

CLÉMENT VIGUIER

MOUNTAIN GRASSLANDS
DYNAMICS: INTEGRATING
PHENOTYPIC PLASTICITY
IN A NEW AGENT-BASED
MODEL

ABSTRACT

Mountain grasslands provide numerous ecosystem services that need fine understanding and characterisation to be assessed and predicted. The vulnerability to climate change and the complexity of mechanisms driving alpine community dynamics require the development of new tools to predict the dynamics of these communities facing new conditions. Moreover, individual variation has large effects on community responses to external condition changes, as shown by multiple empirical studies but often overlooked in modelling approaches. In addition to these effects, intra-specific variability has contrasting potential impacts on coexistence mechanisms that need to be disentangled.

To answer both the need for a dynamic model of species rich communities and the integration of individual level , the model *MountGrass* was developed. It is designed around two main components: (1) a closed strategy space allowing a efficient representation of high species diversity, and (2) a plastic allocation mechanism integrating trade-offs between active and structural tissues, as well as between shoot and root tissues. In a first result part, after a parameter filtering step, the combined effects of allocation rules, species strategy and phenotypic plasticity on individual plants are studied. In a second part, the effect of plasticity is then studied at the scale of the community.

This work demonstrates the importance of phenotypic plasticity both at the individual scale and its role for community dynamics. While further work is needed to fully capture plasticity mechanisms, the model provides sound starting point to further explore the role of intra-specific variability in coexistence mechanisms, the resistance and resilience to drought events, or the detection of regime shift in this type of systems.

ACKNOWLEDGEMENTS

I LOVE YOU ALL, BUT I LOVE YOU MORE MOM.

CONTENTS

I Introduction	VII
1 Context	1
1.1 Global change: how to describe the future of alpine ecosystems?	1
1.2 The need for new mechanistic models	5
2 Aims, Objectives, and Overview	8
2.1 Aims: understanding and prediction	9
2.2 Objectives: a new agent-based model for plant community dynamics	9
2.3 Thesis overview	10
II Background: community dynamics, traits and phenotypic plasticity	15
1 Understanding community dynamics and properties: drivers and theories	17
1.1 Community assembly and coexistence	17
1.2 The complexity of diversity	21
2 How to represent plant community	22
2.1 The continuity of functional ecology	23
2.2 How trade-offs make strategy space	25
2.3 How traits link to ecosystem properties	28
2.4 Modelling diverse plant community	31
3 The importance of phenotypic plasticity as a specific case intra-specific variability	33
3.1 Intra-specific variability change the rules	33
3.2 Phenotypic plasticity: a specific case of intra-specific variability	36
3.3 Toward an integrative framework of plant strategy and phenotypic plasticity	41
3.4 How phenotypic plasticity affect ecosystem properties and dynamics	43
III Modelling alpine grasslands with MountGrass, a generic framework integrating phenotypic plasticity	53
1 Alpine environment: conditions, resources, and perturbations	55

1.1	The scales of alpine grasslands	55
1.2	Resources: light and water	56
1.3	Perturbations: frost, grazing, and mowing	56
2	Multi-dimensional strategy space, carbon pools, and trade-offs	57
2.1	Multi-dimensional strategy space and allocation pools	57
2.2	Craft a trade-off: active and structural tissues	60
3	Modelling phenotypic plasticity	61
3.1	Plasticity as a strategy: between species memory and individual experience	62
3.2	Driving rules of allocation	64
4	ODD description of the model <i>MountGrass</i>	66
4.1	Model overview	66
4.2	Design concepts	70
4.3	Details	72
4.4	Limitations and problems	82
IV	Individual performance: strategy and plasticity	87
1	Model properties and individual responses	89
1.1	Parametrisation and sensitivity analysis	89
1.2	Individual level behaviour and properties of plastic allocation algorithm driven by the plant memory	101
2	Individual performance, plasticity and variable conditions	110
2.1	Individual performance: between strategy, memory and plasticity	110
2.2	Plasticity and variability of conditions	122
2.3	From model behaviour to competition and coexistence in the real world	136
V	Community dynamics	139
1	Community level simulations: non plastic community	141
1.1	Parameter filtering	141
1.2	Non plastic communities	142
2	Plasticity: impact on species fitness and diversity	142
2.1	Plasticity and diversity	142
VI	Synthesis & Outlook	147
1	Synthesis	149
1.1	Modelling diverse community	149
1.2	Effect of plasticity of mountain grasslands properties	149
1.3	On plasticity modelling	149
1.4	The limit of the species.	150

2 Outlook	151
2.1 Competition and feedback	151
2.2 Extend to climate change effects	152
3 Extensions	153
3.1 Ecology of plasticity: plasticity as a trait	153
3.2 Include nitrogen: source of trade-off	153
3.3 For more interaction	154
Glossary	157
4 Index	157

I

INTRODUCTION

1 CONTEXT

1.1 Global change: how to describe the future of alpine ecosystems?

1.1.1 The value of ecosystems: from properties to services

Everyone has a particular relationship with nature. The vision we put behind this word depends on the way we experienced nature, it can be temperate or tropical forests, mountain rivers or cliffs on the ocean littoral, bird songs or wind between stones. Anyone that shares one of these visions wants to preserve natural systems. But facing this emotional perception and inner desire to see these ecosystems be preserved, there are other forces that push in opposite direction. The reduction of biodiversity is increasing at dangerous rates, the deforestation threatens the largest forest systems, insects are less and less presents and animals are repelled to fragmented and diminishing habitats. Logics, other than emotional attachment and will to protect nature, impact all natural systems around the world because they are driven by other interests. To be protected, the natural systems needed a way to be integrated within these strong driving logics. The notion of **ecosystem services** was developed by [costanza_value_1997](#) to capture the value of **ecosystems**. It encompasses the benefits humans extract from ecosystems. It enables a categorisation of services and their quantification (that can go to the monetisation), and therefore allows them to be taken into consideration in the global logic of capital, investment and value.

A NEW LOGIC

The notion of ecosystem services aims to capture the value of ecosystems, but what is this value? In other words, what benefit do nature provide us? If ones could be tempted to answer that the value of an ecosystem cannot and/or should not be measured, it is clear that all ecosystems do not benefit to humans in the same way, and that these differences could be quantified. Facing the diversity of ecosystems, and the diversity of services they provide, we can try to develop a short answer for the object of study to this document: mountain grasslands.

SERVICES

The term **mountain grasslands** designates, in this document, all grasslands, below and above the treeline, that have short growing seasons delimited by snow-covered periods and experience high variations in temperature and water availability. This term is intentionally generic as the scope of this work is relatively broad and theoretical.

Mountain grasslands provide numerous services, that can be divided into multiple categories such as provision, cultural and regulating services (see figure 1.1). Provision services are related to the quantity and quality of primary resources the grasslands provide. Fodder production and quality are the main

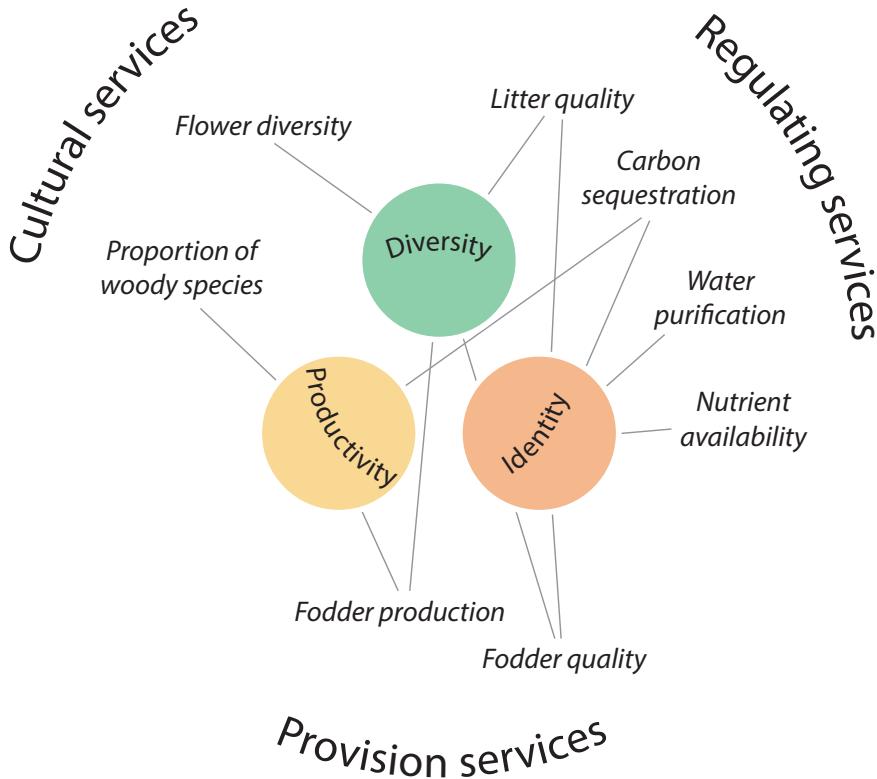


Figure 1.1: Three forms of plasticity in models.

measures of provision services. Other services can be included in this category: diversity of flowers and phenology for flower production for instance. Productivity is also interesting to assess carbon capture, a regulating service. Soil nutrient availability and water filtering are other regulating services impacted by the identity and diversity of species populating mountain grasslands. Finally, cultural services, related to tourism activity and landscape appeal are also related to grasslands species diversity.

In case of terrestrial ecosystems, vegetation cover is often central because of: its role of primary production, and the fact that vegetation community informs on the properties of the abiotic and biotic conditions. Moreover, most of studies on services from terrestrial ecosystem are interested in plants and soil invertebrate [de_bello_towards_2010](#) revealing the importance of vegetation in the provision of ecosystem services. In addition, in alpine habitats plant communities are susceptible to be the first impacted by the global change because they cannot escape changes in conditions and are the target of management practices linked to fodder production. All these arguments support the interest of studying the vegetation dynamics for the assessment of ecosystem services.

The ecosystem services are tightly related to the **ecosystem properties** (as illustrated in figures 1.1) (S. Lavorel and Garnier 2002; Díaz, Sandra Lavorel, Bello, et al. 2007) that can be extracted from the description of the grassland communities. Ecosystem properties are features of the community that characterise it and arise from the characteristics of all parts of the system or how they combine. The main properties of a plant community are captured in the following concepts:

PROPERTIES

- **identity:** the identity of the community refers to the dominant species (or

directly its characteristics) of the community that transfers its traits to the whole community. It can also refer to mean traits (with community-weighted mean measures) of a community. In this document, identity will often be used to talk about the resource use strategy (more or less exploitative). While this notion can encompass multiple traits and measures, it is practical to use one term to identify components of the community description that can be attributed to a species¹;

- **diversity**: diversity plays a large role in the provision of multiple services, and is related to other properties of the community. Diversity can be expressed in term of species richness or functional diversity², and by a wide range of indexes that are not discussed here. Despite a lot of nuances between these notions, they are often tightly correlated and diversity will be discussed in term of the number of species or functional volume in the rest of this document.
- **productivity**: productivity captures the capacity of the system to produce organic matter in a given timespan. It is an ambiguous term as it can refer to the abiotic environment, to a species or a community property or even to a service. I will try to limit its use to the species or community relative vegetative biomass in a given condition.

Linking ecosystem services to ecosystem properties is essential both for the understanding of processes controlling these services and for an easier quantification of such services. This is particularly important for the prediction of services levels to plan management practices in the context of global change. Some ecosystem services are here linked to the main community properties as illustrated in figure ???. Because services are hard to assess, ones can take advantage of this link and assess levels of ecosystems services thanks to a detailed description of the community; of both its structure and properties. The structure is defined by the relative abundance of the different species of the community, and properties result from the combination of the structure and the specific characteristics of present species. Multiple drivers affect the relative abundance and characteristics of a given species, from abiotic filtering processes to biotic interactions. So, ecosystem services also largely depend on abiotic factors (S. Lavorel and Garnier 2002). Therefore, there is a tight link between drivers, community structure and properties, and ecosystem services (see figure 1.2) that can be exploited to predict changes in ecosystem services (Lamarque et al. 2014).

The evaluation of ecosystem services relies on a precise description of the ecosystem abiotic and biotic properties. In mountain ecosystems, the plant community is the most dynamic and complex driver of ecosystem services, but direct links can be drawn between the fine description of the community and the ecosystem services. Understanding and prediction the main variables dynamics that capture those links is necessary to efficiently predict changes in ecosystem services levels. Plant communities are complex interconnected systems. In order to evaluate ecosystem services, they can be summarised by three main types of variables that capture different dimensions of such systems: the diversity, the productivity, and the identity. But grassland communities are natural systems driven by environmental variables, and changes in these drivers can lead to changes in services because of this link.

¹ in opposition to variables that are related to a system, e.g. diversity cannot be expressed for a species alone

² each measure depending on the functional space that is considered

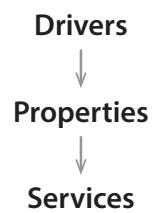


Figure 1.2: Link between abiotic drivers, community properties and ecosystem services.

1.1.2 Global change: what changes and what consequences

Mountain grasslands are maintained by strong climatic constraints that limit growth rate and lifeforms ([koorner_alpine_2003](#)), but also frequent grazing or cutting perturbation regimes that strongly limit the growth woody species and favour low stature species or rapid growth herbs (Díaz, Sandra Lavorel, McINTYRE, et al. 2007). But these drivers are changing at alarming rates with negative consequences on levels of ecosystem services [schroter_ecosystem_2005](#). Moreover, mountain grasslands are suspected to be very vulnerable ([schroter_ecosystem_2005](#); Engler et al. 2011) due to higher variations in water availability regimes and specific warming processes (Mountain Research Initiative EDW Working Group 2015), stronger isolation (island effect due to rise in temperature) and reduction of the grazing pressure.

The rise of carbon dioxide in the atmosphere due to human activities has a large impact on climate. The constant increase in mean temperature is the most known and easily observable phenomenon (see figure 1.3). But mountain grasslands will also suffer from more frequent and severe drought event, but also precipitation events ([beniston_climate_1997](#); Solomon, Change, and I 2007; Intergovernmental Panel on Climate Change 2014). They will also experience longer growing seasons and stronger invasive pressure from alien species and species from a lower altitude.

CLIMATE CHANGE

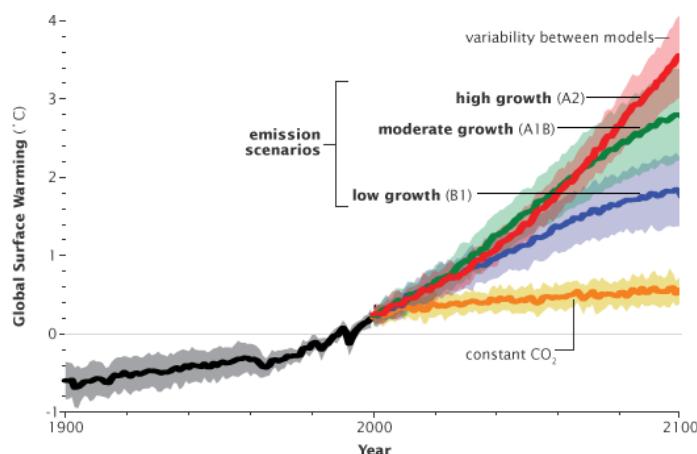


Figure 1.3: Historical models and projection scenarios for global mean temperature from Solomon, Change, and I 2007

In this context, the aptitude to plants to adapt to such changes and to cope with new competitors, no more filtered out by climatic conditions, will greatly determine the response of alpine communities (Alexander, Diez, and Levine 2015).

In addition to changes in climate, land use is also modified. Land-use, mowing or grazing in alpine grasslands, is a great filter for slow-growing perennial species that try to accumulate biomass over multiple seasons. Because of such asymmetric effects, land-use acts as a strong driver and can cause mountain grasslands communities to shift along service gradients ([schirpke_multiple_2012](#)). Land-use abandonment is suspected to greatly impact the invasion dynamics as it removes the pressure of biomass removal ([carboni_simulating_2017](#)).

LAND-USE MUTATIONS

Global change is a source of considerable changes, both in mean regimes, but also frequency and amplitude of climatic events. In addition to changes in

the climatic environment and resource availability, mutation of management of mountain grasslands will also affect community dynamics and particularly competition hierarchy. These modifications of strong drivers will have large effects on plant communities, and therefore their attributes and services they provide.

Mountain grasslands provide numerous services, that can be assessed thanks to the main attributes of the plant community. But global change threats these systems, and as consequence, the ecosystem services we take benefit of. We need tools to anticipate the effects of global change on these services and eventually adapt the management of mountain grasslands.

1.2 The need for new mechanistic models

1.2.1 The limit of classic patterns

The world is changing at a fast rate (**butchart_global_2010**; Intergovernmental Panel on Climate Change 2014), but most importantly in ways never experienced by living species in recent history. So, anticipating the effects of new environmental conditions on vegetation community cannot be built on the observation of previous or existing states. Extrapolation of complex system behaviour is generally not a good predictor of its actual behaviour. The complexity of the prediction goes beyond the multiplicity of dimensions impacted by the global change (rising mean temperature, frequency, and amplitude of drought events, reduction of cutting frequency or grazing abandonment, etc...), as the drivers often interact, synergise or negate themselves.

A NEW WORLD

To answer this challenge, large-scale experiments are conducted such as Cedar Creek experiment in the United-States, or JENA experiment in Germany. These experiments give high-value experimental data for various conditions and a variety of species, where interactions can be studied as well as management effects. Transplant experiments are also conducted to investigate the effects of temperature rise on the productivity, diversity, and identity of the community (example for SLA response Scheepens, Frei, and Stöcklin 2010, or see **debouk_functional_2015** for an increase in productivity and decrease in diversity, as well as a shift toward more acquisitive species).

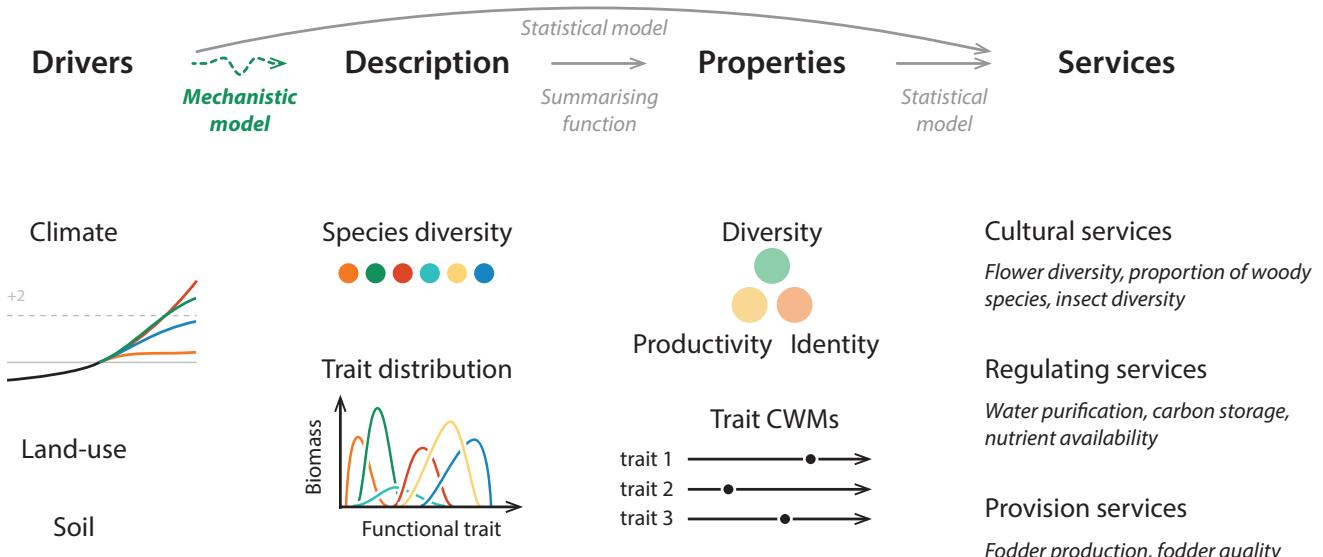
FIND BALANCE

But these common garden or transplant experiments also show contrasting response, that can come from opposite responses between intra-specific level and inter-specific level (Jung et al. 2014), between low and high elevation (changes in identity and contrasting effect in diversity between altitudes, observation data in Rosbakh, Bernhardt-Römermann, and Poschlod 2014) or between effects (see effect of warming and carbon dioxide on phenology in Reyes-Fox et al. 2016).

To accurately predict the future dynamics of grasslands communities, we need to be able to find the balance between dominant effects and eventually identify the interactions. For such complexity, empirical studies provide required and fundamental knowledge of processes and basic differences between effects, but no consensus can be made (Merilä and Hendry 2014) and new approaches need to use.

An additional argument for the use of alternative approaches is the uncertainty around the climatic scenarios (see figure 1.3). Indeed, the future of the planet atmosphere, and by consequence climate, is mainly depending on how

we are capable of changing our dependency to fossil energy (Intergovernmental Panel on Climate Change 2014). The will to adjust management scenarios to the future of vegetation community ([schipe_multiple_2012](#)) also require extensive experimentation (Rodriguez, Van Oijen, and Schapendonk 1999; Martin et al. 2012; Deléglise et al. 2015).



Mechanistic approaches allow better linking of drivers with community dynamics. This link can then be used to assess the level of ecosystem services as illustrated in figure 1.4.

1.2.2 When phenotypic plasticity makes things complicated

Within the context of climate change, the ability of species to adapt has a great influence on the response of the community. Indeed, the capacity of species to adjust to variations in drivers, via genetic variability and mutations, or thanks to plastic mechanisms, will certainly buffer the response of the community to changes in climate or land-use. [morin_comparing_2009](#) highlight stronger responses to climate change from vegetation communities within niche-based distribution models than within process-based models that capture adaptation mechanisms. More mechanistic processes should be included in these approaches ([evans_toward_2016](#)) to take into account adaptation mechanisms and interactions between species (Gilman et al. 2010). Plasticity can also change the competition intensity that increases negative effects of climate change (Hänel and Tielbörger 2015), while it can in other cases shift interactions from competition to facilitation Callaway, Pennings, and Richards 2003.

Phenotypic plasticity adds another level of complexity to the dynamic of communities and the interacting drivers. Statistical or expert based prediction cannot handle such complexity and mechanistic approaches have great potential to model complex systems.

Figure 1.4: From drivers of community dynamics to ecosystem services. The effects of main drivers (climate and land-use) on grasslands dynamics is captured thanks to mechanistic approaches to predict the composition and structure of the community. This description can then be used to assess the levels of ecosystem services through statistical models, to evaluate climatic scenarios or alternative land-use practices.

1.2.3 The rise of individual-based approaches

Individual-based-models (IMBs) let the complex behaviours of systems made of numerous interacting agents emerge from individual functioning. This type of modelling is extremely well adapted to the modelling of plant communities as we have a fairly good understanding of plant functioning and parameters are relatively easy to measure. The dynamics of essential resources is also relatively easy to compute. Yet, this apparent simplicity is relative (to animal modelling for example) and numerous models have been developed with various simplification hypothesis. Most of these hypothesis deal with the essential resources: light is often ignored in grasslands, while forest models focus on this aspect of resource competition. These hypotheses are most of the time justified, and the choice depends on the focus of the modelling exercise, and the importance of the given variables for the dynamics of the system.

These models are used to investigate the effect of climate change in the study of Rodriguez, Van Oijen, and Schapendonk 1999 with the model LINGRA-CC and show an increase in productivity. But the link to land-use practices is always questioned, and in this example, the increase in productivity allows a higher cutting frequency. Alternative scenarios are also explored in other grassland models Taubert, Frank, and Huth 2012; Taubert 2014; Maire, Soussana, et al. 2013; Maire, Gross, et al. 2009. Forest modelling present also numerous implementations of individual-based models (see Falster et al. 2016; Maréchaux and Chave 2017 for recent forest model examples).

BETWEEN CLIMATE AND LAND-USE

Other models based on processes can be used to study long term dynamics in the context of climate change in mountain ecosystems. It can be to study patterns of diversity (**boulangeat_fate-hd_2014**) or the impact of evolutionary processes on adaptation to climate change (Cotto et al. 2017).

1.2.4 Gaps to fill

A wide range of models has been developed to better understand biological processes involved in plant growth and population dynamics and the impact of climate change and land-use on these dynamics. They spread from organ-based models to functional types approaches. As the scale increases, the resolution diminishes and the verticality of processes is rarely taken into consideration. This is rarely a problem in stable conditions because the lower levels are implicitly integrated into the grain of larger processes (like the leaf gas exchanges regulation processes are ignored at the scale of the population). But two aspects can limit such simplification: (1) if the process is ignored instead of being integrated into higher level function (e.g.: stomatal regulation is often not modelled because it is assumed that it is correlated to photosynthetic activity, either because it is limiting the photosynthesis when the vapour pressure deficit is high, or it is down-regulated to avoid water loss when photosynthesis is limited by other factors). However, phenotypic plasticity is often ignored but not translated into the hypothesis of the model. Moreover, variables that are directly impacted by this process are explicitly represented (unlike stomatal conductance with stomatal regulation processes) leading to a misrepresentation of these variables (especially root:shoot ratio (RSR) or strategic traits like SLA). (2) if the non-modelled process has a great impact on the dynamic of the system.

Among models that target grasslands ecosystems (or more specifically) there is a dichotomy between growth models that are mainly interested in individual processes and species dynamics [lohier_analyse_2016](#); Soussana et al. 2012; Taubert 2014, and models interested in species-level processes and community dynamics ([boulangeat_fate-dh_2014](#); Cotto et al. 2017). The former focus on the individual growth of a limited number of species. They take into account fine-scale resource dynamics and interactions driven by explicit strategies and precise plant functioning. These models are on the side of the spectrum of the development models that often focus on a single species. The productivity of the system is often the primary concern and questions relative to the management of these systems are privileged over questions concerning climate change (but see Rodriguez, Van Oijen, and Schapendonk 1999, but still with the perspective of productivity). The latter is more interested in larger scale dynamics driven by the climate and evolutionary processes. The questions interrogated with these models are therefore more often relative to climate change and adaptive dynamics of the communities and the effects on community diversity and identity. These models are closer to DGVMs despite finer scale interactions. This dichotomy highlight the lack of integrative models that support community dynamics at long time scale with modelling of processes at the individual scale, based on explicit resource dynamics. The explicit modelling of the link between plant strategies, plant functioning, resource dynamics and plant growth allows a solid integration of plant interaction and external drivers (via the effect of resource dynamics and plant growth). Moreover, phenotypic plasticity can be integrated at the plant level, while its complex effects are emergent. Finally, considering the growth of individuals, the strategies of species and the dynamics of the population is required to build predict of all facets of mountain grasslands communities (diversity, productivity, and identity) that can integrate both management practices and climate scenarios.

DICHOTOMY BETWEEN MODELS

Because models have often practicality objectives, it is easier to develop a model that can be calibrated with species-specific empirical data. They can also be calibrated with Bayesian procedures and pattern-based approaches [hartig_statistical_2011](#). As a consequence, these models often integrate a limited number of species or functional types. This requirement of calibration limits the number of species simulated. To model diverse communities and evolutionary processes, this species diversity is required and generic framework must be adopted to avoid the calibration of individual species. Such diversity is observed in DGVMs that integrate trade-offs and multiple strategic axis ([kleidon_global_20000](#); Pavlick et al. 2013).

WHERE IS THE DIVERSITY

Mechanistic models are great tools and can be used to explore the uncertain future of mountain grasslands ecosystems. Bridges between individual-centred and generic community dynamics approaches must be built to take into account the complexity of population dynamics emerging from fine-scale interactions and plant functioning, driven both by the environmental conditions and species strategies. Considering both levels is compulsory to capture the complexity of responses of vegetation communities exposed to diverse drivers.

BUILDING BRIDGES

2 AIMS, OBJECTIVES, AND OVERVIEW

2.1 Aims: understanding and prediction

Global change is probably the biggest challenge humanity has to face at the beginning of this millennium. Actions are urgently needed to reduce the release of carbon dioxide but also mitigate the effect of climate change on natural and semi-natural systems. While solutions for the former must be found in technology, economics, and sociology, the ecology can help with the latter. But it requires an understanding of how the drivers impacted by global change will impact these ecosystems. The multiplicity of environmental drivers impacted by global change - whose effects can synergise or balance themselves -, in addition to complex structure and dynamics of natural systems make this understanding hard to build and to summarise.

To go beyond traditional pattern-driven ecology and overcome the difficulty of combined causes leading interacting effects, mechanistic approaches should be privileged.

The functioning of individuals living in these communities and the dynamics of the resources should be at the core of the new approaches to better understand the trajectories of the ecosystems.

2.2 Objectives: a new agent-based model for plant community dynamics

Traditional empirical approaches of observation and controlled experiments provided valuable information on the functioning of grassland ecosystems. However, they lack the power to understand intricate systems and predict their dynamics, especially in case of uncertain scenarios.

Modelling approaches must be used to build understanding and predictions of natural ecosystems dynamics driven by changing environmental drivers. These models should include a diversity of drivers as well as the diversity and the intrinsic complexity of these systems.

In order to compensate a long development time and to extend the reach of simulation experiments, models should try to be generic in structure and flexibility at use, while being specialisable thanks to parameters or simple equation changes.

2.2.1 Generic framework for multi-species and plastic plant modelling

In the context of mountain grasslands, showing unique levels of diversity despite strong environmental drivers, species diversity cannot be ignored to predict the response of the community. This diversity must be translated into species-specific functioning differences leading to diversity in niches and possible responses. In addition to species level dynamics driven by these differences, intra-specific responses cannot be ignored, and a phenotypic plasticity mechanism is needed.

2.2.2 Effect of phenotypic plasticity on plant growth, community properties, and dynamics

Intra-specific variations are expected to play an important role in the response of mountain grassland communities to global change. The effects of phenotypic plasticity and other sources of variations must be disentangled. Explicit integration of species-specific phenotypic plasticity in a plant community model will help identify and understand these effects.

As multiple services derive from the main properties of the vegetation of mountain grasslands, it is crucial to establish how phenotypic plasticity specifically impact these properties. Because these properties depend both on properties of the individuals and the relative abundance and diversity of species, effects on processes at both individual and community scales must be investigated.

2.3 Thesis overview

The rest of this thesis is divided into five chapters. The following chapter II, in the form of a literature review, introduces the concepts and knowledge that support the approach developed in later chapters. The chapter III develops the generic framework for plant functioning and phenotypic plasticity from the concepts established in chapter II and further extended. Chapters IV and V present respectively individual and community scale results of simulations made with the developed model *MountGrass* on the effects of phenotypic plasticity on main plant community properties. Finally, the final chapter discusses the outcomes of this work and possible paths to follow from the presented conclusions. Extensions to develop on the model are also proposed.

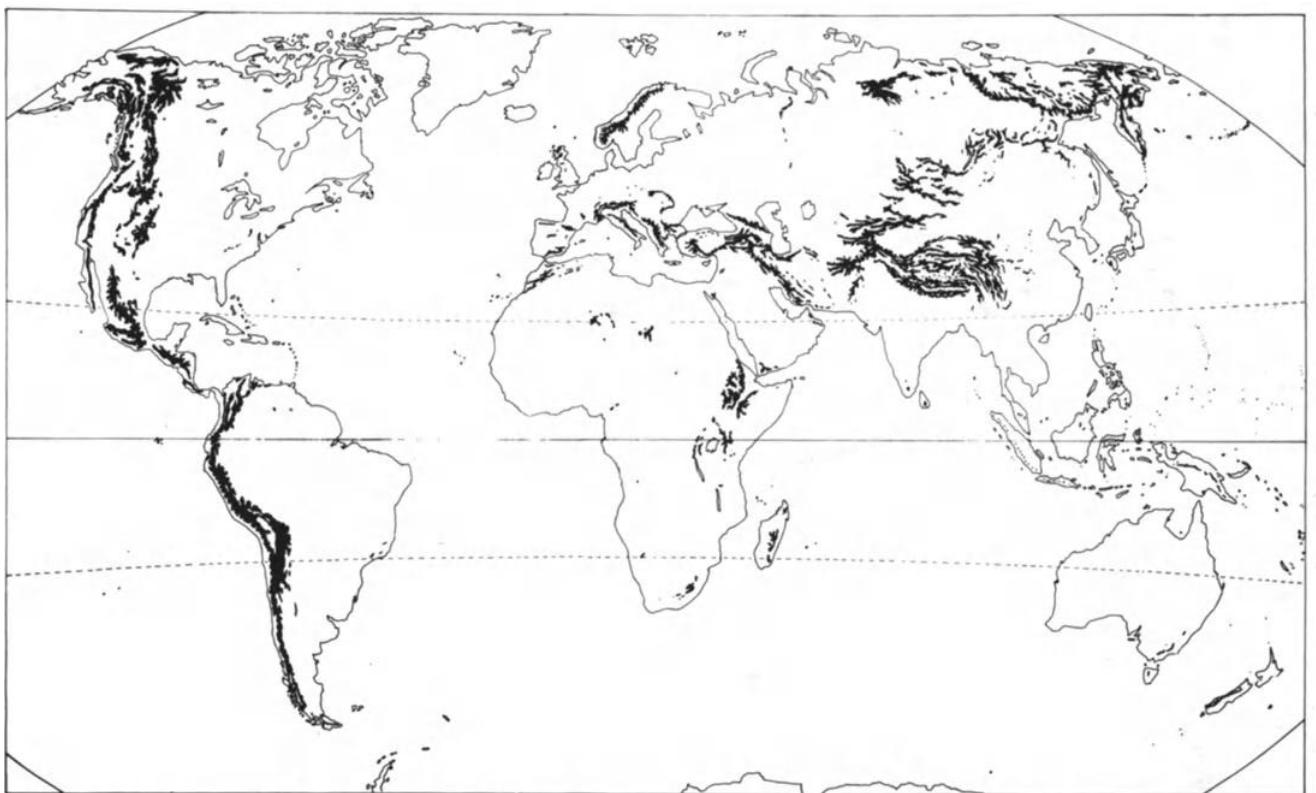


Figure 2.1: Distribution of alpine habitats. Alpine habitats shelter unique and rich ecosystems providing numerous services to human populations. Climate change and mutations of land-use practices threaten these dispersed and fragile habitats. From [korner_alpine_2003](#)

BIBLIOGRAPHY

- Alexander, Jake M., Jeffrey M. Diez, and Jonathan M. Levine (2015). "Novel competitors shape species/' responses to climate change". en. *Nature* 525.7570, pp. 515–518.
- Callaway, Ragan M., Steven C. Pennings, and Christina L. Richards (2003). "Phenotypic plasticity and interactions among plants". *Ecology* 84.5, pp. 1115–1128.
- Cotto, Olivier et al. (2017). "A dynamic eco-evolutionary model predicts slow response of alpine plants to climate warming". en. *Nature Communications* 8, ncomms15399.
- Deléglise, Claire et al. (2015). "Drought-induced shifts in plants traits, yields and nutritive value under realistic grazing and mowing managements in a mountain grassland". *Agriculture, Ecosystems & Environment* 213, pp. 94–104.
- Díaz, Sandra, Sandra Lavorel, Francesco de Bello, et al. (2007). "Incorporating plant functional diversity effects in ecosystem service assessments". en. *PNAS* 104.52, pp. 20684–20689.
- Díaz, Sandra, Sandra Lavorel, Sue McINTYRE, et al. (2007). "Plant trait responses to grazing – a global synthesis". en. *Global Change Biology* 13.2, pp. 313–341.
- Engler, Robin et al. (2011). "21st century climate change threatens mountain flora unequally across Europe". en. *Global Change Biology* 17.7, pp. 2330–2341.
- Falster, Daniel S. et al. (2016). "plant: A package for modelling forest trait ecology and evolution". en. *Methods Ecol Evol* 7.2, pp. 136–146.
- Gilman, Sarah E. et al. (2010). "A framework for community interactions under climate change". *Trends in Ecology & Evolution* 25.6, pp. 325–331.
- Hänel, Sabine and Katja Tielbörger (2015). "Phenotypic response of plants to simulated climate change in a long-term rain-manipulation experiment: a multi-species study". en. *Oecologia* 177.4, pp. 1015–1024.
- Intergovernmental Panel on Climate Change, ed. (2014). *Climate Change 2013 - The Physical Science Basis: Working Group I Contribution to the Fifth Assessment Report of the Intergovernmental Panel on Climate Change*. Cambridge: Cambridge University Press.
- Jung, Vincent et al. (2014). "Intraspecific trait variability mediates the response of subalpine grassland communities to extreme drought events". en. *J Ecol* 102.1, pp. 45–53.
- Lamarque, Pénélope et al. (2014). "Plant trait-based models identify direct and indirect effects of climate change on bundles of grassland ecosystem services". en. *PNAS* 111.38, pp. 13751–13756.
- Lavorel, S. and E. Garnier (2002). "Predicting changes in community composition and ecosystem functioning from plant traits: revisiting the Holy Grail". en. *Functional Ecology* 16.5, pp. 545–556.
- Maire, Vincent, Nicolas Gross, et al. (2009). "Trade-off between root nitrogen acquisition and shoot nitrogen utilization across 13 co-occurring pasture grass species". en. *Functional Ecology* 23.4, pp. 668–679.
- Maire, Vincent, Jean-François Soussana, et al. (2013). "Plasticity of plant form and function sustains productivity and dominance along environment and competition gradients. A modeling experiment with Gemini". *Ecological Modelling* 254, pp. 80–91.
- Maréchaux, Isabelle and Jérôme Chave (2017). "An individual-based forest model to jointly simulate carbon and tree diversity in Amazonia: description and applications". en. *Ecol Monogr* 87.4, pp. 632–664.
- Martin, G. et al. (2012). "Simulations of plant productivity are affected by modelling approaches of farm management". *Agricultural Systems* 109, pp. 25–34.

- Merilä, Juha and Andrew P Hendry (2014). "Climate change, adaptation, and phenotypic plasticity: the problem and the evidence". *Evol Appl* 7.1, pp. 1–14.
- Mountain Research Initiative EDW Working Group (2015). "Elevation-dependent warming in mountain regions of the world". en. *Nature Climate Change* 5.5, pp. 424–430.
- Pavlick, R. et al. (2013). "The Jena Diversity-Dynamic Global Vegetation Model (JeDi-DGVM): a diverse approach to representing terrestrial biogeography and biogeochemistry based on plant functional trade-offs". *Biogeosciences* 10.6, pp. 4137–4177.
- Reyes-Fox, Melissa et al. (2016). "Five years of phenology observations from a mixed-grass prairie exposed to warming and elevated CO₂". en. *Scientific Data* 3, p. 160088.
- Rodriguez, D., M. Van Oijen, and A. H. M. C. Schapendonk (1999). "LINGRA-CC: a sink-source model to simulate the impact of climate change and management on grassland productivity". en. *New Phytologist* 144.2, pp. 359–368.
- Rosbakh, Sergey, Markus Bernhardt-Römermann, and Peter Poschlod (2014). "Elevation matters: contrasting effects of climate change on the vegetation development at different elevations in the Bavarian Alps". en. *Alp Botany* 124.2, pp. 143–154.
- Scheepens, J. F., Eva S. Frei, and Jürg Stöcklin (2010). "Genotypic and environmental variation in specific leaf area in a widespread Alpine plant after transplantation to different altitudes". en. *Oecologia* 164.1, pp. 141–150.
- Solomon, S., Intergovernmental Panel on Climate Change, and Intergovernmental Panel on Climate Change Working Group I (2007). *Climate Change 2007 - The Physical Science Basis: Working Group I Contribution to the Fourth Assessment Report of the IPCC*. Assessment report (Intergovernmental Panel on Climate Change).: Working Group. Cambridge University Press.
- Soussana, Jean-François et al. (2012). "Gemini: A grassland model simulating the role of plant traits for community dynamics and ecosystem functioning. Parameterization and evaluation". *Ecological Modelling* 231, pp. 134–145.
- Taubert, Franziska (2014). "Modelling and Analysing the Structure and Dynamics of Species-rich Grasslands and Forests". PhD thesis. Osnabrück.
- Taubert, Franziska, Karin Frank, and Andreas Huth (2012). "A review of grassland models in the biofuel context". *Ecological Modelling*. 7th European Conference on Ecological Modelling (ECEM) 245, pp. 84–93.

II

BACKGROUND: COMMUNITY DYNAMICS, TRAITS AND PHENOTYPIC PLASTICITY

This chapter is dedicated to the review of the literature and aims to introduce the concepts and hypotheses used and interrogated in following chapters. A link between properties of the community and the ecosystem services is first drawn, then I examine the use of functional traits to represent plants, plant functioning, and communities. Finally, the impact of intra-specific variability, in particular phenotypic plasticity, on community properties is interrogated.

While this thesis is a modelling thesis, it is not a modelling textbook, and rather than an exhaustive description of the different types of models the focus will be given to selected modelling examples close to the context of this work.

1 UNDERSTANDING COMMUNITY DYNAMICS AND PROPERTIES: DRIVERS AND THEORIES

1.1 Community assembly and coexistence

1.1.1 Filtering processes: from potential to realised niche

A community is defined by the ensemble of species that coexist within the same space and time intervals. Communities were first viewed as a group of species that have evolved together to survive within specific conditions. To maintain itself within the community, each species need to grow during the vegetative phase, survive and reproduce. These steps of the life cycle result from the coordination of multiple physiological processes, supported by the extraction and use of essential resources: light, water, and nutrients. A part of community ecology sees communities as discrete entities with specific characteristics. This view is particularly practical for management as the community type can be associated with certain properties and services, or even particular dynamics and management systems. This view is the base of phytosociology as it is still used. While a discrete approach to community ecology provides practical categorisation, it ignores the fundamental dynamic nature of living systems. In a context of global changes, considering the dynamics of plant communities is crucial to predict how these systems will react to conditions never experienced. Another approach to community ecology considers that communities emerge from the distribution of individuals of a species, distribution controlled by its genetic and physiologic characteristics and its interactions with other species (Gleason 1926, Whittaker 1975). The distribution of individuals depends on how it is affected by abiotic conditions and interaction with other species or biotic conditions. The joint effects of the abiotic and biotic environment are captured by the concept of niche ([elton_1927](#)). The **niche** of a species is defined by how a species population reacts to abiotic and biotic conditions (resource, competition, predation, survival) and how it impacts its environment. Defining the niche of a species is primarily defining the barriers that constraint the distribution of the individuals of the species.

PLANT COMMUNITY

The **abiotic filtering** designates the non-biological variables that prevent the establishment of a species in a habitat. This term generally refers to climatic conditions and resource availability because temperature, water, nutrient and light availability are the main variables that constrain the plant development. Other abiotic factors can be considered, such as salinity L. Poorter and Bongers 2006 or soil properties (pH). These variables determine if a plant (depending on its specific properties) can establish a given habitat without any biotic interactions. These filters define, for a given habitat, the pool of species (or individuals if genetic variations are considered) that can grow and reproduce in this habitat without interaction. The ensemble of habitats a species can invade if only the abiotic factors are considered is called the **potential niche** (see figure 1.1).

ABIOTIC FILTERING

In addition to this large scale filters, another barrier may prevent a species to invade a habitat: its access. Indeed, dispersion plays a major role in the geographical extent of a distribution area of a species. Dispersion barriers such as mountains, seas or ocean prevent uniformisation of vegetation and reduction of global diversity. Such limits explain the existence of endemic species that grow only in a few locations, despite a larger potential distribution area (defined by potential niche).

DISPERSION FILTERING

Finally, the main factor that can affect the ability of a plant species to establish, is living interactions. For plant species, herbivory and competition are the most important factors, but other forms of interaction can affect the potential niche. The resulting niche, after all filtering processes, is called the **realised niche**. Competition affects the growth of the focal plant indirectly by reducing the availability of resources, increasing the stress of the plant and reducing its niche (see the interaction between species 1 and 3 in figure 1.1). Competition interactions are major factors shaping vegetation community and are extensively studied both with theoretical (Chesson 2000a; Amarasekare 2003) and empirical approaches ([kunstler_plant_2016](#)).

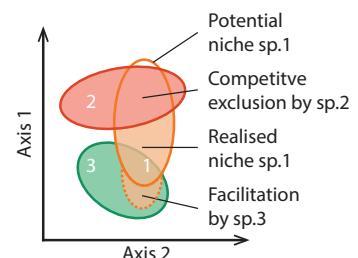
BIOTIC FILTERING

Figure 1.1: The potential niche of the **focal** species is reduced by competition interaction with **species 2**, but extended by facilitation interaction with **species 3**. This representation of the niche requires the knowledge of the effects of both abiotic factors and all pairwise interactions with other species. A more mechanistic approach of the niche should be considered in IMBs.

Similarly **facilitation** interactions also affect indirectly the levels of resources experienced by the focal plant, but in a way that is positive for the focal plant. So they widen the realised niche outside the potential niche (see the interaction between species 1 and 3 in figure 1.1). There are hypothesised to be larger along a stress gradient, where competition interactions are filtered out because they do not allow species maintenance and only positive interactions remain. Such relationships are dependent on the pair of species considered and may change depending on conditions (Callaway, Pennings, and Richards 2003).

FUNDAMENTAL NICHE

From the point of view of the focal plant, these interactions only exist through the changes in resource availability (even if plants are able to identify their neighbours). In this sense, we can see potential and realised niches as displacements of the fundamental niche (niche defined in term experienced conditions, stresses and resources) within spaces defined by abiotic variables or biotic variables. From this framework, the fundamental niche, or conditions experienced by the focal plant, is the stronger representation of the species niche and the realised niche (abiotic and biotic filters on the niche) emerge from the effects of external factors on this experienced environment.

This point of view should be adopted in models (Berger et al. 2008) because it allows the representation of both abiotic and biotic factors in a shared and generic framework. This is an improvement in comparison to models requiring

a matrix of interaction coefficient between species. Such matrix, in addition to being hard to parametrise, cannot be used in a framework of dynamic strategies. Modelling effort should instead be on explicit temporal and spatial dynamics of resource dynamics. Plant interactions would be captured by the effects of plant functioning (reduction of resource levels in relation to plant growth and resource use) on these dynamics ([morin_comparing_2009](#); Berger et al. 2008).

The concept of ecological niche serves as a great tool for theoretical research on coexistence. It encompasses in a convenient way both abiotic and biotic filters of one species distribution. While traditional view of the niche requires considering both abiotic filters and pairwise interaction, fundamental niches and resource dynamics modelling offer an alternative to model realised niche as an emergent property of the model.

1.1.2 The complexity of coexistence

If ones want to better understand and predict dynamics of complex systems, they first need to understand how such complex is assembled. Niches can be used to characterise a range of habitats a plant can live in, but because of complex inter-specific interactions, determining the final composition of a community from the list of species that can live in this habitat is not easy. If it is easy to observe diverse ecosystems (from bacteria to plants, insects or algae), it is challenging to determine the processes that 1) group the entities together (in time and space), 2) maintain an apparent stability in the group composition (at least at a certain spatial and temporal scale). We can imagine biotic filtering as a physical filter, the same way the abiotic filter is often illustrated, but this image does not translate the dynamic and complex nature of underlying processes. Biotic filtering emerges as the result of all the interactions between the entities that make it through the other filters. And how these interactions, direct or indirect, play together determines the stability of the diversity.

THE QUESTION OF COEXISTENCE

To predict the outcome of competition interactions multiple theories have been developed. Among these theories, we can cite two that have a different perspective on the same question: how do species sharing essential resources coexist in a homogeneous environment?

Chesson 2000b tends to have a population dynamic view of the system and identifies two types of processes that promote coexistence: (1) stabilizing mechanisms, (2) equalizing mechanisms. The former are required to stable coexistence as it a condition of invasibility. In other words, plants can coexist only if one species can invade the other. The condition to such invasion is that the species at low-density growth better than the species at high density. This is the case if intra-specific competition is higher than the inter-specific competition. Equalizing mechanisms are processes that diminish the fitness differences between the species, without ensuring stable coexistence. This framework is extended by [adler_niche_2007](#) in the modern coexistence theory. It states that niche differences (Levine and HilleRisLambers 2009) and fitness differences are the two mains axis of species coexistence. They make the assumption that niche differences define the relative strength of inter-specific versus intra-specific competition. The larger the differences between niches, the thinner is the overlap, and the weaker the inter-specific interactions. Therefore, this can be related to

stabilizing mechanisms in Chesson 2000a. On the other end, fitness differences also impact coexistence. The lower the differences, larger are the chances species coexist. The importance of niche differences required for stable coexistence decrease with the decrease in fitness differences.

In the other hand, Tilman elaborates a theory [tilman_resource_1982](#); [tilman_plant_1988](#) around resource use more in line with the idea of fundamental niche expressed in the previous paragraph, the contemporary niche theory. Species are characterised by the impact they have on the resource, and they use the resource for growth. Competition is in favour of the species with the lowest requirement for the resource because competition leads to resource deprivation it can survive. But coexistence if there is more than one limiting resource. In this case, coexistence can be achieved if species have a stronger impact on the resource from which they benefit the most (and intersecting zero net growth isolines).

These two theories give strong conditions for stable coexistence, however, they required simplifying hypotheses (all other things being equal, homogeneous environment) that are not met in natural environments. Despite their different approaches, these theories can be united as demonstrated by [letten_linking_2017](#) if the impact and benefit coefficients from contemporary niche theory are translated into niche and fitness differences. Despite this unified theory, they applied to a too limited range of situation to be applicable in the context of diverse mountain grasslands.

Plant community requires strong coexistence mechanisms to maintain species richness. Single theories fail to predict high diversity observed in plant communities such as natural mountain grasslands. However, high dimension coexistence processes and complexity seems to be an answer to the biodiversity paradox. In addition to niche based coexistence processes, other mechanisms that promote coexistence must be considered.

1.1.3 Variability and dynamics: driven by the resource

Resource dynamics, even with constant influxes, seems to be the key to understanding plant interactions and dynamics according to Tilman [tilman_plant_1988](#). Can the resource distribution in time and space explain coexistence?

In Tilman's perspective, resources are driven by two things, external influx and internal (to the system) consumption or cycle. The system structure and composition is responsible for resource dynamics as much as external influx. And these dynamics alter the structure of the community and change the hierarchy within the community. This cycle is well illustrated by the cycles we can observe in forest systems and gaps models. Mature forests produce big trees that fall down and create perturbation within the system. The resulting hole in the canopy allows for pioneer species to invade this space without competition. While they grow, other slower species are in shadows and must tolerate this competition, and grow enough to out-compete first established species. Because there is a trade-off between potential growth and shade tolerance allowing this cycle to set up, there is a succession dynamic after each perturbation of the systems. These local events of perturbation support coexistence at a large scale, a coexistence that can be captured by spatially explicit models Jérôme Chave 1999; Falster et al. 2016.

COMMUNITY DYNAMICS

Such drastic dynamics do not exist in mountain grasslands communities. But the natural temporal variability of resources due to contrasted seasons also drives diversity in growth strategies. Coexistence comes to the existence of multiple climatic contexts at the same place (but not the same time). As plants cannot be the most competitive species for any given condition in the whole range of conditions experienced in mountain habitats, there is a succession of species at the top of competition hierarchy (Adler et al. 2006) (see figure 2.5 for illustration). The diversity of flowering periods in figure 1.2 is an evidence of this succession dynamics.

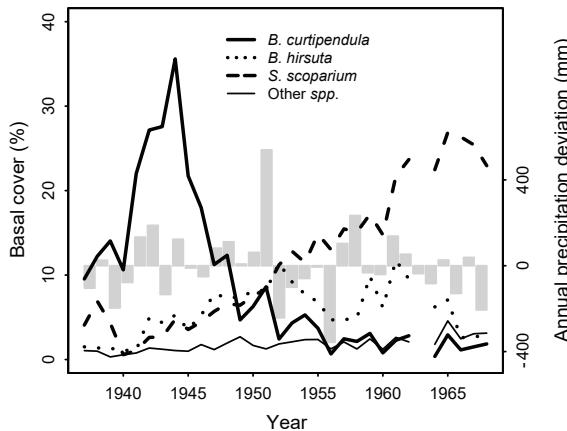


Figure 1.3: Changes in observed basal cover for 3 grassland species. This variation in hierarchy illustrates the succession in grassland communities and the storage effect due to the stabilizing effect of climatic variation promoting coexistence. See details in original study by Adler et al. 2006.

This mechanism promoting coexistence because of succession dominance driven by temporal changes in environmental condition is called storage effect. The species grow when the conditions match their niche and store the gains to wait until next favourable conditions. This term is generally applied to yearly variations, but the idea can be applied for variations within a growing season, allowing growth and storage until next season.

The temporal variations have a stabilizing effect on coexistence [tilman_plant_1984](#) but maybe more intuitively, spatial heterogeneity also promotes coexistence. Indeed, spatial variations of conditions at small scale create multiple niches that allow for diversity if measured at a higher scale. This spatial heterogeneity can be overlooked, but in the context of mountain grasslands, where plants are generally small due to high-stress levels and a very fine scale heterogeneity resulting from the terrain texture, it can play as a strong stabilizing mechanism.

