

Coping with climate change.

**Implications of the Yellow-Bellied Marmot's (*Marmota flaviventeris*)
body mass evolution in the last half-century.**



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Comprehensive exam proposal

by

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Introduction

Climate change

Climate change is unequivocally recognized as one of the most pressing challenges of our time. Its impacts (e.g., melting polar ice caps, rising sea levels) are well documented and increasingly evident. More specifically, this global phenomenon is characterized by **rising temperatures, changing season lengths, increased environmental variability and unpredictability, and a growing frequency and severity of droughts and extreme weather events** (Intergovernmental Panel On Climate Change (Ipcc) 2022).

Climate change impacts on human society are, for example, well represented in the city of Ottawa. Temperature, snowing and raining trends in the Canadian capital over the last century, reveals worrying figures (Walsh and Patterson 2022), and future projections are not much more reassuring (Zhai et al. 2019). One good illustration of that is the management of the Rideau canal ice skating rink. Indeed, in later years the opening of the world's longest ice staking risk has been more and more uncertain, and its future is unfortunately but, fatally, questionable.

But above all, as shown by countless studies, climate change deeply impacts the vast majority of Earth's ecosystems (Intergovernmental Panel On Climate Change (Ipcc) 2022). These profound ecological upheavals put numerous species at risk, which must act accordingly to avoid extinction, either by dispersing or adapting (Gienapp and Brommer 2014). It is crucial to improve our comprehension of how natural population cope with this rapid and unpredictable changes in order to conduct efficient conservation policies.

Body mass as a Life-History Traits (LHT)

Life history traits (LHTs) are phenotypic characteristics that directly impact an individual's survival and reproduction success, thereby determining its overall “**fitness**” (Roff 1992). The concept of life history theory relies on the hypothesis that organisms have limited resources and must allocate them strategically between competing functions such as growth, maintenance, and reproduction (Roff 1992; Stearns 1992).

Various traits can be considered as LHTs, as they represent the outcome of this resource allocation process. Among these, body mass plays a crucial role in many species. Individuals with greater body mass will have more energetic reserves, making them more resilient to

environmental pressures, as they will be able to buffer poor years in terms of resources (!!).

Additionally, individuals with greater energy reserves have a better capacity to reproduce and are often more attractive to potential mates, leading to higher reproductive success. Body mass is considered to be an “honest signal” of an individual capacity to accumulate resources and its quality as a reproducer (!!). Therefore, body mass can be considered a key LHT influencing directly both survival and reproduction in many species.

However, as said before, energy quantity are limited and cannot be invested only in body mass. Moreover, a too large body mass can become a handicap (Jebb et al. 2021). Even though some handicaps have been theorized to be an asset in sexual selection (Zahavi 1997), beyond a certain threshold, too large individuals will be counter-selected. Hence, as predicted in Life History Theory, trade-offs must be made between available energy and individuals’ performances to find the optimal trait value, here, body mass.

Link with hibernation

The role of body mass as a LHT is particularly important for a specific group of species: **Hibernators**. To survive unfavourable seasons, some species disperse to milder environments, others cope with it and find ways to survive through (!!). A strategy adopted by some species to survive through is to enter a specific state of torpor, called “**hibernation**”.

This strategy consists of reducing the metabolism to a minimum survival level. More specifically, “Fat-storing” hibernators (in opposition to “Food-storing hibernators”) will sustain on their energetic resources, stocked in their white adipose tissue, through the harsh season, then emerge at the start of the favourable season (Carey et al. 2003; Geiser 2013; Nedergaard and Cannon 1990).

Furthermore, this energy gathering must be done in a short amount of time, as hibernators are active only for a small part of the year, during which they must forage as efficiently as possible to gain enough mass to survive through next hibernating season. Some fat-storing hibernators nearly double their weight during a 4 months active season (Armitage 2014; Carey et al. 2003).

Hence, not only a prerequisite adaptation is an important body mass, but also a sufficiently efficient metabolism to gain weight quickly. These prerequisite represent a lot of challenges and specific adaptation. Body mass and metabolism are therefore highly constrained in hibernating species (!!).

Additionally, for some species, reproduction occurs right at the onset of the active season. This means that before entering into hibernation, individuals must reach a body mass sufficient not only to survive, but also to have to reproduce directly after. Body mass is therefore a keystone LHT for hibernating species' bio-demography.

Meanwhile, Climate change is expected to have significant impact on such LHTs. It has been theorized that changes in body size could be a third universal response to climate change, alongside modification in phenology and geographic range (Daufresne et al. 2009; Durant et al. 2007; Gardner et al. 2011; Visser and Both 2005).

Expected effect of global warming on body mass

As said earlier, a key aspect of climate change is a rise in global average temperatures — hence the expression *global warming* (a term often invoked by climate skeptics during unusually cold spells¹).

This overall temperature increase is suspected to influence phenotypic traits such as body mass and size, though the precise direction of these changes remains uncertain. Some authors argue that a shrinking body size might be a universal response to climate change (Daufresne et al. 2009). This hypothesis is based on Bergmann's rules, which states that smaller body size are favoured in warmer environment as a higher surface-to-volume ratio facilitates heat dissipation (Bergmann 1847). In other words, in warmer environments, an expected adaptive response would be a shrinking body size.

However, as noted by Gardner et al. (2011), a lack of large-scale comparative studies prevents us from confirming that this response is universal. In addition to that, in 2022, the IPCC's report stated that "evidence is weak for a consistent reduction in body size across taxonomic groups in terrestrial animals" (Intergovernmental Panel On Climate Change (Ipcc) 2022; Siepielski et al. 2019).

On the other hand, several studies at higher latitudes and altitudes yield opposite results (i.e., increasing body mass in response to climate change, Guillemain et al. 2010; Ozgul et al. 2010; Sheridan and Bickford 2011; Yom-Tov et al. 2008). In these regions, climate change is a synonym of milder conditions. Hence, individuals have access to a large food supply for a longer time and the severity of the harsh season is reduced, which overall is less energetically demanding. Ultimately, these new conditions enable individuals to grow larger.

¹"[...] Large parts of the Country are suffering from tremendous amounts of snow and near record setting cold. [...] Wouldn't be bad to have a little of that good old fashioned Global Warming right now!" U.S. President, Jan 20, 2019.

Phenotypic plasticity vs microevolution

Evolution by natural selection was defined by Darwin (1859) as so: individuals that are best adapted to their environment will have better survival and reproductive success. This process results in a population composed mainly of individuals adapted to their environment. So, observed phenotype in natural populations is expected to be the best fit for a specific environment.

However, when this environment changes, as expected with climate change, individuals have two solutions to avoid disappearance: **disperse** to a more favorable environment, or **adapt** to their new conditions through phenotypic changes (Gienapp and Brommer 2014).

To adapt, there are two further possibilities: **phenotypic plasticity**, defined as a change in phenotype expressed by a given genotype, which allows for a rapid response within an individual lifetime, is highly flexible and does not involve any changes at the genetic level (!!); and **microevolution**, defined as a change in allele frequencies in a population over time (!!). When an individual with a better-fitted phenotype for its new environment appears, it would have a better survival and more reproductive success. If this advantage relies on a heritable genetic difference (i.e., transmitted to its descendants, Lynch and Walsh (1998)) the new genotype is going to rapidly increase in proportion in the population, ultimately replacing the old one.

Hence, adaptation through microevolution can be slow but is a long-term solution, better suited for a persistent ecological change. However if the change is transient, plasticity might be better suited. As noted by DeWitt et al. (1998) and Gardner et al. (2011), phenotypic plasticity solely is unlikely to be the most optimal long-term response to climate change as it is usually a transient answer to a temporary change, presenting costs and limits (DeWitt et al. 1998). Furthermore, if the optimal response to the new environment is a canalized phenotype (i.e., very low phenotypic variance), plasticity can even be maladaptive (Nussey et al. 2007). Therefore, the expected optimal answer to a long-term environmental change, as those expected with climate change, is evolution through natural selection.

Phenotypic plasticity and microevolution are not expected to be mutually exclusive. This is particularly evident in highly plastic traits such as body mass which can vary significantly up and down throughout an individual's life in response to among- and within-year changes in environmental conditions but can also change via microevolution at the population level over the same time period.

Nevertheless, as the consequences of these mechanism can be highly different on the long term (evolution being more permanent than plasticity), quantifying the extent to which each of these mechanisms contributes to the observed change over a long study period remains a challenging but fundamental task to understand the adaption and evolution of species.

Quantitative Genetics and Animal Models

As we have discussed, climate change is expected to cause long-term environmental changes. Thus, natural population will need to adapt to these new conditions in order to persist (!!). As mentioned earlier these adaptations can involve shift in geographical ranges, phenology, or phenotype (Dufresne et al. 2009; Durant et al. 2007; Gardner et al. 2011; Visser and Both 2005). As climate change imply deep and long-term environmental modification, we expect natural population to **evolve**, allowing them to be better suited to new environments (Darwin 1859; Sih et al. 2011).

However, estimating evolutionary signal in natural conditions can be complicated, and an observed phenotypic change is not necessarily due to evolution (i.e., phenotypic plasticity). Fortunately, quantitative genetic provides us a robust and well-established method to address this by decomposing the total phenotypic variance (V_P) into its genetic (V_A) and environmental (V_{PE}) components: $V_P = V_A + V_{PE}$ (Lynch and Walsh 1998; Wilson et al. 2010).

A well-known statistical method to estimate these variance components from observed phenotypic data is the **Animal Models** (Kruuk 2004). This method allows a robust estimation of the genetic variance in a trait affected by a large number of genes, each with small effects (Kruuk et al. 2014; i.e., a “quantitative trait,” Lynch and Walsh 1998). An Animal model is a specific kind of mixed model using individual identity as a non-independant random effect, linked to a relatedness matrix between each individual, extracted from the population pedigree (i.e., parental links between each individuals in the population population, Lynch and Walsh (1998)).

