-cost policy in terms of land use, thus resulting in a production possibility frontier. While land-use policies are key, other articles investigate monetary based policy instruments to conserve biodiversity. In this approach, [Drechsler et al. \(2007\)](#) develop a single-species, spatially explicit meta-population model of butterflies living in an agricultural landscape in Germany. Taking into account species dynamics, agricultural constraints, and heterogeneous land quality for agricultural and conservation purposes, they design means of determining cost-effective solutions to biodiversity conservation through conservation payments. Based on this framework, they show that patch-specific conservation payments can increase ecological benefits up to 50% compared to uniform strategies. Co-leading a European strand of literature on the conservation of species in a working landscape is [Mouysset et al. \(2012\)](#). In this paper, a spatially explicit model of 620 small French agricultural areas is coupled with a public decision maker who aims at preserving diversity under budgetary constraint. Farmers decide their management schemes under uncertainty and with no specific regards to biodiversity, apart from economic incentives. The model is used to evaluate various policy scenarios pertaining to farm management and the impact on common farm bird species. Optimal policies such as tax and subsidies to promote biodiversity conservation are derived.

The third group is the largest group (51% of our database) from our classification, spanning over the whole temporal distribution (1973-2021) and a median year in 2005. It is mostly concerned with the notion of harvesting, i.e, removing a portion of the biodiversity variable for beneficial use. The measure of this beneficial use tends to be monetary, and the problem is framed as the maximization of the profit or utility of a set of agents derived from the flow of the biodiversity variable, mostly population, raising thus a cost-benefit problem. The notion of harvesting is mostly applied to two particular sets of species, endangered/remarkable species and invasive species, that are characterized with opposite properties : the former is an economic "good", the latter is an economic "bad".

In the endangered/remarkable case, a good example is [Skonhoft \(1999\)](#). Considering the case of African wildlife, especially large mammals, and factoring in land-use costs, non-consumptive benefits, nuisance costs and harvest profits, Skonhoft examines the dynamics of a single species' population and its optimal harvesting scheme, in a deterministic framework. This paper can be seen as one of the most refined versions of the work of [Cla](#) and later on [Swanson \(1994\)](#), who examines the optimal harvesting of African elephants in the context of land-use pressures, later on refined by [Alexander \(2000\)](#) through the integration of non-consumptive values. In this strand of literature, the institutional arrangements between stakeholders are refined, thus examining the equilibria between poachers and locals, the potential for tourism revenue, and the interaction of conservation measures and harvesting. What is key is the Human around interactions surrounding the resource, rather than the resource's intrinsic dynamics, such as migration or uncertain population dynamics.

Invasive species, whether present in agricultural, forestry or wildlife settings, are one of the earliest application cases of bioeconomic modeling for terrestrial ecosystems. In these settings, a resource owner (mostly farmers or foresters) are concerned with the spread of a single invasive species. In this case, optimal control methods are developed to compute the optimal amount of surveillance and detection, pesticides use, preventive cuts or harvests, to prevent damages from invasions. A typical example can be found in [Jayasuriya et al. \(2011\)](#). In this article, a state of the art population dynamics, seed-bank model is applied to a crop invader. This invasive species spreads stochastically, depending on both its intrinsic growth rate, and the agricultural crop growth rate. Using dynamic stochastic programming, the authors show that control measures are always beneficial, and that if eradication is too costly, it still pays to maintain infestation at low levels. While agricultural damages are of interest, the value of ecological degradation from biological invasions are also considered. For example, [Taylor and Hastings \(2004\)](#) investigate the spread of *Spartina Alterniflora*, an invasive grass species, in Willapa Bay in the state of Washington in the USA. This species, subject to a density-dependent, age-structured growth function, is to be removed, and the authors investigate the least cost strategy, for the sake of the preservation of the local landscape.

Eventually, **the fourth group** spans all over our temporal distribution, with a median year in 2002. This group focuses on more specific biodiversity dynamics. It tends to focus on uncertain biodiversity dynamics, and to a lesser extent, multiple species relationships (mostly predator-prey, but incorporating some mutualistic configurations). In this context, decision makers are concerned with the optimal harvesting of a stochastic population that can be an economic good or

bad. It is therefore no surprise that forestry economics are more represented in this setting. A typical example can be found in [Lin et al. \(1996\)](#), where a density-dependent stochastic growth model governs the evolution of forest stands characterized by their diversity. The question, akin to Faustmann's seminal interrogation ([Faustmann, 1849a](#)), sums up to when is it optimal to harvest this uneven-aged stands forest? Taking into account age and species diversity modifies the optimal harvesting rule.

Forests can also be the habitat to stochastic populations of invasive species. For example, [Epanchin-Niell et al. \(2014\)](#) focus on bark beetles and wood borers, that may invade forests. The authors focus on the optimal surveillance strategy to develop in order to prevent a detrimental forest invasion in New Zealand. The program's costs are weighed against the benefits (in the form of forgone damages) from earlier detection. Their appraisal of the relative costs and benefits from surveillance suggest that implementing the program is always beneficial, under all considered scenarios. Eventually, a small last strand of literature of this fourth group focuses on the economic implications of the stochastic nature of biodiversity on the provision of ecosystem services. Using a single species, closed-form mathematical framework, [Augeraud-Véron et al. \(2019\)](#) investigate the value of biodiversity as an insurance device for agricultural production, as it decreases agricultural productivity volatility. In a similar fashion, [Baumgärtner \(2007\)](#) characterizes the insurance value of biodiversity in the provision of monetary values ecosystem services, not specifically agriculture. Biodiversity conservation therefore becomes a financial product, akin to financial insurance.

Not surprisingly, articles from the first two groups (related to conservation paradigm) were published in economics journals for a first half, and non-economics journals for the second half (mainly Ecology journals but also Sustainability science journals). This testifies of the explicit integration of other disciplines in the study of conservation issues, while modeling approaches remain anchored by the economics methodology. On the contrary, articles from the last two groups (related to the harvesting paradigm) display a disciplinary distribution skewed towards Economics journals. This dominance of economics journals is consistent with the methodological specifications of the bioeconomic models within these corpus, which are directly in line with economic theory.

Eventually, the second group of our overall cartography displays an over-representation of European researchers (63% of the corpus, compared to 37% of the database) as well as an under representation from North-American research. A European strand thus emerges out of this corpus, led by Drechsler, Wätzold and Mouysset¹¹, focusing on biodiversity conservation in agricultural settings.

¹¹These authors are the top 3 of the most credited authors in the corpus, thus each representing

The other corpus do not display a significantly different geographical distribution from the full database.

6 Discussion

1 Bioeconomic models as tools to manage social-ecological systems

Designing sustainable development paths in the context of the ecological crisis requires identifying sustainable dynamics or equilibria, which could be defined as the long-term behaviors needed to maintain both socioeconomic and ecological systems. To characterize such sustainable states and their underlying drivers, an adequate understanding and representation of the relationships between society and ecosystems are required ([IPBES, 2016](#)). In this respect, we are forced to deal simultaneously with considerations of economic and ecological dynamics as well as their mutual interactions by integrating feedback effects and interdependences between the ecological and socioeconomic systems ([Carpenter et al., 2009](#); [Figueiredo and Pereira, 2011](#); [Perrings, 2011](#)). Since the modelling communities in the natural and social sciences are relatively isolated from each other, substantial research efforts have to be done to overcome linguistic, epistemological, technical and other hurdles between the disciplines to provide a consistent framework ([Rindfuss et al., 2004](#))

The bioeconomic mathematically-based method reviewed in this article fits perfectly with this objective. By modelling complex structures and interactions within social-ecological systems, this type of model investigates how people perceive their well-being, how people make decisions to enhance their well-being, how it is affected by environmental conditions, how people may adapt their behaviour as their environment changes and how policies might be designed to be ecologically and economically efficient and socially accepted.

The set of models we reviewed in this article reveals nevertheless a plurality in the way that individuals and groups value nature, especially pending on contexts and scales. By combining methodology-based and narrative-based analysis, our cartography showed that this plurality of understandings, to perceive the interactions between human and nature within the social-ecological systems, can be embodied within two main and opposite human-nature paradigms. The first one is related to harvesting while the second one to conservation.

The harvesting paradigm resonates with the early nature paradigm where nature is mainly seen as wild nature, grasped in its emblematic dimension. In this

10% or more of the publications.

context, living elements are linked to socioeconomic decisions without considering any of their ecological features nor economic particularities except their direct and visible benefits. It can be understood as the modernization of the "conservationist" movement in the United States in the late XIXth century, championed by Gifford Pinchot, who conceived Nature through its instrumental value for humans and adopted a model of rational planning for resource use ([Banzhaf, 2019](#)). This conception of nature underlies international institutions such as the World Wide Fund for Nature founded in 1961 or emblematic public policies such as the Endangered Species Act established in 1973 in the USA.

Beside this harvesting paradigm, the conservation group derives from a second paradigm stemming from the early "preservationist" movement, whereby Nature should be conserved for its own sake, led by naturalist John Muir ([Banzhaf, 2019](#)). This paradigm was reshaped by the concept of biodiversity in the 90's. Popularized in 1992 by the United Nations Earth Summit in Rio, the concept of biodiversity captures both the notion of biological diversity and its ongoing situation of crisis ([Robin and Libby, 2011](#)). This new concept has implied two switches. First, it appears crucial to extend the conception of biological diversity by incorporating genetic, population, and ecosystem diversity to the classical species diversity, and by moving from emblematic nature to common and ordinary nature. Second, such an explicit context of ecological crisis calls unambiguously for protection.

Today both paradigms coexist in the mathematically-based bioeconomic modeling framework. In this perspective, this method seems to offer an up-to-date and promising context to think and assess the management of terrestrial social-ecological systems. In the 1990's and 00s, the biodiversity crisis spurred social demand and agenda setting in environmental policy, thus accelerating the development of this method.

2 Discussion about the recent and on-going decline

Despite a large increase in the 90's and 00's, our review reveals a decline over the last few years (since 2008), which is surprising as the question of sustainably managing terrestrial social-ecological systems is far from being solved. Indeed this situation is quite unusual for a methodology associated with such a burning issue and which seems to offer an up-to-date framework. This recent decrease might suggest a lag between the questions opened by the sustainable management of terrestrial social-ecological system and the answers brought by this bioeconomic mathematical modeling method. Understanding such a lag constitutes a determinant methodological stake with two implications: (i) defining the insights of the bioeconomic methodology to the knowledge in the field of economics, (ii)

identifying perspectives of development of this methodology in regards with this ecological crisis.

A crucial perspective to investigate these questions is to proceed to a similar analysis of the neighboring methodological corpuses, including agro-ecological and land use change models, declarative bioeconomic models, simulation-based models (*i.e* without mathematical specifications, biodiversity and ecosystem service quantification articles etc). Studying the technical features of such methods and their changes would be a determinant piece of information to understand the coexistence of the different methods aiming at studying the management of terrestrial social-ecological systems. To complete this analysis, it may be informative to extend this overall comparative analysis to the habitat, by distinguishing marine and terrestrial habitats¹². Indeed the nature paradigm might slightly differ among these habitats, due to differences in the intensity of the competition between nature and society. The most adequate methodology to investigate the bioeconomic question can thus be different. This might be an explanation of the differences in the development of such a methodology for marine and terrestrial resources

Among the neighboring methods, one of them merits specific attention, namely the correlative models (see for example [Leclère et al. \(2020\)](#), who use a wide array of Integrated Assessment Models (IAMs) and biodiversity models (BDMs) to evaluate biodiversity decline scenarios). This method, widely popular in ecological sciences, also fits with social expectations for decision-makers regarding social-ecological system management ([IPBES, 2016](#)). Indeed, there is a social demand for data-driven models since these ones look more realistic and reliable to make management decision. In this perspective, the correlative approaches based on large datasets might look more accurate to design concrete public policies and management strategies than the process-based model such as the mathematical bioeconomic models. Moreover, the IPBES report highlights the need for user-friendly modeling tools to be successfully used by decision-makers. Correlative models are based on mathematical tools since statistical analysis relies on mathematical foundations. However, by emphasizing the results on the data instead of the mathematical foundations, such tools look more understandable than process-based approaches which emphasize the equations of processes and frequently provide results in terms of stylized facts. By emphasizing the central-

¹²Terrestrial bioeconomic models applied to terrestrial biodiversity management seem to represent a smaller yet comparable share of the literature as bioeconomic models applied to marine ecosystems. A search on SCOPUS yields 418 articles (respectively 212) against 407 (respectively 229). (We used the following query : *TITLE-ABS-KEY (bioeconomic AND model)* as well as *TITLE-ABS-KEY (bioeconomic AND modeling)* selecting the appropriate keywords pertaining to both sub-fields. Because the selection was operated using keywords, some articles applied to marine ecosystems can remain. The aim of these numbers is to gauge the magnitude order of the different literature strands

ity of mathematics in the method, the process-based models are less accessible to a non-specialist audience. These specificities might explain the relative decline of mathematical bioeconomic models which may have benefited integrated data-driven approaches.

Despite substantial advantages, data-driven correlative approaches have to deal with several difficulties. First, they usually rely on specific and user-friendly softwares. While many tools are open-source and freely accessible, access to proprietary softwares can be attained through financial support from funding sources such as the UN, the World Bank and the Convention on International Trade in Endangered Species ([IPBES, 2016](#)). Similar problems emerge to access some datasets since some of them remain costly. To overcome this difficulty, it is possible to use different platforms collecting biodiversity and ecosystem services datasets at large scale. However their use is not always easy since inconsistencies and a lack of complementarity persist and interfere with an optimal use of the data. Second, correlative models are calibrated with existing data. Therefore it is impossible to model unexpected effects which never happened in the past. Yet there is an urgent need from the stakeholders to identify early tipping points as proxy of regime shifts to avoid crisis before its emergence ([Zimmermann et al., 2009](#)).

Mathematical bioeconomic models offer promising answers to these two limits. First the approach is less dependant to datasets and softwares. Second the modeling of the explicit processes makes possible the integration of events out of the set of calibration, including crisis effects. In this context, we understand that bioeconomic models offer a complementary tool to the popular correlative models. Actually, a variety of modelling approaches may often be available for addressing the social-ecological system research questions. As mentioned by the IPBES report ([IPBES, 2016](#)), debates about the use of correlative versus process-based models are frequently polluted by misconceptions about the utility of these models. Yet, many modelling exercises have clearly illustrated the benefits of combining multiple model types since it improves the quality of the management of social-ecological systems by providing complementary understandings of the research question and limitating uncertainty ([Cheaib et al., 2012; Gritti et al., 2013; van Oijen et al., 2013](#)).

Due to this complementarity, we support that the ongoing decline of the mathematical bioeconomic method is not desirable and merits to be reverse. This reversal calls therefore for improvements in adequacy between the method and the social demand from decision-makers.

3 Challenges for bioeconomic models

Our methodological cartography makes possible to observe that bioeconomic models have gradually improved over time. First, we highlighted the evolution of the model features in both paradigms. While earlier models lacked an inclusion of uncertainty, whether through a stochastic component in the ecological or economic model or a sensitivity analysis on the model parameters, they gradually evolved to take into account several forms of uncertainty, for example in the form of stochastic population dynamics ([Bulte and Kooten, 1999](#)) or uncertainties in the value of ecosystem services ([Augeraud-Véron et al., 2019](#)). However, in line with [Drechsler \(2020\)](#), it appears that uncertainty remains to be systematically integrated and considered as a major modeling component.

Second, the bioeconomic method has gradually encompassed the spatial dimension, and recognized its importance in both model components. Following [Sanchirico and Wilen \(1999\)](#), the spatial component has been integrated on the ecological side, in the form of a "patchy resource", paving the way for spatially differentiated population dynamics, namely meta-populations. The use of spatially differentiated data for ecological processes, including different habitat qualities, has gradually increased as well as spatially differentiated economic components. Third, a variety of actors have been gradually included, ranging from a single resource owner to complex property rights settings (local conservation agencies and communities competing for the resource ([Skonhoft, 1998](#)) , neighboring farmers facing a common threat of invasive species ([Fenichel et al., 2014](#))) and political settings (with the integrated management, by a social planner, of heterogeneous farmers ([Mouysset et al., 2014](#)) through public policies). Eventually, the process-based models we reviewed have gradually included some key components of correlative-methods, in order to be applied to real world settings and to provide policy guidance and evaluation. For example, species-area relationships have gradually been included. Notably [Davis et al. \(2006\)](#) investigate efficient conservation measures in a utility maximizing framework, where they used a species-area relationship to measure the value of conservation in the Sierra Nevada bioregion of California, instead of designing a fully tractable species model. These different improvements testify ways to better fit real-world conditions and thus answer to social expectations from decision-makers managing concrete terrestrial social-ecological systems.

However these improvements do not totally overcome the challenges. For example, explicit geographic economic components remain mostly absent from our sample and constitute an on-going challenge of bioeconomic models. Likewise, the articulation of actors within social-ecological systems remains an unsolved question since a wider variety of actors, especially local stakeholders and house-

holds, could better be taken into account.

Beyond these methodological examples in direct lines with our cartography, we can point out some more general fruitful avenues for future methodological improvements of bioeconomic models to better fit with stakeholders' needs. First, regarding the human-nature paradigm: indigenous standpoints and different value cultural systems should more systematically integrate into bioeconomic models which remain for now grounded in Western-Occidental ethics ([Kneese and Schulze, 1985](#)). Especially, spirituality underpinning the value of nature could be integrated, although existing works such as [Lopes and Atallah \(2020\)](#) investigate trade-offs and perform valuations of spirituality based on the framework pioneered by [Krutilla \(1967\)](#). Second, regarding the methodology : a more systematic use of statistical approaches developed by correlative models to calibrate or interpret the process-based modeling might provide the in-real anchorage desired by decision-makers. Third, regarding the communication platform. Inspired by correlative models, it would probably be strategic to provide easy-to-use softwares which generate the results of simulations pending on a set of parameters that the user can change. Even if the results can be expressed in stylized facts, this way of communication is not operational for practitioners.

4 Technical limitations and perspectives

Our cartography relies on a review which might be discussed at several levels. First of all, in spite of our efforts, we could not access *Forest Science*, a leading review in forestry. Therefore, forestry is underepresented in our sample. That being said, a sizeable share (16%) of our sample focuses on the topic. Second, our review procedure encompasses several criteria with a high level of generality. This methodological choice aims at filling a gap in the literature since most of the reviews focus on a smaller level. However, refining our methodological criteria, such as the ecological models we considered (population dynamics v. others) or the framing of uncertainty (absent or present, it can be refined through an analysis of stochasticity, sensitivity analysis) might help to precise the groups depicted by the MCA and thus help to connect our cartography to existing reviews, such as [Eppink and van den Bergh \(2007\)](#) and [Castro et al. \(2018\)](#). Third, our methodological characterization relies on a K-modes algorithm, an extension of K-means. Although well performing, the potential of such algorithms is limited by the number of observations. Given the number of variables and potential values, our sample size could limit the power of the K-modes algorithm in retrieving the structure of the dataset. In this perspective, other classification algorithms merit investigation to assess the robustness of our cartography. Fourth, our narrative elicitation with text data can be viewed as coarse, given that it only

encompasses word counts and lexical groups. Further analysis should deploy a more comprehensive method to analyze narratives quantitatively, and select a limited sample to conduct in depth analysis of narrative structures. Moreover other semantic fields might be investigated to precise the narrative underlying the different groups.

Finally, we adopted here a method perspective to cartography the bioeconomic models. We complete this perspective with some sociological information related to geographical origins of the researchers and to disciplines in which articles are published. However at this stage this information remains scarce and merits to be deepened by a specific sociological analysis. For example, connections between labs and institutions as well as between researchers measured by professional relationships (PhD student & supervisor), and citation networks such as in [Smessaert et al. \(2020\)](#) could yield interesting results. Regarding disciplinary aspects, an epistemological discussion would constitute a valuable addition in this economic field at the interface with natural sciences. All these elements might be used to precise our cartography but might also be the basis of new cartographies which could be confronted to the methodological ones. Convergences or divergences between technical, sociological and epistemological stakes might be this way highlighted.

Acknowledgements

We are thankful to two anonymous referees and the editors for their valuable suggestions and comments.

A Appendix

A Article selection equation on SCOPUS

In order to select articles, we performed a research on SCOPUS using the following query :

TITLE-ABS-KEY (biodiversity AND (ecological-economic OR bio-economic OR economic) AND modeling) AND (LIMIT-TO (SUBJAREA , "ENVI") OR LIMIT-TO (SUBJAREA , "AGRI") OR LIMIT-TO (SUBJAREA , "SOCI") OR LIMIT-TO (SUBJAREA , "EART") OR LIMIT-TO (SUBJAREA , "ECON") OR LIMIT-TO (SUBJAREA , "ENER") OR LIMIT-TO (SUBJAREA , "ENGI") OR LIMIT-TO (SUBJAREA , "COMP") OR LIMIT-TO (SUBJAREA , "MATH") OR LIMIT-TO (SUBJAREA , "DECI") OR LIMIT-TO (SUBJAREA , "BIOC") OR LIMIT-TO (SUBJAREA , "MULT") OR LIMIT-TO (SUBJAREA , "BUSI") OR LIMIT-TO (SUBJAREA , "ARTS"))

TITLE-ABS-KEY (bioeconomic AND modeling) LIMIT-TO (SUBJAREA , "ENVI") OR LIMIT-TO (SUBJAREA , "AGRI") OR LIMIT-TO (SUBJAREA , "SOCI") OR LIMIT-TO (SUBJAREA , "EART") OR LIMIT-TO (SUBJAREA , "ECON") OR LIMIT-TO (SUBJAREA , "ENER") OR LIMIT-TO (SUBJAREA , "ENGI") OR LIMIT-TO (SUBJAREA , "COMP") OR LIMIT-TO (SUBJAREA , "MATH") OR LIMIT-TO (SUBJAREA , "DECI") OR LIMIT-TO (SUBJAREA , "BIOC") OR LIMIT-TO (SUBJAREA , "MULT") OR LIMIT-TO (SUBJAREA , "BUSI") OR LIMIT-TO (SUBJAREA , "ARTS"))

TITLE-ABS-KEY (bioeconomic AND model) LIMIT-TO (SUBJAREA , "ENVI") OR LIMIT-TO (SUBJAREA , "AGRI") OR LIMIT-TO (SUBJAREA , "SOCI") OR LIMIT-TO (SUBJAREA , "EART") OR LIMIT-TO (SUBJAREA , "ECON") OR LIMIT-TO (SUBJAREA , "ENER") OR LIMIT-TO (SUBJAREA , "ENGI") OR LIMIT-TO (SUBJAREA , "COMP") OR LIMIT-TO (SUBJAREA , "MATH") OR LIMIT-TO (SUBJAREA , "DECI") OR LIMIT-TO (SUBJAREA , "BIOC") OR LIMIT-TO (SUBJAREA , "MULT") OR LIMIT-TO (SUBJAREA , "BUSI") OR LIMIT-TO (SUBJAREA , "ARTS"))

B Lexical groups

Agriculture : agriculture, agricultural, crop, rangeland, livestock, forage, fallow, farmland, grassland, oats, agri, farmers, grazing, crop, livestock, farming, wheat, crops, farm, cropping, rangeland, grazing, stocking, alfalfa, wheat, agro, crofting,

pastures, ranchers, range, grasslands

Forest : Trees, stand, tree, forest, forestry, basal, spruce, even-aged, uneven-aged, forests, timber, diameter, wood, pine, faustmann, volume, reforestation, rotation, rotational, acacia, forested, lichen

Invasive species : Invasive species, invasive, rabies, invasion, invasivespecies, invader, coevolution, non-endemic, nis, eradication, pest, mountainpinebeetle, weevil, disease, gypsymoth, weevil, oats, weed, herbicide, invader, rabies, pathogens, invasivespecies, indigenous, barrier, infestation, alien, beaver, calvescens, eradicate, host, resistance, infestations, pesticide, pests, invasions, weed, weeds , pest, nonindigenous, pathogen, invaders, spartinaalterniflora, spartina, beetle, endemic, emeraldashborer, beetles, avena, rodent, serratedtussock, tuberculosis, miconiacalvescens, vaccine, insects, spread, vector-borne, epidemiology, quarantine, trap

Endangered/remarkable species : Endangered species, remarkable, trophy, tiger, endangeredspecies, warbler, moose, illegal, threatened, threats, endangered, elephants, butterfly, wildlife, game, poachers, wolf, reindeer, poaching, wolves, elephant, bushmeat, ivory, black, hunt, canislupus, hunters, bear, serengeti, tigers, deer, rhino, extinction, endangered

Policy : Policy, subsidy, tax, tradable, subsidies, instruments, policy, policies, payments, taxes, market, markets, incentives, payment, permits, taxes, incentive, funding, budget, budgets, conflict, conflicts, bonus, planner, taxation, property, market-based, contracts, interventions, intervention, strategy, propertyrights, tax-subsidy

Risk : risk, uncertainty, insurance, markov, option, resilience, stochastic, probabilities, uncertain

Conservation : conservation, park, reserve, sites, restoration, planning, conservationplanning

Harvesting : harvesting, harvests, harvest, hunting

C Supplementary tables

	<i>Biodiversity measure</i>	<i>Proxy measure</i>	<i>Biodiversity state variable</i>
Biodiversity characterization	Per se Proxy	No - per se Habitat Economic activity Conservation budget Not specified	Population Species Community (species & population) Not specified
	<i>Biological diversity</i>	<i>Biodiversity contribution level</i>	
	Functional Genetic Functional & genetic	Single species Multiple species Unknown	
Ecological specifications	<i>Dynamics</i>	<i>Spatiality</i>	<i>Uncertainty</i>
	Pop. dyn. Other	Explicit Implicit Absent	Stochastic Deterministic
Bioeconomic linkage specifications	<i>Biodiversity monetarization</i>	<i>Bioeconomic problem</i>	<i>Biodiversity stake</i>
	Yes No	Cost-benefit analysis Cost-effectiveness analysis	Constraint Objective Other
Economic specifications	<i>Dynamics</i>	<i>Spatiality</i>	<i>Uncertainty</i>
	Static Dynamic	Explicit Implicit Absent	Stochastic Deterministic
General characteristics	<i>Solving method</i>	<i>Data anchorage</i>	<i>Model use</i>
	Closed form Numerical solution Both	Theoretical Empirical Both	Normative Descriptive
	<i>Equilibrium</i>		
	General Partial		

Table 1.A: List of the methodological criteria and their related items used to perform the methodology-based cartography. In grey stand the criteria which have been excluded after the sensitivity analysis of the MCA.

<i>Economic journals</i>	<i>Count(60%)</i>	<i>Ecology journals</i>	<i>Count(26%)</i>
Ecological Economics	44	Ecological Modelling	15
American Journal of Agricultural Economics	19	Biological Conservation	11
Journal of Environmental Economics and Management	19	Ecological Applications	11
Environmental and Resource Economics	15	Canadian Journal of Forest Research	6
Resource and Energy Economics	15	Journal of Applied Ecology	6
Land Economics	11	Conservation Biology	4
Environment and Development Economics	6	Forest Science	4
Journal of Bioeconomics	6	Diversity and Distributions	3
Journal of Environmental Management	6	Forest Ecology and Management	3
Agricultural Economics	4	Biological Invasions	2
Review of Agricultural Economics	4	Conservation Letters	2
Australian Journal of Agricultural Economics	3	Ecology Letters	2
Journal of Economics	3	Biodiversity and Conservation	1
Journal of Forest Economics	3	Commonwealth Forestry Research	1
Agricultural and Resource Economics Review	2	Ecological Indicators	1
American Economic Review	2	Ecology and Society	1
European Review of Agricultural Economics	2	Environmental Entomology	1
Journal of Economic Theory	2	European Journal of Forest Research	1
Canadian Journal of Economics	1	Journal for Nature Conservation	1
Computational Economics	1	Journal of Economic Entomology	1
Econometrica	1	Journal of Forestry Research	1
Economic Inquiry	1	Journal of Theoretical Biology	1
Economic Theory	1	New Forests	1
Journal of African Economies	1	Silva Fennica	1
Journal of Agricultural and Applied Economics	1	Theoretical Population Biology	1
Journal of Agricultural and Resource Economics	1	Wildlife Biology	1
Journal of Agricultural Economics	1		
Journal of Economic Dynamics & Control	1		
Kiel Working Papers	1		
		<i>Sustainability science journals</i>	<i>Count (10%)</i>
Management Science	1	Natural Resource Modeling	10
MPRA Papers	1	Agricultural Systems	3
MPRA Working Papers	1	Environmental Modeling and Software	2
Oxford Economic Papers	1	Environmental Modeling and Assessment	2
Review of marketing and Agricultural Economics	1	PNAS	2
RFF Discussion papers	1	Australian Journal of Experimental	1
		Agriculture	
Social Choice and Welfare	1	Central European Journal of Operations	1
		Research	
Socio-Economic Planning Sciences	1	Climatic Change	1
Spatial Economic Analysis	1	EcoHealth	1
The Australian Journal of Agricultural and	1	Ecosystem Services	1
Resource Economics			
The B.E. Journal of Economic Analysis and Policy	1	Journal of Biological Dynamics	1
The Journal of Political Economy	1	Journal of Environment Management	1
Western Journal of Agricultural Economics	1	Land Use Policy	1
		Nature	1
<i>Applied Mathematics journals</i>	<i>Count (4%)</i>	Non Linear Analysis	1
Journal of Mathematical Analysis and Applications	2	Operations Research	1
Journal of Mathematical Biology	2	PLoS One	1
Applied Mathematics Letters	1	Proceedings of the Royal Society	1
Biometrics	1	Regional Environmental Change	1
Bulletin of Mathematical Biology	1	Science	1
Computers and Mathematics with Applications	1		
Journal of Optimisation Theory and Applications	1		
Mathematical Biosciences	1		
Mathematical Biosciences and Engineering	1		
Mathematical Models and Methods in Applied Science	1		

Table 1.B: Journal distributions among disciplines

D Supplementary figures

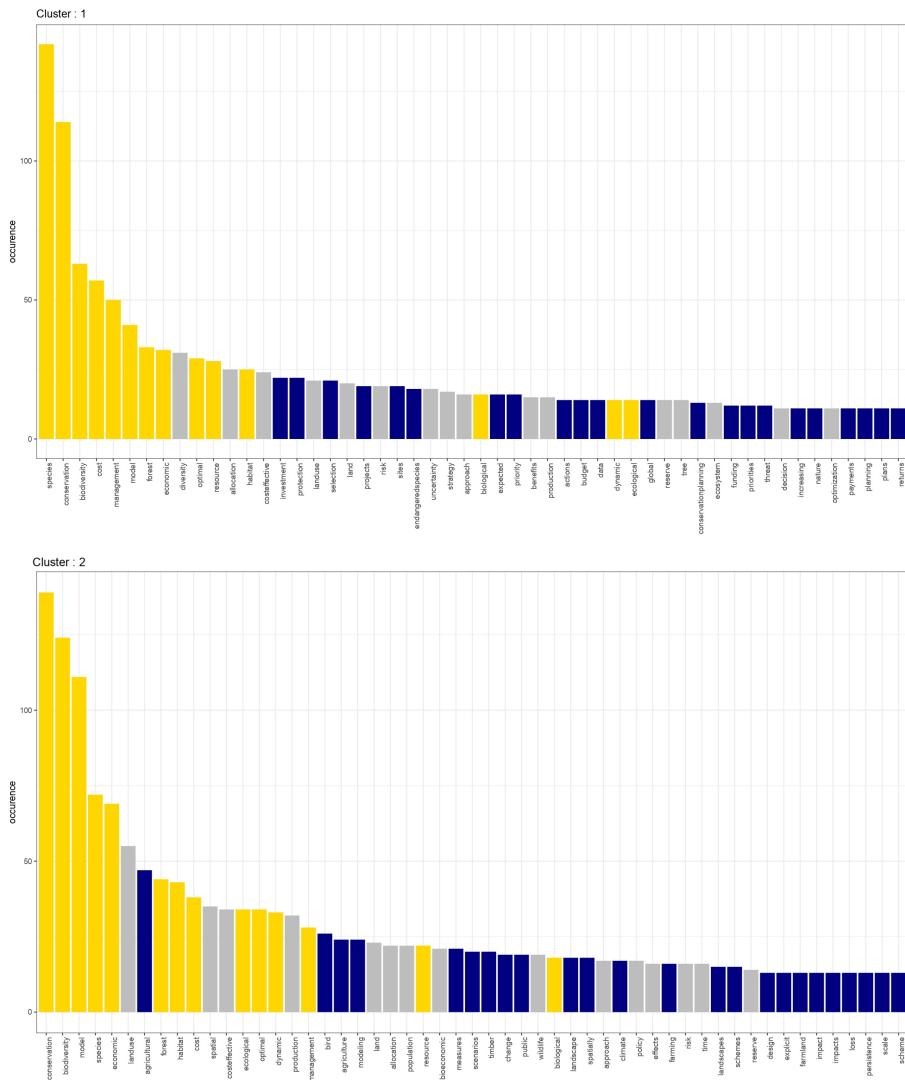


Figure 1.A: Distribution profiles of the 50 words the more frequent for methodology-based groups 1 and 2

In yellow stand the words in common among the 4 profiles. On the opposite in blue stand the words specific to a profile

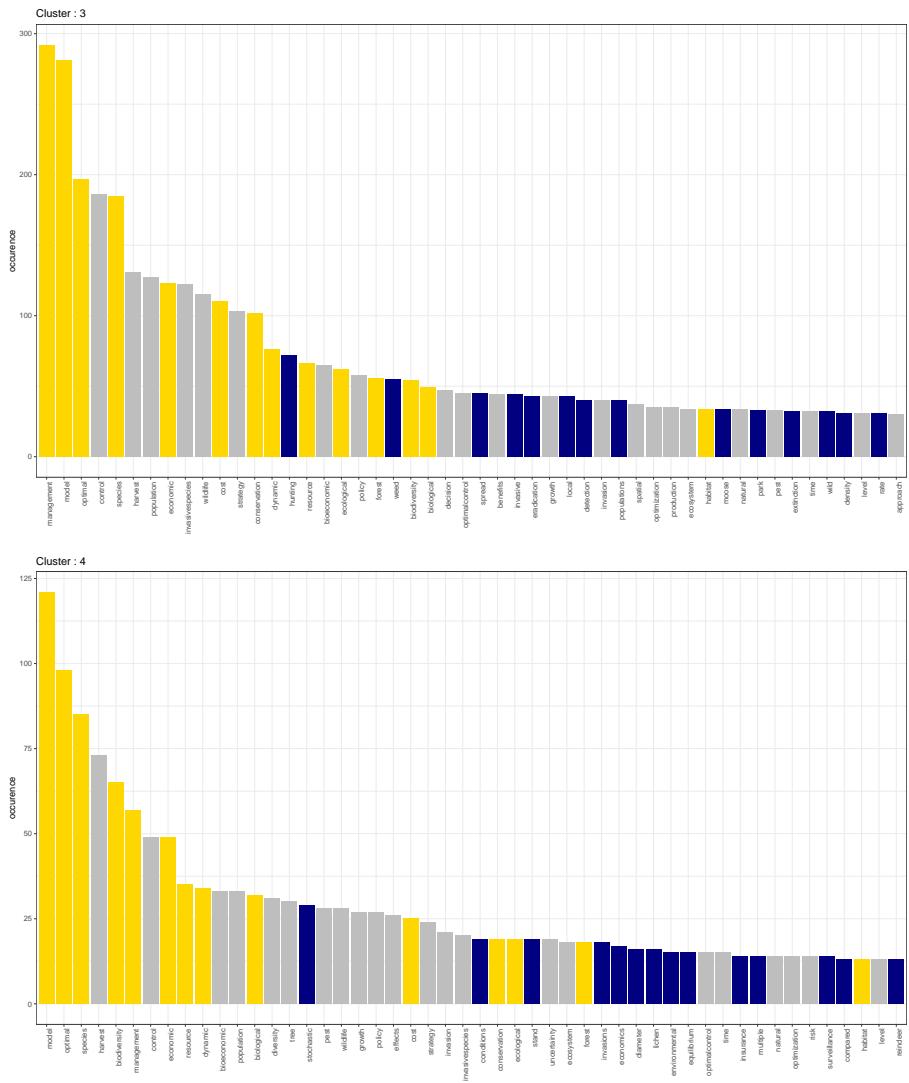


Figure 1.B: Distribution profiles of the 50 words the more frequent for methodology-based groups 3 and 4

In yellow stand the words in common among the 4 profiles. On the opposite in blue stand the words specific to a profile

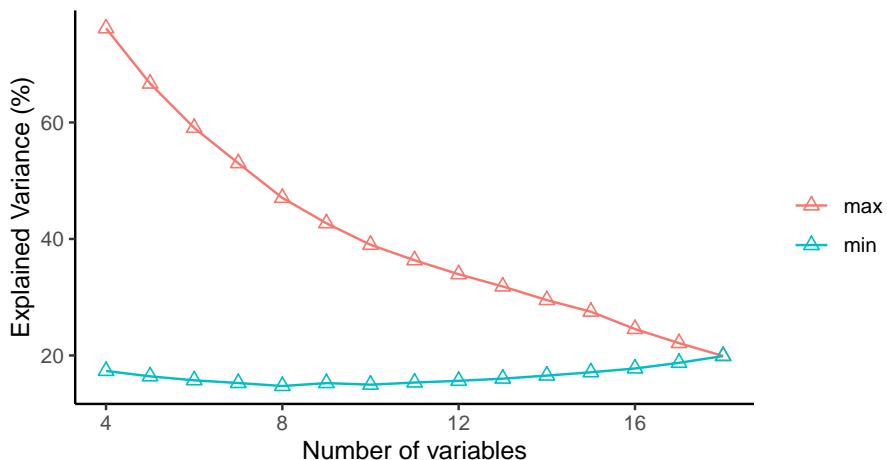


Figure 1.C: Explained variance function of the number of methodological criteria used in the MCA.