Spatial and temporal heterogeneity play a major role in coexistence maintenance by creating various opportunity, or niches, in a given ecosystem. Internal dynamic variation of conditions also support stable coexistence.

1.2 The complexity of diversity

While resource use strategies and resource heterogeneity are important mechanisms for diversity, dispersal processes and meta-community dynamics should also be considered. Grassland communities are not independent of one another, but there are connected by dispersal vectors such as wind and animals. These connections support diversity but not stable coexistence, but remains crucial for community dynamics. Indeed, the link between the community and the meta-community (all connected communities) is a source a species that can be absent from the focal community. Therefore, in case of transition of environmental

TEMPORAL HETEROGENEITY

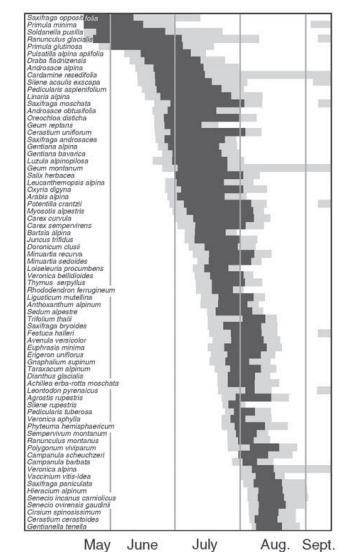


Figure 1.2: Diversity of flowering periods of alpine species. Evidence of succession in grassland ecosystems. From [korner_alpine_2003](#)

SPATIAL HETEROGENEITY

LABGEB SCALES DYNAMICS

conditions, these external species can invade the focal community, accelerating species turn-over compared to a closed community. In the context of global change, it is essential to consider mountain grasslands communities as open systems as the question of invasion by lower altitude species is yet to be solved.

Other larger scale dynamics can impact community dynamics such as species-specific interactions (herbivory or pollination) that lead to dynamic equilibriums. However, modelling such processes are demanding, and while it maintains some diversity, it is not expected to be the main driver of grassland dynamics in the context of global change.

Coexistence theory has difficulties explaining high species diversity in communities like freshwater diatoms or mountain grasslands that compete for a limited number of resources in fairly stable conditions. From the previous paragraphs, it seems that these environments are not that stable and that there are numerous mechanisms supporting diversity. Diversity is highly dimensional as it is stated by Clark et al. 2007. This complexity, that we just have scratched the surface here, is too high for theoretical models to handle. That does not mean they are not useful, but they cannot consider all these processes at the same time. To study diverse communities, it is required to incorporate at least parts of this diversity in mechanistic models. While it increases the modelling work, model's complexity, and difficulty to analyse results, it allows a stronger representation of communities, of their diversity and enables the identification of main processes, and possible interactions:compensations:synergies between these processes.

EMBRACING COMPLEXITY

The evaluation of services relies on a good representation of the plant community and its essential properties. To represent complex interacting systems like vegetation communities, descriptive approaches and theoretical models are not sufficient. The main driving processes must be considered and explicitly modelled. Explicit heterogeneity and dynamics of the resources are key to understand and model filtering processes, coexistence mechanisms, and community dynamics. Modelling both community properties and resource dynamics require an understanding of plant functioning and diverse growth strategies. The challenge of community modelling is not to keep simplicity in its structure, but increase diversity in its processes.

2 HOW TO REPRESENT PLANT COMMUNITY

All plants share the same pool of essential resources and similar physiological processes of assimilation and allocation, however, species differ by their growth rates, niches, and competitive abilities. How do such differences emerge from a common functioning wireframe? It seems that these differences can be explained by differences in parameters that characterise this functioning. So considering this diversity is required to represent the diversity observed in mountain grasslands.

The challenge of modern community ecology is to determine the trajectory

ries the existing ecosystems will follow under new environmental conditions. Species centred approaches, because they are limited to the knowledge of existing response patterns to existing gradients, cannot tackle this problem. How can a new representation of plants enable generalisation of the diversity of plant functioning in new conditions?

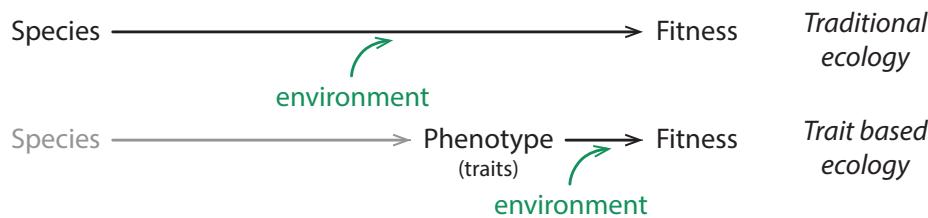
2.1 The continuity of functional ecology

2.1.1 Shift in paradigm: traits and patterns

Classical use of niche theory can be observed in Species Distribution Models (SDMs) that link the probability of presence of one species to a multidimensional description of a habitat. The environmental variables are literally used as the dimensions of the Hutchinsonian niche, and directly link the species to its fitness in a given environment (see figure 2.1, first row). This method is widely used to model environmental niche, but some can also include species interactions to incorporate an explicitly biotic filter. SMDs have good theoretical support and have a lot of practical applications, however, their strength is reduced at the scale of the community where the biotic filtering processes and fine scales dynamics take the advantage over large-scale abiotic filtering. Also, because they require a lot of data for any given species, they lack generalisation properties to be applied to rich communities. Community dynamics require fine-scale plant functioning processes to capture the effects of small scales variability and plant interactions, drivers of coexistence.

A SHIFT NEEDED

This example of modelling approach based on a species centred framework reveals the weaknesses of this framework. The distribution of a species along gradients, or its niche, while it can be captured by abiotic variables, is primarily determined by the fitness components (and whether or not they lead to a positive fitness): growth, survival, reproduction. These variables are not intrinsic properties of species but emerge from the interaction between physiological processes (carbon assimilation by photosynthesis, water absorption, organic matter allocation, etc...) and the environmental conditions. Only considering these processes allow to explicit and decompose plant functioning, and therefore model it in new combinations of environmental conditions.



Most of plant species share the same growth, survival and reproduction processes, but they still differ in these aspects as a function of the abiotic and biotic environment. The solution to shift from species centred paradigm, and its couple habitats-species (or species-environment-abundance like in SDMs), is to explicit the phenotype of these species. By using functional traits to define the phenotype of a species, ecologist can limit the representation effort to the link

Figure 2.1: The shift toward trait-based ecology allows for the decomposition of the link between species and fitness determined by the environment. On one hand, the link between species and traits is better characterised by standardised protocols and the use of databases such as TRY (TRY). On the other hand, the link between phenotypes (defined by trait values) and fitness can be generalised and the role of environment in this relationship better understood.

between traits and fitness physiological properties (P. B. Reich, Walters, and Ellsworth 1992), and then link species to traits with simpler data collection procedure ([cornelissen_handbook_2003](#)) (see figure 2.1, second row).

This shift in paradigm allows for a simpler and functional representation of plant species, that can be later linked to physiological or ecological processes.

The functional traits allow the decomposition of the link between species and fitness, to gain general understanding instead of specific relationships between species, environment, and fitness. However, this decomposition also breaks down the species, that can no more be described by one word, but needs instead multiple quantitative values to be described. To the singularity of the species is exchanged the multiplicity of traits. The link between species and fitness, now broken down by traits, can be analysed in a new light, parts by parts.

This decomposition allows the identification of relationships between morphological traits (easy to measure) and physiological traits (more interesting but harder to measure) (D. D. Ackerly and Peter B. Reich 1999; L. Poorter and Bongers 2006; Peter B. Reich 2014). Response patterns along climatic gradients have also been identified (Niinemets 2001) increasing the understanding of the role of the functional traits for the performance of plant species.

This trait-based approach, demanding in data collection effort, benefit from the consistency of the measures [cornelissen_handbook_2003](#) allowing pooling of the data into big databases such as TRY or Glopnet. The standardised collection of data all around the globe is a model of centralisation and collection that can lead to major large-scale pattern enhancing the understanding of the functioning of plant communities.

some interspecific patterns:

along climatic (Niinemets 2001; Ian J. Wright, Peter B. Reich, et al. 2004) ,on nitrogen (Dwyer)

but also correlations and intra-specific change of traits along gradients. Is it interesting?

The complexity of coexistence and community dynamics processes could not be captured with traditional species centred ecology. The last two decades saw the rise of functional ecology and its ability to capture quantitatively the relationship between vegetation and abiotic gradients. The capacity to

THE RISE OF FUNCTIONAL TRAITS

ARE THERE PATTERNS?

2.1.2 Traits and competition

If traits can describe a species and capture its functioning, it is logical to consider them to assess competitive interactions. Two visions have been developed to capture relative interactions. As mentioned in paragraph 1.1.2, trait distance can be a measure of competitive strength. This interpretation is an extension of the hypothesis of the limiting similarity that states that two species with similar niches cannot coexist. If plant functional traits can be used to define the niche, then, trait dissimilarity should be a measure of competitive interaction: the greater the dissimilarity, the lower the interaction. Because the competition is proportional to the absolute distance between traits, the relationship between distance and competition strength is symmetrical. On the other hand, some argue that competition interaction are not all symmetrical, but hierarchical, and that some traits can capture the competition sensitivity and others the competition impact ([kunstler_plant_2016](#)), therefore the intensity of the competitive interaction is

not symmetrical and dependent on the relative trait difference, but rather on the relative strength of impact traits compared to sensitivity traits. It seems that the form of the relationship depends on the type of competition mechanism considered. It will be hierarchical if they compete for the exact same resource (light, water), and symmetrical otherwise (temperature resistance, specific predation avoidance, pollinator, etc...).

Understanding how competition (or any other interaction) is regulated by traits is important to determine competition outcomes with alternative methods than pairwise coefficients that require empirical data to determine. Linking traits and strength of competition interaction would also allow the intra-specific variations to be considered. In this case, determining the exact relationship between trait distance and the competitive effect is crucial as it would change the effect of intra-specific variability (see Hart, Schreiber, and Levine 2016 for example).

But these interactions are not only symmetrical or asymmetrical, there can be non-transitivity promoting dynamic stable coexistence (Levine, Bascompte, et al. 2017), or be context dependent (Callaway, Brooker, et al. 2002). Moreover, the nature of the competition relationship (dissimilarity or hierarchy) depends on traits considered (Bennett et al. 2016). Due to their complexity, interactions cannot be summarised by single trait value comparison but is multi-dimensional (Kraft, Godoy, and Levine 2015). However, traits can inform competitive interaction by informing the plant functioning and the use and effects on the resource.

Traits can be a good proxy for competitive interaction but the relationship between trait differences and competition intensity depends on the competition process. If the interaction is transitive, a strong asymmetric pattern can be observed between interaction effects and trait differences, while symmetric interaction reveals niche differentiation processes. Despite these observed relationships, the specificity and multiplicity of trait-mediated interactions promote the use of mechanistic solutions to capture the multi-dimensional and context-dependent nature of plant interactions.

The paradigm shift toward functional ecology allowed the shift from discrete to a continuous representation of species. This change makes easier the representation and study of plant communities, especially along conditions or management gradient. Traits are also used to study plant interactions. Trait approaches offer a functional link between morphology and physiology that has great potential in generalising environmental effect on the phenotype-fitness relationship. However, the need for multiple traits to capture plant niche differences or similar response patterns of multiple traits suggests underlying structure within trait assemblage. Understanding this structure and how it relates to community dynamics external drivers is crucial in the representation of diverse communities.

2.2 How trade-offs make strategy space

2.2.1 Trade-offs: capture constraints on species differences

The functional link that is observed between some morphological traits and physiological traits suggests underlying processes that link these traits together. It appears that multiple traits are correlated together at the global scale between species (P. B. Reich, I. J. Wright, et al. 2003; Ian J. Wright, Peter B. Reich, et

al. 2004; Jerome Chave et al. 2009; Peter B. Reich 2014) and within species (Hu et al. 2015). This correlation between functional traits of the leaf was described at a global scale by Ian J. Wright, Peter B. Reich, et al. 2004. The **Leaf Economic Spectrum** (LES), defined by these correlations between multiple traits, draws a continuum of strategies. It spreads from species with high resource acquisition rates and rapid growth rates but low tissue lifespan, to species with longer tissue lifespan but lower growth rates. This is a clear description of a **trade-off** between strategies, opposing exploitative strategies (high Specific Leaf Area (SLA), high Leaf Nitrogen Content (LNC) and low Leaf LifeSpan (LLS)) to conservative strategies.

This axis of differentiation allows ecologists to link quantitative measures to types of strategies that better capture diversity of strategies than discrete typology. These strategies are translated into traits, traits that can be translated into physiological processes parameters, then into components of fitness.

In addition to a quantitative measure of species-strategies, such trade-offs simplify a lot trait-based approaches. While many variables can be measured on one individual, correlations between these variables reduce the number of dimensions to consider. This simplification cannot be better illustrated by the work of Diaz et al. 2004 that demonstrate the existence of two major axes of "evolutionary specialisation" that explain most (41%) of trait variability: size-related traits, and resource use speed traits. Similar evidence is also found on a global scale in addition to evidence for high levels of coordination between axes (Díaz et al. 2016).

Similar correlations could be found in roots P. Ryser 1996; Peter B. Reich 2014 but patterns are generally weaker, certainly because of more fragmented data and interactions with micro-organisms that alter the link between morphology and function of roots.

The existence of such trade-off can be explained by constraints that shape the distribution of trait distributions. Trait-function relationships are often depicted as bell-shape with an optimum (Cécile Hélène Albert, Thuiller, Yoccoz, Soudant, et al. 2010). I rather think that trait and function are linked by monotonous functions, but traits are generally not independent and another monotonous trait-function relationship can constrain the first function. For example, the exchange function of the leaf (and photosynthesis activity), is negatively linked to the thickness of the leave (promoting thin leaves for a higher light capture and photosynthetic activity, but the lifespan and mechanical support of the leaf require denser leaves to be viable. This trade-off in functions, linked by a trade-off in traits (the leaf cannot be both thin and light in one hand, and robust and self-supporting in the other), lead to the emergence of a strong constraint ("hard frontier" in figure ??) on one side of the relationship, while competition processes out-select combination of traits that are not relevant on the other side ("hard frontier" in figure ??).

Trait-based ecology rapidly lead to the observation of trait correlations and trait syndromes between plants. These axes of differentiation emerge from processes that constraint plant strategies. Global characterisation of these constraints should allow a better representation of plant functional diversity.

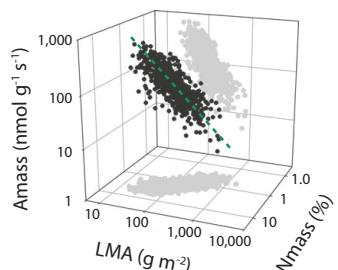


Figure 2.2: Three dimensions of the LES. Correlation of Leaf Mass Area, assimilation rate per mass unit and nitrogen concentration. This correlation reduces three dimensions (more dimensions not shown) into one axis (- -).

STRATEGIES

EMERGENCE

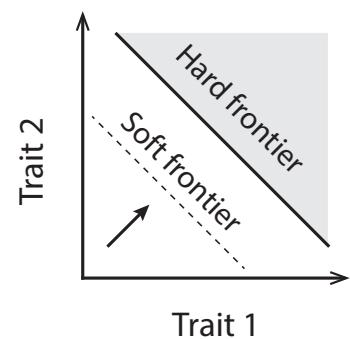


Figure 2.3: Emerge of trade-off between traits because of hard physical-biologival frontiers, and "soft frontier" due to selection.

2.2.2 Strategy-spaces made of trade-offs

Plant diversity is expressed, and visible to anyone, by the variation in shapes and colors, scents and growth forms, but this diversity is the demonstration of the multiplicity of strategies. In an early attempt to make sense of this diversity of strategy Grime 1977 theorise the existence of two type of constraints that shape plant communities: perturbations and stress. The perturbation axis captures the variability of community drivers, while the stress axis captures how conditions facilitate or make difficult plant establishment. They draw a two-dimensional space where three regions can be invaded¹, corresponding to three different strategies: competitive (C) in low stress-low perturbations region, stress tolerant (S) in high stress-low perturbations region, ruderal (R) in low stress-high perturbations region, forming Grime's triangle (see figure 2.4).

Grime's triangle set the basis for strategy space, and the broad meaning of *stress* and *perturbations* terms allow them to be applied to various conditions. However, the diversity of types of stresses (drought, cold, nutrient availability) and perturbations (predation, fire, avalanches etc...) cannot be specifically captured by such wide concepts. Westoby 1998 highlight the difficulty to use such space and its incapacity to explain some patterns. According to him, a strategy space² should:

- "express meaningful differences in ecological behaviour between species";
- allows to "position a plant species from anywhere in the world within";
- be composed of attributes that "require little enough effort to estimate";
- lets "possible to quantify the extent to which the [strategy-space] captures variation in other plant attributes".

He proposes to use functional traits to meet these criteria of functional differences, generalisation, and practicality. Three traits capture the components of Grime's triangle:

- Specific Leaf Area (denoted L): captures the speed of return of investment of carbon in leaf, as latter highlighted in the LES. High SLA is generally associated with competitive species that capture a lot of light and have a high growth rate. At the other end of the spectrum, low SLA species are more stress tolerant. This axis is the practical equivalent to the axis CS in Grime's triangle.
- Height at maturity(H): the race to the light (but not time fixed as the protocol for functional trait measure encouraged it), but also capture ruderal axis (time interval between perturbations)
- Seed mass (S): expresses the capacity of a species to invade freshly disturbed environments or the competitive advantage seedlings possess with a larger starting carbon pool. This trade-off between the competitive strength of seedlings against the chance of invading freshly disturbed environment capture well the CR axis of Grime's triangle.

The LHS strategy space proposed by Westoby has the advantage to be easily measurable and to allow comparison between species all around the globe (Pierce et al. 2013).

FROM THEORY TO TRAITS

¹ high stress and high perturbation regions does not allow establishment

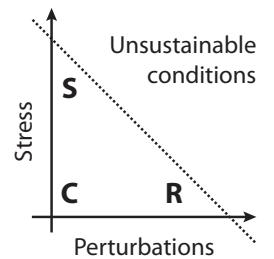


Figure 2.4: Grime's triangle. Competitive (C), stress tolerant (S), and ruderal (R) strategies are dominant in the three regions of the perturbations-stress space.

² called Plant Ecology Strategy Scheme (PESS) in his paper

This approach can be further extended with multiple traits. Indeed, global datasets and databases of functional traits reveal global scale correlations between traits. These correlations, or trade-offs, simplify the representation of plant species (Díaz et al. 2016) and translate fundamental axes of strategy differentiation (**reich_world-wilde_2013**). Yet, plant communities exhibit extraordinary species and functional diversity suggesting that not all traits are correlated. Trade-offs emerge because of hard (physical, chemical or biological) and soft (competitive pressure) constraints on combinations of functional traits (see figure ?? and Shipley et al. 2006). Therefore, for a given couple of traits, the physical independence of traits and the independence of ecological processes they are involved in should ensure the absence of trade-offs between those. While some traits are related to multiple physiological processes (a composite trait like SLA is involved in water regulation, but also light capture and tissue toughness), traits are often specific to one or two processes.

These processes must have an ecological significance, so filtering processes are the best candidate to observe such trade-offs. Against climatic filters, plants can either escape (*i.e.* finish a life-cycle before the filtering event) or avoid/resist (develop specific tissues or strategy to pass the filter). This can be observed for drought (Kooyers 2015) or frost (**korner_alpine_2003**). The resource use strategies and reproductive strategies are also orthogonal (**diaz_global_2014**). From this, a generic principle can be formulated stating that the number of observable trade-off in an ecosystem is close to the number of constraining processes. It is supported by the observation that a limited number of traits (or dimension, or trade-off) is often enough to capture the diversity of vegetation community as in Laughlin 2014.

The independence of strategic trade-off justifies that these trade-offs can be used as independent dimensions of a **strategy space**, defining the diversity of strategies present in a community.

The existence of such low dimensional strategy spaces have been observed at large scale (Pierce et al. 2013; Díaz et al. 2016) but also at smaller scales, where the CSR triangle could be identified in ecosystems where precipitation stress and grazing perturbation are shaping the community Frenette-Dussault et al. 2012.

The multiplicity of processes shaping vegetation systems leads to similarly constrained diversity in plant strategies. These strategies are captured in a strategy space drawn by independent trade-offs tightly related to functional traits. These functional trade-offs have great potential in the representation of a functioning plant diversity, while parameter set allows easy characterisation of species and communities.

GENERALISATION OF STRATEGY SPACES

EMPIRICAL EVIDENCE

2.3 How traits link to ecosystem properties

Now that functional traits, trade-offs and strategy spaces are identified as good candidates to characterised the plant functioning and differentiate species, can we link functional traits to **ecosystem properties** and by extension to ecosystem services.



Figure 2.5: Empirical evidence of the CSR triangle in natural communities. The CSR triangle is created by the translation of a multivariate analysis into a coordinate system (see Pierce et al. 2013 SI for details). "CSR classification of six species of the genus Poa from lowland (left column; P. annua, P. nemoralis, P. supina) and alpine (right column; P. alpina, P. chaixi, P. laxa) northern Italy. Grey circles represent the strategies of individuals, and black circles the mean strategy for the species based on the individuals included in the analysis." from Pierce et al. 2013.

2.3.1 Mass Ratio Hypothesis, Community Weighted Means, and functional identity

As explained, plant species, based on their identity, provide ecosystem services. Some of these services are direct consequences of the characteristic of the species and their functioning. Because of that, Grime 1998 formulates the **Mass Ratio Hypothesis** that states:

... the extent to which a plant species affects ecosystem functions is likely to be closely predictable from its contribution to the total plant biomass. - Grime 1998

Because functional traits are quantitative variables, they can be manipulated more easily than factors. Therefore, while phytosociology describes vegetation communities with broad types and approximate abundances, trait-based ecology benefit from this continuity to characterise mean properties of community. The **Community Weighted Mean** of a functional trait is the average of species-specific trait values weighted by the relative abundance of each species, and correspond to an extended mathematical application of the mass ratio hypothesis when functional traits are linked to services. These summary variables define the communities in a quantitative way similar to the functional trait for species. In addition to be quantitative, it is functional and responses to disturbing factors can be predicted (Lavorel and Garnier 2002).

According to the **Mass Ratio Hypothesis**, some properties of the community directly scale to the characteristics of the most abundant species. In this hypothesis, the **functional identity**, defined by functional trait values, has more importance than the identity of the species. Community Weighted Mean measures generalise this hypothesis using mean species trait values. While these tools can link community composition to ecosystem properties and services,

they require precise measures of plant functional traits to be reliable.

2.3.2 Benefits of diversity

Certain processes are determined by the most abundant species of a community, but other services and functions may result from the properties of the group. **Diversity** is the most important property of an ecosystem or a community for a wide audience. This measure is peculiar to groups of organisms and plays a major role in its functioning and the services it provides. Diversity can refer to species richness or functional diversity. The former quantifies the number of species present in a habitat and can take into account the relative abundance of the species. Many indexes can be used to measure this variable representing different perspective or aspect of the metrics (see Chalmandrier 2015 for exhaustive information).

Functional traits and functional diversity can be used to estimate certain ecosystem services. For example, the diversity of phenology captured in flowering periods (see figure 1.2) is an indicator of the recreational function of mountain grasslands.

But diversity also supports indirectly functions and other properties of the system. Multiple mechanisms explain this multiplicity contained in the measure of diversity.

First importance of species richness is found in the insurance effect that prevents the loss of a function or a service with the loss of a species by ensuring that multiple species provide such function or service (see • species in figure 2.6). Because insurance effect relies on the redundancy of function, this effect is better captured by species richness than functional diversity. Another way of seeing this notion is the selection effect that states that increasing diversity increases the potential number of services provided by the community (see • species in figure 2.6), as each species added can provide new function/service (or at worst reinforce already present ones). When the function or service is directly linked to a trait value, this selection effect is directly captured by an increase in functional diversity.

Functional diversity is also associated with higher productivity and resistance to invasion or perturbation events due to niche complementarity mechanisms.

Species richness and functional diversity are often strongly correlated, but they do not capture the same services or effect. Functional diversity is a strong indicator of niche complementarity and its benefits.

2.3.3 Productivity: both community property and ecosystem services

Productivity of a plant community is mostly sensitive to abiotic conditions, precipitation, nitrogen, and temperature being the main variables influencing this productivity. Because of that, there is a large contrast between ecosystems in contrasting environmental conditions (tropical forests and mountain grasslands being two extremes). These differences can be observed in the distribution of functional traits of species, size and resource use related traits being the most eloquent ones.

While community productivity depends heavily on environment properties (climate, soil), it is also dependent on the community, its richness, and the domi-

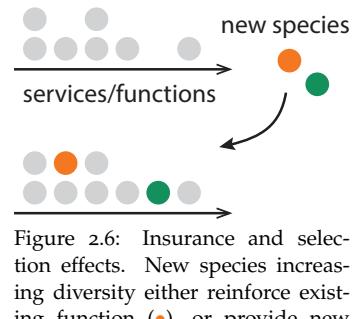


Figure 2.6: Insurance and selection effects. New species increasing diversity either reinforce existing function (•), or provide new function (•)

nant species. The abundance of highly productive species, characterised by high nutrient content, fast-growing and exploitative strategies is responsible for most of a community productivity. Nevertheless, it is hard to disentangle the link between the productivity of the habitat and the productivity of the species living this habitat.

The productivity has another ambiguity: it is both a property of the system and a service. It's a property, and is important in ecosystem services assessment as some services will scale with productivity (*e.g.* carbon storage). But it is also a service, it measures fodder production in grasslands, but wood production in forests, etc...

Productivity is at the same time a property of the habitat, of the community and a service. While the role of abiotic factors is prominent, the effect of the dominant species and the community structure on productivity should not be ignored.

2.3.4 Trade-offs in ecosystem properties

Traits can be linked to ecosystem services by statistical framework ([lavorel_how_2014](#)). But, in the same way there are trade-offs between traits, the ecosystem services provided by an ecosystem are also constrained. Understanding these trade-offs and the dynamics of the community dynamics allows to capture these trade-offs between services bundles (Lamarque et al. 2014). This link should encourage ecologist to focus on the development of methods to link drivers of ecosystems to community dynamics, to predict changes in ecosystem services (see figure 1.4 in chapter I).

In addition to facilitate the study of the effect of abiotic conditions and biotic interaction, functional traits can be used to describe the community and its main properties to evaluate ecosystem services. Statistical links that can be used to determine these links, and research effort should better focus on the dynamics of grasslands communities and the changes in main properties.

2.4 Modelling diverse plant community

Modelling mainly consist in deciding what is important considering and worth representing. The choice of how an entity or a mechanism is represented also correspond this decision making. While considering vegetation community the choice can be on the resources needed, the type of perturbation, or the part of the life cycle of most importance. For vegetation models for the study of community properties and dynamics, the representation of the interactions of multiple species is key. The strategy-space concept offers a great solution to both the interactions and the diversity of species, while also informing the modellers of the communities' properties.

2.4.1 How strategy space open vegetation modelling

In a mechanistic model with multiple species, strategy-spaces are simplified ways to define multiple species. Species identity is fully defined by its position in this space of species-specific parameters. This is a great advantage compared to traditional approaches of vegetation models that rely on strong knowledge

THE POSITION MAKES THE SPECIES

of represented species. Because mechanistic models function with shared biological and ecological processes, the differences of behaviours between species emerge not from the functions but from the species-specific parameters. Therefore, to properly model a species behaviour, in addition to having properly modelled the processes, all species-specific parameters for all species must be determined. This step requires a large investment of time and resources and is proportional to the number of species. Strategy space based on trade-offs enable the representation of multiple species, in a closed environment. A greater effort is required to establish such strategy space, as it needs identification of strong trade-offs and the delimitation of ranges along the axes of strategic differentiation. But once established, an infinity of species can populate this robust space without the threat of Darwinian demons. This subject is further discussed in the following chapter (chapter III, section 2). Because of that strategy space are great tools to consider a diversity of species, when the identity of species is not of primary interest.

While I have no knowledge of living species being projected on a strategy space used in a simulation model, it can be imagined with a projection of measured traits on traits used in the model (even if there can be some discrepancies between the two spaces), in the way of Pierce et al. 2013.

Dynamics Global Vegetation Models tend to use such strategy spaces to model high diversity with a limited number of traits. These models, mainly represented by the model of Kleidon and Mooney 2000, and extensions Reu et al. 2011; Pavlick et al. 2013. They use 12 to 15 traits in their strategy space. These traits can be grouped in: allocation traits, tolerance to climatic conditions, resource efficiency, reproduction strategy and tissue turn-over. All these traits are linked to trade-offs in the formulation of the model. A general observation we can make is that these trade-offs often take the form of greater growth or efficiency against greater resistance to stress. This is similar to observed strategies in drought environment (Kooyers 2015). These models (Reu et al. 2011; Pavlick et al. 2013) demonstrate the ability to capture diversity and climatic response patterns, better than plant functional types, with a limited number of traits.

IN DGVMs

Such approaches are also used to study more specific mechanisms like fire perturbations (Scheiter, Langan, and Higgins 2013). In this case, specific traits are included. The adaptive value of the traits in such frameworks thanks to the inclusion of genetic optimisation processes. This kind of approach is a first step in the understanding of the effect of drivers on community property responses. However, the large scale of these models often does not allow to look at small scales interactions and dynamics.

IN IBMs

Individual-Based Models are great tools to model community dynamics incorporating local interactions and small-scale dynamics. Because they are interested in smaller systems, IMBs often do not use strategy-spaces and prefer species-specific parametrisation (lohier_analyse_2016; Soussana et al. 2012; Taubert 2014). This is often explained by the focus on heavily manage grasslands with objectives of productivity that need precise predictions and model a limited number of species. But strategy spaces are used in IBMs to understand diversity patterns in diverse systems such as savannahs (Reineking et al. 2006) or forest (Falster et al. 2016). These approaches successfully describe the diversity and encourage us to use such strategic differentiation spaces.

Higher diversity can be achieved in these models, but numerous species can be discarded. The benefit of a smooth continuum in strategies is that it avoids strong dominance and shifts. Also, the perception of finer changes in the community is possible, while small default in species parametrisation of species centred models could lead to either no shift (one species dominates and is not sensitive to drivers) or drastic responses (the shift in dominance is abrupt because of no intermediary species).

2.4.2 How models inform us on properties and dynamics

reineking: the shift in dominance along a gradient maire : trade-offs
reu ?

schafter: value of a trait context dependant.

The use of strategy spaces in models allows the representation of high diversity in a common plant functioning framework requiring a limited number of parameters. Such approaches are very useful to follow the dynamics of communities in a mechanistic framework. Individual models tend to ignore such simplifications procedure and relies on the direct measure of traits of interest because they generally integrate a limited number of species. IBMS can take advantage of trade-offs and simple strategy spaces to model diverse communities at small scales while keeping biological mechanisms at their core. However, model-based of strategy space tend to consider mean individuals and ignore the individual variations.

3 THE IMPORTANCE OF PHENOTYPIC PLASTICITY AS A SPECIFIC CASE INTRA-SPECIFIC VARIABILITY

3.1 Intra-specific variability change the rules

3.1.1 Increasing interest in intra-specific variations

Trait approaches emergence lead to a better understanding of general patterns of community responses to drivers and of trade-offs in plant functioning. But with the accumulation of large trait databases, the importance of **intra-specific variability** could not be ignored.

The extent of the intra-specific variation is a big question as some ecologists point out because trait-based approaches make sense only if inter-specific differences are greater than intra-specific differences. While this can be discussed, high functional variability within the species would weaken theories and generalisation based on mean traits. Violle et al. 2012 suggest that the extent of within-population variability relatively to within-community variability should be considered and avoid mistakes in the estimation of coexistence mechanisms. EXTEND

Ignoring intra-specific variability lead to underestimation of niche overlapping, plastic response to neighbours or the fraction of resource a species can used. Multiple studies focused on the extent of functional intra-specific variability (Cécile Hélène Albert, Thuiller, Yoccoz, Soudant, et al. 2010; Cécile Hélène Albert, Thuiller, Yoccoz, Douzet, et al. 2010) and how to disentangle this variability from species turn-over (Lepš et al. 2011) in community response. These studies show contrasting results between traits and levels. Cécile Hélène Albert, Thuiller, Yoccoz, Douzet, et al. 2010 demonstrate a within-species variability explaining between 20% and 40% of total trait variance, and Siefert et al. 2015 note similar levels, but this fraction tends to decrease with the increasing community diversity. They also show that the strategic differentiation between exploitative and conservative species is robust to these variations. It appears that all traits are not variable to the same degree and traits like SLA, height, LNC and LDMC are relatively variable while leaf morphology traits variability is lower Siefert et al. 2015.

The variability of multiple traits certainly impacts the functional diversity (Bello et al. 2011; Cécile H. Albert et al. 2012). All indexes are not sensitive to the same degree, with single trait measure being the most sensitive, but should be used carefully to draw an interpretation of ecological pattern linked to functional diversity. To overcome this difficulty and disentangle the effects of the different forms of functional diversity specific indexes are developed (Bello et al. 2011).

The relative extent of intra-specific variability depends on the trait, spatial extent, and species richness, but not on climatic conditions (Siefert et al. 2015) suggesting general mechanisms

The fact that some traits are variable, while others are not, implies that some mechanisms structure this variability. A way to identify such effects is to look if variability is structured along environmental gradients, suggesting adaptation mechanisms.

Along such gradients trait variability for traits like SLA (H. Poorter, Niinemets, et al. 2009) of leaf mass fraction (LMF) (H. Poorter, Niklas, et al. 2012) follows similar patterns as inter-specific response (Niinemets 2001), with increasing SLA along precipitation and temperature gradient, and decreasing SLA along radiance gradient (leaf mass fraction shows similar responses). These responses suggest strong constraints (similar to the ones that shape inter-specific differences) shaping this variability. However, species may vary in their response (Kichenin et al. 2013). This contrast can be explained by differences in position around a bell-shaped response curve around the optimum (see Cécile Hélène Albert, Thuiller, Yoccoz, Soudant, et al. 2010 for more details). Kichenin et al. 2013 argue that it is not the case because alongside a wide altitudinal gradient the response curves observed for any trait or species are not bell-shaped.

This additional level of variability is not always in the same direction as community response driven by turn-over (Cécile Hélène Albert, Thuiller, Yoccoz, Soudant, et al. 2010; Kichenin et al. 2013; Jung, Cécile H. Albert, et al. 2014) leading to difficulties to predict the response of the community. These levels must be disentangled, in order to do that, mechanisms underlying intra-specific variability must be understood. This is particularly important because they have multiple effects on how we model community dynamics and understand coexistence mechanisms Bolnick et al. 2011; Viole et al. 2012.

After the emergence of trait-based ecology and its high potential, the re-

cent focus on intra-specific trait variability questions the strength of mean species approaches. While intra-specific variability does not negate numerous conclusions from previous work, because of its large extent and how it alters functional diversity, its effects on community dynamic processes must be interrogated, and underlying mechanisms investigated.

3.1.2 Contrasting effects of intra-specific variations

Intra-specific variability impacts coexistence mechanisms and community properties in multiple ways. The following paragraphs are not an exhaustive list of all ways intra-specific variations affect community properties or coexistence mechanisms, but a few contrasting examples to emphasise the need for better identification and understanding of underlying mechanisms.

Hart, Schreiber, and Levine 2016 use a mathematical model to investigate the impact of intra-specific variations on coexistence. They demonstrate the negative effect of intra-specific variations by the intermediate of Jensen inequality effects, that leads to an under-estimation of competitive dominance because of the non-linearity. I argue that this problem is partially solved if the estimation of the mean parameter is empirical, in this case, the variability is taken into account. Moreover, this certainly applies to genetic variations. Indeed, the plastic response would emerge if there are changes in conditions, changes that certainly affect the competitive interactions. Such change (justifying a change in phenotype), is likely to alter the competition relationship as modelled. This supposition is supported by the fact that variations away from the mean are generally due to a distance from the optimum of the niche, and therefore, certainly a lower fitness and lower competitive ability.

JENSEN INEQUALITY

ISV also effect of abiotic filtering

NICHE

effect on realised niche

neighbours: avoid or increase competition

specifically on diversity

CONTRASTING EFFECTS

callaway 2003 from competition to facilitation.

Bolnick et al. 2011 (Hart, Schreiber, and Levine 2016) (Courbaud, Vieilledent, and Kunstler 2010) (Turcotte and Levine 2016) (Roscher et al. 2015) (Valladares et al. 2015) ([barabas_effect_2016](#)) (Jung, Violle, et al. 2010)

The intra-specific variability has been observed to be an important part of community functional diversity, but also a way the community responds to changes in conditions. In addition to the empirical evidence of this importance, theoretical approaches support contrasting effects of such variations on coexistence mechanisms, evolutionary processes and community responses to climate event or invasion. It is crucial to disentangle different sources of intra-specific variability in order to understand potential effect on ecosystem dynamics.

3.1.3 Beyond the mean and the bell-shape: towards more mechanisms in representing intra-specific variability

There is a difference between how we observe ISV, and why it emerges. What is random? Therefore it is ... not good ... to apply such simplification of random effect onto theoretical models to predict the effect of intraspecific variability

...

with strong assumptions (observed on the functional trait in wide spatial range, applied to interactions in homogeneous context) on how they translate onto interactions (done in Halt, check Bolnick) Cécile Hélène Albert, Thuiller, Yoccoz, Soudant, et al. 2010 bell shape intra-specific response pattern along gradient, but doesn't stand according to Kichenin et al. 2013. Depends on trait and gradient... cannot assume that, need real quantification. ok if not a gradient response (or gradient is not known). Bell shape can emerge from non-measured gradient with linear response.

Dewitt and Barabas.

The same way the neutral theory is simplifying and brings little understanding to underlying processes and relies on strong hypothesis, considering intra-specificity as a purely random mechanism is insufficient.

Bell shape does not appear in altitude gradient... inconsistencies between theory and empirical data

Strong theoretical hypothesis

refer to asymmetric and symmetric competition

If most of the changes are plasticity or selection: it changes the effects on interactions and niche.

What are the possible effects? probably it does not affect interaction like (Hart, Schreiber, and Levine 2016) supposes (even if they talk about variations, their conclusions may not be extendable to plastic variations). May change a lot the balance between abiotic filtering and biotic filtering.

– go to individual mechanisms, evolution could tackle genetic variations, physiology and ecology on ontogeny, and evolution and ecology on phenotypic plasticity

Simple approaches to intra-specific variation constitute an improvement over mean approaches as they highlight processes ignored until now. However such approaches overlook the structure of the variability and underlying processes, leading to simplistic representations and potentially misinterpret the role and effect of this variability.

Ecology shifted from species to traits syndromes with great success, but the intra-specific variability constitutes a great challenge for generalisation of observed patterns. By overlooking the processes that structure intra-specific variations, we might lose the capacity to properly interpret the role of variability and refine our understanding of community functioning. The complexity of living communities requires to go further down and consider the individual scale. This is made possible by the accumulation of more and more numerous and detailed data, the improvement of statistical and new simulation tools. The question of the sources and drivers of intra-specific functional variability seems crucial to rise to the challenge it issues.

3.2 Phenotypic plasticity: a specific case of intra-specific variability

Until now, the processes at the origin of intra-specific variability has not been discussed, but to understand how it can alter community properties it is necessary to differentiate the different sources of intra-specific variations as they work

in different ways.

3.2.1 The different sources of intra-specific variability

Intra-specific variation can be caused by two mechanisms: genetic variation and phenotypic plasticity. Genetic variation occurs when individuals from the same species have different genotypes, leading to different phenotypes. On the other hand, phenotypic plasticity implies that the same genotype can lead to different phenotypes. Plasticity can involve epigenetic mechanisms (beaman_evolution_2016; Zhang et al. 2013; Adrienne B. Nicotra et al. 2015) that blur the frontier between the two forms of intra-specific variability as epigenetic is an inheritable form of plasticity. It is transmitted to descendants but unlike genetic mutation is reversible. To keep things simple, epigenetic phenomena will not be discussed here.

Genetic variability (as well as epigenetic) can be detected in case of origin specific response, while if the variability is explained by the treatment, it is a plastic response (Frei, Ghazoul, and Pluess 2014), and a large fraction of the variability observed in grasslands species is a plastic response rather than genetic variation alone (Frei, Ghazoul, and Pluess 2014; Merilä and Hendry 2014).

A. B. Nicotra et al. 2010 provide a good review of plasticity mechanisms and the importance for the adaptation to climate change. They advocate plasticity in functional traits should be considered in mechanistic models as they may play a central role in the speed and adaptiveness of community response to climate change.

Intra-specific variability can be decomposed in two main types: genetic variability that seems to be closer to random processes envisioned in simple models of intra-specific variability, and phenotypic plasticity that specifically links variations of phenotype to differences in external conditions. These mechanisms of variations are under the control of both evolutionary and molecular processes, that need to be better understood to be disentangled and to better predict their effects on community dynamics.

3.2.2 What is phenotypic plasticity?

Plasticity is a source of intra-specific variability, but biological processes leading to changes in phenotype can be complex. These paragraphs try to disentangle the different forms of plasticity and the underlying mechanisms.

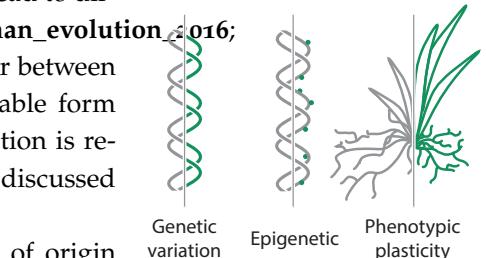


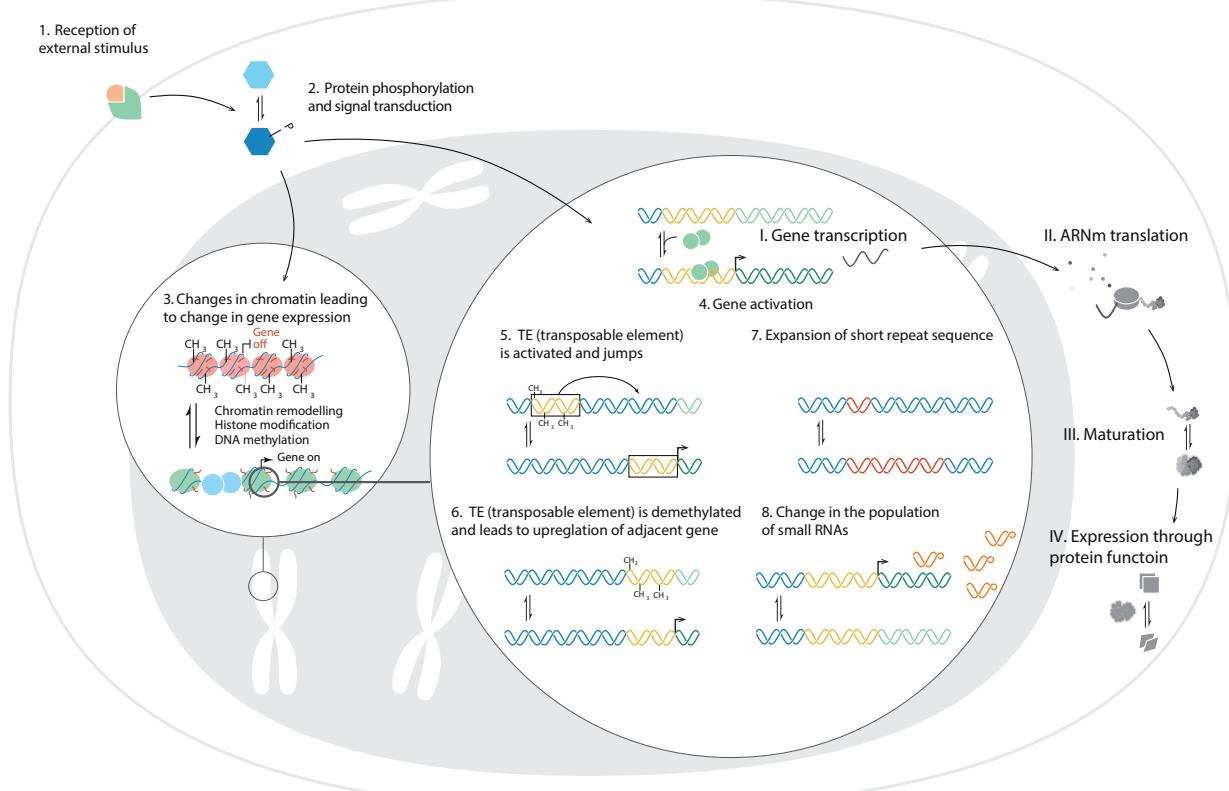
Figure 3.1: The three main sources of intra-specific phenotypic variability: genetic, epigenetic and phenotypic plasticity. Phenotypic plasticity can involve epigenetic mechanisms.

Molecular basis of phenotypic plasticity

The phenotypic plasticity lies both in the perception of external conditions through sensor organs and signalling pathways (auxin pathway for light, root gravitropism for gravity ...), and the integration of this information to alter the development plan. This integration must be coordinated at the scale of the plant according to rules or objectives, which partly explore in this work, but ultimately is applied at the cell levels.

Because of the complexity of these pathways and our partial understanding of these mechanisms, we will not attempt to model them. However, I hope that this little overview of molecular mechanisms at the scale of the cell will give the reader an idea of the processes behind the abstract concepts used in this manuscript.

The processes of information gathering (through specific organs, cells or organelles) and integration of this information finally leading to changes in the phenotype visible at the macro-scale result from similar events at the cell scale. The external signal is captured by a specific receptor at the cell membrane (1), then integrated through phosphorylation cascade (2) leading to numerous alterations of the gene expression sequence (I to IV) because of regulation mechanisms (3 to 8). These regulation mechanisms are diverse, from chromatin changes (3) modifying the accessibility of certain genes, to other gene regulation processes (5 & 6) or post-transcription regulations (7 & 8).



Phenotypic plasticity is the effect of environment on the link between genotype and phenotype. Plasticity can itself be decomposed in active plastic response that change the internal status of the individual (under genetic control) and passive response that results from the inevitable effect of the environment of the traits on the individual.

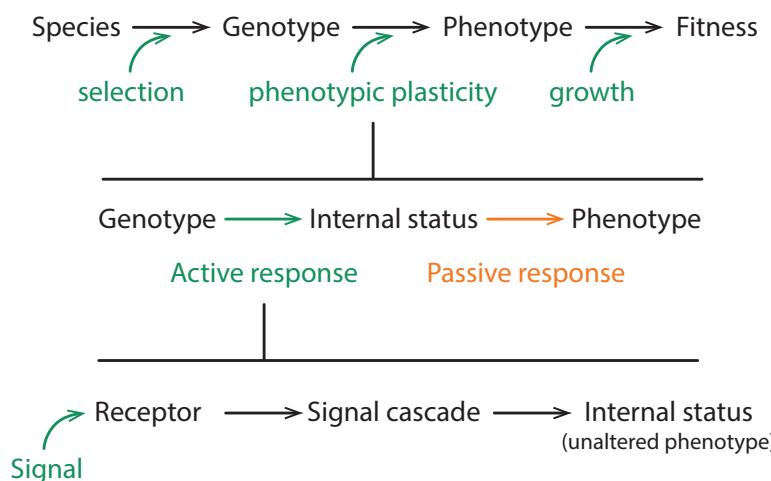
These regulations of gene expression affect the plant behaviour and development. These regulations are reversible (their effect may not be reversible) but can also be inheritable (*i.e.* 6). The type of regulation depends on the targeted genes, the duration of the regulation, and other factors. This multiplicity of regulatory processes at the scale of cells, in addition to the interconnectivity of genes, signalling pathways and tissues interactions, demonstrate an extraordinary potential for the regulation of both functioning and phenotype of plants. Therefore it seems that the molecular basis does not limit the plasticity, but it is rather the difficulty to anticipate the future and to define the best strategies that limits the benefits of phenotypic plasticity.

The diversity of mechanisms and scales (both spatial and temporal) these processes can act inside of plant gives an idea of the diversity of strategies a plant can deploy to face changes in its environment. Considering this complexity, only a small fraction can be explored in such model as *MountGrass*, but hopefully, it will help make progress in our understanding of the role of these molecular mechanisms at the scale of the community.

Phenotypic plasticity is the capacity of a species to produce individuals with the same genotype but different phenotypes. This difference in phenotype should be an active process, not the results of direct alteration of the phenotype by external factors without changes in internal functioning. This change in internal functioning process has the objective¹ to match the phenotype with expected future conditions to maximise the individual fitness. The expression "expected future conditions" is key here, as it is this projection that drives the plasticity.

Active plasticity is used for predominantly anticipatory, and often highly integrated, phenotypic changes in response to some environmental cue or signal, and reflect modifications of developmental pathways and regulatory genes. Forsman - 2014

Passive plasticity, on the other hand, may stem from direct environmental influences on chemical, physiological and developmental processes, and is generally not considered anticipatory, but a mere consequence of the environment, such as stunted growth owing to low resource levels.



FORMS OF PLASTICITY

¹ in the sense it has been selected because it provides this capacity

Figure 3.2: Decomposition of phenotypic plasticity as a step between the genotype and the fitness. Phenotypic plasticity is the effect of environment on the link between genotype and phenotype. Plasticity can itself be decomposed in active plastic response that change the internal status of the individual (under genetic control) and passive response that results from the inevitable effect of the environment of the traits on the individual.

Active and passive plastic response can be discriminated by the position of the control: internal for the active plasticity, or external for the passive response. In the case of active plastic response, the signal from environment must be integrated (from physical or chemical to information) then transferred to response organs. These organs respond to the integrated signal by changes in their expression levels (*internal status* in figure 3.5) as summarised in figure 3.3.

Changes in phenotypes are controlled mainly by changes complex development processes. These processes involve numerous proteins and signaling pathways. Genes expression of proteins (transcription factors, enzymes, signalling proteins...) is controlled by specific mechanisms with various degrees of speed and duration (instantaneous regulation response, to inherited epigenetic adaptation). Some of these molecular processes are detailed in box ?? in relationship with gene expression pathway (see also A. B. Nicotra et al. 2010).

Active phenotypic plasticity is an integrative process at the scale of the individual that aims for an improvement of plant fitness by the adjustment of its morphology according to environmental cues. It often relies on multiple regulation processes. Modelling the extent and the rules of such mechanism

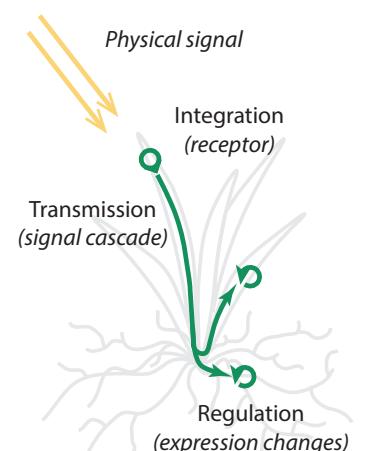


Figure 3.3: Mechanism of active plasticity. Integration of a physical (or chemical) signal, transmission and regulation of phenotype through regulation of gene expression, or post-transcription regulations.

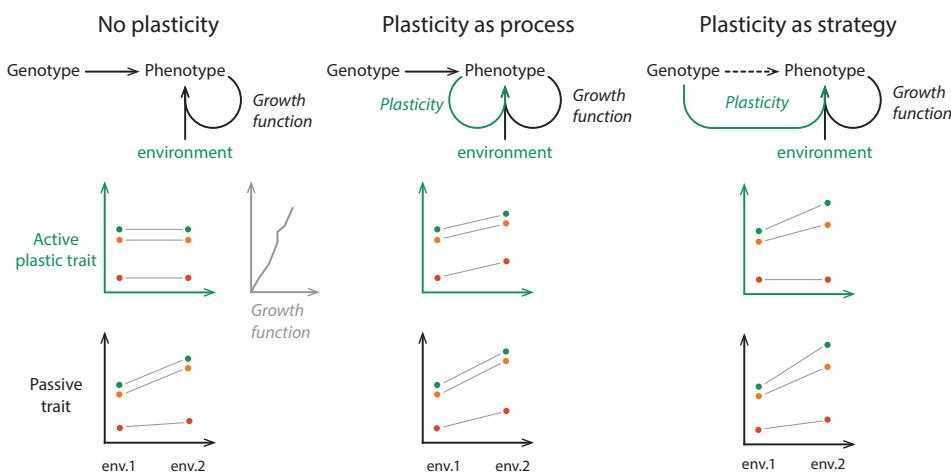
is not an easy task that might depend on the context and the framework used.

3.2.3 How to model phenotypic plasticity

A plastic response can involve numerous genes interaction in networks of regulatory pathways. The objective of an ecological model is not to reproduce this complexity, but the basic behaviours emerging from this biological complexity¹. The basic components of the active plastic response are the perception of the external signal, its integration into meaningful information and the transformation into phenotype modification.