This method has the advantage of being relatively simple to employ, enabling genetic parameters estimation directly from phenotypic data. Only parental links between individuals need to be known, making this method applicable to wild populations (Kruuk 2004; Lynch and Walsh 1998).

As emphasized by Kruuk et al. (2014), there is a pressing need for quantitative genetics studies on long-term wildlife populations, as the most common problem in such studies is the lack of statistical power, which can be resolved thanks to the quantity of data brought by

long-term studies. Such studies would improve our understanding of the relationship between animals and their environment, as well as the genotype-phenotype-environment relationship, especially in a context of global change.

Traits coevolution

A well-recognised challenge when studying evolution in natural context, is to consider the genetic correlation between several traits, causing **traits coevolution** (Gould and Lewontin 1979; Roff 1992). Indeed a trait cannot evolve independently without impacting other traits. This constraint narrows the range of possibility and reachable outcomes in the adaptive landscape (Arnold et al. 2001; Gould and Lewontin 1979; Teplitsky et al. 2014).

When studying the evolution of a specific trait (especially ones having important phenotypic consequences), failing to account for its link with other traits is an oversimplification. This misconception can bias not only our understanding of the causes and consequences of phenotypic change, but also the estimations of its evolutionary potential [Teplitsky et al. (2014); !!!]. To effectively study traits' evolution, it is essential to consider that selection generally acts on multiple traits simultaneously, as phenotype is the result of a combination of various traits (Phillips and Arnold 1989).

A continuation of this reasoning is the extended **Pace Of Life Syndrome** (POLS) suggesting that life-history strategy, physiological and behavioral traits coevolve in response to the environment (Dammhahn et al. 2018; Réale et al. 2010).

So, if climate change induce LHT modifications, it is to expect that other key physiological or behavioural trait will coevolve with it. For example, it has been observed that certain behavioural types will be consistently associated with certain Life-History strategies (Biro and Stamps 2008; e.g., individual with faster life-cycle will tend to be bolder Stamps 2007; Wolf et al. 2007).

As such, changes in LHTs are expected to have even greater impacts on the global phenotype. A robust method to study such changes are **Multivariate Animal models** (i.e., an Animal model with multiple phenotypic traits as dependant variables) which allows us to estimate the genetic covariance between each trait [!!!].

However such models are heavily data-consuming, and the main reason that significant results with such methods are quite rare today is that only a few studies have enough data to support the statistical power required for these complex models. Again, using long-term datasets brings a lot of expectation to address such limitations (Teplitsky et al. 2014).

Individual variation in their plasticity

As explained above, evolution and plasticity are not mutually exclusive, but even more so, plasticity itself can evolve (!!).

Plasticity is usually studied using **Reaction Norm framework** (!!), i.e., studying the value of a phenotypic trait (e.g., body mass) in response to an environmental proxy (e.g., temperature, precipitation). A plastic response correspond to a different phenotypic value associated to a different environment. In this reasoning, a trait is plastic if the slope of the reaction norm is different from 0 (Nussey et al. 2007).

A reaction norm has two parameters: **Elevation**, which is the expected phenotypic value in the average environment; and the **Slope** corresponding to the linear regression of the phenotype over the environmental gradient. In statistical terms, these parameters correspond respectively to the “Intercept” and the “Slope” of the linear regression of the phenotype over the environment.

This is how phenotypic plasticity is usually studied. however, as emphasized by Nussey et al. (2007), a lot of information is loss when plasticity is only studied at the population level (i.e., fixed linear model of the phenotype over the environment). First using a linear mixed model is crucial to estimate different elevations for each individual, this allows the model to account for the individual variability in their phenotype in the average environment. However, that kind of model doesn't allow to estimate individual-level difference in their plastic response to the environment, which can be biologically significant and must be accounted for.

Nussey et al. (2007) proposed a framework to estimate that individual variance in plasticity, which is now fairly accepted and used: **Random Regression**. Let's go back to the Linear Mixed Models (LMMs), the most classical type of mixed models is to add a random individual term on the intercept to estimate the “individual variance” by allowing the model to fit one intercept per individual, however in this compilation, the slope is the same for each individuals; now, if we add a random term also on the slope, not only the model estimate one intercept per individual, but also one slope (so, one reaction norm) per individual, estimating thus the individual variance in the phenotype expected in the average environment (*elevation*) but also the individual variation in phenotypic variation (*slope*) which is commonly called **I * E**.

Once we've estimated this between-individual variance in plasticity, we can use quantitative genetic methods using Random Regression Animal Model: “RRAM” (Nussey et al. 2007) to

decompose this variance into it's genetic and environmental parts in order to get the between-individual genetic variation in plasticity, commonly referred to as $\mathbf{G} * \mathbf{E}$. From that we can estimate the heritable variation in reaction norm's slope and have an idea of the evolutionary potential of a trait's phenotypic plasticity itself.

From a theoretical point of view, this method allows for proper estimations of a trait's $I * E$ and $G * E$. However, as underlined by Ramakers et al. (2023), an important limitation is the environmental proxy used for the reaction norm. Indeed, if this latter isn't good enough (i.e., too far from the real predictor of the trait's plastic response), an important part of the actual individual variation in their plasticity is missed, this is what Ramakers et al. (2023) called the "hidden $I * E$ ".

Furthermore, it is often impossible to identify the real driver of plasticity in natural conditions, as natural environments are exceptionally complex systems, and individuals generally have to react to a combination of environmental variables rather than one. Therefore, the real driver of plasticity is often unknown, unmeasurable and a combination of a lot of different variables. In response to that, Ramakers et al. (2023) proposed a new method, using Environment Specific Mean phenotype (ESM). Although they have shown that this method is indeed efficient, they emphasized that we still need a really good knowledge of the study system, and specific conditions to be more efficient than classic the random regression method. Hence, although it's a progress, more work is needed to deal with this $I * E$ detection problem in natural environments.

The good news is that a promising, fairly new, statistical method could bring new interesting insights to that matter: the "Double Hierarchical Generalized Linear Model" (DHGLM). DHGLM is a kind of mixed model estimating the distribution for both the mean and the residual variance (Lee and Nelder 2006). A DHGLM estimates the residual variance for each individual, then estimates the variance in this residual variance. In other words, DHGLM estimates the *variance in the residual variance* hereafter referred to as V_{V_e} . This corresponds to the variance in the proportion of residual variance attributable to each individual.

We are confident that DHGLM can be applied to develop a more powerfull method to detect individual variation in plasticity ($I * E$) in natural systems. Indeed, a V_{V_e} significantly greater than 0 indicate some disparities in the individual contributions to the residual variance. To put it differently, it shows that the range of phenotypic values expressed by each individuals is different, which can be interpreted as an individual variance in their plastic response, so $I * E$.

Although a non-zero variance isn't necessarily due to the presence of $I * E$ in the system, it would justify further investigations by testing different environmental proxies. In contrast, except in very specific conditions, almost impossible in natural conditions (i.e., **perfect** contrary effect), a V_{V_e} equal to 0 indicate that the relative contribution to the residual variance in phenotype isn't significantly different between each individual, and so that there's no individual variation in plasticity in the studied system.

An important feature of this method is that it should be free from the problems of bad environmental proxies as it detects $I * E$ from the structure of the residual variance in phenotype. And obviously, as for the random regression with the RRAM, we can use quantitative genetics to also detect $G * E$ if $I * E$ is detected by combining a DHGLM with an Animal model, giving a “Double Hierarchical Animal Model” (DHAM).

This is a very promising method, however, as most new statistical methods, it is often poorly understood/used. Hence, a general framework, including clear applications showing how DHGLM should be used to detect $I * E$ would be really helpful for the community, especially with the rising need to study efficiently phenotypic plasticity and its evolution, in natural condition in a context of climate change.

Species and study

Since 1962, a wild population of Yellow-Bellied Marmots (*Marmota flaviventris*) is monitored continuously, initiated by Kenneth Armitage, and now supervised by Julien Martin and Daniel Blumstein, at the Rocky Mountain Biological Laboratory (RMBL) in Gothic, Colorado, USA ($38^{\circ}56'34'' - 38^{\circ}59'13''$ N / $106^{\circ}58'60'' - 107^{\circ}0'45''$ W). This study is the second longest-term wild mammal population monitoring in the world.

Yellow-bellied marmots is a ground-dwelling sciurid (Rodentia, Sciuridae) inhabiting alpine habitats in western North America. Their life cycle is divided between an “active season” representing approximately a third of the year (from May to September) where individuals must forage to reach a threshold body mass in order to survive hibernation for the remainder of the year (September to May) (Armitage 2014).

Marmots experiences high seasonal fluctuation in body mass, with a critical threshold to be reached before the onset of hibernation in order to first, survive through the next active season and second, have enough energy left for reproduction (which occurs in the first weeks of the active season, Armitage 1965, 2014). Hence, body mass is a critical LHT for the marmots.

Yellow-bellied marmots lives in colonies usually composed by one or more matrilineal with one adult male, multiple adult females and their offspring (Armitage 2014). Our population is composed of 7 main and 31 smaller colonies divided between an “up” and a “down valley” (Figure 1) with an elevation difference around 300 m (“up” = 3,000 m; “down” = 2,700 m).

The altitude differences between the two valleys implies some differences in weather like delayed snowmelt and vegetation growth onset, or temperature differences up to 2 °C (Armitage 2014; Blumstein et al. 2004). In consequence of these weather differences, a delayed emergence up to two weeks in the up-valley can be observed (Blumstein 2009; Monclús et al. 2014). These two different condition offers an amazing opportunity to test the impact of environmental differences on several factors while working in natural conditions.