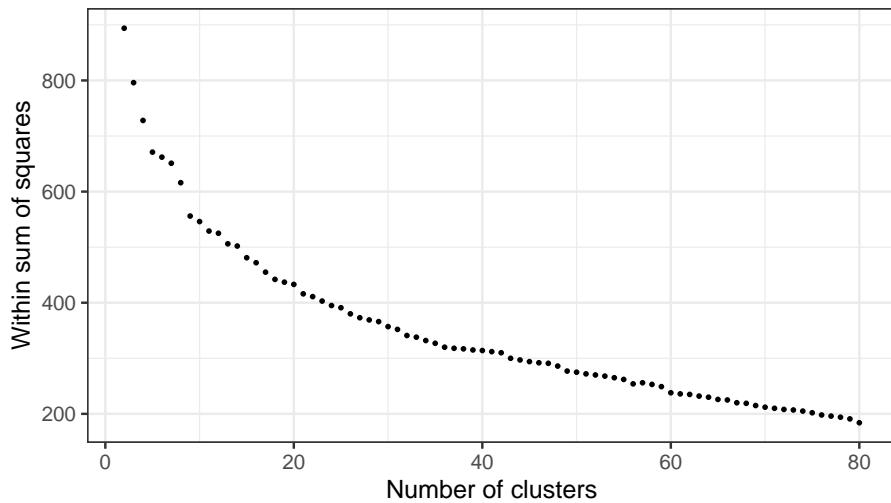


Figure 1.D: Cost function of the K-modes.

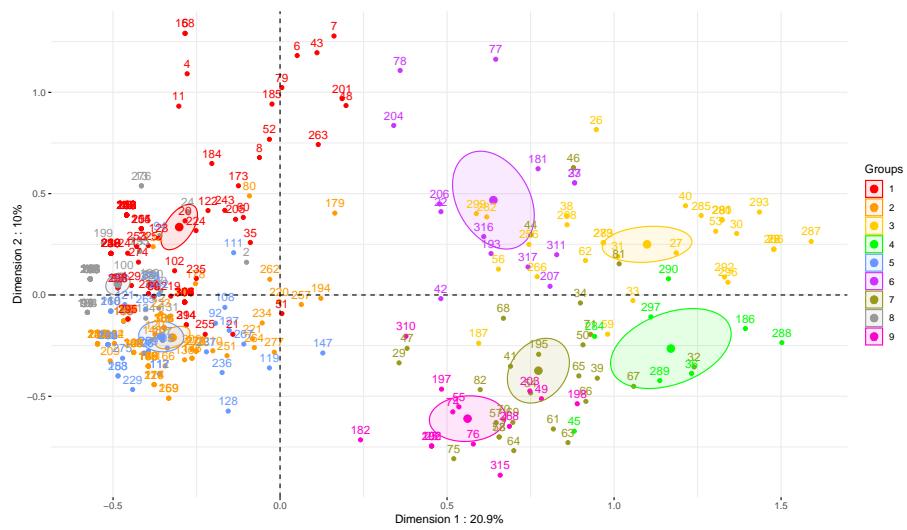


Figure 1.E: Multiple Correspondence Analysis (MCA) running on 12 methodological criteria and 9 groups.

Chapter 2

The wildfire-habitat connectivity dilemma: a graph theoretical approach to landscape management

This article is co-authored with L. Mouysset and under review at

Abstract

Background: Fuel treatment operations help to mitigate the spread and severity of wildfires in numerous ecosystems. As they aim at fragmenting the fire landscape, they also fragment wildlife posing a dilemma for land managers. We use graph theory on simulated to gain a general understanding of the allocation of treatments over space and time and the corresponding landscape properties with various habitat connectivity targets.

Results: Our results show that all initial landscapes converge to steady-state landscape cycles. Optimal trajectories significantly reduce wildfire risk while safeguarding habitat connectivity. As the policy budget increases, more risk reduction is achieved, albeit with a decreasing marginal efficiency. As habitat targets increase, increasing the budget is of no effect, and risk increases. Landscapes are less risky, more fragmented, and diverse when the budget is large and biodiversity targets are low, while they are more compact and less diverse when the opposite is true. Treatment allocation follows graph centrality measures, and central cells are treated first. When biodiversity targets increase, central cells are no longer treated as they decrease habitat connectivity. Treatment is reshuffled to the edges of the landscape.

Conclusion: Computational experiments generalize existing results. Using graph theory, general insights can be gained, and help managers faced with multiple objectives in forested landscapes. From a policy perspective, in the face of climate change, increasing treatment budgets should be a priority to avoid increasing damages. A key guideline is treating *adolescent* successional stages and fragment *mature* patches to mitigate risk and guarantee the connectivity of wildlife habitat.

Keywords : Fuel treatment, connectivity, wildfire risk, wildlife habitat, spatial optimization, graph theory

- We have a problem at extending
- Edge effects
- Difficulty to extend the critical node detection problem with no special structure graphs
- Betweenness centrality is not enough

Solution : get rid of the large scale and focus on the small scale, and say future research avenues involve understanding the transitional path.

1 Introduction

Hazardous and intense wildfires destroy forest cover¹, threaten forest resilience and can cause ecosystem shifts, ranging from changes in forest structure to changes towards non-forest ecosystems (Coop et al., 2020). Additionally, intense wildfires cause human damages, in the form of direct asset losses: in 2018, wildfires in California have caused \$ 27 billion (Wang et al., 2021). Indirect costs are also of concern, especially related to wildfire smoke : increases in PM 2.5 concentrations have important health impacts (Burke et al., 2023; Heft-Neal et al., 2023), smoke directly affects recreation values in the US, amounting to \$USD 2.3 billion in welfare losses (Gellman et al., 2024). Aside from directly measurable costs, wildfires also cause dramatic impacts on biodiversity across taxa (Wintle et al., 2020), through direct population losses and durable habitat disruption (Ayars et al., 2023).

In a business as usual scenario in terms of forest management, wildland-urban interface expansion and climate change, these direct and indirect costs and damages to both humans and non-humans are expected to increase drastically. Decades of wildfire suppression have created a “wildfire deficit”, which increases the probability, extent and severity of wildfires in the western United States (Kreider et al., 2024). European forests are not adapted to climate change induced wildfire risks (Khabarov et al., 2014), in terms of species composition and use of fuel management operations. Mechanical thinning, prescribed burns, and sometimes, logging, have been leveraged to decrease the fuel load in risky areas and theoretically decrease the probability and severity of burns upon wildfire occurrence². In numerous regions, such as conifer forests in California (Vaillant et al., 2009; Kalies and Yocom Kent, 2016; Low et al., 2023), eucalypt forests in South Western Australia (Burrows and McCaw, 2013; Boer et al., 2009; Florec et al., 2020), southern Europe (Fernandes et al., 2013), evidence shows that fuel treatments, can mitigate wildfire intensity and spread. Land management agencies have historically implemented these policies in Australia (Burrows and McCaw, 2013), Europe, and the United States (and are projected to ramp up, for example under the Infrastructure Investment and Jobs Act of 2021 in the US). While potentially useful, the use of these treatments is still hindered by numerous obstacles (Miller et al., 2020) and remains insufficient³. Additionally, the extension

¹From 2001 to 2023, forest loss attributed to wildfires amounted to 138 million hectares (roughly 33% of the surface of the European Union) (Tyukavina et al., 2022)

²The efficiency of these measures depends on environmental and terrain variables. For example, prescribed burns are efficient every 1-4 years in reducing risk and severity only in the case of non-extreme weather conditions, and when the terrain ruggedness is limited (Bradstock et al., 1998)

³However, recent bills have been passed in the US (Infrastructure Investment and Jobs Act of 2021) and California to ramp up the use of prescribed burns - such as the bugdet act of 2022,

of wildland-urban interfaces (WUI) increases the extent of potential damages as well as ignition probabilities (Radeloff et al., 2018).

As global warming affects water supply and fuel moisture (Jolly et al., 2015; Abatzoglou and Williams, 2016; Ruffault et al., 2018), it is projected to increase the frequency, severity, and magnitude of wildfires (Dupuy et al., 2019; Wasserman and Mueller, 2023). Recent wildfire events in California (since 2018), in Australia (2019-2020), and in Europe (France, Portugal, Greece in 2022) have epitomized these trends. Moreover, wildfires and climate change are endogeneously linked in a positive feedback loop : large wildfires are of importance in the face of climate change; as they release large amounts of greenhouse gases (1.7GtC per year on average between 2003 and 2022) and reduce the extent of terrestrial carbon sinks (Zheng et al., 2023; Friedlingstein et al., 2023; Byrne et al., 2024).

In the face of a growing threat to human assets and biological diversity, increasing the efficiency of fuel treatments to manage multiple objectives is paramount. A decision framework that accounts for wildfire processes and biological diversity drivers is paramount to deliver policy recommendations that simultaneously achieve wildfire damage reduction and protect biological diversity (Driscoll et al., 2010). Among the decision levers, the extent and location of treatments are key variables.

By changing the structure of the landscape, fuel management operations may reduce the risk and associated damages of wildfires. Treatments achieve larger risk reduction when located close to the values at risk instead of being dispersed across the landscape (Ager et al., 2007; Williams et al., 2017; Florec et al., 2020). However, they also affect the structure of biodiversity habitat, notably, its structural connectivity (Taylor et al., 1993). Maintaining habitat connectivity, through wildlife corridors, landscape links, and ecoducts (Turner, 2005; Turner and Gardner, 2015), is instrumental in mitigating the biodiversity crisis. Species richness and diversity are intimately linked to landscape connectivity (Olds et al., 2012; Tian et al., 2017; Velázquez et al., 2019) and are necessary to maintain ecosystems in the future. Fragmentation, conditional on habitat surface being constant, may enhance biodiversity (Tischendorf and Fahrig, 2000; Hu et al., 2012; May et al., 2019). However, it is often accompanied with habitat loss, detrimental to biodiversity (Fahrig, 2003). The use of fuel management operations alters the structure of the landscape e.g. both habitat and matrix⁴, in terms of temporal and spatial variation in landscape configuration and composition. As habitat is altered, so is the surrounding matrix, which can impede species movement (Eycott et al., 2012; Kuefler et al., 2010) and alter evolution and selective regimes (Cheptou et al.,

committing \$2.8 billion to the Governor's Wildfire and Forest Resilience Action Plan - and limiting liabilities in the case of wildfire escape (see California Senate Bill SB-332) on private land.

⁴e.g. land use or cover, or environmental conditions that differ from either species' habitat or reference natural conditions (Fletcher et al., 2024)

2017).

The impact of fuel treatments on biodiversity remains a debated topic. Evidence suggests that maintaining a variety of vegetation types and ages on a patchy landscape maintains a 'fire mosaic' (Sitters et al., 2015) (e.g. landscape level variations in habitat types that provide habitat to an ecological community) or that fuel treatment can be beneficial to wildlife (Saab et al., 2022; Loeb and Blakey, 2021) and even restore local populations (Templeton et al., 2011). On the other hand, treating at too high a frequency may be detrimental to biodiversity (Bradshaw et al., 2018), as vegetation with extensive juvenile period may disappear, and fauna that rely on them as well⁵, or high frequency treatment favors the invasion of fire tolerant, fire-enhanced weed species (van Wilgen, 2013).

Hence, fragmenting the wildfire risk poses significant threats to biodiversity in forest landscapes. Nonetheless, there may exist a range of spatial allocation patterns that take into account the location of protected species and can reduce threats to both assets and biodiversity (Ager et al., 2007; King et al., 2008; Rachmawati et al., 2018).

Eventually, wildfire risk and potential damages pose a significant challenge in terms of policy-making. As wildfire risks and potential damages are spatially heterogeneous, and as wildfires spread, they create a large spatial externality. Indeed, individual risk reduction (e.g. self-protection) is hampered by the influence of neighbors on individual risk, which results in the under provision of risk reduction (Shafran, 2008; Costello et al., 2017). Additionally, in a risky (e.g. stochastic) context, risk aversion may further reduce self-protection this phenomenon when financial insurance is limited (Ehrlich and Becker, 1972)⁶. Finally, the magnitude of potential damages (Costello et al., 2017) as well as the large information requirements for efficient fuel treatment planning warrant a collective approach.

In this context, we study the spatial patterns of treatment allocation that diminish potential damages from wildfires in where fire spread is governed by patch connectivity, while safeguarding biodiversity habitat connectivity, from a central decision maker perspective.

A substantial literature has applied optimization techniques to tackle the spatial allocation of fuel treatments. Analytical (Finney, 2001), simulation-based (Finney, 2007; Rytwinski and Crowe, 2010) or mixed-integer programming techniques (Wei et al., 2008) have solved the allocation of treatments in a static framework. Given the dynamic nature of fuel growth, studies based on mixed-integer dynamic programming (Wei et al., 2008; Minas et al., 2014; Rachmawati et al.,

⁵For example, in Australia, species such as *Banksia baueri*, *B. nutans* and *B. baxteri* would disappear, threatening tammar wallabies, quokas and honey possums (Bradshaw et al., 2018)

⁶This is particularly the case in California, where repeated fire episodes have pushed insurers to spike contract premiums, or to not renew contracts- non renewal rates went from 11% in 2018 to 13% in 2021

2015, 2016) have studied the temporal and spatial allocation of fuel treatments on real and simulated landscapes. While they solve the spatial treatment allocation problem in forests, these articles fail to acknowledge the multiple uses and objectives land planners have to consider, such as habitat conservation. Several articles have devoted their attention to the spatial allocation of treatments while conserving habitat, and investigated the trade-offs between risk reduction and biodiversity conservation, using spatial heuristics (Calkin et al., 2005; Lehmkuhl et al., 2007) and linear programming (Williams et al., 2017; Rachmawati et al., 2018).

Most of the existing literature focuses on case studies and lacks a general interpretative framework to generalize its results. Graph theory offers a toolbox suited to analyze the properties of connected cells or patches of land with varying characteristics, and has extensively been applied in landscape ecology (Urban and Keitt, 2001; Minor and Urban, 2008; Rayfield et al., 2016). Conrad et al. (2012) and Jafari and Hearne (2013) use a specific graph theory algorithm - a network-flow model - to find the optimal subgraph of corridors connecting habitat areas. Their approach optimally connects patches of habitat spread across the landscape for a given species, in a reserve-network design problem fashion. Our approach adopts a more holistic perspective, as it emphasises the degree of connectedness between habitat cells, thus allowing for a multi-species and multi-scale perspective, instead of a corridor for a single species.

Recent research focusing on the allocation of fuel treatments has leveraged tools from graph theory (Matsypura et al., 2018; Pais et al., 2021a). Reconciling habitat and wildfire risk mitigation using graph theory is a recent research endeavor (Rachmawati et al., 2018; Yemshanov et al., 2022) and has focused on specific case studies.

In this article, we focus on the dynamic and spatial dimensions of the problem (thus abstracting from the stochastic components) and leverage graph theory to study the general patterns of treatment allocation emerging from a multi-objective, dynamic, and integer landscape management problem, governed by connectivity.

To do so, we first compare the optimal allocation of treatments using repeated static optimization and heuristic dynamic programming on a 5 period horizon on representative subsamples of small scale landscapes with an exhaustive range of habitat connectivity constraint. We show that for realistic biodiversity habitat constraint levels, the constraint imposed on the evolution of the forest results in similar structures for repeated myopic and dynamic optimization. Therefore, we analyse the treatment allocation and landscape structures emerging in the long run using repeated myopic optimization for all the possible initial landscape con-

figurations, in a graph theoretical framework. We explicit the trade-off between risk reduction and biodiversity habitat, in the form of a production possibility frontier (PPF). We characterize the landscapes using a range of ecological indicators and find general mechanisms and guiding principles applicable to a broad class of settings, to guide decision-makers and foster new efficient multi-objective graph theory algorithms. Finally, we test our predictions from a small scale landscape to simulated realistic large scale landscapes (10,000 cells) with varying composition and spatial autocorrelation, and compare them with different intuitive policy recommendations.

Our contributions are several. First, we provide a spatial framework to understand the trade-offs between wildfire risk reduction and biodiversity conservation. Second, we leverage the constraints imposed on a dynamic spatial system to show that repeated optimization performs relatively well compared to dynamic programming. Third, using graph theory, we derive general principles regarding the spatial characteristics of landscapes and treatments from an exhaustive set of theoretical landscapes to guide policymakers as well as future research in heuristics to reconcile conflicting land-based phenomena.

2 Methods

1 Model description

We consider landscapes represented by a regular grid of $n \times n$ standardized area cells in period t by \mathbf{A}_t with a forest seral stage succession module. Each cell $a_{ijt} \in \mathbf{A}_t$ with $\{i, j\} \in \{1, \dots, n\}^2$, at time t is characterized by a successional stage: *juvenile*, *adolescent*, or *mature*, which translates into 3 numerical age classes ranging from 0 to 2. Each transitionary seral stage has the same duration⁷, hence at each time step, it changes stage until it is in the *mature* stage, where it remains (eq 2.1)

We use a stylized representation of the link between vegetation age, habitat, and wildfire risk (figure 2.1).

First, we assume a cell offers suitable wildlife habitat once it is *adolescent* (eq. 2.2). Second, a cell can turn at critical risk of wildfire during a normal hot season when its successional stage is *mature* (eq. 2.3). We assume an Olsen-type model of flammability, where age class is the main predictor of flammability (Olson, 1963; McCarthy et al., 2001; McColl-Gausden and Penman, 2019). A cell remains at

⁷For example, in Australia, McColl-Gausden and Penman (2019) use quasi evenly spaced age classes for heathland, tall-mixed, foothills, forby and wet vegetation types (see table 1); on the other hand, in coniferous forests in Western US (Washington and Oregon), Thomas (1979) developed a successional stage description for wildlife habitat management, still used by the USDA. 40 year transitional classes can be made grouping *grass-forb*, *shrub-seedling* and *pole-sapling* together and *young*. *Maturity* is reached at 80 until 159 years old, where it mutates into *old-growth*

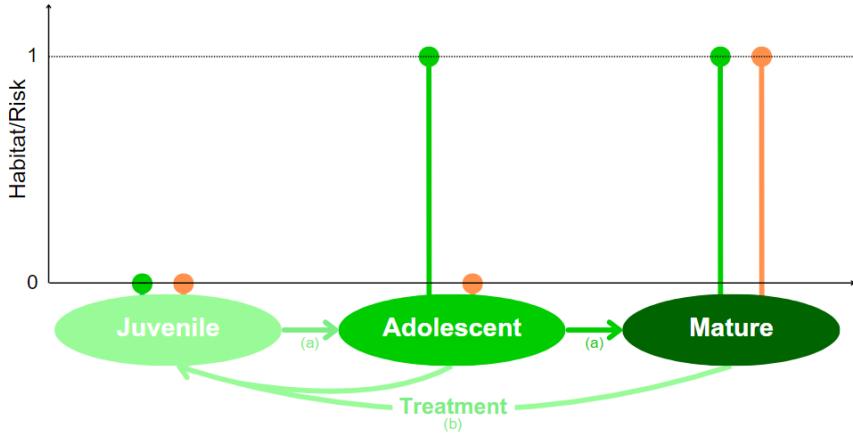


Figure 2.1: Illustration of the successional stages and the link between successional stage, habitat and wildfire risk using a discretized Olson-type relationship

At the bottom, the dynamics of the model are illustrated. First, successional stages transition (step (a)), then treatment is applied (step (b)). At the top, the link between successional stage, habitat and high risk. In green, a *habitat* variables turns to 1 when a cell is *adolescent*, and in orange, a *high risk* dummy turns to 1 when a cell turns *mature*

high risk as long as it is in the *mature* age class.

Finally, we consider fuel treatment to be a binary decision e.g. treatment is absent or present and there is no extensive margin, hence a treatment binary variable $x_{ijt} \in \{0, 1\}$ represents the treatment status in cell a_{ij} a time t . The decision maker first observes the transition to the next successional stage, then decides upon treatment. Treatment can happen at any successional stage : whether at an *adolescent* stage, in anticipation of a cell becoming *mature* and turning at *high risk* in the next period, or as an immediate strategy upon becoming *mature* and thus, *high risk*. Upon treatment, a cell successional stage is reset to *juvenile* (eq. 2.1). Figure 2.1 illustrates the dynamics of the model.

Given a patch a_{ijt} and treatment status x_{ijt} in period t , equation 2.1 summarises the successional dynamics, and equations 2.4 and 2.3 summarize the link between successional stage, habitat, and high risk: $\forall t, \forall \{i, j\} \in \{1, \dots, n\}^2$

$$a_{ijt+1} = \max((a_{ijt} + 1)(1 - x_{ijt}); 2) \quad (2.1)$$

$$\text{Habitat } (a_{ijt}) = \begin{cases} 1 & \text{if } a_{ijt} \geq 1 \\ 0 & \text{otherwise} \end{cases} \quad (2.2)$$

$$\text{Risk } (a_{ijt}) = \begin{cases} 1 & \text{if } a_{ijt} \geq 2 \\ 0 & \text{otherwise} \end{cases} \quad (2.3)$$

We use a network structure to apprehend the landscapes. From the matrix \mathbf{A}_t , we form two graphs: $\mathcal{B}_t = (V_{\mathcal{B}_t}, E_{\mathcal{B}_t})$, the graph of suitable habitat cells and

$\mathcal{F}_t = (V_{\mathcal{F}_t}, E_{\mathcal{F}_t})$, the graph of high risk cells. First, the vertices of each graphs are the suitable habitat cells e.g $V_{\mathcal{B}_t} = \{(i, j) \text{ such that } \text{Habitat}(a_{ijt}) = 1\}$ and the high risk cells, respectively e.g. $V_{\mathcal{F}_t} = \{(i, j) \text{ such that } \text{Risk}(a_{ijt}) = 1\}$.

Second, vertices are connected if they are within a Moore (or 8-cell) neighborhood of each other and share the same status. Therefore, notice that $\mathcal{F}_t \subset \mathcal{B}_t$. Figure 2.2 illustrate the mechanism from the landscape in matrix form \mathbf{A}_t with age classes ranging from 0 to 2, to graphs \mathcal{B}_t and \mathcal{F}_t .

We use this 8-cell neighborhood for evaluating biodiversity habitat and wild-fire risk within a common a spatial framework, using the same adjacency properties. Regarding biodiversity, we focus on general characteristics related to landscape structural connectivity rather than functional connectivity, as we are agnostic about effective species (Fahrig et al., 2011). We assume that species are able to disperse from one patch to another, and that habitat quality is uniformly distributed conditional on habitat being available.

We consider the wildfire risk through the lens of potential spread, influenced by fuel, wind direction and terrain. We abstract from wind patterns and terrain, to focus on fuel connectivity⁸. Consistent with the literature (see Peterson et al. (2009), Pais et al. (2021b); Gonzalez-Olabarria et al. (2023)), a wildfire can spread in any direction, conditional on neighbor cells with high risk.

To assess the connectivity of \mathcal{F}_t and \mathcal{B}_t , we use a global connectivity indicator. As connectivity can be measured in numerous ways in graph theory, we use this metric as it satisfies criteria pertaining to its evolution when vertices and edges are removed (Pascual-Hortal and Saura, 2006) when using graph theory applied to landscape ecology. Additionally, it offers a reformulation of the metric used in previous work closely related to ours (Minas et al., 2014; Rachmawati et al., 2016) (see appendix A for a demonstration). We define the global connectivity index of a given graph $\mathcal{G} = (\mathcal{V}, \mathcal{E})$ as⁹:

$$H(\mathcal{G}) = \text{card}(V) + 2\text{card}(E) \quad (2.4)$$

Let a *patch* be a collection of connected cells of suitable wildlife habitat. This indicator considers that a habitat cell is connected to itself (i.e, within a habitat patch, there is no barrier) and whether it is connected to other cells. It implies lower connectivity when the distance between habitat cells increases, attains its maximum value when a single habitat patch covers the whole landscape, indicates lower connectivity as the habitat is progressively more fragmented, considers negative the loss of a connected or isolated cell, and detects as more important the loss of

⁸Note that our framwork is amenable to prevailing wind patterns and terrain ruggedness, as the graph adjacency matrix can change from a Moore adjacency to any pattern influenced by environmental features

⁹With *card* being the cardinal operator in set theory and denotes the number of elements in a set

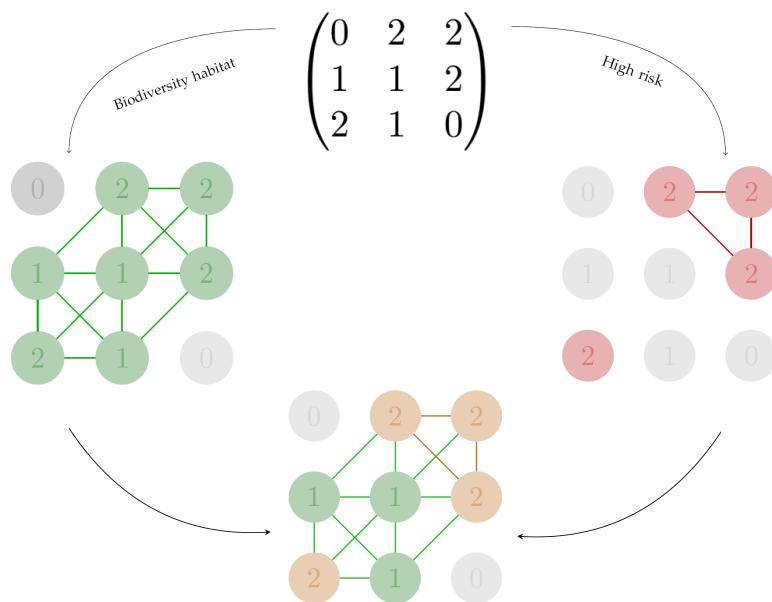


Figure 2.2: Illustration of the suitable habitat and high risk graphs for $n = 3$

The first layer is the values from a raster \mathbf{A}_t of age classes in a forest landscape. It is turned into two different graphs.

In the left graph, the green vertices are $V_{\mathcal{B}_t}$ and support biodiversity habitat, while on the right graph, red vertices are $V_{\mathcal{F}_t}$ display high risk. Green and red links are respectively $E_{\mathcal{B}_t}$ and $E_{\mathcal{F}_t}$. The high risk graph has two components (top right corner with 3 nodes, and bottom left corner with 1 node), while the biodiversity habitat graph only has one.

Cells for which the value is 0 are not considered as nodes for both graphs, and are thus not connected to the rest of the graphs.

In the final landscape, because $\mathcal{F}_t \subset \mathcal{B}_t$, the landscape where orange cells are high fuel load and also support biodiversity habitat (e.g. $a_{ijt} \in V_{\mathcal{B}_t} \cap V_{\mathcal{F}_t}$)

bigger patch, and less important steppingstone cells or patches.

Our global connectivity indicator is similar to the notion of *energy of a graph* (Gutman, 2001), which can be understood as a measure of connectedness (highly connected graphs tend to have high energy) for graphs. However, we differ from Gutman (2001) by including self-loops as habitat cells and patches are connected to themselves. Our formulation of H reframes a quadratic form from the adjacency matrix of a graph grid structure (appendix A). The adjacency matrix displays interaction among nodes that are neither purely constructive or destructive, as some combinations of active neighboring nodes will add to global connectivity, while other combinations may subtract global connectivity. In all the landscape sizes we used, eigenvalues of the adjacency matrix were both positive and negative, leading to indefiniteness (see figure 2.A). Therefore, H is not globally convex nor concave.

2 Social planner decision : the high-risk /connectivity dilemma

2.1 Dynamic decision problem

A social planner tries to minimize the global connectivity index of the high risk graph, using fuel treatments (eq. 2.5). However, when implementing treatment, a cell's successional stage is reset to *juvenile*, thus destroying biodiversity habitat. In coherence with real world applications, the social planner is faced with a temporal budget constraint (e.g. the sum of treatments $\sum_{ij} x_{ijt}$ must be lower or equal to the *Budget* - eq. 2.9) as well as an ecological constraint, in terms of biodiversity habitat connectivity (e.g. the global connectivity of biodiversity habitat $H(\mathcal{B}_t)$ must be larger than constraint *Biod* - eq. 2.7). Both the ecological and budget constraint need to be satisfied in each period.

For the sake of the analysis, we focus on two layers of complexity : time and space. We do not include a stochastic component related to wildfire risk e.g. we adopt a deterministic framework where the value at risk (global connectivity of risky cells) is weighed against the loss in biodiversity habitat connectivity. Additionally, we consider a homogeneous distribution of treatment costs across the landscape e.g. cost of treatment in each cell through time is 1. We come back to this assumption in the discussion. Monetary benefits are also homogenously distributed across the landscape, and normalized to 1. Note that there is, however, heterogeneous returns to treating across the landscape : some cells will contribute more than other to global connectivity. Finally, given that the planning horizon is finite, we do not discount future high risk connectivity scores and assume each period is equally important in decision making.

The optimization problem is :

$$\min_{\{\{x_{ijt}\}_{(i,j)}\}_{t=1}^T} \left[\sum_{t=1}^T H(\mathcal{F}_t) \right] \quad (2.5)$$

Such that:

$$\mathbf{A}_0 \text{ given} \quad (2.6)$$

$\forall t \in \{1, \dots, T\}$:

$$H(\mathcal{B}_t) \geq Biod, \quad (2.7)$$

and $\forall (i, j) \in \{1, \dots, n\}$:

$$a_{ijt+1} = \min((a_{ijt+1}(1 - x_{ijt}); 2), \quad (2.8)$$

$$\sum_{i,j} x_{ijt} \leq Budget, \quad (2.9)$$

$$x_{ijt} \in \{0, 1\} \quad (2.10)$$

Notation	Concept
\mathbf{A}_t	Landscape matrix representing successional stage at time t
a_{ijt}	Cell (i, j) of landscape with value $\in \{0, 1, 2\}$
x_{ijt}	Treatment status $\in \{0, 1\}$ of cell (i, j) at time t
H	Global connectivity measure
$\mathcal{F}_t = (V_{\mathcal{F}_t}, E_{\mathcal{F}_t})$	Graph of high risk cells
$\mathcal{B}_t = (V_{\mathcal{B}_t}, E_{\mathcal{B}_t})$	Graph of suitable habitat cells
$Biod \in \{0, \dots, \max H(\mathcal{B})\}$	Level of habitat global connectivity constraint
$Budget \in \{1, 2, 3, 4\}$	Level of the budget constraint
$n \in \{4, 100\}$	Size of the landscape
$c = 3$	Number of age classes
$T \in \{5, 10\}$	Planning horizon

Table 2.1: Summary of model variables and functions

As common in the literature, we can express the budget as a share of land being treated ranging from 5 to 25% of the surface area (when $n = 4$). These values encompass historical and projected policies in Australia (Burrows and McCaw, 2013), the United States (Office, 2019) and Southern Europe (Fernandes et al., 2013).

Additionally, we solve the problem for a range of possible habitat connectivity values, ranging from 0 to the maximum possible habitat connectivity for each landscape size n .