Every growth model is plastic. Every growth model predicts different phenotypes for plants sharing the same phenotype (often just defined by the species affiliation) growing in different condition. But most of this plasticity is passive, and it could be encompassed in this personal definition of the notion of **growth function** (see figure ??). However, among vegetation models only some of them claim to include phenotypic plasticity (Maire et al. 2013), why so? What criterion can be used to distinguish active from passive plasticity in the context of plant modelling?

The use of information from the environment to change the phenotype in order to have a better fitness is active plasticity. But in practice (in models)(Maire et al. 2013), often nothing really separates the two as plasticity is often modelled as a general mechanism shared by all species (but see **jablonka_adaptive_1995** for discrete strategies in clonal plants) and local environmental variables are used to determine the phenotype of a plant in both cases. Only the justifications and the forms of the linking functions are different, and they may involve different traits. This idea is illustrated in figure 3.4, where the phenotype is first defined by the genotype then controlled by the growth function as a function of current phenotype and environment (see figure 3.4, left column). There are no differences between plasticity of two species if two species have the same phenotype, then in a similar environment, they would express the same plastic response. I argue that plasticity, to be considered as an active process, should be under a genetic control (*i.e.* species-specific parameter). This means that, despite a shared rule and similar phenotype, the plastic would be different and would depend on a species-specific parameter.



¹ this biological complexity can be explained by the simplicity and a limited number of basic biological units living organism are made of, and the emergence through a simple mutation-selection operation. This complexity can be mimic by simpler and freer mathematical design.

REFERENCE AND PLASTIC TRAITS

Figure 3.4: Three forms of plasticity in models.

Moreover, no integration function,

Two questions emerge from this: if growth function and plastic response are different (conceptually), how to determine each of these functions? How the genetic control affect the phenotypic response? Or why would it be beneficial to have multiple rules - non-discrete perspective on plasticity?

we want active plasticity, what's distinguish plasticity types is not the frame of reference, but the strategy: it's a choice: determine by another trait that characterises the response.

what makes it plastic: find the invariance. Laughlin? (what's invariance anyway)

defined by variable (ref plasticity is good in variable) and limiting variables (resource, temp, perturbation) = drivers

PLASTICITY RULES: A QUESTION OF DRIVERS

functions: reaction norms (Feller et al. 2015) thickness and light. Nice, but doesn't work with multiple drivers and composite traits

Or general rules : give an optimum phenotype. More deterministic approach. But optimum = general rule. Conceptually, if there is an optimum: why do something else ? Empirical studies: can be maladaptive, plasticity is a bet, and often ignore

resources, but also risks (frost, grazing): alter cost and gains. Multi-process plasticity, with relative weights. see chapter VI, section 3.

Or perfect optimisation with perfect estimation of conditions.

3.3 Toward an integrative framework of plant strategy and phenotypic plasticity

Adaptive plasticity in models is often a layer on top of the species strategy, it acts more like a new mechanism, rather than a strategy within the already existing growth process. To interrogate the plasticity as a dimension of plant growth and an evolutionary process (Bradshaw 1965) (see also work of Scheiner ([scheiner_genetics_2002](#) ; Scheiner and Lyman 1989; Scheiner and Holt 2012)), or better understand the cost and limits of plasticity DeWitt, Sih, and Wilson 1998; Callahan, Maughan, and Steiner 2008; Auld, Agrawal, and Relyea 2009, or the effect of plasticity on coexistence and community dynamics Hart, Schreiber, and Levine 2016, plant strategies and plasticity need to be blended together in an integrative framework.

3.3.1 Plastic strategies

Resource-use and allocation strategies have been related to environmental conditions in both empirical (Ian J. Wright and Westoby 2002; D. Ackerly 2004; L. Poorter and Bongers 2006), conceptual (Grime 1977; Westoby 1998) and modelling(Kleidon and Mooney 2000; Scheiter and Higgins 2009; Reineking et al. 2006) studies. Moreover, functional traits show evidence of intra-specific changes along environmental gradient (Kichenin et al. 2013) and intra-specific economic spectrum (Hu et al. 2015), and constraints that shape main ecological trade-offs are certain to also constrain individual traits. Therefore, if strategies vary between and within species along environmental gradients, it makes sense to

imagine that plasticity as changes in strategic traits. This goes beyond changes in spatial allocation(Schapendonk et al. 1998), or parameters not identified as strategic traits Lohier et al. 2014; Feller et al. 2015. Considering strategic traits is not common practice because it blurs the limits between species that are not well identified by these traits any more¹.

However, while this interpretation makes sense, the species and the individuals do not have the same constraints, and plasticity cannot be as large as intra-specific diversity as there are limitations to plastic development (DeWitt, Sih, and Wilson 1998; Auld, Agrawal, and Relyea 2009). Moreover, it seems that rules that drive plastic may not be the same as the ones that drive intra-specific genetic variations and inter-specific differences(Peter Ryser and Eek 2000), explaining contrasting response along gradient or between experimental drought treatment (Kichenin et al. 2013; Jung, Cécile H. Albert, et al. 2014). This difference is probably more important for grass species than trees (Franklin et al. 2012) because of a lower scale difference between growth and selection processes.

¹ especially when a relatively low number of species-specific traits are considered

Phenotypic plasticity tends to maximize resource acquisition and growth rate in the short term, whereas the higher tissue-mass density and the longer leaf lifespan of shade-tolerant species indicate reduced loss rates as a more advantageous species-specific adaptation to shade in the long term. - Peter Ryser and Eek 2000

3.3.2 Plasticity as a strategy

Most models consider plasticity in traits or carbon partitioning as a general behaviour that is present or absent for all considered species. While this discretisation of the phenomenon is not problematic, and rather informative for a single plant or monoculture simulations Maire et al. 2013, it ignores the question of the adaptive value of plasticity and does not allow a continuous representation of plasticity.

Intuitively phenotypic plasticity is a mechanism that increase fitness and has a positive adaptive value (increases the chance to be selected). However multiple costs and limits have been identified, both biological (**dewitt_cost_1998**; Auld, Agrawal, and Relyea 2009; Callahan, Maughan, and Steiner 2008) and ecological (**dewitt_cost_1998**; **scheiner_genetics_2002** ; Auld, Agrawal, and Relyea 2009; Scheiner and Lyman 1989; Scheiner and Holt 2012; Van Kleunen and Fischer 2005), limiting the extend of plasticity observed in nature and differences between species (in grasslands see Peter Ryser and Eek 2000).

...

COST AND LIMITS

These limitations, in addition to indicating the processes that should be included in dynamic models involving phenotypic plasticity, show that plasticity should be continuous. Indeed, costs of plasticity can increase with the amplitude of the plastic response and/or the complexity, therefore reducing the adaptive value of plasticity. Because non-linearity can be expected between the amplitude of plastic response and both fitness increase and cost, the adaptive value of plastic response can switch from positive to negative depending on its amplitude. Such behaviour would justify a non-discrete plastic response (or variable sensitivity for polyphenism) to be captured in a model.

CONTINUOUS PLASTICITY

As mentioned, ecological processes can favour or limit the selection of plasticity as any other trait. The idea of plasticity as a trait under genetic control is not

FROM PROCESS TO STRATEGY

new. Anthony Bradshaw was probably the first to defend this idea of genes controlling the variability of phenotypes.

But it is rarely implemented in individual or community growth model. This can be explained by the fact that plasticity is often seen as a process, rather than a strategy (see the previous paragraph). In individual-based models, plasticity as a process is often considered because of the relatively low number of species, and scientific question not focusing on ecological aspects. In models that consider the dynamics of diverse communities under drastic changes, integrating the plasticity as a strategy is crucial. This can be done by the use of species-specific traits that control the amplitude and/or direction of the response (see more details in chapter III). In population models, plasticity is often considered as a source of variation equivalent to intra-specific genetic variations and is modelled by a distribution function. DeWitt 2016 proposes approaches with higher moments and environment dependent distribution to integrate plasticity into such models. In development models, Bayesian models offer a unifying framework to combine inherited information and environmental cues (Stamps and Frankenhuus 2016).

This shift is also important because if genes control plasticity, plasticity can also alter evolutionary process and therefore the response to climate changePfennig et al. 2010; Matesanz, Gianoli, and Valladares 2010; A. B. Nicotra et al. 2010.

Plasticity is a complex matter, both a growth process that alters strategies and a strategy itself. New simulations tools for understanding community dynamics should try to both include multiple coexistence mechanisms and plant strategies, and focus on individual level mechanisms of competition, growth, and survival. This can only be achieved in a constraint high dimensional strategy space based on physical and biological trade-offs. Individual-level modelling allows the integration of multiple sources of intra-specific variability: genetic diversity and phenotypic plasticity. Phenotypic plasticity being driven by the perception of the environment, it cannot be simply described by normal random distribution and should receive more attention. This focus is particularly important considering both the lack of understanding of this phenomenon and the consequences for plant communities.

3.4 How phenotypic plasticity affect ecosystem properties and dynamics

The difficulty to model phenotypic plasticity, more precisely to integrate multiple aspects of the complexity of phenotypic plasticity in the context of community dynamics, is limiting the current knowledge of the impact of this mechanism on community composition, properties, and dynamics under global change. In this paragraph, I try to identify the mechanisms by which phenotypic plasticity impacts plant communities, and to determine if there are unresolved questions or paradoxes, or incomplete conclusions. The focus will be given to the main properties of the grassland communities: diversity, productivity, and identity.

3.4.1 Contrasting effect on diversity

Diversity is a complex subject as discussed earlier in section 1, resulting from various processes and measured by many indicators. Therefore, there are many ways the plasticity can affect diversity. Also, the scope at which diversity is considered may change the effect of plasticity as the balance between may driving mechanism is shifted (see Chalmandrier 2015 for the importance of the scale of diversity). I will try to keep it simple and focus on measures of diversity at the scale of the community.

Species diversity is driven on two levels, at large scales by abiotic conditions and filtering, and at a lower scale, within this large potential niche defined by abiotic conditions, by competition and facilitation interactions. From this point of view, plasticity certainly increases the potential niche both along environmental conditions axis, but also along variation axis (species might be more or less sensitive to changes in conditions), therefore enlarging **niche** superposition (Violle et al. 2012). This effect should, in theory, increase potential diversity as more species can potentially live in any given environment (Lepik, Liira, and Zobel 2005; Jung, Cécile H. Albert, et al. 2014), but the effect of biotic interactions must be considered before drawing any conclusion of the effect of plasticity on realised diversity. The effect of plasticity on interactions is much harder to predict. According to **adler_coexistence_2007** increase in niche difference and decrease in average fitness differences would increase stable coexistence.

The impact of plasticity mechanism on stabilizing effect is also hard to anticipate. It will likely be negative because established species may better fill any potential gap and prevent low-density positive effect and therefore invasion (Berg and Ellers 2010). On the contrary, reduction of fitness difference due to plasticity could lead to stronger coexistence between species. Yet, the reduction of fitness differences is not guaranteed and in case of asymmetric gain (relative to strategies), plasticity could reduce realised diversity by increasing competitive exclusion. There are here multiple effects (figure ?? on species diversity that needs to be disentangled. Recent review (Turcotte and Levine 2016) of these effects show no consensus on the effect of phenotypic plasticity on stable coexistence.

SPECIES DIVERSITY

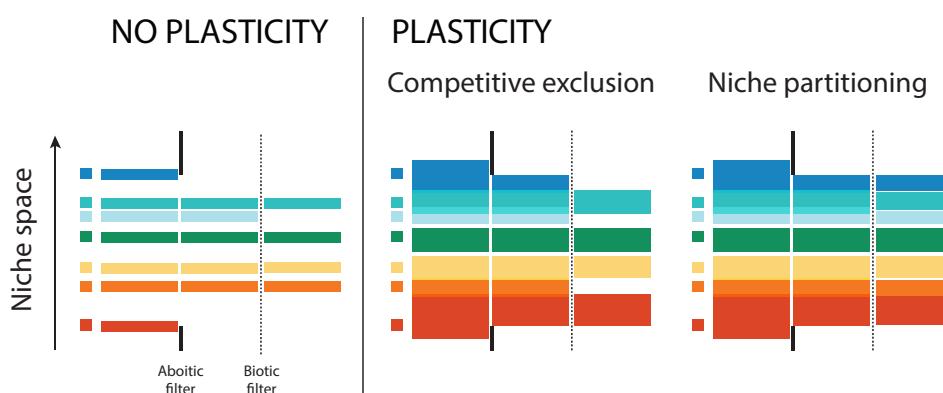


Figure 3.5: Phenotypic plasticity can affect filtering processes in diverse ways, making difficult the understanding of the role of plasticity in diversity maintenance.

But plasticity responses not only depend on abiotic condition, but also on the neighbourhood that affects local environment (Sultan 1995) at a fine scale. Because of plasticity, these interactions can even shift from competition to fa-

cilitation Callaway, Pennings, and Richards 2003. A novel difficulty arises with the evidence that the identity of the competitor affects plastic response (Callaway, Pennings, and Richards 2003; Abakumova et al. 2016), but it is likely that such interaction is related to traits and therefore impact on resource (Callaway, Pennings, and Richards 2003).

Species diversity often comes with functional diversity, however, phenotypic plasticity affect plant traits and is likely to affect functional diversity (Cécile H. Albert et al. 2012). Plasticity can lead to a convergence or a divergence of functional traits, decreasing or increasing functional diversity. In an experiment with legumes species Roscher et al. 2015 observed these two phenomena on different types of traits, between monoculture and mixture. The convergence of canopy filling and vertical growth traits suggests that competition stresses the different species on light competition, leading to a reduction of working strategies along these dimensions. Whereas, relatively, the other aspects of plant development are less constraint, or species experience diverse and contrasting conditions in mixtures than in monocultures.

FUNCTIONAL DIVERSITY

Phenotypic plasticity is expected to increase the potential niche of species and reduce the filtering effect of abiotic conditions. However, the effect on biotic interaction makes no consensus and is likely to vary depending on the identity of the competitors, and the relative effect on trait differences. The balance between stabilizing niche differences and average fitness differences is crucial to determine the final impact on stable coexistence. The effects on functional diversity are also diverse but mainly depends on the plastic rules leading to convergence or divergence of traits.

3.4.2 Productivity always improved?

There is still debate on the effect of phenotypic plasticity of mechanisms driving species diversity, but is the question of the effect on productivity solved?

Plasticity is a mechanism that emerges in a situation where the plants can increase their fitness in response to environmental conditions. This increase in fitness is often due to higher resource use or resource foraging efficiency and therefore better growth rate (observed in models (Maire et al. 2013) and empirical studies (Hamann et al. 2016)). This leads to higher individual productivity. It is especially true when resources are varying and these variations can be anticipated Richter et al. 2012.

STABILITY

However, has mentioned earlier, plasticity comes with inherent costs, related to the biological machinery needed to sense and process the signals and alter the phenotype. This costs, if the plant does not take advantage of the plasticity (no variability, in its niche) to increase (or maintain) growth rate will impact the productivity.

COSTS AND LIMITS

The unreliability of environmental cues is a limit of plasticity, and it can lead to maladaptive changes in phenotypes, but this is a marginal behaviour, and maladaptive plasticity is expected to be eliminated by an evolutionary process in fairly constant conditions. However, in the context of climate change, the reliability of these cues may decrease and leads to maladaptive responses.

If unnecessary costs and unreliable cues can impact overall plant efficiency, adaptive plasticity can also hurt productivity while increasing fitness. Indeed,

as evolutionary models and game theory predict, competition can lead to lower efficiency than optimum arrangement. Competition leading to lower resource availability, plastic species may have an aggressive plastic response leading to a stronger competitor but with less effective resource use.

Biodiversity - productivity

DIVERSITY AND PRODUCTIVITY

3.4.3 Community identity shift

The third main property of grassland communities is the **identity** of the dominant species (or average species if CWMs are considered). Phenotypic plasticity can impact community identity in two ways: (1) by shifting the identity of present species, (2) by altering the output of filtering processes in favour of different traits.

The first effect makes sense only in the context of a change in condition. Drought experiments in mountain grasslands show an intra-specific shift toward higher LDMC and lower SLA (Jung, Cécile H. Albert, et al. 2014). Other empirical studies show uncoupled response between above- and below-ground organs, shifting the strategy of the species ([freschett_plasticity_2014](#)).

A modelling experiment shows that the phenotypic plasticity is required to correctly model the dominance pattern along cutting frequency gradient (Maire et al. 2013), illustrating the second effect.

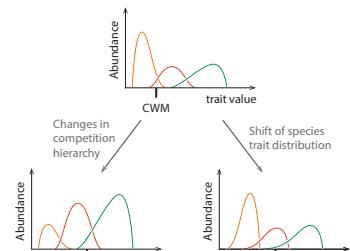


Figure 3.6: Phenotypic plasticity effects on community identity. Phenotypic plasticity can modulate community-identity response to a change in conditions by two distinct but non exclusive mechanisms: altering the competitive hierarchy and species abundances (left) or shifting the individual species identity (right)(Dwyer, Hobbs, and Mayfield 2014).

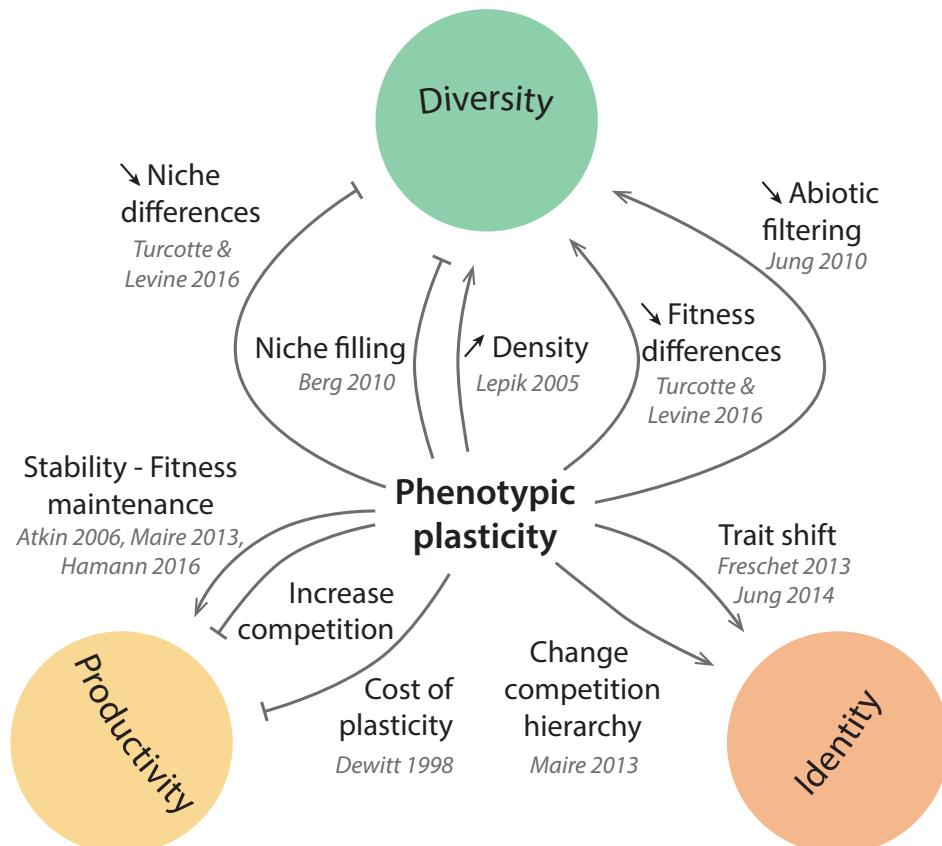


Figure 3.7: Effect of phenotypic plasticity on the three main community properties. Phenotypic plasticity can impact these properties through multiple processes that may have contrasting effects. To determine the overall effect of plasticity on community response to changes in drivers (climate and land-use) we need to integrate all these effects.

3.4.4 Phenotypic plasticity effect on individuals and communities

Plasticity is a complex matter, both with a growth process that alters strategies and a strategy itself. New simulations tools for understanding community dynamics should try to both include multiple coexistence mechanisms and plant strategies, and focus on individual level mechanisms of competition, growth, and survival. This can only be achieved in a constraint high dimensional strategy space based on physical and biological trade-offs. Individual-level modelling allows the integration of multiple sources of intra-specific variability: genetic diversity and phenotypic plasticity. Phenotypic plasticity being driven by the perception of the environment, it cannot be simply described by normal random distribution and should receive more attention. This focus is particularly important considering both the lack of understanding of this phenomenon and the consequences for plant communities.

BIBLIOGRAPHY

- Abakumova, Maria et al. (2016). "Plasticity in plant functional traits is shaped by variability in neighbourhood species composition". en. *New Phytol* 211.2, pp. 455–463.
- Ackerly, David (2004). "Functional strategies of chaparral shrubs in relation to seasonal water deficit and disturbance". *Ecological Monographs* 74.1, pp. 25–44.
- Ackerly, David D. and Peter B. Reich (1999). "Convergence and correlations among leaf size and function in seed plants: a comparative test using independent contrasts". en. *Am. J. Bot.* 86.9, pp. 1272–1281.
- Adler, Peter B. et al. (2006). "Climate variability has a stabilizing effect on the coexistence of prairie grasses". en. *PNAS* 103.34, pp. 12793–12798.
- Albert, Cécile Hélène, Wilfried Thuiller, Nigel Gilles Yoccoz, Rolland Douzet, et al. (2010). "A multi-trait approach reveals the structure and the relative importance of intra- vs. interspecific variability in plant traits". en. *Functional Ecology* 24.6, pp. 1192–1201.
- Albert, Cécile Hélène, Wilfried Thuiller, Nigel Gilles Yoccoz, Alex Soudant, et al. (2010). "Intraspecific functional variability: extent, structure and sources of variation". en. *Journal of Ecology* 98.3, pp. 604–613.
- Albert, Cécile H. et al. (2012). "On the importance of intraspecific variability for the quantification of functional diversity". en. *Oikos* 121.1, pp. 116–126.
- Amarasekare, Priyanga (2003). "Competitive coexistence in spatially structured environments: a synthesis". en. *Ecology Letters* 6.12, pp. 1109–1122.
- Auld, Josh R., Anurag A. Agrawal, and Rick A. Relyea (2009). "Re-evaluating the costs and limits of adaptive phenotypic plasticity". en. *Proceedings of the Royal Society of London B: Biological Sciences*, rspb20091355.
- Bello, Francesco de et al. (2011). "Quantifying the relevance of intraspecific trait variability for functional diversity". en. *Methods in Ecology and Evolution* 2.2, pp. 163–174.
- Bennett, Jonathan A. et al. (2016). "The reciprocal relationship between competition and intraspecific trait variation". en. *J Ecol* 104.5, pp. 1410–1420.
- Berg, Matty P. and Jacintha Ellers (2010). "Trait plasticity in species interactions: a driving force of community dynamics". en. *Evol Ecol* 24.3, pp. 617–629.
- Berger, Uta et al. (2008). "Competition among plants: Concepts, individual-based modelling approaches, and a proposal for a future research strategy". *Perspectives in Plant Ecology, Evolution and Systematics. Space matters - Novel developments in plant ecology through spatial modelling* 9.3–4, pp. 121–135.
- Bolnick, Daniel I. et al. (2011). "Why intraspecific trait variation matters in community ecology". *Trends in Ecology & Evolution* 26.4, pp. 183–192.
- Bradshaw, A. D. (1965). "Evolutionary Significance of Phenotypic Plasticity in Plants". In: *Advances in Genetics*. Ed. by E. W. Caspary and J. M. Thoday. Vol. 13. Academic Press, pp. 115–155.
- Callahan, Hilary S., Heather Maughan, and Ulrich K. Steiner (2008). "Phenotypic Plasticity, Costs of Phenotypes, and Costs of Plasticity". en. *Annals of the New York Academy of Sciences* 1133.1, pp. 44–66.
- Callaway, Ragan M., R. W. Brooker, et al. (2002). "Positive interactions among alpine plants increase with stress". en. *Nature* 417.6891, pp. 844–848.
- Callaway, Ragan M., Steven C. Pennings, and Christina L. Richards (2003). "Phenotypic plasticity and interactions among plants". *Ecology* 84.5, pp. 1115–1128.

- Chalmandrier, Loïc (2015). "From communities to meta-communities : decrypting diversity patterns". PhD thesis.
- Chave, Jérôme (1999). "Study of structural, successional and spatial patterns in tropical rain forests using TROLL, a spatially explicit forest model". *Ecological Modelling* 124.2–3, pp. 233–254.
- Chave, Jerome et al. (2009). "Towards a worldwide wood economics spectrum". en. *Ecology Letters* 12.4, pp. 351–366.
- Chesson, Peter (2000a). "General Theory of Competitive Coexistence in Spatially-Varying Environments". *Theoretical Population Biology* 58.3, pp. 211–237.
- (2000b). "Mechanisms of Maintenance of Species Diversity". *Annual Review of Ecology and Systematics* 31.1, pp. 343–366.
- Clark, James S. et al. (2007). "Resolving the biodiversity paradox". *Ecology Letters* 10.8, pp. 647–659.
- Courbaud, Benoît, Ghislain Vieilledent, and Georges Kunstler (2010). "Intra-specific variability and the competition-colonisation trade-off: coexistence, abundance and stability patterns".
- DeWitt, Thomas J. (2016). "Expanding the phenotypic plasticity paradigm to broader views of trait space and ecological function". *Curr Zool* 62.5, pp. 463–473.
- DeWitt, Thomas J., Andrew Sih, and David Sloan Wilson (1998). "Costs and limits of phenotypic plasticity". *Trends in Ecology & Evolution* 13.2, pp. 77–81.
- Díaz, Sandra et al. (2016). "The global spectrum of plant form and function". en. *Nature* 529.7585, pp. 167–171.
- Diaz, S. et al. (2004). "The plant traits that drive ecosystems: Evidence from three continents". en. *Journal of Vegetation Science* 15.3, pp. 295–304.
- Dwyer, John M., Richard J. Hobbs, and Margaret M. Mayfield (2014). "Specific leaf area responses to environmental gradients through space and time". *Ecology* 95.2, pp. 399–410.
- Falster, Daniel S. et al. (2016). "plant: A package for modelling forest trait ecology and evolution". en. *Methods Ecol Evol* 7.2, pp. 136–146.
- Feller, Chrystel et al. (2015). "Mathematical Modeling of the Dynamics of Shoot-Root Interactions and Resource Partitioning in Plant Growth". *PLoS ONE* 10.7, e0127905.
- Franklin, Oskar et al. (2012). "Modeling carbon allocation in trees: a search for principles". en. *Tree Physiol* 32.6, pp. 648–666.
- Frei, Esther R., Jaboury Ghazoul, and Andrea R. Pluess (2014). "Plastic Responses to Elevated Temperature in Low and High Elevation Populations of Three Grassland Species". *PLOS ONE* 9.6, e98677.
- Frenette-Dussault, Cédric et al. (2012). "Functional structure of an arid steppe plant community reveals similarities with Grime's C-S-R theory". en. *J Veg Sci* 23.2, pp. 208–222.
- Grime, J. P. (1977). "Evidence for the Existence of Three Primary Strategies in Plants and Its Relevance to Ecological and Evolutionary Theory". *The American Naturalist* 111.982, pp. 1169–1194.
- (1998). "Benefits of Plant Diversity to Ecosystems: Immediate, Filter and Founder Effects". *Journal of Ecology* 86.6, pp. 902–910.
- Hamann, Elena et al. (2016). "Evidence of local adaptation to fine- and coarse-grained environmental variability in Poa alpina in the Swiss Alps". en. *J Ecol* 104.6, pp. 1627–1637.
- Hart, Simon P., Sebastian J. Schreiber, and Jonathan M. Levine (2016). "How variation between individuals affects species coexistence". en. *Ecol Lett* 19.8, pp. 825–838.
- Hu, Yu-Kun et al. (2015). "Novel evidence for within-species leaf economics spectrum at multiple spatial scales". *Front. Plant Sci.* P. 901.
- Jung, Vincent, Cécile H. Albert, et al. (2014). "Intraspecific trait variability mediates the response of subalpine grassland communities to extreme drought events". en. *J Ecol* 102.1, pp. 45–53.
- Jung, Vincent, Cyrille Violle, et al. (2010). "Intraspecific variability and trait-based community assembly". en. *Journal of Ecology* 98.5, pp. 1134–1140.
- Kichenin, Emilie et al. (2013). "Contrasting effects of plant inter- and intraspecific variation on community-level trait measures along an environmental gradient". en. *Funct Ecol* 27.5, pp. 1254–1261.

- Kleidon, Axel and Harold A. Mooney (2000). "A global distribution of biodiversity inferred from climatic constraints: results from a process-based modelling study". en. *Global Change Biology* 6.5, pp. 507–523.
- Kooyers, Nicholas J. (2015). "The evolution of drought escape and avoidance in natural herbaceous populations". *Plant Science* 234, pp. 155–162.
- Kraft, Nathan J. B., Oscar Godoy, and Jonathan M. Levine (2015). "Plant functional traits and the multidimensional nature of species coexistence". en. *PNAS* 112.3, pp. 797–802.
- Lamarque, Pénélope et al. (2014). "Plant trait-based models identify direct and indirect effects of climate change on bundles of grassland ecosystem services". en. *PNAS* 111.38, pp. 13751–13756.
- Laughlin, Daniel C. (2014). "The intrinsic dimensionality of plant traits and its relevance to community assembly". en. *J Ecol* 102.1, pp. 186–193.
- Lavorel, S. and E. Garnier (2002). "Predicting changes in community composition and ecosystem functioning from plant traits: revisiting the Holy Grail". en. *Functional Ecology* 16.5, pp. 545–556.
- Lepik, Mari, Jaan Liira, and Kristjan Zobel (2005). "High shoot plasticity favours plant coexistence in herbaceous vegetation". en. *Oecologia* 145.3, pp. 465–474.
- Lepš, Jan et al. (2011). "Community trait response to environment: disentangling species turnover vs intraspecific trait variability effects". en. *Ecography* 34.5, pp. 856–863.
- Levine, Jonathan M., Jordi Bascompte, et al. (2017). "Beyond pairwise mechanisms of species coexistence in complex communities". en. *Nature* 546.7656, pp. 56–64.
- Levine, Jonathan M. and Janneke HilleRisLambers (2009). "The importance of niches for the maintenance of species diversity". en. *Nature* 461.7261, pp. 254–257.
- Lohier, Théophile et al. (2014). "Explaining ontogenetic shifts in root–shoot scaling with transient dynamics". en. *Ann Bot* 1, mcu128.
- Maire, Vincent et al. (2013). "Plasticity of plant form and function sustains productivity and dominance along environment and competition gradients. A modeling experiment with Gemini". *Ecological Modelling* 254, pp. 80–91.
- Matesanz, Silvia, Ernesto Gianoli, and Fernando Valladares (2010). "Global change and the evolution of phenotypic plasticity in plants". en. *Annals of the New York Academy of Sciences* 1206.1, pp. 35–55.
- Merilä, Juha and Andrew P Hendry (2014). "Climate change, adaptation, and phenotypic plasticity: the problem and the evidence". *Evol Appl* 7.1, pp. 1–14.
- Nicotra, A. B. et al. (2010). "Plant phenotypic plasticity in a changing climate". *Trends in Plant Science* 15.12, pp. 684–692.
- Nicotra, Adrienne B. et al. (2015). "Adaptive plasticity and epigenetic variation in response to warming in an Alpine plant". en. *Ecol Evol* 5.3, pp. 634–647.
- Niinemets, Ülo (2001). "Global-scale climatic controls of leaf dry mass per area, density, and thickness in trees and shrubs". *Ecology* 82.2, pp. 453–469.
- Pavlick, R. et al. (2013). "The Jena Diversity-Dynamic Global Vegetation Model (JeDi-DGVM): a diverse approach to representing terrestrial biogeography and biogeochemistry based on plant functional trade-offs". *Biogeosciences* 10.6, pp. 4137–4177.
- Pfennig, David W. et al. (2010). "Phenotypic plasticity's impacts on diversification and speciation". *Trends in Ecology & Evolution* 25.8, pp. 459–467.
- Pierce, Simon et al. (2013). "Allocating CSR plant functional types: the use of leaf economics and size traits to classify woody and herbaceous vascular plants". en. *Funct Ecol* 27.4, pp. 1002–1010.
- Poorter, Hendrik, Ülo Niinemets, et al. (2009). "Causes and consequences of variation in leaf mass per area (LMA): a meta-analysis". en. *New Phytologist* 182.3, pp. 565–588.
- Poorter, Hendrik, Karl J. Niklas, et al. (2012). "Biomass allocation to leaves, stems and roots: meta-analyses of interspecific variation and environmental control". en. *New Phytologist* 193.1, pp. 30–50.
- Poorter, Lourens and Frans Bongers (2006). "Leaf Traits Are Good Predictors of Plant Performance Across 53 Rain Forest Species". en. *Ecology* 87.7, pp. 1733–1743.

- Reich, P. B., M. B. Walters, and D. S. Ellsworth (1992). "Leaf Life-Span in Relation to Leaf, Plant, and Stand Characteristics among Diverse Ecosystems". *Ecological Monographs* 62.3, pp. 365–392.
- Reich, P. B., I. J. Wright, et al. (2003). "The Evolution of Plant Functional Variation: Traits, Spectra, and Strategies". *International Journal of Plant Sciences* 164.S3, S143–S164.
- Reich, Peter B. (2014). "The world-wide 'fast–slow' plant economics spectrum: a traits manifesto". en. *J Ecol* 102.2, pp. 275–301.
- Reineking, Björn et al. (2006). "Environmental variability and allocation trade-offs maintain species diversity in a process-based model of succulent plant communities". *Ecological Modelling*. Pattern and Processes of Dynamic Mosaic Landscapes – Modelling, Simulation, and Implications 199.4, pp. 486–504.
- Reu, Björn et al. (2011). "The role of climate and plant functional trade-offs in shaping global biome and biodiversity patterns". en. *Global Ecology and Biogeography* 20.4, pp. 570–581.
- Richter, Sarah et al. (2012). "Phenotypic plasticity facilitates resistance to climate change in a highly variable environment". en. *Oecologia* 169.1, pp. 269–279.
- Roscher, Christiane et al. (2015). "Contrasting Effects of Intraspecific Trait Variation on Trait-Based Niches and Performance of Legumes in Plant Mixtures". *PLoS ONE* 10.3, e0119786.
- Ryser, P. (1996). "The Importance of Tissue Density for Growth and Life Span of Leaves and Roots: A Comparison of Five Ecologically Contrasting Grasses". *Functional Ecology* 10.6, pp. 717–723.
- Ryser, Peter and Liina Eek (2000). "Consequences of Phenotypic Plasticity vs. Interspecific Differences in Leaf and Root Traits for Acquisition of Aboveground and Belowground Resources". *American Journal of Botany* 87.3, pp. 402–411.
- Schapendonk, A. H. C. M et al. (1998). "LINGRA, a sink/source model to simulate grassland productivity in Europe". *European Journal of Agronomy* 9.2–3, pp. 87–100.
- Scheiner, Samuel M. (2002). "The genetics of phenotypic plasticity. VII. Evolution in a spatially-structured environment". *Journal of Evolutionary Biology* 11.3, pp. 303–320.
- Scheiner, Samuel M. and Robert D. Holt (2012). "The genetics of phenotypic plasticity. X. Variation versus uncertainty". en. *Ecology and Evolution* 2.4, pp. 751–767.
- Scheiner, Samuel M. and Richard F. Lyman (1989). "The genetics of phenotypic plasticity I. Heritability". en. *Journal of Evolutionary Biology* 2.2, pp. 95–107.
- Scheiter, Simon and Steven I. Higgins (2009). "Impacts of climate change on the vegetation of Africa: an adaptive dynamic vegetation modelling approach". en. *Global Change Biology* 15.9, pp. 2224–2246.
- Scheiter, Simon, Liam Langan, and Steven I. Higgins (2013). "Next-generation dynamic global vegetation models: learning from community ecology". en. *New Phytol* 198.3, pp. 957–969.
- Shipley, Bill et al. (2006). "Fundamental trade-offs generating the worldwide leaf economics spectrum". *Ecology* 87.3, pp. 535–541.
- Siefert, Andrew et al. (2015). "A global meta-analysis of the relative extent of intraspecific trait variation in plant communities". en. *Ecol Lett* 18.12, pp. 1406–1419.
- Soussana, Jean-François et al. (2012). "Gemini: A grassland model simulating the role of plant traits for community dynamics and ecosystem functioning. Parameterization and evaluation". *Ecological Modelling* 231, pp. 134–145.
- Stamps, Judy A. and Willem E. Frankenhuys (2016). "Bayesian Models of Development". *Trends in Ecology & Evolution* 31.4, pp. 260–268.
- Sultan, S. E. (1995). "Phenotypic plasticity and plant adaptation*". en. *Acta Botanica Neerlandica* 44.4, pp. 363–383.
- Taubert, Franziska (2014). "Modelling and Analysing the Structure and Dynamics of Species-rich Grasslands and Forests". PhD thesis. Osnabrück.
- Turcotte, Martin M. and Jonathan M. Levine (2016). "Phenotypic Plasticity and Species Coexistence". *Trends in Ecology & Evolution* 31.10, pp. 803–813.
- Valladares, Fernando et al. (2015). "Species coexistence in a changing world". English. *Front. Plant Sci.* 6.

- Van Kleunen, Mark and Markus Fischer (2005). "Constraints on the evolution of adaptive phenotypic plasticity in plants". en. *New Phytologist* 166.1, pp. 49–60.
- Violle, Cyrille et al. (2012). "The return of the variance: intraspecific variability in community ecology". *Trends in Ecology & Evolution* 27.4, pp. 244–252.
- Westoby, Mark (1998). "A leaf-height-seed (LHS) plant ecology strategy scheme". en. *Plant and Soil* 199.2, pp. 213–227.
- Wright, Ian J., Peter B. Reich, et al. (2004). "The worldwide leaf economics spectrum". en. *Nature* 428.6985, pp. 821–827.
- Wright, Ian J. and Mark Westoby (2002). "Leaves at low versus high rainfall: coordination of structure, lifespan and physiology". en. *New Phytologist* 155.3, pp. 403–416.
- Zhang, Yuan-Ye et al. (2013). "Epigenetic variation creates potential for evolution of plant phenotypic plasticity". en. *New Phytol* 197.1, pp. 314–322.

III

MODELLING ALPINE GRASSLANDS WITH MOUNT- GRASS, A GENERIC FRAMEWORK INTEGRATING PHE- NOTYPIC PLASTICITY

The objective of this chapter is to develop the core concepts of the model, introduced in the previous chapter, and to explain the structure and design choices made during the model development. The first part focuses on the general context of alpine grasslands and some coexistence mechanisms at stake. The following part details the definition of the strategy space and the modelling of phenotypic plasticity while introducing the key concepts of species memory and individual experience. Finally, the last part is a detailed description of the model following Grimm et al. 2006 recommendations.

1 ALPINE ENVIRONMENT: CONDITIONS, RESOURCES, AND PERTURBATIONS

1.1 The scales of alpine grasslands

The **scale** is a determinant variable in the quantification of mechanisms that structure ecological communities Bello et al. 2013, and therefore in modelling approaches. It is chosen based on structures that the modeller intends to explore and determine the upper limit of mechanisms the model can reproduce. Large scales will favour geo-climatic and dispersal effects Kleidon and Mooney 2000 while small scales will focus on direct plant interactions processes or resource heterogeneity Soussana et al. 2012; Maire, Soussana, et al. 2013; Taubert 2014. This is true for spatial scale, but also temporal scales. Because mechanisms studies at large scales like dispersion, invasion, speciation occur over long time scales whereas mechanisms occurring at smaller spatial scale, like competition, facilitation, disturbances play a role on shorter time scales, spatial and temporal scales are often correlated. The scales are also dependent on the studied environment. There is a high contrast between highly productive environments like tropical forests and unproductive environments like mountain grasslands. The dimensions of individuals themselves are a constraint on the scales: while tropical trees grow few tens of meter high and above one meter diameter alpine grasses do not exceed half a meter **korner_alpine_2011** Similar differences of the order of magnitude can be observed for life cycles between long live tree and annuals or bi-annuals grasses. The focus of this work being on plant functioning and interaction mechanisms, the scale of the model will be around the meter, while the temporal scale will be in the order of the season.

THE SCALE

The same way the scales are constraint by the size or length of the individual, the **resolution** should be close to the size of the modelled entities. The resolution is also determined by the focus of the model: interactions between individuals must be distinct and not blended to hope see the emergence of spatial patterns. Cell size and time step length should be small enough to take into account heterogeneity that is an important driver of diversity. For these reasons, the spatial resolution is set up to the centimetre and the temporal resolution to the day (can be changed but processes might not scale well).

THE RESOLUTION

Once the resolution is fixed there is always the temptation to increase the size, or scale, of the system. This should be avoided for two main reasons. (1) the increase of scale with relatively fine resolution lead to a high increase in computational power required for simulations that are already complex. (2) there is a high chance that the processes modelled at fine resolution lose their sense when scale increases. Indeed, as mentioned, the importance of processes at stake is often dependant on the scale the system is studied. The effect of higher scale processes are often taken into account in inputs or parameters Calibration these parameters against certain data is a way to better understand these processes lagarigues

COMPLEXITY AND PERFORMANCE

1.2 Resources: light and water

As mentioned in the previous chapter, resource fluctuations, heterogeneity, and competition are important factors for coexistence. Unlike animals, plants mainly compete for the same resources: light, water and nutrients. Light is the source of energy that allows the transformation of inorganic carbon into organic matter through photosynthesis. Water has multiple functions in plants: transport, structural support, and oxygen supply for photosynthesis. Nutrients are used in the construction of cells and cell walls, and especially the production of proteins that act as cell machinery.

1.3 Perturbations: frost, grazing, and mowing

The most notable specificity of mountain grasslands is the climate. While there is a wide range of mountain grasslands type, the focus of this document is in French Alps. The alpine climate in France is characterised by cold winter with snow precipitations and dry summer. The growing season is relatively short and spread between May and October in low altitude, and June and September in high altitude **need some charts here**. The particularity of this habitat is the presence of snow cover during winter that protects soil, rhizomes, and seeds from negative temperatures. Because of this, seasons are decomposed in the model based on the snow melt in spring/summer and the first snowfall in winter. While a rise in temperature is needed to allow snow to melt down, frost event can occur after the beginning of the season. Such events represent strong environmental filter for non-adaptive plants that do not invest in specific resistance mechanisms to favour early germination and growth. Therefore there is a strategic trade-off between germination date and early growth with frost resistance.

CLIMATE

Another specificity of alpine grasslands is that they are subject to changing management practices. Mountain grazing by domestic cattle was fairly common in the Alps, but changes in agricultural practices and a decrease in productivity due to drought lead to less and less grazing or mowing for fodder in alpine grasslands. These two types of management have different impacts on the community. While mowing is non-specific and favours small plants, herbivory is known to be specific when the production is greater than the grazing pressure. Leaves with high nutrient content and low structural tissues content are generally favoured because of high input and high digestibility. The grazing pressure

MANAGEMENT

plays as ... mechanisms as it favours conservative species with lower competitive ability.

OTHER FORM OF HERBIVORY

2 MULTI-DIMENSIONAL STRATEGY SPACE, CARBON POOLS, AND TRADE-OFFS

2.1 Multi-dimensional strategy space and allocation pools

within species LES

2.1.1 The strategy space in *MountGrass*

In an ecological agent-based simulation model a species will be defined by its values for the species-specific parameters. They can be estimated from experimental data Taubert 2014; Maire 2009; Lohier et al. 2014 or be picked from a strategy axis Reineking et al. 2006; Kleidon and Mooney 2000 composing a strategy space Westoby 1998. The diversity of the species pool will depend on the number of values for each of these specific parameters, or traits, and the number of these traits. Each trait increasing the dimension of the strategy space Laughlin 2014. The ambition of this model being to simulated rich plant communities, the definition of these axes is crucial. Trade-offs between traits are excellent applicants for these specific parameters as they reduce the dimensionality of phenotypes to a small number of dimensions Wright et al. 2004; Díaz et al. 2016; Peter B. Reich 2014 while keeping the information of traits needed to describe the plant functioning. Trade-offs emerge from ecological and physical or biological constraints, by considering these constraints Darwinian demons are avoided.

WHAT IS A STRATEGY SPACE

While considering too many axes does not improve community description, a certain number is needed to have strategic diversity Laughlin 2014. This is intuitively explained by the fact that each trade-off is closely related to a particular aspect of fitness or mechanism for coexistence (e.g. reproduction, competitive ability, resistance to resource shortage, predation, etc.). In this model, multiple aspects of plant life are represented: germination with the germination rate for storage effect Chesson 2000; Adler et al. 2006, dispersion with seed mass Westoby 1998 or tissue construction cost P. B. Reich, Walters, and Ellsworth 1992; Wright et al. 2004; Peter B. Reich 2014. Main components of plant growth and life history are covered by such trade-offs and driven by mechanisms shared by all vegetation systems. Because of that, the model has a great potential of genericity and diversity. It can be easily adapted to other plant communities with specific calibration, and extended to couples of biological process and differentiation axis (e.g. root herbivory and associated resistance carbon pool). The the trade-offs used in the model are detailed in the model description below ².

² see section 4.

These axes should, in such models, be independent, (*i.e.* it is physically and biologically possible for a plant to take any position in the space drawn by two given axis) and result from physical or biological laws (ensuring that impossible strategies are indeed excluded from the model). First, it is a condition for parsimony of the model. The second and more interesting reason is that any trade-off emerging from the model should have an ecological interpretation Maire, Gross, et al. 2013.

One way of constraining plant strategies to certain axes is to consider allocation trade-offs Kleidon and Mooney 2000; Reineking et al. 2006. An allocation trade-off is the translation of the mass conservation rule that prevents the allocation of biomass to distinct carbon pools. If biological functions are related to organic matter pools (photosynthesis to leaves, water and nutrient uptake to roots), then the sum of biomass to invest in each carbon pool (therefore in each function) cannot exceed the total available biomass: leaving the plant with a choice on the balance between the different functions. Allocation trade-offs have the advantage to be easily implemented and be intuitive. By design, a partitioning factor value corresponds to a position on the related strategic axis. In *MountGrass*, five main trade-offs are captured by allocation trade-off: (1) development vs reproduction: partitioning factor between reproduction and maintenance of vegetative tissues (when plant is mature), (2) persistence vs dispersion: partitioning of reproduction biomass between persistence (storage) and production of new propagules (seed/clone production), (3) aboveground vs belowground competition: investment between shoot and root, Kleidon and Mooney 2000; Reineking et al. 2006; Taubert 2014(4) slow vs fast: construction cost trade-offs between active and structural tissues in both shoot and root and (5) growth vs resistance: partitioning between stored biomass and frost resistance carbohydrates Cai et al. 2004. This last trade-off can be extended to other carbon pools of specific resistances, for example to herbivory. Modification of these coefficients during life history is a way to introduce plasticity in the model. The rules driving such changes for some of this partitioning parameters are described in the following section.

One of these trade-offs, (4), is key and related the construction cost of organs (independently leaves and roots). Highlighted at the global scale and for leaves, the Leaf Economic Spectrum Wright et al. 2004 draws a strategic differentiation axis from conservative slow species and exploitative fast species. The construction cost has long been identified as a factor of strategic differentiation in plant communities Westoby 1998. This strategic axis, being related to many functional traits: SLA, LDMC, LNC, leaf longevity, Amass, etc. Wright et al. 2004 is of crucial importance. First, these traits are closely related to the characterisation of plant communities and the assessment of services Grime 1998. Second strong links and correlations can be made between these soft traits physiological traits **wright_worldwide_2004**; Craine et al. 2002; Peter B. Reich et al. 2003. Finally, a species resource use strategy is closely related to its responses and vulnerability to changing conditions Poorter et al. 2009; Dwyer, Hobbs, and Mayfield 2014; Deléglise et al. 2015. The traits related to this trade-off play a major role both in individual growth and physiology and in community services and response to a gradient. Therefore it is essential to the model. Questioning the underlying

mechanisms for such strong trade-off is necessary to implement satisfying representation in the model.

Change this: maybe start with Shipley results, then composite stuff. Question: should it be here, or in the following part ?

These trade-offs between highly productive tissues with low construction cost and short lifespan called exploitative, and more conservative strategy with longer lifespan but lower productivity are mainly observed thanks to soft traits such as SLA for LNC Wright et al. 2004. Mechanistic models require traits related to physiology and organ performance Soussana et al. 2012; Lohier et al. 2014, but a link can generally be done between these traits and soft traits. However, traits such as SRL or SLA are composite traits emerging from different organ properties P. Ryser 1996; John et al. 2017, where tissue density and organ thickness are the main determinants. "A necessary trade-off between allocation to structural tissues versus liquid phase processes" has been identified by Shipley et al. Shipley et al. 2006 as one of the two main factors for the leaf economic spectrum to emerge. Such allocation trade-off can indeed explain differences in construction cost as the liquid phase corresponding to the "active" part of plant tissue, the cell content, have much lower dry volumetric mass than its "structural" counterpart, the cell-wall. Also, active tissues containing the protein machinery for photosynthesis and water absorption, a higher proportion of high protein concentration tissue would be correlated to higher nitrogen concentration in the organ on the "fast-slow" spectrum, along with a higher mass-based photosynthetic rate Peter B. Reich 2014. On the other end, the structural tissues give the organ a higher lifespan Mediavilla, Escudero, and Heilmeier 2001; P. Ryser 1996 that compensate for lower productivity Westoby, Warton, and Peter B. Reich 2000. Such trade-off can be apply to both shoot and roots Craine et al. 2002; Tjoelker et al. 2005; Peter B. Reich 2014. From that, the decomposition of organs between active and structural tissues constitutes a strong basis to model construction cost trade-offs as the main parts of the global strategy space.

The Similar axis of differentiation has been demonstrated for roots [tjoelker_linking_2015](#); Peter B. Reich 2014. The necessity for independent similar axis for leaves and root can be discussed with respect to coordination between shoot and root activities. Because perfect equilibrium cannot be guaranteed in all conditions, strict coordination cannot be taken as a principle for the reduction of strategy space. Moreover, empirical results suggest small deviations from coordination are common Freschet, Kichenin, and Wardle 2015. The leaf economic spectrum being conserved at the intra-specific level Hu et al. 2015 is another reason to include such trade-off as it would be a good basis for phenotypic plasticity Freschet, Bellingham, et al. 2013.

The use of allocation trade-offs allows the construction of a generic multi-dimensional strategy space where a high diversity of species can potentially coexist. Because this space is based on physic laws, it ensures the non-existence of Darwinian demons and does not limit the species or individual plants to tested parameters and strategies. To be complete the link between carbon pool allocation and physiology must be determined within the respect of similar biological or physical laws.

2.2 Craft a trade-off: active and structural tissues

Allocation trade-offs offer great flexibility and are easily understood and implemented. However, when they control the value of traits (SLA or SRL) involved in multiple processes, a balance must be found to avoid that: (1) one process is ignored because has a low relative importance onto fitness (becoming useless to the model), (2) the effects of processes involved show strong response curves to the allocation and there is only one global¹ optimum. The idea behind a trade-off is that multiple positions are viable in different conditions or in association with other strategies. The leaf-economic spectrum, in addition to relying on the active-structural tissue trade-off, also requires "*an evolutionary trade-off between leaf photosynthetic rates, construction costs, and leaf longevity*". This trade-off is explored in this section of the document.

In the framework of the model, plants share the same global parameters, and the maximum photosynthetic rate should be the same. Because photosynthesis relies on the exchange of gases (CO_2 , O_2 and H_2O) and the interception of light, it is related to exchange area. Considering one shared parameter for maximum area-based potential exchange rate satisfy both the need for a shared parameter and a way for plants to vary their mass based exchange rate by changing its proportion of active tissues. This is in agreement with the LES that a strong relationship between mass-based traits and limited ones for area-based variables Wright et al. 2004, and explain the first part of the trade-off between photosynthetic rate and construction cost. The second part is the relationship with the longevity. The longevity is often correlated to SLA in empirical studies, however, this is mainly explained by differences in tissue density and toughness than in thickness (another component of SLA). For this reason, we can directly link the leaf longevity to active tissue proportion. Respiration is also increased by the increase of the proportion of photosynthetic tissues **kleidon; reich** We have now a trade-off between a gain function (exchange area gain by changes in densities) and a cost function (tissue turn over and respiration). This should be enough to explain different strategies **westoby** However, the model needs internal limits to avoid the gain function to lead to only active tissue organ (or only structural). These limits are required to allow individuals or species to change position along these axes (plasticity or strategic shift). The convex shape of gain function in association with a minimal cost (minimum turn-over cost above maximum potential gain) is enough to limit the allocation to structural tissues only. To avoid allocation to only active tissue, that would correspond to an organ made of protoplasts, the cost function needs higher than the potential gain. To achieve that an exponential function is chosen. This choice ensures that the potential gain function has an optimum different from the borders. (**see figure**).