The marmots’ hibernation life-cycle is highly environment-dependent. Indeed, their phenology (i.e., onset and end of the active season) seems to be mediated mostly by weather variable such as temperature and snow cover (Armitage 2014), as for the rest of high-altitude ecosystems (Inouye and Wielgolaski 2003).

Body mass increase in Yellow-Bellied Marmots

An important body mass increase has been observed in this population over the past half-century, estimated around 600 g, representing almost 20% of total individuals’ body mass (Birot, Blumstein & Martin, Manuscript in progress, Figure 2a). Precedent studies concluded that most of this major change was due to phenotypic plasticity (Ozgul et al. 2010). This hypothesis is indeed relevant, with climate change, active season is getting longer and population faces milder conditions (e.g., higher temperature, less snow). Hence, marmots has more time to forage and gain weight, and the hibernation period is getting shorter, meaning less time for individuals to lose mass. With these new conditions, individuals are automatically getting heavier, makes sense!

However, with now almost 15 additional years of data, we reanalyzed the body using animal models to properly assess the genetic part of this change and estimate explicitly the trait’s evolutionary signal of 199 adult females between 1965 and 2022 (657 observations). Our results show an increase, at the genetic scale, estimated around 400 g, with a heritability of 56%, meaning that around two third of the observed body mass increased is in fact genetic (Birot Blumstein & Martin, Manuscript in progress, Figure 2b). With these results, it is reasonable to conclude that plasticity is not the only process causing this phenotypic shift, but that evolution also plays a crucial role here.

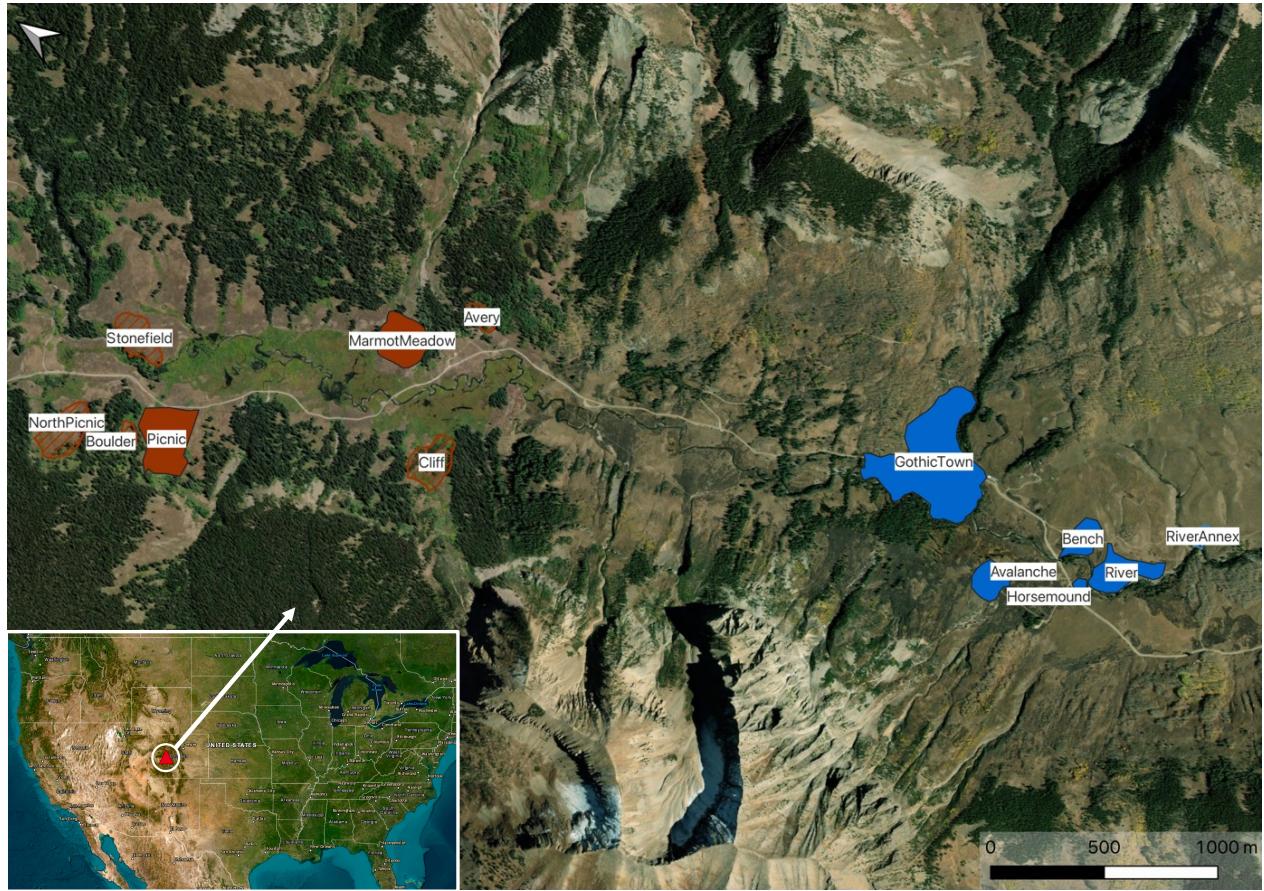


Figure 1: Red colonies represent the “up” valley, blue ones represent the “down” valleys. Plain background polygons represent the seven main colonies. The map was created with QGIS software (QGIS Development Team 2024) and the base map comes from ESRI (“GIS Mapping Software, Location Intelligence & Spatial Analytics | Esri” n.d.).

Furthermore, although the lengthening active season is indeed a good potential explanation for the body mass increase through phenotypic plasticity, it doesn't match with the observed evolutionary pattern. Indeed, the observed increasingly milder conditions in parallel to this change in body mass should decrease selection on body mass, as it lower the survival pressure through hibernation. The observed body mass increase here should be expected with an increasing pressure over winter survival. It is clear that the lengthening active season and global milder condition are not the only drivers of body mass changes, since we observed both plasticity and microevolution.

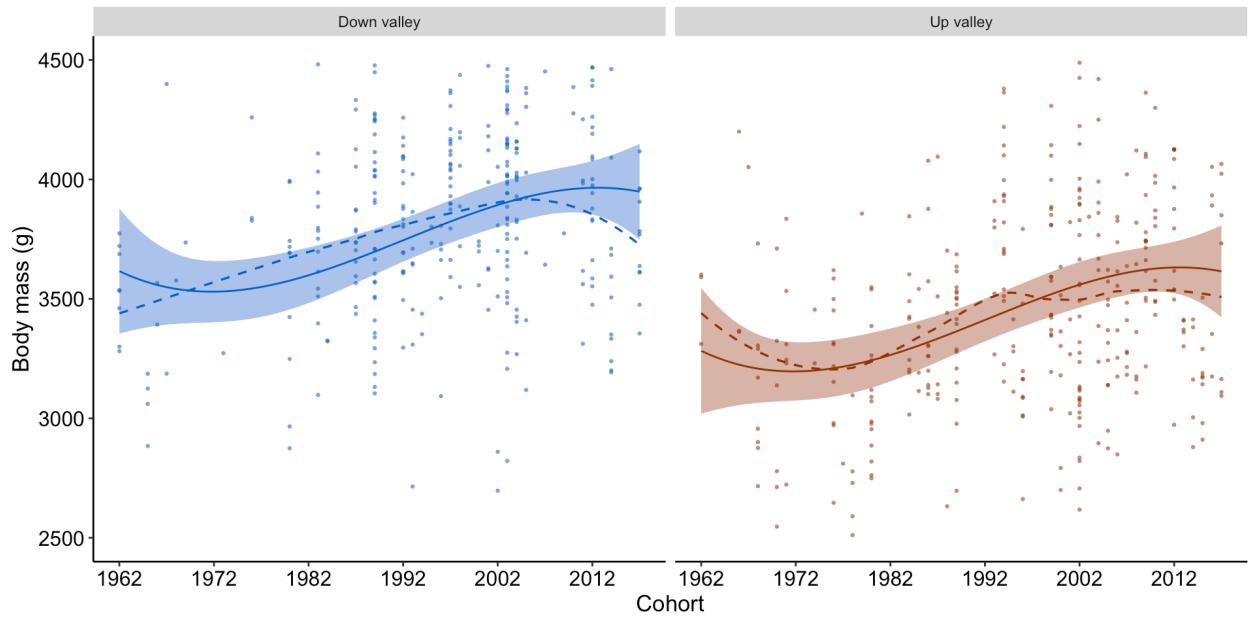
As explained earlier, body mass is a keystone phenotypic trait for the yellow-bellied marmots. It is therefore crucial to understand how this trait and this population is responding to climate change, both for conservation purposes and better comprehension of the links between phenotype and environment. There is a pressing need to explore which environmental factors may have triggered this shift, the mechanisms behind this increase, and the potential implications for the population's future to better understand how can natural population cope with climate change.