2.2 Non-convexity and dimensionality curse

Our problem can be classified as a *critical node detection problem*, i.e, a problem of locating the vertices that best degrade connectivity, such that the number of components increase, and within remaining components, nodes with the largest *centrality* are targeted (Arulselvan et al., 2009). As definitions of *graph centrality* matter, we refine our approach in section 4.2. Problems of the critical node class are computationally difficult (e.g. NP - Hard) in a single graph (Arulselvan et al., 2009; Matsypura et al., 2018). Efficient heuristics to find near-optimal solutions exist and leverage perturbations around local solutions (Arulselvan et al., 2009; Zhou and Hao, 2017). Compared to the canonical *critical node detection* problem, our problem features a non-convex objective function, a budget constraint, and a constraint on habitat connectivity, which imposes a constraint on the supergraph of high risk cells. Given our constraints, the behavior of the global connectivity measure H , standard optimization techniques cannot be applied, and heuristics are required.

In dynamic problems, a standard technique is dynamic programming (Bellman, 1957). Dynamic programming provides a temporal decomposition of the initial problem defined over T periods, into T recursive problems, as it relies on the 'optimality principle'¹⁰. A value function V , mapping each possible state of the world e.g. \mathbf{A}_t to the optimal value of the objective function along the planning horizon, is iterated upon to find the optimal policies $x_{ijt}^*(\mathbf{A}_t)$, i.e, the sequence of optimal treatments, and the optimal states $\mathbf{A}_t^*(\mathbf{A}_0)$ resulting from the optimal policies and the initial conditions. However, it is impractical in our case. Our problem suffers a *dimensionality curse* (Bellman, 1957). There are 3^{n^2} values for the state variables¹¹ in each period and the specific nature of our objective function H (e.g. no global convexity) and the discrete nature of the state space make interpolation of a value function impossible¹².

¹⁰"An optimal policy has the property that whatever the initial state and initial decision are, the remaining decisions must constitute an optimal policy with regard to the state resulting from the first decision". (See Bellman (1957), Chap. III.3., p.83)"

¹¹Given that the landscape \mathbf{A} is of size $n \times n$, and that each element of a_{ij} can take $c = 3$ values, there are 3^{n^2} landscape configurations possible

¹²As a matter of fact, with a large number of state variables e.g. a high-dimensional state space, methods such as adaptive sparse grids can be used towith smooth, continuous objective functions (Brumm and Scheidegger, 2017) to circumvent the dimensionality curse. The fact that the input space is an n^2 -dimensional binary Cartesian product and that H is not globally convex hinder the use of such tools.

3 Solution method and computational experiments

Three key features of our problem hint that a dynamic (e.g. that optimizes the objective function over the whole planning horizon) and a repeated myopic solution (e.g. which optimizes the objective function in each period) should be similar. The dynamics occur before the decision is made, therefore the decision maker has full knowledge about the state of the system. The dynamics are simplified and have relatively little depth, as we limit ourselves to 3 age classes. Finally, our intertemporal objective function is additively separable.

Our solution methods resort to two key ingredients : optimization heuristics, and comparison between the dynamic and repeated myopic problem.

First, we circumvent the non-convexity of the global connectivity metric and the high dimension of the state space by using a genetic algorithm ([Holland, 1992](#)) (implemented in R with package `GA` ([Scrucca, 2017](#))) with population size of 200 and 250 iterations. Genetic algorithms are especially suited for high dimensional, combinatorial search spaces¹³ and fare better than a brute force approach, or other heuristics (Particle Swarm Optimization or Simulated Annealing).

Then, we compare the performance of a 5-period objective function to a 5 period repetition of a static objective function. We trade the completeness of dynamic programming for a more manageable approach, where we compare these approaches for 884 randomly drawn landscapes of size $n = 4$. We sample the landscapes according to the distribution of possible landscapes (see figure 2.C). As landscapes with large numbers of *juvenile* and *adolescent* cells are overrepresented, we impose that underrepresented possible landscapes are included at least 2 times in our sample, to disentangle composition (e.g. number of cells of each successional stage) from configuration (location of cells) effects.

We focus $T = 5$ planning horizon for several reasons. First, as the dynamic of our ecological processes comprises 3 stages, using a 5-period horizon allows for each cell to grow from its original stage to *mature*, be treated, and revert to its original stage, e.g. allows for a full successional cycle to be performed. Second, a 5-period horizon corresponds to a long policy horizon, ranging from 25 years to 200 years ([McColl-Gausden and Penman, 2019](#); [Thomas, 1979](#)). Third, for our approach to be useful for policy making given that we abstract from stochastic modifications to the environment (e.g. occurrence of wildfire, spread of invasive species increasing flammability at a given age etc), policies need to be forward looking with enough temporal depth to be relevant and be reevaluated with potentially new initial conditions resulting from environmental perturbations.

Next, with repeated static optimization we increase the size of our sample and temporal depth, to encompass all the possible landscape configurations for land-

¹³Here, the control variable is a $Tn^2 = 5n^2$ binary variable

scape size $n = 4$, over the whole range of possible values for the biodiversity habitat constraint, over $T = 10$ years. Of all the 3^{n^2} initial landscapes combinations possible, we only keep landscapes that are unique up to a permutation¹⁴. This results in a sharp reduction of landscapes to consider from 43,046,721 initial to 5,398,082 unique initial landscapes for $n = 4$. We focus on exact optimal solutions for all the initial conditions of these small-scale landscapes and implement our own solution algorithm in Python 3.9.13 and R 4.3.3¹⁵. We find generally applicable principles for treatment allocation.

Third, we increase the landscape size to $n = 100$, for a sample of 20 large scale (10,000 cells) landscapes with varying compositions and autocorrelation using two-dimensional fractional Brownian motions (table 2.B summarizes their characteristics and figure 2.B illustrates 6 of them). We use neutral landscape models ([Caswell, 1976](#); [Gardner and Urban, 2007](#)) and implement them in R ([Sciaini et al., 2018](#)). Neutral landscape models were designed in theoretical landscape ecology to develop spatial ecology indicators and “evaluate the effects of landscape structure on ecological processes” ([With and King, 1997](#)). Even though they are designed as null models to compare with real landscapes, after ecological processes have shaped them, they provide a useful basis for scaling our analysis. We solve the repeated myopic optimization problem on these 20 landscapes over $T = 10$ periods based on our generally applicable principles, and compare them with other policy scenarios. We compare our principles with (i) a repeated random policy, targeting cells that are either *adolescent* or *mature*, (ii) a gridded treatment policy (as depicted in figure 2.F) with evenly spaced segments of treatment, (iii) a policy that always targets the most central nodes in terms of betweenness centrality, and (iv) a policy which targets the largest degree¹⁶ nodes, in terms of global risk and habitat connectivity measures.

4 Landscape and treatment indicators

4.1 Landscape level indicators

To characterize the managed landscapes, we mobilize several indicators from landscape ecology and graph theory (see appendix C).

First, we account for the high risk and habitat surfaces in the landscape by measuring the number of vertices in each graph. Second, to assess landscape connectivity and fragmentation as well as landscape diversity¹⁷, we use our global

¹⁴That is to say, landscape \mathbf{A}_0 is included in the set of initial conditions \mathcal{I} if and only if for any element \mathbf{A}'_0 in \mathcal{I} , \mathbf{A}_0 is not a permutation (eg can be obtained through rotations or symmetries) of \mathbf{A}'_0

¹⁵Data and code are [publicly available](#)

¹⁶In a graph $\mathcal{G} = (V, E)$, the degree of a node v is the number of edges connected to the node

¹⁷In the context of fire prone ecosystems, the notion of “fire mosaics” ([Bradstock et al., 2005](#))

connectivity metric H (eq. 2.4), as well as the *number of components*¹⁸. To specifically assess landscape diversity, we use the Simpson index (Simpson, 1949) on successional stages stages (eq. A.4)¹⁹. However, the Simpson index does not account for the diversity of spatial patterns: a checkered landscape with two seral stages would be as diverse as a landscape with two large patches for each seral stage, according to the Simpson index. Therefore, we use the landscape shape index (LSI - eq. A.5), a normalized ratio between the perimeter of biodiversity habitat and its area (Patton, 1975; McGarigal and Marks, 1995). To disentangle the correlated effects of perimeter and area that affect the landscape shape index, we use a successional stage heterogeneity index, that averages the probability that, for each cell, neighbors in the 4 cardinal directions share the same successional stage (eq. C). The index ranges between 0, when the successional stage is the same across the whole landscape, to 1, in a checkered landscape. The index assesses whether the landscape is a mosaic (Bradstock et al., 2005), and if it displays structural diversity, conducive to diverse communities and functional diversity.

4.2 Treatment level indicators

To analyze the treatment allocation mechanism, we use the *number of treatments* as well as their *geographic location*. Additionally, we use graph theory measures to assess the location of treatments in relation with the graph structure. To do so, we use different measures of *centrality*, e.g. measures that answer how important a vertex is for the graph structure, and overall, its connectivity. Measures of centrality produce a ranking of vertices, but are not necessarily comparable. Additionally, depending on the measure chosen, different vertices can be the most central. To overcome these limitations, we use 4 measures of vertex centrality.

First, we use *degree centrality*, which measures the degree of vertices. This measure is computationally simple and captures direct centrality effects e.g. how a single vertex interacts with its neighbors, without considering 2nd degree neighbors, or further relationships. Second, we use *betweenness centrality* (see appendix D for a formal definition), measuring the extent to which a vertex is on the shortest path between other vertices. This measure is more computationally intensive and useful to understand through which vertex flows may go through, e.g.

conveys the idea that fire causes variations in successional stages through space thus providing different types of habitat for biodiversity and improving biodiversity

¹⁸A component C_t of graph \mathcal{G}_t is a maximally connected subgraph of \mathcal{G}_t that is not part of any larger connected subgraph. A component is *connected* (for all two vertices $(u, v) \in V_C$, there exists a path in C_t that connects them) and C_t being a subgraph of \mathcal{G}_t , it is *maximal* if there is no other connected subgraph C' of \mathcal{G}_t such that C_t is a proper subgraph of C'_t . Figure 2.2 illustrates this concepts in both the habitat and high risk graphs resp. \mathcal{B}_t and \mathcal{F}_t

¹⁹Similar results can be found with the Shannon index (Shannon, 1948). To avoid issues related to degenerate values and logarithms, we focus on the Simpson index.

wildlife dispersal or wildfire spread. Third, we use *eigencentrality*, which measures the influence of nodes through their connections : a vertex is central if it is largely connected to nodes themselves well-connected. While computationally intensive, it furthers the results from degree centrality. Finally, we use *subgraph centrality* (Estrada and Rodríguez-Velázquez, 2005), which measures the participation of vertices to subgraphs of the graph, and captures the role of a node in local structures, especially suited for networks with well identified, low connected subgraphs. We implement these measures in R using the package `igraph` (Csárdi and Nepusz, 2006)

3 Results

1 Dynamic v. myopic repeated optimizations

As expected, the results of the dynamic and static optimization procedures over 5 periods yield very similar aggregate results in terms of intertemporal global risk connectivity (hereafter *risk*), for different budget (measured in treatment units) and global habitat connectivity constraints (hereafter *habitat constraints*, measured as proportion of the maximal global habitat connectivity attainable), as shown by figure 2.3.

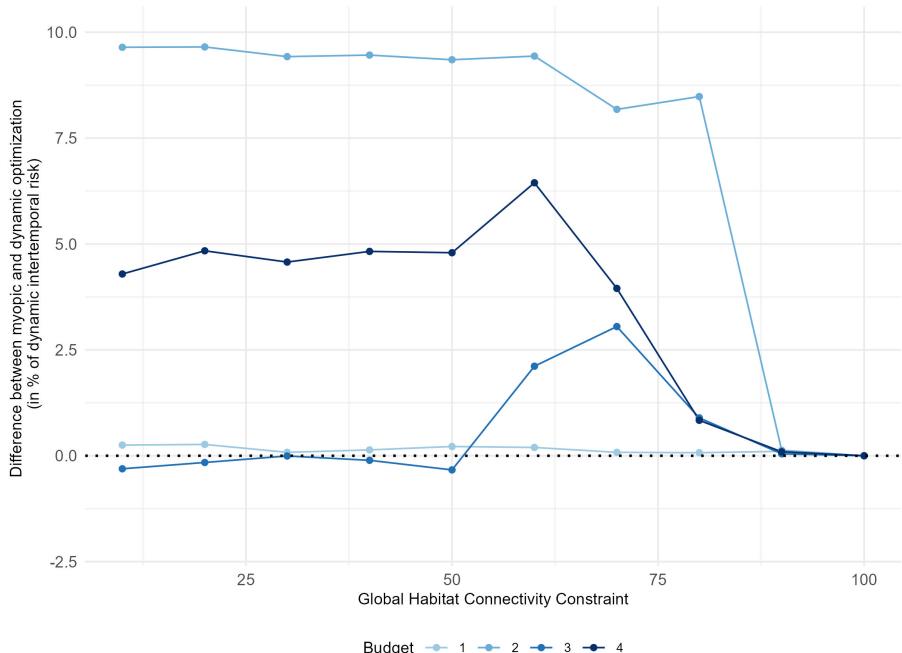


Figure 2.3: Comparison of aggregate intertemporal global risk connectivity for dynamic and repeated myopic procedures

Table 2.C shows the result of a regression analysis of the difference in global risk connectivity between the dynamic optimization and the repeated myopic

procedure (e.g. if the difference is positive, repeated optimization results in a lower intertemporal global risk connectivity), based on initial landscape characteristics, without interaction terms. First, the average risk difference is positive across all habitat constraint levels (given the magnitude of the intercept and constraint coefficients). With larger budgets, the relative performance of repeated myopic optimization increases, while it merely decreases with increases in the habitat connectivity constraint level: although statistically significant, the magnitude is negligible. Finally, the Successional Stage Heterogeneity Index is statistically significant but does not lead to significant effects due to its magnitude. Other models, including interaction terms are presented in appendix F. They all point towards the absence of clear mechanism determining the performance differentials between myopic and dynamic optimization procedures.

2 Wildfire risk reduction and habitat connectivity : a production possibility frontier approach

Figure 2.4 shows the global risk connectivity measure, with varying levels of global habitat connectivity and budget constraint, e.g. a production possibility frontier between risk and habitat connectivity, for the 10 years planning horizon²⁰

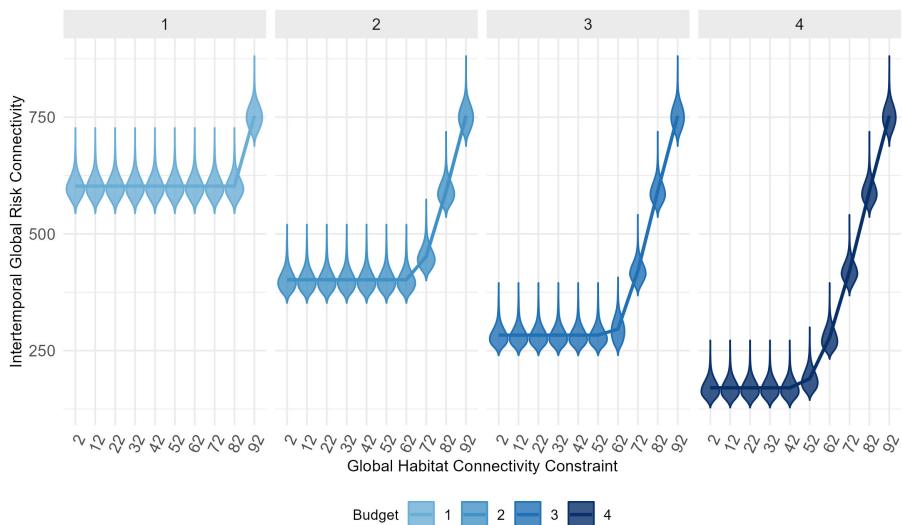


Figure 2.4: Production possibility frontier across global habitat connectivity and budget constraints

For each value of the global habitat connectivity constraint is plotted the distribution of the values in each violin plot. The line refers to the group specific average level of global risk connectivity

Using repeated myopic spatial optimization, reducing global risk connectivity while maintaining global habitat connectivity comes as a trade-off, albeit moderate: indeed, increasing habitat requirements increases the remaining risk, but

²⁰Figure 2.D displays production possibility frontiers for repeated myopic and dynamci procedures and shows the same properties

there are combinations that can satisfy large habitat connectivity and risk reductions.

Budget is a key factor in risk reduction, as it relaxes the trade-off between the two objectives: increasing the budget reduces the wildfire risk while maintaining a range of biodiversity constraints. When habitat constraints are large, however, the marginal effect of budget is limited, production possibility frontiers tend to be identical and a larger remaining risk needs to be accepted. Indeed, when the $Budget = 2$, average risk is maintained at 401 for habitat constraint levels ranging from 2 to 62, while when $Budget = 4$, risk is down to 170 (-58%) for habitat constraint levels ranging from 2 to 42. However, when the habitat constraint is at 72, average risk for $Budget = 2$ is at 451, while it is at 420 for $Budget = 4$ (7% difference). Hence, similar risk profiles can be attained at a lower budget for high habitat constraints. Conversely, as the costs of treatment increase, for a stable budget, the remaining risk increases sharply, and factoring in habitat requirements in the decision-making is not necessary for targets below 82. For example, if costs were to double at $Budget = 2$, the average risk between 2 and 62 would increase by 50% (e.g. at $Budget = 2$, average risk is 401, while it is 602 for $Budget = 1$), and the habitat constraint only becomes stringent at 82.

3 Convergence towards steady states

Our simulations for $T = 10$ and scanning all the possible configurations for landscapes of size $n = 4$ show that 100% of the initial landscapes converge in finite time towards a steady state solution, that minimizes wildfire risk while satisfying budgetary and habitat connectivity requirements (figure 2.5a). Steady states are landscape cycles with finite periods. Landscapes converge to steady state distributions given the bounded nature of the successional dynamics. Analyzing the steady-state cycles (and the unique landscapes that form them) drastically reduces the set of landscapes to analyze: they represent 0.001% of the initial landscapes. Results show that landscapes converge to cycles with equivalent configurations when the cycle period = 1, or have a transitory phase during 1 period, before reverting to an equivalent configuration.

Figure 2.5b shows that conditional on data availability on every patch, the more the decision maker wants to conserve biodiversity, the fewer steady-state landscapes she has to consider. An increase in the habitat requirement reduces the room for maneuver. Indeed, budget acts as a complexifying factor: the larger the budget (relative to costs), the larger the set of steady-states to consider. Aiming for relatively large habitat connectivity reduces the set of viable strategies to be considered and can more efficiently guide policy.

4 Properties of steady state landscapes: surface, fragmentation, and diversity

Figure 2.6 displays, for each global habitat connectivity and budget constraint levels, the most frequent steady-state cycle. Figure 2.7 shows the indicators relative to the surface and components of the high-risk graph and figure 2.8 shows the indicators related to diversity, both for landscapes of size $n = 3$ and 4, averaged over all the steady-state landscape cycles.

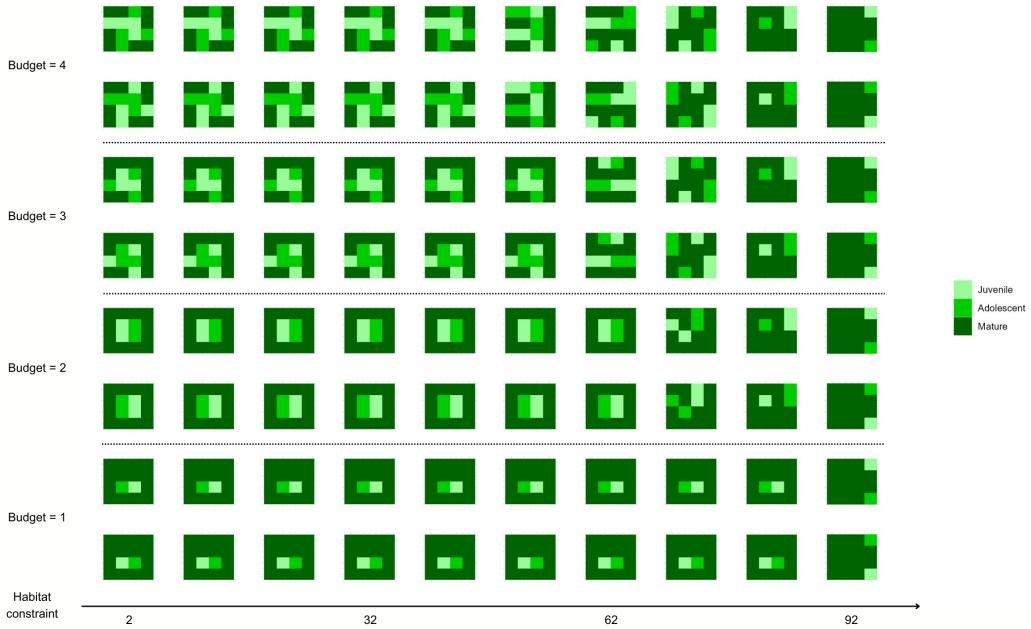
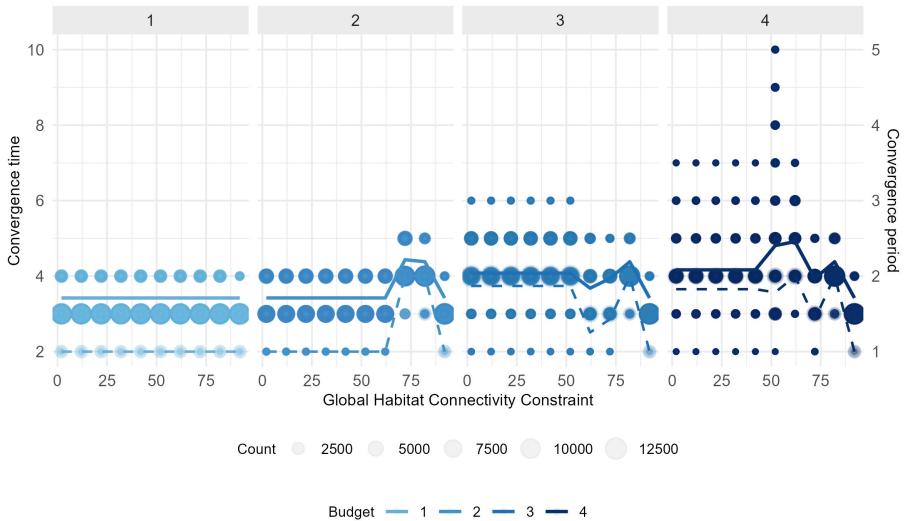


Figure 2.6: Most represented cycles for each global habitat and budget constraint levels

Previous results show that budget further reduces risk, conditional on habitat connectivity constraint being low. Focusing on constraint levels below 50, risk reduction is primarily driven by reduced area (figure 2.7a), and increased fragmentation of the landscape, in the form of increased components number and reduced sized of the components mean size (figures 2.7b and 2.7c). As more connected habitat needs to be protected, the average size of components increases for all $Budget > 1$. For large budgets (e.g. $Budget \in \{3, 4\}$), the average component number starts to fall first (for example, at habitat connectivity constraint level 40 for $Budget = 4$), and then the average risk area increases (for example, at habitat connectivity constraint level 52 for $Budget = 4$). The average component size increases as the number of components decreases for habitat connectivity constraints above 42 : small components either disappear or increase in size, risky cells are reallocated to connect separated components before the high-risk surface increases. This is exemplified by the landscape cycles displayed in figure 2.6 (especialy for panels of $Budget \in \{3, 4\}$).

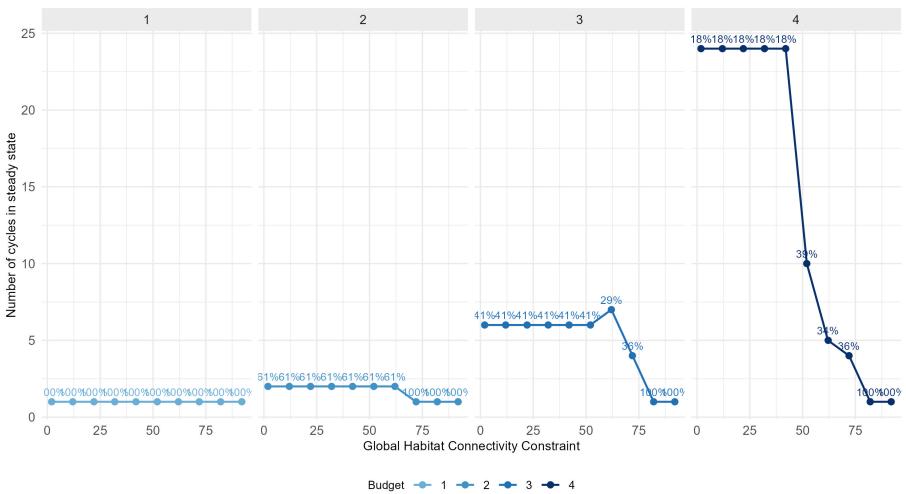
Landscape diversity unambiguously increases with the budget at low habitat connectivity constraint levels (figures 2.8a and 2.8b). As more units are treated, the evenness of successional stages increases in the landscapes, which drives increases in the Simpson Index (fig. 2.8a). At low habitat connectivity constraints, global risk connectivity is diminished through fragmentation of the risky patches. The larger the budget, the more treatment, and the more fragmentation, which increases the structural diversity of the landscapes as cells are less likely to be at the same successional stage in all directions, driving the evolution of the Successional Stage Heterogeneity Index (fig. 2.8b). At low habitat connectivity constraint levels and large budgets, even though the relative area of habitat decreases, the shape of habitat is more irregular (fig. 2.8c). In this context, *adolescent* cells act as stepping stones and corridors between *mature* habitat patches.

When habitat connectivity constraints increase, diversity collapses both quantitatively and qualitatively (fig. 2.8). The Simpson index collapses as land successional stages gradually homogenize (fig. 2.8a) across all budgets. Moreover, landscapes form less of a mosaic, and are more clumpy, as displayed by the LSI and Successional Stage Heterogeneity Index (figs. 2.8c and 2.8b). Overall, for large habitat targets, landscapes tend to homogenize and to be better connected, although less quantitatively and qualitatively diverse.



(a) Convergence times and period across global habitat connectivity and budget constraints

Average convergence time is displayed with full lines and measured on the left *y*-axis, while average convergence period is displayed with dashed lines and measured on the left *y*-axis



(b) Number of cycles in steady state as the global habitat connectivity constraint evolves and across budget constraints

Above each data point is the frequency of the most represented cycle in the data.

Figure 2.5: Steady state cycles: convergence and distribution

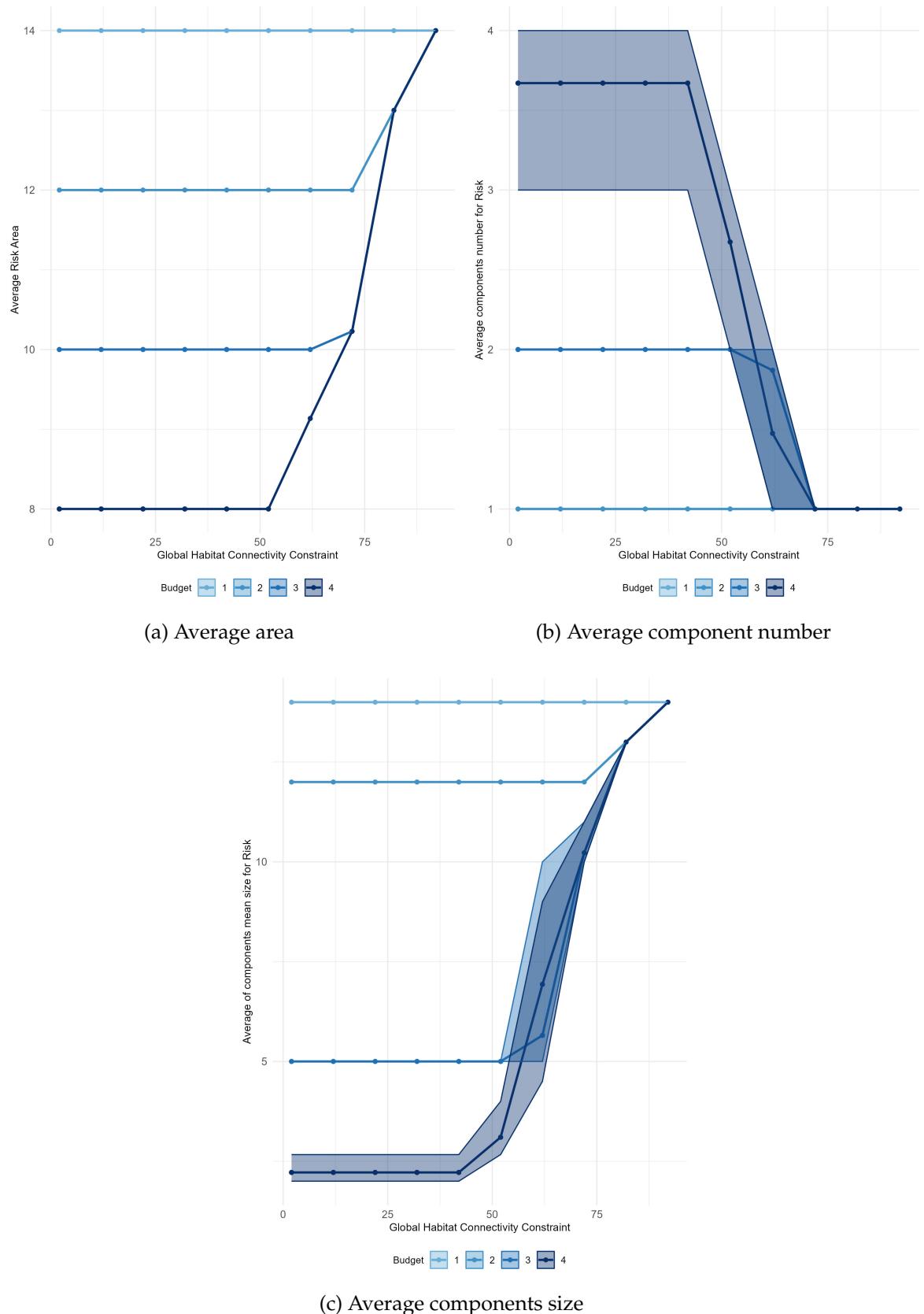


Figure 2.7: Indicators relative to surface and components across habitat and budget constraints

The indicators are averaged across the cycles represented for each habitat and budget constraint levels.

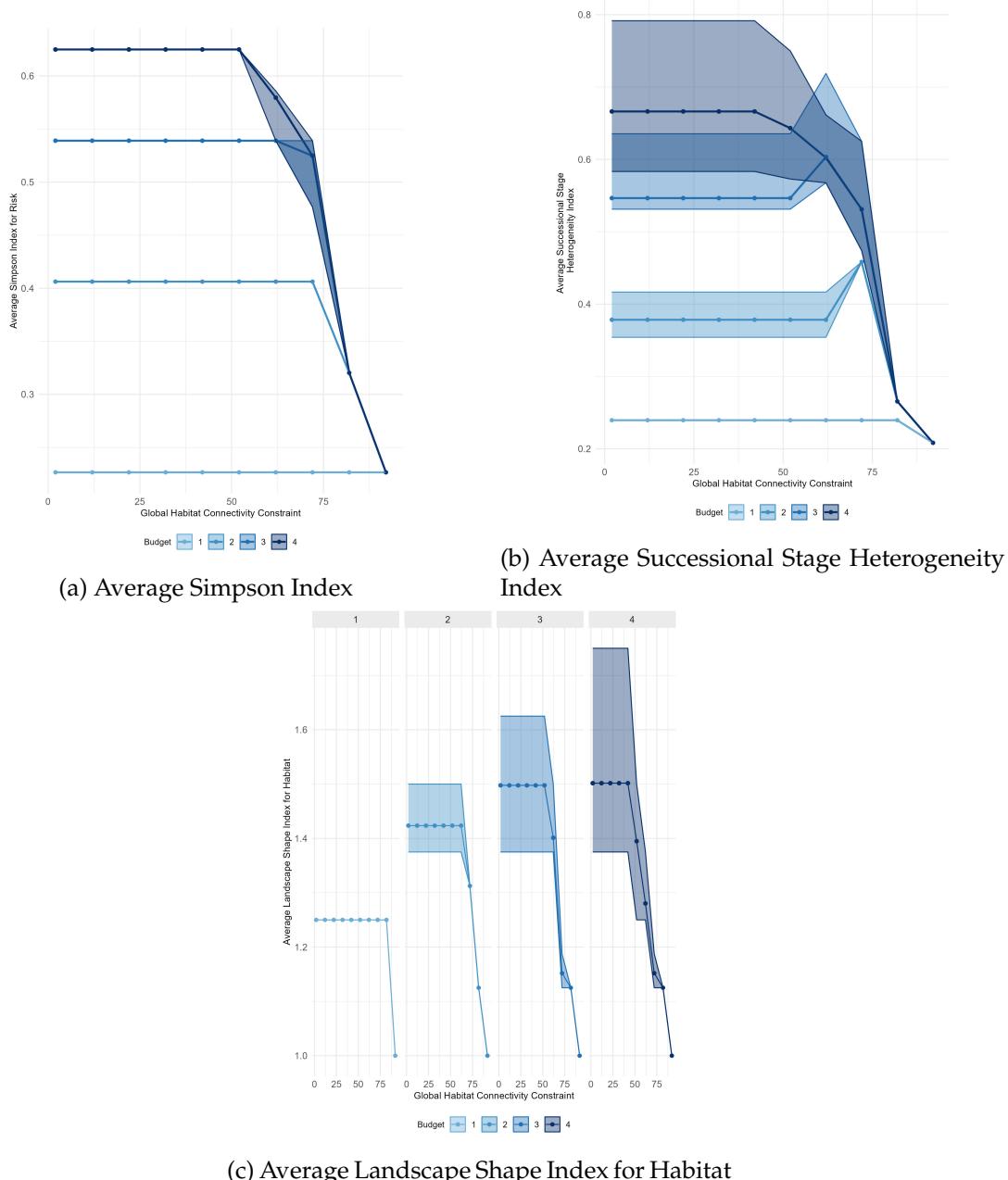


Figure 2.8: Indicators relative to landscape diversity across habitat and budget constraints

The indicators are averaged across the cycles represented for each habitat and budget constraint levels.

5 Spatial allocation of optimal management at the steady-state landscape cycle

Treatments are concentrated on *adolescent* cells across all budget and habitat connectivity constraints (figure 2.9, except at 62, for some steady-state cycles for $Budget \in \{3, 4\}$). This is coherent with the real life practice of fuel management operations, primarily targeting *adolescent* land patches. At level 62, the number of treatment varies between phases of the steady-state cycle for $Budget = 4$, reflecting the increase in habitat connectivity constraint.