Poorter et al. 2009 variations in SLA due to the same extend to LD than Th. Picon-Cochard et al. 2012 root density and root respiration

Active got closer to optimum, but less active and positive gain in more conditions. Can I demonstrate this with formulas ? (gain = function(condition))

¹ I use the term global here to designate the multidimensional space draw by the axis of interest and other variables play a role in involved process (e.g. resource availability, temperature etc...).

depends on resource availability. Changes in resource level imply changes in the slope of gain function and a shift of the organ optimum for tissue allocation. This shift makes more conservative strategies more interesting when resources are scarce, while more exploitative allocation strategies are better for high resource availability. This link between optimum allocation and resource level could be used to define the best phenotype according to experience conditions, but the organ strategy cannot be disconnected from the whole plant strategy and allocation.

The phenotype (within the subspace of vegetative allocation) depends both on the individual efficiency of organs and the balance between shoot and root activity. This balance often used to model plant plastic allocation and considered between light and nitrogen **lohier; soussana** In the context of mountain grasslands and global change, the water... The integration of nitrogen as a limiting solution is discussed in a later chapter. The balance between shoot and root activity being key in overall performance, the root:shoot ratio (RSR) will be determined as a function of estimated availability.

2.2.1 Species memory and phenotype determination

phenotype = ensemble of response trait values. Emerge from default trait + environment.

Composite traits are defined by the interaction of different, independent, driving traits. What is a driving trait? Biology: genetic information. This genetic information is selected by climatic conditions. If we can make a link between optimum value for a trait and environmental conditions, then store external conditions and use link between.

MEMORY OF SPECIES: A DRIVING TRAIT

The decomposition of organs organic matter in active and structural carbon pools makes a link between allocation and physiology and draws a subspace within the strategy space where individuals can move and change their phenotype. Limiting mechanisms restrain the viable options to realistic values along these axes. Within this space, the resource availability and external conditions play a major role in the expression of the strategy.

VEGETATIVE PHENOTYPE AXIS

Flexible, allocation based, diversity and movement

3 MODELLING PHENOTYPIC PLASTICITY

3.1 Plasticity as a strategy: between species memory and individual experience

3.1.1 Concept of active plasticity as a strategy

The active plastic response is highly integrated and involve a lot of regulatory processes. It is impossible to represent all regulatory processes involved in an APR (because of our lack of knowledge and their complexity). Alternatively, the concept of *integrated response* can be conceptualised. It supposes link, or coordination, between the experienced conditions and the phenotypic response. This can be translated, in the model framework, by the existence of an explicit link between a representation of external conditions and a phenotype matching this conditions: the **allocation rule**¹. Another key work is *anticipatory*. It supposes that the plant knows, or at least have an idea of the future conditions. This is really the point of an active plastic response: change the phenotype to better match future conditions. A representation of future is also called a **projection**. The projection and the allocation rule together form the active plastic response. If allocation rule is not obvious and is discussed latter(see paragraph 3.2), the idea of projection is fairly intuitive. The projection will correspond to a value for a given metric that represents the external conditions. It can be resource availability level, temperature, herbivory risk, etc... If such metrics can be given at the community scale, it makes sense to use a plant-centered measure of these variables for two reasons: (1) take into account the spatial heterogeneity, (2) plant experience of conditions is necessarily egocentric. The details on how experienced conditions are interpreted by plants in *MountGrass* are described in section 4.

DECOMPOSITION OF PLASTIC RESPONSE

¹ the use of the word *allocation* is justified here since the phenotypic plasticity in *MountGrass* is reduced to changes in allocation.

Active plasticity is now represented by a projection and allocation rule. However, how a species can control the whole process is unclear. In theory, both projection and allocation rule can species specific. In nature plants generally have structurally similar regulatory processes² and response to external stimuli is translated (and stored temporarily) thanks to the accumulation of chemical compounds(**need-references**). These mechanisms suggest that, while the allocation rules are mainly shared, individuals vary on the information level (i.e. the concentration of phytohormones), or in the context of the model: plant vary in projection. This control of active plasticity is supported by the model design. The number of rules that can drive the allocation is reduced and discrete, while the projection is multi-dimensional (one dimension per external variable considered), continuous and highly flexible with a reduced number of parameters³. For this reason **projection** is chosen to be the **controlling factor** of active plasticity, while the allocation rule is **fixed** and **shared** between all species. Therefore an individual with fixed projection won't be actively plastic, despite the fact that it could express apparent plasticity because of external factors: reduced resource availability, grazing, frost damage, etc... The model has now a concept for active plastic response⁴ controlled by the projection of external condition. The next question that needs to be answered is: how do species differ in their plastic response ?

CONTROL OF PLASTICITY

² see box ??.

³ details in paragraph *estimation of conditions* in section 4.

⁴ In the rest of the document terms *plasticity* or *phenotypic plasticity* will refer to *active plastic response*.

In *MountGrass*, the projection of external conditions is the mean for plants to alter their phenotype in response to changes in experienced conditions. Since the allocation (or driving) rule is shared by all plants, if the projection of external conditions is also shared by all plants, then is the response still active plasticity? The first intuitive answer is *yes*, since the conceptual framework is respected and plants would react to changes in conditions that would affect the projection. But, such response would be equivalent to a direct external control of the climate on the phenotype. In such case, species would not have control on how the phenotype varies, that would be fully controlled by shared projection and shared allocation rule. This is passive plasticity. To have proper active plasticity, the species need to be able to **regulate** the plastic response. If species can regulate plastic response thanks to species-specific parameters, plasticity becomes a **strategy**. This is in agreement with Bradshaw vision of phenotypic plasticity as a trait, or a character, subject to selection and evolutionary processes(A. D. Bradshaw 1965; Anthony D. Bradshaw 2006). How do species regulate plastic response to make it a strategy?

SPECIES SPECIFIC PLASTICITY

Plasticity: expected environment -> phenotype, here phenotype is equivalent to biomass partitioning, that means expected environment -> allocation coefficients. Then memory -> expectations -> allocation. Because low dimensions, and we want diversity, and the link between memory and allocation might not be a function (one memory give exactly one optimum allocation), in the model this relationship is not verified. Species-specific traits are used to allow for different strategies to be associated with the same memory (different plants won't have the same strat, despite sharing the projection)

3.1.2 Projection of external conditions: between species memory and individual experience

The projection is the way plant controls phenotypic plasticity. A projection is an idea of the future based on available information and on the understanding of a phenomenon. Ones could discuss what is the understanding of the climate by plants, while others can focus on how to represent such understanding and state that fine molecular regulatory processes can reproduce and store such information. The focus is on the construction of the projection with respect to the different sources of information a plant has: (1) its experience of climate and external factors, and (2) its ancestors' memory¹.

BUILD A PROJECTION

While, for any given individual plant, the experience of external conditions varies in time, the memory stays fixed. There is a clear contrast between the variable experience of conditions and the fixed species memory. A way to represent different strategies and the level of control the plant apply on projection is to vary, between species, the relative weight of species memory against individual perception. This species-specific parameter, the **confidence in species memory**, sets the **stability** of the projection with respect to individual experience. The capacity to adapt the phenotype to changing conditions is directly linked to the projection changes. High confidence in species memory translates in low amplitude of projection variations, and though in low active plasticity. The calculation of projected resource availability levels, or temperatures, are de-

¹ see paragraph 2.2.1.

tailed in the dedicated paragraph of the model description. The key message is that the species has control on plasticity with both its confidence in species memory and the said memory that alters the projection. The relative impact of memory and confidence is described in figure ??.

3.2 Driving rules of allocation

Allocation rules are determinant in the model behaviour as it is shared by all species, and link the projection of conditions with the phenotype. Multiple options are possible to drive plasticity, but they can be divided into two main categories: (1) determining, (2) directive. Functions from the first category fully determine the phenotype (within the plastic strategy space), while functions from the latter group direct the phenotype toward a sub-space and other parameters are needed to determine the exact new phenotype.

The two type of rules have different strengths and weaknesses that are detailed in table ??

Strength or weakness	Determining	Directive
Phenotype fully determined	●	○
Risk of convergence	●	○
Reduction of functional diversity	●	○
Discrepancy between parameters	●	○
Strong plasticity effect	●	○

Table 3.1: Two types of allocation rules: strengths and weaknesses

One of the assumption made by the conception of plasticity developed in this document is the existence of a tight relationship between experienced condition and fitness, and a subsidiary assumption in the implementation of this plasticity is that this function can be captured, or modelled, by the same functions that drive plant growth. In other words, simulating individual growth, using the estimated/projected conditions as parameters, day by day, is enough to capture the link between environmental conditions (experienced by the focal plant) and plant growth¹.

... How these rules fit into that ?

The driving rule of plasticity defines whether or not the choice of the phenotype is fully determined by the projection of external conditions or also constrained by some species-specific parameters. The effect of this balance between projection and parameters has a large influence on the model behaviour. In any case, the projection is the main control on individual plastic response to change in conditions, offering possibilities to modulate individual plasticity despite an allocation rule shared by all species. The role of both projection and allocation rule will of particular interest during the analysis of the impact of phenotypic plasticity of plant growth and community dynamics.

While the representation of plasticity as a strategy increases both model potential species diversity and potential diversity of response (Peter Ryser and Eek 2000; Kichenin et al. 2013), plasticity itself may reduce diversity. Indeed, plasticity lead to changes in phenotype in response to condition changes, while these phenotypic changes are unlikely to be identical for all individuals, their general convergence points will probably be similar. Plasticity is a mechanism that is

¹ that takes here the value of fitness proxy

likely to contract the space of expressed values for plastic traits. Therefore, it is hard to analyse the effect of plasticity on functional diversity without disentangling the direct effect on the expressed trait values, and the indirect effect by changes in performance and interactions. Nonetheless, some external mechanisms¹ can prevent convergence of phenotype: (1) changes in competitive hierarchy may lead to differences in individual experience of conditions, (2) specificity of the external driver, e.g. selective herbivory of more digest species, (3) relative position of the optimum phenotype.

Asides from these external mechanisms, there are internal controls of active plastic response: the projection and the plastic allocation mechanisms. It is easy to imagine numerous projections and allocation mechanisms, however, they are susceptible of emerging only if they have a positive impact on fitness overall. Considering the diversity of plastic response is a research question in itself, and I will not try to answer it in this document. Nevertheless, the progress in the understanding of the effect of plasticity on performance and potential diversity this work provide will certainly help further work in that direction. In this context, the use of species-specific control over the projection of conditions is already a step forward and prevent total convergence². Indeed, without considering multiple allocation algorithm within a same community, having the plasticity as a strategy (Anthony D. Bradshaw 2006) (controlled by a species-specific trait, as opposed to many existing individual-based-models) allows interesting questions to be addressed. The questions of cost of plasticity is central in the understanding of this mechanisms (DeWitt, Sih, and Wilson 1998; Auld, Agrawal, and Reileya 2009), and could lead to mechanisms of co-selection between resource use or reproductive strategies with plasticity strategies. A first step in this direction consists in looking at how plasticity can have different impacts on performance of species with different strategies (conservative versus exploitative).

¹ impact the drivers of plastic response, not the response itself

² in addition to directive allocation mechanisms, see below subsection 3.2

3.2.1 On the difficulty to match strategy and conditions.

As mentioned, the framework of plasticity developed in this document relies on a strong link between condition estimation and the phenotype, that is supported by the assumption that similar link exist between condition experienced by the focal plant and its fitness. If this assumption is correct, then the initial phenotype (or default phenotype) should match the optimum phenotype defined by this link expressed by the allocation rules and the species-specific memory of conditions. One main difficult emerges here: because the processes involved in plant growth are numerous and complex, it is not possible to determine analytically what phenotype is the best (considering the memory of conditions). This point is discussed in the following paragraph as the understanding of the component of plant performance is a first step to understand the model's behaviour and plasticity mechanisms. Ones could compute the convergence phenotype for a given memory of external conditions for each possible memory combinations, and map the phenotype to the memory. This solution is a good alternative to analytical solution when the later is not possible, but it comes with the disadvantage of very high computational cost that is prohibitive for calibration procedures. When a model is only directive, the

WHY MATCH PHENOTYPE TO MEMORY

Here lies a difficulty, indeed, the design of the model favours modularity and different allocation algorithms do not share the same constraints. 2 things here:

- why having default traits even if they do not match the optimum defined by the allocation rule.
- the difficulty to determine an optimum phenotype with analytics.

Complexity between organ, overall efficiency, and equilibrium. Try to formalise each function and show the link between these elements.

The multi-dimensionality of fitness.

EFFICIENCY AND PERFORMANCE

IS PERFORMANCE FITNESS?

Fitness and competition: where niche theory and coexistence theory coexist ([letten_linking_2017](#))

The projection of external conditions, driving the plastic allocation of organic matter, lies on a balance between species memory and individual experience. Its design makes of plasticity an axis of strategic differentiation alongside the other strategy axis. Thanks to this innovative design, the model can be used to examine the ecological relevance of plasticity in different conditions and in association with different strategies. The effect of allocation rules and projection stability can be explored independently or conjointly for a better understanding of relative importance of allocation and plasticity.

4 ODD DESCRIPTION OF THE MODEL *MountGrass*

This document is a detailed description of the *MountGrass* model. This description is based on the ODD protocol of Grimm et al. The model is inspired by multiple other forest and grassland models (for grassland models see particularly Taubert (Taubert 2014) and Lohier (Lohier et al. 2014)). It differentiates itself from these models by the incorporation of phenotypic plasticity in a generalizing framework for plant functioning. This allows it to be used to both to explore the fundamental effects of phenotypic plasticity the dynamics of rich grass communities and the impact of the phenotypic plasticity on plant interactions. The general approach and the practical details are further detailed in this document.

4.1 Model overview

4.1.1 Model purpose

The development of *MountGrass* is motivated by the need for a flexible tool to explore the complex dynamics of mountain grassland communities, in the context of global change. This tool should, by a better understanding of community dynamics and representation of plant strategies and interaction, also help in the assessment of ecosystem services in new conditions. We believe that to capture the dynamic of such communities, we need to understand and represent first the individual response of plants to fluctuating levels of resources, and the impact of plants on the resources. Individual responses and relative impact should follow general rules of plant physiology but also integrates specific behaviour based on the species resource use strategy and individual characteristics. Therefore the

model should allow following distinct individuals from different groups (e.g. species) in a spatially explicit environment where they compete for resources.

Moreover, since we focus on the community levels, coexistence mechanisms are important and we should include a certain number of these if we want to maintain diversity to observed levels. These mechanisms include: multiple resources competition (water and light), spatial and temporal heterogeneity of resource levels, strategic trade-off between species, perturbation mechanisms (frost, management), link to meta-population, etc...

The model is built to try to satisfy conditions to reproduce and explore mountain grassland community dynamics. In the current version of the model (MountGrass2.0), a generalist approach has been privileged, and focus on some coexistence maintenance mechanisms and integration of phenotypic plasticity framework. In this state, the model has to be seen as a toy model with good generalisation potential. The link between to ecosystem services are not included, but we can easily imagine to compute them from the community trait distribution. All processes and mechanism are detailed below.

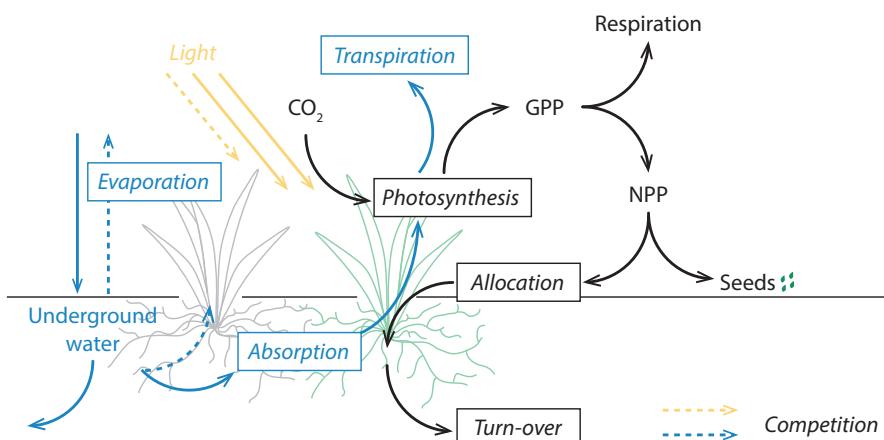


Figure 4.1: Model overview. **Water** and **carbon** cycles are represented. Processes are represented framed and in *italic* by contrast with pools that are not framed and in regular fontface. Dashed arrows indicate loss of resource (for the **focal plant**) due to competition.

4.1.2 State variables

In mountain grasslands individuals (tillers) generally do not grow big and interact only with close neighbours and form little patches. And thus it is possible to represent a rich community at a fairly small scale (\approx dm or m), but the spatial resolution should be relatively fine (\approx cm) to capture inter-individual interactions. Because the model is intended to explore climate change impact on mountain grasslands, it can run on multiple growth seasons separated by snow-covered periods, but must also integrate the intra-seasonal variations at daily scale. Mountain weather (mostly temperature) is known for its large hourly variations, it would, however, require too much computational power to consider such variations. In addition to this argument, we believe that even though they imply physiological flexibility and specific strategies for plants experiencing these conditions, they will not have a huge impact on overall community dynamics changes caused by the climate change. That why hourly variations will not be considered and physiological processes are estimated at the daily timescale.

SCALES

The plants are described in the model by state variables described in table 4.1. The best way to understand how plant are represented is to imagine two homogeneous cylinders on top of each other, the shoot cylinder varying in radius and height representing the light acquisition (and shading) zone, and the root cylinder varying only in diameter (because of shallow soil in mountain ecosystems) representing the water acquisition zone. These cylinders are centred on cells of the torus simulation plan.

In addition to classic variables (age, position, height, diameter, shoot and root biomasses) the plants are described by traits, that can be species-specific or non-specific, others are variable (SLA, SRL) and depend on particular traits that are unique to this model: the **ratio between active tissue and structural tissue** (in shoot and root) (variables $\frac{act}{str\ ag}$ and $\frac{act}{str\ bg}$ in table 4.1). This couple of traits come from the evidence that numerous trade-off observed in leaves can be explained (at least partially) by this allocation trade-off between active tissue producing organic matter, but increasing respiration, and structural tissue that increase tissue lifespan.

PLANTS

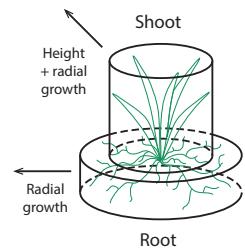


Figure 4.2: Plant geometry and growth axis.

Table 4.1: State variables of individual plants

Variable	Description	Unit
x	x position on the grid	cells
y	y position on the grid	cells
age	age	days
sp	species	-
BM_{ag}	above-ground biomass	g
BM_{agsen}	senescent above-ground biomass	g
SLA_{sen}	senescent above-ground biomass	$cm^2.g^{-1}$
BM_{bg}	below-ground biomass	g
stem	stem biomass	g
$\frac{act}{str\ ag}$	above-ground active on structural biomass ratio	g/g
$\frac{act}{str\ bg}$	below-ground active on structural biomass ratio	g/g
h	height	cm
r	shoot radius	cm
r_r	root radius	cm
$light_{exp}$	above-ground potential resource availability	$gH_2O.\text{leaf area}$
$water_{exp}$	below-ground potential resource availability	$gH_2O.\text{root area}$

Plants are characterised by state variables that describe them individually, but they also share common characteristics with individuals of the same group, (we will refer as *species* to talk about this group in the rest of the document even though it could be a group at another scale (i.e. population, clones). These species are the groups present in the meta-population and that can invade the simulated ecosystem. There are described by multiple traits characterising the strategy of the species (table 4.2).

SPECIES

The seed-bank is the transition state between the different seasons. Individuals may persist thanks to stored resources, but they can also reproduce by the production of new individuals. A lot of grasses use clonal reproduction, in

SEED-BANK

Table 4.2: Species traits

Trait	Range (close range)	unit	trade-off or strategy
seed mass	(0.00001 - 0.001)	g	seed output vs seedling productivity
maturity	-	green biomass	flowering time vs reproduction potential
fract_dev	0-1 (0.05-0.6)	-	blooming vs persistence
fract_rep	0-1 (0-1)	-	reproduction vs persistence
geometric constant (k_g)	(0.1 - 20)	-	competition sensitivity vs self-shading
plasticity stability	0-1 (0.8-1)	-	genetic information vs experience
initial water resource	(0.001 - 0.05)	$gH_2O.cm^{-2}$	water resource niche
initial light resource	(0.001 - 0.05)	$gH_2O.cm^{-2}$	light (in H_2 equivalent) resource niche
$\frac{act}{str ag,d}$	(0.03 - 0.3)	$g.g^{-1}$	active vs structural tissue
$\frac{act}{str gg,d}$	(0.03 - 0.3)	$g.g^{-1}$	active vs structural tissue
mean temp.	(0 - 5)	°C	early vs late germination
germination rate	0-1 (0.5 - 1)	-	good season bet-hedging
thickness	(0.012 - 0.05)	cm	WUE vs light efficiency (not in this version)

addition, or replacement of sexual reproduction. This type of reproduction is characterised by a persistent link between the newly produced individuals and the parent one that allows the two to communicate and exchange resources. Such dynamics are complex and costly to represent as the link between ramets must be stored and strategies defined for the resource distribution (see Oborny 2012) for more details on clonal growth modelling). To avoid too much complexity, it is possible to approximate the representation of clones to big seeds with little dispersion around the parent plant¹. For this reason, reproduction mechanism is reduced to sexual reproduction mechanism with the production of "seeds". Seeds are stored in the seed-bank and only defined by their species and positions.

The soil is an important aspect of the model as it drives (with the precipitations) the water competition between individuals. It is however limited, as in numerous vegetation models, to a grid characterised by its capacity to retain water, and its depth. Only the first component (water retention capacity) is spatially variable and is described by the critical water content (minimum soil water content), the saturation water content (maximum water content, the water non absorbed leaves the system we assume the same root depth for all species), and the current water content (temporally variable, depending on competition, precipitation and evaporation, between the critical and the saturation water content) only dynamic variable among the three.

4.1.3 Process overview and scheduling

As mentioned the model runs at a daily step to capture individual responses to conditions and over multiple seasons to capture long temporal dynamics. Some processes occur (or are evaluated) at the daily time-step, some at the season time-step. The following ordered list presents the different processes and the scheduling over days and season of one simulation.

One season can be divided into the following parts:

¹ This would take advantage of dispersion kernels. Not implemented in the current version. Dispersion is uniformly random within the simulation plan

SOIL

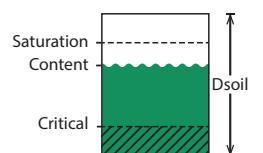


Figure 4.3: Soil section.

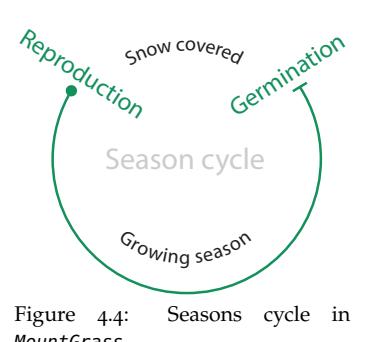


Figure 4.4: Seasons cycle in MountGrass .

- *germination*: marks the beginning of the season when the ground is no more snow-covered;
- *growing season*: consists in daily processes like competition, production of organic matter (OM), allocation, and death lottery;
- *reproduction-invasion-persistence*: marks the end of the season when the first persistent snow-fall occurs. OM invested in reproductive tissues turns into seeds that are sampled to create the seed-bank. Seeds from the meta-population may integrate the seed-bank. Persistent perennials loose most of their biomass but storage (and eventually stem) and regrow from stored organic mass at the beginning of the following season.

The *growing season* part consists in all processes evaluated every day of the growing season. These processes are:

- *light competition*: the individual potential photosynthetic activity is computed based on average daily light and shoot properties;
- *water competition*: evaporation and the individual water update (and potential water uptake) are computed based on potential transpiration, water availability and potential evaporation;
- *production*: respiration and production are computed to give the net productivity in OM;
- *senescence*: based on lifespan a part of tissue is no longer active.
- *death*: death of individuals based on their age and their desiccation stage (number of consecutive days with negative growth).
- *allocation*: allocation of produced OM to the different carbon pools of the plant.
- *grazing/cutting*: (optional) grazing or cutting of plants to a certain height. The grazing can be selective.¹

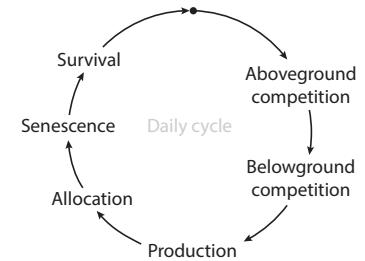


Figure 4.5: Processes in order during the daily cycle.

¹ remarks in grey are features or components implemented in the model but not used and-or calibrated.

4.2 Design concepts

4.2.1 Design concepts

This part clarifies the rules that drive the dynamics of the model.

The purpose of the model is to understand the rules that drive the community responses. We tried making the community dynamics emerge from the underlying processes of plant growth, resource use, and reproduction. That means that population dynamics are at least partially emergent from the surviving and reproducing individuals. *Partially* emergent because it depends on the invasion rules applied to the system. The traits and biomass distribution that describe the community are completely emergent from the individual traits exposed by the individuals and their relative biomass and abundance.

Plants have in theory many options to adjust their phenotype and increase their fitness in response to changes in environmental conditions (resource availability, temperature, ...). High diversity of mountain grasslands suggests that multiple strategies coexist and that individuals do not change to converge toward a

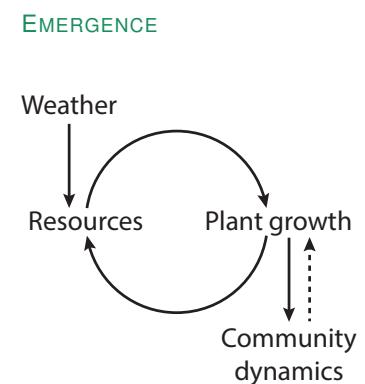


Figure 4.6: Population dynamics emerging from plant growth and weather.

ADAPTATION

unique strategy. These strategies are set up at the species level by the species-specific traits (see table 4.2). Therefore, individuals may only adapt morphological traits but not strategic traits (unless there is an epigenetic mechanism added). These morphological traits are the relative biomass of shoot and root, the relative proportion of active and structural tissues in each leaf, and roots (controlling respectively the SLA and SRL and the overall resource acquisition cost)¹. Geometry traits (distribution of leaves and roots within space) are not considered plastic as grasses have far less control over their geometry than forbs or trees. Root distribution plasticity has been shown to greatly improve the individual and community productivity (Gemini article), but to keep the model (and implementation) simple we will ignore root distribution plasticity and foraging strategies to focus on allocation problems instead of spatial distribution questions. Shallow soils and relative small rooting zone are also arguments to ignore spatial distribution plasticity for roots.

In the model, the realised fitness can be estimated as the capacity of plants to maintain themselves or their descendants through time. It emerges from the productivity, allocation to storage or reproductive carbon pools, and survival. Assessing fitness as the average number of persistent individuals is, however, a bit hazardous in simulations limited in time and to a relatively small spatial scale. Plus, plants cannot easily make a prediction of such variable to adjust their phenotype. They need a proxy function for fitness that integrates measures of external conditions to evaluate the best strategy to develop. As said above, this strategy should be a composite between the species strategy and individual adjustment specific to the individual experience of the environment. Plant fitness is estimated by individual plant thanks to a gain function integrating current phenotype, species strategy, and projection of future conditions. This gain function can take multiple forms and be more or less constraint. In the context of the model, the function should include a measure of productivity that relies on the principle of functional equilibrium - that is the allocation of organic matter to maintain the balance between the shoot activity (transpiration) and root activity (water uptake). This equilibrium can be achieved by changes in shoot:root ratio only, or also changes in active over structural tissues ratio. Further details about the gain function are discussed in the dedicated paragraphs (4.3.3). A more complex form of functional equilibrium incorporating nutrients (like nitrogen) could be added to the framework of this model.

Adaptation or plasticity mechanisms imply that agents have an insight of what will be the future. In *MountGrass* we consider that plants have two main sources of information. The first source of information is the genetic information. Indeed, the evolutionary process of genotype selection has led to the selection of genotypes adapted to the local conditions. This selection relationship can be seen as a link between environmental conditions and genetic information. Because plants cannot fully predict future environmental conditions, they grow following (at least partially) the plan contained in genetic information that match conditions where previous generations grew in. This is an internal *a priori* information about the external conditions. If the conditions where the seed grows change from the conditions its genotype has been selected for, the genetic information does not fit the environmental conditions is not sufficient enough to build a working phenotype. In this case, if the plant has a plasticity capacity, it

¹ and optionally the proportion of stored OM dedicated to frost resistance and not to growth

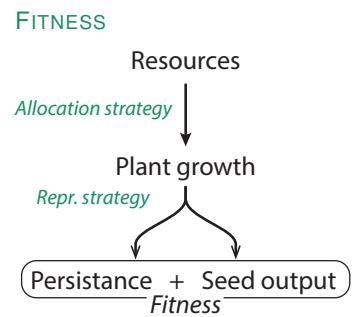


Figure 4.7: Fitness emerges from the plant growth and the plant reproductive strategy.

The plant growth is the result of the interaction of the resource levels, the plant strategy, and the competitors.

PREDICTION

can integrate the second source of information, in the form of the experienced conditions, to its "a priori" and forge a new estimation of what conditions will be. One question emerges to this idea is: how to create an image of future conditions and how to balance the genetic *a priori* information with the experienced information? This balance can be described by a term of "reactivity" that describes the relative weight of genetic and experienced information. A reactive species will give a higher weight to experienced condition information, whereas a stable species will give a higher weight to genetic information.

The way the two source of information are brought together and used to define the plant phenotype is at the core of plant strategy and is the main feature of the model *MountGrass*.

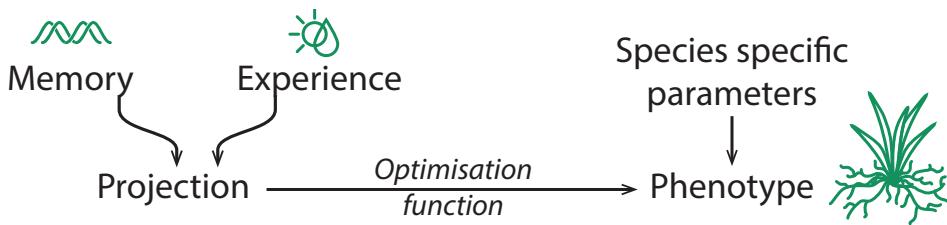


Figure 4.8: Genetic and perceived information are both considered to determine the phenotype.

4.3 Details

Further details on daily mechanisms are described in the following paragraphs.

4.3.1 Initialisation

The model doesn't need particular initialisation if the state of the community species pool, the seedbank, and the soil are given as inputs. Otherwise, a set of $E(n/s)$ individuals are created from a set of s species (randomly generated if not given) and randomly positioned on the soil grid, where s and n are respectively the number of species and the approximate number of individuals within the grid. Soil grid is also randomly generated within default ranges for critical and saturation water contents then slightly smooth, and homogeneously filled ($filling = \frac{w_{cont} - w_{crit}}{w_{sat} - w_{crit}}$).

4.3.2 Inputs

MountGrass needs system state information (individuals, species, seed-bank and soil) and climate data. If the state of the system is not completely given, then the complete state is generated in the initialisation. The daily climate data at must contain the following fields:

- *date*;
- *radiance*, in Watt.m^2 ;
- *precipitation*, in mm;
- *mean temperature*, in K;
- *mean day temperature*, in K;
- *min temperature*, in K;

Figure 4.9: Overview of the model inputs and outputs.

- max temperature, in K;
- relative humidity in %;

Vapour pressure deficit is then computed from temperature and relative humidity.

The climate data must explicitly differentiate the seasons (delimited by the first day of the year without snow and by the first day of the second semester with snow).

4.3.3 Submodels

Individuals from the seed-bank randomly germinate according to their species-specific germination rate. Germination consist of investing a percentage (*mob* parameter) of the seed mass into shoot and root biomass according to default traits. This is coupled with a round of random seed death following uniform law of parameter *seed_{surv}*. Living non germinating seeds stay in the seed-bank until the next season.

GERMINATION

Daily processes

Light competition is central to all vegetation model as it constrains the photosynthetic activity and so plant growth. To avoid costly calculation of ray propagation we assume vertical homogeneous top radiation. Relief and orientation effects are taken into account in the computation of irradiance data.

LIGHT COMPETITION

Light competition sub-model allows calculation of individual potential photosynthesis activity and light at soil surface for evaporation calculation.

Competition for light is calculated independently for each pixel, potential photosynthetic activity is then aggregated at the individual level. Each pixel can be seen as a column of homogeneous layers containing at least one individual (top layer). For each layer, the light transmission is computed based on leaf density.

LIGHT COMPETITION

$$I(h) = I_0 e^{-LAI(h)} \quad (4.1)$$

where $LAI(h)$ is the cumulative LAI at the bottom of layer l (between h and $h + \Delta_h$) defined as the homogeneous layer delimited by the top of consecutive individuals in the same pixel. The LAI is calculated like this:

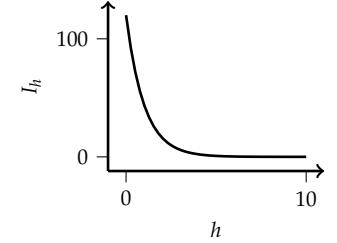


Figure 4.10: Net gain function and its first derivative.

Looks like there is some kind of mismatch here.

$$LAI(h) = LAI(h + \Delta_h) + \Delta_h \cdot \text{pix_width}^2 \sum_{i \text{ in } l} d_i \cdot \text{coverage}_{i,p} \quad (4.2)$$

where d_i is the individual leaf area density corrected by the coverage ($0 < \text{coverage} < 1$) of the pixel p by the plant i , $\Delta_h = (h_l - h_{l-1})$ is the height of the layer l .

Following Thornley and Johnson, the potential photosynthetic leaf activity is calculated as:

$$P_{leaf}(h) = \frac{\alpha \cdot I_{leaf}(h) \cdot P_{max}}{\alpha I_{leaf}(h) + P_{max}} \quad (4.3)$$

where $I_{leaf}(h)$ is the light absorbed by the leaf at height h , α the initial slope of the light response curve and P_{max} the maximum photosynthetic rate per unit of area and unit of time. I_{leaf} is the radiance at the leaf surface, derived by

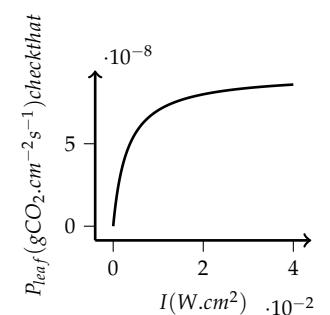


Figure 4.11: Photosynthetic saturation function

correcting the radiance at the top of the layer following the equation used in Taubert with the extinction and transmission coefficients k and m :

$$I_{leaf}(h) = \frac{k}{1-m} I(h) \quad (4.4)$$

The equation (4.3) can be integrated over the leaf surface by mixing it with equations (4.1) and (4.2) to give the total potential photosynthesis for layer l in pixel p :

$$P_{leaf}(p, l) = d_i.coverage_{i,p} \Delta_h(l) \int_{h_{bottom}}^{h_{top}} P_{leaf}(h) \quad (4.5)$$

the total leaf potential photosynthesis is then calculated as follow:

$$PS_{pot} = \sum_{p \text{ in shoot } l \text{ in pixel}} \sum P_{leaf}(p, l) \quad (4.6)$$

Potential photosynthesis must then be converted to potential transpiration to define the water demand. The conversion from photosynthesis to transpiration is done by dividing the potential photosynthesis by the water use efficiency (WUE). The potential activity of leaves are also dependent on the regulation of stomata so the transpiration can be written:

$$transp = \frac{PS_{pot} \cdot g_{red}}{WUE} \quad (4.7)$$

Photosynthesis depends on gazes exchanges at the leaf surface. These fluxes result from relative concentration in carbon dioxide and water, and from the stomatal conductance. Stomatal conductance is reduced and limits productivity when vapour pressure deficit is too high ¹. A linear relationship describe this relationship:

$$g_{red} = 1 + VPD_{g_red} \quad (4.8)$$

STOMATAL REGULATION

¹ g_{red} is set to 1 for current version to avoid potential problems between allocation and regulation

Potential evaporation is calculated for each pixel depending on the light at soil surface:

$$\beta = 0.25 * (1 - \cos(\frac{\theta}{\theta_{sat}} * \pi))^2 \quad if water_{cont} \leq water_{sat} \quad (4.9)$$

$$\beta = 1 \quad otherwise \quad (4.10)$$

$$PET = 0.0023 \sqrt{(T_{max} - T_{min})} * (T_{mean} + 17.8) \quad (4.11)$$

$$evap = PET \cdot \beta \cdot I_{surface} \cdot daylength \quad (4.12)$$

Water competition is also computed at the pixel level. To determine the water uptake, first the individual water demand is computed as the minimum between the transpiration and the potential water uptake. Transpiration demand per pixel is easily calculated by dividing the total potential transpiration by the volume in the pixel $V_{i,p}$ over the overall root volume V_i . Water potential uptake is the product of root area in the pixel and root water uptake rate reduced by the water availability reduction factor U_{lim} , leading to the water demand for

EVAPORATION

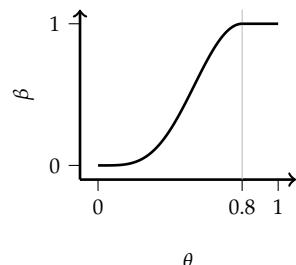


Figure 4.12: Evaporation limitation function.

WATER COMPETITION

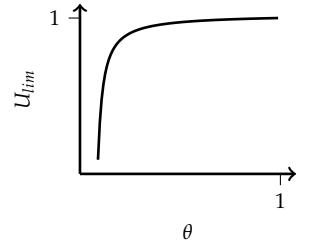
individual i in pixel p :

$$transp_i(p) = transp \cdot \frac{V_{i,p}}{V_i} \quad (4.13)$$

$$Wpot_i(p) = Root_{area}(p) \cdot U_{max} \cdot U_{lim} \quad (4.14)$$

$$Wdem_i(p) = \min(transp_i(p), Wpot_i(p)) \quad (4.15)$$

$$(4.16)$$



where, the limitation function U_{lim} is defined as in (Reineking et al. 2006):

$$U_{lim} = \exp \left(\beta_\theta \left(\frac{1}{\theta_s - \theta_{crit}} - \frac{1}{\theta - \theta_{crit}} \right) \right) \quad if \theta < \theta_{crit} \quad (4.17)$$

$$= 0 \quad otherwise \quad (4.18)$$

Figure 4.13: Water uptake limitation response function to soil saturation

The total water demand per pixel is then the sum of all individual water demand of the pixel and potential evaporation. If the total water demand exceeds the total water availability (W_{av} product of water content and soil volume in the pixel) then the available water is distributed proportionally to the individual demand.

$$Wup_i = Wdem_i \cdot \frac{Wdem_{total}}{\min(Wdem_{total}, W_{av})} \quad (4.19)$$

The potential water uptake (Wup_i), non limited by the transpiration is calculated the same way but considering $Wdem_i = Wpot_i$ in equation (4.19).

Because the water competition is computed at the pixel level, there is no compensation between two pixels containing respectively not enough and too much water.

No radial flow of water between pixel is implemented in the model. This simplification leads inevitably to edge effects, but allows simpler implementation and is partially covered by the effect of the pixel size. Indeed, increasing pixel size would have similar effect in the pixels at the border of the rooting zone than radial flow because it would increase the potential water pool plant has access to.

Once potential and realised transpiration and water uptake are computed, plant productivity can be calculated.

Following previous vegetation models, the respiration is decomposed in growth respiration and maintenance respiration. The first is function of trait values, biomass and temperature:

$$R_m = \left(R_{act} \cdot (Act_{ag} + Act_{bg}) \right) \cdot daylength \cdot T_{effect} \quad (4.20)$$

PRODUCTION, AND RESPIRATION

where R_{act} is the respiration rate of active tissues, and Act_{ag} and Act_{bg} are the active biomass pools in shoot and root.

Net Primary production (in CO_2 equivalent) can then be calculated the difference of GPP and respiration, then converted in OM production thanks to tissue carbon content (under the assumption of fixed carbon content for leaf and roots between species):

$$NPP_{carbon} = (1 - R_g) \cdot (WUE \cdot \min(w_{up}, transp) - R_m) - BM_{total} * Pl_{cost} \quad (4.21)$$

$$NPP_{OM} = NPP_{carbon} \cdot (12/44) / TCC \quad (4.22)$$

Here R_g is a fixed parameter but is set to 0 if the difference between gross productivity ($GPP = WUE \cdot \min(w_{up}, trans_p) - R_m$) and maintenance respiration is negative. Pl_{cost} is the plasticity cost as calculated in the dedicated paragraph below.

Temperature has a effect of plant activity, this effect can be modelled by a bell shape function around an optimum value of 20 °C. See Lohier for details.

TEMPERATURE EFFECT

The projection of environmental conditions is central in any implementaion of phenotype plasticity. Differences between the current perception of environment and the projections lead to adjustment of phenotype to increase fitness. In the model *MountGrass* this projection results from hte averaging of two key concept: memory and perception. The latter is relatively simple to understand and corresponds to the perceived resource availability computed as the mean potential exchange rate per unit of area (total leaf or root area) and per hour(the hourly measure is used instead of daily measure to simulate the ability of plant to perceive the photoperiod. This is an easy way of taking into account one aspect of seasonality without complicating the model. However, it also reduce the range of memory and its impact to determine the phenotype, as an additional information would be needed to define the optimum phenotype: the day length).:

$$light_{exp} = \frac{transp}{exchangearea_{ag}} \quad (4.23)$$

$$water_{exp} = \frac{Wpup}{exchangearea_{bg}} \quad (4.24)$$

$$(4.25)$$

The former is related to the species (or group) history and result from processes of selection and acclimation. It is the default projection of resource availability when the plant is not plastic.

$$light_{est}(t+1) = (1 - \tau).light_{exp}(t) + \tau.light_{memory}.daylength(t+1) \quad (4.26)$$

$$water_{est}(t+1) = (1 - \tau).water_{exp}(t) + \tau.water_{memory}.daylength(t+1) \quad (4.27)$$

Because these are supposed to be expected conditions for the future, other formulation can be used instead of an average that is likely to introduce a lag in estimations. For example the following equation allow for a more stable projection that better fits the slower process of plant physiology adjustments:

$$light_{est}(t+1) = ((1 - \tau_{react}).light_{exp}(t) + \tau_{react}.light_{est}(t))((1 - \tau_{amp}) + \tau_{amp}.light_{memory}).daylength(t+1) \quad (4.28)$$

with τ_{amp} and τ_{react} being respectively amplitude and reactivity where only τ_{amp} is used in the first equation. Such solution could limit sensitivity and phenotypic instability. IN addition, such formulation would also better capture the accumulation of stress signals and would lead to a softer and more stable phenotypic shift.

The estimation of external conditions as expressed here is then used to select the best allocation scheme during the allocation process. Limited here to levels of two resources (light and water), this estimation equation could be extended to other mechanisms such as herbivory risk, frost risk, humidity impact on water pressure deficit.

CONDITION ESTIMATION

Allocation is primordial in plant development and ontogeny. The following paragraph detail the implementation of the plastic allocation in *MountGrass*.

ALLOCATION

Maturity: For most of plants the development cycle is divided in two phases of different durations: the vegetative phase when plant growths organs to gather resources and product OM, and the reproductive phase when plant take advantage of these organs to accumulate carbon and invest them in reproduction mechanisms. Plants are considered mature (they switch from vegetative to reproductive phase) in *MountGrass* when the phenologic variable has reach a species specific threshold. The phenologic variable can be either the age, the height, the biomass, degree.days, in the current version total living biomass is used as trigger for reproductive phase.

Allocation to supporting tissues: Even-though grasses do not grow tall vegetative parts like trees, some grow vertically and they are exposed to stronger winds than most of forest. Therefore they need structural supports¹. Not all grasses grow stem, but they'll have stronger central vein in their leaves to structurally support the weight of leaves. In addition shoots and roots also need supporting tissues for water transport, for this reason the minimal mechanical support needed is calculated as a function of total living biomass:

$$\text{support} = \alpha \cdot (BM_{ag} + BM_{bg})^\gamma \quad (4.29)$$

where α and γ are allometry coefficients.

At each time step we must determine what fraction of new OM will be allocated to tissues growth while the remaining will support these need tissues. This leads to an optimisation problem numerically solved by the function `uniroot`.

¹ This supporting tissue mechanic is also needed to avoid exponential growth rate.

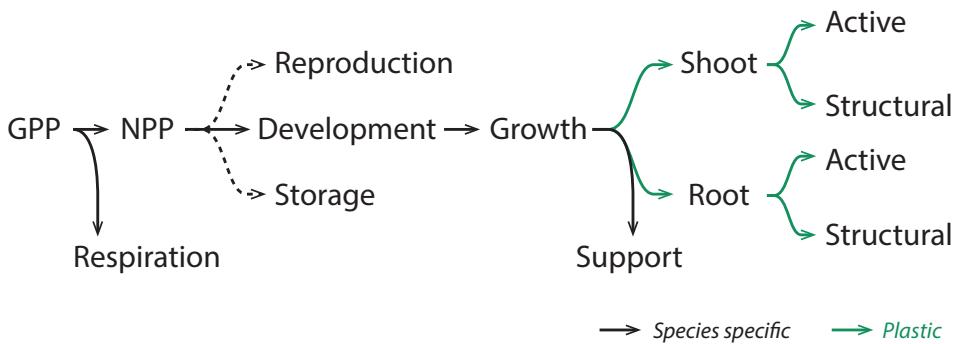


Figure 4.14: Allocation of produced organic matter to different processes and pools.

Allocation to organs: Allocation of produced organic matter is central in vegetation as it shapes the plant and define the strength of the different organs. There are multiple ways to model the distribution of produced organic matter between the plant organs. We believe that such mechanism has great impact on individual development and response to external conditions, and so on community dynamics. To explore the role of this mechanism, multiple options are implemented. The different allocation algorithms are summarised in table 4.3.

There are two major components in the allocation algorithm:

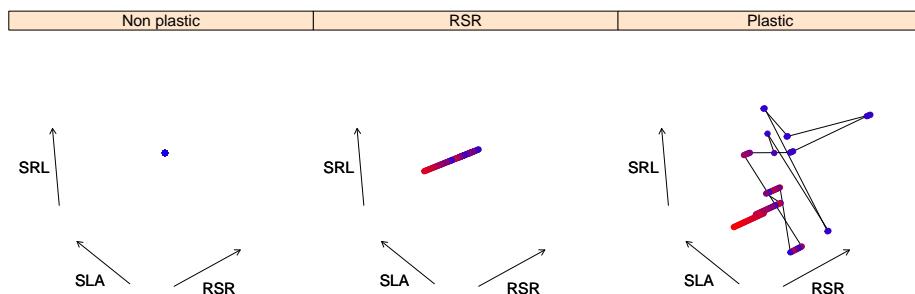
- the objective function;

- the plastic dimensions.

The *objective function*: it is the function that give an fitness estimation or gain metrics for any given phenotype. This function is used to compute the optimum phenotype (phenotype at which the function is evaluated at the maximum value), or rank alternative phenotypes¹.

The *plastic dimensions*: they are the dimensions along which the individual can move. The space defined by these dimensions is the phenotypic space within which each individual plant can look for an alternative phenotype. They do not necessarily fully define a phenotype since some dimensions of the individual's phenotype can be fixed².

The objective of this step of the model is to solve the objective function with



¹ in this case, if not all possible phenotypes are tested, the solution might be only a local optimum. This is the case in *MountGrass*.

² either by shared parameters of species specific ones.

Figure 4.15: Trajectories of a plant in the trait space depending on the plastic dimensions explored.

the unknown variables being the plastic dimensions (RSR, SLA and SRL). In case of simple equations an analytical solution could be used to find an optimum³. However, because the analytical solutions are already non trivial and the model is likely to evolve, a numeric solving method is adopted. **Need to detail the random algorithm.**

Also make a note on multiple optimum and the choice for a 'gradient descent' type of algorithm. Also sensitivity at early stages

Algorithm	Objective	variable RSR	variable SLA-SRL	stochastic
No plasticity	—	○	○	○
Equilibrium	functional eq.	●	●	●
Eq-Fixed	functional eq.	●	○	●
Optimisation	instantaneous gain	●	●	●
Optim-Fixed	instantaneous gain	●	○	●

³ under the condition that such optimum exists. The design of the model should ensure that.

Table 4.3: Allocation algorithms implemented in *MountGrass*

No plasticity allocation: this allocation is very similar to classic vegetation model where the biomass is allocated to the different carbon pools according to species specific parameters. But *MountGrass* differs from other models by the order of the different steps of growth. In this model, the senescence comes between the allocation step and the resource competition-production steps⁴. The partitioning coefficient are directly computed from species default trait to maintain the phenotype after senescence.

Fixed trait allocation: The fixed allocation supposes the allocation on OM to maintain trait values to fixed species specific values. The shoot:root ratio

⁴ see plastic allocation algorithm for explanation

may however change to maintain functional equilibrium. The shoot root ratio is derived from the following equation of the functional equilibrium:

$$SLA \cdot BM_{ag} \cdot light_{est} = SRL \cdot BM_{bg} \cdot water_{est} \quad (4.30)$$

$$\frac{BM_{ab}}{BM_{bg}} = \frac{SRL}{SLA} \cdot \frac{water_{est}}{light_{est}} \quad (4.31)$$

where $light_{est}$ and $water_{est}$ are the estimated resource availabilities.

Plastic trait allocation: Another approach to allocation is to try to optimize phenotype based on a fitness proxy. This proxy can be the sum of NPP, tissue turn-over loss and plasticity cost. But in a complex model like *MountGrass*, plant performance is function of multiple aspects:

- individual organ efficiency;
- relative mass of each organ;
- balance between organ water exchange activities.

And this could be extended to herbivory or frost risks. To take into account all these components, and take advantage of having all processes already made explicit by the implementation in the model, the daily processes of senescence and production are recalculated according to the **estimation of conditions** and the plant phenotype. This function is used to rank different alternative phenotypes (algorithm detailed below).

Plastic trait equilibrium: An alternative approach can be easily derived from the previous one and extend the principle of the first: the functional equilibrium with plastic traits. This approach consists in using the same algorithm as before but rank phenotypes with a function negatively correlated to the difference between estimated shoot and root activity. Such mechanism would nonetheless require the algorithm to look for close solutions within the allocation space to avoid convergence or drift from species strategy. Having non zero cost of plasticity in this approach should limit the drifting of the plant phenotype.

Fixed trait optimisation: This algorithm takes the idea of the optimisation algorithm but limits the plastic traits to the RSR ratio. If we can expect similar response than the fixed trait equilibrium if we suppose that the equilibrium is the main aspect of plant performance, global efficiency being considered in this case the result may vary.

Alternative phenotypes are computed from the actual phenotype and random uniform distribution of available organic matter to the main active and structural carbon pools of the plant.¹ ... This algorithm has the advantage of being relatively cheap compared to other optimization functions, however, its performances are variables and it is very sensitive to the number of samples used. As a consequence there is a trade-off between model stability and performance as a function of the number of samples (*i.e.* alternative phenotypes) considered.

The limits and costs of plasticity have long been discussed in the related literature. If *MountGrass* is intended to be used to examine ecological costs and limits, it has to include physiological aspects of plasticity limits. There are two physiological processes involved in the mechanism of altering a phenotype based on changes in external conditions: sensing and signalling. 'Sensing' relates to the capacity of the individual to perceive environmental conditions. This is related

PLASTIC ALGORITHM

¹ talk about the order senescence production, and the way exchange rates are computed.

PLASTICITY COST

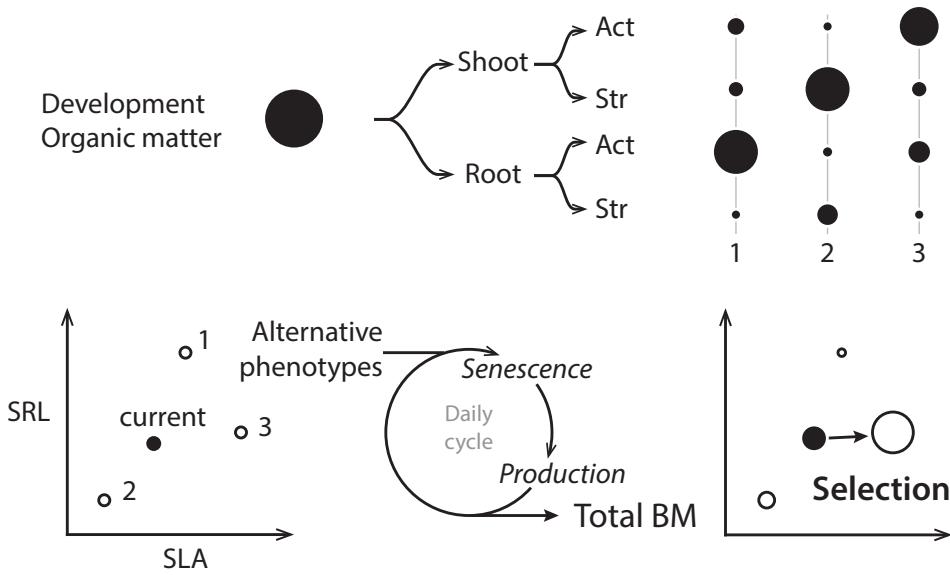


Figure 4.16: Algorithm for the evaluation and selection of randomly generated alternative phenotypes.

to the capacity of the individual to perceive the environment and should, therefore, be considered constant over time. To take into account the cost of precise sensing, the first component of the plasticity cost is proportional to τ .