Data

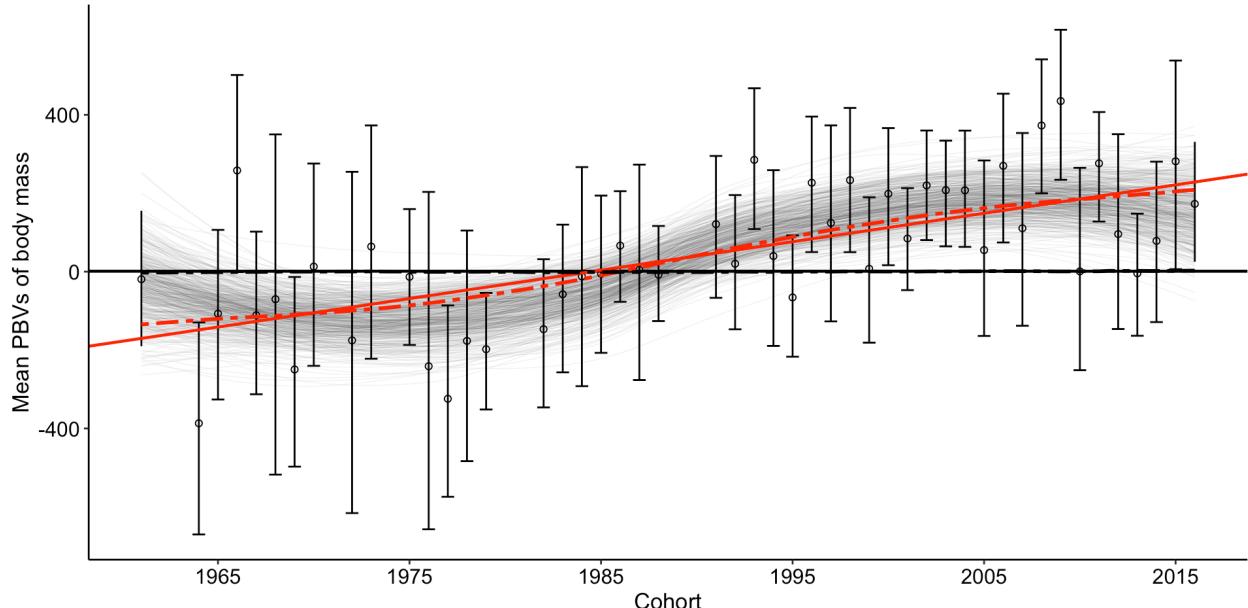
Each Spring-Summer (May - September) since 1962, marmots are regularly trapped (between 1 and 20 times per individual, with an average of 4.5) using baited Towahawk live traps ($81 * 25 * 30\text{cm}$) situated near burrow entrances. If the individual is captured for the first time, it is identified by placing a unique numbered ear tag on both ears, and with a nontoxic black Nyzanol dye fur mark for distant identification during behavioral observations. Over 95% of individuals are captured between the first two years of their lives and thus have known year of birth and age, from that each we define four age classes, the **juveniles** between 0 and 1 year old, **yearlings** between 1 and 2 years old, **subadults** between 2 and 3 years old, and **adult** over 3 years old (Jebb et al. 2021).

Pedigree (i.e., parental links between individuals) is known for most individuals in the population (4,652 individuals to this date). This was first estimated through behavioral observations, meaning that only the mother-daughter links could be determined, then pedigree was estimated with genetic methods (details in Blumstein et al. 2010; Olson et al. 2012).

Behavioral observations and experiments (running speed and Flight Initiation Distance (FID) (!!!)) are conducted all along the season. Upon each capture, individuals are sexed, weighted (initially with a spring scale ($\pm 50\text{g}$) and now with a digital balance ($\pm 10\text{g}$)) measured and DNA samples are taken. More descriptions can be found in Armitage (2014).



(a)



(b)

Figure 2: Adult females' mean cohort's body mass. a) At phenotypic level, trend lines (\pm SE) represent LMM predictions and points shows raw data. b) At genetic level, median trends of the observations (red line) and under a null scenario (black line) are represented, according to linear models, points and error bars represent the median and 95% credibility interval of the posterior mean predicted breeding values for each cohort.

Body mass is a highly plastic trait, particularly for marmots as it experiences considerable fluctuations throughout the active season. It is, for obvious reasons, impossible to record all individual body mass at the same time. Therefore, it is necessary to estimate it for each individual at the same time of the year. Using repeated measures for each individual throughout each active season, a linear mixed model was fitted and its Best Linear Unbiased Predictors was used to extract each individual's body mass on August 15 each year (Jebb et al. 2021; details in Ozgul et al. 2010).

As our study site has been an important scientific station for more than a century, we have various and exhaustive data. By combining multiple sources, such as Billy Barr (a RMBL resident), the National Oceanic and Amospheric Administration (NOAA), the United States Geological Survey (USGS), the United States Department of Agriculture (USDA) and the Oregon State University's PRISM Climate group, Prather et al. (2023) provides us with exhaustive data. We have weather data (e.g., monthly temperatures, snowing, precipitations, season lengths) at our study site from 1975 to 2022.

Data are stored in the R package “ybamaRmot” (Martin 2024), analysis will be perofrmed on R (R Core Team 2024), Animal models will be performed using R package “MCMCglmm” (Hadfield 2010), DHGLM and other bayesian analysis (excluding MCMCglmm) will be done with “brms” (Bürkner 2021), figures will be made with “ggplot2” (Wickham 2016).

Research objective

As explained before, the observed major body mass increase (20%) in Yellow-Bellied marmots has been miss considered. We have shown a strong evolutionary signal in this change, meaning that we need to rethink the evolutionary scenario explaining this phenotypic change.

First, I will study which specific aspects of the marmots body mass has changed. We need to understand if individuals are born with a greater constitution (so the intercept in statistical term), if their growing capacity has increased (slope) or both.

Even though our results indicate a strong genetic variation in body mass, it doesn't explain the entire phenotypic change. Phenotypic plasticity also plays a role here, and to fully understand the population's reaction to climate change, there is a pressing need to develop new methods to detect and study $I * E$. Therefore I am going to develop a framework to use DHGLMs to detect $I * E$ in natural conditions.

Then, using our extensive datasets on the weather and environment of the marmots, I am going to study the changes during the last half-century in order to understand which factors could have triggered this body mass increase.

Finally, we expect that such modification in a crucial LHT in natural population will not be the only major phenotypic change. Therefore, as theorized by the extended POLS, I am going to investigate potential behavioral changes linked to body mass in marmots.

Chapter 1 - Mechanisms behind the body mass increase.

To investigate the body mass increase in the last-half century, the first step is to understand what has exactly changed. Understanding the mechanism of this increase means understand if cohorts have now a bigger constitution than before (i.e., born bigger), or does the growing capacity of each cohorts has increased? To explain that in statistical term, we have to understand if the intercept (constitution), the slope (growing) or both has changed over time cohorts.

Body mass in juvenile was predicted to stay stable as it favour a higher running speed allowing juvenile to escape predators more efficiently during foraging. That prediction was the main explanation for a stable body mass in adults marmots even if it was linked to a better fitness. This stable body mass in Juvenile, holding its evolution in the subsequent age classes, is called the “invisible fraction” and is explained by Jebb et al. (2021).

However, looking at body mass in Juvenile over time cohort for both males and females reveals interesting details (Figure 3). We see that though the mass at birth is indeed relatively stable (or even slightly decreasing), the mass at the end of the individuals first active season shows a similar pattern found in adult females (i.e., cubic effect, Figure 2), although the decrease at the end of the period seems much more pronounced here. Indeed, the body mass at the end of our juveniles’ first foraging season has increased *from 1,206.18 g in 1976 (Males) and 1,130.64 g in 1979 (Females) to 1,519.71 g in 1998 (Males) and 1,363.03 g in 2001 (Females, Estimations from local regression on raw data)*. These changes represent a body mass increase of 23% in 22 years (22 cohorts), meaning that between the mid 1970s and the 2000s, each cohort was in average 1% heavier than the last one at the end of their first foraging season (Figure 3).

The relax of the evolutionary constraint on body mass in Juvenile could be a serious clue to explain the sudden increase observed in adults arround the same period.

First, I will study the body mass patternst through time cohorts at the beginning and end of the active season for each age classes at both phenotypic and genetic level to understand the changes in the last half century. Then, I will investigate the reaction norms (at cohort scale) by looking both the average cohorts mass at these two moments, and slopes over the active season.