As show in figure 2.10 for low habitat connectivity constraints, the budget constraint is saturated, and all of the budget is used. Coherent with the evolution of risk area highlighted in section 4, the number of treatments decreases after the steady state landscape start experiencing an increase in mean component size, and gradually reduce as the habitat connectivity constraint increases (e.g. starting at 62 for $Budget = 2$, or 52 for $Budget = 4$): for large habitat connectivity constraints, the budget constraint is no longer satiated.

Figures 2.11a, 2.Ea, 2.Eb and 2.Ec show that the different centrality measures (e.g. betweenness, subgraph, degree and eigencentrality, respectively), are very correlated, and display identical overall patterns, in terms of relative values of metrics compared to the maximum possible, and in terms of rankings. We therefore focus on betweenness centrality to characterize our results.

Figure 2.11a shows the average betweenness centrality (and the corresponding ranking) of treated cells in risk graphs \mathcal{F}_t . For low levels of biodiversity constraint, cells with the largest betweenness centrality are treated first, as testified by the panel of $Budget = 1$ in figure 2.11a, and illustrated by the average share of treatments per cell in figure 2.11b. Across budgets, the first treatment unit targets the most central cells when available for treatment. In the context of critical node detection, when the ecological requirements are low, the high-risk graph \mathcal{F}_t is primarily considered, and nodes with the most cost-efficient risk reduction, i.e, with the largest betweenness centrality are targeted. Once the most connected cells are treated and the budget constraint relaxes, lower-centrality cells get treated, in a sequential fashion. : for $Budget = 4$ and constraint below 42, the first treatments across cycles are the 1st and 2nd most central locations, but the 4th treatments target the least central cells with non zero centrality. This is epitomized by the top row of figure 2.11b.

When the habitat connectivity constraint increases, several effects come at play. Not only does the number of treatments decrease, but the spatial allocation also changes. When $Budget = 3$, figure 2.10 shows that treatment number remains constant between habitat connectivity constraints 52 and 62, but the spatial distribution of treatments drastically changes (fig 2.11b), as treated cells are

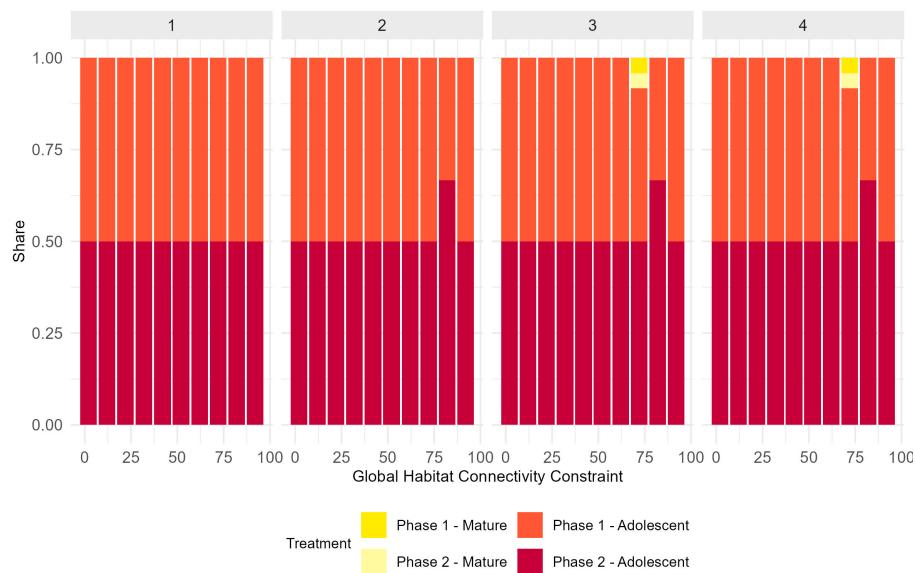


Figure 2.9: Distribution of the successional stages of treated cells in steady state cycles across budget and habitat connectivity constraints

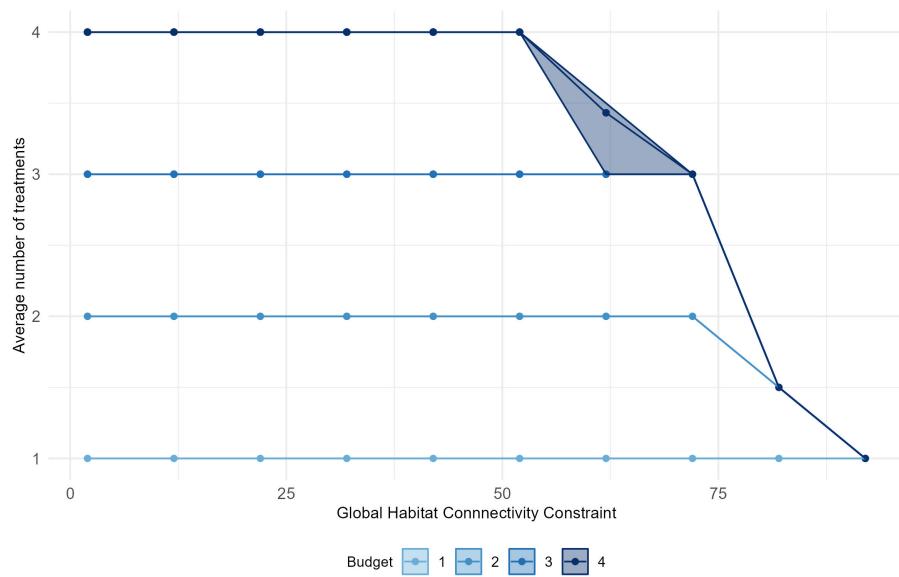
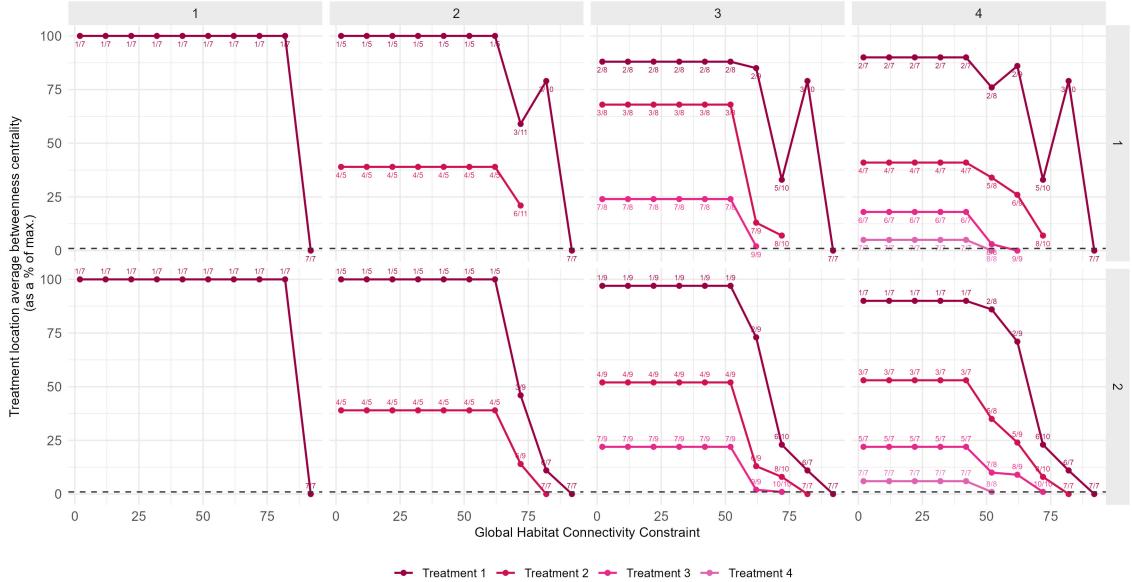
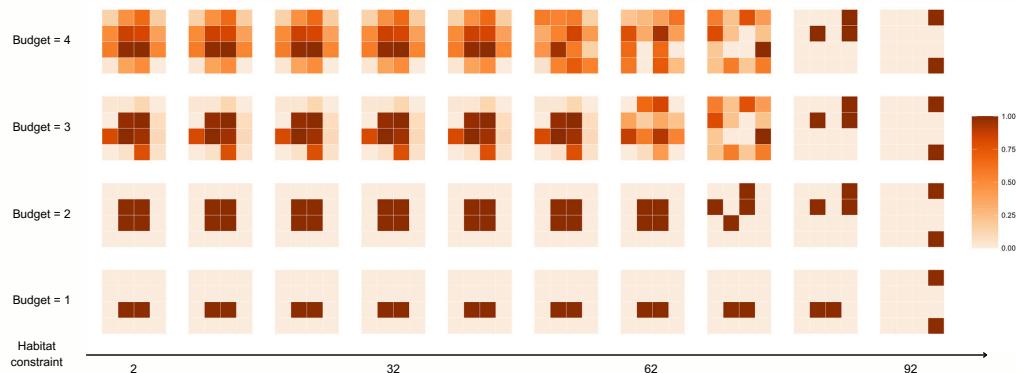


Figure 2.10: Average number of treatments in steady state cycles across budget and habitat constraints.



(a) Average treatment location betweenness centrality across budget and habitat connectivity constraints, and steady-state cycle phases

On the horizontal grid are displayed budget levels, while cycle phases are displayed on the vertical grid. For each steady-state cycle, treatment location is characterized by a betweenness centrality score, as a proportion of the maximal betweenness. Averages across steady-state cycles are displayed. Numbers in 5/7 format refer to the average betweenness centrality ranking of the treatment



(b) Distribution of treatment locations in steady-state cycles across budget and habitat connectivity constraint levels

Each cell is colored as the average of her treatment indicator x_{ijt} over all steady-state : the darker the cell, the higher the frequency of treatment

Figure 2.11: Treatment allocation : centrality

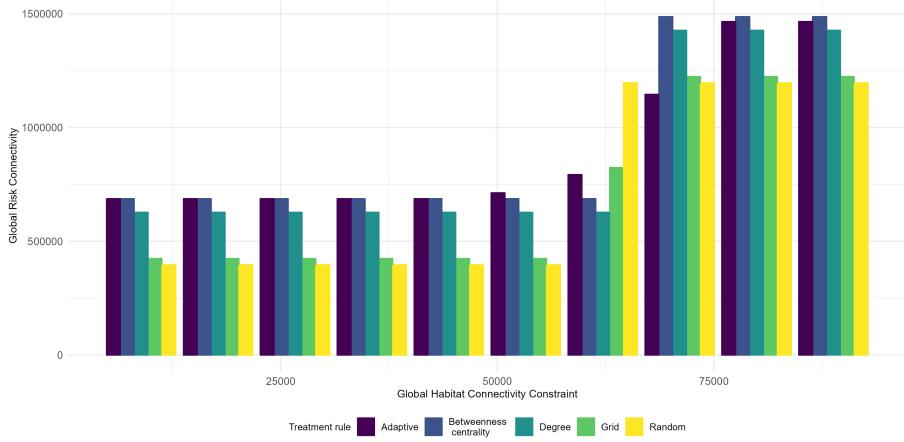
less central, and edge cells are more treated (fig 2.11a) as the relative weight of the habitat graph \mathcal{B}_t increases, treating the most cost-efficient risk-reducing nodes also degrades habitat connectivity.

Therefore, as habitat targets increase, the number of treated cells remains stable but the betweenness centrality of cells decreases : for $Budget = 2$, aggregate betweenness centrality starts declining at 62, for $Budget = 4$ at 42 (figs 2.11 and 2.11b). Then, the number of treated cells decreases : for $Budget = 2$ for example, treated cells start declining at 82, and for $Budget = 4$ at 52 (fig 2.10). Once the number of treated cells has decreased, there is a spike in betweenness centrality of the remaining treated cells : while less area is treated, a more central location is treated (see the panel of phase 1 and $Budget \in \{2, 3, 4\}$ at levels 82). Then, the number of treated cells continues to decrease as the habitat connectivity constraint increases, and less central cells are targeted.

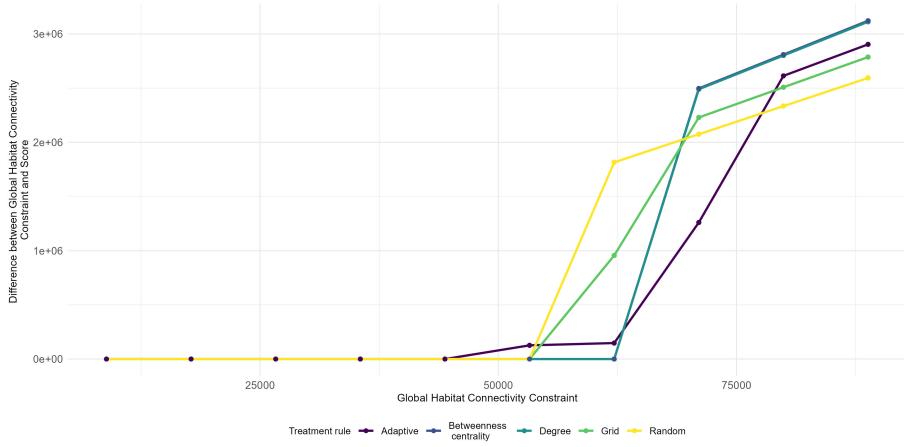
Considering results in terms of pattern convergence and treatments characteristics, general treatment guidelines can be derived and applied to large scale landscape, with a budget comprised between 0 and 25% of the landscape size. As measures of connectivity can be correlated at the small scale, but differ at the large scale, we formulate 4 different strategies, where centrality is defined according to the 4 measures used :

1. For low habitat requirement (e.g. below 50% of maximal global habitat connectivity), the budget constraint must be saturated and target cells with the largest betweenness centrality. For the first 5 periods, target cells can be both *adolescent* and *mature*
2. For intermediary habitat requirements (between 50% and 75% of maximal global habitat connectivity), and low budget, the budget constraint must still be saturated, but for larger budgets, decreased by 25%. 25% of the budget can be spent on high betweenness centrality cells, but 75% of the budget must be spent on average or low betweenness centrality cells
3. For large habitat requirements (between 75% and 100%), the budget constraint must not be saturated: the percentage of treated cells must decrease to an interval between 5 and 15%, and low betweenness centrality must be targeted if any cell is to be treated.

Across all these options, to account for the transitional dynamic, target cells can be both *adolescent* and *mature* in the first 5 periods, and exclusively *adolescent* for the last 5. We use these principles to structure treatment rules for large scale landscapes, and compare them to optimal management as well as other policy rules.



(a) Comparison of policy intertemporal risk and penalty for large scale landscapes



(b) Comparison of global habitat connectivity constraint attainments for large scale landscapes

Figure 2.12: Comparison of policy performances for large scale landscapes

6 Evaluating policy rules

Figure 2.12a and 2.12b shows the performance of the diverse policy options across 20 large scale landscapes on planning horizon of 10 periods. The performance is assessed as the sum of the global risk connectivity across 10 periods and a penalty proportional to the difference between the global habitat connectivity constraint and the global habitat connectivity score across periods.

Results show that the grid and random policy rules fare better than the policies based on graph theory metrics for constraint values below 61,262 (e.g. 70% of maximal global habitat connectivity). At 61,262, they fare worse than the metrics based on graph theoretical measures, but the adaptive strategy extrapolated from small scale results fares worst among graph theory policies. At 71,403 (e.g. 80% of maximal global habitat connectivity), the adaptive strategy fares better than all measures, as only it is the only one that satisfies the habitat connectivity constraint. After 71,403, the adaptive strategy is ranked 3rd after the random and grid strategies, as the difference in global habitat connectivity constraint attain-

ment rises faster than with the random and grid policies.

4 Discussion

1 Confirmation and generalization of existing results

Our analysis of the exhaustive set of initial conditions for small-scale landscapes confirms existing results in the literature. We argue that they bring robust evidence and complement the existing literature to derive general conclusions.

Our model encompasses 3 seral stages and 1 composite vegetation type and proves the convergence of every initial condition to a steady state cycle, irrespective of the initial configuration. We extend [Minas et al. \(2014\)](#) that find convergence patterns for *homogeneous* landscapes only, i.e, landscapes where the initial vegetation age is uniformly distributed. We show that in the event of environmental perturbations that do not disrupt ecosystem dynamics, an appropriate policy can recover the previous equilibrium risk and habitat.

Our production possibility frontier (PPF) between wildfire risk and habitat connectivity is consistent with PFF literature ([Arthaud and Rose, 1996](#); [Calkin et al., 2005](#)). Our results also confirm that trading one objective for the other is not as efficient as increasing the policy budget to reconcile objectives. We show that increasing the policy budget nonetheless has diminishing returns for risk reduction, as highlighted by [Wei et al. \(2008\)](#); [Yemshanov et al. \(2021\)](#) and [Pais et al. \(2021b\)](#).

Our study yields clear results in terms of landscape ecology, leveraging concepts from landscape ecology, and highlighting the spatial mechanisms underlying the shape of PPF. We show that, for small scale, treatment allocation targets the most (between) central nodes first and then focuses on less central nodes (e.g cells closer to the border of the landscape) when habitat goals are low. In doing so, we do find general treatment allocation principles where previous studies on larger landscapes could not ([Minas et al., 2014](#); [Rachmawati et al., 2016](#)), generalize smaller scale ([Konoshima et al., 2008](#)) and case study specific ([Yemshanov et al., 2021](#); [Pais et al., 2021a](#)) results.

Compared to existing studies, our bounded depiction of vegetation dynamics allows, as well as the timing of decision making makes repeated myopic and dynamic approaches almost identical. As we abandon the refinement of dynamics, we are able to analyze the whole range of initial conditions, an endeavor that is seldom possible in dynamic spatial modeling. Using a graph theoretic framework on small-scale landscapes, we show that cell-level metrics help formalize and understand the drivers of treatment allocation and rationalize existing results.

Furthermore, we show that while prioritization approaches based on a graph

theoretic framing fare very well in an unrestricted set-up, including biodiversity habitat targets augments the problem's complexity. As a matter of fact, critical node detection can be efficiently achieved ([Arulselvan et al., 2009](#)), in the presence of budget constraints. However, solving critical node detection on the risk graph \mathcal{F}_t with constraints on the habitat supergraph \mathcal{B}_t remains a challenge. We generalize case studies ([Yemshanov et al., 2022](#)) and show less central risk nodes need to be targeted to achieve risk reduction and safeguard biodiversity habitat.

2 Challenges in generalizing small-scale landscape rules to large-scale systems

As highlighted in subsection 6, the policy based on the results from the small scale landscapes analysis do not improve risk performances compared to random policies. Several reasons explain this phenomenon and serve as guides for future research to finalize this project.

First, on small scale landscapes e.g small scale graphs, different definitions of graph connectivity can be equivalent, and mechanisms to decrease different measures of connectivity tend to have the same results. However, at a large scale, this is not the case, and different connectivity measures are no longer identical. When $n = 4$, treating the cells with large between-centrality tends to increase the number of components of the graph. When using the proposed adaptive strategy (see fig. 2.G) and treating between-central nodes, this results in a large, donut-shaped component : the mean shortest path length between paths is increased, but not infinite, there is always a path. Hence, while betweenness centrality efficiently guides treatments to increase the number of components and reduce betweenness centrality among components on small scale landscapes, it fails at the large scale.

The performance of the gridded treatment policy (figs. 2.12a and 2.F) show that increasing the number of even-sized components is a fruitful policy option. However, the grid does not take into account initial conditions, and therefore does not *optimally* fragment the landscape. Hence, a fruitful avenue for future rules lies in having a twofold hybrid approach. First, part of the treatment budget must be dedicated to fragmenting the landscape among large scale patches. Second, the remaining part of the budget must be used to decrease within component betweenness centrality, such that global connectivity is best degraded. The method for optimal components breaking can be based on the available, budget, the distribution of the patches successional stages, as well as the distribution of the betweenness centralities of edge patches, to optimally size large components.

Second, our policy rule stems from a steady-state analysis of small scale landscapes. We have shown that small scale landscapes converge in finite time to-

wards steady state landscapes. However, the convergence time increased with the available budget, and the convergence patterns period variance increased as well. Hence, as size and budget increase, the importance of the transition towards the steady state increases for dynamic treatment allocation and overall risk reduction. Therefore, one fruitful avenue to guide research on large landscapes can be found in analyzing the transitional dynamics of the small scale landscapes, bearing in mind the difficult scalability of specific connectivity degrading mechanisms.

3 Caveats and methodological perspectives

First, we resort to optimization heuristics in the dynamic and repeated cases. As the dynamic problem is significantly more complex, limiting the number of iterations of the algorithm may result in inaccuracies, which help explain the volatility of the difference in individual risk between repeated myopic and dynamic optimization procedures.

Our analysis tackles the exhaustive set of landscapes of size $n = 4$, allowing us to study the steady-state patterns emerging from any initial condition, replicate existing results in larger landscapes, and shed light on the mechanisms underlying the wildland dilemma.

Increasing landscape size is incompatible with this approach, as the set of possible landscapes becomes quickly unmanageable. To conserve our exhaustive approach, different proof mechanisms would be required. Nonetheless, if landscape size is of the essence for actual policy recommendation, so are other layers of information such as habitat quality, treatment costs, and values at risk heterogeneity. These other layers would reduce the computational burden, and we believe our results, targeting the most cost-efficient, risk-reducing, and habitat-conserving strategies, would still apply.

In our model, we use a simple relationship to characterize the link between the successional stage, habitat formation for a single species, and wildfire risk and severity. This choice is motivated by the existence of a lower bound for a fire return interval and drives our ability to adopt our exhaustive approach. Increasing the number of seral stages would help to complexify the relationships governing habitat formation and wildfire risk and severity: in some ecosystems, wildfire risk and severity may be higher for young vegetation than for older and may not be linear (Taylor et al., 2014). On the other hand, some species may require old-growth forests to survive, not 'young' forests, and old-growth forests may also be more fire-resilient (Lesmeister et al., 2021). As the number of successional stage augments, convergence towards steady-state landscape cycles would take longer, but we hypothesize it would still occur, as long as a final stage can

be reached. Moreover, as long as wildfire risk and habitat quality are in conflict, a trade-off would govern treatment allocation. Multiple successional stages may be targeted for fuel treatment, depending on their location and properties, but we claim the general mechanism would still apply: in a graph weighted for different risk and habitat properties, centrality and connectivity would still guide treatment allocation.

We implicitly assume that focusing on a given species' habitat would also provide habitat for a variety of species and be conducive to functional diversity. However, this does not imply that all species would benefit from maintaining a given habitat type (Saab et al., 2022). Moreover, the lack of structural diversity may cause the trophic web of the targeted species to collapse. Therefore, management objectives should include structural diversity. In this case, landscapes could not satisfy extreme habitat connectivity targets and diversity targets. For intermediate goals, however, we claim that treatment allocation would still aim at fragmenting the landscape, and node centrality and connectivity would still govern allocation.

We chose to abstract from a stochastic ignition process affecting the landscape, and assumed a fully deterministic scenario. In our set-up, we assume that risk causes damages, not the realization of risk, hence we focus on a worst-case scenario. In a stochastic setting where ignition depends on the time since the last occurrence of fire, and/or the quantity of biomass in each patch, treatment location and landscape structure would be modified to account for the "free" treatments caused by fire, and the differential probabilities of wildfire occurrence. In a setting with limited successional stages, our framework is amenable to a stochastic process of wildfire. However, as we derive our conclusions from the steady-state landscape cycles, a complementary analysis of the transitional phases is required to extend our results to the stochastic case.

Focusing on the steady state, we can limit the number of landscapes to be studied. However, in doing so, we abstract away from the transitionary dynamics, which bring a lot of information on the design of optimal policies. Nonetheless, we account for transitory dynamics by relaxing conditions on the distribution of successional stages to be treated.

We use a social planner to determine the optimal allocation of treatments while safeguarding biodiversity habitat connectivity. We adopt this stance because the effects of treatments (or non-treatments) cause spatial externalities in the form of non-rival and non-excludable (e.g. public) goods (e.g. habitat connectivity) and bads (e.g. fire risk). A social planner accounts for these effects and finds the optimal location of treatments. Using a graph theoretic framework for the spatial interactions, and under the rather restrictive assumption of uniform

land ownership²¹, our framework can be mapped to individual decisions, and how to decentralize optimal policies. Indeed, recent advances in economic theory such as [Elliott and Golub \(2019\)](#) map the position of agents in a network of public good benefits, and find how negotiating can improve the collective outcome.

4 Conclusion and policy relevance

While there is a *dilemma* for land managers between lowering wildfire risk and severity and maintaining species habitat connectivity, reconciling the two objectives is not a dead end. This is an important result for land planners as biodiversity habitat targets are gradually included in policy agendas (for example, the recent pledge by the participants to the Conference of Parties on Biodiversity in Montreal to preserve 30% of land and oceans by 2030 for biodiversity²²). It shows that if policymakers can commit to a given budget over time, these biodiversity targets can be reached and a management cycle that minimizes wildfire risk can be implemented in wildlands. Moreover, as steady-state cycles are reached, the uncertainty over future land uses is resolved while achieving policy goals.

In the face of climate change, treatment costs are expected to increase ([Kupfer et al., 2020](#)). The decreasing marginal efficiency of budget to reduce risk highlights that as climate change increases the costs of treatments, risk, and damages will increase at an increasing rate, unless the budget is changed accordingly.

Our analysis shows that budget should be determined by factoring a careful, *ex-ante* analysis of treatment costs, the policy maker's risk aversion towards a measure of wildfire risk and severity, and ecological preferences. Indeed, low budget-to-cost ratios are incompatible with high risk and severity aversions and/or large ecological requirements.

As wildfires and biodiversity habitat destruction are challenges in the face of global warming, finding policy guidance tools is of the essence. Many studies focus on specific case studies or limited ranges of potential initial conditions. We develop a simplified ecological model of habitat and wildfire connectivity to guide policymakers in the form of general principles. Reducing wildfire risk and accommodating wildlife habitat is possible with carefully designed policies, where budget plays a key role. However, it is impossible to achieve drastic risk reduction without harming biodiversity habitat. General principles of treatment allocation in the landscape are derived, and the concepts of graph theory provide an operational toolbox to understand the underlying mechanisms, as well as an opportunity to connect to other branches of policy making such as economics.

²¹Much like Jefferson's ideal *yeoman* democracy, this would amount to divide land in equal size patches owned by different individuals

²²See Target 2 in the [Keunming-Montreal Global Diversity Framework, 2022](#)

Landscape patches that display high wildfire risk successional stages and are well connected e.g. on the shortest path to other such patches should be treated first. When habitat targets are included, tackling lower-risk **patches** is of the essence to maintain habitat connectivity.

Our article summarizes and generalizes how policies should be implemented, both in terms of budgets and spatial allocation, to protect and enhance ecosystem health.

5 Declaration

Acknowledgments

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Data availability

Given its size, steady-state cycle data is available upon request from the authors. Code for replication is available at https://github.com/sim-jean/Landscape_connectivity_dilemma

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Competing interests

The authors declare no conflict of interest.

Contribution

LM and SJ designed the study, SJ ran the computational experiment, SJ and LM analyzed the results and wrote the manuscript.

A Appendix

A Global connectivity index and graph theory

Minas et al. (2014) work with a collection of cells I . This landscape can be represented by a graph structure $\mathcal{G} = (\mathcal{V}, \mathcal{E})$. For each $v_i \in \mathcal{V}$, define a neighborhood²³ of v_i by $\Phi_i = \{v_j \text{ such that } (v_j, v_i) \in \mathcal{E}\}$. Finally, let $Q_{ij} \in \{0, 1\}$ where $Q_{ij} = 1$ if v_i and v_j are connected. Minas et al. (2014) define the following connectivity metric over a landscape:

$$z_t^* = \sum_{i \in I} \sum_{j \in \Phi_i} Q_{ij} \quad (\text{A.1})$$

For the proof, assume $Y \in \{0, 1\}^{n^2}$ such that $Y_i = 1$ if cell i is 'high risk' and 0 otherwise, and that we focus on the high risk graph on the landscape. The argument is identical in the case of mature habitat.

In graph theory, an adjacency matrix \mathcal{K} for an undirected graph is a binary, symmetric, square matrix of dimension $\text{card}(V)^2$ where $k_{ij} = 1$ if vertices i and j are connected, 0 otherwise. In our context, it is clear that $k_{ij} = Q_{ij}$. Equation A.1 can be reformulated as :

$$\begin{aligned} Y' \mathcal{K} Y &= \sum_j \left(Y_j \sum_i Y_i k_{ij} \right) \\ &= \sum_j \sum_i (Y_j Y_i k_{ij}) \end{aligned}$$

And notice that $Y_j Y_i k_{ij} = Q_{ij}$, so :

$$\begin{aligned} Y' \mathcal{K} Y &= \sum_{i \in I} \sum_{j \in \Phi_i} Q_{ij} \\ &= \sum_j \left(Y_j \left(Y_j k_{jj} + \sum_{i \neq j} Y_i k_{ij} \right) \right) \end{aligned}$$

Given the symmetric nature of \mathcal{K} , $\forall i \neq j$, $k_{ij} = k_{ji}$. Each cell is connected to itself

²³Notice that *belonging to the neighborhood of* is a symmetric binary relationship e.g. if $v_i \in \Phi_j \iff v_j \in \Phi_i$, as we are working with undirected graphs

so $k_{jj} = 1$. Additionally, as $Y_i \in \{0, 1\}$ then $Y_i^2 \in \{0, 1\}$:

$$\begin{aligned} Y' \mathcal{K} Y &= \sum_j \left(Y_j^2 + \sum_{i \neq j} Y_i Y_j k_{ij} \right) \\ &= \sum_j Y_j + \sum_j \left(\sum_{i \neq j} Y_j Y_i k_{ij} \right) \\ &= \sum_j Y_j + \sum_j d_j \end{aligned}$$

The first sum is the number of high risk cells, i.e. $\text{card}(\mathcal{V})$. In the second sum, $\sum_{i \neq j} Y_j Y_i a_{ij}$ is the degree of each vertex j excluding self loops. In a graph with no self loops, by definition, $\sum_j d_j = \text{card}(\mathcal{E})$.

Hence, for a set of cells I reformulated in terms of graph theory :

$$\sum_{i \in I} \sum_{j \in \Phi_i} Q_{ij} = \text{card}(V) + 2\text{card}(E) \quad (\text{A.2})$$

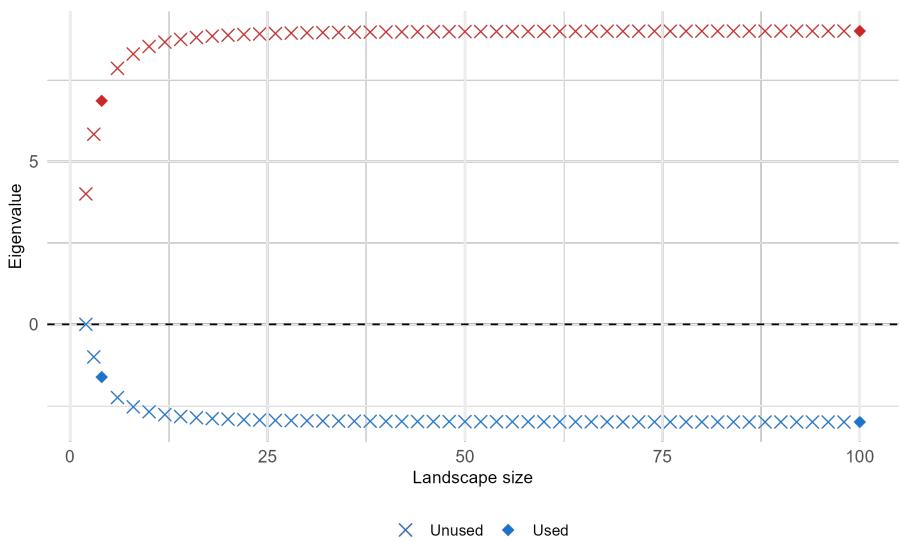


Figure 2.A: Maximum and minimum eigenvalues of \mathcal{K} depending on graph size
In red, the maximum eigenvalues, in blue, the minimal eigenvalues. Diamond-shaped points represent values actually used for the present study. Dotted line at 0

B Large scale landscape characteristics

Age classes			Autocorrelation
<i>Juvenile</i>	<i>Adolescent</i>	<i>Mature</i>	
33%	33%	34%	0.5
10%	45%	45%	0.7
10%	10%	80%	0.9
10%	80%	10%	1.3
			1.8

Table 2.B: Summary of the large scale simulated landscapes characteristics

For each distribution profile governing the number of cells in a landscape, all of the spatial autocorrelation values are applied making 4×5 landscapes

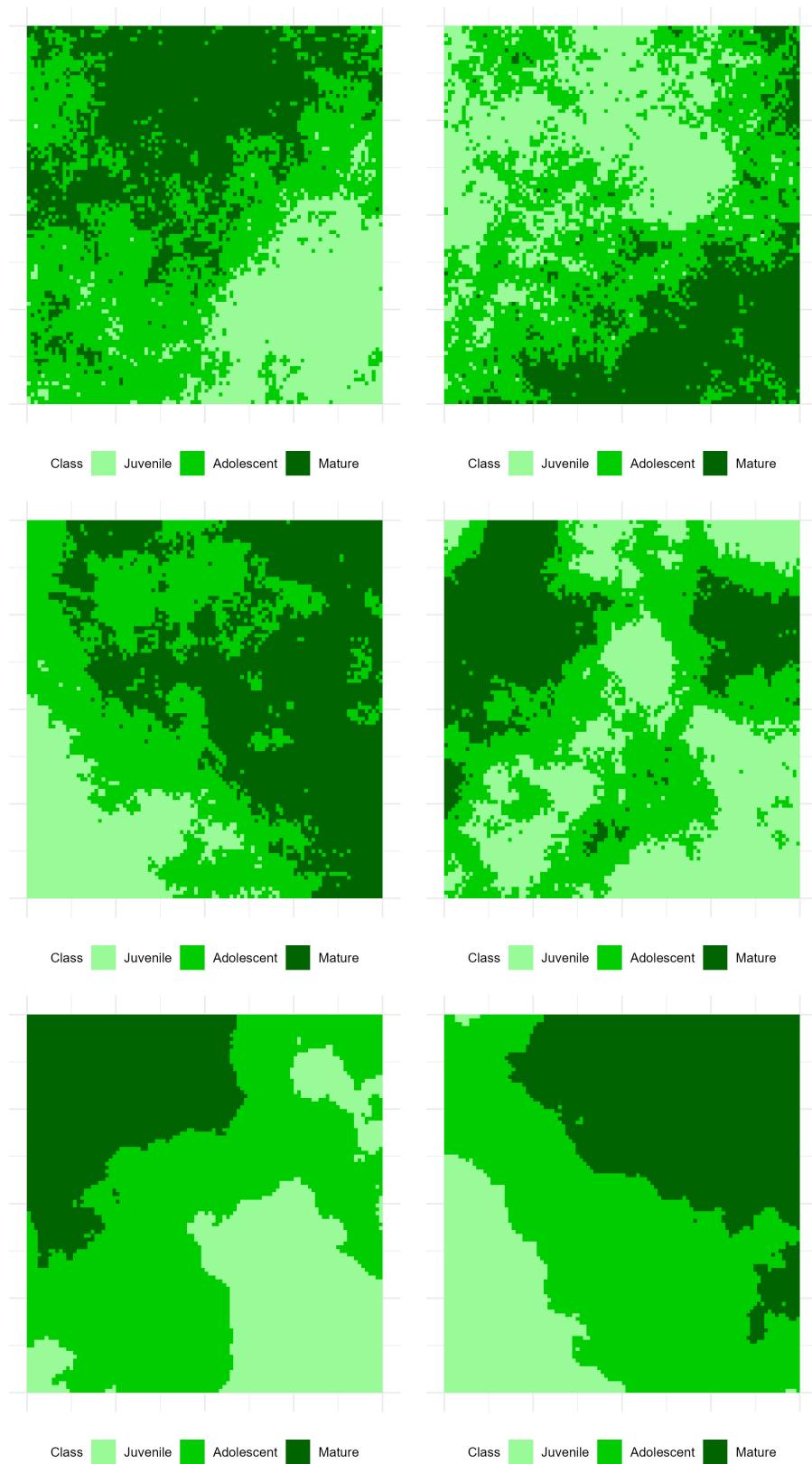


Figure 2.B: Examples of large scale landscapes with uniform distribution (left) and skewed towards *adolescent* (right)and low, middle and large spatial autocorrelation

C Landscape indicators

Area We use the number of vertices (nodes) for both graphs:

$$Area(\mathcal{G}_t) = \text{card}(V_{\mathcal{G}_t}) \text{ with } \mathcal{G}_t \in \{\mathcal{B}_t, \mathcal{F}_t\} \quad (\text{A.3})$$

Simpson diversity index: Let p_i be the proportion of landscape A_t in a given successional stage²⁴, the Simpson diversity index is :

$$SIDI = 1 - \sum_{i \in \{Juv, Ado, Mat\}} p_i^2 \quad (\text{A.4})$$

Landscape shape index: following McGarigal and Marks (1995), the adapted LSI index from Patton (1975) in a raster landscape is:

$$LSI = \frac{0.25 \times \text{perimeter}(G)}{n} \quad (\text{A.5})$$

Where $\text{perimeter}(G)$ is the perimeter of the cells comprised in the graph as vertices.