The other component is related to the capacity of the plant to transmit this knowledge of conditions to change the development plan toward a new phenotype. This cost is proportional to the carbon-based distance (calculated as the difference between proportion of active tissues) between the default phenotype and the alternative (during allocation algorithm) or current phenotype.¹

Plasticity cost is the sum of both component and is proportional to the total biomass since most of the tissues should have the appropriated cell machinery and are affected b plasticity.

¹ We could imagine cost based not on the default, but the previous phenotype, but it would have lead to large phenotypic shifting and convergence.

$$pc_{maintenance} = (1 - tau) * pc_m \quad (4.32)$$

$$pc_{plasticity} = d_{traits} * pc_p \quad (4.33)$$

where d_{traits} is the Euclidean distance between default phenotype and the alternative phenotype in the space defined by the proportion of active tissue for shoot and for roots.

Plasticity in trait suggests that trait values are modified in time. Because plants are described by single values (e.g. one SLA value for all leaves), this values must be updated after the plastic allocation. This values could be updated as the average of old tissue value weighted by old biomass and new tissue value weighted by the freshly produced biomass. This, however, would work only if active on structural tissues ratio linearly linked to others traits. This is not the case, it is then simpler to consider that organs have uniform active and structural distribution. This hypothesis suggests that whenever the allocation scheme change, old tissue reallocate their own biomass to follow the new scheme. Nevertheless, to avoid full plasticity allowed by this hypothesis, the changes in trait carbon pool sizes are limited by the produced biomass available for plant development.

following the following survival probabilities:

From this, supposing homogeneous distribution of active and structural tis-

TRAIT UPDATE

sues within an organ allows to directly link the size of the carbon pools to average traits by the following relationships:

$$SLA = \frac{1}{(th.p_{act_{shoot}} \cdot \rho_{as} + th.(1 - p_{act_{shoot}}) \cdot \rho_{ss}) \cdot V_t} \quad (4.34)$$

$$SRL = \frac{1}{(s_r.p_{act_{shoot}} \cdot \rho_{ar} + s_r.(1 - p_{act_{shoot}}) \cdot \rho_{sr})} \quad (4.35)$$

Senescence is the process of ageing of tissues. This process usually occurs at the scale of an individual organ (e.g. a leaf), however, *MountGrass* does not consider organs independently because it would be complex and computationally expensive to follow multiple leaves and roots for all individuals. So the process is considered homogeneous over all tissues. To emulate the senescence process senescence is calculated from the tissues lifespan, giving :

$$sen_{leaf} = \frac{1}{LLS} \quad (4.36)$$

$$sen_{root} = \frac{1}{RLS} \quad (4.37)$$

Because *MountGrass* does not contain any mechanism preventing plant from growing only active tissues¹, it is necessary for this cost function to make this strategy unreliable. The is then expressed as follow:

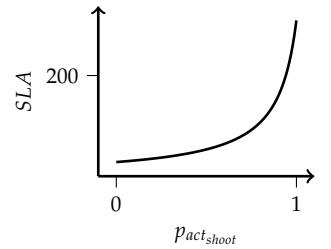


Figure 4.17: Specific Leaf Area as a function of the proportion in active tissues in shoot

SENECENCE

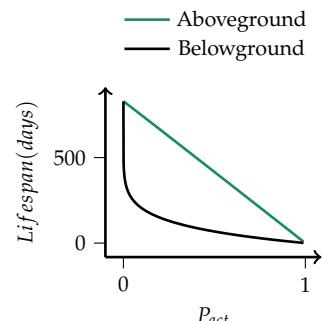


Figure 4.18: Lifespan of organs as a function of proportion of active tissues.

¹ it was intended to make the WUE negatively correlated to the amount of structural tissue per area.

$$LLS = LSS_{s0} * (1 - p_{act_{shoot}}^{LSS_1}) \quad (4.38)$$

$$RLS = LSr_{s0} * (1 - p_{act_{root}}^{LSR_1}) \quad (4.39)$$

where LLS and RLS are respectively the leaf and the root lifespans calculated as negative log-linear relationships with the proportion of active tissue.

Root senescent tissues disappear from the system. Information about senescent aboveground biomass is stored, but senescent biomass effect of light competition is ignored in this version because as it is implemented senescent tissues appear early in plant development and have large negative effect on light absorption.

To the natural senescence and artificial cost of having only active tissue, an additional component can be added to the turn-over rate: the negative NPP. In case of negative NPP, the biomass will be taken from the already allocated following the shoot:root ratio. This can lead to a lower overall productivity (negative growth during unproductive periods) but also changes in the equilibrium if tissue have different efficiencies.

Death is modelled as in Reineking (Reineking et al. 2006). Age and desiccation (negative NPP) are the two reasons why a plant can die. The two death mechanism are simulated by independent random lotteries following the following survival probabilities:

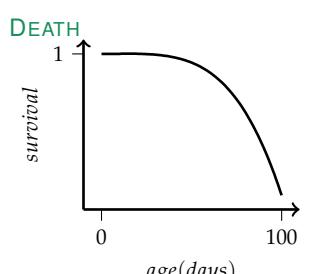


Figure 4.19: Age related survival probability function

$$P_d = \exp \left(- \left[\left(\frac{des}{\alpha_d} \right)^{\gamma_d} - \left(\frac{\max(des-1, 0)}{\alpha_d} \right)^{\gamma_d} \right] \right) \quad \text{if } NPP \leq 0 \quad (4.40)$$

$$= 1 \quad \text{otherwise} \quad (4.41)$$

$$P_a = \exp \left(- \left[\left(\frac{age+1}{\alpha_a} \right)^{\gamma_a} - \left(\frac{age}{\alpha_a} \right)^{\gamma_a} \right] \right) \quad (4.42)$$

State of dead individuals is stored until the end of the season when seeds are stored in the seed bank. Seeds of dead individuals then join other seeds.

Sexual & clonal reproduction: reproduction is handled at the end of the season. To limit the number of parameters reproduction is limited to the division of the invested biomass in reproduction by the species-specific seed biomass into a round number of seeds (the number of seeds per plant could also be a differentiation axis). Clonal reproduction is not explicitly represented but can be mimicked with bigger seeds and by adding a dispersion process around the parents. The seeds then are added to a potential seed-bank. This potential seed-bank is sampled, after eventual invasion, and merged with the existing seed-bank.

REPRODUCTION & PERSISTENCE

Persistence Some grasses are perennial and persist over the cold season. This is allowed in the model by investment in storage tissues instead of reproductive tissues. At the end of the season, marked by the first snowfall, these plants (with non-null storage biomass) lose their living and supporting biomass, but will regrow from a large pool of stored organic matter.

Explore management effect on the community is one of the aims of the *MountGrass* model. The management of mountain grassland will be explored only of the aspect of biomass removal, as productivity changes can be explored by changing the parameter values as the nutrients are not explicitly modelled. The management sub-model is not detailed here but it is based on the mapping of biomass and target trait (e.g. the fraction of structural biomass as a proxy for digestibility). Both cutting and grazing can be modelled but require management plan in the form of calendar of management operation and a cutting height or harvest objective.

GRAZING/CUTTING

4.4 Limitations and problems

4.4.1 Link to the real world and data

The generalized framework introduced in *MountGrass* allows to create a rich community in a high number of dimension strategy space, it, however, comes with downsides.

One of the first problems is that some parameters (not explicitly detailed here) are hard to access (e.g. tissue density of active, or structural, tissue). It makes the calibration long as the incertitude for some parameters is very high. This is problematic when calibration is made difficult by a large execution time (see subsection below).

Another issue with such model is that the high dimensionality of the species strategy space allows a lot of different strategies that are not viable. This could

be overcome by selection mechanism over multiple plots, but again require a lot of simulation. Moreover, there are dependencies between viable strategies and parameter values that make it hard to restrict meta-community to viable species to set-up calibration runs.

It is possible to extract summary statistics from the model output and compare them to information from collected data making calibration and community analysis easy. However going from the data to feed the model is harder, indeed without a great knowledge of a species it is hard to define its representation within the model framework. To do so would require the knowledge of the plasticity capacity to set the reactivity, anatomical traits to define default ratios of active over structural tissues, and climatic niche to define the *a priori* estimation of external conditions. Without making a direct association with real species, it is possible and interesting to try to reproduce some strategies and explore their response to various conditions.

4.4.2 Technical problems

The model is implemented in R with some limiting function using RCPP to speed up the process. Simulations are fairly slow compare to theoretical C++ equivalent code. The main problem is the choice of the data structure. Indeed agents are stored in data.frames that are often modified with the `mutate` function, that makes the implementation much easier and the code readable, but slow down the execution due to constant condition checking on operations. This makes calibration routine methods almost impossible to use as they demand a very number of runs to be efficient.

The slowness of the model also limit to simple algorithms for the research of favourable positions in the allocation space.

BIBLIOGRAPHY

- Adler, Peter B. et al. (2006). "Climate variability has a stabilizing effect on the coexistence of prairie grasses". en. *PNAS* 103.34, pp. 12793–12798.
- Auld, Josh R., Anurag A. Agrawal, and Rick A. Relyea (2009). "Re-evaluating the costs and limits of adaptive phenotypic plasticity". en. *Proceedings of the Royal Society of London B: Biological Sciences*, rspb20091355.
- Bello, Francesco de et al. (2013). "Hierarchical effects of environmental filters on the functional structure of plant communities: a case study in the French Alps". en. *Ecography* 36.3, pp. 393–402.
- Bradshaw, A. D. (1965). "Evolutionary Significance of Phenotypic Plasticity in Plants". In: *Advances in Genetics*. Ed. by E. W. Caspari and J. M. Thoday. Vol. 13. Academic Press, pp. 115–155.
- Bradshaw, Anthony D. (2006). "Unravelling phenotypic plasticity – why should we bother?" en. *New Phytologist* 170.4, pp. 644–648.
- Cai, Qingsheng et al. (2004). "Changes in Freezing Tolerance and its Relationship with the Contents of Carbohydrates and Proline in Overwintering Centipedegrass (*Eremochloa ophiuroides* (Munro) Hack.)" *Plant Production Science* 7.4, pp. 421–426.
- Chesson, Peter (2000). "General Theory of Competitive Coexistence in Spatially-Varying Environments". *Theoretical Population Biology* 58.3, pp. 211–237.
- Craine, J. M. et al. (2002). "Functional traits, productivity and effects on nitrogen cycling of 33 grassland species". en. *Functional Ecology* 16.5, pp. 563–574.
- Deléglise, Claire et al. (2015). "Drought-induced shifts in plants traits, yields and nutritive value under realistic grazing and mowing managements in a mountain grassland". *Agriculture, Ecosystems & Environment* 213, pp. 94–104.
- DeWitt, Thomas J., Andrew Sih, and David Sloan Wilson (1998). "Costs and limits of phenotypic plasticity". *Trends in Ecology & Evolution* 13.2, pp. 77–81.
- Díaz, Sandra et al. (2016). "The global spectrum of plant form and function". en. *Nature* 529.7585, pp. 167–171.
- Dwyer, John M., Richard J. Hobbs, and Margaret M. Mayfield (2014). "Specific leaf area responses to environmental gradients through space and time". *Ecology* 95.2, pp. 399–410.
- Freschet, Grégoire T., Peter J. Bellingham, et al. (2013). "Plasticity in above- and belowground resource acquisition traits in response to single and multiple environmental factors in three tree species". en. *Ecol Evol* 3.4, pp. 1065–1078.
- Freschet, Grégoire T., Emilie Kichenin, and David A. Wardle (2015). "Explaining within-community variation in plant biomass allocation: a balance between organ biomass and morphology above vs below ground?" en. *J Veg Sci*, n/a-n/a.
- Grime, J. P. (1998). "Benefits of Plant Diversity to Ecosystems: Immediate, Filter and Founder Effects". *Journal of Ecology* 86.6, pp. 902–910.
- Grimm, Volker et al. (2006). "A standard protocol for describing individual-based and agent-based models". *Ecological Modelling* 198.1–2, pp. 115–126.
- Hu, Yu-Kun et al. (2015). "Novel evidence for within-species leaf economics spectrum at multiple spatial scales". *Front. Plant Sci.* P. 901.
- John, Grace P. et al. (2017). "The anatomical and compositional basis of leaf mass per area". en. *Ecol Lett* 20.4, pp. 412–425.

- Kichenin, Emilie et al. (2013). "Contrasting effects of plant inter- and intraspecific variation on community-level trait measures along an environmental gradient". en. *Funct Ecol* 27.5, pp. 1254–1261.
- Kleidon, Axel and Harold A. Mooney (2000). "A global distribution of biodiversity inferred from climatic constraints: results from a process-based modelling study". en. *Global Change Biology* 6.5, pp. 507–523.
- Laughlin, Daniel C. (2014). "The intrinsic dimensionality of plant traits and its relevance to community assembly". en. *J Ecol* 102.1, pp. 186–193.
- Lohier, Théophile et al. (2014). "Explaining ontogenetic shifts in root–shoot scaling with transient dynamics". en. *Ann Bot* 1, mcu128.
- Maire, Vincent (2009). "Des traits des graminées au fonctionnement de l'écosystème prairial : une approche de modélisation mécaniste". français. PhD thesis. Université Blaise Pascal - Clermont-Ferrand II ; Université d'Auvergne - Clermont-Ferrand I.
- Maire, Vincent, Nicolas Gross, et al. (2013). "Disentangling Coordination among Functional Traits Using an Individual-Centred Model: Impact on Plant Performance at Intra- and Inter-Specific Levels". *PLoS ONE* 8.10, e77372.
- Maire, Vincent, Jean-François Soussana, et al. (2013). "Plasticity of plant form and function sustains productivity and dominance along environment and competition gradients. A modeling experiment with Gemini". *Ecological Modelling* 254, pp. 80–91.
- Mediavilla, S., A. Escudero, and H. Heilmeier (2001). "Internal leaf anatomy and photosynthetic resource-use efficiency: interspecific and intraspecific comparisons". en. *Tree Physiol* 21.4, pp. 251–259.
- Picon-Cochard, Catherine et al. (2012). "Effect of species, root branching order and season on the root traits of 13 perennial grass species". en. *Plant Soil* 353.1-2, pp. 47–57.
- Poorter, Hendrik et al. (2009). "Causes and consequences of variation in leaf mass per area (LMA): a meta-analysis". en. *New Phytologist* 182.3, pp. 565–588.
- Reich, P. B., M. B. Walters, and D. S. Ellsworth (1992). "Leaf Life-Span in Relation to Leaf, Plant, and Stand Characteristics among Diverse Ecosystems". *Ecological Monographs* 62.3, pp. 365–392.
- Reich, Peter B. (2014). "The world-wide 'fast–slow' plant economics spectrum: a traits manifesto". en. *J Ecol* 102.2, pp. 275–301.
- Reich, Peter B. et al. (2003). "Variation in Growth Rate and Ecophysiology among 34 Grassland and Savanna Species under Contrasting N Supply: A Test of Functional Group Differences". *New Phytologist* 157.3, pp. 617–631.
- Reineking, Björn et al. (2006). "Environmental variability and allocation trade-offs maintain species diversity in a process-based model of succulent plant communities". *Ecological Modelling*. Pattern and Processes of Dynamic Mosaic Landscapes – Modelling, Simulation, and Implications 199.4, pp. 486–504.
- Ryser, P. (1996). "The Importance of Tissue Density for Growth and Life Span of Leaves and Roots: A Comparison of Five Ecologically Contrasting Grasses". *Functional Ecology* 10.6, pp. 717–723.
- Ryser, Peter and Liina Eek (2000). "Consequences of Phenotypic Plasticity vs. Interspecific Differences in Leaf and Root Traits for Acquisition of Aboveground and Belowground Resources". *American Journal of Botany* 87.3, pp. 402–411.
- Shipley, Bill et al. (2006). "Fundamental trade-offs generating the worldwide leaf economics spectrum". *Ecology* 87.3, pp. 535–541.
- Soussana, Jean-François et al. (2012). "Gemini: A grassland model simulating the role of plant traits for community dynamics and ecosystem functioning. Parameterization and evaluation". *Ecological Modelling* 231, pp. 134–145.
- Taubert, Franziska (2014). "Modelling and Analysing the Structure and Dynamics of Species-rich Grasslands and Forests". PhD thesis. Osnabrück.
- Tjoelker, M. G. et al. (2005). "Linking Leaf and Root Trait Syndromes among 39 Grassland and Savannah Species". *New Phytologist* 167.2, pp. 493–508.
- Westoby, Mark (1998). "A leaf-height-seed (LHS) plant ecology strategy scheme". en. *Plant and Soil* 199.2, pp. 213–227.

- Westoby, Mark, David Warton, and Peter B. Reich (2000). "The Time Value of Leaf Area." *The American Naturalist* 155.5, pp. 649–656.
- Wright, Ian J. et al. (2004). "The worldwide leaf economics spectrum". en. *Nature* 428.6985, pp. 821–827.

IV

INDIVIDUAL PERFORMANCE: STRATEGY AND PLASTICITY

The chapter contains the main results from simulations experiments at the individual scale. It provides insight on the impact of plastic allocation algorithm on individual growth and potential effects on community properties.

The first part is dedicated to the parameter filtering and the study on individual growth in a stable environment. The second part examines response of individual root strategies to two gradients of water availability: (1) with constant influx but differences in means simulating spatial heterogeneity, (2) with shared mean influx, but contrasting rate of reduction of precipitation simulating the reduction of available resource during the growing season.

1 MODEL PROPERTIES AND INDIVIDUAL RESPONSES

The first part of the chapter is dedicated to the parameter filtering process, the sensitivity analysis and basic model behaviour.

1.1 Parametrisation and sensitivity analysis

Calibration, or **parametrisation**, is an essential step in the development of an agent-based model. ABMs are often characterised by multiple processes, and though parameters, at individual levels. The results of these processes (depending of parameter values) from numerous individuals combine to produce the group or community behaviour. Because there are interactions between the processes and between the agents, the overall behaviour of the group (often the subject of interest) is sensitive to these parameters. For the same reasons, an incredible variety of results could be produced with ABMs if the parameters were not chosen in order to produce sensible responses to simulated conditions. The aim of the calibration is to determine, from the *a priori* knowledge of the processes and parameters, and the comparison with data, the best values for the model parameters. This step often goes along with a sensitivity analysis that determine the relative sensitivity of variables of interest to specific parameters.

Because of their nature, ABMs often model processes for which the parameters are either unknown, or hard to access (because at the individual scale). In such cases, advance calibration techniques like pattern oriented modelling(hartig; Grimm et al. 2005) can be developed. However, such method require a high number of simulations and relatively precise simulation parameters. Because the implementation in R makes the model relatively slow, and because available datasets, despite being very interesting lack information on sensitive parameters, a less robust but less expensive approach is chosen: **parameter filtering** at the individual scale. The focus of the part of this work on the individual growth, and the will for more individual-centric approach also support this choice.

For similar reasons of computational cost, the **sensitivity analysis** is realised *a posteriori* on calibration runs.

1.1.1 Method

Pot data consists in total biomass and root shoot ration (RSR) data of 11 species grown in pots by Peterson and Billings (Peterson and Billings 1982). This dataset has the advantages of being grass species grown in a described steady environment with two conditions of watering with measures of essential components of growth: biomass and RSR.

POT DATA

Simulated plant grow in square pots 9 cm wide and 12 cm deep. The soil is characterised by the following parameters: critical soil water content: $0.1m^3.m^{-3}$, and saturation water content: $0.1m^3.m^{-3}$. Simulation time of 111 days of 15 hours is divided between the growing phase of 48 days, followed by the treatment phase when plant are water (soil saturation) either once a week or once a day. The light level and water influx are simulated following the experimental conditions (Peterson and Billings 1982) by a lighting of 1850 Watts per square meter, and soil saturation. Plants have default geometry parameters, reproduction is ignored and it is assumed that plants do not stop their growth.

POT SIMULATION

The whole filtering process has been implemented in R. Model parameters are sampled following the LHS method (from `lhs` package) within parameter ranges (described in table 1.1) defined both thanks to the literature and constraints dictated by desired behaviours from the model. When necessary the sample is log transformed. Because of strong relationship between exchange rate parameters and cost of exchange area, exchanges rates parameters are expressed on a mass basis for sampling then transformed into an area basis for the model. To avoid extreme RSR ratios, the ratio between the mass based exchange rate parameters is limited between 0.1 and 10.

PARAMETER FILTERING PROCESS

As explained in previous chapter, species specific parameters are required to model plant growth. These parameters are sampled at the same time that the parameters of the model, according to ranges detailed in table chapter III, 4.2. Once the parameters are generated, a first filtering is applied to save simulation time and avoid unrealistic trait values. The computed initial trait values considered out of range (see table ?? for ranges extracted from LES data (Wright et al. 2004) in the alpine biome) are excluded, modifying the initial distribution of the parameter values (see figure 1.2). These two steps lead to the creation of a list of n independent parameter sets that are then used for individual pot simulations following Peterson and Billings 1982 experiment setup.

The results from the finished simulations (i.e. the plant lives until the end and do not exceed model's internal size limits) are then compared to the experiment data species by species. The parameters of logistic distributions are computed from the species means and standards deviations for RSR and total biomass. The use of this distribution form is justified by the intrinsic form of the RSR variable and the need to reject negative values for total biomass variable. A parameter set is accepted for one species if it lies within a 95% range of the calculated distribution for both RSR and total biomass in wet and dry conditions.

The parameter filtering procedure is applied on the three main allocation algorithms: *non plastic*, *fixed-equilibrium* and *plastic-optimisation*.

Relative importance of variables in the selection process is investigated with the packages `randomForest`. A random forest analysis (depth = 5, number of trees = 300) is performed on a balance dataset composed by all selected parameter sets

SENSITIVITY ANALYSIS

Table 1.1: Global parameters of *MountGrass* with units and extreme values used during the parameter filtering process.

name	min	max	unit	full name
u_max	0.36	10	cm ³ .cm ⁻² .h ⁻¹	Maximum root uptake rate
beta_o	0.002	0.2	AU	Soil absorption limitation strength
P_max	0.00001	0.0001	gCO ₂ .cm ⁻² .s ⁻¹	Maximum photosynthesis
alpha	0.00001	1.0001	AU	Photosynthesis curvature
mob	0.0005	1	fraction total green biomass	Maximum growth rate
m	0.1	0.5	AU	Leaf light transmittance
r_g	0.1	0.5	gC.gMO ⁻¹ .h ⁻¹	Growth respiration rate
r_i	0.003	0.03	gC.gMO ⁻¹ .h ⁻¹	Active tissue respiration rate
ls_so	5.7658	7.9628	day	Log of maximum shoot lifespan
ls_s1	-1.2325	0	day	Shoot lifespan slope
ls_ro	4	7	day	Log of maximum root lifespan
ls_r1	-1.5	0	day	Root lifespan slope
sd_s_rate	0.05	1	per year	Seed survival rate
WUE	0.001	0.01	GCO ₂ .gH ₂ O ⁻¹	Water Use efficiency
LCC	0.39	0.5	gC.gOM ⁻¹	Leaf carbon content
alpha_d	10	30	AU	Drought mortality
gamma_d	1	3	AU	Drought mortality
th	0.0124	0.0437	cm	Leaf thickness
s_r	0.0019	0.05	cm ²	Root section (area)
rho_as	0.005	0.1	g.cm ⁻³	Volumic mass of shoot active tissue
rho_ss	0.8	1.5	g.cm ⁻³	Volumic mass of shoot structural tissue
rho_ar	0.005	0.1	g.cm ⁻³	Volumic mass of root active tissue
rho_sr	0.8	1.5	g.cm ⁻³	Volumic mass of root structural tissue
vt_s	0.7	0.75	AU	Volume occupied by the tissue in total leaf volume
k_os	0.001	0.01	cm ³ .cm ⁻³	Shoot volume occupancy
k_or	0.01	0.5	cm ³ .cm ⁻³	Root volume occupancy
k	0.4	0.6	AU	Light extinction parameter

trait	min	max	unit
SLA	20	400	cm ² .g ⁻¹
SRL	1000	15000	cm.g ⁻¹
LLS	10	400	days
RLS	100	1200	days

Table 1.2: Extreme values of traits related to exchange area per biomass and organ longevity for both shoot and root.

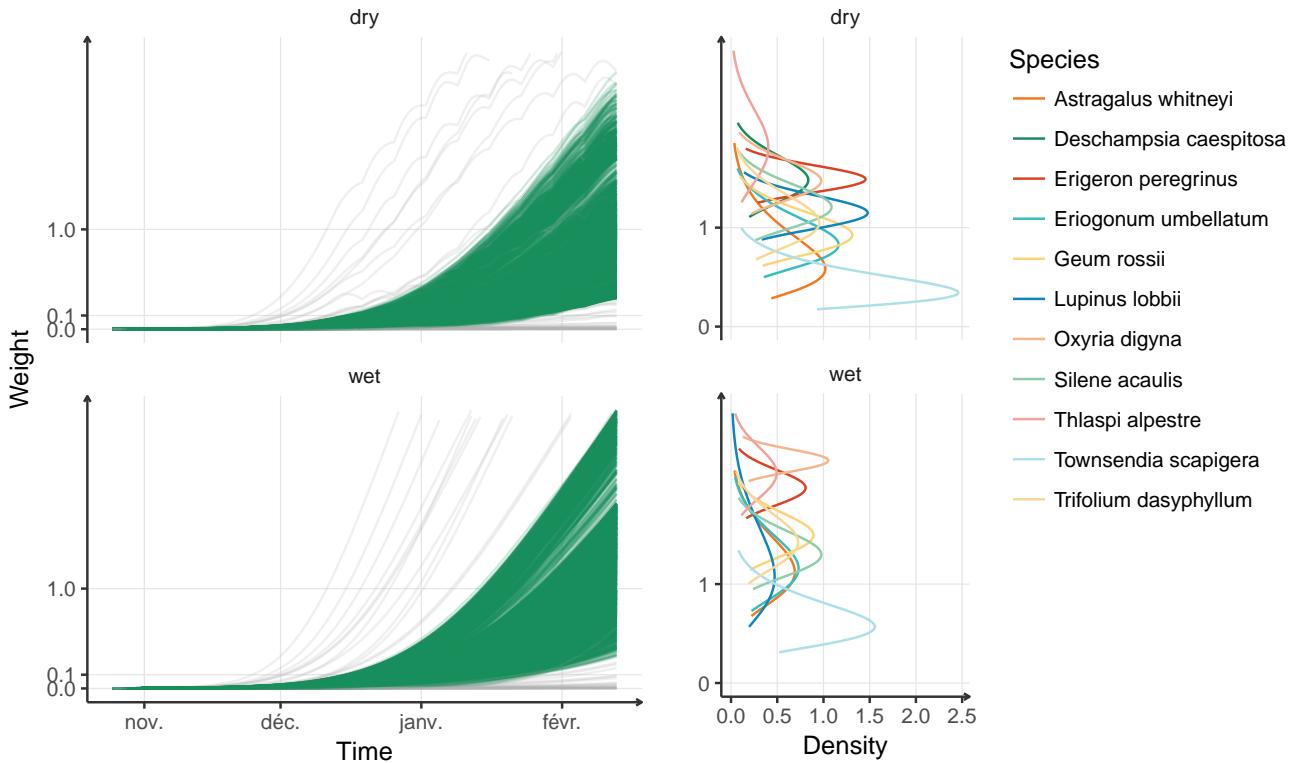


Figure 1.1: Comparison of simulated weights with distribution of weights of real alpine species for contrasting conditions.

and a random sample of rejected sets of equal size. Importance is assessed on the results of the random forest.

1.1.2 Results

Parameter filtering process resulted in the selection of a low number of parameter sets (below 0.2%) for each allocation algorithms (table 1.1.2). This number is below the sum of accepted parameter sets per species because a parameter set can match to multiple species. Not all species contribute to the same extend to the filtering process. *Astragalus whitneyi* accounts for a high percentage of accepted parameter sets, while no parameter set could match 2 species (*Oxyria dignya* and *Deschampsia caespitosa*). The former is characterised by wide distribution in both conditions for the two variables of interest (weight and RSR), while the latter show relatively tight distribution with little overlap between the conditions for the both variables (see figure 1.1 for comparison between simulations and data for total weight).

SELECTION RATE

Despite the low selection rate, a difference can be noted between the *fixed-equilibrium* algorithm and the two other algorithms with a accepted rate of 0.14 % against 0.09% and 0.10% (table 1.1.2). This difference cannot be explained by a significantly better selection rate for specific species, but rather higher rates for all species.

Most of parameter sets are not shared between the algorithms (*i.e.* around respectively and third and a quarter of accepted parameter sets are shared between *non plastic* allocation and *fixed-equilibrium* allocation calibrations), despite that the distribution of parameter values that are not shared are very similar and

Table 1.3: Acceptance rate per species for the 3 main allocation algorithms. Because some parameter sets match multiple species, the total number and rate of accepted parameter sets is lower than the sum of accepted parameter sets per species. All rates are given in %.

species	non plastic		fixed-eq		plastic	
	n (2M)	rate	n (2M)	rate	n (200,000)	rate
Silene acaulis	227	0.02	396	0.04	55	0.03
Trifolium dasypphyllum	271	0.03	317	0.03	45	0.02
Geum rossii	51	0.01	72	0.01	12	0.01
Thlaspi alpestre	342	0.03	360	0.04	59	0.03
Deschampsia caespitosa	-	-	-	-	-	-
Eriogonum umbellatum	500	0.05	805	0.08	118	0.06
Townsendia scapigera	593	0.06	930	0.09	107	0.05
Astragalus whitneyi	1570	0.016	2424	0.24	318	0.16
Lupinus lobbii	678	0.07	868	0.09	123	0.06
Erigeron peregrinus	1	<0.01	-	-	-	-
Oxyria digyna	-	-	-	-	-	-
Total	4233	0.43	6172	0.62	837	0.42
Accepted	924	0.09	1416	0.14	200	0.10

do not show any clear pattern (data not shown).

Out of the 31 parameters, 6 show graphical response of selection rate (see figure 1.2), and only u_{max} and P_{max} present a possible optimum different from limit values. The relative importance of the parameters is better explored in sensitivity analysis.

A total of 12 parameters show a relative influence on selection rate for at least one of the algorithm. These parameters are divided between model parameters and species parameters. Species parameters show influence only for the *non plastic* allocation algorithm. Model parameters express relatively similar importance for all three algorithms. The respiration rate of active tissues (r_1) is the most sensitive parameters (see figures 1.2 and 1.1.2). Other sensitive parameters are related to water availability (β_0), organ exchange rates (P_{max} and u_{max}) and soil coverage by roots (ρ_{ar} and k_{or}).

SENSITIVITY ANALYSIS

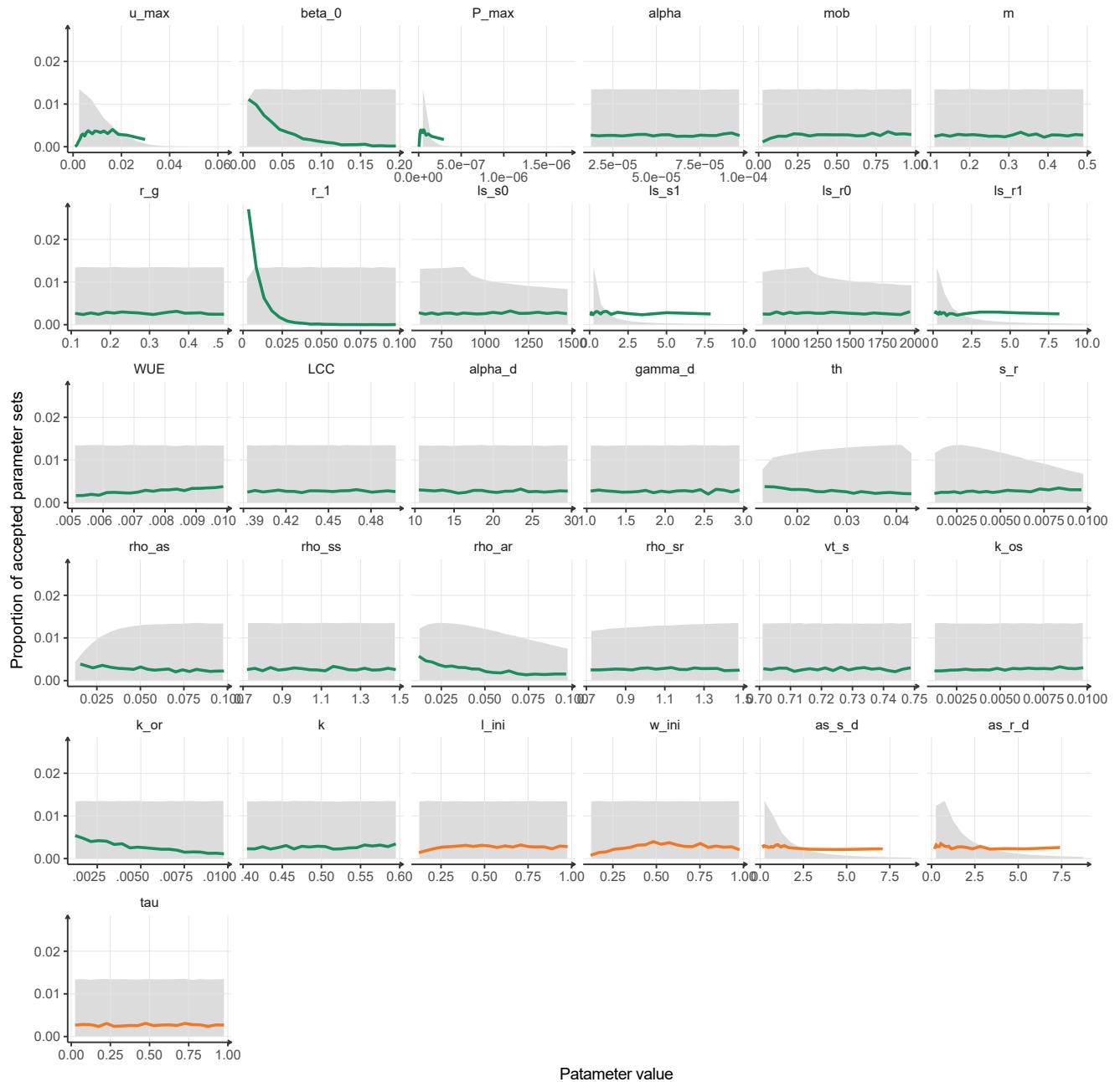


Figure 1.2: Selection rate (coloured lines) per parameter (global and species specific) for the individual growth. The grey area illustrates the prior distribution after the first filtering step (see method form ore details. Non plastic).

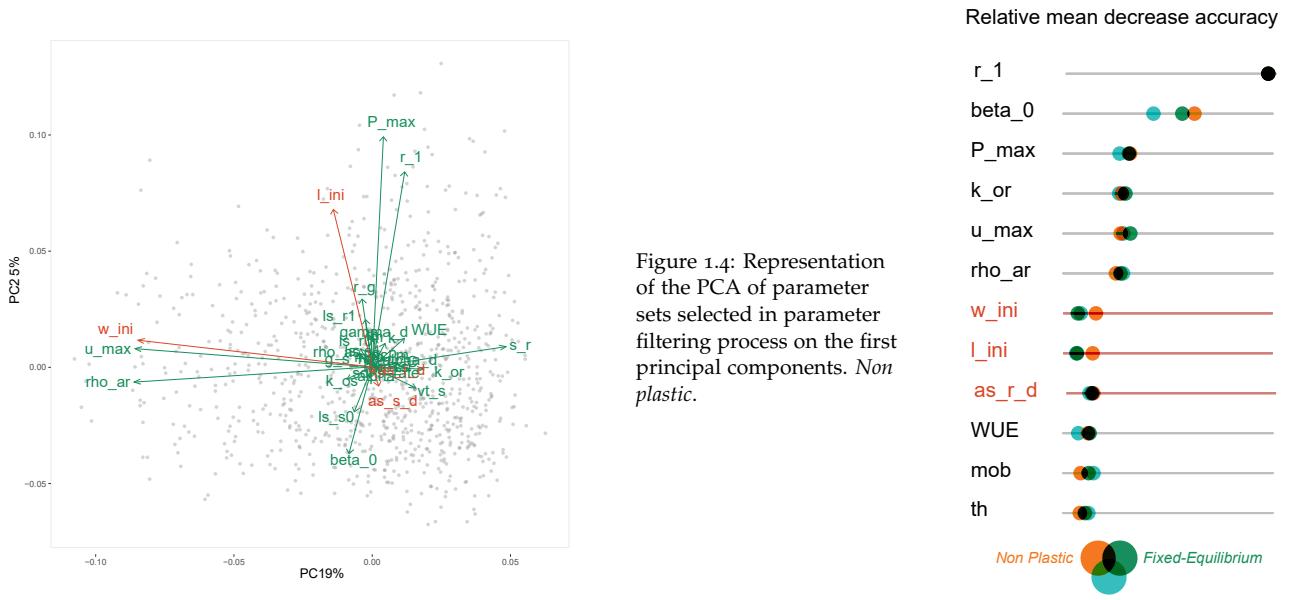


Figure 1.4: Representation of the PCA of parameter sets selected in parameter filtering process on the first principal components. *Non plastic*.

The PCA performed for *non plastic* algorithm only on parameter values reveals that important parameters are also the dominant variables that shapes the selected subspace. The two first axis explain only 14% of variance. The first one is related to the root activity and efficiency (u_{max} , l_{ini} , ρ_{ar} and s_r), the second is in line with global efficiency and resource availability.

The parameter filtering process is based on individual species, ... Species cannot be distinguished on these two main component space, neither on species specific parameters space (l_{ini} , w_{ini} , w_{ini} & l_{ini} , as_{s_d} , as_{r_d} , as_{r_d} & as_{s_d}) despite small variations in distribution shapes and ranges between species (data not shown).

For each algorithm the response of the two filtering variables (weight and RSR) are plotted against the most important variables in figures 1.5 and 1.6.

VARIABLE RESPONSES

The total biomass is particularly sensitive to the tissue respiration cost (r_1), but also to the maximum exchange rate parameters. There is a notable difference in growth maxima between the two conditions in favour of the wet condition, in line with observed data. This difference is observed for the three algorithm that differ mainly by the amplitude of the biomass ranges (need data). Growth response curves are similar for all allocation algorithm. Growth is only weakly related to species specific parameters. Total biomass under *Plastic-optimisation* algorithm seems to be more sensitive to variables influencing the exchange area per unit of biomass.

The species specific parameter τ controlling the balance between genetic and environmental control does not emerge as a influencing parameter at the global scale for any of the two flexible allocation rules.

Root:Shoot Ratio (or RMF in figure 1.6) strongly responds to species specific parameters under *non plastic* allocation because the memory parameters (l_{ini} and w_{ini}) are the means plants control their RSR. For other allocation rules, species specific parameters have little control over RSR. Surprisingly, the photosynthetic capacity has stronger influence on the ratio than the root maximum exchange rate.

Figure 1.3: Relative importance of main parameters for selection under the three main allocation algorithms: (*non plastic*, *fixed-equilibrium* & *plastic*).

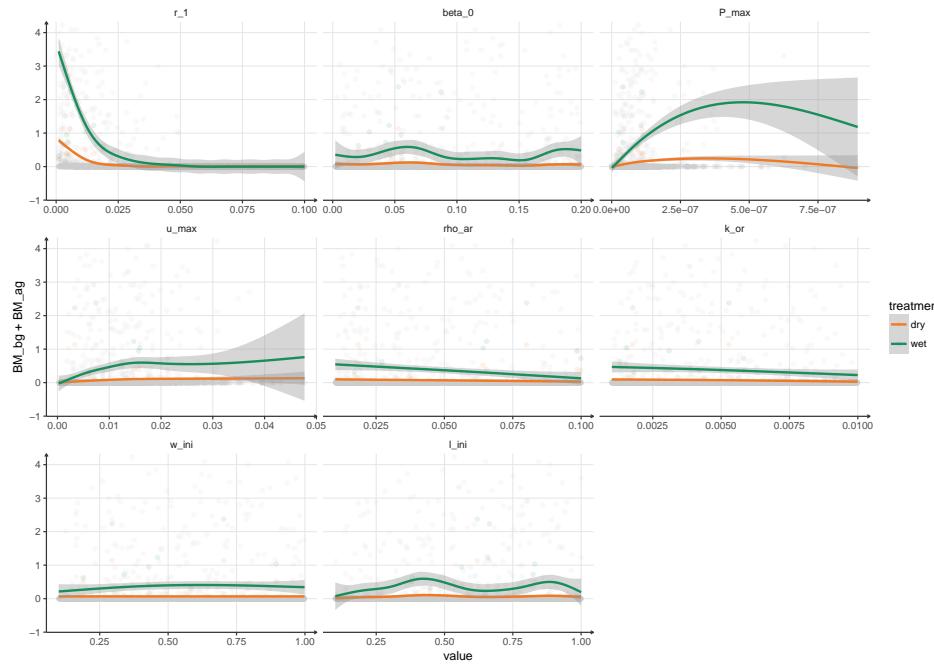


Figure 1.5: Main parameters effect on the total plant biomass. Non plastic. One dot represents a parameter set. Not all parameter set are represented as the y axis is limited around the smooth function (loess). Coloured points represent selected parameter sets in the two treatments (dry and wet).

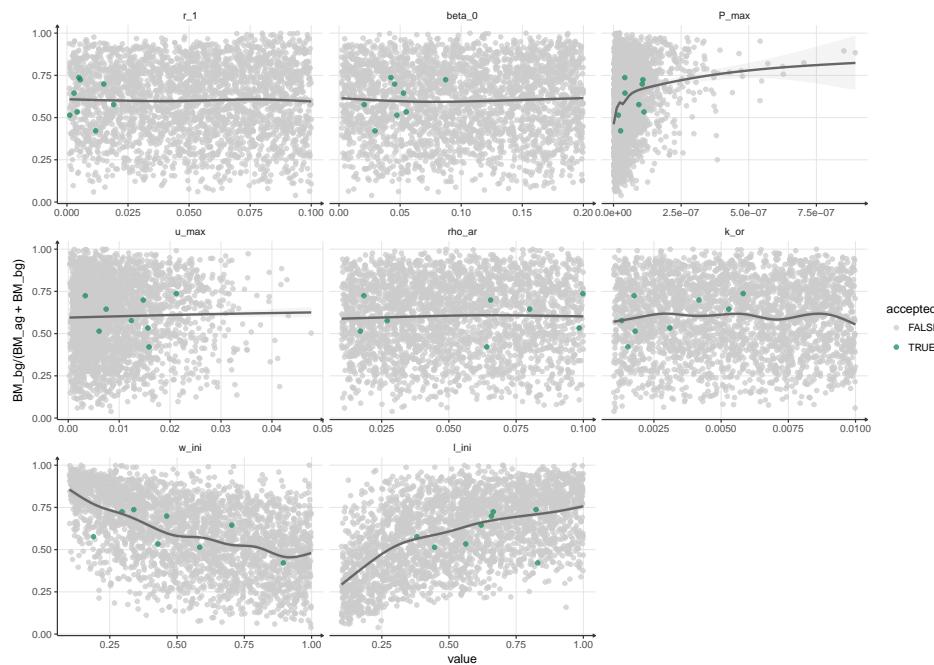


Figure 1.6: Main parameters effect on the total plant Root Mass Fraction (RMF). Non plastic

Little to no difference in RSR is expected for *non plastic* allocation rule since allocation promoted a fixed phenotype, but both *fixed-equilibrium* and *plastic-optimisation* allocation rules allow for changes in RSR. Nevertheless, no stable change in RSR is observed in any of the simulations. Fluctuations are present but consist in stable oscillations between two fixed values (see figure 1.7), synchronized with water variations. These rapid adaptations of the relative proportion of roots denote a high flexibility of plant phenotypes in *MountGrass*.

ROOT SHOOT RATIO AND PLASTICITY

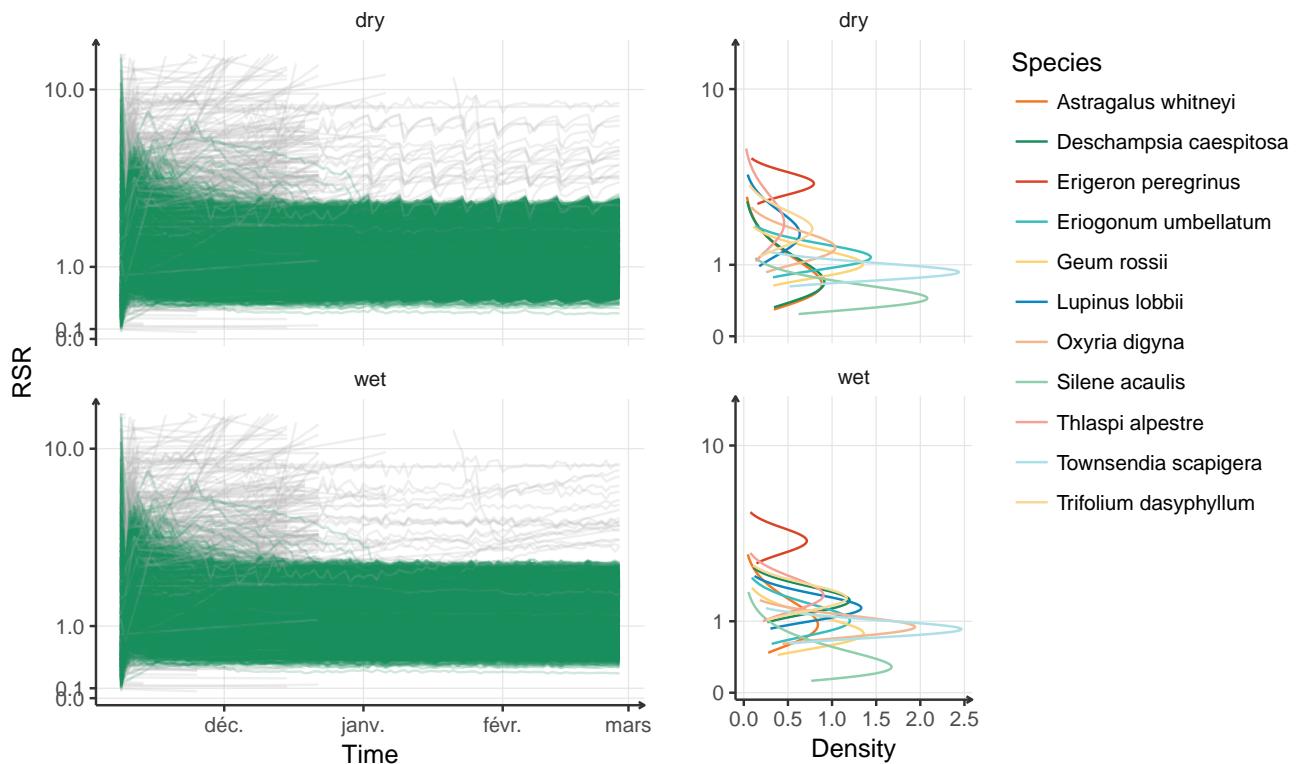


Figure 1.7: Comparison of simulated values of RSR with real species RSR in two contrasting conditions. Because there is no plasticity or ontogeny, the simulated plant do not express any changes in RSR. *Fixed-equilibrium*.

1.1.3 Discussion

The relative low selection rates for all allocation rules highlight the complexity of fitting such complex model to empirical data, despite the relative simplicity of the data. This difficulty seems to lie in two factors: the high number of parameters and the lack of stable changes in RSR. This last point is further discussed in the following paragraphs. Nevertheless, plant growth is reproduced in two contrasting conditions for multiple species, and while plastic algorithms have a greater potential for growth (more high growth rate), this is not systematic and the absence of clear pattern for the most influencing parameters, such as maximum exchange rates and respiration rates, indicates that such high growth depends on a combination of parameter values. I believe that the shape of gain and cost functions along the functional trade-off between active and structural

GROWTH AND STRATEGY SPACE

tissues plays a determining role in the growth. A trade-off function with a wider viable range is more likely to be selected as more strategies would grow (therefore reducing the relative sensitivity to species-specific parameters). Considering the exponential shape of the turn-over function (one of the main cost with respiration), the width and height of the trade-off (or net gain function) is probably more strongly linked to the gain functions (exchange rates) and linear cost function (respiration), explaining little effect of parameters related to lifespan (already preselected otherwise). There is a strong dependency between viable strategies (and as a consequence functional potential diversity) and the main trade-off between resource acquisition and efficiency.

Filtering the parameter sets based on all species instead of individually would have been ideal to quantify this link and better calibrate the model. However, such approach would have required many more simulations, when the parameter filtering method was chosen for its low computational cost. Moreover, considering the number of species-specific parameters, fitting the strategy subspace (at least default active tissue allocation parameters, the memory of resources and stability) of 11 species to the data in combination with more than 20 models parameters is near impossible. Ones should have had first determined the relative positions of the species within the said strategy space before any global calibration routine. Nonetheless, species-specific parameters have an influence on model main variables. Memory parameter affected the RSR in the context of *non plastic* allocation rule (see figures 1.7 and ??, while the default proportion of active tissues in roots was an influencing parameter in all algorithms (figure 1.1.2, `as_r_d`). Therefore, they should be analysed in further simulations within the same set of model parameters.

Because of the model complexity and the number of species specific parameters, in addition to long simulation time, Bayesian calibration could not be performed. In the Bayesian paradigm the information is contained in the data and revealed by the structure of the model. An alternative modelling approach is to used the parametrisation phase to accept certain parameter sets, and learn about the system through simulation experiments. The simulated data is analysed rather than empirical data. The patterns emerging from the simulation experiments inform us on the impact of the modelled mechanisms (even if they do not totally match the data). Therefore the model is still an understanding tool and can inform the effect of plasticity on ecological processes.

The growth is reproduced in contrasted conditions, but only partially as one per parameter set is tested. The number of species and dimensions in the strategy space would not allow for a calibration of all species for one parameter set. The plastic response of the root:shoot ratio is not correctly reproduced and would require a different implementation (stress based). However the plasticity as implemented improved the acceptance rate because of a better growth. Therefore the effects of plasticity can still be investigated with simulation experiments.

If the parameter filtering step does not result in the selection of optimum values for all parameters, it provides information on the main mechanisms influence plant growth. Indeed, the relatively high importance of parameters related to water shows the importance of the resource on the model behaviour. Both water availability (water absorption limitation, exchange rate) and root mass and construction parameters are important to match the empirical data. Considering

that the calibration relies on experiment data of drought events, it is no surprise that parameters related to water economy show strong influence on the selection rate and model behaviour. In the context where the model has been developed, water shortage is expected to be an important factor for the community dynamics. In this perspective, the ability of *MountGrass* to reproduce the differences in productivity between both conditions, and the relative sensitivity to water related parameters is an advantage. The link between water resource, species strategy, plant performance and phenotypic plasticity is explored more in details in the following section.

The sensitivity of the different variables to the parameters align with the two criterion of selection (that work with the independence of trade-off). In contrast with forest, the light is not the most important factor and water plays a more limiting role. A particular focus on below-ground resources should drive the simulation experiments with this model.

As mentioned earlier in this discussion, the model is not able to produce any shift in RSR in different water treatment. It is not a surprise for *non plastic* algorithm, but the filter was still applied on this criterion to allow the comparison with plastic algorithm and to be able to measure the improvement in selection rate. However, even plastic algorithms do not show strong enough response to water treatment in term of RSR. A strong and good (in the sense it would have matched the data) is larger in amplitude and more stable in time. Such processes generally amplify with time, *i.e.* when the number of drought event increases, the response (allocation to roots) increases (relative to default phenotype). Unlike natural systems, plants in *MountGrass* fluctuates between two "states", or phenotypes associated to the dry and the wet conditions. The value of the RSR following a drought event is reached after the first week without water. This can be explained by two main mechanisms that are related but have contrasting implications. The quickness in response to the changing conditions is allowed by relatively high assimilation rates. While the net growth rate is limited by the comparison during the filtering process of the total weight of plants with the empirical data, the assimilation rate is not and can be compensated with relatively high turn-over rate. Net growth rate being equal, species with higher assimilation rate will have higher **phenotypic flexibility** (higher fraction of biomass to invest in carbon pool of choice) than species with lower assimilation rate (but lower turn-over. This flexibility, similar to reallocation, allows changes in RSR, but not the accumulation of biomass in roots. Unfortunately, both the constant turn-over rate implemented in the model, and the selection toward "wide and high" gain functions limit control on this aspect.