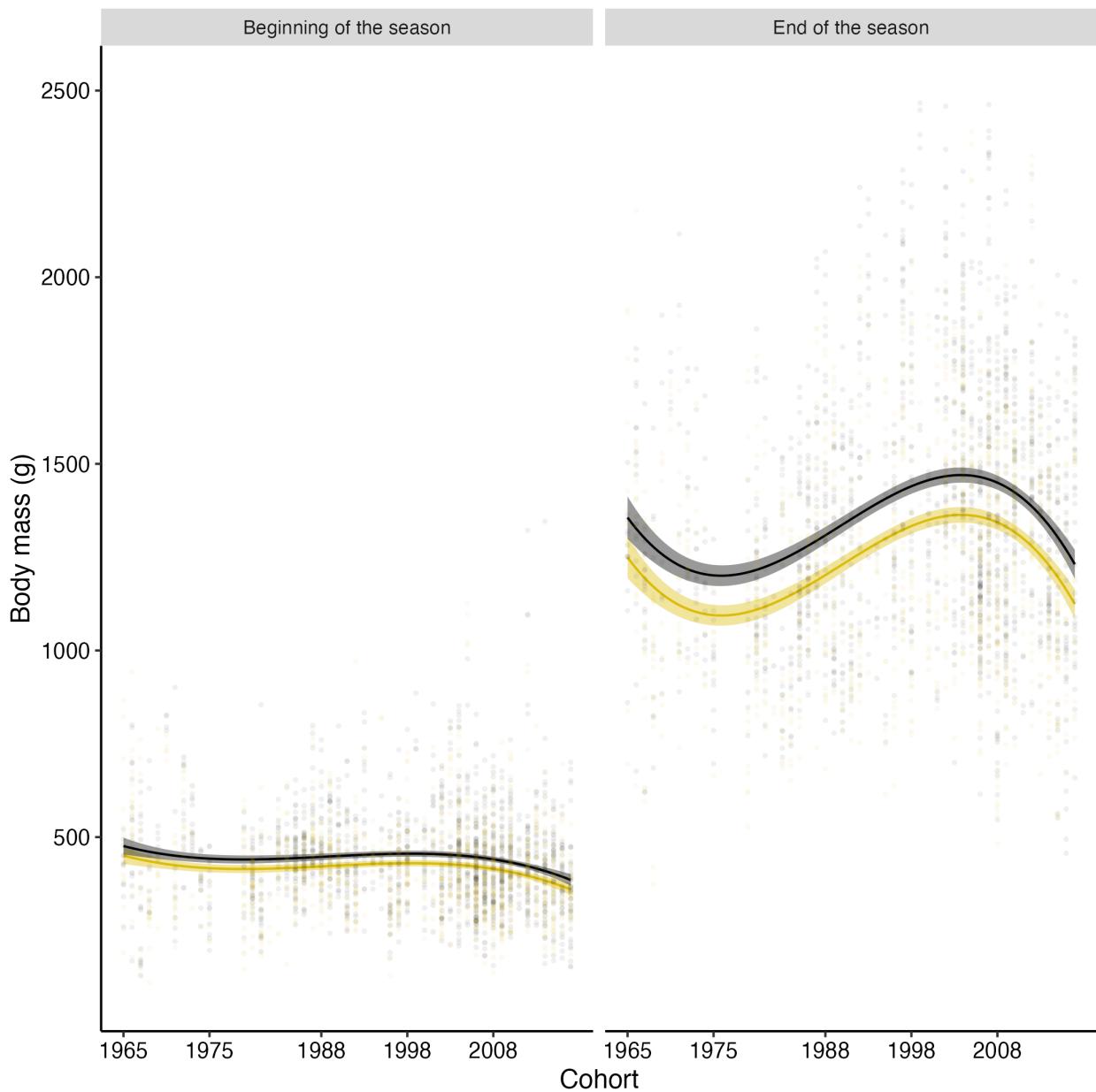


Figure 3: Body mass trend over time cohort for females (yellow) and males (black) Juveniles compared between the beginning (birth weight) and end of their first active season (mass on August 15th).

Chapter 2 - Detecting individual variation in plasticity with DHGLMs.

As explained previously, detecting more efficiently individual variation in plasticity ($I * E$) in natural condition is essential to better understand individual responses to climate change. However, although some good methods exists today (Nussey et al. 2007; Ramakers et al. 2023), a lot of biases coming from environmental proxies still limit these methods today. As said in the introduction again, we believe that using Double Hierarchical Generalized Models (DHGLM) can drastically reduce these problems and help us improve the efficiency of $I * E$ detection in natural populations.

DHGLMs will estimate the residual variance for each individuals, then the variance in this residual variance: V_{V_e} . When fitting a DHGLM on a focal phenotypic trait with multiple values for each individuals, if there is no $I * E$ (i.e., each individual's phenotypic response will be the same, Figure 4 a), there will be no variance in the residual variance as each individual will exhibit the same range of phenotypic values (Figure 4 b), so $V_{V_e} = 0$.

However, if there's individual variation in their plastic response ($I * E$) for the focal phenotypic trait (Figure 4 c), there, the residual variance won't be the same for each individual, hence there will be a V_{V_e} different from 0 (Figure 4 d).

Therefore, although a V_{V_e} significantly different from 0 isn't a proof of $I * E$ in itself (as it could come from something else), such result show that investigation on this focal trait are worth doing. On the contrary, unless in very specific case, almost never met in natural conditions (i.e., exact opposite effect, Figure 5 c, d), leading to false negatives, a $V_{V_e} = 0$ most likely indicate that there is no individual variation in their plastic response for the focal trait.

Finally, it is worth noting a potential limitation with this method that must be taken into account before performing such analysis to avoid biases. This method can only work with a balanced setup. By that understand individuals' phenotypic responses sampled on the same range of dependant variables vamues. Indeed, if the sampling range isn't the same for individuals, then the range of phenotypic values exhibited by them will be different, whether there is $I * E$ or not (Figure 6 c). In such a scenario, we would also have significant variance in the residual variance even if there is no actual effects (Figure 6 d). In other words, a false positive.

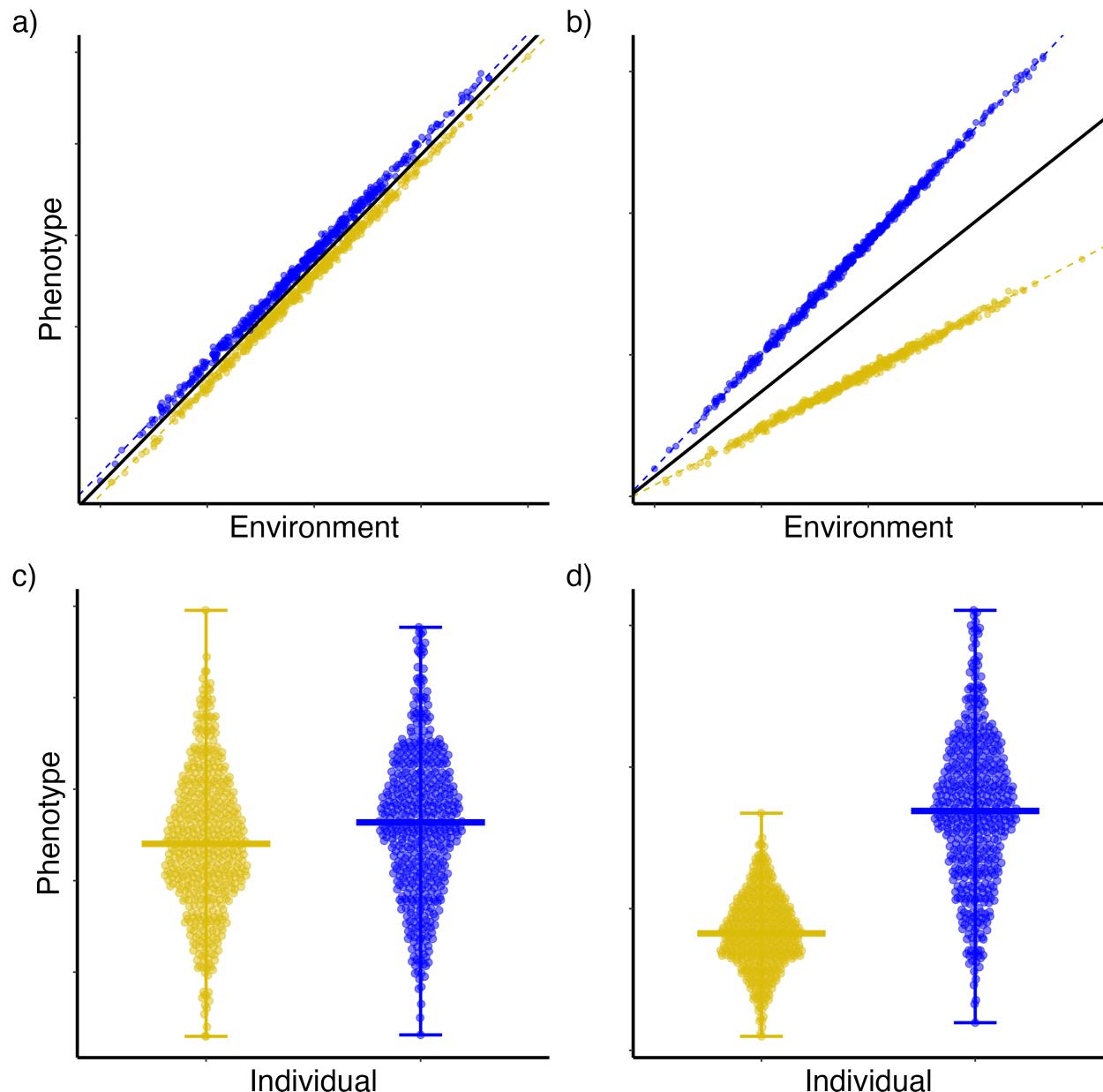


Figure 4: Reaction norms for two individuals (blue and yellow) for a trait a) without and b) with individual variations in plasticity. c) In the absence of I^*E , both individual express the same range of phenotypic values. d) However, with I^*E , individuals express different range of values.