Successional Stage Heterogeneity Index: let d_{ij} be a binary variable such that $d_{ij} = 1$ if patch i and j share the same successional stage. Define \mathcal{J} as the set of neighbors in 4 directions (north, south, east, west) of cell i ²⁵. The successional stage heterogeneity index is :

$$SSHI = 1 - \frac{1}{N} \sum_{i=1}^N \left(\frac{\sum_{j \in \mathcal{J}_i} d_{ij}}{\text{card}(\mathcal{J}_i)} \right) \quad (\text{A.6})$$

²⁴Let $Juv = \{a_{ijt} \text{ such that } a_{ijt} = 0\}$, then $p_{Juv} = \frac{\text{card}(Juv)}{n^2}$

²⁵The set \mathcal{J}_i varies with cell i to account for edge effects

D Treatment centrality indicators

Betweenness centrality: take a graph $\mathcal{G}(V, E)$ and let σ_{st} be the total number of shortest paths from node s to t and $\sigma_{st}(v)$ be the number of those paths that pass through v , for $\{s, t, v\} \subset V$, betweenness centrality is given by :

$$g(v) = \sum_{s \neq v \neq t} \frac{\sigma_{st}(v)}{\sigma_{st}} \quad (\text{A.7})$$

Eigencentrality: let $\mathbf{A} \in \mathcal{M}_{n,n}$ be the adjacency matrix of graph and $a_{i,j} = 1$ if vertices i and j are connected. Let $\lambda \in \mathbb{R}^{n^2}$, and a vector $\mathbf{x} \in \mathbb{R}^{n^2}$, such that $\lambda \mathbf{x} = \mathbf{A}\mathbf{x}$ e.g λ is the eigenvalue of matrix \mathbf{A} . Using this eigenvalue, *centrality scores* are computed as :

$$\text{score}(x_i) = \frac{1}{\lambda} \sum_{j \in V_{\mathcal{G}_t}} a_{i,j} x_j \quad (\text{A.8})$$

Subgraph centrality: In a graph, a *walk* is a sequence of adjacent vertices in a graph. A *closed walk* is a walk with identical beginning and ending vertices, and can be of order k e.g. of length equal to k edges. The number of *closed walks of order k* is found using the adjacency matrix \mathbf{A} of a graph. Let $\mu_k(i)$ be the number of closed walks of order k starting at i :

$$\mu_k(i) = (\mathbf{A}^k)_{i,i}$$

Subgraph centrality is defined as :

$$SC(i) = \sum_{k=0}^{\infty} \frac{\mu_k(i)}{k!} \quad (\text{A.9})$$

[Estrada and Rodríguez-Velázquez \(2005\)](#), who define this notion, show that it can be reformulated with the eigenvalues and eigenvectors of the adjacency matrix \mathbf{A} of a graph $\mathcal{G}(V, E)$ of order n . Let v_1, \dots, v_n be a an orthonormal basis of \mathbb{R}^N composed of eigenvectors of \mathbf{A} associated to the eigenvalues $\lambda_1, \dots, \lambda_N$, and let v_j^i be the i -the component of v_j , then subgraph centrality can be expressed, for all $v \in V$:

$$SC(i) = \sum_{j=1}^N (v_j^i)^2 e^{\lambda_j} \quad (\text{A.10})$$

E Additional figures

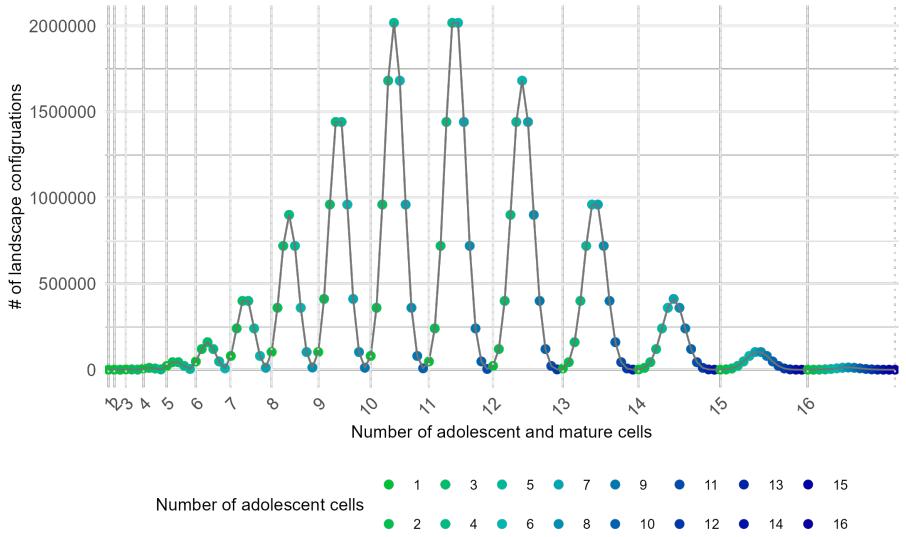


Figure 2.C: Distribution of number of landscapes depending on number of *juvenile*, *adolescent*, and *mature* cells for sizes $\in \{3, 4\}$

Each number on the x axis represents the cumulated number of *adolescent* and *mature* cells among the landscape. Between each number on the x -axis is the number of *adolescent* cells among the number of cumulated *adolescent* and *mature* cells. For example, the highest point of the distribution, between 11 and 12, represent the number of possible combinations of landscapes with a cumulated number of *adolescent* and *mature* cells of 11, with 5 and 6 *adolescent* cells e.g. 6 and 5 (respectively) *mature* cells

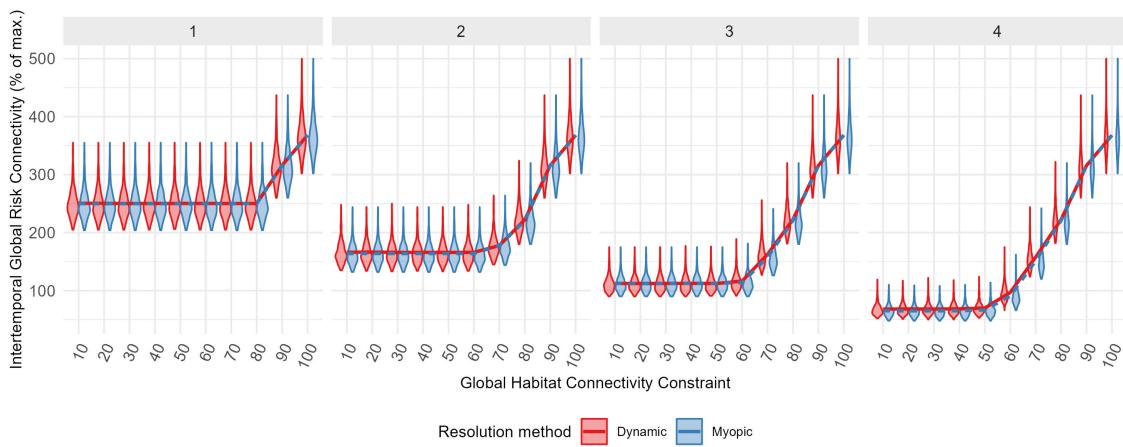
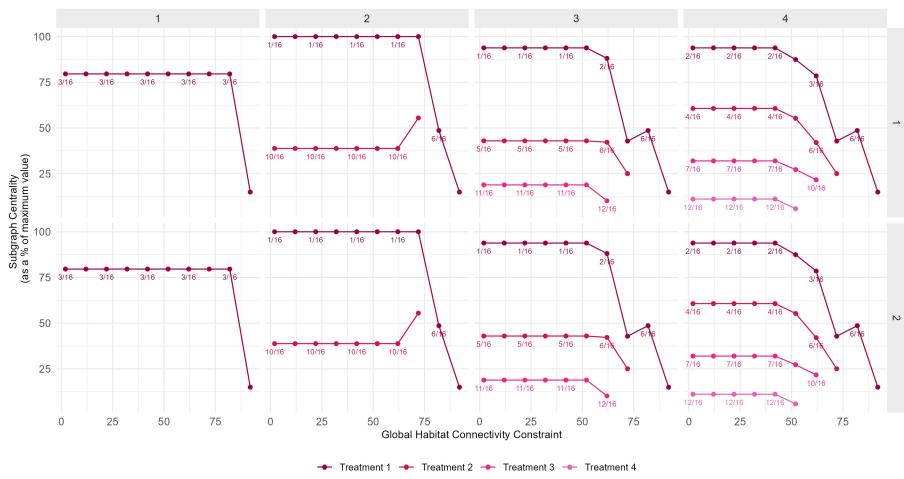
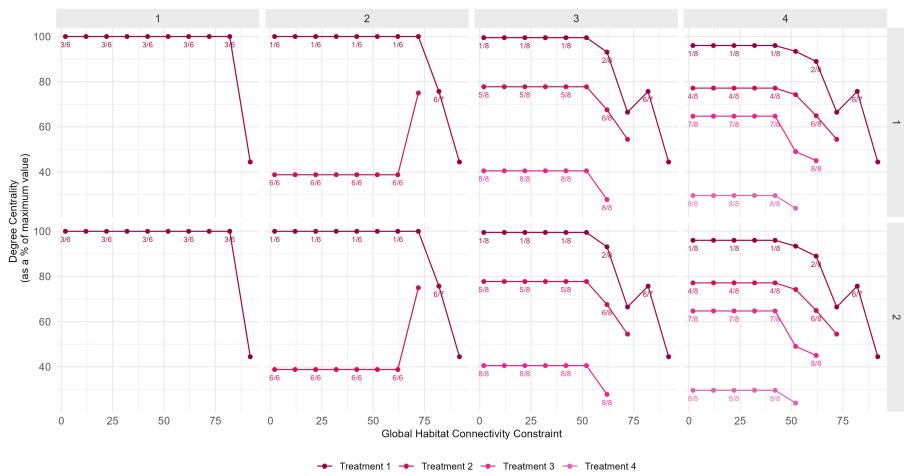


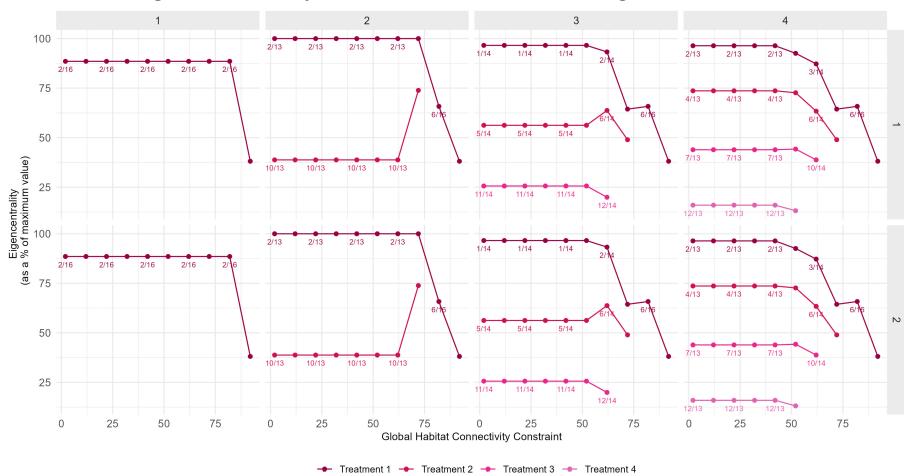
Figure 2.D: Production possibility frontiers between global risk connectivity and global habitat connectivity constraints across budget constraint levels for the sample of representative landscapes of size $n = 4$ between repeated myopic and dynamic optimization procedures for $T = 5$



(a) Subgraph centrality of treated vertices (average value and rank)



(b) Degree centrality of treated vertices (average value and rank)



(c) Eigencentrality of treated vertices (average value and rank)

Figure 2.E: Average subgraph and degree centralities, and eigencentrality, across steady-state cycles for budget and global habitat connectivity constraints

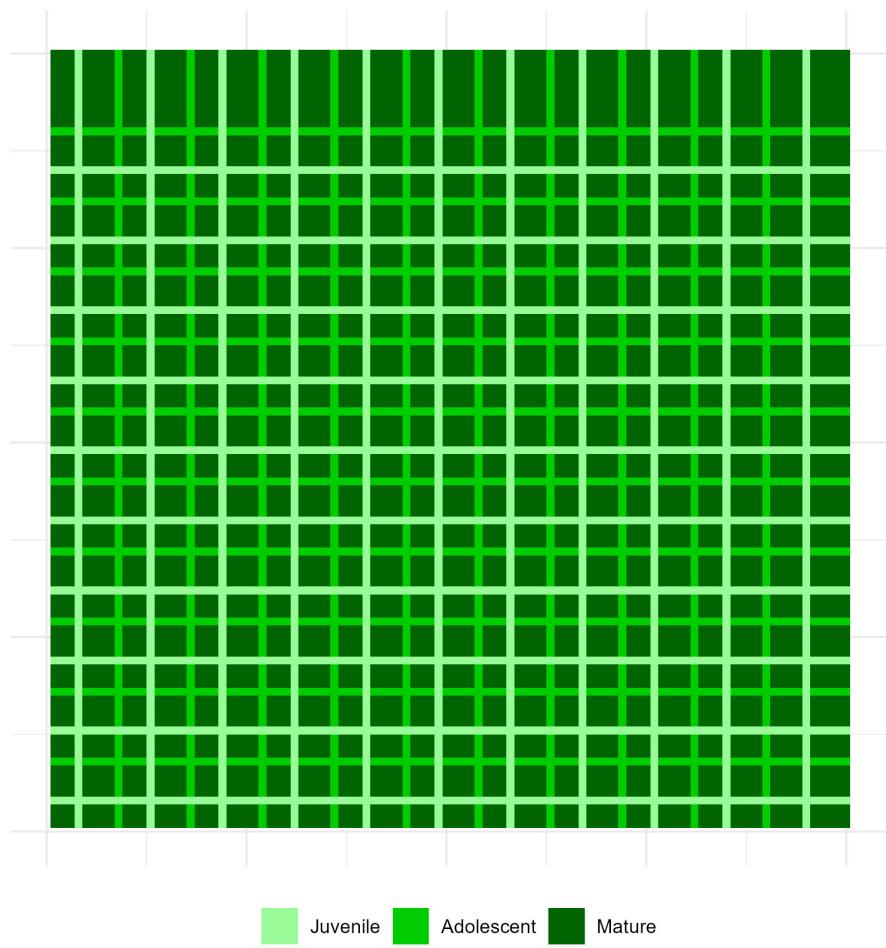


Figure 2.F: Illustration of grid treatment rule

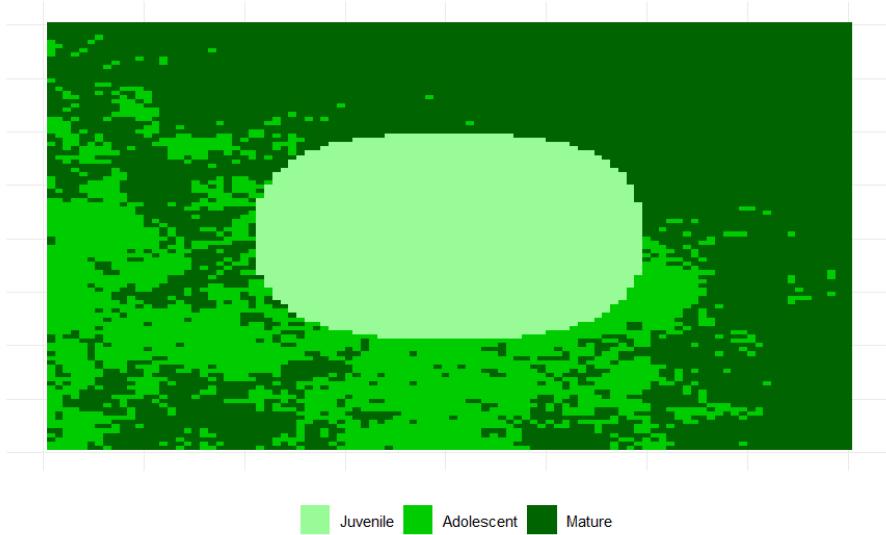


Figure 2.G: Example of location of treatment with adaptive policy on large scale landscape

F Additional tables

The main model is :

$$\begin{aligned} DiffRisk_i = & \beta_0 + \beta_1 Budget_i + \beta_2 Constraint_i + \beta_3 Number2_i \\ & + \beta_4 LSI_i + \beta_5 Simpson_i + \beta_6 SSHI_i + \beta_7 NumberComponents_i \\ & + \beta_8 GlobalRiskInitial_i \end{aligned} \quad (\text{A.11})$$

A second model is tested :

$$\begin{aligned} DiffRisk_i = & \beta_0 + \beta_1 Budget_i + \beta_2 Constraint_i + \beta_3 LSI_i + \beta_4 Simpson_i \\ & + \beta_5 SSHI_i + \beta_6 GlobalRiskInitial_i + \beta_7 Constraint_i \times Budget_i \\ & + \beta_8 LSI_i \times SSHI_i + \beta_9 LSI_i \times Simpson_i \\ & + \beta_{10} Constraint_i \times GlobalRiskInitial_i \\ & + \beta_{10} Constraint_i \times GlobalRiskInitial_i \\ & + \beta_{11} Budget_i \times GlobalRiskInitial_i \\ & + \beta_{12} Constraint_i \times Budget_i \times GlobalRiskInitial_i \end{aligned} \quad (\text{A.12})$$

Table 2.C: Summary of model A.11: linear regression of risk differences between optimization procedures on landscape characteristics

	<i>Dependent variable:</i>
	<i>DiffRisk_i</i>
Constraint	−0.009*** (0.001)
Budget	0.753*** (0.030)
Number of 2s	0.034 (0.081)
LSI	−0.139 (0.230)
Simpson	−0.654 (0.618)
Successional Stage Heterogeneity Index	−0.990** (0.490)
Number of components	−0.048 (0.075)
Global Risk Connectiivty	0.002 (0.020)
Constant	1.183*** (0.270)
Observations	25,840
R ²	0.028
Adjusted R ²	0.027
Residual Std. Error	5.375 (df = 25831)
F Statistic	91.470*** (df = 8; 25831)

Note:

*p<0.1; **p<0.05; ***p<0.01

Table 2.D: Summary of model A.12: linear regression of risk differences between optimization procedures on landscape characteristics

	<i>Dependent variable:</i>
	<i>DiffRisk_i</i>
Constraint	−0.008* (0.005)
Budget	0.661*** (0.108)
LSI	0.584 (0.643)
Succesional Stage Heterogeneity Index (SSHI)	1.087 (2.203)
Simpson	−1.612 (2.532)
Global Risk Connectivity	−0.022 (0.023)
Constraint × Budget	0.0002 (0.002)
LSI × SSHI	−1.446 (1.605)
LSI × Simpson	0.201 (2.018)
Constraint × Global Risk Connectivity	0.0002 (0.0004)
Budget × Global Risk Connectivity	0.017** (0.009)
Constraint × Budget × Global Risk Connectivity	−0.0002 (0.0001)
Constant	0.662 (0.647)
Observations	25,840
R ²	0.028
Adjusted R ²	0.027
Residual Std. Error	5.375 (df = 25827)
F Statistic	61.872*** (df = 12; 25827)

Note:

Chapter 3

Fences : the economics of movement in mobile public goods

Abstract

This article examines the management of spatially distributed renewable resources—specifically wildlife and infectious diseases—through the lens of economic and spatial analysis. I focus on "bads" like invasive species and diseases, which cause economic and ecological harm, and utilize population control and fencing as central mechanisms. I analyze how fencing influences resource flow and connectivity. On the one hand, in the presence of ecological and economic heterogeneities, fencing can be used to leverage spatial arbitrage opportunities. On the other hand, while promoted as a tool to incentivize the internalization of costs associated with "bads", they may undo what Nature has rightfully done. In this sense, while fencing may be welfare improving in a setting with initially poor connectivity, an uncoordinated use of fencing, although welfare improving, is not welfare maximizing. The study develops a theoretical model that integrates aspects of stock and patch connectivity management and explores both cooperative and non-cooperative management strategies. The findings indicate that optimal management often requires a nuanced understanding of the spatial dynamics and economic costs associated with different control strategies. We present a series of propositions that characterize the conditions under which fencing and resource control strategies can be optimized, including the interaction effects of exclusionary and trap effects. This article contributes to the literature by highlighting the role of spatial heterogeneity in the management of renewable resources and providing insights into the formulation of more effective environmental policies, as it analyzes how to design policies on a subset of the landscape, to maximize economic and ecological benefits.

JEL codes : Q20, Q24, R12

Keywords : spatial resource management, invasive species; fencing and control strategies; optimal management; non-cooperative equilibrium; second-best policy.

1 Introduction

In the 1600s, the Ma'ohi, the Indigenous People of the Society Islands in French Polynesia (?) built vast fish traps, using organic fences, stakes, and poles. On the island of Huahine, stones set vertically, forming V-shaped enclosures trapped schools of fish coming down to sea, from a shallow salt water lake. Fish were pulled towards the sea with the tides, and became trapped in basins. Fish were then harvested using nets in the shallow lake. Managing the fish stock for the community amounted to more than harvesting. Trapping, thus reducing the extent of fish school mobility, was instrumental¹.

This should ne moved to the conclusion section : works with goods, but also with mixed goods and bads

Centuries later, in the US populations of white tailed deers have skyrocketted to an estimated 36 million, with exceptionally high densities in the South East ([Hanberry and Hanberry, 2020](#)). At high densities, deer populations threaten the regeneration of forests as they influence species composition and abundance through browsing, hence damaging people's properties ([Hanberry and Abrams, 2019](#)). Moreover, risks of zoonosis and epidemics increase with large populations. While large scale culling policies have been implemented, landowners have increasingly resorted to other methods, such as repellents, or fencing. Eight-foot or higher woven-wire fences have been used to protect agricultural land such as orchards as well as private homes, to limit the damage done by growing deer populations ([Caslick and Decker, 1979](#)). Eventually, during the COVID 19 pandemic between 2019 and 2023, international airports and ports were shutdown, and extensive lockdown policies were implemented worldwide. By avoiding contact between infected and non-infected people, these policies aimed at slowing the spread of the pandemic², while managing the extent of the economic losses associated with frozen national and international economies.

These three examples display cases of management of spatially distributed renewable resources. Indeed, fish deer populations, and pandemics grow through time, depending on the size of the population. Moreover, they move through oceans, land, jurisdictions and countries. These examples highlight that the management of spatially distributed renewable resources, whether goods or bads, involves at least two layers : managing the stock, and how it moves through space. Indeed, fishing culling, and curing act as stock management measures,

¹Modern applications, such as fish fences on Pacific islands, are detrimental to seascapes connectivity, and destroy the sea bed, see [Exton et al. \(2019\)](#)

²In a given population, where successive infections are possible, lockdown policies aim at diminishing the basic reproduction number \mathcal{R}_0 , which measure "expected number of infections generated by a single and (typical) infected individual during their entire infection period" see Saldan and Velasco for a primer SIR modeling applied to COVID 19

while weirs keep the fish in a given area, repellents and fences keep the deers away, and lockdowns avoid spread from infected to non infected people. Finally, in all cases, policies aimed at managing the movement of the resource are more efficient in one way than the other : weirs avoid outflow of fish, but allows inflow; wildlife exclusion fencing often have doors to let animals escape, and to a certain extent, people were prohibited from entering a country more than leaving one during the COVID 19 pandemic.

First, the decentralized management of spatially distributed renewable resources is made difficult by the spatial externality they generate. When communities compete for mobile fish, they anticipate part of the school to migrate to other communities, and tend to overharvest, as they do not have secure property right over the whole resource through time ([Kaffine and Costello, 2010](#)). In the case of deers, free riding on neighbor's culling may deter people to cull the population to efficient levels ([Costello et al., 2017](#)). In this sense, patch connectivity, in a non cooperative setting, generates inefficiencies. As a consequence, fences appear as welfare improving, as they diminish patch connectivity and therefore contribute to solving the spatial externality. If a fish stock no longer migrates, communities would tend to harvest it in a more sustainable way. If on a given property, deers have no chance of re-entering, then one may undertake efficient culling measures. However, from a welfare perspective, fencing may undo what Nature has rightfully done. Considering spatial heterogeneity in marginal returns to harvesting or culling, and biological productivity, a resource may flow naturally to where it is best managed. In this case, although fencing can solve the spatial externality and promote efficient resource use, it would not maximize welfare. Second, spatially distributed renewable resources live on intricate institutional maps, between private and public land and sea. As a result, optimal harvesting and fencing may be difficult to decentralize. Hence, figuring the second best policy mix to best manage spatially distributed renewables is a challenge.

This approach can be viewed as application of the spatial trade literature to ecological networks. For example, [Donaldson and Hornbeck \(2016\)](#) shows that railroads have a global effect, as they change the "market access" of each county, accounting that local changes in "market access" have spillover effects onto other counties. More generally, to understand the general equilibrium effect of domestic policies on international trade patterns, the use of a structural gravity model is inevitable (e.g. 'the new quantitative trade model' e.g. [Arkolakis et al. \(2012\)](#)). However, the gravity equation fails at identifying the impact of country specific determinants of trade flows, e.g. multilateral resistance terms ([Anderson and van Wincoop, 2003](#)). In this article, I analyze the changes in local fencing patterns have local and spillover effects, and can be seen as changing multilateral resistance terms in an ecological context, and show how they affect each patch, under

various management regimes.

In this article, I focus on the management of “bads”, e.g. species that cause economic damages. This includes rodents, feral pigs, deers, or predators in areas where native species prey are threatened. I develop a theoretical model à la [Costello et al. \(2017\)](#), to understand the interplay between stock and patch connectivity management. Species are harvested, grow and disperse through space, according to immutable environmental factors and expenditures that change connectivity, e.g. fences. Fences have two effects : they keep the bad out (*exclusionary effect*), and they keep the bad in (*trap effect*). In what follows, I assume the exclusionary effect dominates the trap effect. In most cases, exclusionary fencing keeps predators, or damaging species out, while allowing entrapped animals to leave the area³..

First, I study the optimal policy mix between stock and dispersal rate management. When costs of control are heterogeneous, the sole owner leverages the spatial arbitrage opportunity, and fences only have an exclusionary effect, the sole owner redirects the population stock to where it controlled at the cheapest cost. In doing so, she reduces the population in more expensive patches further than when connectivity is absent. Allowing for resource redispach, she controls more of the species. When fencing has both an exclusionary and trap effect, cost heterogeneity does not suffice to redirect resource. If biological productivity is larger in relatively costlier patches, trapping them can increase the aggregate cost of the invasive species. Therefore, fencing only occurs when biological productivities and control costs are inversely correlated.

Second, I characterize the non cooperative equilibrium in harvesting and fencing. When fencing only displays an exclusionary effect, and fencing is costless, every patch owner fences to the maximum. In doing so, they isolate their patch from the rest of the landscape, and control as if they were isolated from other patches. While this results in a more efficient level of control than in the case of uncontrolled spatial dependence, this is not welfare maximizing : as a matter of fact, the non cooperative equilibrium, while solving the spatial externality, does not leverage the spatial arbitrage opportunity provided by heterogenous costs of controlling and biological productivities. When fencing displays (unequal) exclusionary and trap effects, best response functions are non monotonous. In this case, increasing fencing is not always optimal, and the Nash equilibrium results in suboptimal fencing, although closer to the optimal solution.

Third, decentralizing the optimal policy on public and private land may prove impossible. Therefore, I investigate the second best allocation, where some patches of land can enforce the optimal policy mix, while others cannot, and fencing is

³This can be viewed as an ecological version of inward and outward multilateral resistance terms ([Anderson and van Wincoop, 2003](#))

restricted. I generalize insights from [Costello and Polasky \(2008\)](#) and [Costello et al. \(2017\)](#) to understand how to optimally control the stock and connectivity when only a subset of patches can be regulated. I show that implementing the first best policy mix, which reshuffles resources to the most cost effective patch is always best. The second allocation is decentralizing an uncoordinated equilibrium, as the spatial externality is resolved and future damages and costs are internalized. However, when a policy maker can only choose 1 instrument, decentralizing optimal fencing with uncoordinated control is the worst outcome, while decentralizing optimal control with uncoordinated fencing is not the worst outcome. Finally, I use a simplified empirical application, using simulated data, to illustrate the optimal control and fencing in the presence of cost and biological heterogeneity, as well as the non cooperative equilibrium. Additionally, I characterize the welfare effects of different management scenarios, depending on the starting policy ground.

This article proceeds as follows : section 2 draws lessons from the existing literature, section 3 explains the models main mechanisms. In section ??, I establish results for the optimal fencing and controling of a public bad, while section ?? looks at the uncoordinated equilibrium. Eventually, section ?? illustrates the findinds, section ?? concludes. Proofs can be found in the appendix (see section A).

2 Related literature

There is a vast literature that investigates the optimal control, eradication and detection of invasive species (see Epanchin Niell for a review). A much scarcer one looks at the spatial nature of the management of public bads and/or invasive species. Early approaches,[Huffaker et al. \(1992\)](#),[Bhat et al. \(1996\)](#) analyze various management regimes (cooperative, isolated, and coordinated) to deal with the presence of beavers on private land. Movement between patches corresponds to a density dependent pattern, which is, funny enough, an adaptation of Stenseth's "*social fence*" hypothesis [Stenseth \(1988\)](#). In this framework, migration is entirely driven by relative densities. Therefore, optimal stock management needs to account for these migratory effects. With this analysis, [Huffaker et al. \(1992\)](#) and [Bhat et al. \(1996\)](#) limit themselves to two patches, for analytical and computational tractability. A different approach, viewing space as a continuum, has considered options to halt the progression of an invasive species, using barreer zones, to ultimately slow the rate of spread [Sharov and Liebhold \(1998\)](#). While theoretically appealing, this approach may not be suited for operational concerns, whereby optimization on a continuum space is difficult, especially in various directions. In the wake of [Brown and Roughgarden \(1997\)](#), [Bulte and](#)

van Kooten (1999), numerous models of invasive species have been developed in economics, taking advantage of familiar optimization structures. For example, Blackwood et al. (2010) develop a linear quadratic framework to study the control of an invasive plant species. Taking advantage of the stock independent nature of migration patterns and of the linear quadratic structure, the authors solve the control and prevention problem at a large spatial scale. In more recent work, Costello et al. (2017) develop a large scale model of public goods, characterized by exogenous dispersal, and analyze the potential for eradication in a connected landscape. In doing so, they analyze the effects of varying connectivity parameters, without acknowledging for the potentially endogenous nature of dispersal. Finally, a wealth of papers, in the wake of Sanchirico and Wilen (1999), several papers (Albers et al., 2010; Ambec and Desquilbet, 2012) have investigated the use of policies to halt the spread of invasive species, including mandatory refuges, albeit uniform. While these articles view dispersal as a characteristic that can be influenced, they do not consider the optimal management, or lack thereof, of dispersal. Finally, Janmaat (2005) highlights the role of dispersal in a fishery, and other parameters, to assess the extent of the tragedy of the commons. Interestingly, in that article, Janmaat states that “*until ‘fences’ are available to contain the ‘wandering’ offspring, management zones would have to be large. This would minimize the spillover, bringing the incentives of the ‘owner’ into line with maximizing the total return generated by the resource*”. The contribution of the article is framed as how to adapt regulation to a given migration pattern. In this article, I reverse the approach for terrestrial species, of sufficient size such that their dispersal can (more or less) be managed. In this article, I build on these frameworks by using a discretized, raster-type landscape, with metapopulation dispersal across patches. Instead of analyzing how policies should adapt to dispersal, and I analyze how policies can shape dispersal, and what happens in the case where management is incomplete.