This generalised high phenotypic flexibility allowed by high assimilation rates to compensate high turn-over rate highlight a problem within the calibration. The reproduction of growth patterns gives us confidence in the good functioning parameter filtering process, so wrong priors are certainly the cause of this behaviour. The uncertainty around the exchange rates for shoot and roots lead to the definition of relatively wide priors informed by parametrised models (Kleidon and Mooney 2000; Reineking et al. 2006; Taubert 2014). In the other hand, the turn-over parameters are relatively well informed by modelling approaches but also empirical studies (Ryser and Urbas 2000; Wright et al. 2004; Tjoelker et al. 2005; Luke McCormack et al. 2012), leading to more constrained priors. The value of these priors is not discussed, it is rather how they are translated

PHENOTYPE FLEXIBILITY

within the context of the model leading to an over-estimation of the cost of leaf senescence. Because the lifespan is integrated at the daily time-step as a constant turn-over rate, instead of a late decrease in biomass as in natural systems, the biomass is reduced early in the growth (from day 0). This can be a problem when the growth is non linear, especially when growth is higher early in the growth period. In this context, fairly narrow priors can lead to an over-estimation of the turn-over cost as the non linear growth is not properly integrated by the integration of the tissue senescence. This over-estimation is then compensated, during the parameter filtering process by a higher assimilation rate and a higher tissue flexibility.

The particular design of the experiment from Peterson and Billings 1982 with cycling wet and drought periods can also explain this effect. Other experiment designs with shifts in the mean influx of water would limit the role of the phenotypic flexibility and show more consistent differences in RSR between wet and dry conditions.

Moreover, the fact that plants are more productive during periods where they may not want to invest in roots reduces the possibility for a strong durable shift of RSR. Indeed, a plant would drift to higher RSR if it was more productive when pursuing the high RSR phenotype than when pursuing the low RSR phenotype. This last point mentions the "will" of the plant, in the context of *MountGrass* this target-phenotype is encoded in the projection of external conditions. Because this projection is daily based by design, the accumulation of drought stress is not translated in the internal projection variables of the plant (like it can be with the accumulation of phyto-hormones.). This limitation highlights a big difference between simulated plants in *MountGrass* and natural plants. While solutions to overcome this problem can easily be imagined(see equation 4.28 in 4.3.3), they would require more parameters and introduce more complexity to the analysis. This model provides a first approach to phenotypic plasticity in grassland models and the formulation of the projection, key element of the phenotypic plasticity, is certainly a starting point for further development. Nevertheless, the differences in response to the parameters between the three allocation rules, despite shared plant functioning, demonstrate the importance of plasticity itself. And simplification of the processes should not be a reason to not explore its effects. The fact that the parameter tau has a relatively small impact on selection rates also support the need to better understand all strategic axis before focusing on the effect of projection. While there are many ways of simulating the phenotypic plasticity, the parsimony is privileged. This simple representation is enough to understand the effects of active plastic allocation in association with the other strategic differences between species.

The high flexibility of the plant phenotype given by the high assimilation and turn-over rates reduces the inertia of the model and its capacity of modelling lasting changes in RSR. The modelling of the plastic response also reduces the capacity of the model to well capture changes in RSR.

The parameter filtering process successfully capture the growth pattern, showing convincing patterns of parameter sensitivity and variable response. However, limitations in the plastic response modelling, coupled with high phenotypic flexibility and a particular experiment-design do not allow a solid representation of the RSR differences between the conditions. Nevertheless,

less, the *MountGrass* still offers a way to interrogate the effect of plasticity on growth patterns, optimum strategies and potential diversity.

1.2 Individual level behaviour and properties of plastic allocation algorithm driven by the plant memory

Calibration and sensitivity analysis give information on the main processes of plant growth, but the general effects of the allocation rules on plant growth are not fully identified. Moreover, because the parameter filtering processes was limited to individual plants and the effects of the species specific parameters are depending on the other parameters of the model, the effects of these species specific parameters should further be investigated. The objective of this part is to set better understanding on the role of the **allocation rules** and species **memory** on plant development as basis for interpretation of plasticity effects in following chapters.

The challenge of the framework presented in the paragraph 2.2.1 under *plastic-optimisation* is to control the phenotype with the values of the memory. The risk of this approach is to have too tight estimation function of the fitness (or driving function) and to see the convergence of all species (with different memory values) toward the same phenotype (same allocation of active and structural tissues in roots and shoot). The extend to which different species memory lead to different phenotypes under full genetic control (not influenced by the external conditions) is explored through simulation experiment under *plastic optimisation* allocation algorithm with no effect of conditions on traits ($\tau = 1$), only on growth.

1.2.1 Method

The effect of allocation rule on phenotypic development is investigated thanks to pot simulations (see Methods in 1.1) of 100 days in 3 watering treatment: 2mm, 8m and 16mm per day. To avoid drift in the phenotype due to allocation algorithm (see paragraph 2.2.1 on phenotypic determination), simulations where run a first time, then rerun with default specific traits matching traits at the end of the first simulation set. All four algorithms are simulated. To reduce the number of simulations 100 parameter sets are selected randomly within the accepted parameter sets for the *non plastic algorithm*.

ALLOCATION ALGORITHMS

Memory of external conditions plays a determining role in phenotypic development under *plastic-optimisation* allocation rules. The effect of the memory alone (environmental cues ignored by setting tau to 1) on the default emerging phenotype is explored for diverse memories (9 values on the two axis from 0.1 to 1 later scaled to the maximum area exchange rates for model parameter set considered, or 81 values) for each accepted parameter set. The effect of the memory values on the final position of plants in the phenotypic space are visualised by fitting loess curves between memory values and individual trait values.

MEMORY & PHENOTYPE

1.2.2 Results

The allocation algorithm affect the way the organic matter in distributed between the different tissues of the plant. With partitioning coefficient pre-established for

ALLOCATION ALGORITHMS

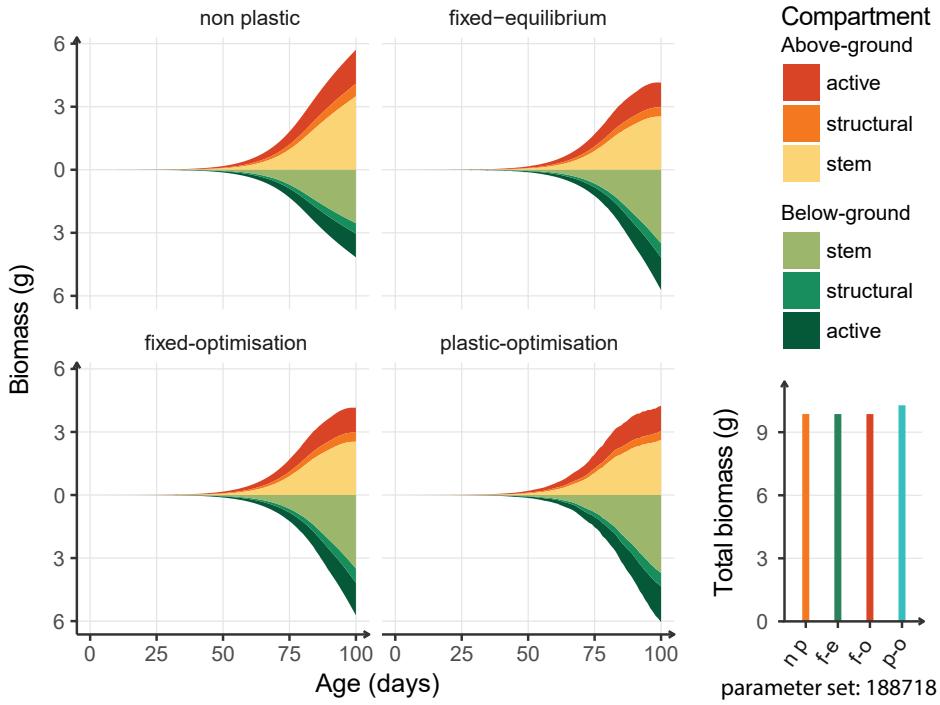


Figure 1.8: Effect of the different allocation algorithms on the different biomass compartments of the plant. The fraction of organic matter allocated to the stem (ensemble of supporting tissues for shoot and roots) are increasing over time for all algorithms. The *non plastic* algorithm show constant allocation coefficients between above-ground and below-ground compartments and between active and structural tissues. All others show different coefficients for the above-ground - below-ground partitioning, and the *plastic-optimisation* algorithm have changing proportion of active and structural tissues. The bottom-right panel show the total biomass for the four allocation algorithms after 100 days.

the given conditions, the algorithm show very similar performances (see figure 1.8). The difference in allocation algorithm is mostly noticeable in figure 1.8 mostly on the shift toward root allocation at the end of the simulation when the water becomes to be limiting. The plant under *plastic-optimisation* allocation benefit from a light improvement in performance (mean: +10%, median: +3.4% relative to *non plastic*).

The *plastic-optimisation* algorithm allows changes in the proportion of active tissues in organs. This may have repercussions on the allocation between shoot and root, but also can lead to non specific variability within plants with no perception of resource fluctuations ($\tau\alpha = 1$). The median variability of the RMF (root mass fraction) along the 100 simulated days is 0.015, that is five times higher than the variability of the other plastic algorithms (*fixed-optimisation* and *fixed-equilibrium*) (see table 1.4). This variability is much higher (around 0.028) for the plastic plants in all three plastic algorithm, while it is null for the *non plastic* allocation rule. The range of the RMF follows similar trend, with higher value for the *plastic-optimisation* than the other algorithms when plant do not perceive the resource fluctuations, and wide range for all plastic allocation algorithms when plants take into account the changes in light and water resources.

algorithm	sd		range	
	$\tau\alpha = 0$	$\tau\alpha = 1$	$\tau\alpha = 0$	$\tau\alpha = 1$
none	$< 10^{-12}$	$< 10^{-12}$	$< 10^{-12}$	$< 10^{-12}$
fixed-equilibrium	0.0278	0.00212	0.173	0.0155
fixed-optimisation	0.0279	0.00221	0.173	0.0161
plastic-optimisation	0.0283	0.0150	0.174	0.0839

Table 1.4: Median of variability and range of the RMF for simulations of 100 days, for 100 different parameter sets and three different water treatments (2, 4 and 8 mm per day), in the four different allocation algorithms. sd: standard deviation.

The plastic algorithms show similar levels of variation and range, while the *non plastic* one is stable as expected. The *plastic-optimisation* allocation show more

instability for non plastic plants ($\tau = 1$) but that is lower than the variability observed in plastic plants ($\tau = 0$). The allocation (and therefore phenotype) is controlled by the allocation rules (plastic dimensions and objective functions) and the estimation of conditions. Before investigating the effects of varying conditions, it is important to understand the effect of memory on plant strategy and phenotype.

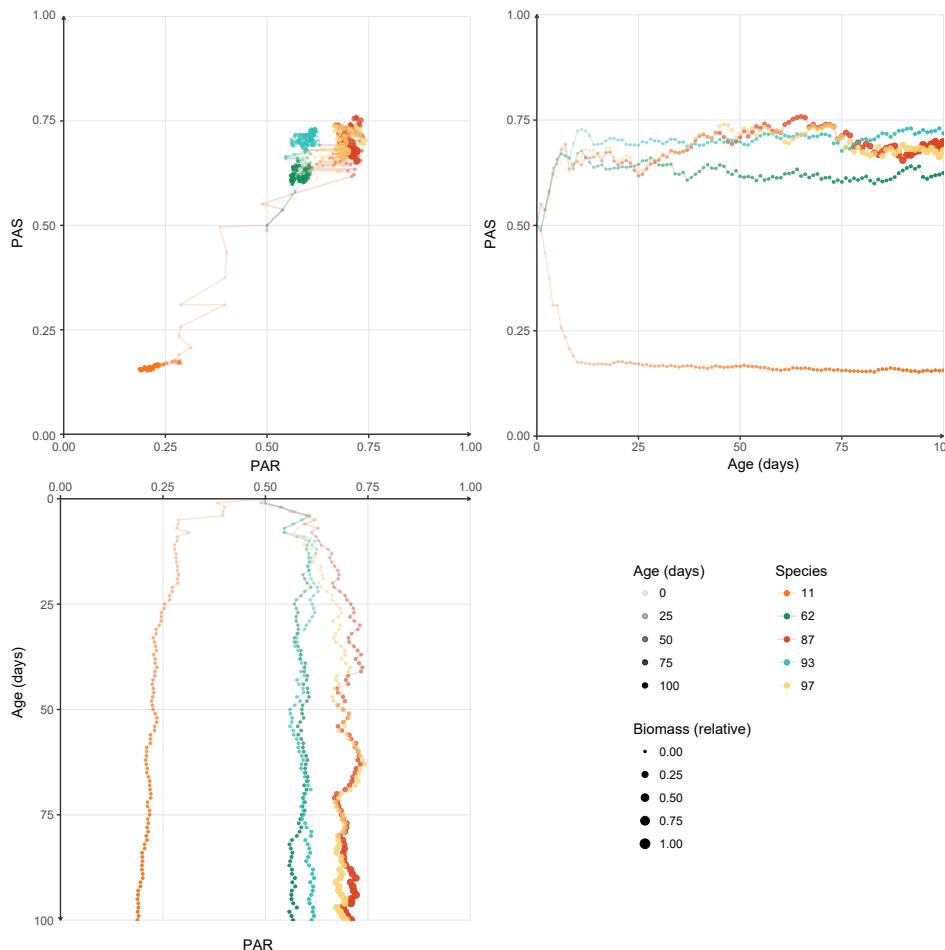


Figure 1.9: Trajectories along time in the strategy space of 5 plants with different memories. After 10 days, all plants have converged toward the estimated optimum.

The kinetic of the phenotypic shift is first visualised for one parameter set on the two main phenotypic axis (proportion of active tissues in roots: PAR and proportion of active tissues in shoot: PAS). From the same starting point the five species show distinct rapid shift toward segregated subspace of the 2D strategy space. The equilibrium point is reached in approximately 10 days for all 5 species. Despite constant memory, variations are visible on both tissue allocation traits of roots and shoot. These variations lead to partial overlap but the five species are distinct on the 2D space.

MEMORY AND PHENOTYPE

The memory of resource availability is a strong enough driver to alter the default phenotype of a species. The effect of the two components of the memory (memory of water availability and memory of light availability) on the three main traits is explored through local regressions. The proportion of active tissues in roots increases to a plateau with increase in water availability memory (figure 1.10). This response pattern is consistent between all parameter sets, but the

starting points and slopes may differ. The same pattern is observed between light availability memory and proportion of active tissues in roots (data not shown). The allocation convergence in the root is also influenced by the increase in light availability memory. An increase in the latter leads to a smooth increase in the former (see figure 1.11) with less drastic response than the water. This response is mirrored in shoot allocation response to increase in water availability memory (data not shown). Both organs react in symmetric ways to increases in resource availability. The RSR has a negative log response to water availability memory (positive in the case of light availability memory).

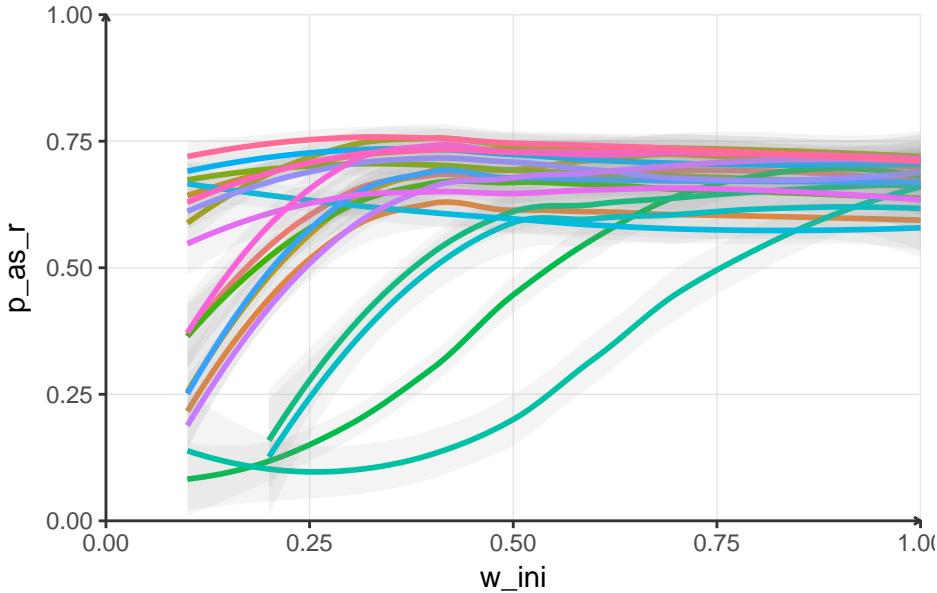


Figure 1.10: Effect of memory of water availability on proportion of active tissues in roots. *Plastic-optimisation*. Each line correspond to a local regression fitted for all memory combinations for a given parameter set. Water availability memory is given in percentage of maximum exchange rate, absolute values may change between parameter sets.

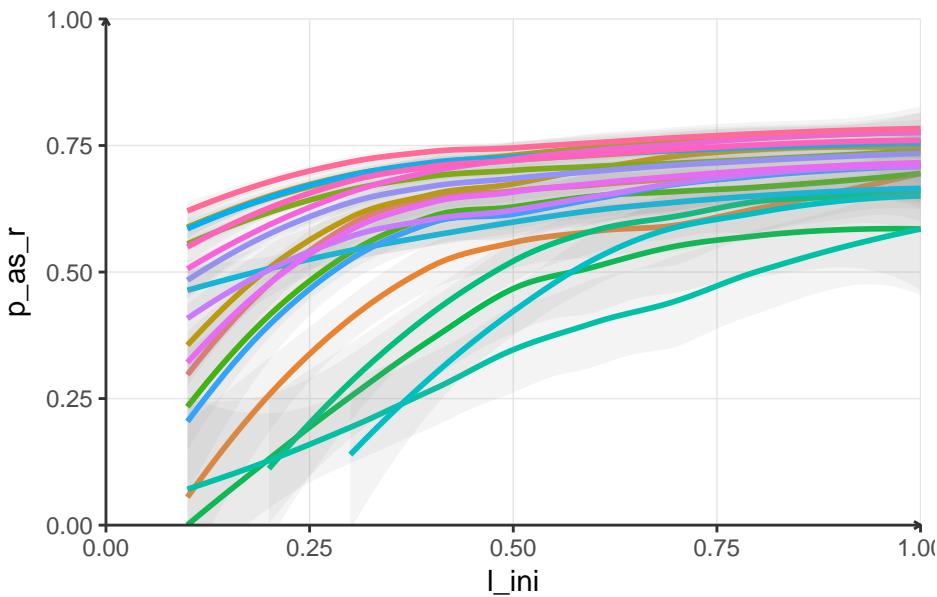


Figure 1.11: Effect of memory of water availability on proportion of active tissues in shoot. *Plastic-optimisation*. Each line correspond to a local regression fitted for all memory combinations for a given parameter set. Light availability memory is given in percentage of maximum exchange rate, absolute values may change between parameter sets.

The combine effect of the two axis of plant resource availability memory is observed by plotting the phenotypes (on the 2D space of active tissue allocation)

of four contrasting memories for all parameter sets (figure 1.12). There is clear clustering of the four memory profiles, with some overlaps due to the fact that multiple parameter sets are plotted at the same time. The memory of low availability (●) has a much larger distribution area than others, suggesting the relative instability of this profile within the "estimated net gain landscape". Memory of low availability for both resource drives plant toward very conservative strategies (need some values here) than other strategy. High expected availability of at least one resource increases allocation to active tissues to both organs. This confirms the positive effect of complementary resource (light for roots and water for shoot) of active tissue allocation in organs (see figure 1.11). Because of this, there is no highly unbalance phenotypes with high contrast between organ specific allocation emerging from the *plastic-optimisation* allocation in *MountGrass*. There is general coordination, but the balance between resource availability memories still impacts the position on the 2D, illustrated by the absence of overlap between low light - high water (●) and high light - low water (○) phenotypes. In case of high resource availability and coordination, high investment in active tissues for both organ is achieved (●) and high light - high water), but the range of values is similar than for unbalanced memories (●) and high light - low water (○).

1.2.3 Discussion

The pre-calculation of phenotypes, avoiding any phenotypic drift, allows for all allocation rules to grow plants with close performances. Nevertheless, the plastic algorithms show changes in RMF at the end of the simulation when the light:water balance starts to shift. This consistent shift in RMF for all three plastic allocation rules (with low variation of the other plastic dimensions) suggests the sensitivity and importance of this phenotypic axis. On the other hand, the other plastic dimensions benefit the plant growth suggesting that they also play a role in the tissue efficiency. While both the RMF and the proportion of active tissues can change the exchange area, only the proportion of active tissues can change the tissue efficiencies. Because the RMF shows similar levels of variation and range in both *fixed* algorithms (RMF is the only plastic dimension) and *plastic-optimisation* algorithm (see table 1.4) the allocation of active tissue in the latter algorithm does not compensate for change in root:shoot allocation and is not used to increase the area of the limiting organ. This is confirmed by the fact that memory of low-light conditions (● in figure 1.12) lead to lower allocation to active tissues than high light conditions (○). In the case of fully plastic plant trying to optimise their growth, the vegetative phenotypic dimensions do not fulfil the same functions: the RMF is used to adjust the balance between the resource exchanges while the change in active tissue proportions are related to the tissue and whole plant efficiency. This contrast in functions looks opposed to what is often observed in empirical studies where shoot:root ratio and SLA (here controlled by the proportion of active tissues) respond in the same direction to increase the leaf area and compensate low incident light (Ryser and Eek 2000; Poorter, Niinemets, et al. 2009; Poorter, Niklas, et al. 2012). This discrepancy reveals a limitation within the plastic-allocation algorithm: the balance function is mostly supported by changes in root:shoot ratio while the proportion of active tissues (controlling SLA and SRL) controls the tissues efficiency. The low proportion of active tissues in low resource (● in figure 1.12) indicates a selection of

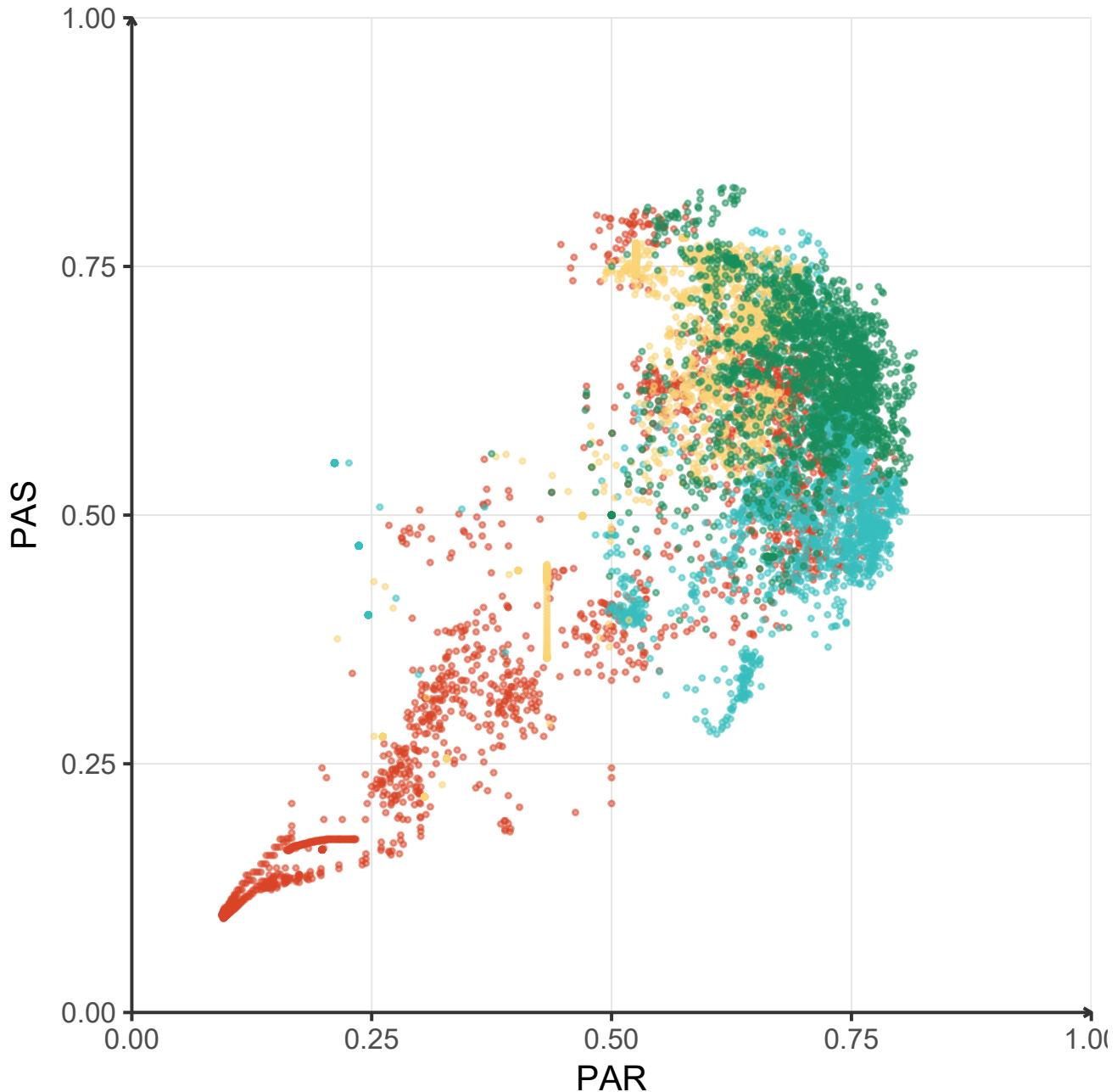


Figure 1.12: Impact of species memory on final phenotype in case of fully plastic allocation. *Plastic-optimisation*. Each point corresponds to a plant phenotype for a memory syndrome for a given parameter set. Colours denote the memory syndromes.

- low light - low water,
- HIGH light - low water,
- low light - HIGH water,
- HIGH light - HIGH water.

more conservative phenotypes when the resource is scarcer. This is in agreement with the Grime's triangle (Grime 1977) and large scale empirical studies (Wright et al. 2004). In contrast with the conclusions of Ryser and Eek 2000, here the full phenotypic plasticity of the *plastic-optimisation* algorithm is driven by similar constraints than the long term selection processes. This can be explained by the design of the trade-offs that drive the gain function (see chapter III). Therefore there are strong constraints on the tissue allocation, but low constrains on the root-shoot allocation. Additional constraint of this dimension can be added by considering other functions of each organs (such as nitrogen absorption by roots), or more artificially by increasing the cost of the displacement along the RMF axis. Also, it appears here that studying the long term effect of a fixed estimation of conditions is probably not the best way to understand how the plastic responses of plants to an abrupt change in conditions. However, in *MountGrass*, the plasticity is driven by the same mechanism, so such interpretations can be made. But, this discrepancy suggests that mean phenotype and plastic responses should probably not be driven by exactly the same mechanisms.

In addition to this imbalance in constraints, the mean organ approach can also explain this behaviour. Approximating the properties of the canopy by considering one mean organ leads to: a low impact of the plastic allocation on the SLA and SRL if the already existing compartments are large relative to the growth, a high importance of old tissues, while most of the exchange activity is generally produced by freshly grown tissues. Also, the rapid growth and turnover in numerous parameter sets also authorises rapid plastic response on the RMF dimension (see also the rapid oscillations in the figure 1.7 top left panel), diminishing the need for tissue specific adjustments. A stronger calibration of gross production and turn-over rates, as mentioned in the previous section, should reduce this effect. Finally, the optimisation function may be too strong and plants may not always go for the optimum allocation but for the fastest and most competitive choice (see Farrior 2011; Dybzinski et al. 2011; Farrior 2014). If this is not a problem in the context of this simulation where the memory is used to drive the default phenotype of the plant, it would be problematic in the context of plastic responses.

The different allocation algorithms impact the vegetative phenotype in different ways, but with similar performance when any phenotypic drift is avoided. But, the plasticity along the three main dimensions of the plant vegetative phenotype (PAR, PAS & RMF) seems to have different objectives. While the RMF is the main adjustment variable to respond to changes in equilibrium, the proportion of active tissues is more closely related to the amount of resources and tissue efficiency. However, it does not reproduce increases in organ area by changes in traits when the related resource is limiting. Multiple factors can explain this partial discrepancy with empirical results. The model still can be used to better understand the role of the memory as a driver for the phenotypic development, and the effects of the plasticity (particularly the RMF dimension) on plant performances.

The *plastic-optimisation* allocation algorithm allows for interesting insights on how the different resources affect the theoretical optimum phenotype. The increase in resource levels leads to an increase in the allocation of organic matter to the active tissues. While this is commonly demonstrated, the indirect effect of one resource on an organ that is not limiting for this resource is less often

STRATEGIES AND COORDINATION

studied. A higher perceived resource availability drives plants to have a higher proportion of active tissues in both gathering (*i.e.* leaves for an increase in light availability) and other organs (*i.e.* roots for an increase in light availability). The direct effect on the related organ shows a rapid shift from low to a maximum value. This rapid shift can be explained by the fact that the increased resource availability both increases the slope of the exchange rate per biomass (gain function) and reduce the importance of the maintenance costs relative to the productivity, favouring the exploitative strategies.

In the other hand, the indirect effect of an increased resource level on the non gathering organ can be explained by two mechanisms: a shift in the limiting organ requiring an increase in the exchange area of the newly limiting organ or an increasing gross productivity reducing the need for efficient organs. The former mechanism is related to the equilibrium maintenance. The balance between the two organs can be maintained by increasing the exchange area of the newly limiting organ (or reducing the exchange area of the non limiting organ, see Liu and Stützel 2004 for an example, or Grassein, Till-Bottraud, and Lavorel 2010). However this type of response is unlikely considering this implementation of phenotypic plasticity. The changes in exchange area are mostly driven by the organ biomass rather than its proportion of active tissues (see previous paragraph). The later mechanism is more in line with the observations of the behaviour of *MountGrass* (figures 1.12 & 1.11). It explains the increase in active tissues in both organs by an increase in the exchange rate of the gathering organ and in the productivity at the plant scale, decreasing the relative importance of maintenance costs and allowing for a more exploitative strategy of the organs.

Such allocation pattern could explain coordination between organs, as the cost of the respiration and turn-over are compensated globally by the gross productivity, and allows divergence from the optimum of the isolated organ functioning (see chapter III for details on the trade-offs at the organ's scale). However this coordination along a fast-slow axis asks the question of the stability of this strategy. Indeed, the high investment in active tissues observed suggest that the turn-over and respiration costs are high, and a loss in efficiency based on an incorrect estimation of condition could have strong negative effects.

The allocation trade-off allows for strategies from the fast-slow spectrum to arise for the shoot and roots based from the perceived condition availabilities with some degrees of coordination, in a coherent framework. Such allocation mechanism can explain coordination thanks to shared cost and increase efficiency when the resource is available. The potential instability of the phenotypes may lead to discrepancies between the optimum defined by the plastic-optimisation algorithm and the realised performance landscape.

The model *MountGrass* brings a new approach to agent-based models and plasticity by integrating the resource availability estimation directly as a parameter for the plant development strategy. Despite requiring certain adjustment for an integration with full plasticity (in RMF and organ specific traits), it reproduce a certain pattern of coordination and overall resource use strategy along resource gradients. It also makes a bridge between the mechanistic approaches, that use species specific parameters measure on individual plants, and species distribution models (SDMs) that focus on abiotic conditions¹ and how species distribution match climatic variables. This new framework can allow more exploration at bigger scales with numerous species, that is often the limitations of

THE MEMORY CONCEPT

¹ new SDMs now integrate biotic interactions as well as other ecological processes, as suggested by Guisan and Thuiller 2005.

such agent based models. However, to make this step, further work is needed on the general assumption that these estimation of conditions coupled with the gain function give good proxy to the plant development. There must be a strong positive correlation between the memory, the developed phenotype and the plant performance.

While this verification seems obvious, difficulties can arise if you consider plant with different levels of plasticity. A non plastic plant will certainly require the same memory as a plastic plant that will be able to adjust this memory. The former should conciliate the memory (and therefore the phenotype) matching the conditions of it growing period with values that limit risks of negative growth outside this favourable period. A mean value of the experienced condition during the growing period is certainly a good value for the memory. This also rise the question of the ontogeny in these models that often consider fixed allocation parameters. In *MountGrass*, ontogenetic shifts can be mimic under *plastic-optimisation* by having default allocation parameters different from the ones computed by the optimisation algorithm¹. In the other hand, plastic plants should better have a memory that matches the conditions at the early stages of growth, and let the plasticity drive the allocation for the continuation of the development. Also, while the structure of the model lets a door open for the integration of heritability mechanisms (through epigenetic modifications) that are expected to play an important role in the adaptation to the global change, those differences between plastic and non plastic plants may impact the integration of plasticity. This argument also encourage to find alternative solution to model plastic traits. Based on the review by Crisp et al. 2016, the concept of memory can be conserved, but adapted to be more driven by stress levels and stress response/recovery than actual resource availability values. The knowledge of molecular mechanisms of the plant functioning must better inform the modelling routine that is too focused on mathematical and theoretical approaches. The advantage of such specific memory mechanism is that it can be stress specific² and allows the integration of heritability.

The concept of memory, even if it allows the contrasting phenotype in a continuous space, should take a different form to suit multiple plasticity strategies and integrate a form of heritability. The molecular mechanisms of plastic responses are better understood and provide solid foundations for new organ specific plasticity.

The model *MountGrass* integrates trade-offs in resource use driven by the memory in resource availability. The investigation of the allocation patterns driven by the *plastic-optimisation* algorithm under the assumption of maximisation of the daily growth demonstrates different roles of the phenotypic axes: the RMF largely controls the equilibrium between shoot and root total activities, while the proportion of active tissues are related to the tissue efficiency as well as the overall plant efficiency and resource use strategy. While the fast-slow gradient along resource gradients is reproduced, and organ partial coordination explained, plastic responses to answer quick changes in resources are likely to not be reproduced due to a lack of constraints on the RMF dimension. The effect of the different algorithm, plasticity strategy and resources affect the plant performance still have to be investigated. Despite the evidences that the *plastic-optimisation* allocation mechanisms needs adjustments, the *fixed-equilibrium* algorithm offer a great tool to study the effect of

¹ limited here by a first simulation cycle, see methods for details.

² as suggested in the chapter III.

plasticity on plant performances and best strategies.

2 INDIVIDUAL PERFORMANCE, PLASTICITY AND VARIABLE CONDITIONS

The previous section highlighted the ability of the model to model growth, but also the importance of species specific parameters. While the plasticity mechanism did not replicate to a full extend (stable and higher amplitude) the phenotypic changes between the different conditions, there were some changes both in traits and in growth, leading to a higher selection rate. Considering the importance of species specific parameters and their potential impact on growth, these differences between plastic and non plastic allocation rules should be investigated in an extended manner. The specific roles of strategy and memory on the multiple components of plant growth need to be disentangled to draw better hypotheses on the role of phenotypic plastic on plant performance and coexistence. The role of resource availability on these mechanisms also needs to be interrogated. The effect of plasticity on coexistence can also be approached with respect to relative performances and contraction of the strategy space.

This chapter tends to answer these questions with simulations of individual plants with diverse strategies and under multiple allocation rules. To simplify the approach and focus on the interaction between species strategies and allocation algorithm, the plasticity will be model as discrete mechanism ($\tau = 0$ for all plastic allocation algorithms).

2.1 Individual performance: between strategy, memory and plasticity

This first subsection focuses on the link between the phenotype and the plant performance. The plasticity and allocation mechanisms can affect both the link between phenotype and performance and the distribution of the existing phenotypes.

2.1.1 Method

Because little differences are found between accepted parameter sets for the three main algorithms, parameter sets selected for the *non plastic* algorithm are used for all algorithm. To reduce the number of simulations but have a measure of the genericity of the observed patterns, 20 parameter sets are selected among the accepted parameter sets for the *non plastic* allocation algorithm. As mentioned in the previous section, the parameter sets have been selected for only one species-specific and therefore an additional step was used to filter out the parameter sets

PARAMETER SETS

that could lead to high biomass values. For each parameter set, simulations of diverse phenotypes run for 100 days of 15 hours with favourable temperature conditions (20 °C) along resource availability gradient. The parameter sets are selected based on the maximum biomass of all simulated plants. One parameter set is randomly selected for each of the 20 brackets between 0 and 2 grams of total biomass.

To better understand what make a plant perform in the model, a multitude of phenotype needed to be tested. Tested phenotypes are distributed regularly along the three axis of the strategy space (proportion of active tissues in root, proportion of active tissues in shoot, proportion of roots) between extreme values (respectively (0.1, 0.99), (0.1, 0.99) and (0.1, 0.9)) for a total of 3375 combinations (15^3). Because the RSR is defined by the memory, and in this set of simulation experiments the RSR is defined before, the species memory needs to be computed afterwards. There is an infinite number of couple of memory values that can match a given RSR. Also, the projection of conditions is sensitive to both memory and experienced conditions, therefore the choice of memory can affect the relative sensitivity of species to changes in external conditions and alter the model behaviour. Because the role of memory is not the focus here, and because there is much more focus on the role of the plasticity as a mechanism (as opposed as a strategy with various values of tau), the parameter tau is set to 0. This ensures that only the starting phenotype and the experienced conditions play a role in plant performance.

For each phenotype a pot simulation is ran for 100 days of 15 hours under 4 millimetres rainfall and 120 Watt per square metres and per hour with the 4 main allocation algorithms (*non plastic*, *fixed-equilibrium*, *fixed-optimisation* and *plastic-optimisation*). Two resource levels are tested for each simulations. The low resource availability conditions correspond to a reduction by a factor 4 of resource influx, but the day length was conserved.

To visualise the performance landscape (plant performance relative to biggest plant as function of its phenotype) the performance of best phenotypes are projected against the 3 plans that compose the phenotypic space. Such projections are preferred to 3D alternatives as they work better with static visualisation and when most of the space is occupied. Alternative axis are defined to facilitate the interpretation and description of the performance landscape: the organ strategy plane(PAR-PAS plane) can be transformed into strategy balance (difference between PAS and PAR) and "speed" (in sense of Reich (Reich 2014))(mean allocation to active tissues).

To study the potential effect of resource availability and or allocation mechanism on the link between strategy and performance, an aggregated measure is designed: the **gravity center** of the phenotypic space is defined as the average phenotype weighted by the relative performance of each phenotype. It can be defined with respect to the initial strategy, or to take into account the plasticity, to the final position in the phenotypic space. Shift of this gravity center within the projection space inform of translation of the performance landscape.

Biomass measures are relative to best performing non plastic plant (to remove the general parameter set effect on growth) and compare (within each condition) the effect of allocation algorithm.

STRATEGY SPACE SAMPLING

SIMULATION SET-UP

PROJECTIONS

NORMALISATION

2.1.2 Results

The effect of species specific parameter on growth are first studied with the analysis of the performance landscape drawn by the growth of plants uniformly distributed in the strategy space.

PERFORMANCE LANDSCAPE

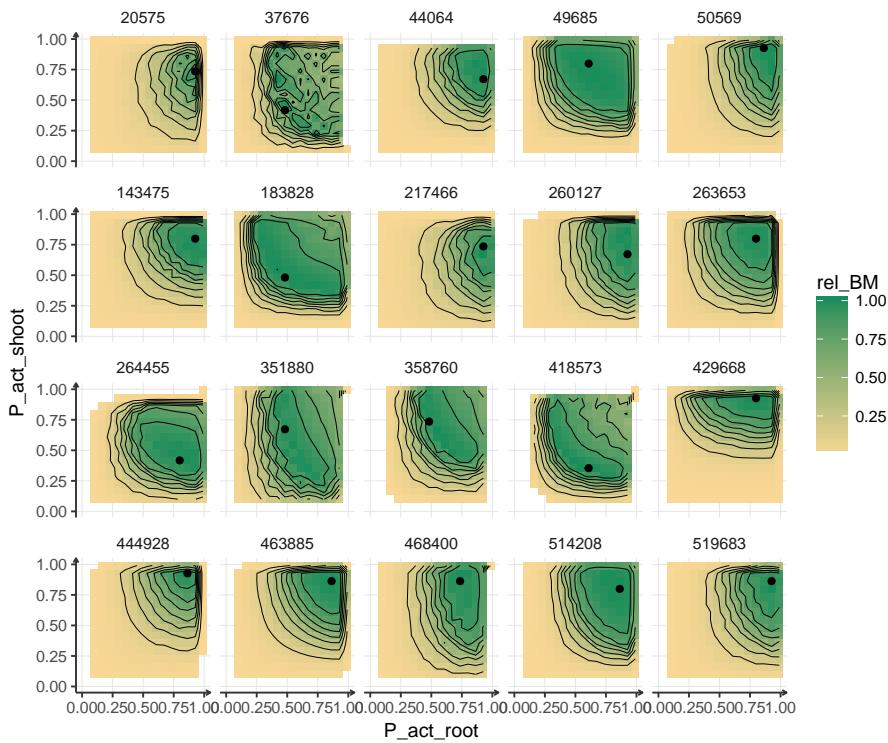


Figure 2.1: Projection of best phenotypes (varying RMF) on the 2D PAR-PAS plane for each parameter set. Points identify the optimums. *Non plastic.*

On the tissue allocation plan (proportion of active tissues in leaves and roots) (see figure 2.1), the best performing phenotypes present a bean shape. This shape covers a good fraction of the space, in the center and sometimes top-right corner (high active tissue allocation) of the 2D space, while other corners are ignored. Too low values for any of the organs lead to a limited growth. For certain parameter sets, the top-right corner, corresponding to high resource acquisition strategies, has lower growth values than the center. They have lower growth values than phenotypes with similar values for one of the organ and lower value for the other organ.

Projection of the best phenotypes over the three planes also gives information on the importance of the ignored variable on each plane. If the contrast between the growth projected phenotypes is high, at least on the main dimension is crucial for the growth. If the contrast is lower when the variable is ignored (*i.e.* the best value is used) then the projected variable is likely to be important. The projection on PAR-RMF and PAS-RMF (see figure ??) planes show higher contrast between phenotypes relatively to PAR-PAS plane, therefore the RMF is a more sensitive variable than the allocation factors to active tissues in organs.

Introducing resource availability variations and plasticity can impact the shape of the performance landscape.

OPTIMUM SHIFTING

A shift of gravity centres can be observed between the two resource levels in all allocation algorithms (see figure 2.2, the four panels). *Non plastic and fixed*

algorithm show similar trends with an increase of proportion in active tissues in both organs. This change toward more exploitative tissues is consistent and is observable for all parameter sets but one. The *plastic-optimisation* algorithm show drastically different responses of gravity center of phenotypes. There is little change in shoot proportion of active tissues, but a consistent reduction of active tissues in root system, and a reduction of root mass fraction (data not shown). These two responses indicates a net reduction of root activity in favour of shoot activity. Two things must be taken into consideration while looking at these results: (1) the gravity center is computed from final position into the phenotypic space, not the starting position, (2) because *plastic-optimisation* algorithm allows changes in traits that are represented (PAR and PAS), shifts along these axis can be driven by the plasticity mechanisms and not necessarily only performance differences. Similar representation of the gravity center computed from the initial phenotype (not shown) shows similar response for the three first algorithms, and no apparent shift for the *plastic-optimisation* plasticity.

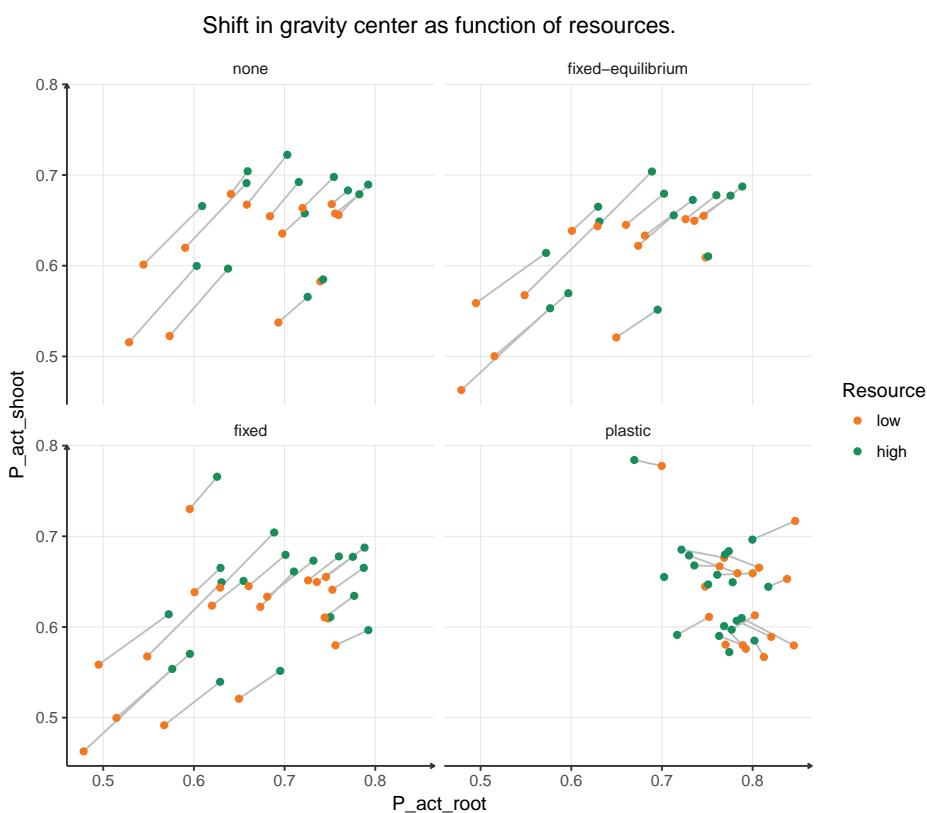


Figure 2.2: Shift on the 2D phenotypic space of the center of gravity as function of resource availability. The center of gravity is defined as the average phenotype weighted by the relative biomass, and characterises the performance landscape. *Non plastic*.

Non plastic and *fixed* plasticities respond the same way to a shift in resource availability. However, we can note that the gravity centres have lower proportion

of active tissues for *fixed* allocation algorithm compared to the *non plastic* one.

Plastic allocation lead to an improvement in mean biomass of all individuals for all three plastic allocation algorithms (see figure 2.3). The *fixed-equilibrium* plants are 2.5 times bigger in average than *non plastic* plants (in low resource conditions, and up to 7 times bigger for *plastic-optimisation* plant. These ratios are relatively similar for high resource availability.

PRODUCTIVITY CHANGES

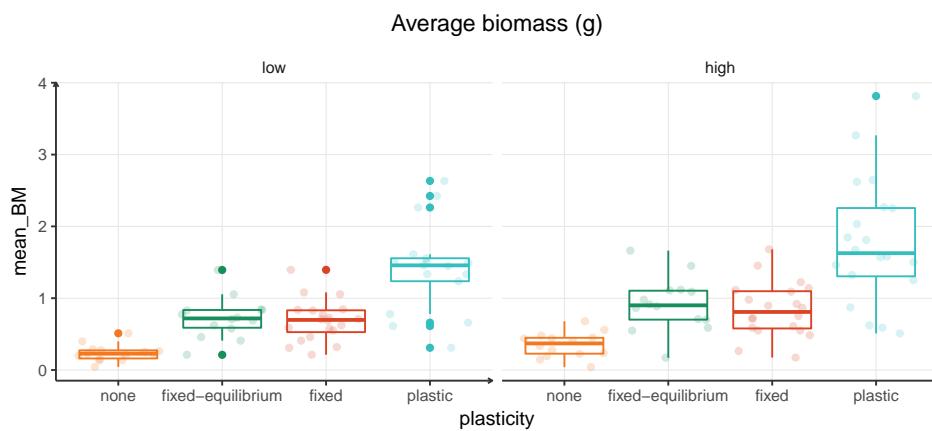


Figure 2.3: Mean relative biomass as a function of allocation algorithm and resource level.

However, the maximum biomass is only marginally improved with an increase of 6% for *fixed-equilibrium* and 8% for *fixed-optimisation* in low resource condition (see figure 2.4). These percentages drop to less than 1% in high resource availability conditions. The *plastic-optimisation* algorithm even lead to a decrease in the maximum biomass averaging 10% and 13% respectively in low and high resource availability conditions.

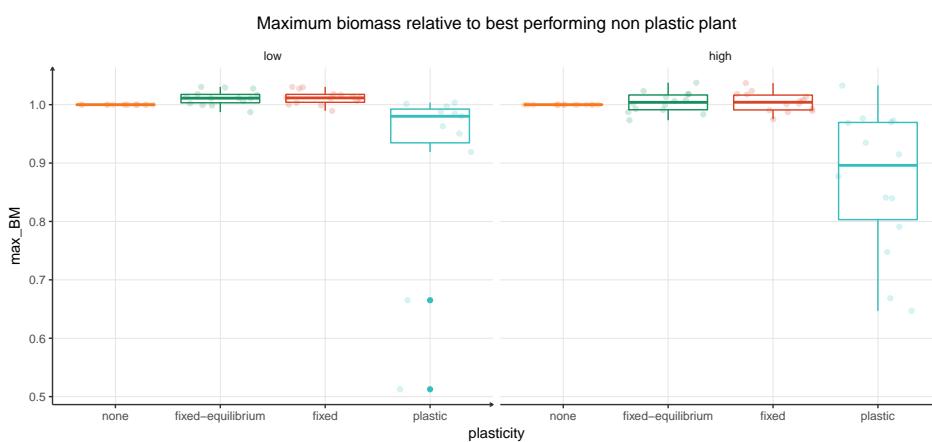


Figure 2.4: Maximum biomass relative to the *non plastic* simulations, as a function of allocation algorithm and resource level.

The effect of plasticity on the potential diversity is estimated by looking at the species that reach the range of 90% to 100% of the maximum biomass within the specific conditions (for each parameter set, algorithm and condition separately).

The number of species within this range is extremely low in *non plastic* allocation algorithm simulations, with 1.4% and 2.1% respectively for low and high resource conditions. This percentage is greatly improved by plastic allocation al-

PHENOTYPIC CONVERGENCE

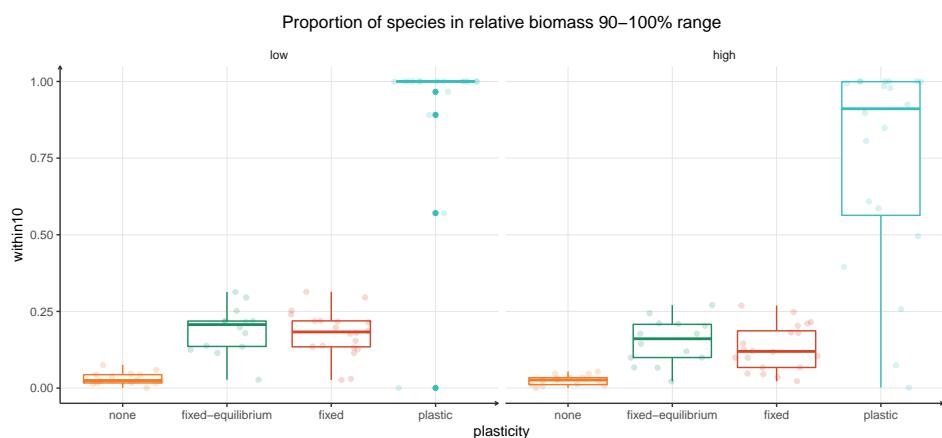


Figure 2.5: Number of species within the range of 90% to 100% of the maximum biomass, as a function of allocation algorithm and resource level.

gorithm and reach in average 9% to 15% of species in *fixed-equilibrium* and *fixed-optimisation* algorithm, while in can reach up to 100 % for *plastic-optimisation* algorithm, with a mean proportion of species with a top performance around 72% in low availability condition, and up to 82% in high resource availability condition.

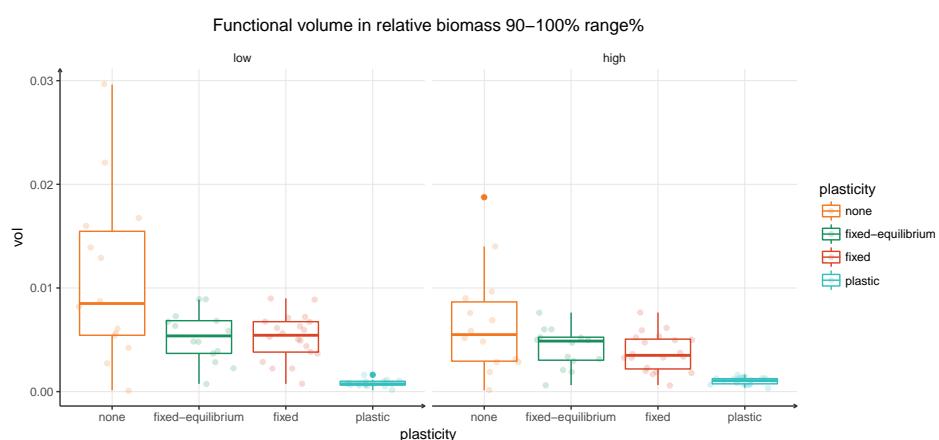


Figure 2.6: Functional volume occupied by the species within the range of 90% to 100% of the maximum biomass, as a function of allocation algorithm and resource level.