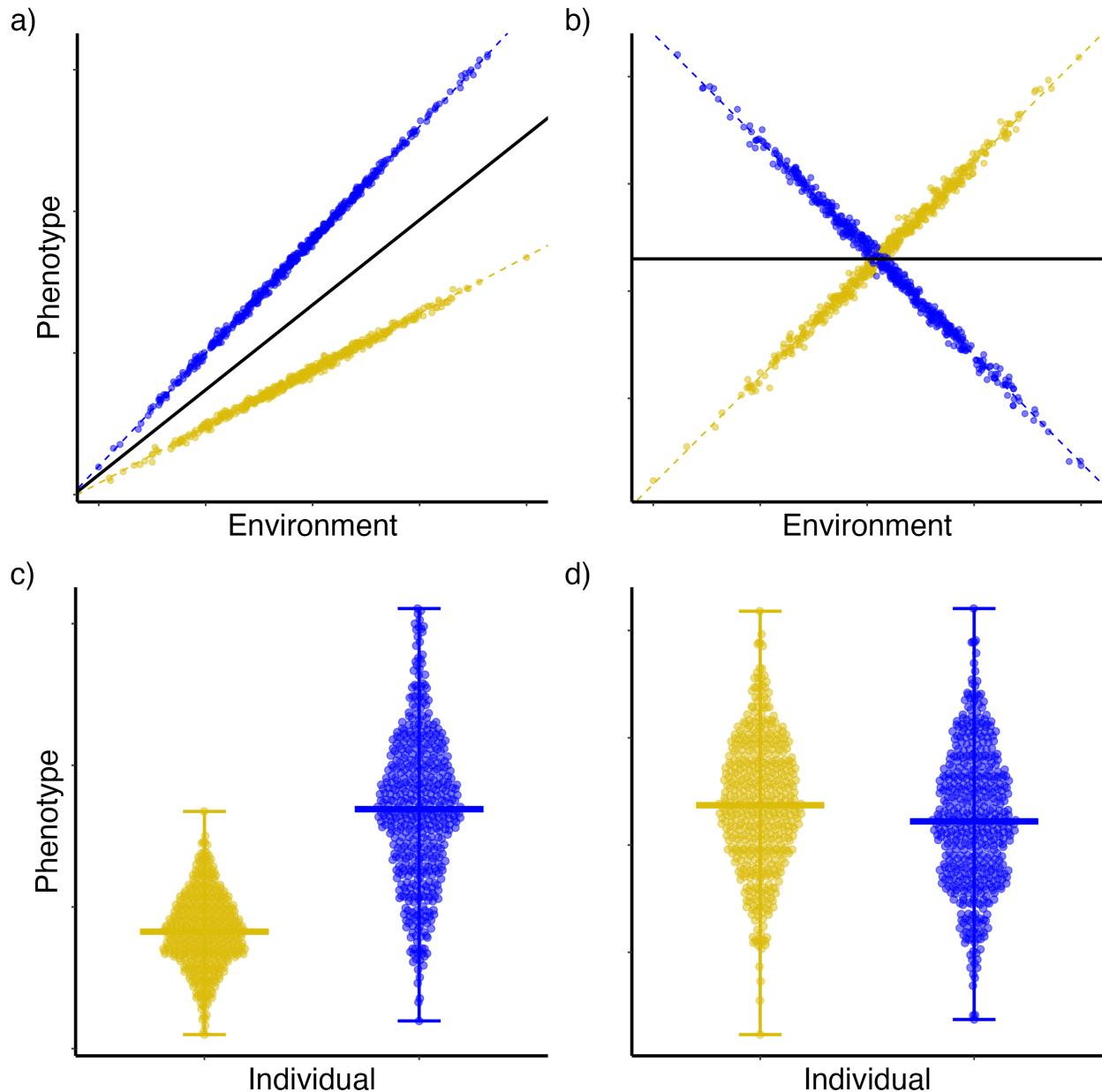


Figure 5: Reaction norms for two individuals (blue and yellow) for a trait a) with individual variation in their relationships and b) with the exact opposite effects. Here, although individuals in the first case c) express different ranges of phenotypic value, d) in the second case, they express the same range despite the $I * E$.

Based on that, I will simulate phenotypic and environmental values for populations with and without $I * E$, with balanced and unbalanced models. On these simulated populations, I will fit DHGLMs models, in a Bayesian framework using R package, brms (Bürkner 2021), using stan software (Carpenter et al. 2017). I will use the models V_{V_e} estimation to assess the detection capacity of this method. Then I will then apply this method on our Yellow-Bellied Marmots population to illustrate it with a real condition example.

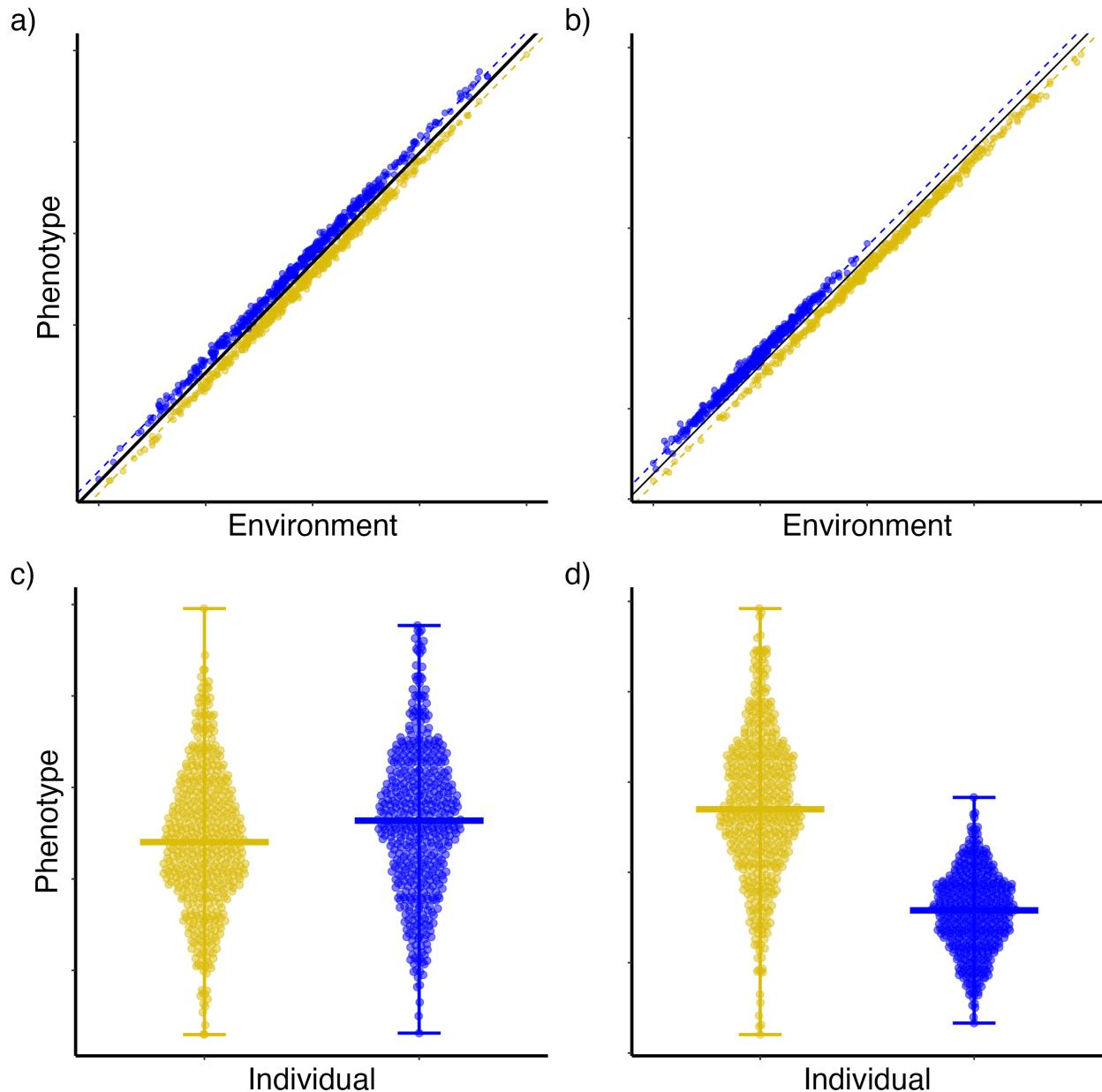


Figure 6: Reaction norms for two individuals (blue and yellow) for a trait without individual variation in their plasticity in a) balanced and an b) unbalanced design. c) In the balanced design there is indeed no individual difference in the range of expressed phenotypic values. d) But in the unbalanced design, although both individuals have the same slope (i.e., phenotypic response), there is a difference in the range of expressed phenotypic values.

Chapter 3 - Identifying predictors of increased body mass.

In the introduction, I explained the link between climate and body mass, especially in hibernating species such as marmots. I also discuss about the different hypotheses in the litterature about the links between climate change and body size. Knowing all that, the main hypothesis to explain the body mass increased in Yellow-Bellied Marmots in the last half-century is a change in climate over the years.

Using weathers data on our study site from 1975 to 2013, I performed some preliminary multivariate anaylyses (Principal Components Analysis, PCA with “ade4” package, Dray et al. 2023) from which I obtained a “seasonal gradient” (Figure 7) with high values corresponding to harsh years (i.e., long and snowy winter, lower temperatures) and lower values to milder years (i.e., long active season, warmer temperatures).

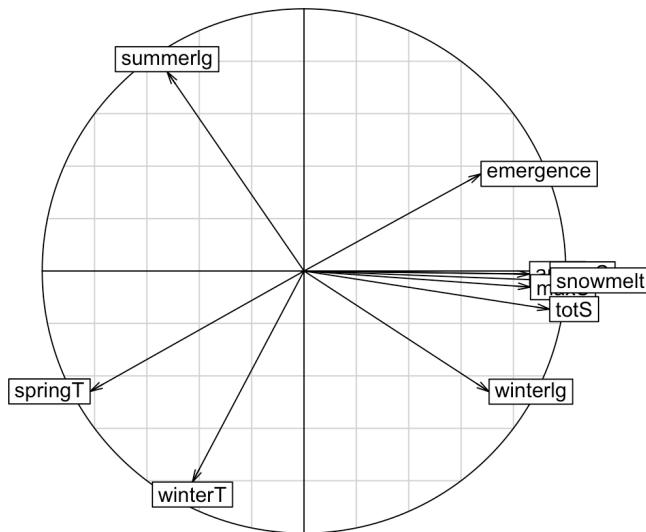


Figure 7: Correlation circle from the PCA of 10 different weather variables at RMBL between 1975 to 2013.

Looking at this seasonal gradient through time reveal interesting variation (Birot, Blumstein & Martin, Manuscript in progress, Figure 8). This, happening at the same period as body mass, justify further investigation. I will use more extensive and detailed weather data at the study site from 1975 to 2022, collected and treated by Prather et al. (2023). I will perform deeper multivariate analysis and test for different predictors for body mass changes within weathers variables.