3 A dynamic spatial model of renewable goods management : fencing and controlling

1 Spatial ecology

Assume N patches indexed $i \in \{1, \dots, n\}$ with a renewable resource. In a given period, the resource stock X_{it} is harvested by h_{it} , and grows according to the remaining stock, defined as $e_{it} = X_{it} - h_{it}$, such that the pre-migration population in patch i in $t + 1$ is $g_i'(e_{it})$ such that $g_i'(e_{it}) \geq 0, g_i''(e_{it}) \leq 0$.

Moreover, after the resource grows, it disperses through space (see fig. 3.A for a summary of the model timing). This is consistent with continuous metapopula-

tion models (Sanchirico and Wilen, 1999; Bulte and van Kooten, 1999), although discretized (Costello et al., 2017). I assume that dispersal exclusively depends on exogenous, immutable environmental characteristics, and fencing. Density effects on migration rates are not considered in this model.

Dispersal rates between patches depends on directional fencing expenditures in both patches, with $d_{ijt+1} \equiv d_{ijt+1}(f_{it}^j, f_{jt}^i)$, where f_{it}^j measures the amount of fencing in patch i in direction of patch j . The inflow of invasive species from i to j , $d_{ijt+1}(f_{it}^j, f_{jt}^i)$ decreases with f_{jt}^i . I call this the “exclusionary effect”: fences keep nuisances out of j . When fencing in i at f_{it}^j , the outflow of invasive species from i to j decreases as well, as species get trapped in i . This effect is the “trap effect”: fences trap the nuisance in. In real life applications, either the exclusionary or trap effects tend to dominate. Indeed, actual traps have a limited exclusionary effect, while exclusionary fencing always features a potential escape for trapped animals. Nonetheless, I focus on a symmetric case : in this set-up, fences keep as much in as they keep out. Fencing reduces the inflow from i to j at a decreasing rate, whether it is undertaken in patch i or j . The rate of patch retention d_{iit+1} is the remainder after migrations from i to j . Dispersal rates are ultimately affected by immutable environmental factors (landscape discontinuities such as roads, rivers, moutains; altitude and terrain ruggedness etc). These immutable factors are pairwise symmetric. Eventually, dispersal rates sum to 1. Therefore :

$$d_{ijt+1} : \mathbb{R}^+ \times \mathbb{R}^+ \rightarrow [n_{ij}, m_{ij}] \subset [0, 1] \quad (3.1)$$

$$\underbrace{\frac{\partial d_{ijt+1}}{\partial f_{jt}^i}}_{\text{Exclusionary effect}} = \underbrace{\frac{\partial d_{ijt+1}}{\partial f_{it}^j}}_{\text{Trap effect}} \leq 0 \quad (3.2)$$

$$\sum_{j \neq i}^N d_{ijt+1}(f_{it}^j, f_{jt}^i) + d_{iit+1} = 1 \quad (3.3)$$

Where n_{ij}, m_{ij} are the immutable bounds to dispersal rates, and second-order derivatives are (weakly) positive.

2 Spatial economy

The presence of bads is costly in each patch via two channels, modeled as in Costello et al. (2017). First, the presence of bads implies property specific control expenditures. The larger the stock, the lower the marginal cost of control, hence accounting for a stock effect, where the marginal cost of control $c_i(s)$ is decreasing with stock size, $c'_i(s) < 0$. The total cost of controlling down to residual stock e_{it} is $\int_{e_{it}}^{X_{it}} c_i(s) ds$.

Additionally, the presence of the residual stock causes heterogeneous marginal

damages (for example, deers cause more damages to orchards and managed forests than to meadows) $k_i(s)$, which increase with stock size $k'_i(s) > 0$, resulting in convex damages. The total damages caused by the residual stock is $\int_0^{e_{it}} k_i(s) ds$.

Eventually, fencing is costly, with heterogeneous costs (driven by terrain, difficulty of access, type of fence etc) across patches. The marginal cost of fencing $\gamma_i(s)$ is weakly increasing with fencing $\gamma'_i(s) \geq 0$ and $\gamma_i(0) > 0$. The total cost of fencing is $\sum_{j \neq i} \int_0^{f_{it}^j} \gamma_i^j(s) ds$.

The total cost in each patch i and period t is :

$$C_i(e_{it}, X_{it}, f_{it}^1, \dots, f_{it}^j) = \int_{e_{it}}^{X_{it}} c_i(s) ds + \int_0^{e_{it}} k_i(s) ds + \sum_{j \neq i} \int_0^{f_{it}^j} \gamma_i^j(s) ds \quad (3.4)$$

The patch-period specific cost depends on current patch specific decisions, as well as past decisions by other agents, which influence the stock of bad in patch i at the beginning of period t .

4 The decentralized equilibrium

I assume that the N patch owners can decide individually their fencing level, as well as their control level. The decision is complicated because each player needs to know how the initial stock level across the whole landscape, as well as the distribution of marginal cost and damages, growth and fencing costs. Moreover, it has to anticipate how these decisions will affect the future : a lower level of control today will lower the future marginal cost of control, but may cause larger damages.

Mention the equilibrium concept?

Finally, the strategy space is two-dimensional : indeed, each player not only has to chose their control level (as in [Costello et al. \(2017\)](#)) but also their fencing level.

To simplify the decision, I develop a two-stage dynamic game, according to Figure 3.A. In each period, patch owners first decide on the fencing level in each direction, conditional on fencing decisions in other patches. After having observed the fencing levels in every direction, and how dispersal is affected, they decide on the control level in their patch. This game is solved using backwards induction. Best response functions for controls are computed, conditional on fencing levels. Because each patch owner's decision depends on the control level, a subgame non cooperative equilibrium emerges where equilibrium controls only depend on fencing levels, and patch specific characteristics $\hat{e}_{it}(\mathbf{f})$ as in ([Costello et al., 2017](#)). Taking this equilibrium result, patch owners choose the level of fencing to minimize their total cost.

- First, establish the second step equilibrium
- Then Establish the first stage equilibrium
- Should I enunciate the FOCs first?

A Appendix

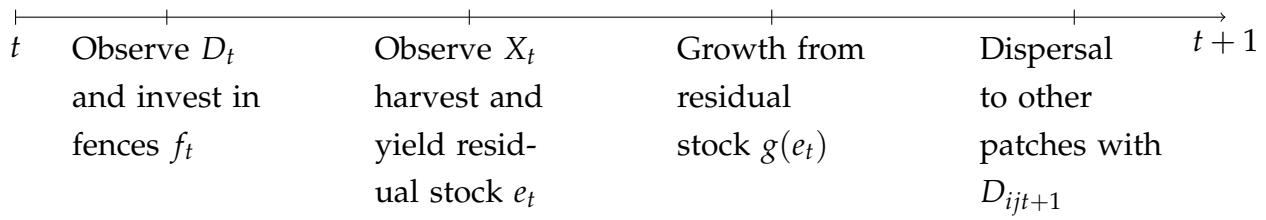


Figure 3.A: Timing of the model

A Proofs

A.1 Proof of concavity

Starting from the first order conditions in equations ?? and ?? and omitting time subscripts in t (but remaining in $t + 1$), the second order derivatives are :

$$\frac{\partial^2 C_i}{\partial e_i^2} = -c'_i(e_i) + k'_i(e_i) + \delta \left[g''_i(e_i)(1 - d_{ij}(f_i, f_j))c_i(X_{it+1}) + (g'_i(e_i)(1 - d_{ij}(f_i, f_j)))^2 c'_i(X_{it+1}) \right] \quad (\text{A.1})$$

$$\frac{\partial^2 C_i}{\partial f_i^2} = \gamma'_i(f_i) + \delta \left[g_i(e_i) \left(-\frac{\partial d_{ij}}{\partial f_i} c'_i(X_{it+1}) - \frac{\partial^2 d_{ij}}{\partial f_i^2} c_i(X_{it+1}) \right) + g_j(e_j) \left(\frac{\partial d_{ji}}{\partial f_i} c'_i(X_{it+1}) + \frac{\partial^2 d_{ji}}{\partial f_i^2} c_i(X_{it+1}) \right) \right] \quad (\text{A.2})$$

$$\frac{\partial^2 C_i}{\partial e_i \partial f_i} = \delta g'_i(e_i) \left[-\frac{\partial d_{ij}}{\partial f_i} (c_i(X_{it+1}) + c'_i(X_{it+1})g_i(e_i)) + c_i(X_{it+1}) \frac{\partial d_{ji}}{\partial f_i} g_j(e_j) \right] \quad (\text{A.3})$$

$$\frac{\partial^2 C_i}{\partial f_i \partial e_i} = -\frac{\partial d_{ij}}{\partial f_i} \delta g'_i(e_i) [c_i(X_{it+1}) + g_i(e_i)c'_i(X_{it+1})(1 - d_{ij}(f_i, f_j))] \quad (\text{A.4})$$

Chapter 4

Little downside and substantial gains result from farming of *Totoaba Macdonaldi*

This article is under review at *NPJ Ocean Sustainability* and is joint work with Julia M. Lawson (co-first author), Andrew Steinkruger, Miguel Castellanos-Rico, Garrett M. Goto, Miguel A. Cisneros-Mata, Erendira Aceves Bueno, Matthew M. Warham, Adam M. Sachs and Steven D. Gaines

Abstract

Illegal wildlife trade poses a growing threat to species globally. Where bans or policy instruments have failed, conservation farming has been considered, which aims to reduce illegal poaching by “flooding the market” with farmed product. However, predicting if farming will succeed necessitates a holistic understanding of how supply and demand interact and how markets will respond. Poaching and illegal trade for totoaba (*Totoaba macdonaldi*), currently dominated by a Mexican monopolist cartel, has continued unabated despite half a century of prohibitions on international trade and domestic fishing. We investigate if farming can reduce poaching and support a healthy wild population by extending a flexible bioeconomic model of a three-stage illegal supply chain: poachers sell to traders (i.e., middlemen or cartels) who sell to end-markets. While we show under the monopolist a large stable wild population is maintained, this outcome is sensitive to cost parameters. Introducing farming decreases poaching by 29% or increases poaching by 6%, and results are robust to changes in cost parameters. Our results upend previous assertions that certain strategic responses will undermine conservation efforts and always result in population collapse. Furthermore, our quantitative framework can be adapted to evaluate conservation farming for other species and market structures.

Keywords :

1 Introduction

Illegal wildlife trade is a multi-billion dollar industry that drives biodiversity loss through unsustainable harvest ([T Sas-Rolfes et al., 2019](#)), spreads zoonotic disease ([Bell et al., 2004](#)), and threatens animal welfare([Baker et al., 2013](#)). The Convention on International Trade in Endangered Species of Wild Fauna and Flora (CITES) provides a regulatory framework that aims to ensure that international trade of wild animals and plants does not threaten their survival. Yet, for many species, regulatory interventions such as trade bans and controls have failed, and illegal trade in black markets continues to flourish ([Challender and MacMillan, 2014](#); [Challender et al., 2015a](#)). In such instances, supply-side interventions such as conservation farming can theoretically bolster conservation by “flooding the market” with farmed products, leading to reduced market prices and lower poaching incentives ([Gentry et al., 2019](#); [Phelps et al., 2014](#); [Tensen, 2016](#)). Supply-side interventions have occasionally succeeded at reducing poaching and recovering wild populations – e.g., vicuña and spotted cat ([IUCN, 2000](#); [Sahley et al., 2007](#)) – but they have also failed – e.g., green python, African elephant ([Lyons and Natusch, 2011](#); [Hsiang and Sekar, 2016](#)). Uncertainty around conservation outcomes from market-based approaches has led to continued reliance on trade bans and controls that are often ineffective at reducing poaching. Determining whether farming will succeed or fail requires a holistic understanding of a specific illegal wildlife market¹, including the interplay between market conditions and ecological criteria ([Challender et al., 2015b](#)). Studies have pointed to a common set of farming pitfalls. Species with slow individual growth rates and low fecundity are often unable to grow supply quickly enough to displace illegal products. Further, if poaching is very inexpensive, it is impossible for farming to undercut prices [6,8](#) – e.g., dried seahorses are ‘free’ to poach when retained as bycatch ([Lawson et al., 2017](#)). Demand-side concerns are focused on substitutability between farmed and wild products. Consumers of wildlife for medicinal or conspicuous purposes often prefer wild products for greater perceived potency or associated social status ([Dutton et al., 2011](#); [Gratwicke et al., 2008](#); [Fabinyi, 2012](#)). Here, we develop a quantitative framework that comprehensively considers all these pitfalls while accounting for detailed species-specific and market information.

Another critical factor in driving the success or failure of farming is market structure: illegal markets are often characterized by imperfect competition – where an individual trader or a small number of traders (i.e., middlemen, cartels, gangs, or other criminal organizations) dominate illegal trade and exert significant control over market prices. A bioeconomic model that predicts how imperfectly competitive markets will respond to competition from farming was developed almost

two decades ago ([Bulte and Damania, 2005](#); [Damania and Bulte, 2007](#)). Predicted strategic responses depend on how a trader chooses to compete with farming. If a trader responds by price setting (an aggressive response where the trader tries to undercut farmed prices and take market shares), then poaching pressure will increase and can lead to the collapse of the wild population. On the other hand, if traders respond by quantity adjustment (a mutually beneficial response where the trader competes on the amount of output produced, letting market prices adjust), poaching pressure is reduced and wild populations have the possibility to increase. This model has been widely used to both justify ([Biggs et al., 2013](#); [Abbott and van Kooten, 2011](#)) and discourage ([Tensen, 2016](#)) prospective farming initiatives. The authors of the original bioeconomic model concluded that farming is a perilous coin toss ([Bulte and Damania, 2005](#); [Damania and Bulte, 2007](#)). Here, we expand upon this model and reach a different conclusion: that farming can maintain large, stable wild population sizes that are robust to changes in cost structure under both types of competition. Furthermore, quantity adjustment yields substantial decreases in poaching and is the more likely response because prices and profits are higher than under price setting ([Singh and Vives, 1984](#)).

We explore the biological and economic performance of conservation farming for totoaba swim bladder in the context of illegal poaching and trade under different market conditions ([Froehlich et al., 2017](#)). Specifically, we examine the evolution of poaching and wild totoaba biomass, as well as prices and profits for different economic actors. The lifecycle for totoaba has been successfully closed in aquaculture, and the species is currently farmed in Mexico for domestic meat production. Totoaba is endemic to Mexico's Gulf of California and is threatened by a lucrative illegal international trade for its large swim bladder ([C4ADS, 2017](#); [env, 2019, 2016](#)). A single totoaba swim bladder can sell for up to \$80,000 USD per kilogram in Chinese end markets, where it is purchased for special occasions, gifting, and speculative investment ([ElephantActionLeague, 2018](#); [Sadovy de Mitcheson et al., 2019](#); [Martínez and Alonso, 2021](#)). For nearly half a century international trade for totoaba has been prohibited, and the legal totoaba commercial fishery has been closed. However, illegal fishing and trade continue and are controlled primarily by a single criminal organization (a cartel) that will likely respond strategically to farming ([Damania and Bulte, 2007](#); [Felbab Brown, 2022](#))

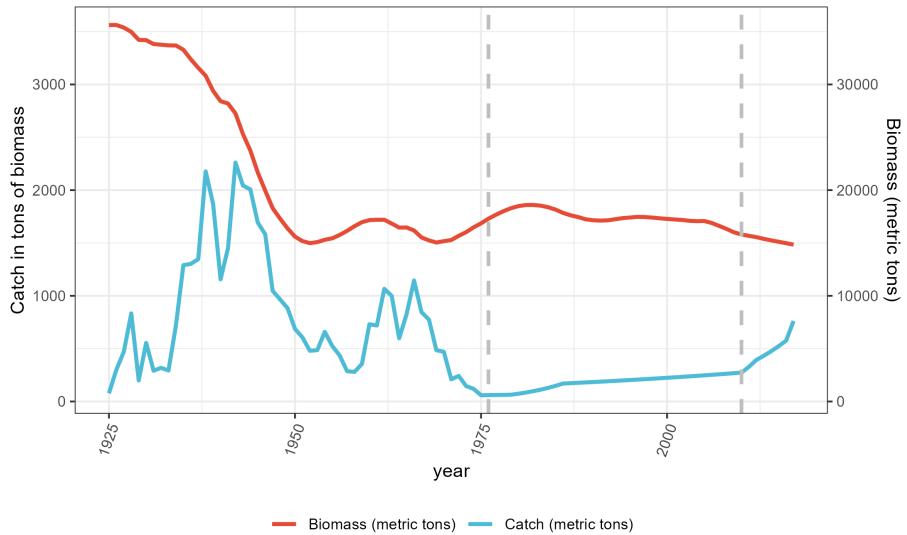


Figure 4.1: Evolution of totoaba population and catch over time

Dashed lines represent listing as CITES Appendix II species, and cartel takeover, respectively

There is an urgent need to reduce poaching for totoaba, as the vaquita (*Phocoena sinus*), a porpoise also endemic to the upper Gulf of California, is caught as bycatch in gillnets used to catch totoaba. The vaquita is on the brink of extinction as there are now fewer than fifteen individuals remaining (Rojas-Bracho et al., 2022). Furthermore, illegal trade has had negative social welfare consequences, as cartels are increasingly extorting Mexican fishing communities (Felbab Brown, 2022). Despite Mexico's attempts to stop totoaba poaching through various enforcement mechanisms, the country recently received wildlife trade sanctions for taking inadequate action (Rojas-Bracho and Reeves, 2013; CITES, 2023). Conservation farming presents a legal alternative to reduce illegal fishing by manipulating market structure.

We assemble and leverage a unique wealth of information on the totoaba stock, poaching sector, and farming sector to estimate the effects of market structure on poaching harvest and stock biomass. We focus on the market structure that best characterizes the totoaba trade – a vertical monopoly where a single monopolist trader controls the entire supply chain – and evaluate how this trader will respond strategically to competition from farming. We also show how to identify an effective policy space, where all supply, demand, and market structure parameters align to ensure that conservation farming will reduce poaching. Our results challenge long-standing model conclusions (Bulte and Damania, 2005; Damania and Bulte, 2007), thereby disrupting widely-held beliefs about the impacts of conservation farming. In particular, previous studies cautioned that when a trader responds to farming through price setting, the wild population always declines dramatically. In contrast, we find that for totoaba, price setting can maintain a stable and large population given that as the population size decreases, fishing

costs increase. To ensure low retail prices, traders must limit the price they pay to poachers and maintain a viable wild population.

2 Methods

We examine the effect of market structure and competition on poaching a population of wild animals using the logistic growth function (Figure 4.2). The poaching harvest function intersects with population growth producing stable and unstable equilibria. If poaching pressure is high relative to population growth (i.e, when demand is large, or poaching costs are low), a single stable equilibrium point is observed with a low wild abundance (an overharvested population). In the opposite scenario, where poaching pressure is low relative to population growth (i.e, when demand is small, or poaching costs are prohibitive), a single stable equilibrium point is observed with a high wild abundance (a healthy population). Between these extremes, two or three potential equilibria can emerge, with uncertain results that depend on the initial size of the population: a large initial population will result in a high abundance equilibrium point, and a small initial population will result in a low abundance equilibrium point.

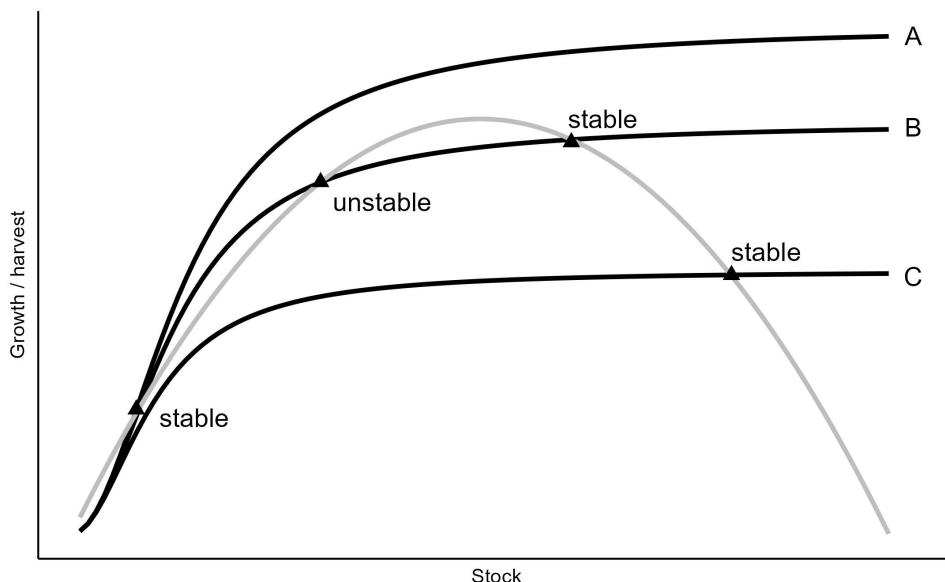


Figure 4.2: Schematic of equilibrium points under different poaching harvest functions

Logistic growth function (light gray) showing equilibria points resulting from three hypothetical poaching harvest functions (black). (A) a single low stable equilibrium point; (B) uncertain outcome, three interior equilibria two of which are stable and one unstable and separating. The long run equilibrium point will depend on the initial size of the population. A large initial population will result in a high abundance equilibrium point, and a small initial population will result in a low abundance equilibrium point; (C) a single high stable equilibrium point.

To assess expectations for totoaba, we first calculate equilibrium points for the stock in the absence of conservation farming under vertical monopolistic conditions (hereafter referred to as monopolistic conditions for ease) (Figure 4.3). A

single trader exists in a single location where he is the sole buyer, typical of endemic species such as totoaba (Wyatt et al., 2020; Martinez-Alvarado and Martinez, 2018). The trader sells poached harvest on an end market where prices and quantities can be manipulated.

Next, we add conservation farming to the monopolistic market structure, creating a duopolistic market (Figure 4.3). We calculate equilibrium points for the totoaba stock if a monopolistic trader responds to conservation farming either in a way that is (a) mutually beneficial by quantity adjustment or (b) aggressive by price setting. From a policy assessment perspective, any scenario where poached harvest produces a single high stable equilibrium point, and the monopolist cartel loses income, presents clear conservation and social welfare benefits.

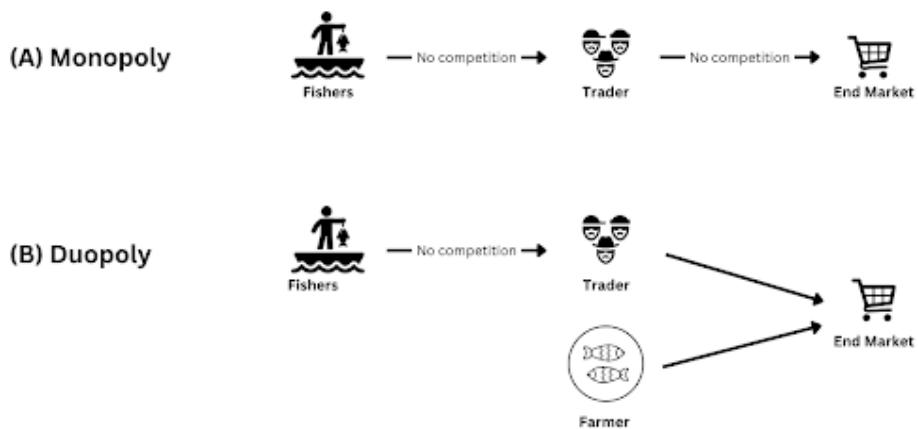


Figure 4.3: Schematic of monopoly and duopoly market structures

(A) monopolistic conditions, where fishers sell to a single trader where they are the sole buyer. This single trader sells poached harvest on an end market where they can manipulate prices and quantities. (B) Next, we add duopoly with farming: A monopolistic trader responds to conservation farming either in a way that is mutually beneficial by quantity adjustment or aggressive by price setting.

Here we briefly discuss our methods with an emphasis on the empirical application. Information on our theoretical conclusions from the bioeconomic model we revisited, lemmas and proofs can be found in the Appendix, section B. Table 4. J summarizes all the functions of the model.

1 The Poaching Model

The growth of the fish stock follows a logistic curve and the stock is poached following a Gordon-Schaefer production model. Totoaba population growth parameters were obtained from the 2017 stock assessment, where the carrying capacity (K) was 20,226 mt, and the stock biomass in 2017 was 14,844 mt (Cisneros-Mata, 2020). The intrinsic rate of population increase (r) was predicted using the *FishLife* package in R, which estimates growth parameters using totoaba-specific life

history data from *FishBase* (Thorson et al., 2017). The growth equation is :

$$g(x) = rx \left(1 - \frac{x}{K}\right) \quad (2.1)$$

using a predicted r of 0.20. We do not consider potential effects of hyperstability of the stock resulting from poaching on seasonal spawning aggregations (Erisman et al., 2011) or age structure.

Poachers optimally determine their effort to maximize their profit, with constant catchability, σ , and stock biomass, x , obtained from the 2017 stock assessment⁴⁶, and a linear quadratic cost of effort function, E . The poaching equation is $q = \sigma x E$ where $\sigma = 0.00002$.

Poachers are faced with a linear quadratic cost function $C(E) = W_1 E + W_2 E^2$. We calculated two poaching cost parameters W_1 (the linear coefficient of the cost function) and W_2 (the quadratic coefficient of the cost function) by (a) estimating total and average annual operating costs of the fishing fleet using semi-structured interviews conducted by the authors of this study; and (b) calibrating a linear quadratic cost function that matches historical data and predicts future cost evolution.

We conducted semi-structured interviews in the upper Gulf of California with two fishing cooperative leaders and four fishers in July and August 2018. These interviews informed annual poaching costs: food and fuel, labor, gear replacement, and bribes paid to fisheries officials. The fishery operates over six months with a variable number of active vessels, monthly fishing days, and sets per day (Cisneros-Mata, 2020). Poaching costs also include annual fleet-wide costs related to gear confiscations, vessel replacement, and fines, adapted and extrapolated from a summary of law enforcement actions provided by Mexico (noa, 2018). The cost per fishing trip was estimated to be \$5,051.26 during the low season (January and June), \$8,385.34 during the mid-season (February and May), and \$14,386.7 during the high season (March and April) (supplementary table 4. F). In our analysis we reconstructed a linear quadratic cost function with cumulative effort. We considered effort in each season cumulative with effort in less intense seasons. We used a low-season average cost for effort levels between 0 and low-season effort; for effort levels between low-season effort and cumulated low and mid-season efforts, we used a mid-season average cost.

We estimated the corresponding poaching cost parameters to match the observed average cost and modeled marginal costs at historical levels (resulting in cost parameters $W_1 = 12,200$ & $W_2 = 0.57$). Our low sample size precludes a robust statistical estimation of these cost parameters, e.g., of the historical cost function and of the evolution of costs if the fishery were to increase. To account for this uncertainty, we run a sensitivity analysis on two dimensions of costs.

First, we use different estimates for the average cost and reconstructed total costs, ranging from -10% to $+30\%$ of our high season average cost estimates. Second, we test weights for the linear and quadratic costs, ranging from a purely linear cost ($W_1 = 14,386,7$, $W_2 = 0$) to a purely quadratic cost ($W_1 = 0$; $W_2 = 3,74$).

The resulting poaching profit function is calculated as follows:

$$\Pi = p\sigma xE - W_1E - W_2E^2 \quad (2.2)$$

Traders operate on the end market, taking prices as given (competitive scenario) or determining prices (monopolistic scenario) to maximize profits. Traders face a linear demand function. We estimate a linear demand function by regressing price data on estimated catch from 2014 to 2017⁴⁶, yielding the equation $p(q) = \alpha - \beta q$ where the intercept, α , is \$1,625,837 USD and the slope coefficient, β , is \$1,563.75 USD (see supplementary table 4. G). Price data were obtained from available literature that provided estimated weight and value of totoaba maw seizures ^{24,26,50,51}. In addition to the literature review, valuable insights were obtained through personal communication with Wild Aid Investigators (pers. comm. Anonymous Wild Aid Investigators, 2018) as well as with local fishers and cooperative leaders in the upper Gulf of California, as previously described. The information shared by investigators and stakeholders was aggregated with the existing data from the literature. To ensure consistency and comparability, we standardized the weight measurements to grams and the currency values to US dollars. We assume that annual catch reaches the market during the same year, i.e, there is no stockpiling. As data are notoriously difficult to acquire for illegal trade, we pool observations and estimate a stationary demand function (supplementary table 4. G).

Traders buy totoaba from poachers at **price** s (USD/metric ton). The price paid to poachers balances demand from traders and supply to poachers. It decreases as the population increases, as fishing becomes less demanding. Traders also pay a unit transaction cost c (USD/metric ton), which we conservatively estimated to be zero. At a minimum this unit transaction cost includes transport (land and air travel), and payment to two or three ‘runners’ who carry up to ten swim bladders each (pers. comm. Anonymous Wild Aid Investigators, 2018). We know through anecdotal evidence that unit transaction costs c are likely large ([ElephantActionLeague, 2018](#)). However, due to scarce evidence, we used a value of $c = 0$ thus adopting a conservative strategy.

2 The Farming Model

We use a linear profit model for aquaculture and estimate a unit farming production cost parameter v (USD/metric ton) using annual operational costs (labor,

feed, vessel fuel, facility and administrative fees), as well as annual maintenance of pens (including cleaning) and vessels, using information provided by existing aquaculture facilities (supplementary table 4. H). Population growth rates differ in the wild and in captivity. Using captive growth rates obtained from personal communication with totoaba aquaculture producers, we consider harvestable size to be between 4.5 and 5 years old (an adult weight of 21.43 – 27.2 kg), associated with a swim bladder size between 417 – 529 g (supplementary figure 4.E). A minimum farmed harvestable size of 4.5 years closely corresponds to the mean swim bladder size (500 g) and estimated adult totoaba size (25.7 kg), as reported in surveys of individuals harvested in the wild ([Cisneros-Mata, 2020](#)). We considered this to be the size at which farmed totoaba would be competitive with the average wild-caught totoaba. We assume that aquaculture operates on a homogenous rotation ([Faustmann, 1849b](#); [Mitra and Wan, 1986](#)). The implications of this assumption are discussed in the appendix A.3. We compute the farming cost per metric ton as the capitalized sum of annual costs over 4.5 years at a 10% interest rate.

3 Demand

We use a linear demand function in the case of the vertical monopoly, estimated using price and catch data from 2014 to 2017 (see table 4. G), such that $p^w = \alpha^w - \beta^w q^w$. Upon the introduction of aquaculture, following [Singh and Vives \(1984\)](#) and [Damania and Bulte \(2007\)](#), we include a substitutability parameter γ , which measures the imperfect substitutability between farmed and wild products in the linear demand functions. When farmed products are introduced, the linear demand function is modified such that $p^i(q^i, q^k) = \alpha_i \beta_i q^i + \gamma q^k$ where q^i and q^f indicate the supply from the wild (w) and the farmed supply (f). This demand system emerges from a linear quadratic utility function in Supplementary Text (section 1.3.2). When demand intercept α_i s are equal, and own price effect $\beta_i = \beta_j = \gamma$ are equal, products are perfect substitutes. When demand intercepts are equal, but own price effects differ ($\beta_i \neq \beta_j$), then $\frac{\gamma^2}{\beta_i \beta_j}$ denotes the degree of product substitutability. At present, there has been no stated preference investigation for wild and farmed totoaba swim bladders in Chinese end-markets, although we know that the end-market economic value for fish maw is determined by taxon, size, and thickness of swim bladder. Investigative work in Mexico reports that it is challenging to distinguish between wild and farmed specimens ([ElephantActionLeague, 2018](#)). Therefore, we assume high substitutability (75% product substitutability) and check for smaller substitutability values in our sensitivity analysis (Figure 4.7) (see supplementary table 4. I) for a list of parameters).

3 Results

1 Totoaba stock under monopoly is sensitive to cost structure

We revisit and expand upon a bioeconomic model developed nearly two decades ago which differentiates between poachers and traders and develops a three-stage game (Bulte and Damania, 2005; Damania and Bulte, 2007). The totoaba is an endemic species that is illegally traded by a single trader, a monopolist, who dominates the market. This is the market structure that best characterizes the present consolidated totoaba trade (Felbab Brown, 2022). In this setting, a single monopolist trader restricts the supply of wildlife products to consumers, leading to increased prices and profits for the monopolist.

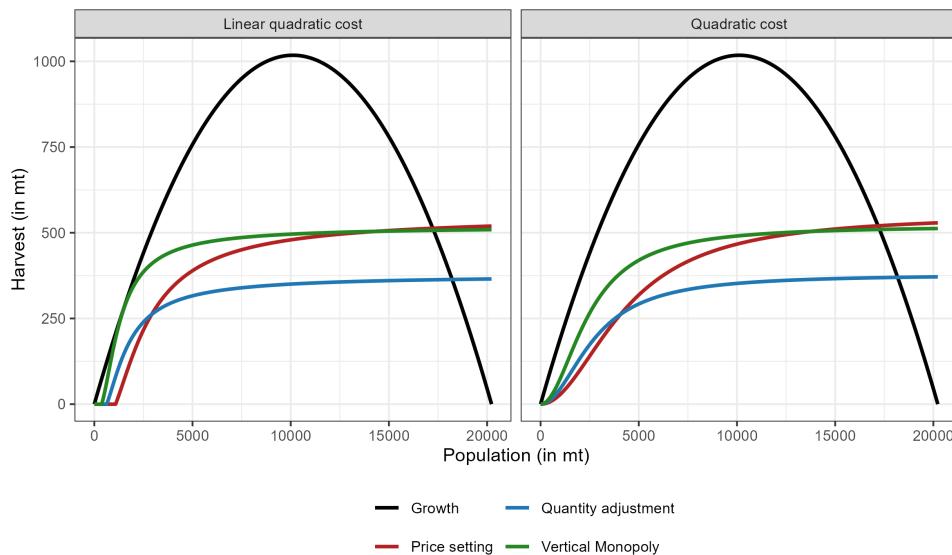


Figure 4.4: Equilibrium points for wild totoaba stock under different market structures with (left) a linear quadratic cost structure, and (right) a quadratic cost structure.

Logistic growth function (black) for *Totoaba macdonaldi* wild stock biomass with intersecting colored lines representing different market structures and competitive responses. Harvest under the status quo vertical monopoly is represented by the green curve. When conservation farming is added to the monopoly scenario the trader can respond either in a mutually beneficial way by adjusting the quantity supplied given a market price (quantity adjustment, in blue). Alternatively, the trader can respond aggressively and try to set a price that undercuts the price of farmed products, resulting in increased poaching (price setting, in red)

We initially calculate equilibrium points for totoaba assuming a quadratic cost structure, consistent with the original model, before calculating equilibrium points under a linear quadratic cost structure (Figure 4.4). Under the quadratic cost structure used in the original bioeconomic model, the totoaba wild stock biomass remains at a high steady-state equilibrium of 17,259 mt. However, we expand upon the quadratic cost structure, introducing a linear quadratic cost structure to account for energy costs associated with fishing. A linear quadratic cost structure more accurately represents new poachers being recruited to the

fishery as fishing opportunities increase (Péreau et al., 2012; Clark, 2007).