The functional diversity, estimated with the approximate volume of the top phenotypes, follows a opposite trend, with the highest value for the *non plastic* allocation algorithm. *Fixed* algorithms present half the functional volume of the *non plastic* algorithm, and the *plastic-optimisation* algorithm has extremely low values five times lower than the *non plastic* ones.

2.1.3 Discussion

The study of the performance landscape puts in light the different components of **plant performance**. To understand how plasticity can play a role, it is important to understand what make a phenotype a good phenotype. On one hand, the extend of strategies (plan PAR-PAS in figure 2.1) with high relative growth (green area) is high when the best RMF is considered, while this is greatly reduced on plans that integrates RMF variability (see figure 2.7). This result suggest the high

COMPONENTS OF PERFORMANCE

importance on this axis for the plants performance. This can be explained by a stronger effect of this dimension on the exchange area through changes in organ masses, instead of organ densities (affected by PAR and PAS). The RMF fraction impacts the plant performance in two ways: by changing the **equilibrium** between shoot and root exchange activities, and by changing the global carbon loss rate (respiration and tissue turn-over) if the organ differ on this aspect. These two components may have opposed directions, as the limiting organ may also be the least efficient, and therefore the RMF could be greatly constrained if the two aspects have similar importance. The effect on the equilibrium is likely be more important as a wide range of RMF values can be observed for numerous datasets (data not shown), and plant with uncoordinated (low-PAS & high-PAR, or high-PAS & low-PAR, *see figure 2.7*) organs still present high biomass values, suggesting that the respiration and turn-over loss are less important than a balanced resource acquisition.

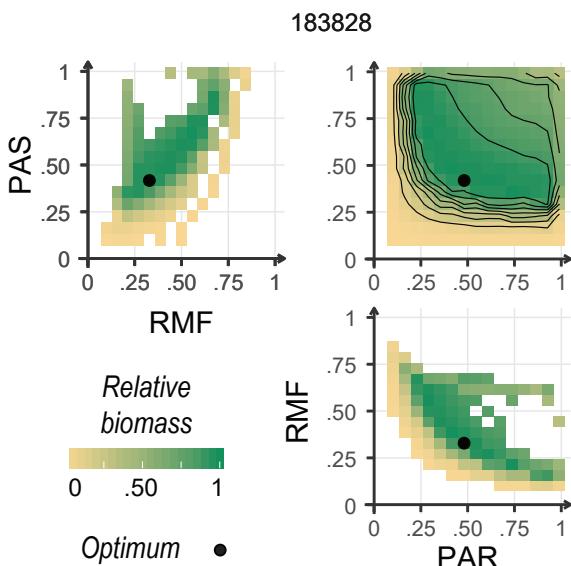


Figure 2.7: Projection for the parameter set 183828 of best phenotypes (according to the variable that is ignored) on 2D plans of the phenotypic space. The dots represent the optimum phenotype. White space indicates the absence of phenotypes able to survive until the end of the simulation (100 days). RMF: root mass fraction, PAR: proportion of active tissues in roots, PAS: proportion of active tissues in shoot.

On the other hand the organ specific strategies are also important as low values for any of the organs (leaves or roots) lead to very low growth. Extreme high values can also be limiting, suggesting the existence of an optimum of the proportion of active tissue for the tissue efficient. This **optimum tissue efficiency** results from trade-off between active and structural tissues, driven by the relative importance of carbon gain (increased exchange area with active tissues) and carbon loss (increased respiration and turn-over with proportion of active tissues) that depends on models parameters and resource availability (that change the exchange rate).

However, meeting these tissues specific optima might not be sufficient, as the bean shape of the best phenotypes suggests, another component is relevant. Low values of proportion of active tissue in one organ can be compensated by a high allocation of active tissues in the other organ that allows a higher allocation in the low exchange rate organ. This confirms the importance of the equilibrium over the tissue specific strategies. But the shape also reveals a last component of the plant performances. The fact that species with high values of proportion of active tissues in both organs have lower biomass, is certainly due to a limitation of both resources (equilibrium is assumed), reducing the overall efficiency.

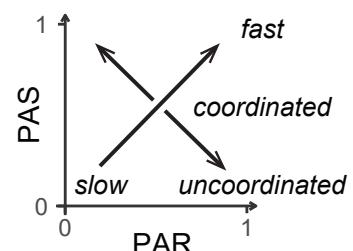


Figure 2.8: Alternative axes to describe the plant phenotypes on the plan PAR-PAR (PAR: proportion of active tissues in roots, PAS: proportion of active tissues in shoot). The slow-fast axis refers to the proportion of active tissues (close to the fast-slow strategies of Reich 2014), while the orthogonal axis show how coordinated the plant is (*see Freschet, Swart, and Cornelissen 2015 for similar concept*).

From this visualisation of plant biomass as a function of the phenotypes, three main components play a role. The **equilibrium**, mostly driven by the changes in RMF is essential to the plant growth. This is explained by a reduction of the exchange rate of the non limiting organ that greatly reduces its organ specific efficiency (see figure ??). This **organ tissue efficiency**, driven by its effective exchange rate, respiration and turn-over, is also an important component of plant performance. Low values of allocation of active tissues greatly reduces this efficiency, but it can be compensated by bigger organs. However, such mechanisms can affect the overall efficiency defined as the average mean of organ realised efficiencies (taking into account resource limitations) weighted by the organ masses. Finally, the **speed** of the plant, or the overall resource acquisition rate, admits an optimum that is between an over-capacity leading to a co-limitation of resource on both organ reducing their individual efficiencies, and the under-capacity, leading to a sub-optimum use of resources and letting space for competition.

The phenotypic plasticity allows species to move within this performance landscape along certain axis. It is often perceived with a species-centric perspective, that is to say, that plasticity is seen as variations in the species mean phenotype. However, in the context of community ecology, it is also interesting to try to see how it not only affect individual species but shape the community distribution in the strategy space. The plasticity relies on changes of default phenotypes toward "better" strategies in the context of the given conditions, therefore it implies that if it exists an optimum subspace (one strategy or an ensemble of strategies) species will converge toward this subspace, distorting the functional space. Environmental variations and plant interactions aside, in a constant environment the **performance landscape** is fixed. As a consequence, the plasticity benefits to the plant in a static manner, that is to say, it is only a tool to reach a better phenotype where the plant stays in if conditions do not change. This can be related to spatial heterogeneity that would lead individuals from the same species to adopt different phenotype to acclimate to the particular conditions of their spatial situation. It is opposed to the perception of a more dynamic phenotypic plasticity as a tool for a given individual to cope with temporal variations in environmental conditions. These two aspects are further discussed in the following section, while the effects of the contraction of the phenotypic space are discussed now.

As just mentioned, the plasticity can be seen at the scale of the species assembly¹ as a contraction of the phenotypic space of the species assembly. This contraction has two main effects: the reduction of potential functional diversity and a reduction of growth rate differences. There is here an emerging trade-off between the **species diversity**, supported by lower fitness differences, and **functional diversity**, reduced by the contraction of the phenotypic space. However, if the plasticity reduces greatly the potential functional diversity (volume of the whole phenotypic space without considering filtering based on relative fitness), the realised diversity (expressed as the functional diversity of the species within the 90%-100% maximum biomass range) is less impacted because a large parts of the phenotypic space have low growth rate in the given conditions. Nevertheless, there is a reduction of the diversity of expressed phenotypes. Indeed, in this scenario of "extreme" plasticity ($\tau = 0$) the convergence is important on plastic dimensions while partial convergence would be enough to have good fitness

CONVERGENCE TO SUBSPACE

¹ here I draw a distinction between species assembly that refers to all present species, and community that refers to the interacting individuals of the present species. However, some interpretations can be translated to communities.

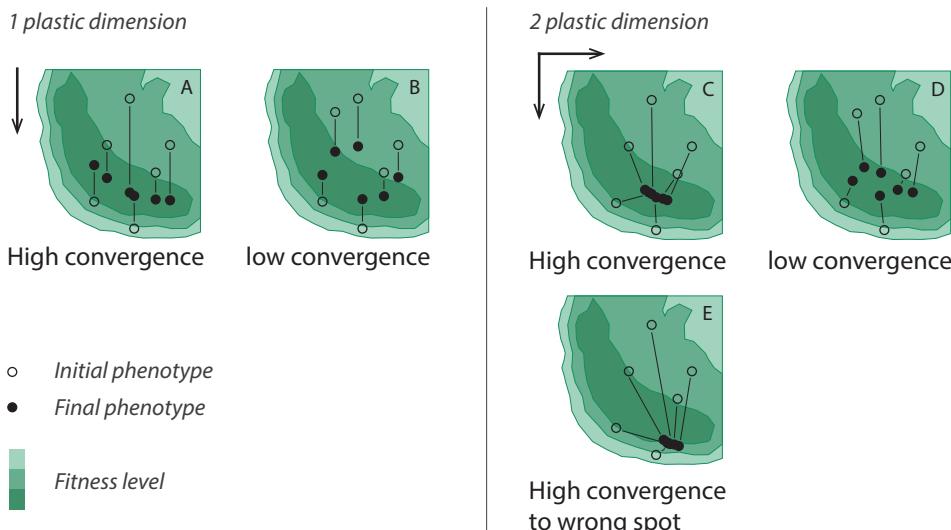


Figure 2.9: Convergence patterns on a 2D phenotypic fitness landscape, with 1 plastic dimension (A & B) or 2 plastic dimensions (C, D & E). Plasticity can lead to high convergence (A, C and E) with potentially high fitness evenness, especially in space with numerous plastic dimensions (A & E), this is problematic especially if the point of convergence is not the optimum (E). Limits to high convergence are necessary to allow realistic functional diversity with plasticity (B & D).

(see conceptual figure 2.9). Lower convergence on plastic dimension should lead to less compact phenotypic subspace while keeping relative fitness evenness. In the case of *fixed-equilibrium* and *fixed-optimisation* allocation mechanisms, this reduction of diversity is lower because only one axis is plastic.

A reduction of the phenotypic convergence can be achieved by other allocation mechanisms, differences in projection (different τ values leading to different projections) and plasticity costs. In heterogeneous system, this convergence is expected to be lower as heterogeneity will lead to different projections. The constraints imposed by fixed traits also reduce the risk of convergence (lower convergence in panel A than in panel C in figure 2.9), and other dimensions than the 3 studied here can be involved in the definition of the optimum (chemical traits for example) lead to larger optimum sub-space.

The question of diversity is essential in ecology, it often refers to species richness, or different indexes to measure this richness. In the context of the ecosystem functioning and service, the functional diversity is often preferred to qualify the community. To measure the functional diversity, the selection of the measured traits has an importance then algorithms can be used, considering the relative species abundances or distances between measures ([laliberte_distance-based_2010](#)).

Here, in the context of simulations with species diverging only of the 3 vegetative phenotypic axes, the functional diversity is expressed an estimation of the functional volume occupied by the species with top performance within this 3D space. But, because under certain allocation algorithms, some dimensions are plastic, it is difficult to study how plasticity impacts the functional diversity. The plasticity of an axis can lead to convergence and certainly reduce the potential functional diversity as only a subspace is considered. This is problematic in this context because there is a high convergence due to the specific implementation of the plasticity based on a shared gain function (equilibrium or growth-optimisation), but is certainly reduced if the algorithm allow contrasted responses (with response curves for example), or optimum phenotypes depending on the other traits as it is the case for *fixed* algorithms (as opposed as the *plastic-optimisation* algorithm). Because of this phenomenon of convergence, that cannot be totally avoided and is inherent in plasticity, the functional

ON DIVERSITY

diversity should be considered in relation with the species diversity. The functional volume is reduced by a factor between 1 and 2, while the species richness is increased by a factor from 5 to 10 (see figures 2.1.3, & 2.1.2). While the ratio between functional and species diversity decreases in plastic conditions, the overall effect of diversity could be positive, especially if there are other traits (non considered here) correlated to the initial phenotype.

The convergence of the phenotypes to a sub-space of lower performance lead to an increase in the mean biomass (see figure 2.4). However, the maximum biomass is only marginally improved in *fixed* plastic allocation simulations, and reduced in *plastic optimisation* allocation simulations. This two contrasted results, show different effects of plasticity. The light increase can be due to either a dynamic gain or a static gain. The **dynamic gain** can emerge because the plant growth affects the resource availability, changing the optimum phenotype, and allowing plastic plants to follow these changes during time. It could also result from a **static gain** because the phenotypic plasticity allows a better resolution in tested phenotypes (the plastic axis are continuous while the phenotypic space sampling was discrete). The role of plasticity and dynamic gain is explored in the following sections with temporal resource heterogeneity.

LIMITED GAIN

The reduction of the maximum biomass highlights the difficulty to find the optimum phenotype. Because, the growth mechanisms are reproduced in an exact manner in the plasticity algorithm, this mismatch is certainly due to a difficulty to project the future of resource availability. Because of that, it is possible that the gain in maximum biomass, mentioned above, due to static or dynamic gain is greater than it appears. The particular case of mis-projection in *plastic-optimisation* simulations is discussed in the following paragraph.

The phenotypic plasticity can lead to a certain degree of convergence, especially if the target phenotypes defined by the implementation of the plasticity are more condition-specific than species specific. While it can have strong effects on the functional diversity for the plastic traits, it also leads to high species richness due to a convergence toward a more performing subspace, and can potentially increase the total functional diversity if other traits are considered and the phenotypes more constrained.

Plastic-optimisation algorithm is characterised by a high convergence of the species within the phenotypic space, high mean biomass but maximum biomass lower than best *non plastic* phenotype, and high potential species diversity. The convergence is expected and explained by the fact that all three traits are plastic and all species (for a given resource level) experience similar conditions leading to the computation of the same optimum. The absence of plasticity cost limiting the convergence leads to a phenotype concentration toward this optimum. This convergence explains both the high potential species diversity, as all species have very similar growth rate, and the relatively high mean biomass because only few species did not survive or had very little growth rate.

PLASTIC EXHAUSTION

The fact that this plasticity does not translate into higher maximum biomass is surprising, especially considering the fact that RMF plasticity improves maximum biomass (see figure 2.3). Lag in adaptation is often identified as a limit of plasticity (DeWitt, Sih, and Wilson 1998; Van Kleunen and Fischer 2005), nevertheless, in a constant resource influx experiment, and considering the high phenotypic flexibility of plants in *MountGrass*, this explanation is unlikely. Another

problem highlighted with plasticity is its adaptiveness. Evolutionary speaking, it is hard to imagine the emergence and maintenance of a plasticity mechanism (in a given context) if it is no adaptive. Yet, such process could be maladaptive in a new context. Because plasticity is not emerging, but imposed by the simulation set up, its adaptiveness can be interrogated. Here adaptiveness do not refer to a reduction of fitness due to plasticity, but to the capacity of the plastic mechanism to define an optimum (or at least better) phenotype. Plasticity as implemented in model has no explicit bias and all mechanisms involved in plant growth are simulated by the allocation algorithm. The sampling of phenotypes is random and could be source of uncertainty, but it is uniform and no consistent drift is likely to emerge from the noise introduced by such sampling. The last aspect of plasticity that can affect the adaptiveness of plasticity is the estimation of conditions. The estimation of conditions is based on resource levels experienced by the plant and by definition are exact, therefore the problem lies in the projection of these conditions and how they translate into resource uptake.

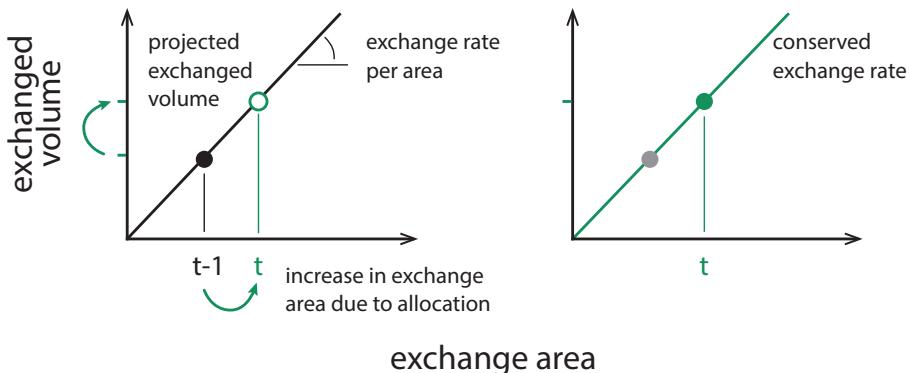


Figure 2.10: Projection of the water volume exchange after increase in exchange area at equilibrium and with no limitation.

In *MountGrass* the resource availability is coded as an uptake rate per day and per unit of exchange area, and is computed as the resource uptake divided by the exchange area. This resource availability is supposed constant, and plants make the assumption that increasing their exchange area leads to a proportional increase in resource volume exchanged (see figure 2.10).

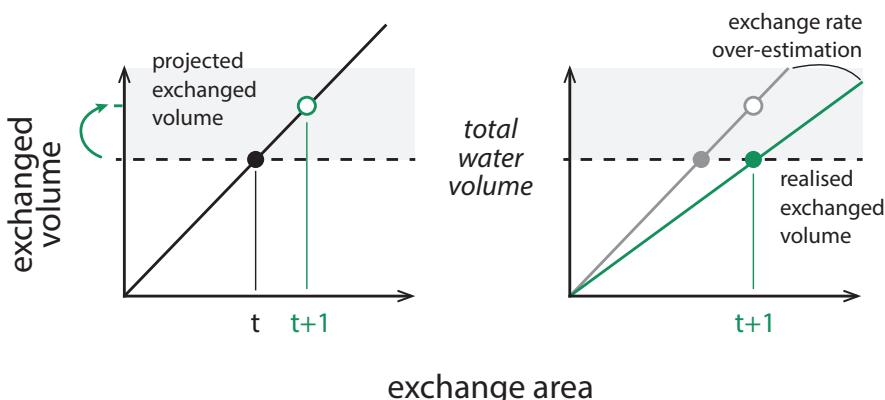


Figure 2.11: Projection of the water volume exchange after increase in exchange area when total available water volume is limiting. The water volume exchanged cannot exceed the total available water volume, leading to a systematic over-estimation of water availability and offset between shoot and root activity.

However, in the case where a plant already absorbs all the available resource, then this assumption is not respected, and the uptake rate per area is lower than expected (see figure 2.11 right panel, realised exchanged volume does not

match the projection because it cannot exceed the total volume of available water). This gap between perception and actual resource availability occurs because the plant is not able to perceive that the limitation cannot be compensated by a higher investment in the limiting organ. This behaviour explain a very high investment toward root and root active tissues in low resource conditions under *plastic-optimisation* allocation (figure 2.2). This gap¹ is the cause of the **plastic exhaustion** phenomenon. Indeed, this constant over-estimation leads to constant discrepancy between the estimated optimum phenotype and the actual phenotype, and a larger allocation to root active tissues. This effect is particularly noticeable in the context of pot simulations where the water pool is limited. The absence of plasticity costs also favours such extreme behaviour.

Despite this particular seemingly non-adaptive behaviour, the *plastic-optimisation* algorithm is still interesting to study in community simulations. First, the presence of plasticity cost should limit such extreme behaviours. Second, in a context of competition in a larger environment, this aggressive search behaviour is likely to be an advantage against individual with less aggressive, or stable strategy. Finally, this mechanism emerges in constant influx conditions that allow growth, but its emergence should be reduced in variable environment where water shortage leads to reduced growth.

The plastic exhaustion mechanism seems to be contradictory with the previous observation that the proportion of active tissues does not increase when the related resource is limiting (see figure 1.12), but it can be argued that it such extreme case, if the RMF has extreme values and is constrained by differences in tissue efficiencies. This is not verified however.

Plastic-optimisation simulations expose this phenomenon with large effects, but it is probably present for simulations with other plastic allocations but with smaller effects. The difference in magnitude can be explained by a less effective growth in early stages of development for *fixed* plasticities (when *plastic-optimisation* is more efficient than *fixed* plasticities) that delay the time when the total volume is reached (time t in figure 2.11), and in average lower active tissue allocation in roots that leads to lower loss due to non-equilibrium.

Plastic exhaustion is a specific limit of phenotypic plasticity as implemented in *MountGrass* that relies on the assumption of constant exchange rate per exchange area. It has a large effect in the specific case of pot simulations. However, this phenomenon can be mitigated by plasticity cost linked to changes in traits, and can have adaptive value in a context of competition. Therefore, I argue that *plastic-optimisation* algorithm has low information value in the context of pot simulations with constant resource influx, but should still be studied in the context of community dynamics.

As expected the resource availability and the resource balance are key components of the plant growth, to which the plant phenotype needs to match. Aside from the increase in biomass, an increase in "speed" of optimum phenotypes can result from higher resource availability. This observation is in agreement with empirical data that demonstrate higher SLA and faster physiology in favourable conditions. This aspect was less obvious in the response of species under *plastic-optimisation* allocation that shifted more in term of balance and RMF. This may be due to a change in the relative balance between both resources as their availability (from the plant perspective) are linked to the global resource levels by non linear relationships.

¹ this is different from a lag because it is not the result of slow changes in phenotype but comes from a default in the estimation of optimum phenotype.

The fact that plastic plants (for *fixed* allocation algorithms) show shifts of optimum strategies toward more exploitative phenotypes, in addition to the *non plastic* optimum shifts, in conditions of higher productivity demonstrate the importance of these strategies for the plant growth. However, the extend of this effect of conditions on optimum phenotype is susceptible to vary along a gradient. Indeed, because of the non linearity of relationships between resource levels and exchanges rates, and between exchange rates and growth rates, the link between the optimum phenotype and a resource gradient is likely to be non linear itself. In addition, phenotypic plasticity might also change the sensitivity of the phenotype to the resource level.

The study of the performance landscape highlight the importance of the RMF phenotypic dimension to regulate the balance between shoot and root. In the other hand, the proportion of active tissues in the two compartments impacts the performance in a more subtle way with possibility of balance between the two organs thanks to the RME. It also highlights the existence of a tight sub-space with higher performances, leading to high convergence under plastic allocation. This convergence can be problematic if the plasticity fails to be adaptive under specific circumstances. It can lead to a reduction of the functional diversity, but also increases the species diversity, altering the species-functional diversity relationship with potential consequences on the community functioning. The performance landscape is sensitive to the global resource availability, altering the relative fitness of species and their competitive relationships. This effect must be further studied to understand the link between the optimum phenotypes and the resource availability heterogeneity, and how plasticity can impact this link.

2.2 Plasticity and variability of conditions

The heterogeneity of conditions is an essential mechanism for plant coexistence. Plasticity is likely to alter the effect of this heterogeneity on plant coexistence and relative performance. The impact of plasticity on this relationship between spatial and temporal **heterogeneity** of resources (here limited to water) and strategy dominance is explored with the model *MountGrass* .

How does plasticity impact the performance of the different phenotype along a resource gradient? How can these potential changes affect the identity, diversity and productivity of mountain grassland communities?

2.2.1 Method

Because the coordination is shown to be less important than the equilibrium, the below-ground resource acquisition is expected to be important in mountain grassland under climate change scenarios, and an extensive simulation plan comes with high computational cost, only root strategies are sampled and studied in this part. Considering the structure of the model, the conclusions about the root compartment can certainly be extended to the shoot compartment.

For each of the 20 selected parameter sets, the growth of 400 plants (20 PAR values between 0.25 and 0.95, and 20 memory values between 0.1 and 1) is simulated for 100 days in square pots of 12 centimetres deep and 90 centimetres

SIMULATION SET-UP

wide (to avoid quick self-competition) in a temperature of 20 degrees celsius during the day of 15 hours, and 10 degrees during the night. The radiance is set to the high values of 122 Watt per hour and per square metre. Because *fixed* algorithms showed similar results, and the *plastic-optimisation* algorithm show strange results, only two allocation algorithms are simulated: *non plastic* and *fixed-equilibrium*.

Spatial heterogeneity of water level is mimicked by a gradient of water influx. The growth of all 400 species described above are simulated for *non plastic* and *fixed-equilibrium* algorithm independently in separated simulations where the water influx is regularly sampled between 0.05 and 7 mm per day (20 values).

SPATIAL HETEROGENEITY

Similar set-up is used for temporal heterogeneity simulations. Because the range of water influx used in the previous simulation is too wide, a lower value is chosen as the mean water influx. This value of 1.3mm per day corresponds to a point around which there is variations in the optimum strategies for most parameter sets. It is also relatively close to average rainfalls in the Alps during summer.

TEMPORAL HETEROGENEITY

2.2.2 Results: gradient of homogeneous precipitation conditions

To study the effect of plasticity on community identity along a precipitation gradient, we can look at the position of the optimum strategy (PAR) along such gradient with different allocation algorithms.

OPTIMUM STRATEGY

The effect of allocation algorithm is observed on all species by plotting the position of the median *optimum* along the watering gradient that translates what part of the strategy spectrum (from conservative to exploitative) benefit from the simulation conditions. At the low end of the gradient, conservative species exhibit higher growth than exploitative species with a median optimum around 25% of active tissues in roots for both the *non plastic* and the *fixed-equilibrium* plastic allocation. In the other end of the spectrum, for watering values above 1 mm per day, the *optimum* reaches a high point (median around 90% of active tissues for both algorithms) demonstrating better performance of the exploitative species in high resource availability conditions. There is not apparent differences between algorithms and the optimum is conserved along the gradient. There is a similar shift with an increase of optimum water availability memory for *non plastic* algorithm.

The memory of water availability of best performing phenotypes increases along the gradient under *non plastic* allocation (see figure 2.13, left panel), but the plasticity negates the effect of this species specific parameter and no clear pattern can be observed (right panel).

BIGGER PRODUCTIVITY?

The total cumulative biomass of all plants increases along the precipitation gradients. The plastic simulations have a cumulative biomass that is twice the biomass of *non plastic* simulations.

The effect on the maximum biomass is also investigated. For most simulation the maximum biomass is unchanged, and the median of the maximum biomass follow the same path for both conditions. However, the 75th and the 95th percentiles of plastic simulations shows a high increase in maximum biomass.

WHAT ABOUT DIVERSITY?

Similarly to the previous results, the potential diversity is estimated with the number of species, or the functional volume, of the species within the 90%-100%

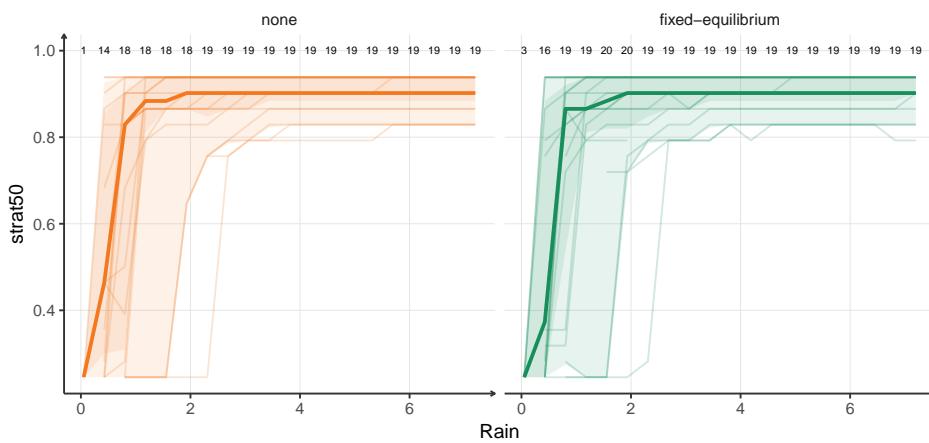


Figure 2.12: Median (dark line -) optimum root strategy along the water treatment gradient for - *non plastic* & - *fixed-equilibrium* allocation algorithms. The light lines (-) correspond to the 20 independent parameter sets. The color ribbon marks the band between the 5th and 95th percentiles.

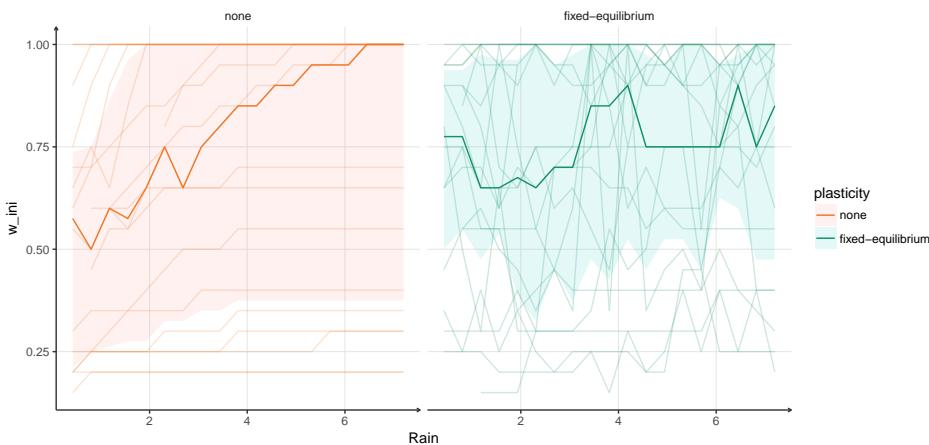


Figure 2.13: Median (dark line -) optimum water availability memory along the water treatment gradient for - *non plastic* & - *fixed-equilibrium* allocation algorithms. The light lines (-) correspond to the 20 independent parameter sets. The color ribbon marks the band between the 5th and 95th percentiles.

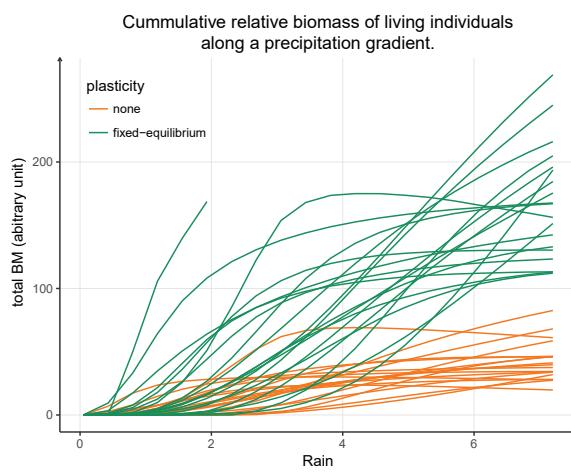


Figure 2.14: Total biomass of all individual along a precipitation gradient for all tested parameter sets. Colour distinguishes plasticity treatments: - *non plastic* & - *fixed-equilibrium*.

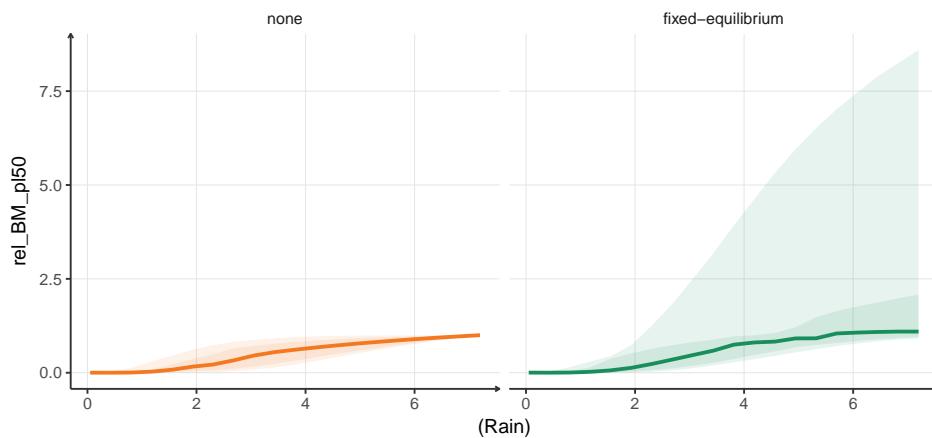


Figure 2.15: Maximum biomass relative to the best performing plant in the most favourable condition for each parameter set, along a precipitation gradient. Colour distinguishes plasticity treatments: - *non plastic* & - *fixed-equilibrium*.

range of the maximum biomass for the given conditions.

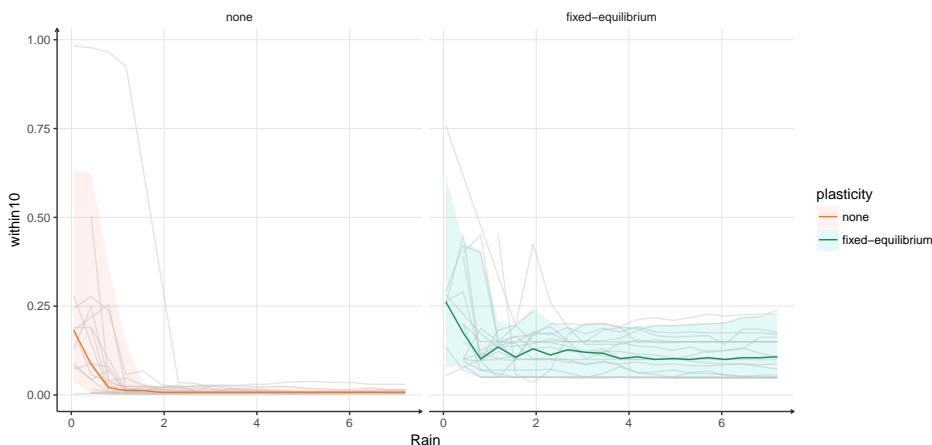


Figure 2.16: Species richness of the species within the range 90%-100% of highest biomass for any given condition (parameter and precipitation) along a precipitation gradient. Colour distinguishes plasticity treatments: - *non plastic* & - *fixed-equilibrium*.

The species richness decreases along the gradient for the two plasticity treatments. The medians of species richness reach the low point for the same precipitation values than the medians of the optimum reach the highest values. The *fixed-equilibrium* simulations show highest species richness along the whole gradient (except for one parameter set).

The functional volume occupied by the top species, also decreases for both plasticity when the precipitations increase. For low watering values, the functional volume of *non plastic* simulation is higher, however this difference disappear when both group of simulations reach low functional diversity.

The median performance of the best performing phenotypes for each conditions of the gradient are compared with and without RMF plasticity (*fixed-equilibrium*) along the gradient. It is limited to the best phenotypes to mimic a degree of biotic filtering. The plasticity greatly enhance the ability of the plants to maintain a high growth, often comparable to the one of the best phenotype, along the gradient. The extreme low value of the gradient show no differences between phenotypes because the water level does not allow plant to grow, only the organic matter contain is the seed is used, leading to similar outputs. As seen in figure 2.12, the best phenotypes often share the same strategy, but differ in

POTENTIAL NICHES

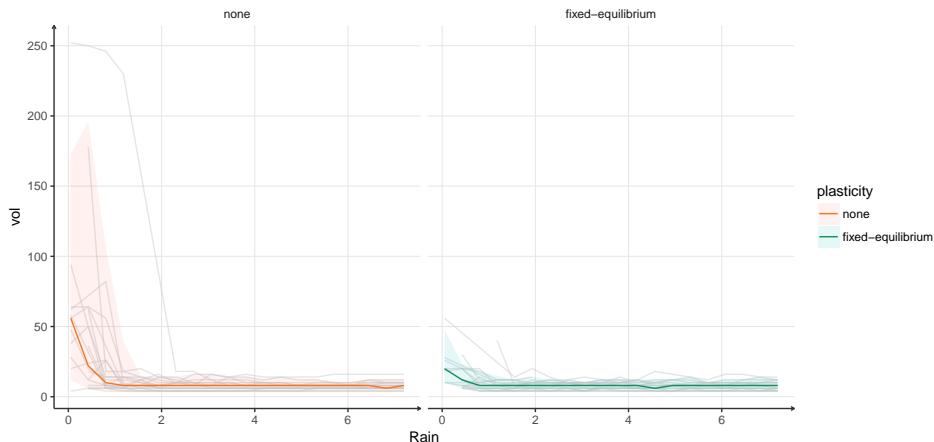


Figure 2.17: Estimation of the functional volume occupied by the species within the range 90%-100% of highest biomass for any given condition (parameter and precipitation) along a precipitation gradient. Colour distinguishes plasticity treatments: - *non plastic* & - *fixed-equilibrium*.

memory for resource availability (figure 2.12). Under *fixed-equilibrium* allocation (right panel), the left end of the gradient (except the first value of precipitation) shows more contrast in the performances in strategies, while the right end (more water) shows very little contrast. This is very different from the *non plastic* allocation results (left panel) that show large differences between the best strategies along the whole gradient.

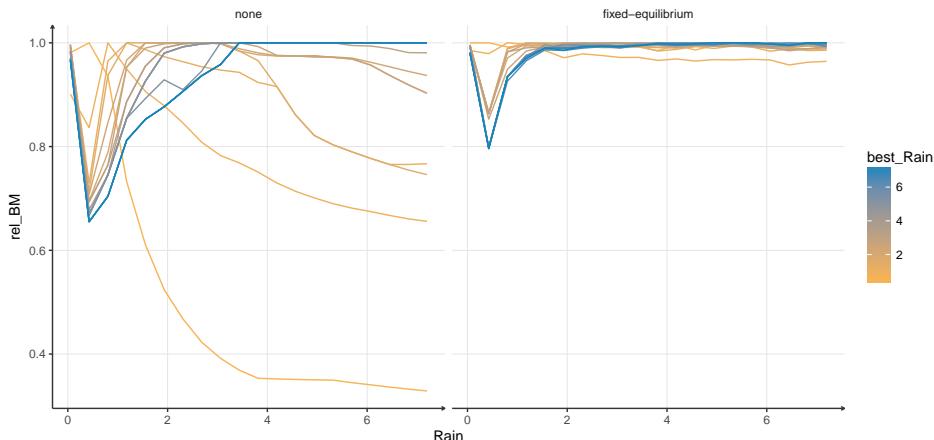


Figure 2.18: Median relative performance of best phenotypes along a precipitation gradient for 20 parameter sets.

2.2.3 Discussion: gradient of homogeneous conditions

Along the watering gradient, the **optimum strategy** (active tissue allocation in roots) changes from conservative toward exploitative. This shift demonstrates that the trade-off between active and structural tissues allocation allows different strategy to dominate in contrasting conditions (**wright_worldwide_2006**). This shift occurs for low values of the gradient and exploitative strategies are dominant over a large part of this gradient. The shape of the relationship results from the gradient including high precipitation values. Also, the low resolution of the strategies (15 values for the proportion of active tissues in roots) limits the possible number of different dominant strategies along such gradient. Because we can see a wide range of optimum of resource-use strategies, and the relation-

STRATEGY SHIFT

ship between this variable and the resource availability is certainly continuous, we can be confident in the pattern observed and a positive relationship between these two variables.

The optimum memory also shows a shift, from low water availability memory in low water conditions, to high values in high water availability conditions. This response is less strong than for the optimum strategy certainly because the relationship is more linear. The fact that the plasticity negates this relationship confirms the role of this variable as a control on the RMF, control that is no longer useful under plastic allocation with no cost.

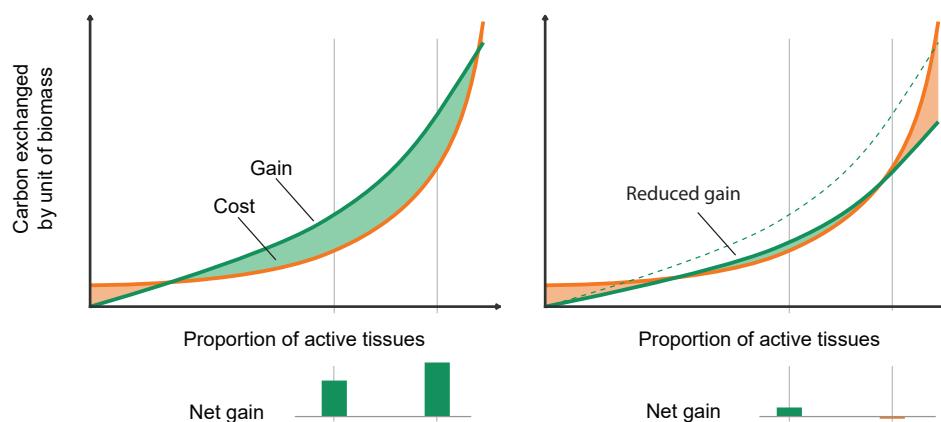


Figure 2.19: **Gain** and **losses** curves along the allocation strategy axis for one organ. Left panel corresponds to a high resource availability, the right panel illustrates the effect of a 30% loss of gain due to a reduction of the resource availability. The bottom bar plots represent the net gain of two distinct phenotype in the two conditions.

The cumulative productivity of all species combined is largely improved by the plasticity for all parameter sets (see figure 2.2.2). However, the best total biomass is only improved for a fraction of these parameter sets (see figure 2.15), while it remains similar in most of others parameter sets between *non plastic* and *fixed-equilibrium* simulations. This observation supports the idea that the total biomass is mostly increased due to an improvement of plants with a non optimum phenotype, and not an improvement of the best phenotypes. Moreover, while the number of species reaching high performance levels increases with plasticity (corroborating the previous conclusion), the functional diversity does not increase. We can conclude that the phenotypic plasticity leads to a convergence of the plants toward good performance phenotypes. Therefore, the productivity gain provided by the plasticity comes mainly from the convergence toward the best fixed phenotype, and cumulative static gain.

STATIC GAIN

This form of gain provided by the phenotypic plasticity can be called **static gain**, as it is capped by the best performing constant phenotype. It is illustrated in the figure 2.20. The gain only comes from the transition from a sub-optimum phenotype toward the best one (see **starting phenotype A** shifting toward the green one). The gain can be quantify as the total biomass difference between the non plastic species (- - dashed line) and the plastic one (- continuous line). In constant condition, the best phenotype cannot benefit from the plasticity because it has the highest growth rate along time and has no static gain (**starting phenotype B**).

The static gain allows species to persist in environmental conditions that do not fit their initial phenotype (**A** – continuous line), and go through the **abiotic filter** while the non plastic equivalents **A - - dashed line**) cannot. This reduction of the impact of the abiotic filter can have large impacts on the community

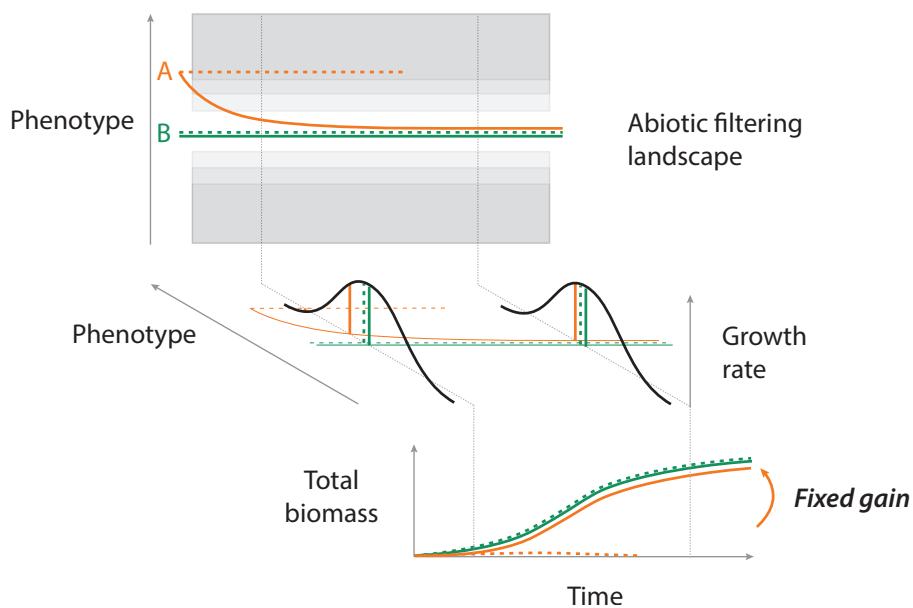


Figure 2.20: Conceptual representation of the abiotic filtering in constant conditions and the illustration of the static gain. The top panel represent the abiotic filtering landscape with a central valley, and the trajectories of the species. The middle panel is the growth rate as a function of the phenotype for two positions in time. The bottom panel illustrates the growth curves for the different phenotypes. Two alternative position **A** and **B** represent a sub-optimum and the optimum position without plasticity (dashed lines). Alternative plastic trajectories are represented by continuous lines.

properties.

The phenotypic plasticity of the RMF leads to an important widening of the potential niche. This is explained by the removal of one constraint of the niche. Indeed, with cost-free plasticity in RMF, the equilibrium is almost guaranteed for all species and the resource-use strategy is the only limitation of a species niche. Because the best proportion of active tissues is the same for a long portion of the gradient (high water availability)(see figure 2.12), along the gradient most of best phenotypes share this resource-use strategy. Therefore, along this same gradient, if the RMF axis is ignored, the different species have equivalent phenotypes (except for a few first growing days).

This niche widening has for consequence a higher niche overlapping. This overlapping can be translated into lower niche differences, as the niche are now discriminated only on one dimension, and into lower fitness differences as the species with similar strategies but different memories have close performances under plastic allocation. According to Turcotte and Levine 2016 these two processes have opposed effects (see figure 3.7 in chapter II). On one hand the reduction of the niche differences diminishes the positive effect of the spatial heterogeneity on diversity. On the other hand, the reduction of fitness differences reduces the competitive exclusion of the non dominant species and limits the abiotic filtering. The current simulations do not allow to tell which effect will be the strongest at the community scale. The cost of plasticity should nevertheless ensure some degree of niche differentiation. In addition to these two effects, the widening of the potential niche corresponds to a reduction of the abiotic filtering pressure, and should promote the diversity as more species can potentially invade an habitat.

The reduction of the abiotic filtering implies a potential increase in the biotic filtering due to the limited carrying capacity of the habitat. This could raise the competition intensity, especially at the beginning of the growing season, and eventually change the dominant species if this increase is strong enough to alter the competition outcome toward more competitive species. Unless the eventual

NICHE WIDENING

COMPETITION EFFECT

new dominant species has a dramatic effect of the overall productivity or diversity, the effect of the phenotypic plasticity through the competition should be positive or null on these properties. But because the phenotypic plasticity does not alter directly the optimum phenotype in temporally fixed conditions, the impact on the dominant resource-use strategy should be rather limited.

While the effects at the community level are still hard to defined, the phenotypic plasticity can alter the dynamic at the meta-community scale. Phenotypic plasticity, by reducing the abiotic filtering effect, allows for stronger link between the communities as the chance to transfer from one community to another are higher. Therefore this mechanism has a positive effect on the stability of the ecosystem as species from unperturbed communities can invade, and partly sustain the properties and services of the perturbed community.

META-COMMUNITY DYNAMICS

In a context of global change, to survive two options are possible: (1) migrate to new habitat with suited conditions, (2) adapt to new condition in the same habitat. In this context, the phenotypic plasticity facilitates both the adaptation to new conditions in the same spatial habitat and the adaptation to a new habitat that match the ideal conditions only to a certain degree. By facilitating the adaptation to new conditions or habitat, the phenotypic plasticity reduces the relative importance of climatic variables compared to the competition as already suggested by empirical results (Alexander, Diez, and Levine 2015).

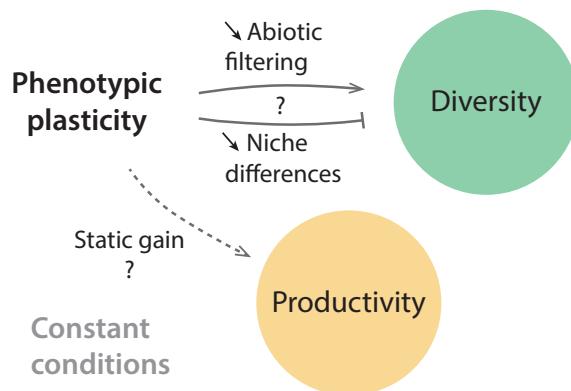


Figure 2.21: Effect of the phenotypic plasticity on the main properties of the grassland communities in constant conditions.

In constant environmental conditions, the phenotypic plasticity already has an impact on species performances and interactions through the static gain it provides to species with the good resource-use strategy. It increases the fitness evenness of species by reducing the abiotic filtering dimensionality. This leads to a convergence of the phenotypes and a widening of the potential niche. The effects at the community level are hard to anticipate, but they will be largely dependent on the competitive interactions and how they are affected by the plasticity. The reduction of the abiotic filtering will likely increase the species diversity, but the functional diversity might not follow this trend due to functional convergence. The effects on the other component of the ecosystem properties cannot be fully determined and greatly depend on the outcome of the competitive interactions. In temporally heterogeneous conditions, the phenotypic plasticity may play a larger role and greatly mediate the inter-specific interactions.

2.2.4 Results: gradient of heterogeneous precipitation conditions

This part of the chapter present the results at the individual scale along a gradient of temporal increasing variability of the underground resource (increasing negative slope of water influx).

The maximum biomass (relative to the *non plastic* best performance in constant watering conditions) decreases along the gradient for all allocation algorithms (see figure 2.22). The *non plastic* algorithm show a drastic drop after the fourth level of variation, while the fixed-trait algorithms show better performances in this part of the gradient. The *plastic-optimisation* algorithm show low growth for all conditions, but a more stable performance.

In contrast with the stable conditions (see previous results) the plasticity provide, in non constant conditions an great improve in the maximum biomass for most parameter sets.

PRODUCTIVITY

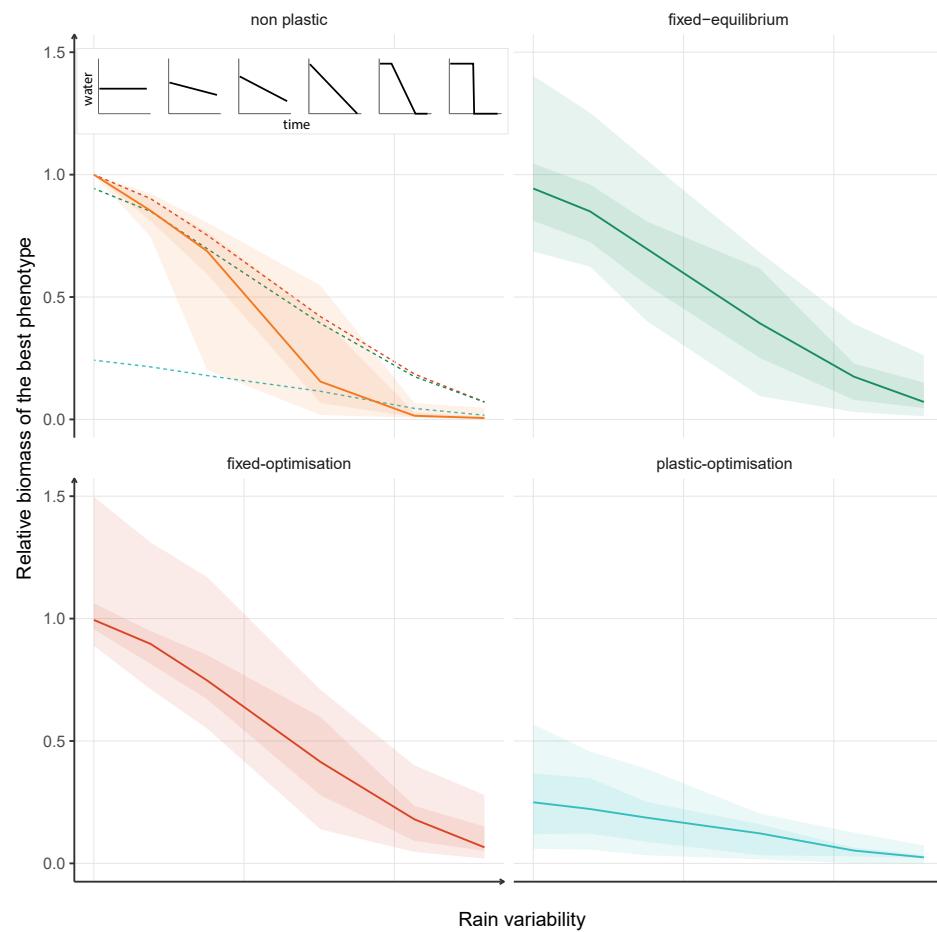


Figure 2.22: Biomass variation of the best performing species along a gradient of resource variability for four plasticity treatment. The upper panel of the top-left frame illustrate the water influx as a function of time along the variability gradient. The dotted lines in the top-left frame indicate the median biomass for the other three algorithms.

In addition to a reduction of biomass, the increasing slope of the water influx reduction lead to a shift of the optimum strategy in *non plastic* simulations (see figures 2.23 & 2.24) toward more conservative strategies. The median optimum value shifts from 0.85 to 0.75 then 0.35 in extrem conditions. This reduction of optimum toward more conservative strategies is offset in most of *fixed-equilibrium* and *fixed-optimisation* simulations where the median optimum value for the PAR stays above 80%. Only a small reduction (around 25%) of the 5th and 25th

IDENTITY

percentiles of the optimum root strategy can be observed between the extreme conditions for these two algorithms.