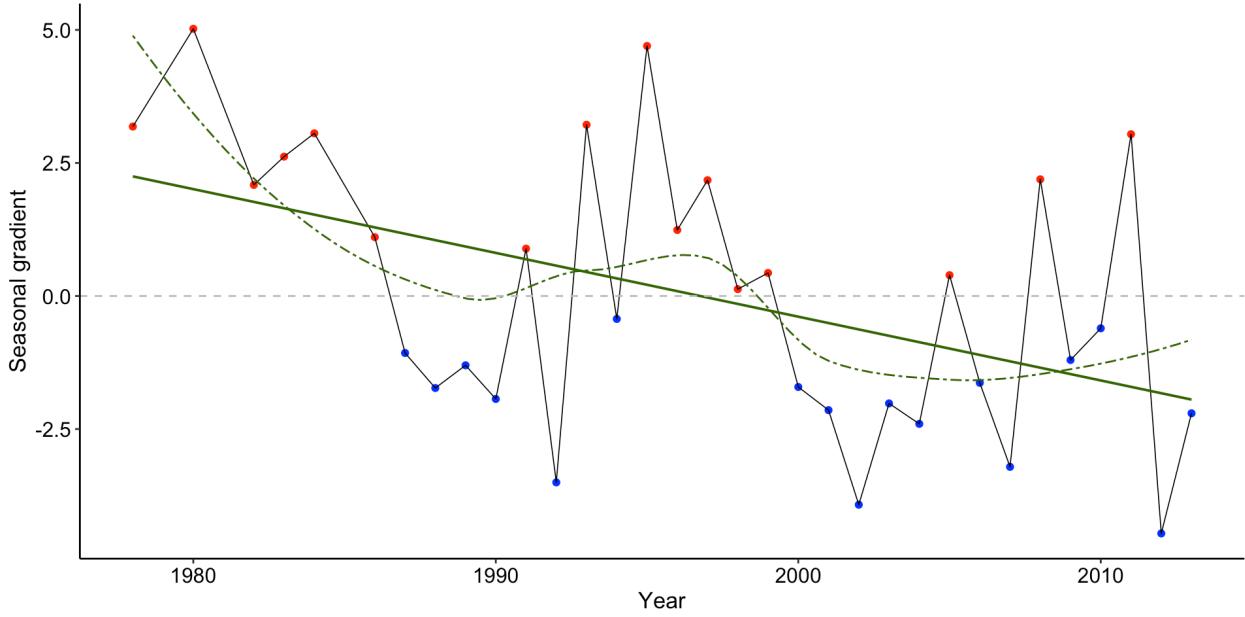


Figure 8: Seasonal gradient variation at RMBL from 1975 to 2013.

In a second time, I will test another hypothetic predictors. Body mass was previously hypothesized to be stable in juveniles as an anti-predator strategy (Jebb et al. 2021), however, it seems that juvenile body mass has also changed (Figure 3). Therefore, a potential explanation for the global body mass change is a change in the predation pressures, allowing juvenile body mass to increase and so impacting the other age classes.

Hence, using observation data at RMBL, I will look for changes in predators populations or behaviour between the 1970s and the 2000s and test for potential effect on the body mass of the different age classes.

Chapter 4 - Balance between body condition and experience as predictors of marmots' behavior.

Following extended Pace Of Life Syndrome framework (POLS, Dammhahn et al. 2018; Réale et al. 2010), we expect to see correlation between body mass and individuals' behavior. I will study that using different behavioral observation (i.e., two minutes focal observation with detailed ethogramms) and experiments (Flight Initiation distance) on our individuals. However, by looking at some preliminary results, although some results are statistically significant, the biological effect seems fairly small (Figure 9; Figure 10).

For example, the proportion of time "stand looking" is impacted by body mass dependant especially during the two first years of life, but this relationship is less and less important through age classes, to become almost flat in adults (Figure 10).

Although the impact of body mass, with bigger individuals spending more time looking is easily explainable as heavier individuals are slower and thus must stay more vigilant to avoid predation, this age effect is interesting to note and advocate for the hypothesis that with age, experience prevail on physical condition to dictate behavior, with older individuals more careful than juveniles, as already suggested by Jebb et al. (2021).

Using various and extensive behavioral data on our individuals, I am going to describe the impact of body mass at the beginning and end of the active season on behavior and its potential correlations with different behavioral types between age classes and sexes. The aim of this chapter will be to discriminate the effects of body condition (i.e., body mass) and experience (i.e., age class) as behavior predictors.

Understand the links between behavior and body mass is crucial to truly comprehend the impacts of such phenotypic shifts for the population's future.

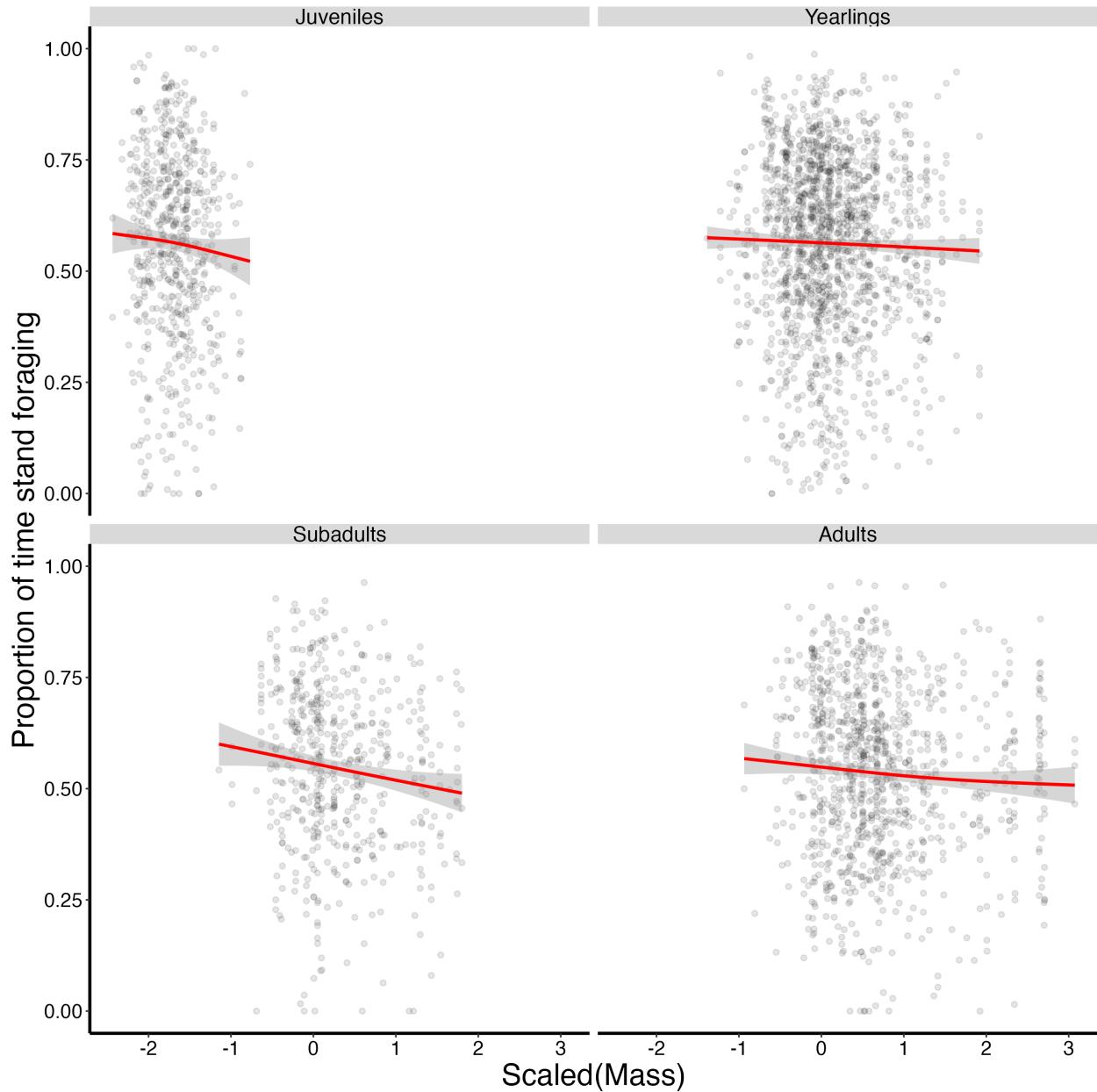


Figure 9: Proportion of time spent stand foraging (within two-minute focal observations) as a function of scaled mass on August 15th across age classes: Juveniles (0-1 year), Yearlings (1-2 years), Subadults (2-3 years), and Adults (3+ years). Red lines represent local regressions, and points represent raw data.