We find that under monopoly the linear quadratic cost structure is sensitive to cost parameter specifications, where relatively small changes in cost parameters can cause multiple steady states to emerge (Figure 4.5). If an increase in poaching comes at a small cost increase compared to historical average costs, the aggregate cost is close to linear (e.g. $W_2 = 0.47$) and below, compared to baseline $W_2 = 0.57$). In this case, a low steady-state equilibrium of 1,106 mt, an unstable intermediate equilibrium arises at 1,842 mt and a high stable steady-state equilibrium of 17,277 mt in the vertical monopoly. Our model uses the best available information on the totoaba fishery, but uncertainty surrounding the projected evolution of fishery-wide poaching costs warrants a cautious assessment of monopoly performances: while it could maintain a healthy population, it can also lead to stock collapse.

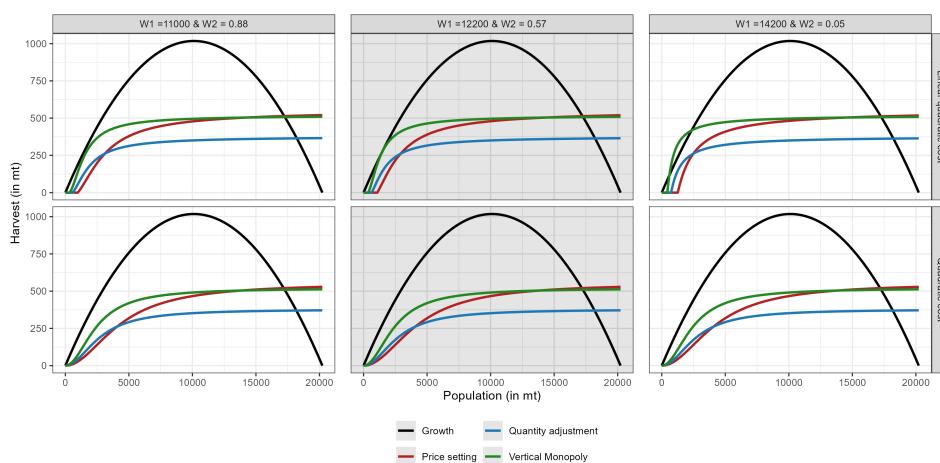


Figure 4.5: Sensitivity of equilibrium points to cost structure for wild totoaba stock

Logistic growth function (black) for *Totoaba macdonaldi* wild stock biomass with intersecting colored lines representing different market structures and competitive responses. Harvest under the status quo vertical monopoly is represented by the green curve. When conservation farming is added to the monopoly scenario the trader can respond either in a mutually beneficial way by adjusting the quantity supplied given a market price (quantity adjustment, in blue). Alternatively, the trader can respond aggressively and try to set a price that undercuts the price of farmed products, resulting in increased poaching (price setting, in red). Cost parameters W_1 and W_2 correspond to the linear quadratic cost structure. In the top panel, equilibria are displayed for the linear quadratic cost, on the bottom, for a quadratic cost. On the left panel, the quadratic component is large, and vertical monopoly maintains a healthy stock. Center panel highlights the baseline scenario. In the right panel, the cost structure is close to linear. In this case, the vertical monopoly may lead to drastic stock decline.

2 Farming produces conservation benefits

While our results show that totoaba stock may remain healthy under the current monopolistic market conditions, these results are sensitive to changes in poaching costs (Figure 4.5). Therefore, we ask if conservation farming can improve upon

the status quo by producing a robust single high stable equilibrium point and reduced cartel profits.

We add conservation farming to the monopolist model and now have two ‘firms’ – a trader and a farmer – competing on a duopolistic market. When farming supplies legal product to end-market consumers, the demand for illegal product will fall, assuming that wild and farmed products are substitutable (an assumption we explore later). The monopolist trader can respond to competition in two ways: a mutually beneficial way by adjusting the quantity supplied given a market price (quantity adjustment), or alternatively, an aggressive way that tries to select a price that undercuts the price of farmed products (price setting). In both scenarios, the trader and farmer choose a quantity supplied simultaneously, without knowing how the other will respond.

Illegal markets are almost always characterized as competing through quantity adjustment (Poret, 2009; Flores, 2016). Under the assumption that products are substitutable, it is more profitable – and therefore more likely – for both firms to compete through quantity adjustment (Singh and Vives, 1984). When goods are substitutes, if both firms restrict the quantities supplied, they both enjoy higher prices. If they flood the market, prices and profits collapse. In the case of totoaba, we find that if traders respond through quantity adjustment under the linear quadratic cost structure, then the wild stock biomass increases by 5.45% (compared to a monopoly) to a steady state equilibrium of 18,220 mt, or to 90% of carrying capacity (Table 4.6). This represents a reduction in poaching harvest of 28.27% and \$195.16 million USD of annual lost profit to the trader.

Even if traders respond aggressively through price setting, considered a less likely response (Singh and Vives, 1984) a single high equilibrium emerges (Figure 4.4). Price setting is considered a much less likely response to competition because the trader would face steep profit losses. Under the high steady-state equilibrium with the linear quadratic cost structure, wild stock biomass decreases by 0.24% relative to monopoly, to a steady-state equilibrium of 17,235 mt, or to 85% of carrying capacity (Table 4.6). Although the high steady-state reflects a relatively small increase in poaching harvest by 5.85%, it would result in \$313.84 million USD of annual lost profit to the cartel, making this strategy unlikely.

Our current specifications for totoaba show that price setting leads to a slight increase in poaching pressure, however, we argue that price setting does not universally lead to increased poaching pressure, challenging a key conclusion from the original bioeconomic model (Bulte and Damania, 2005; Damania and Bulte, 2007). Farming puts an upper bound on the price traders can pay to poachers in order to remain competitive. When the cost of farming becomes lower than the combined cost of poaching and trading, price-setting competition does not inevitably result in the overexploitation of the wild stock. This is because

Scenario	Poached harvest (in mt)	Farmed harvest (in mt)	Steady state population (in mt)	Illegal profit (in million USD)	Farming profit (in million USD)	Fishing profit (in million USD)	Aggregate profit (in million USD)	Illegal profit change (in million USD)	Variation in ss. pop.	Poaching change (%)
Vertical Monopoly	507.04	0.00	17277.0	402.02	0.00	1.22	403.24	0.00	0%	0%
Quantity adjustment	363.71	333.60	18220.5	206.87	174.03	0.57	381.46	-195.16	5.46%	-28.27%
Price setting	536.70	430.05	17235.0	88.18	58.74	3.57	150.49	-313.84	-0.24%	5.85%

Bioeconomic performance

Figure 4.6: Economic and ecological performance of different market regimes

when farming costs are low, traders have an incentive to maintain large stocks by poaching less to remain competitive with farmers. This limits the price paid to poachers. On the other hand, when farming costs are large, traders have an incentive to poach more, paying a larger price to poachers while remaining competitive with the farming sector. In the case of totoaba, species specific traits and market characteristics result in a slight increase in poaching in the price setting scenario. However, if the carrying capacity were smaller, or demand lower, the price-setting equilibrium would result in conservation benefits.

While we focus on the effect of conservation farming on a monopolistic market structure, given that this scenario best represents the totoaba fishery today, the effect of conservation farming on market structures can be explored in different contexts. We model alternative market structures, including scenarios with multiple competing traders or multiple competing farmers, and find that if the number of farmers exceeds the number of traders, poaching levels will decline (supplementary figure 4.A). Additionally, if farming is taken over by monopolists, we find that poaching is reduced and the wild population increases (supplementary figure 4.B).

3 An effective policy space for farming.

Our analysis provides a quantitative framework that can identify an effective policy space where all supply, demand, and market structure parameters align to ensure that conservation farming will reduce poaching, improving greatly on the original bioeconomic model and the limitations of binary qualitative approaches (Phelps et al., 2014; Tensen, 2016; Bulte and Damania, 2005; Damania and Bulte, 2007; Challender et al., 2019). This bioeconomic model allows researchers to quantify: (a) how much cheaper farming must be relative to poaching to be competitive; (b) how much of a demand increase can be absorbed by farming; and (c) how substitutable must wildlife products be for farmed products to displace wild products. Critically, we also explore how the interaction between these fac-

tors may affect outcomes. We explore how sensitive the results are for totoaba, offering general and totoaba-specific policy solutions to help ensure that conservation farming remains in the effective policy space.

We find that the cost of conservation farming for totoaba can be high and still competitive with poaching, but this is contingent on the cost for traders also being high (supplementary figure 4.D). Traders inherently rely on poachers to obtain totoaba, and if farming is expensive this forces traders to pay poachers higher prices. If traders compete with poachers under the more likely quantity adjustment response, the population remains healthy, even increasing by nearly 6% from the monopoly steady state. However, if traders compete with farmers by price setting, the low prices paid to poachers can incentivize poachers to increase fishing pressure in order to maintain payouts. This can lead to a decrease in the wild population biomass modestly by 0.24% from the monopoly steady state. Policymakers can support farming success by subsidizing farming to keep the cost low while maintaining enforcement to keep the cost of poaching high (for totoaba this includes marine patrols, fisheries closures, and gillnet bans). To mitigate the possibility of stock decline under the less-likely price-setting response, we identify that maintaining conservation farming below \$77,339 USD per mt of totoaba (amounting to a 14% subsidy on unit production cost) will prevent any increase in poaching pressure under either competitive response, assuming no effect of law enforcement in our baseline model.

Our results confirm that high substitutability is critical to conservation farming success and leads to larger conservation benefits in the quantity-setting equilibrium, under the assumption that demand remains stable (Figure 4.7). Fish swim bladders have a wide variety of uses and values, and it is possible that farmed totoaba swim bladders may enter into these different product streams ([Sadovy de Mitcheson et al., 2019](#)). In the case of no substitutability, two separate, non-competitive markets emerge. In this scenario the status quo is maintained, both firms set high prices, and traders continue to operate as a monopoly because farmed product does not compete with wild product. At the other extreme, in the case of perfect substitutability, consumers prefer the cheaper option without any preference of source. This increases the intensity of potential price-setting competition between firms and further depletes the stock in this case. To comply with CITES captive breeding guidelines totoaba must be identified as farmed ([CITES, 2019](#)), and distinguishing between products to meet regulatory obligations can artificially lower substitutability. Outcomes vary under intermediate states of substitutability. For low to medium substitutability (i.e., 10 – 50%) traders and farmers are still likely to limit quantity: undercutting a competitor would yield significant profit losses. For high substitutability (i.e., 90%) there is an incentive to compete for market control either by price setting or quantity adjustment, which

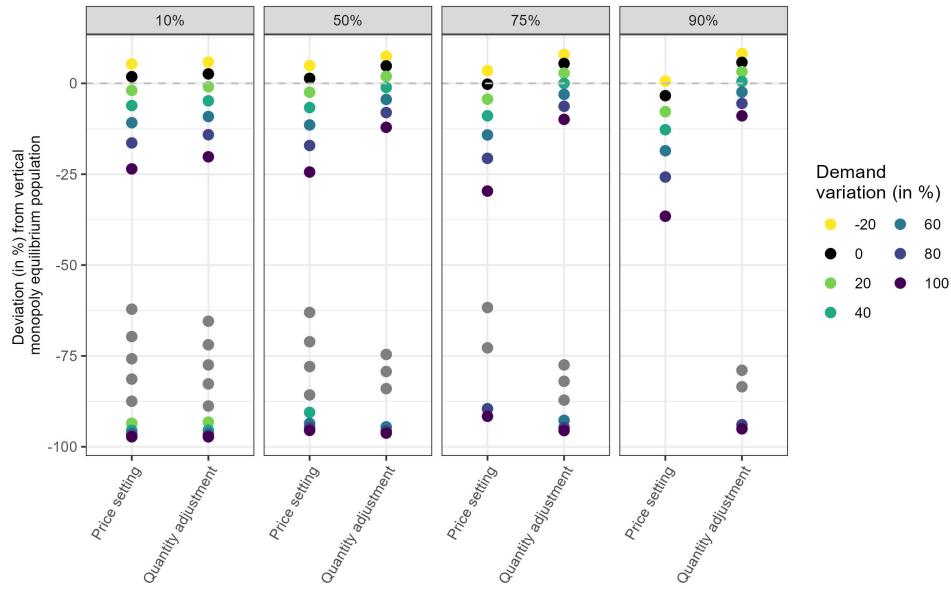


Figure 4.7: Interaction between substitutability and demand under duopolistic competition

Each panel represents a different substitutability between farmed and wild product: large (90% substitutability), baseline (75%), medium (50% substitutability), and low (10% substitutability). Our baseline results are in the 75% substitutability case, with zero demand variation (black dots). When conservation farming is added to the monopoly scenario the trader can respond aggressively and try to set a price that undercuts the price of farmed products (price setting), alternatively the trader can respond in a mutually beneficial way by adjusting the quantity supplied given a market price (quantity adjustment). We simulate a change in end-market demand ranging from a reduction in demand by 20% to an increase in demand up to 100%, in increments of 20%. One, two, or three potential equilibria can emerge. Where three equilibrium points emerge, we color only the high and low stable equilibria (unstable equilibria are indicated in gray). The dotted horizontal lines indicate the status quo monopoly equilibrium population (in the absence of conservation farming). Points closer to 0 represent a high stable equilibrium point, whereas points closer to -100 represent a population collapse stable equilibrium point.

reflects our main results.

The value of totoaba swim bladder is tied to rarity, and while demand evolution is an open empirical question, we test the sensitivity of our results to simultaneous changes in demand and substitutability (Figure 4.7). Totoaba swim bladder purchases are ‘conspicuous consumption,’ luxury products commonly purchased for social status and speculative investing by wealthy consumers ([Sadovy de Mitcheson et al., 2019](#); [Veblen, 2023](#)). A decrease in swim bladder price resulting from conservation farming may actually undermine the desirability of totoaba swim bladders in Chinese end markets, given that the high monetary value is linked to high social status ([Jenkins, 2016](#)). However, some increase in demand may be expected if a legal product becomes available, as law-abiding consumers will be more likely to purchase wildlife products when those products are traded and purchased legally ([Phelps et al., 2014](#)). Under our high substitutability as-

sumption (75%), competition through quantity adjustment can withstand a 40% increase in demand, whereas competition through price setting is not robust to demand increases. For price setting, a demand increase of 40% would cause the equilibrium population to decrease by 10% from the monopoly status quo, increasing poaching by 216 mt.

There is a much higher threat to the wild population if demand increases under low to medium substitutability (i.e. 10 – 50%), given that this additional demand cannot be fully met by farmed product (Figure 4.7). In the best-case and most likely scenario, medium substitutability (50%) can meet a 20% increase in demand if competition occurs through quantity adjustment, although uncertain outcomes (e.g. high and low steady states) start to emerge if demand increases by 60% or more. In the worst-case scenario, if competition occurs through price setting and products have medium substitutability (50%), any increase in demand reduces the wild population from the status quo. While increases in demand of 20 – 40% still produce a single high equilibrium point, the population size is lower than under monopoly. Furthermore, if demand increases beyond 80%, uncertain outcomes emerge, with the wild population either stabilizing at a high equilibrium point (14,322 mt in the price setting scenario; 15,886 mt in the quantity adjustment scenario) or being pushed to a low equilibrium point (ranging from 763 mt in the quantity adjustment scenario; 909 mt in the price setting scenario). We recommend that stated preference investigations on wild versus farmed product should be undertaken in Chinese end-markets and that these investigations include questions focused on perceived social status benefit and legality ([Hinsley and 't Sas-Rolfes, 2020](#)).

4 Conclusion

Our results show that conservation farming presents a potentially high reward intervention. If traders respond to competition from farming by quantity adjustment, the wild totoaba stock is predicted to increase by 5.45% relative to the status quo monopoly, to a high stable biomass of 18,220 mt (90% of carrying capacity). In addition to improving the totoaba wild stock, this quantity adjustment response will decrease poaching by 28.27% relative to the status quo. If traders respond by price setting, the wild stock biomass decreases by less than 1% to a high stable biomass of 17,235 mt (85% of carrying capacity). Economic theory concludes that quantity adjustment is the more likely outcome because restricting quantities allows both farmers and traders to collect higher profits ([Singh and Vives, 1984](#)). Conservation farming presents a more robust outcome to the status quo monopoly market structure (where a single trader dominates the market), as the wild totoaba reaches either a low or high stable equilibrium biomass

depending on the poaching cost structure. We find that if products have high substitutability they are more likely to maintain a high stable equilibrium. Further, under a quantity adjustment response, highly substitutable products can better maintain this high stable equilibrium for demand increases up to 40%. Our results are sensitive to changes in substitutability and increases in demand, therefore we encourage a thorough understanding of end-market demand before implementing conservation farming for totoaba.

We revive an existing bioeconomic model and reach different and optimistic conclusions about the potential for conservation farming to reduce poaching and maintain a healthy wild population. We provide a novel framework to objectively assess the potential effects of farming by grounding our analysis in detailed species ecology and market data. Furthermore, our approach provides a rigorous alternative to existing qualitative frameworks that are unable to analyze the interaction between multiple variables. While our analysis focuses on totoaba, the bioeconomic model is flexible and can be applied more broadly to other species and contexts to examine the effect of conservation farming on a wild population.

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Contributions and data availability

J.M.L and S.J contributed to this work equally. J.M.L., S.J., M.C-R., G.M.G., M.A.C-M., E.A-B, and S.D.G. contributed to writing the manuscript. J.M.L., S.J., A.S., and S.D.G. contributed to study conception and design. All authors contributed to data acquisition and analysis. All authors approve of the submitted manuscript.

The data that support the findings of this study are available [here](#)

The code used for this study is publicly available on [Github](#) and [archived here](#)

A Appendix

A A theoretical model of poachers, traders, and farmers

Our framework follows [Damania and Bulte \(2007\)](#), with a poaching cost structure adapted to fisheries. The model develops a three-stage dynamic, game theoretic, bioeconomic model. The value chain for poached animal products comprises poachers, middlemen traders, and end markets. As a small number of actors characterizes many wildlife markets, the model features a vertical monopoly and looks at the consequences on wildlife population stocks of the introduction of a farmed substitute. In this setting, farmers compete on end markets with traders in quantity and price. In the original model, price competition unambiguously results in larger harvests than in the vertical monopoly case. Therefore, while quantity competition reduces poaching, the threat of a population collapse in the price-setting case should warrant a cautious approach to conservation farming. We argue that this conclusion is erroneous, as the intricacies of imperfect substitutability and market dynamics have not been properly accounted for in the original model. As a matter of fact, standard economic intuition regarding price-setting competition in the homogeneous goods case does not directly apply here, as fishing costs rise as the stock decreases, limiting the ability of the trader to flood the market. We show that scenarios exist where any type of competition unambiguously leads to positive conservation outcomes, i.e, reduced poaching and larger steady-state stocks. We amend the original results and use this model for simulation.

First, poachers illegally harvest wildlife resources. Second, they sell their catch to a monopsonistic buyer. Third, the buyer sells catches on a monopolistic market, which is not accessible to poachers. We label this value chain 'vertical monopoly' as a reference case. We then look at the impact of introducing a competitor on the end market, the farming sector.

A.1 Entry in the fishery and poaching supply

We denote the fishing effort by E , which is measured in the number of vessel trips. Entry in the poaching sector, \dot{E} , is a function of payoff and an adjustment parameter. Harvest, q , follows the Gordon-Schaefer dynamic biomass model $q = \sigma x E$, with σ the (stock-independent) catchability coefficient, and E , effort. The payoff is determined by the price paid to the poachers s minus the cost of effort. We adopt a disaggregated view of the fishery, and consider increasing marginal costs of effort, as individuals have to be attracted from other activities with increasing opportunity costs. To account for energy costs, we derive a modified version of this model using a linear-quadratic cost function (see [37, 53]). Entry happens as

long as the profit of the marginal poacher is positive :

$$\dot{E} = \eta \frac{d\Pi}{dE} = \eta \frac{d}{dE} [sq - W_1 * E - W_2 E^2] \quad (\text{A.1})$$

The resource stock biomass x follows a logistic growth curve and is harvested. Overall, the dynamics are:

$$\dot{x} = g(x) - q = rx \left(1 - \frac{x}{K}\right) - \sigma x E \quad (\text{A.2})$$

Where r is the intrinsic population growth rate, and K is the carrying capacity.

Fishermen enter the fishery as long as the marginal profit from selling to traders along the vertical value chain is positive. As the resource is in open access from the fishermen poachers maximize their instantaneous profit with respect to effort. The optimal effort and aggregate supply of poached fish is:

$$\frac{d\Pi}{dE} = 0 \Rightarrow E^* = \max \left(0, \frac{s\sigma x - W_1}{2W_2}\right) \quad (\text{A.3})$$

$$\Rightarrow q^* = \max \left(0, \frac{s\sigma^2 x^2 - W_1 \sigma x}{2W_2}\right) \quad (\text{A.4})$$

Given the linear quadratic nature of the costs, there is no effort or catch for low stock levels and/or low prices. Effort and catch increase with the price paid to poachers, s .

A.2 Traders as vertical monopolists, without farming

We introduce a trader who has market power on the end-market (monopoly) and on the primary market, making it a “vertical monopoly”. The trader has to set price s on the primary market to clear the poaching market. On the end market, we assume the trader faces a linear inverse demand :

$$P^m = \alpha^m - \beta^m q^W \quad (\text{A.5})$$

Trading an illegal commodity incurs transaction costs c . Hence, the monopoly profit can be written as :

$$\Pi^m = (\alpha^m - \beta^m q^W - c - s)q^W \quad (\text{A.6})$$

The optimal level of output is :

$$q_m^W = \frac{\alpha^m - c - s}{2\beta^m} \quad (\text{A.7})$$

Using the poachers' supply, it must be that in equilibrium, the supply of the

monopolist trader equals the supply of the poachers. The price paid to poachers s balances supply and demand (consistent with equation 13 in [Damania and Bulte \(2007\)](#)). Substituting s^* into equation A.7 yields the quantities of poached product in the vertical monopoly scenario :

$$\text{Price paid to poachers : } s_m^*(x) = \frac{W_2(\alpha_m - c) + \beta^m(W_1\sigma x)}{\sigma^2 x^2 \beta^m + W_2} \quad (\text{A.8})$$

$$\text{Poaching : } q_m^*(x) = \frac{\sigma^2 x^2 (\alpha_m - c) - W_1\sigma x}{2(\sigma^2 x^2 \beta^m + W_2)} \quad (\text{A.9})$$

First, note that equation A.9 is consistent with equation 14 in [Damania and Bulte \(2007\)](#), as the limiting case where $W_1 = 0$ and $W_2 = W$.

A.3 Captive breeding, imperfect competition and conservation

In this part of the model, a farmer can grow and sell totoaba. The theoretical model focuses on the duopolistic competition between the two actors on the end market for totoaba. As products are strategic substitutes, it is natural to investigate the case where Cournot competition arises. Indeed, when products are substitutes, each firm tries to maximize its residual demand (25). Nonetheless, given the asymmetric nature of costs, we also investigate Bertrand competition, as [Damania and Bulte \(2007\)](#).

Introducing aquaculture : the aquaculture farm needs to determine the optimal harvest age, based on the intrinsic growth rate in the pen, and expected prices. A sizeable literature has shown that rotation time is invariant to market structure in forestry applications ([Faustmann, 1849b](#); [Mitra and Wan, 1986](#)) although quantities can be modified. The optimal rotation literature confirms the existence of a Faustmann rotation, where a set of T^* pens are equally distributed among each age class (1 pen per age class until T^*). While it is arguably unrealistic to expect this structure for an inherited forest, it is reasonable to assume that a farm would *ex-ante* determine this rotation period given the expected price schedule over time. We assume that the aquaculture farm aims at producing a product that is as similar as possible from a biophysical stand-point and thus determines T^* . As we consider a stationary demand function, one can write the farming problem as a linear profit maximization problem, where the unit cost of production equals the capitalized sum of annual average variable costs over T^* periods. Therefore, we assume that an aquaculture firm can raise totoaba at cost v and sell it to the market:

$$\Pi^F = (P^F - v)q^F \quad (\text{A.10})$$

With v the unit cost per ton of totoaba, corresponding to the capitalized sum of annual costs.

Utility maximization and demand functions : upon the introduction of farmed goods, the inverse demand functions change. We use a model consistent with (Singh and Vives, 1984), where a representative consumer maximizes a quadratic and strictly concave utility function subject to prices:

$$\max_{q^W, q^F} V = \alpha^W q^W + \alpha^F q^F - \left(\frac{\beta^W(q^W)^2 + 2\gamma q^W q^F + \beta^F(q^F)^2}{2} \right) - p^W q^W - p^F q^F \quad (\text{A.11})$$

Two inverse demand functions emerge, that the traders and farmers face :

$$P^W = \alpha^W - \beta^W q^W - \gamma q^F \quad (\text{A.12})$$

$$P^F = \alpha^F - \beta^F q^F - \gamma q^W \quad (\text{A.13})$$

Where W, F refers to wild and farmed. We assume $\gamma > 0$ e.g that goods are substitutes. When $\alpha_W = \alpha^F$ and $\beta^W = \beta^F = \gamma$, the goods are perfect substitutes. When $\alpha^F = \alpha^W$, but $\beta^F \neq \gamma$ or $\beta^W \neq \gamma$, $\frac{\gamma^2}{\beta^W \beta^F}$ measures the degree of product differentiation.

Rearrange the initial inverse demand functions into direct demand functions:

$$q^W = a^W - b^W P^W + e P^F \quad (\text{A.14})$$

$$q^F = a^F - b^F P^F + e P^W \quad (\text{A.15})$$

With $a^i = \frac{\alpha^i \beta^j - \alpha^j \gamma}{\beta^i \beta^j - \gamma^2}$, $b^i = \frac{\beta^j}{\beta^i \beta^j - \gamma^2}$ and $e = \frac{\gamma}{\beta^i \beta^j - \gamma^2}$

Cournot competition in the retail market: assume that the two firms compete by setting their quantities. We solve the multi-stage game using backward induction. First, we derive the supply function resulting from Cournot competition. Second, we find the price paid to poachers so that the quantities supplied by the traders on the end market equal the quantities supplied by poachers.

Taking the inverse demand functions and plugging them into the profit functions:

$$\begin{aligned} \Pi^F &= (\alpha^F - \beta^F q^F - \gamma q^W - v) q^F \\ \Pi^W &= (\alpha^W - \beta^W q^W - \gamma q^F - s - c) q^W \end{aligned}$$

In a Cournot equilibrium, each firm takes its competitor's quantity as given, and picks optimal reaction functions.

Solving for the Nash equilibrium using reaction functions, each firm supplies:

$$q_c^W = \frac{2\beta^F(\alpha^W - (s + c)) - \gamma(\alpha^W - v)}{4\beta^W\beta^F - \gamma^2} \quad (\text{A.16})$$

$$\tilde{q}_c^F = \frac{2\beta^W(\alpha^F - v) - \gamma(\alpha^W - s - c)}{4\beta^W\beta^F - \gamma^2} \quad (\text{A.17})$$

Now, we find the equilibrium price paid to poachers for each unit of totoaba $s_C^*(x)$ by equating \tilde{q}_c^W and q^W , and find the Nash equilibrium supply functions.

In the Cournot equilibrium:

$$\text{Price paid to poachers: } s_C^*(x) = \frac{2W_2(2\beta^F(\alpha^W - c) - \gamma(\alpha^F - v)) + W_1\sigma x(4\beta^F\beta^W - \gamma^2)}{4W_2\beta^F + \sigma^2x^2(4\beta^F\beta^W - \gamma^2)} \quad (\text{A.18})$$

$$\text{Poaching: } q_C^{W*}(x) = \frac{\sigma^2x^2(2\beta^F(\alpha^W - c) - \gamma(\alpha^F - v)) - 2\beta^FW_1\sigma x}{4W_2\beta^F + \sigma^2x^2(4\beta^W\beta^F - \gamma^2)} \quad (\text{A.19})$$

First, including a linear component for energy in the poaching cost significantly raises the price paid to poachers (when $W_1 > 0$). Second, poaching decreases with the degree of substitutability between farmed and wild products (γ), and increases with the production cost of farmed products v . On the other hand, it increases with demand for the wild product α^W . For low stock values, poaching can be null since the production costs increase as stocks diminish. In the polar quadratic cost case (e.g. $W_1 = 0$), our results differ from [Damania and Bulte \(2007\)](#) by a magnitude effect. Nonetheless, the results stand :

Lemma 1: *Assume the market is large, i.e., the residual demand for large stock levels is large enough. For any given wildlife stock, poaching levels in equilibrium with captive breeding will be lower than those without captive breeding, if the introduction of captive-bred animal products has no impact on the parameters of the original inverse demand function for wild animal products.*

See Appendix B.1. for proof of Lemma 1

Bertrand competition in the retail market

Interior solution: the two firms compete by setting their prices. This section investigates a potential interior equilibrium, where both producers operate on the market.

Using demand functions instead of inverse demand functions:

$$q^F = a^F - b^F P^F + e P^W$$

$$q^W = a^W - b^W P^W + e P^F$$

With $a^i = \frac{\alpha^i \beta^j - \alpha^j \gamma}{\beta^i \beta^j - \gamma^2}$, $b^i = \frac{\beta^j}{\beta^i \beta^j - \gamma^2}$ and $e = \frac{\gamma}{\beta^i \beta^j - \gamma^2}$

Firms set their prices. The Bertrand profit equations are :

$$\Pi^F = (P^F - v) q^F = (P^F - v)(a^F - b^F P^F + e P^W)$$

$$\Pi^W = (P^W - (s + c)) q^W = (P^W - (s + c))(a^W - b^W P^W + e P^F)$$

Solving for the reaction functions :

$$r^F(P^W) = \frac{a^F + b^F v + e P^W}{2b^F} \quad (\text{A.20})$$

$$r^W(P^F) = \frac{a^W + b^W(s + c) + e P^F}{2b^W} \quad (\text{A.21})$$

Finding the interior solution for the Nash Equilibrium :

$$P_B^F = \frac{2b^W(a^F + vb^F) + e(a^W + b^W(s + c))}{4b^F b^W - e^2}$$

$$P_B^W = \frac{2b^F(a^W + b^W(s + c)) + e(a^F + vb^F)}{4b^F b^W - e^2}$$

The equilibrium price paid to poachers is determined by equating the quantity supplied by the trader in Bertrand duopoly and the quantity supplied by the poachers and yields the quantity supplied yields :

In the **Bertrand equilibrium** :

$$\text{Price paid to poachers } s_B^*(x) = \frac{2W_2 b^W [b^F(2a^W + ev) + ea^F + c(e^2 - 2b^W b^F)] + W_1 \sigma x (4b^F b^W - e^2)}{\sigma^2 x^2 (4b^F b^W - e^2) + 2W_2 b^W (2b^F b^W - e^2)} \quad (\text{A.22})$$

$$\text{Poaching : } q_B^{W*}(x) = \frac{b^W [\sigma^2 x^2 (b^F(2a^W + ev) + ea^F + c(e^2 - 2b^W b^F)) - W_1 \sigma x (2b^F b^W - e^2)]}{2W b^W (2b^W b^F - e^2) + (4b^F b^W - e^2) \sigma^2 x^2} \quad (\text{A.23})$$

We amend the original results from [Damania and Bulte \(2007\)](#) with the concurring Lemma 2:

Lemma 2: *With Bertrand competition, if the introduction of captive-bred products has no impact on the parameters of the demand function for wild*

animal products, poaching levels with captive breeding are ambiguous. The driver of the equilibrium is the cost ratio between aquaculture and the illegal poaching sector, i.e, v and $c + s(x)$

- *For relatively low ratio values (i.e, $c + s(x) \gg v$), poaching is unambiguously lower than without captive breeding for any given wildlife stock*
- *For intermediate ratio values, poaching is larger (for $x < \tilde{x}$), then lower (for $x > \tilde{x}$), than without captive breeding (with \tilde{x} such that $q_B^{W*} = q_m^W$)*
- *For large values of unit farming costs, poaching is unambiguously larger than without captive breeding for any wildlife stocks*

See appendix B.2 for proof of Lemma 2.

Our results significantly differ from [Damania and Bulte \(2007\)](#), as Bertrand competition does not unambiguously lead to more extraction. Indeed, poaching functions are ambiguously ranked, and the final location of the steady state depends on the species intrinsic growth rate r and carrying capacity K .

With low farming costs, traders have an incentive to maintain large stocks. As the price paid to poachers is inversely related to the size of the stock, low harvest maintains large stocks and thus limits the price paid to poachers. Given its operational costs, it is the only way for the trader to remain competitive with the farming sector. On the other hand, when farming costs are large, the traders are incentivized to harvest more, as they can afford to pay a larger price to poachers while remaining competitive with the farming sector.

Corner solution: in a perfectly substitutable framework, a corner solution emerges if one firm has a lower marginal cost than the other: if farmed and wild animal products were perfect substitutes and farmed products unambiguously cheaper to produce, poaching would cease. In the context of imperfectly substitutable goods, this result is challenged. For poaching to cease, it must be that :

$$v = -\frac{1}{e}(2(a^W - cb^W) - \frac{1}{b^F}(ce + a^F)) \quad (\text{A.24})$$

In our setup, the marginal cost of production for farming would need to be **negative** for poaching to stop¹. Moreover, as substitutability increases, this cost lowers. The relative cost of trading poached goods plays a minor role.

¹If consumers enjoy a numeraire good, they must receive compensation to consume the farmed good such that they increase their numeraire consumption to make up for the imperfectly substitutable nature of the farmed good.