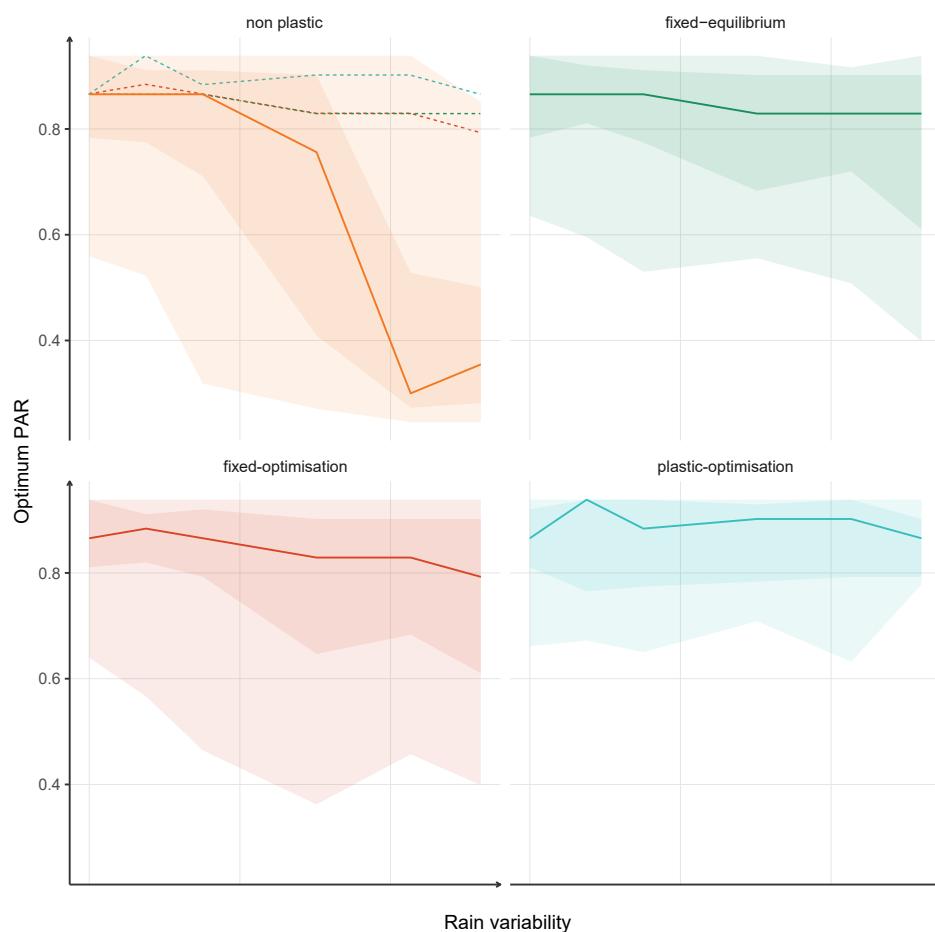


Figure 2.23: Strategy (PAR: proportion of active tissues in root) shift of the best performing species along a gradient of resource variability for four plasticity treatment. The dotted lines in the top-left frame indicate the median PAR for the other three algorithms.

This shift in optimum strategy can better be observed on the plan of the proportion of active tissues in root (PAR) and root mass fraction (RMF) in figure 2.24 where all trajectories¹ along the variability gradient are plotted. *Non plastic* allocation trajectories by a linear shift toward more conservative strategies with higher allocation to roots, while *fixed-equilibrium* and *fixed-optimisation* trajectories are non linear and can be divided into two phases: (1) increase in RMF, (2) reduction of PAR. *Plastic-optimisation* algorithm shows no consistent pattern in trajectories.

¹ trajectory of the optimum, not of the species.

Along the gradient, the performance and the identity of the best phenotype were greatly altered by the water variability, but the phenotypic plasticity mitigate these effects. The number of species is fairly stable for all algorithms but the *non plastic* that show an improvement in species diversity. The fixed allocation algorithm have between 15% & 20% of the species within the 90%-100% range of the maximum biomass for the conditions, while the *non plastic* allocation show this level only of the extreme variable case, but otherwise is limited to a few percents. The functional diversity is stable along the gradient and is similar for all algorithm (*plastic-optimisation* algorithm being exclude)(data not shown).

DIVERSITY

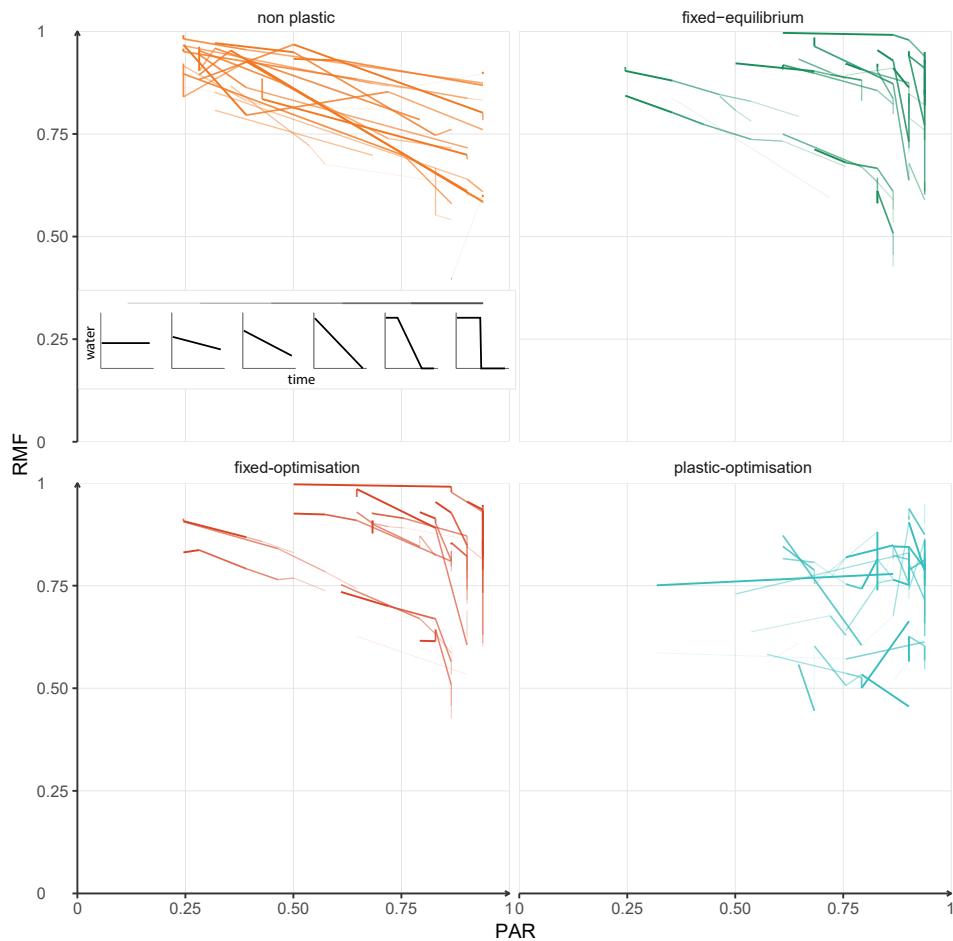


Figure 2.24: Best phenotypes along water resource variability gradient. Thinner lighter lines indicate low water variability, while the thicker lines indicate strong temporal heterogeneity.

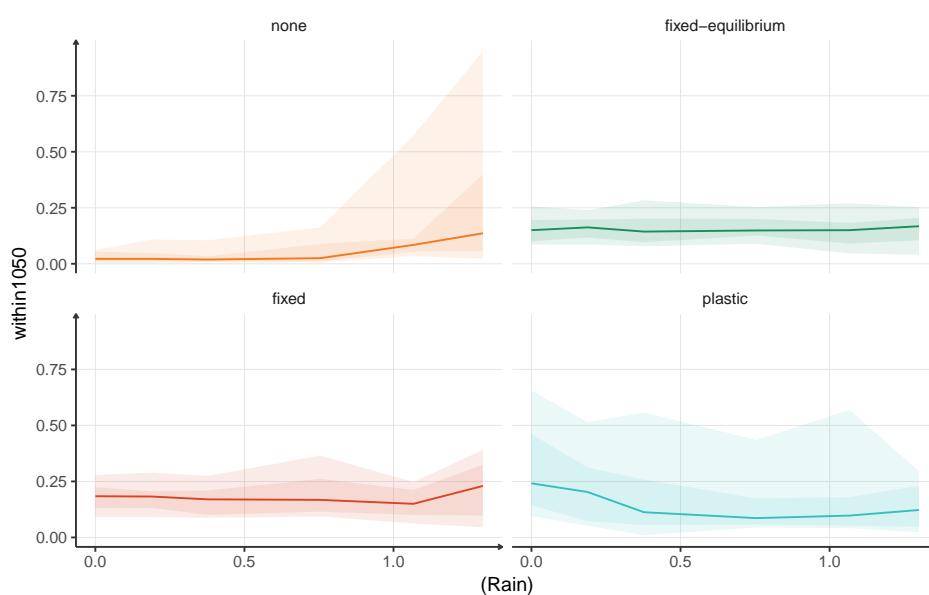


Figure 2.25: Species richness of the species within the range 90%-100% of highest biomass for any given condition (parameter and precipitation) along a water resource variability gradient. Colour distinguishes plasticity treatments: - *non plastic* & - *fixed-equilibrium*.

2.2.5 Discussion: gradient of temporal variations

Before analysing the effect of the different plastic algorithm, the *non plastic* simulations show interesting patterns. The first thing inform us on the relative importance of growing versus surviving. Because the mean water influx over the simulated period is conserved, an increase in the influx negative slope means that during the first half the plants have more available water, while they have less during the second half (relative to constant influx). The decreasing biomass along the gradient suggests that this additional water (and therefore potential growth) does not compensate for a reduction in growth during the drought period. Therefore, it is more important for the plant fitness to limit losses during scarce period instead of maximizing growth during favourable periods.

However, this conclusion should be mitigated by the fact that some reproduction strategy still can benefit from early exploitative strategies and there are not considered here because the success is measure by the biomass afeter 100 days. This effect can be explained by the fact that even the most exploitative species are not fully developed when the resources are the most available, therefore the potential compensation by an early growth is limited. This is the case in the context of mountain grasslands, and some specific life cycle strategies take advantage of this particularity like the *galanthus* species (see figure ??) with early development(schroder_modelling_2014), or development from bulb that allows for an early and rapid growth.

The reduction of biomass along the gradient, even in plastic conditions, supports the idea of a lack of compensation mechanisms between favourable and unfavourable drought periods. In addition to this conclusion, the better performances of the conservative species in low water conditions relative to exploitative species (see figure 2.12), suggests that the best strategy is determined by the optimum resource-use strategy of the most important growth period (the drought period). However, another mechanism can be involved and is relative to the capacity of the species to perform well both in conditions that suit its phenotypes, and in conditions that do not suit its phenotype. In a variable environment, the phenotype does not match conditions different from its niche center because of the optimum resource-use strategy may change, or because the balance between root and shoot activity changes. Both can be important. As said, the former is suggested by previous results, however under plastic allocation the optimum resource-use strategy is maintained at high values of proportion of active tissues. Therefore, the difference in optimum resource-use strategy along the gradient cannot explain alone why conservative strategy are better in contrasted environment. The other explaining mechanism is a better resistance to variability from conservative species, relative to exploitative. The variability in water conditions while the light conditions are fairly constant leads to a shift in the balance in the availability of the two resources. If the phenotype is fixed, a shift from balanced conditions (equilibrium between shoot and root is respected for the given phenotype in these conditions) to unbalanced conditions (no more equilibrium for the same given phenotype) can be seen from the plant perspective as a reduction of the overall gain function stronger than the actual decrease in resource, because the non limiting organ is in over-capacity. Therefore, to the given plant, the decrease in resource is stronger than the actual decrease in resource, and the resource-use strategy shift towards a more conservative strategy

RESISTANCE TO VARIABILITY



Figure 2.26: *Galanthus nivalis* is an example of species that develop early in the season to avoid competition and benefit from the high resource availability.

than a balanced phenotype would require.

While the phenotypic plasticity does not provide any clear advantage to a type of species in constant conditions, in changing conditions the phenotypic plasticity benefit to the exploitative species that are able to maintain good performances in contrasted conditions (see figure 2.23). Because the plasticity gain is assymetric, the phenotypic plasticity alters the identity of the community and promotes exploitative species in variable conditions while conservative species are favoured under *non plastic* allocation. In a more theoretical view, the plasticity can be seen as an alteration of the strength of the stress as shaping factor of community (Grime 1977). If the plasticity allows the exploitative species to better support stress, in modelling studies that do not take into account such plasticity, the amplitude of the stress needed to see a shift in the community identity may be under-estimated.

WHO BENEFIT FROM THE PLASTICITY?

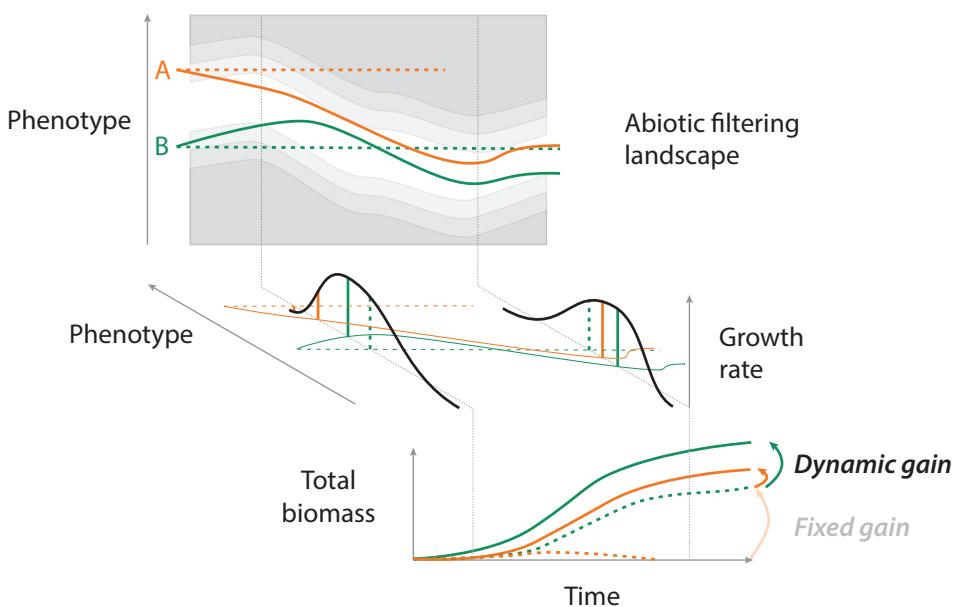


Figure 2.27: Conceptual representation of the abiotic filtering in variable conditions and the illustration of the dynamic gain. The top panel represent the abiotic filtering landscape with a sinuous valley, and the trajectories of the species. The middle panel is the growth rate as a function of the phenotype for two positions in time. The bottom panel illustrates the growth curves for the different phenotypes. Two alternative position A and B represent a sub-optimum and the optimum position without plasticity (dashed lines). Alternative plastic trajectories are represented by continuous lines.

Plasticity has a positive effect on exploitative strategies in low resource availability conditions. As said earlier in the document, plant performance depends on multiple things: the effectiveness of organs, the global resource availability and the equilibrium. If plasticity improve the performance of the exploitative species, it is unlikely to be because of the contraction of the space since it is the optimum only that is looked at. Also means it is probably dynamic gain, if the filter was static, plasticity shouldn't have changed much. And there is gain, even in the high end of the gradient where the optimum does not move, saying that the optimum is fairly conserved despite increase in resource (but low resolution of the strategy space, and match the function).

Change in optimum strategy can be explained by: an assymmetry in efficiency (better a but more conservative in exploitative favourable conditions than exploitative in conservative favourable conditions) or an assymetry in imbalance cost (better be imbalanced when conservative than when exploitative). The first option is not consistent with previous results (see subsection 2.1, figures ...) that show higher fitness for species more exploitative than the optimum, compared to species more conservative. In the other hand, following results show posi-

ABOUT DIVERSITY IMPROVEMENT IN VARIABLE CONDITIONS

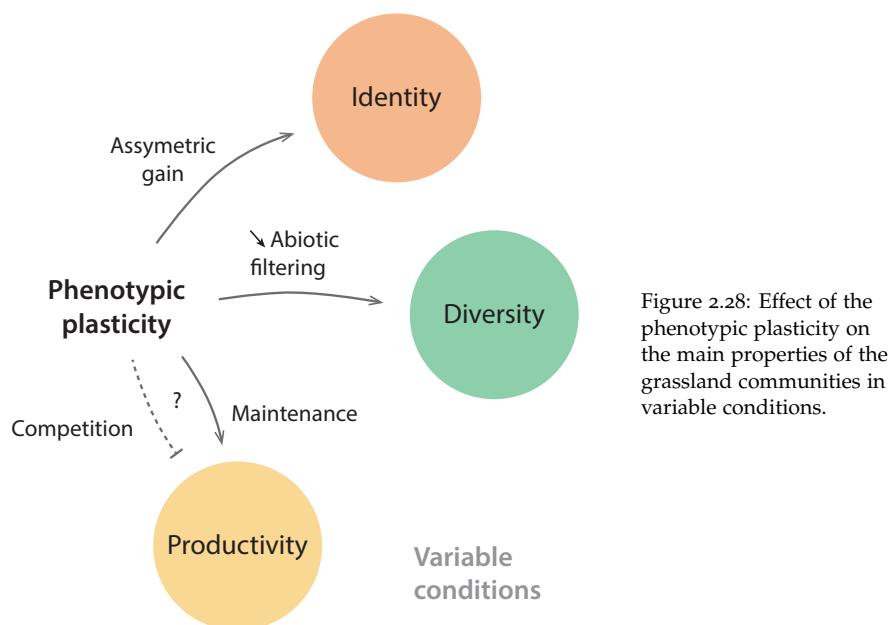
tive effect of functional equilibrium over conservative strategies. Nevertheless, this effect certainly results from an artefact and the contraction of the phenotypic space not in favour of exploitative strategies. The sensitivity of exploitative strategies to low conditions is visible in figure ??

Shift of RMF then strategy explain quite well that the equilibrium is more important than the resource usage and the organ efficiency. How does that inform us on the real world ?

- potentially reduces the meta community diversity if spatial heterogeneity has a less drastic effect on strategic dominance. - talk about that in diversity part /!

may come from a lack of coordination with shoot. Since shoot activity is suppose to be relatively high. == might not have been a good choice to look only at root strategy. But, since there is adjustment of RMF that allow to maintain equilibrium and resource usage, it should be fine to interpret these results.

Phenotypic plasticity give exploitative species an advantage in variable conditions because their growth rate rely more on productivity and therefore equilibrium than conservative species.



The phenotypic plasticity implemented in *MountGrass* improve the relative performance of multiple strategies by concentrating the plant toward a subspace of higher performance for most of plants. Convergence to a smaller subspace can be assimilated to reduction in phenotypic diversity, but it reduce performance heterogeneity and should favour local plant diversity. However, this effect should be limited by plasticity cost. Indeed, if the growth gain due to plasticity is only static, any species with a fixed phenotype closer to the optimum than the focus species has a better growth rate and exclude the focus species. . a few words on dynamics... Meta-community diversity is however reduces by the reduction of potential axis for niche differentiation. Plasticity costs and limits should play major role in the balance between these mechanisms. Community level simulations are needed to further understand the cumulative role of competition, spatial and temporal variability and plasticity

costs on phenotypic plasticity influence on plant community dynamics.

2.3 From model behaviour to competition and coexistence in the real world

*

2.3.1 Plasticity: new functional diversity

functional diversity of plastic traits? Should them be excluded?

Impact of traits, and abundances: the need to account for it!

May still be useful especially for invasion, and works well despite low flexibility (see (Forsman 2014)). May allow more diversity if some correlations with other non plastic traits.

2.3.2 Plasticity as a strategy: cost and correlations

and limits ? what about exhaustion

One of the argument to say this is new, however not really explored, neither with plasticity cost perspectives (a bit with plasticity limits) or with tau. However, used extreme cases: give better understanding and necessary before finer analysis. Still, there are hypothesis on the effect on diversity and the role in phenotypic stability (attention: isn't it just because the formulation of projection is wrong that we can make these conclusions ?).

WHO BENEFIT FROM PLASTICITY?
COST OF PLASTICITY
PLASTICITY AS A STRATEGY

2.3.3 Plasticity and competition: changes in interactions

What about the continuous τ gradient ?

EXTENDED INTERPRETATIONS

What about interactions and cycles ? Little has been discussed on the dynamic of the resource and how it could affect coexistence. Imagine that with cycle, reproduction timing has an importance here...

plasticity will change: performance, sensitivity and impact of the resource.

BIBLIOGRAPHY

- Alexander, Jake M., Jeffrey M. Diez, and Jonathan M. Levine (2015). "Novel competitors shape species/' responses to climate change". en. *Nature* 525.7570, pp. 515–518.
- Crisp, Peter A. et al. (2016). "Reconsidering plant memory: Intersections between stress recovery, RNA turnover, and epigenetics". en. *Science Advances* 2.2, e1501340.
- DeWitt, Thomas J., Andrew Sih, and David Sloan Wilson (1998). "Costs and limits of phenotypic plasticity". *Trends in Ecology & Evolution* 13.2, pp. 77–81.
- Dybzinski, Ray et al. (2011). "Evolutionarily Stable Strategy Carbon Allocation to Foliage, Wood, and Fine Roots in Trees Competing for Light and Nitrogen: An Analytically Tractable, Individual-Based Model and Quantitative Comparisons to Data". *The American Naturalist* 177.2, pp. 153–166.
- Farrior, Caroline E. (2011). *Resource limitation in a competitive context determines complex plant responses to experimental resource additions.*
- (2014). "Competitive optimization models, attempting to understand the diversity of life". en. *New Phytol* 203.4, pp. 1025–1027.
- Forsman, A. (2014). "Rethinking phenotypic plasticity and its consequences for individuals, populations and species". en. *Heredity*.
- Freschet, Grégoire T., Elferra M. Swart, and Johannes H. C. Cornelissen (2015). "Integrated plant phenotypic responses to contrasting above- and below-ground resources: key roles of specific leaf area and root mass fraction". en. *New Phytol* 206.4, pp. 1247–1260.
- Grassein, Fabrice, Irène Till-Bottraud, and Sandra Lavorel (2010). "Plant resource-use strategies: the importance of phenotypic plasticity in response to a productivity gradient for two subalpine species". en. *Ann Bot* 106.4, pp. 637–645.
- Grime, J. P. (1977). "Evidence for the Existence of Three Primary Strategies in Plants and Its Relevance to Ecological and Evolutionary Theory". *The American Naturalist* 111.982, pp. 1169–1194.
- Grimm, Volker et al. (2005). "Pattern-Oriented Modeling of Agent-Based Complex Systems: Lessons from Ecology". *Science. New Series* 310.5750, pp. 987–991.
- Guisan, Antoine and Wilfried Thuiller (2005). "Predicting species distribution: offering more than simple habitat models". en. *Ecology Letters* 8.9, pp. 993–1009.
- Kleidon, Axel and Harold A. Mooney (2000). "A global distribution of biodiversity inferred from climatic constraints: results from a process-based modelling study". en. *Global Change Biology* 6.5, pp. 507–523.
- Liu, F. and H. Stützel (2004). "Biomass partitioning, specific leaf area, and water use efficiency of vegetable amaranth (*Amaranthus* spp.) in response to drought stress". *Scientia Horticulturae* 102.1, pp. 15–27.
- Luke McCormack, M. et al. (2012). "Predicting fine root lifespan from plant functional traits in temperate trees". en. *New Phytologist* 195.4, pp. 823–831.
- Peterson, K. M. and W. D. Billings (1982). "Growth of Alpine Plants under Controlled Drought". *Arctic and Alpine Research* 14.3, pp. 189–194.
- Poorter, Hendrik, Ülo Niinemets, et al. (2009). "Causes and consequences of variation in leaf mass per area (LMA): a meta-analysis". en. *New Phytologist* 182.3, pp. 565–588.
- Poorter, Hendrik, Karl J. Niklas, et al. (2012). "Biomass allocation to leaves, stems and roots: meta-analyses of interspecific variation and environmental control". en. *New Phytologist* 193.1, pp. 30–50.

- Reich, Peter B. (2014). "The world-wide 'fast–slow' plant economics spectrum: a traits manifesto". en. *J Ecol* 102.2, pp. 275–301.
- Reineking, Björn et al. (2006). "Environmental variability and allocation trade-offs maintain species diversity in a process-based model of succulent plant communities". *Ecological Modelling*. Pattern and Processes of Dynamic Mosaic Landscapes – Modelling, Simulation, and Implications 199.4, pp. 486–504.
- Ryser, Peter and Liina Eek (2000). "Consequences of Phenotypic Plasticity vs. Interspecific Differences in Leaf and Root Traits for Acquisition of Aboveground and Belowground Resources". *American Journal of Botany* 87.3, pp. 402–411.
- Ryser, Peter and Pille Urbas (2000). "Ecological Significance of Leaf Life Span among Central European Grass Species". *Oikos* 91.1, pp. 41–50.
- Taubert, Franziska (2014). "Modelling and Analysing the Structure and Dynamics of Species-rich Grasslands and Forests". PhD thesis. Osnabrück.
- Tjoelker, M. G. et al. (2005). "Linking Leaf and Root Trait Syndromes among 39 Grassland and Savannah Species". *New Phytologist* 167.2, pp. 493–508.
- Turcotte, Martin M. and Jonathan M. Levine (2016). "Phenotypic Plasticity and Species Coexistence". *Trends in Ecology & Evolution* 31.10, pp. 803–813.
- Van Kleunen, Mark and Markus Fischer (2005). "Constraints on the evolution of adaptive phenotypic plasticity in plants". en. *New Phytologist* 166.1, pp. 49–60.
- Wright, Ian J. et al. (2004). "The worldwide leaf economics spectrum". en. *Nature* 428.6985, pp. 821–827.

V

COMMUNITY DYNAMICS

This second result chapter examines the effects of phenotypic plasticity at the scale of the community. Another parameter filtering processes is performed and described in the first section of this chapter. The second part focuses of the effects of plasticity of the main properties of the community. The impact of plasticity on species diversity is particularly investigated. This chapter gives a glimpse of the potential of the model to answer various questions around the role of intraspecific variations on diverse community properties.

1 COMMUNITY LEVEL SIMULATIONS: NON PLASTIC COMMUNITY

1.1 Parameter filtering

1.1.1 Method

Weather data for the time period between 1959 and 2014 has been computed by the MeteoFrance model SAFRAN by ... using GPS coordinates, slope, azimuth and horizon computed from a digital elevation model. These parameters were also used by the model CROCUS to compute snow accumulation and snow melting. These high frequency data (resolution under 1h) have been averaged on a daily time-step and used to compute input variables for *MountGrass*. The snow in particular defines the length of the growing season starting with the first snow melt of the year and finishing the day of the first snow fall of autumn or winter.

WEATHER DATA

The simulated years above 2014 are randomly sampled from the existing dataset between 1995 and 2014.

Community level parameter filtering is conducted for a new table of parameter sets. These parameter sets are ... from accepted parameters and joined with LHS random sampling for five community level parameters: seed germination density, drought mortality, ageing mortality, plasticity cost for environmental sensing and plasticity cost for trait changes (see chapter 4 for details).

PARAMETER FILTERING

Few words on why plasticity cost parameters: time limits, distinguish the benefit of plasticity itself, not combined effect. Should have done simulations with no cost to have an idea of plasticity cost effect.

The simulations run over 300 hundreds years for 6 sites described in table ?? on squares of ... square centimetres. The simulation is stopped and the parameter set rejected if no individual persist and the seedbank is empty. The seedbank is composed of seeds contained in the seedbank and seeds from the metacommunity. The total of seeds is defined by the seed germination density and the area simulated. The seeds from the simulated community represent up to 80 % of the seedbank, less if the seed production is limiting. The first ... years are not taken into account in the filtering process to let the community settle.

1.1.2 Results

SIMULATIONS

Simulations done. Need to illustrate the results.

On stability and on diversity (functional and species)

EFFECT OF PARAMETERS

Random forest approaches like sensitivity analysis at individual scale.

1.2 Non plastic communities

Trade-off, diversity, stability ...

Is there a selection of some parameters ? Are there ecological trade-off (resource use strategy and reproduction) emerging from the model ?

ECOLOGICAL TRADE-OFF ?

2 PLASTICITY: IMPACT ON SPECIES FITNESS AND DIVERSITY

Plasticity in integrated framework and full community simulations. Plasticity mechanisms, but also plasticity as a strategy (look at the cost and tau).

Effects on productivity and coexistence. Difference in the correlation ?

Effect of tau on persistence.

2.1 Plasticity and diversity

Now

2.1.1 Method

To test the effect of plasticity on coexistence and community dynamics, runs from the parameter filtering are used as starting points to limit the simulation time of the stabilisation phase. For each parameter set tested, 6 different sites were tested during the calibration phases, 77 parameter sets were accepted and a sample of ... were tested, resulting in ... communities. Each of those is the starting point of three parallel runs that differ only by the allocation algorithm used: *non plastic*, *fixed-equilibrium* and *plastic-optimisation*. The *fixed-equilibrium* is favoured to *fixed-optimisation* algorithm because previous part of the document focused on this algorithm and because it is simpler to analyse. The *plastic-optimisation* algorithm is still simulated, despite the relatively poor performance results observed in constant conditions and the high convergence, because the introduction of plasticity cost, continuous species specific plasticity ($0 < \tau < 1$), and temporal and spatial heterogeneity should mitigate the negative sides of this allocation mechanism and give information of processes at stake.

SIMULATIONS

The plasticity costs (maintenance: related to the value of τ , and displacement: relative to changes in phenotypes) defined in the parameter sets are applied to all algorithms. In *non plastic* simulations, this results in artificial additional costs to species with low values of τ but with no potential gain from plasticity as the allocation is non plastic.

The differences of effects between the different types of plasticity on the variables of interest are computed unpaired Wilcoxon tests assuming an independence of the the different data points. This assumption of statistical independence is justified by the normalisation for each parameter set of the variables relative to the mean of the *non plastic* group. This normalisation allows to compare the simulations between the parameters sets. The interactions and other level (site, and autocorrelations) are discussed later in this section.

The normalisation $Vn_{a,p,t,s}$ of the variable V for the allocation algorithm a , the parameter set p , the time t and the site s is given by the following formula:

$$Vn_{a,p,t,s} = \frac{V_{a,p,t,s}}{\bar{V}} \quad (2.1)$$

$$\bar{V} = \frac{\sum_{a=\text{non plastic}} Vn_{p,t,s}}{n} \quad (2.2)$$

where n is the number of observations for the *non plastic* algorithm.

2.1.2 Results

The level of coexistence is evaluated by the number of distinct species that manage to maintain at least one individual or produce at least one seed at the end of the season. This criterion allow to ignore the potential non stable diversity introduced by the "meta-community invasion" (sampling of species in the meta-community pool) and to consider species that can be filtered out sue to seed mortality. The number of species increases in almost all simulated years and sites for both plastic allocation algorithms, with a median of 1.5 times the number of species in *non plastic* simulations (see figure 2.1.2). This factor can go up to 6 for *fixed-equilibrium* and 9 for *plastic optimisation*.

The effect of plasticity on coexistence is driven by the benefits of plasticity at the individual scale. These benefits are mitigated by the cost of plasticity, particularly the maintenance cost that affect all species relatively to their potential plasticity.

Low values of plasticity maintenance cost (see figure ??) show higher diversity for both plastic allocation algorithms. This trend is consistent across sites despitess some inter-annual variability in the diversity. The effect is a bit less stronger for *fixed-equilibrium* than for *plastic-optimisation* (as already observed in figure 2.1.2).

The mechanisms through which the phenotypic plasticity impacts species richness are multiple (refer to the figure 3.7 in chapter II). However it is hard to disentangle them all.

The density can be affected by the phenotypic plasticity leading to higher species diversityLepik, Liira, and Zobel 2005, higher density leading to the sampling of more species. The density, estimated by the number of individual after the reproduction phase (persisting individuals and produced seeds), is consistently higher in *plastic* simulations, but the difference is relatively low (around

STATISTICAL TESTS

EFFECT ON COEXISTENCE

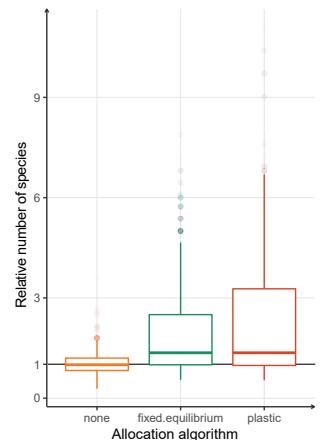


Figure 2.1: Relative species richness in the three plasticity treatment. To negate the variability due to the parameter sets, the realised number of species is divided by the median number of species in *non plastic* treatment for each parameter set. The variability is due to random invasion and climatic variability (inter-sites and inter-seasons).

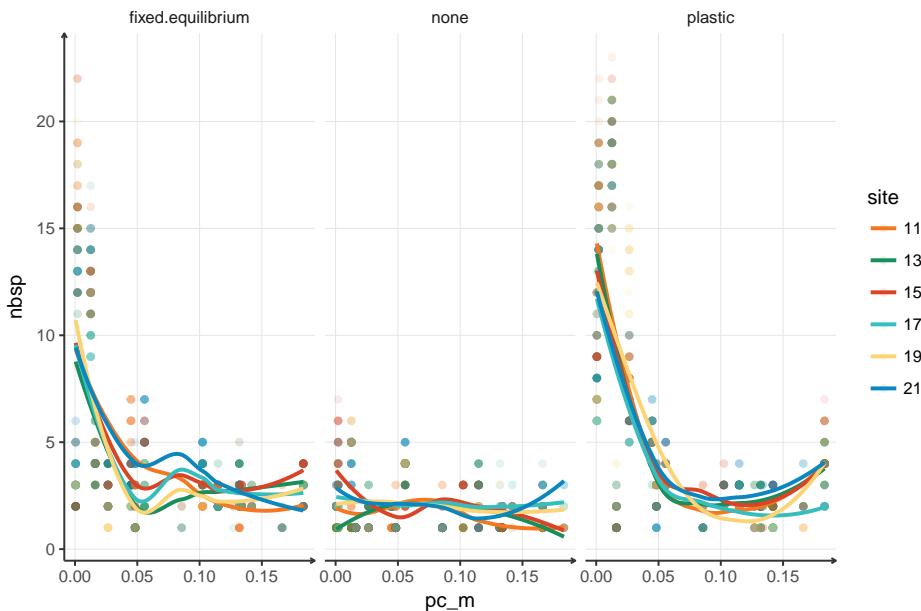


Figure 2.2: Effect of the cost of plasticity-maintenance on the absolute number of species in the three plasticity treatments. Individual season values (points) and site-specific trends (gam smoothing line) are represented.

3% higher than the *non plastic* median density) and an order of magnitude lower than inter-annual and inter-site variations that can go up to 40% difference relative to the median density (for any given parameter set).

If the plant estimated density increases, the number of species relative to the plant density ... with plasticity.

Figure 2.4: Species richness relative to plant density for the three plasticity treatments.

Productivity is also susceptible to be impacted by the phenotypic plasticity at the community level. Multiple mechanisms can be involved, but in any case higher productivity is achieved by high efficiency in the resources given, and plasticity can affect this efficiency at individual level (with positive effects as observed in section 2) or at the community scale with changes in the dominant species and competition intensity.

The productivity of the *non plastic*, *fixed-equilibrium* and *plastic-optimisation* allocation algorithm show little differences. The *non plastic* simulations average biomass tend to be a bit higher in certain cases. Like the diversity and density, the normalised yearly average biomass does not show great variations between plasticity, but higher variability between sites and seasons. *Non plastic* and *fixed-equilibrium* median are quite similar, and the *plastic-optimisation* show lower productivity than the other two algorithms.

The allocation algorithm is expected to alter the fitness of potentially plastic plants. The selective effect of the allocation algorithms is investigated by plotting the τ value of species that are maintained in only one of the algorithms. Because of the plasticity cost, the selection of species with low values of τ signifies an improvement of the fitness due to plasticity. The distribution of τ is fairly high for *non plastic* species and almost 75% of the species have a value above 0.8, whereas *fixed-equilibrium* specific species have lower values ranging from 0.2 to

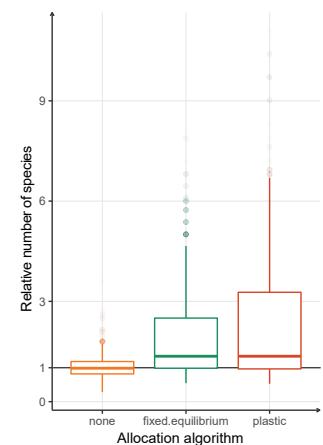


Figure 2.3: Relative plant density in the three plasticity treatment. To negate the variability due to the parameter sets, the realised number of plant is divided by the mean number of plant in *non plastic* treatment for each parameter set. The plant density is estimated with the output of the reproduction process.

PRODUCTIVITY

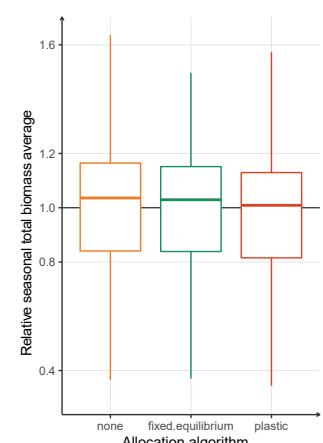


Figure 2.5: Average total biomass relative to *non plastic* simulations, in

1 with the median around 0.7 and the *plastic-optimisation* species have even lower values with a median around 0.55.

Select different strat? meh, from very site specific strats (one dominant species), to more variable within the site, but less differences between the strats. Shift from beta diversity to alpha diversity.

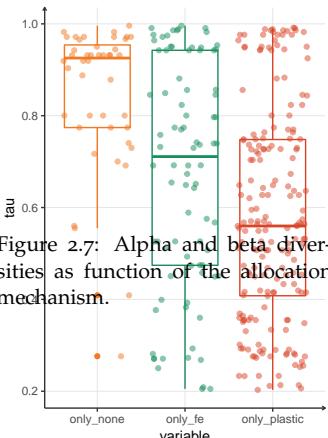
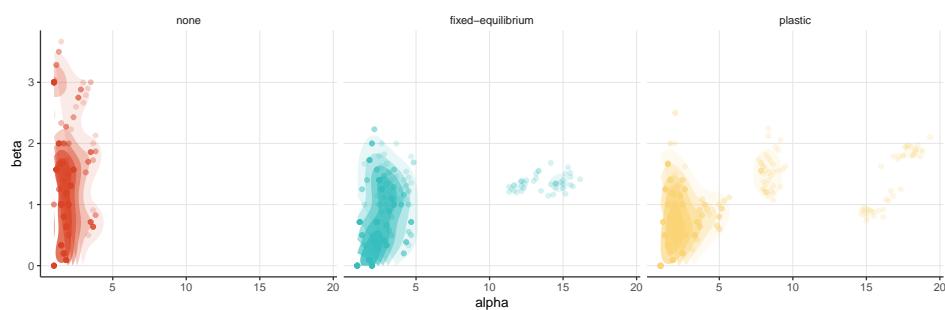


Figure 2.6: Plasticity levels of species that are present in only one type of plastic treatment.

VARIABLE STRATS
DIFFERENT DIVERSITIES

2.1.3 Discussion

Plasticity allows the emergence of new phenotypes that are plastic. Leading to higher density in competitors, and greater evenness. Hard to detect because not the same number of species, or require other experiments.

also, Jung et al. 2014 show contrasting response between species and within species - might not be the best

Plasticity allows for bigger niche (variability dimension), more chance to build enough "growth potential" to persist. Otherwise, other species that are dominant, because other species can settle, take advantage of it. Should look at the growth rates hierarchy for a couple of simulations.

Interaction between plasticity effects and parameter sets, but the interest here, even if it is interesting. Higher variance due to site and weather. The almost perfect knowledge of models allows an extremely precise decomposition of the effects, but at the risks of loosing broad effects. The difficulty is to measure the relative strength of these effects, and generalise. But, by essence, the parameters are suspected to have a significant effect that can be identified, otherwise it would not have been included in the model.

2.1.4 Phenotypic plasticity and global change

climate
management

COMPETITION EVENNESS

SHIFT IN STRATEGY

DIFFERENT MECHANISMS FOR
DIFFERENT DIVERSITIES

INDEPENDENCE OF POINTS

BIBLIOGRAPHY

- Jung, Vincent et al. (2014). "Intraspecific trait variability mediates the response of subalpine grassland communities to extreme drought events". en. *J Ecol* 102.1, pp. 45–53.
- Lepik, Mari, Jaan Liira, and Kristjan Zobel (2005). "High shoot plasticity favours plant coexistence in herbaceous vegetation". en. *Oecologia* 145.3, pp. 465–474.

VI

SYNTHESIS & OUTLOOK

1 SYNTHESIS

Point out the novelty, achieved work

begin to fill the gaps of community dynamics: (Berger et al. 2008): effect on local environment, adaptive behaviour and below-ground.

Plus: lack of diversity.

1.1 Modelling diverse community

It grows.

It's diverse.

It's stable.

It's driver dependant? (at least at individual level).

1.2 Effect of plasticity of mountain grasslands properties

Did a bit what I accused other model to do: have a discrete conception of plasticity. However, the results at the community scale are encouraging and demonstrate the interest of such approach.

1.2.1 Identity

1.2.2 Productivity

1.2.3 Diversity

1.3 On plasticity modelling

One strong assumption this modelling relied on was the existence of a strong link between fitness and environmental condition. This has been proven to be partially true as *MountGrass* was able to express improvements in fitness thanks to plasticity. However, in some situations, the plasticity leads to reduction in fitness, or eventually to complete phenotypic dead-end. The temporal dimension of plant growth, and the difficulty to capture that makes this assumption hard to maintain in such complex systems with strong dynamics. Moreover, such assumption does not necessarily take into account competitive behaviours better captured by game theory and other modelling approaches Farrior 2011; Dybzinski et al. 2011.

1.3.1 How to make it work better, with what consequences

Despite talking about molecular basis of the plasticity, did not really make use of this knowledge. Should better use biological idea (even if a bit more complex). Especially for memory and recovery: idea of stress, memory and loss of memory (recover) to avoid maladaptive responses (Crisp et al. 2016) ! ! ! seee also cues reliability (Simons 2014) and (Scheiner and Lyman 1989; Scheiner 2002; Scheiner and Holt 2012; Scheiner 2013)

ON MEMORY

Non composite traits (here SLA plastic only because of density, but does not consider thickness plasticity).

ON PLASTIC TRAITS

Fitness proxy ? yeah Ryser and Eek 2000 resource use optim by pl, but Franklin et al. 2012 too variable ... stess response oriented.

ON DRIVERS

Nitrogen instead of water, leaves do not respond to water changes (unless low nitrogen because water limitation is more nitrogen limitation really (Farrior 2014)), does not work well with integrative vision to ignore nitrogen.

1.3.2 Genericity and extensions

About extensions, but the process, despite failing completely capture real growth dynamics (but we already discussed why), built the fundations for resistance-risk plasticity with accumulation of risk cues, extend risk avoidance - resistance trade-off (similar than droudtg see Kooyers 2015?) for herbivory.

1.4 The limit of the species.

Refer to the litterature review part.

In this work, but also because of the improvement of molecular biology, and the deeper and deeper dive ecology is doing within individual, the limits of species are fuzzy (started with trait and the introduction of continuity). At some point, there will be a need for a way to go back from the a space of numerous continuous dimensions to the species. Also, understanding species as evolving 3D objects, where the different aspects of intra-specific variations play different shaping roles.

2 OUTLOOK

Further reading, thinking and rambling about what's developed in the papers.

2.0.1 Better calibration

Better implementation rcpp or data table struture, plasticity mech, to allow bayesian and pattern-oriented calibration. A lot of species and a lot of parameters. Difficult exercice. But strategies should be limited by functions and trade-off, just need to calibrate shared (process related) parameters and not species specific (strategic) paramters. This is what allows the modelling of diverse community.

2.1 Competition and feedback

This document focuses on how the plant are doing with the given resources (arrow in fig in margin). However, a key element in competition and resource dynamics (point that separate Tilman appraoches from Chesson) is the impact of plant on resource (fig in margin). Both are fundamental for the understanding on plant interactions, and I argue that understanding the former is necessary to understand the later and have a global view on plant competitive interactions on resources. blablabla competition experiments, resistance to resource shortening (Tilman) and relative homogeneity of resource (homogeneous in influx, content, starting pool, ... ?). Using the term homogeneous allows to use fixed terms and processes, while to me there is a ambiguity around competition that can be seen as: (1) the impact on growth, (2) the winner out of a competitive scenario (with resource shortening). In this later case, the approach of part 4 (?) has limited interpretation since they are not competing. We can intuitively imagine (from our understanding of model's functioning) that there is a hierarchical effect on growth, but that is probably reversed in case of (1) shared resource pool (big plant may have access to bigger resource pool in open environment), (2) sufficiently quick resource shortening to lead to death events.

in margin: figure resource and interaction.

figure competition decomposition of fitness (growth and survival), and growth related to resource pool (try to have graph approach).

competition change vegetation response to climate change (Loon et al. 2014)

2.2 Extend to climate change effects

How plasticity actually affect the effects of climate change: mitigate or amplify, risk of critical transition.

drought resistance experiments to be done.

Higher diversity: higher risk of invasion?

Take advantage of simulated scenarios of climate change.

3 EXTENSIONS

This section is meant to include thoughts and ideas on how to extend *MountGrass* but that could not be included in the first versions of the model for various reasons. Despite not being included, these extensions are interesting from a scientific or technical point of view, and I hope these notes can be useful to anyone interested in *MountGrass* or individual based vegetation modelling.

3.1 Ecology of plasticity: plasticity as a trait

Show that plasticity could be modelled as a continuous strategy. But did not explore enough trade-offs with other traits, or as a propertie of the community.

Simulations with more species, similar trait but plasticity: trade-off and selection, role of plasticity cost on these relationship.

Effect of plasticity of resistance and resilience. Suppose a great effect on resistance and resilience, but will certainly depends on the selection or history of plastic species.

Use the model as a evolution model with mutations in plastic parameters (tau and functions).

3.2 Include nitrogen: source of trade-off

As seen previously in chapter , the emergence of trade-off in growth strategy in the actual framework actually rely on a strong genetic constraint over plant plasticity. Indeed, without plasticity cost and low reactivity there would be a high rate of phenotypic convergence of individuals from different species. This is explained by the existence of optimum carbon partitioning (for a given size) in a stable environment. The coexistence of different resource use strategies (exploitative vs conservative) is allowed only through temporal variations and non equilibrium state. This is quite common since a lot of models will predict rapid dominance of one entity in case of equilibrium (need references here).

Multiple questions arise from this observation: are the conclusions of this work still interesting in the understanding of the coexistence mechanisms? (I hope I did convince you in the dedicated part of this document, see .. for more details), is it possible to see coexistence of multiple strategies in a temporally stable environment? how can we produce trade-off by including only one more resource?

In the following paragraphs I try to answer these questions with theoretical arguments and suggestions on how to integrate them in *MountGrass* .

Coexistence mechanisms are listed and detailed in the introduction of this thesis (see chapter 1). Here I focus on the efficiency of tissues... Nitrogen based,

why coexistence ? different phenotype correspond to different limiting resources and for different resource availabilities, different phenotype will optimize the return cost of tissues. Nitrogen also allow the model to have an extra dimension into strategy: WUE (local scale) versus NUE (global scale) (element of reflexion in Maire's thesis).

Its also can be related to

Original idea was to have specific carbon pools for different function, and weight the relative allocation based on gain projections.

3.2.1 Resistance carbon pools

Pierik and Testerink 2014 the art of being flexible: multiple resistances

3.3 For more interaction

This model, thanks to paired simulations should be used to explore the effect of plasticity on interactions and competition.

Understanding impact of plasticity on fundamental interaction could nourish the theoretical work on coexistence by linking mechanistic model observations and understanding with more abstract work on the basis of coexistence.

game theory farrior and dybinsky: evolutionary dynamics

BIBLIOGRAPHY

- Berger, Uta et al. (2008). "Competition among plants: Concepts, individual-based modelling approaches, and a proposal for a future research strategy". *Perspectives in Plant Ecology, Evolution and Systematics*. Space matters - Novel developments in plant ecology through spatial modelling 9.3–4, pp. 121–135.
- Crisp, Peter A. et al. (2016). "Reconsidering plant memory: Intersections between stress recovery, RNA turnover, and epigenetics". en. *Science Advances* 2.2, e1501340.
- Dybzinski, Ray et al. (2011). "Evolutionarily Stable Strategy Carbon Allocation to Foliage, Wood, and Fine Roots in Trees Competing for Light and Nitrogen: An Analytically Tractable, Individual-Based Model and Quantitative Comparisons to Data". *The American Naturalist* 177.2, pp. 153–166.
- Farrior, Caroline E. (2011). *Resource limitation in a competitive context determines complex plant responses to experimental resource additions*.
- (2014). "Competitive optimization models, attempting to understand the diversity of life". en. *New Phytol* 203.4, pp. 1025–1027.
- Franklin, Oskar et al. (2012). "Modeling carbon allocation in trees: a search for principles". en. *Tree Physiol* 32.6, pp. 648–666.
- Kooyers, Nicholas J. (2015). "The evolution of drought escape and avoidance in natural herbaceous populations". *Plant Science* 234, pp. 155–162.
- Loon, Marloes P. van et al. (2014). "How light competition between plants affects their response to climate change". en. *New Phytol* 203.4, pp. 1253–1265.
- Pierik, Ronald and Christa Testerink (2014). "The Art of Being Flexible: How to Escape from Shade, Salt, and Drought". en. *Plant Physiol*. 166.1, pp. 5–22.
- Ryser, Peter and Liina Eek (2000). "Consequences of Phenotypic Plasticity vs. Interspecific Differences in Leaf and Root Traits for Acquisition of Aboveground and Belowground Resources". *American Journal of Botany* 87.3, pp. 402–411.
- Scheiner, Samuel M. (2002). "The genetics of phenotypic plasticity. VII. Evolution in a spatially-structured environment". *Journal of Evolutionary Biology* 11.3, pp. 303–320.
- (2013). "The genetics of phenotypic plasticity. XII. Temporal and spatial heterogeneity". en. *Ecol Evol* 3.13, pp. 4596–4609.
- Scheiner, Samuel M. and Robert D. Holt (2012). "The genetics of phenotypic plasticity. X. Variation versus uncertainty". en. *Ecology and Evolution* 2.4, pp. 751–767.
- Scheiner, Samuel M. and Richard F. Lyman (1989). "The genetics of phenotypic plasticity I. Heritability". en. *Journal of Evolutionary Biology* 2.2, pp. 95–107.
- Simons, A. M. (2014). "Playing smart vs. playing safe: the joint expression of phenotypic plasticity and potential bet hedging across and within thermal environments". en. *J. Evol. Biol.* 27.6, pp. 1047–1056.

GLOSSARY

active plasticity Change in phenotype controlled by internal regulation processes. Opposed to passive response. *i.e.* change in SLA when light is limiting is an active plastic response.

allocation rule The allocation rule is the set of rules that determine the target phenotype of a plant considering its actual phenotype, the biomass available and the projection of external conditions. It can be decomposed in two main parts: the plastic dimensions, and the fitness proxy function (or gain function). Allocation rule is also designated as allocation algorithm, plasticity rule or plasticity algorithm.

Plasticity

4 INDEX

community weighted mean, 29
diversity, 30, 44
dynamics global vegetation models, 32
individual-based models, 32
individual-based-models, 7
leaf economic spectrum, 26
mass ratio hypothesis, 29
productivity, 30
abiotic filtering, 18
abiotic filter, 127
allocation rules, 101
allocation rule, 62
confidence in species memory, 63
controlling factor, 62
costs, 42
diversity, 3
dynamic gain, 119
ecosystem properties, 2, 28
ecosystem services, 1
ecosystems, 1

equilibrium, 116, 117
facilitation, 18
fixed, 62
functional diversity, 117
functional identity, 29
gravity center, 111
growth function, 40
heterogeneity, 122
identity, 2, 46
intra-specific variability, 33
memory, 101
mountain grasslands, 1
niche, 17, 44
optimum strategy, 126
optimum tissue efficiency, 116
organ tissue efficiency, 117
parameter filtering, 89
parametrisation, 89
performance landscape, 117
phenotypic flexibility, 99

plant performance, 115
plastic exhaustion, 121
potential niche, 18
productivity, 3
projection, 62
realised niche, 18
regulate, 63
resolution, 55
scale, 55
sensitivity analysis, 89
shared, 62
species diversity, 117
speed, 117
stability, 63
static gain, 119, 127
strategy space, 28
strategy, 63
trade-off, 26