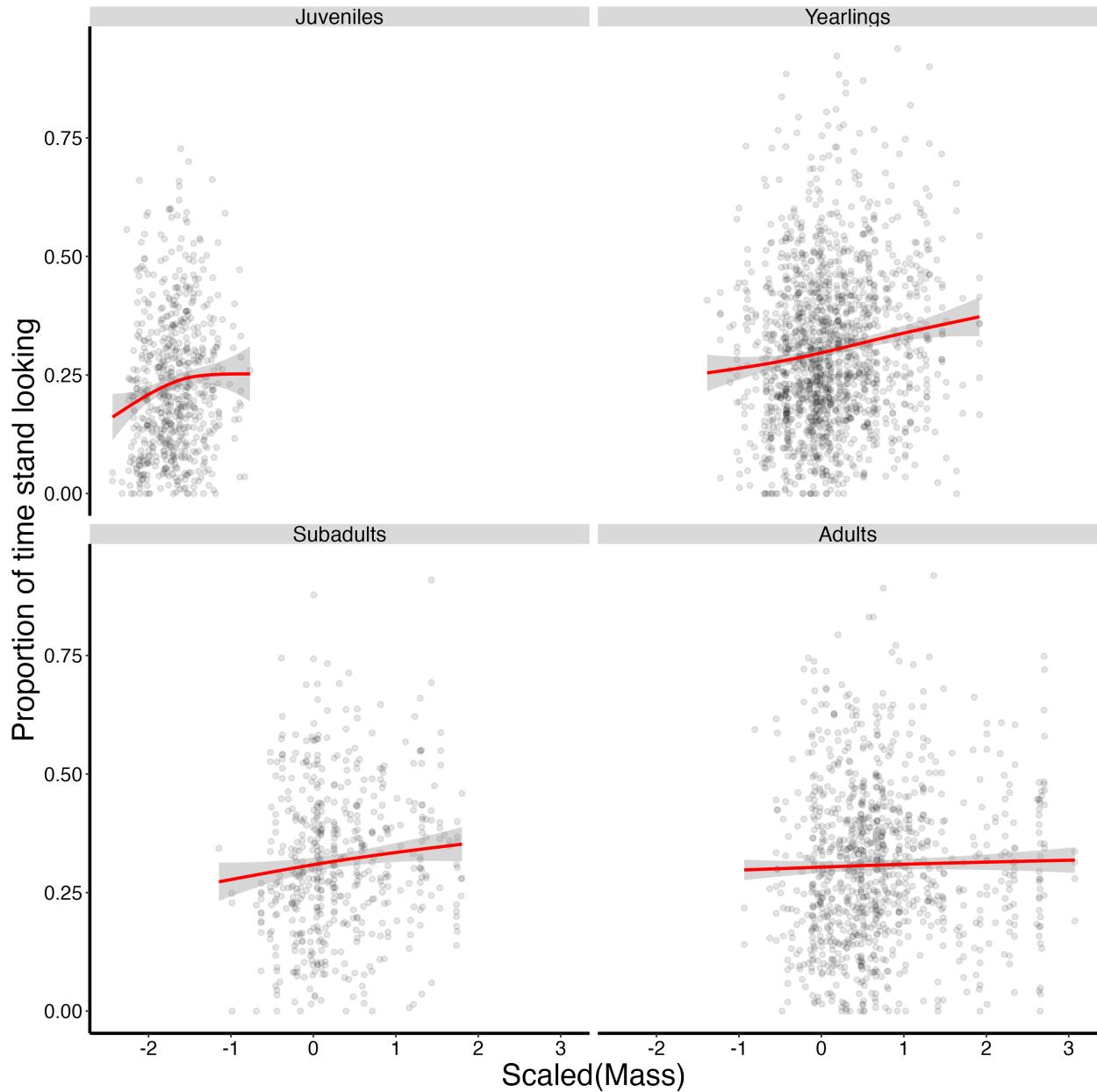


Figure 10: Proportion of time spent stand looking (within two-minute focal observations) as a function of scaled mass on August 15th across age classes: Juveniles (0-1 year), Yearlings (1-2 years), Subadults (2-3 years), and Adults (3+ years). Red lines represent local regressions, and points represent raw data.

Significance and impacts

This project provide a rare opportunity to study a remarkable response to climate change at both phenotypic and genetic levels. The observed body mass increase in Yellow-Bellied Marmots -representing a substential 20% rise in adult females- occured at a notable speed, seemingly, three decades, between the 1970s and the early 2000s.

Using one of the most extensive and detailed long-term datasets on a natural population, this research enables an unparalleled exploration of wild populations' response to climate change. This research allows for powerful analyses, typically unfeasible in natural systems.

By investigating how the body mass has changed (i.e., basis constitution or growing capacity), developping new methods to detect individual variation in plasticity, and identifying predictors of this shift, this research will contribute crucial knowledge about the complex interplay between genotype, phenotype and environment. Furthermore, exploring balance between body condition and life experience as predictors of individual behavior will improve our understanding of adaptive strategies in wild population.

By studying how natural population respond to rapidly changing environmental conditions, this project aims to give critical insights for conservations strategies and predictions. Understanding the causes and consequences of phenotypic response to climate change is essential for predicting future viability of natural populations.

We hope that fidings from this research will help inform conservation policymakers by improving projections of species' adaptive capacity and resilience to climate change, offering guidance for managing biodiversity in a rapidly changing world.

Potential side projects

1. Buffer environmental instability by increasing your body mass: an application of conservative bet-hedging in a hibernant rodent.

Bet-hedging strategy can be summarized as betting on the best fitness for the long term, even though it means lower a bit your immediate fitness, in order to cope with an unpredictable environment (Starrfelt and Kokko 2012). With climate change, environment are less predictable than ever, therefore, bet-hedging is expected to be a relevant adaptive strategy in such context. Let's take body mass increase as an example again.

On the one hand, an increase body mass, over a certain threshold, can become a handicap for certain species. For example, heavier individuals can be more subject to heat stress; they can be more subject to predation as heavier individuals are slower and more palatable.

On the other hand, an increased body mass, being synonyme of a more important energy stock, allow individuals to “buffer” in a poor year in term of resources. This feature can be very important for species living highly constrained environment, such as high altitude (Inouye and Wielgolaski 2003). Therefore, increase your body mass to “buffer” years with less resources can be expected, especially for hibernating species in alpine environment.

Furthermore, larger body mass can have lower temporal fitness variation as the fitness cost of a poor year would be buffer. Indeed, by increasing the global average body mass in the population, we can expect less phenotypic variation in the population due to environmental change, which should, all else being equal, favoured by selection (Cohen 1966). This correspond to a “Conservative bet-hedging strategy” (Childs et al. 2010).

However, reducing the phenotypic variance is again an important risk as it reduce the adaption capacity facing changing environment, just like domestication endangered a lot of species by lower their adaption capacity to new environments due to hyper specialisation (!!).

Indeed, specializing phenotype, in that case “betting” on the long-term fitness, there is a risk of a mismatch with the environment (Stenseth and Mysterud 2002; Visser and Both 2005). This is even more relevant in a context of highly variable and unpredictable environment such as expected with an extremely fast climate change. In such conditions, betting on the future can end in evolutionary traps, where maladapted populations can no longer cope with their environmental constraints and risk local extinctions (Robertson et al. 2013; Schlaepfer et al. 2002).

I will test the hypothesis that a population average increase in body mass can be interpreted as a conservative bet-hedging strategy to cope with environmental variability on the long-term. To do so, I will simulate different bet-hedging scenario to compare there predictions to the observed patterns in our population.

2. Identify key patches for metapopulation persistence using Social Network Analysis methods. (*Continuation of a previously started work at NTNU with Dr. Yimen Araya-Ajoy*)

A metapopulation is defined as a set of subpopulations distributed across various patches, more or less interconnected (!!). Links between subpopualtions (i.e., migration fluxes) in a metapopulations is crucial for its survival over time as it maintains genetic diversity (!!). If a subpopulation finds itself isolated from the rest of the network, lack of genetic diversity putting it at risk of extinction (!!).

To various extent, a metapopulation can be viewed as a population-scale network (i.e., a set of patches connected by edges of varying intensity). Therefore, Social Network Analysis (SNA, !!) can offers a valuable approach to study metapopulation dynamics by identifying key patches contributing to network connectivity and resilience.

Using data from a long-term study on a wild house sparrow metapopulation inhabiting an archipelago in the district Helgeland, northern Norway (!!), I conducted preliminary analysis to test these analysis. Initial findings (e.g., non-random structural patterns and hierarchical order in island selection for emigration, independant from geographic distance) provides encouraging insight about migratory behavior within metapopulation.

Focus on understanding the migration dynamics and identifying critical patches in metapopulation can offer insight for policymakers. We hope to developp new methodes allowing to take effective measures, applicable more efficiently with less need of extensive data. This would help to preserve metapopulation, safeguarding genetic diversity, connectivity and better understanding their resilience capacity in context of changing environment. I am planing to test the efficiency of different tools already developps for SNA, applied to metapopulation dynamics, and try to developp new ones from simulations.

Expected Products

Introduction (*MSc second year project*); *Manuscript is being prepared for submision to Evolution*

1. Beyond plasticity: Study of the evolutionary character of the increase in body mass in the Yellow-bellied Marmot. Augustin Birot, Dan Blumstein and Julien Martin

Thesis

2. What has changed in Yellow-bellied Marmots' body mass. Augustin Birot, Dan Blumstein and Julien Martin
3. Break free from bad envrionmental proxies when detecting I^*E , the DHGLMs solution. Augustin Birot, Ned Dochtermann and Julien Martin
4. Why have Marmots only gotten bigger now? Augustin Birot, Dan Blumstein and Julien Martin
5. Balance between body condition and experience as predictor of marmots' behavior. Augustin Birot, Vincent Careau, Denis Réale, Dan Blumstein and Julien Martin

Side projects

6. Buffer environmental instability by increasing your body mass: an application of conservative bet-hedging in a hibernant rodent. Augustin Birot, Dan Blumstein and Julien Martin
7. Identify key patches for metapopulation persistence using Social Network Analysis methods. Augustin Birot, Yimen Araya-Ajoy, Bernt-Erik Sæther, Henrik Jensen, Jonathan Wright, Pierre Bize and Julien Martin

Timeline

PhD - First year	Fall 2024				Winter 2025				Spring-Summer 2025			
	Sep	Oct	Nov	Dec	Jan	Feb	Mar	Apr	May	Jun	Jul	Aug
Task												
Professional												
TA (Biostats)												
Grants & scholarships application												
Conference & Networking												
Program												
Proposal - Submission												
Comprehensive exam												
TAC meeting												
Research												
Proposal - Literature review												
Proposal - Writing (Thesis introduction)												
MSc Project (Paper 1) - Finis analysis												
Paper 1 - Writing												
Paper 1 - Manuscript submission												
Field season												
	Working on it											
	<input checked="" type="checkbox"/> Deadline											
PhD - Second year	Fall 2025				Winter 2026				Spring-Summer 2026			
Task	Sep	Oct	Nov	Dec	Jan	Feb	Mar	Apr	May	Jun	Jul	Aug
Professional												
TA (TBD)												
TA (TBD)												
Grants & scholarships application												
Conference & Networking												
Honors student supervision												
Program												
Class (TBD)												
TAC meeting												
Research												
Chapter 1 - Analysis												
Chapter 1 - Writing (Paper 2)												
Paper 2 - Manuscript submission												
Chapter 2 - Analysis												
Field season												
	Working on it											
	<input checked="" type="checkbox"/> Deadline											

Figure 11: Proposed timeline for the first and second year.

Figure 12: Proposed timeline for the third and fourth year.

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