A.4 Steady state equilibria

Given the inverted U-shape of the logistic growth function, several steady-state equilibria can arise. First, if the *harvest function* (that is increasing and concave) is *steeper* than the growth function at low stock levels, there can be (i) no equilibrium if the harvest at $K/2$ is larger than the growth rate, (ii) one bifurcation point (tangent harvest and growth functions at $K/2$, and (iii) two equilibria, with one stable and one unstable. If the *growth function is steeper* than the growth function at low stock levels, there can be (i) a single equilibrium, (ii) a bifurcation point and an equilibrium, (iii) three interior equilibrium, with only two being stable (see figure 4.2 for an illustration)

A.5 Extensions

An oligopoly model: we extend our model to gauge the impact of the number of traders and farmers. We denote by \mathcal{I} the set of individual traders $i \in \mathcal{I}$ and by \mathcal{J} the set of individual farmers $j \in \mathcal{J}$. The demand functions are :

$$P_k^W = \alpha^W - \beta^W \sum_{i \in \mathcal{I}} q_i^W - \gamma \sum_{j \in \mathcal{J}} q_j^F \quad (\text{A.25})$$

$$P_l^F = \alpha^F - \beta^F \sum_{j \in \mathcal{J}} q_j^F - \gamma \sum_{i \in \mathcal{I}} q_i^W \quad (\text{A.26})$$

Cournot oligopoly : each farmer and trader maximizes profits by taking as given its competitors' quantity commitments. We assume traders and farmers are homogeneous, i.e for each type of producer, costs are identical :

$$\begin{aligned} i, j \in \mathcal{I}, i \neq j, c_i &= c_j = c \\ k, l \in \mathcal{J}, k \neq l, v_k &= v_l = v \end{aligned}$$

Assuming that $\text{card}(\mathcal{I}) = N$ and $\text{card}(\mathcal{J}) = M$, the profit functions for each farmer and trader can be written as :

$$\Pi_i^W = \left(\alpha^W - \beta^W(N-1)q_{\bar{i}}^W - \beta^W q_i^W - \gamma M q^F - s - c \right) q_i^W \quad (\text{A.27})$$

$$\Pi_k^F = \left(\alpha^F - \beta^F(M-1)q_{\bar{k}}^F - \beta^F q_k^F - \gamma N q^W - v \right) q_k^F \quad (\text{A.28})$$

Where $q_{\bar{i}}^W$ denotes the quantities sold by all other traders different from trader k (and $q_{\bar{k}}^F$ for farmers different from farmer k). Given that all players in each type

are identical cost-wise, the reaction functions are :

$$\forall i, j \in \mathcal{I} : q_i^W = q_j^W = q^W = \frac{\alpha^W - (s + c) - \gamma M q^F}{(N + 1)\beta^W} \quad (\text{A.29})$$

$$\forall k, l \in \mathcal{J} : q_k^F = q_l^F = q^F = \frac{\alpha^F - v - \gamma N q^W}{(M + 1)\beta^F} \quad (\text{A.30})$$

The **Cournot-Nash equilibrium** is :

$$\text{Poaching} : q_{\text{Cournot}}^W = \frac{\beta^F(M + 1)(\alpha^W - (s + c)) - \gamma M(\alpha^F - v)}{\beta^W\beta^F(M + 1)(N + 1) - \gamma^2 NM} \quad (\text{A.31})$$

$$\text{Farming} : q_{\text{Cournot}}^F = \frac{\beta^W(N + 1)(\alpha^F - v) - \gamma N(\alpha^W - (s + c))}{\beta^W\beta^F(M + 1)(N + 1) - \gamma^2 NM} \quad (\text{A.32})$$

(A.33)

The primary market (between poachers and traders) must clear, and $s(x)$ equates supply and demand:

$$Nq_{\text{Cournot}}^W = q^W \quad (\text{A.34})$$

$$\iff s^{C^*}(x) = \frac{2W_2 N [\beta^F(M + 1)(\alpha^W - c) - \gamma M(\alpha^F - v)]}{\Omega} + \frac{W_1 \sigma x (\beta^F \beta^W (M + 1)(N + 1) - \gamma^2 NM)}{\Omega} \quad (\text{A.35})$$

Where $\Omega = \sigma^2 x^2 [\beta^F \beta^W (M + 1)(N + 1) - \gamma^2 NM] + 2W_2 N (M + 1) \beta^F$. Solving for the equilibrium quantity, the quantity supplied on the market by individual traders is :

$$q_{\text{Cournot}}^W = \frac{\sigma^2 x^2 [\beta^F(M + 1)(\alpha^W - c) - \gamma M(\alpha^F - v)] - \sigma x W_1 N (M + 1) \beta^F}{\sigma^2 x^2 (\beta^F \beta^W (M + 1)(N + 1) - \gamma^2 NM) + 2W_2 N (M + 1) \beta^F} \quad (\text{A.36})$$

In our case study, when $c = 0$, it shows that when the number of farmers is larger than the number of traders, the introduction of farming generates larger steady-state stocks. An interesting perspective is when there remains 1 sole trader, and the number of farmers increases: in this case, poaching is drastically cut down, as shown in Figure 4.A. When the number of traders is larger than the number of farmers, steady-state stocks decrease. In our context, when the number of traders is limited, increasing the number of farming facilities is a safe way to guarantee conservation outcomes.

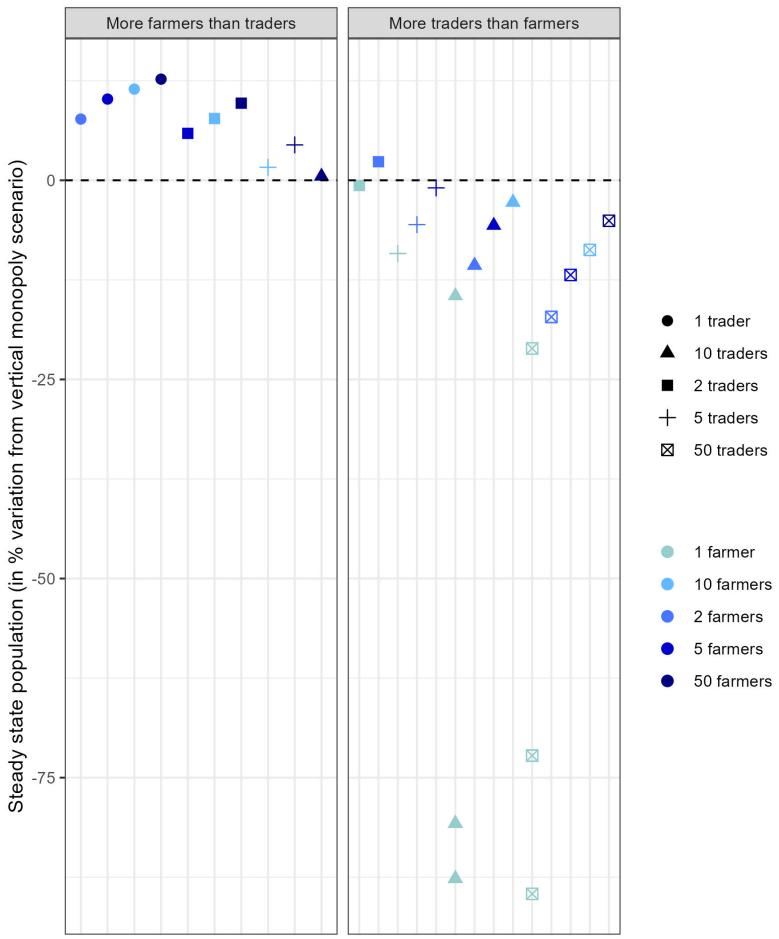


Figure 4.A: Steady state outcomes when multiple traders and multiple farmers are considered (an oligopoly) in the quantity adjustment scenario.

The left panel shows the steady state of the wild *Totoaba macdonaldi* population when there are more farmers than traders. The right panel shows the steady state of the wild population when there are more traders than farmers

Bertrand oligopoly : using the same notations as previously, the demand functions can be written as :

$$\forall i \in \mathcal{I} : q_i^W = q^W = \frac{1}{N}(a^W - b^W P^W - eP^F) \quad (\text{A.37})$$

$$\forall j \in \mathcal{J} : q_j^F = q^F = \frac{1}{M}(a^F - b^F P^F - eP^W) \quad (\text{A.38})$$

Using these demand functions and solving for the reaction functions in each case yields :

$$r^F(P^W) = \frac{a^F + b^F v + eP^W}{2b^F} \quad (\text{A.39})$$

$$r^W(P^F) = \frac{a^W + b^W(s + c) + eP^F}{2b^W} \quad (\text{A.40})$$

These reaction functions are the same as in the duopoly case (see eq. A.21). This result shows that aggregate production is invariant to the number of farmers or

traders as long as both are present on the market. Moreover, the individual production for traders is $\frac{1}{N}q_B^W$ and $\frac{1}{M}q_B^F$ with q_B^W and q_B^F referring to the duopoly equilibrium quantities for poached and farmed productions. In a Bertrand equilibrium, irrespective of the number of players, price-setting competition pushes the price to its minimum such that both firms still operate (given that traders have a stock-dependent production cost). Increased competition in the form of more players cannot push the prices further down. Therefore, aggregate output remains the same and individual production is divided among players.

This result further contradicts the results in [Damania and Bulte \(2007\)](#), as the authors find that increasing the number of players in a Bertrand set-up has detrimental effects on the steady-state stock. We find no effect, consistent with the theory and intuition.

Trader take-over of the aquaculture sector : in this section, we look at the 'extended cartel' scenario, where the vertical monopoly takes over the ownership of the aquaculture firm.

To gain intuition, assume poached and farmed products are perfect substitutes. On the one hand, the vertical monopoly has two production technologies: poaching (with a variable marginal cost, as the price paid to poachers depends on the population stock) and farming (with a constant marginal cost). In this case, the vertical monopoly equates the marginal costs across production units; that is, it buys a poached product to poachers up until the marginal cost of an extra poached unit equates to that of a farmed unit. In this case, if the marginal cost of farming is lower than market prices absent farming, then poaching goes down. Notice that the only way for traders to limit the price paid to poachers is to maintain a healthy stock. Therefore, the new equilibrium population stock is larger than the initial stock, and poaching is lower.

Now consider the case at stake, where products are imperfect substitutes. In this case, the extended cartel does not only equate marginal costs, as marginal revenues diverge across products. We use the following model to investigate the resulting equilibrium. Let the profit of the extended cartel be:

$$\Pi(q^F, q^W) = (\alpha^W - \beta^W q^W - \gamma q^F - (s + c))q^W + (\alpha^F - \beta^F q^F - \gamma q^W - v)q^F \quad (\text{A.41})$$

The extended cartel maximizes its profit with respect to the poached and farmed products. The poached production it sells on end markets is :

$$q^W = \frac{\sigma^2 x^2 (\beta^F (\alpha^W - c) - \gamma (\alpha^F - v)) - W_1 \beta^F \sigma x}{2(\beta^F W + \sigma^2 x^2 (\beta^F \beta^W - \gamma^2))} \quad (\text{A.42})$$

Figure 4.B shows that if the 'extended cartel' scenario arises, poaching goes

down, and the steady-state population increases.

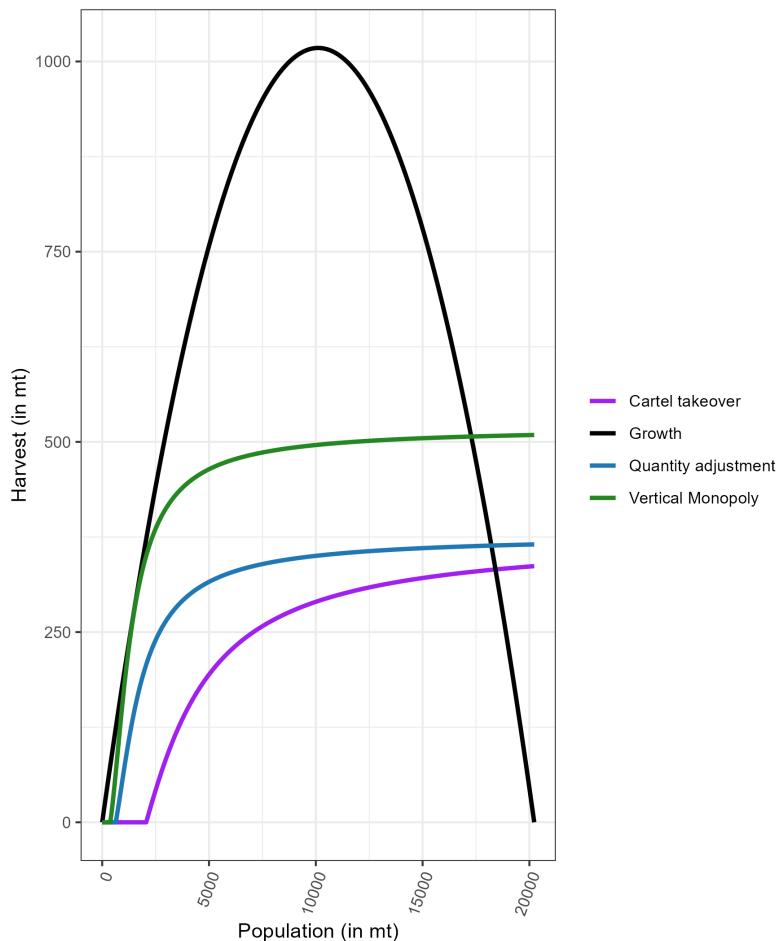


Figure 4.B: Steady-state equilibrium for the wild stock of *Totoaba macdonaldi* in the ‘extended cartel’ scenario, where the vertical monopoly takes over the ownership of farming operations

B Proofs to lemmas

B.1 Lemma 1 : content and proof

Assume $\alpha^W = \alpha^m$ and $\beta^m = \beta^W$, i.e., that the demand faced by the monopolist is the same as in the duopolistic case. Comparing monopoly and Cournot harvest functions:

$$\begin{aligned} q_m^W &\geq q_c^W \\ \Rightarrow v &\leq \bar{v} = \alpha^F - \frac{\gamma(\alpha^m - c)\sigma^2 x^2 - W_1 \sigma x}{2\beta^m \sigma^2 x^2 + 2W} \end{aligned}$$

First, look at when $x \rightarrow 0$:

$$\lim_{x \rightarrow 0} \bar{v} = \alpha^F$$

This requires that farming costs are lower than the choke price for consumers on their market. This condition is necessary for a farm competitor to enter the market.

Second, acknowledge that the second part of the equation is weakly decreasing, but non-increasing. Assuming the carrying capacity goes to infinity, it is limited by :

$$\lim_{x \rightarrow \infty} \bar{v} = \alpha^F - \gamma \frac{(\alpha^m - c)}{2\beta^m}$$

As fish abundance increases, the price paid to poachers decreases, as there is less scarcity. From equation (A.18), when $x \rightarrow \infty$, the price paid to poachers drops to 0. Moreover, notice that the last term in parenthesis is equation (A.7) for $s = 0$. Therefore, it means that the residual willingness to pay, when the poachers behave like a monopoly and $x \rightarrow \infty$, is larger than the unit cost of farming.

If the market is truly duopolistic, in the sense that the poachers could not manage the stock such that they depress demand so much as to kick their competitor out of the market, then Cournot competition unambiguously leads to lower poaching levels than a monopoly does.

B.2 Lemma 2

Assume that the demand parameters are unchanged by the introduction of farmed substitutes, that is to say $\alpha^W = \alpha^m$ and $\beta^W = \beta^m$, and use the definition of the coefficients for the direct demand function:

$$a^j = \frac{\alpha^j \beta^i - \alpha^i \gamma}{\beta^j \beta^i - \gamma^2}; \quad b^j = \frac{\beta^i}{\beta^j \beta^i - \gamma^2}$$

$$a^m = \frac{\alpha^m}{\beta^m}; \quad b^m = \frac{1}{\beta^m}$$

For $i, j \in \{W, F\}$ and m the monopoly case. To establish Lemma 2, we compare q_B^W and q_m^W . Equation (A.9) can be rewritten as :

$$q^m(a^m, b^m) = \frac{\sigma^2 x^2 (a^m - b^m c) - b^m W_1 \sigma x}{2\sigma^2 x^2 + 2W b^W}$$

Therefore:

$$q_m^W \geq q_B^W$$

$$\Rightarrow v \leq \frac{a^m - b^m c}{b^W b^F e} \left[\frac{2W_2 b^W (2b^F b^W - e^2) + (4b^F b^W - e^2) \sigma^2 x^2}{2\sigma^2 x^2 + 2b^m W_2} \right] - \frac{W_1 \sigma x [(4b^F b^W - e^2)(b^m - b^W) + e^2 b^W]}{b^W b^F e (2\sigma^2 x^2 + 2b^m W_2)}$$

$$- \frac{e a^F + c(e^2 - 2b^W b^F) + 2b^F a^W}{b^F e}$$

Notice that this equation can be reframed as :

$$F(x|c) \geq v \text{ where } F(x|c) = \Phi \frac{\eta + \mu x^2}{\theta + \nu x^2} - \frac{\kappa x}{\omega x^2 + \epsilon} - \zeta$$

And :

$$\Phi = \frac{a^m - b^m c}{b^W b^F e}, \eta = 2W_2 b^W (2b^W b^F - e^2), \mu = (4b^W b^F - e^2) \sigma^2,$$

$$\theta = 2W_2 b^m, \nu = 2\sigma^2, \zeta = (ea^F + c(e^2 - 2b^W b^F) + 2b^F a^W)$$

$$\kappa = \frac{W_1 \sigma [(4b^F b^W - e^2)(b^m - b^W) + e^2 b^W]}{b^F * e},$$

$$\omega = 2b^W b^F e \sigma^2 \text{ and } \epsilon = 2b^m b^W b^F e W_2$$

Analysis of $\Phi_{\theta+\nu x^2}^{\eta+\mu x^2}$: if $\mu\theta - \nu\eta < 0$, the first component of $F(x|c)$ is decreasing:

$$(4b^W b^F - e^2)b^m - 2(b^W b^F - e^2)b^W < 0$$

$$\iff \frac{\gamma^2}{\beta^m (\beta^W \beta^F - \gamma^2)^3} [\beta^m \beta^F + \gamma^2 - 4\beta^F \beta^W] < 0$$

Under the assumption that $\beta^m = \beta^W = \beta^F = \beta$, it is clear that

$$\frac{\gamma^2}{\beta(\beta^2 - \gamma^2)} (\gamma^2 - 3\beta^2) < 0$$

as $\gamma < \beta$. Therefore, $\Phi_{\theta+\nu x^2}^{\eta+\mu x^2}$ is decreasing $\forall x$

Analysis of $\frac{\kappa x}{\omega x^2 + \epsilon}$: the second component of $F(x|c)$ is increasing for $x \leq \sqrt{\frac{\epsilon}{\omega}}$, and decreasing after, since $x \in \mathbb{R}^+$. Noticing that $\kappa < 0$:

- For $x \in [0, \frac{1}{\sigma} \sqrt{W_2 b^m}]$, $\frac{\kappa x}{\omega x^2 + \epsilon}$ is negative and decreasing
- For $x > \frac{1}{\sigma} \sqrt{W_2 b^m}$, $\frac{\kappa x}{\omega x^2 + \epsilon}$ is negative and increasing

Conclusion : overall, $F(x|c)$ is such that :

- For $x \leq \frac{1}{\sigma} \sqrt{W_2 b^m}$, the first component is decreasing, while the second component is increasing
- For $x \geq \frac{1}{\sigma} \sqrt{W_2 b^m}$, the first component is decreasing and the second component is decreasing

Hence, $F(x|c)$ is bounded above by $\max(F(0|c), F(\frac{1}{\sigma} \sqrt{W_2 b^m}|c))$, and bounded below by $F(K|c)$ where K is the system carrying capacity. Therefore:

1. If $v < F(K|c)$, then Bertrand harvest is always lower than monopoly harvest

2. If $F(K|c) < v < F(0|c)$, then Bertrand harvest starts by being lower than in the monopoly case, but gets larger for large stock values.
3. Eventually, if $F(0|c) < v$, then Bertrand harvest is always larger than in the monopoly case

Figure 4.C illustrates this lemma with our parameter specification.

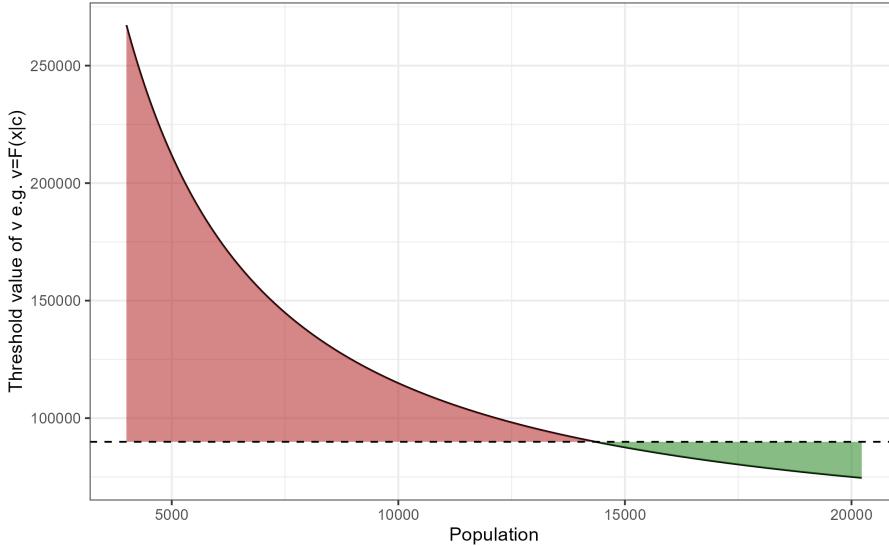


Figure 4.C: Evolution of the threshold v to compare vertical monopoly and price setting (Bertrand) harvest functions

In green, vertical monopoly harvests more than in the price setting equilibrium. For larger population values, in red, price setting leads to more harvest than the vertical monopoly. This illustrates our main specification and property 2 above.

Corner equilibrium: for a corner solution to emerge, it must be that $q_B^{w*} = 0$,

$$v = v(x) = \frac{W_1(2b^F b^W - e^2)}{\sigma x b^F e} - \frac{2b^F a^W + ea^F + c(e^2 - 2b^W b^F)}{b^F e} \quad (\text{A.43})$$

Equation A.43 shows that for low stock values, costs can still be positive and poaching disappear. However, to ensure that poaching is *never* beneficial in the Bertrand equilibrium, it must be that $v = \min v(x) = -\frac{2b^F a^W + ea^F + c(e^2 - 2b^W b^F)}{b^F e}$. In this case, the subsidy rate is so high that production is always beneficial for the farmer, and prices are too low for the trader to compete. In our baseline specification, this would amount to $v = -720,855$ USD.

C Supplementary Figures and Tables

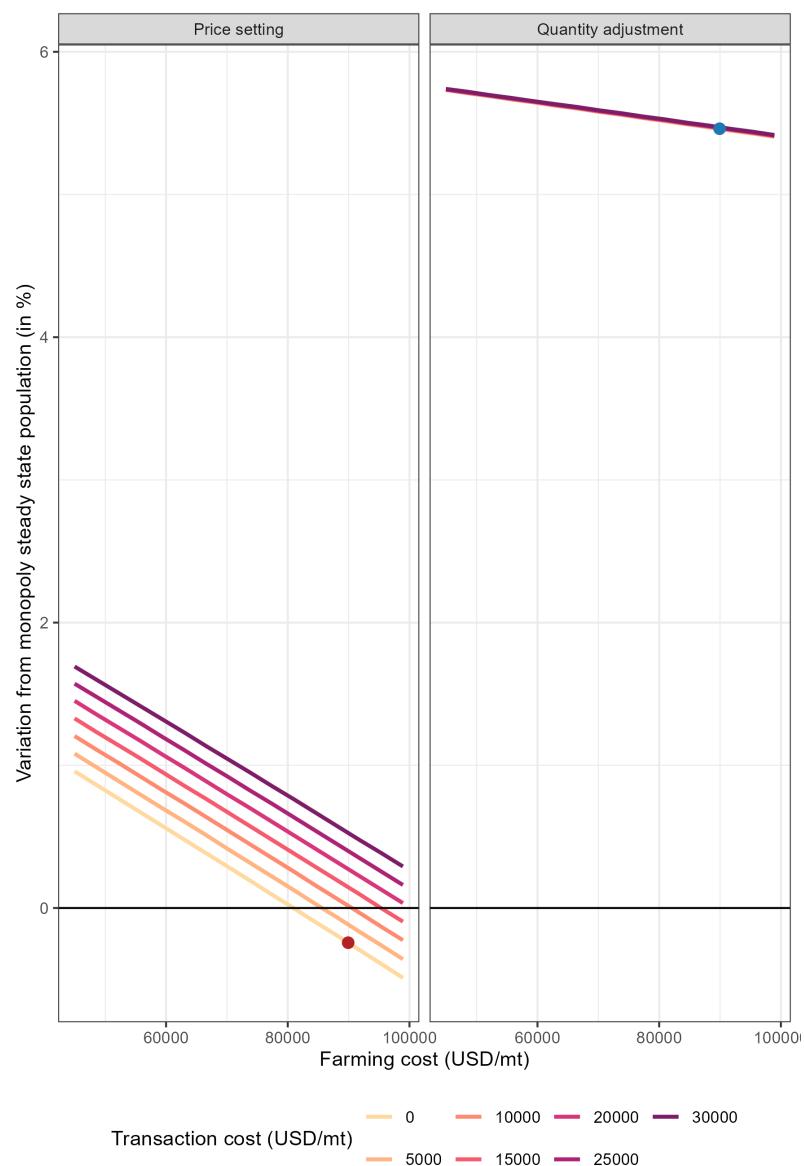


Figure 4.D: Percent change in steady state population across scenarios, following the joint evolution of illegal transaction and farming costs

Red and blue dots represent baseline results in the price setting and quantity adjustment scenarios

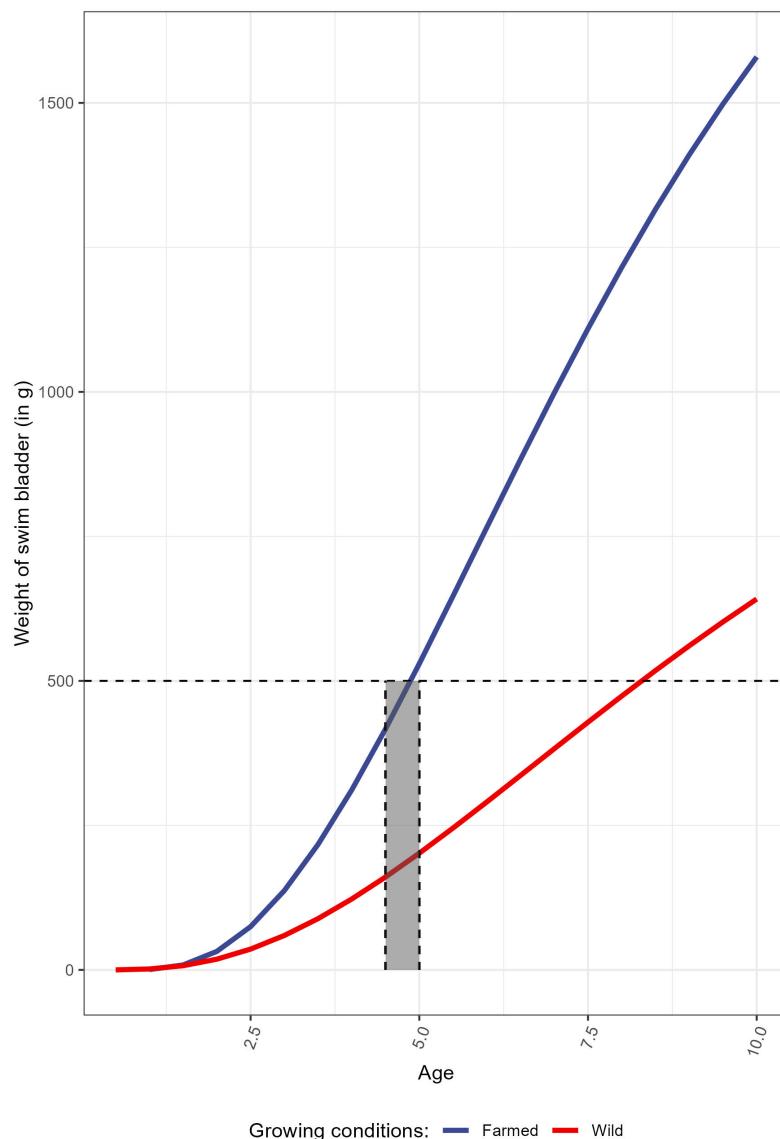


Figure 4.E: Von Bertalanffy Growth curves for wild and farmed *Totoaba macdonaldi* under different growing conditions

Gray box indicates the range of ages that possess a 500 gram swim bladder. The wild individual growth curve was calibrated with information from the stock assessment, while the farmed individual growth curve was calibrated using

Variable	Low Season	Mid Season	High Season	Source
Vessels	5	20	50	Cisneros-Mata (2020)
Days per month	4	12	14	Cisneros-Mata (2020)
Total fleet days year	20	240	700	Cisneros-Mata (2020)
Food fuel day	525	525	525	Semi-Structured Interviews
Totoaba gearset	2	3	6	Cisneros-Mata (2020)
Gear loss day	0.5	0.5	0.5	Semi-Structured Interviews
Gearset vessel per day	2	3	3	Cisneros-Mata (2020)
Gear replacement	1600	1600	1600	Semi-Structured Interviews
Bribes/year	600	7200	21000	Semi-Structured Interviews
Average cost (per vessel day)	8385.34	14386.69	5051.26	Authors' calculation

Table 4. F: Supporting information for the calculation of the *Totoaba macdonaldi* poaching cost parameters (W_1 and W_2)

The methods section details how and when semi-structured interviews were conducted.

<i>Dependent variable:</i>	
Price	
Catch	-1,563.752** (725.985)
Constant	1,625,837.000*** (406,789.500)
Observations	45
R ²	0.097
Adjusted R ²	0.076
Residual Std. Error	431,737.700 (df = 43)
F Statistic	4.640** (df = 1; 43)

Note:

*p<0.1; **p<0.05; ***p<0.01

Table 4. G: Regression output for the linear demand estimation calculated by regressing price data on catch data.

Data were obtained from the available literature that provided estimated weight and value of *Totoaba macdonaldi* maw seizures on estimated *Totoaba macdonaldi* catch from 2014 to 2017 obtained from a recent stock assessment. The methods section details where information was obtained from.

Variable	Value	Source
Sphere	1.00	Earth Ocean Farm Video, 2022
Capacity per sphere (t)	144.00	Earth Ocean Farm Video, 2022
<i>In \$USD</i>		
Maintenance year	12500.00	Felipe Ramirez, InnovaSea, 2018
Cleaning year	5000.00	Felipe Ramirez, InnovaSea, 2018
Vessel maintenance/year	10000.00	Tyler Korte, BlueOcean Mariculture, 2018; Fernando Cavalin, Earth Ocean Farms, 2018
Fuel year	25122.50	Author's Calculations
Feed	312480.00	Tyler Korte, BlueOcean Mariculture, 2018
Labor	1580000.00	Authors' calculations
Facility lease	150000.00	Cygnus Ocean Farms, 2017
Admin.	50000.00	Cygnus Ocean Farms, 2017
Operational costs	2145102.50	Authors' calculations
Operational costs (per t & year)	14896.55	Authors' calculations

Table 4. H: Supporting information for the calculation of the *Totoaba macdonaldi* farming cost parameter (v)

Annual cost estimates were obtained from informants and converted to \$USD. Capacity of each farming pen was obtained from Earth Ocean Farms, and an annual cost 706 per tonne of totoaba was calibrated using personal communications with totoaba aquaculture producers.

Parameter	Value	Concept	Units
α	1,625,836.98	Demand model : intercept	USD
β	1,563.75	Demand model : coefficient	USD/metric ton of biomass
γ	1,354.25	Demand model : substitutable good coefficient	USD/metric ton of biomass
r	0.20	Intrinsic growth rate	unitless
K	20,226.00	Carrying capacity (in metric tons)	metric tons of biomass
σ	2×10^{-5}	Catchability	% of biomass/vessel trip
$AvgCost$	14,386.69	Average cost per vessel trip at historical value	USD/vessel trip
W	3.75	Quadratic cost parameter - Quadratic cost function	USD vessel trip ⁻²
W_1	12200.00	Linear cost parameter - Linear quadratic cost function	USD/vessel trip
W_2	0.57	Quadratic cost parameter - Linear quadratic cost function	USD vessel trip ⁻²
v	89929.92	Unit cost of farming	USD/metric ton of biomass
i_r	0.10	Interest rate	%
Age	4.50	Age of farmed totoaba	Years
c	0.00	Unit cost of trading	USD/ metric ton of biomass

Table 4. I: Summary of *Totoaba macdonaldi* ecological and market parameters for model calibration

The methods section details where information was obtained to estimate each parameter, as well as relevant equations.

Concept	Formula	Reference
<i>Fishery</i>		
Growth	$\dot{x} = rx(1 - \frac{x}{K}) - \sigma xE$	eq. A.2
<i>Poaching</i>	s is price paid to poachers	
Harvest technology	$q = \sigma xE$	
Profit	$\Pi = s \times (\sigma xE) - W_1 E - W_2 E^2$	
Poached harvest	$q^W = \frac{s\sigma^2 x - W_1}{2W_2}$	eq. A.4
<i>Vertical monopoly scenario</i>		
Demand	$P^m = \alpha^m - \beta^m q$	eq. A.5
Profit	$\Pi^m = (P^m - s - c)q$	eq. A.6
Supply on end market	$q_m^*(x) = \frac{\sigma^2 x^2 (\alpha_m - c) - W_1 \sigma x}{2(\sigma^2 x^2 \beta^m + W_2)}$	eq. A.9
<i>Duopoly</i>		
Aquaculture profit	$\Pi^F = (P^F - v)q^F$	eq. A.10
Demand for imperfect substitutes	$P^W = \alpha^W - \beta^W q^W - \gamma q^F$	eq. A.12
	$P^F = \alpha^F - \beta^F q^F - \gamma q^W$	eq. A.13
Quantity adjustment (Cournot) supply	$q_C^{W*}(x) = \frac{\sigma^2 x^2 (2\beta^F (\alpha^W - c) - \gamma (\alpha^F - v)) - 2\beta^F W_1 \sigma x}{4W_2 \beta^F + \sigma^2 x^2 (4\beta^W \beta^F - \gamma^2)}$	eq. A.19
Price setting (Bertrand) supply	$q_B^{W*}(x) = \frac{b^W [\sigma^2 x^2 (b^F (2\alpha^W + ev) + ea^F + c(\sigma^2 - 2b^W b^F)) - W_1 \sigma x (2b^F b^W - e^2)]}{2Wb^W (2b^W b^F - e^2) + (4b^F b^W - e^2) \sigma^2 x^2}$	eq. A.23

Table 4. J: Summary of the key functions in the model

For model conclusions, the plotted functions are growth, vertical monopoly end market supply (q^m), quantity adjustment end market supply (q_C^W) and price setting end market supply (q_B^W)

Conclusion

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