Title: Precipitation drives global variation in natural selection

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One sentence summary: Global and local climate conditions predict variation in natural selection across diverse plant and animal populations.

Abstract:

Climate change has the potential to affect the ecology and evolution of every species on Earth. While the ecological consequences of climate change are increasingly well documented, the effects of climate on the key evolutionary process driving adaptation—natural selection—is largely unknown. We report that aspects of precipitation and potential evapotranspiration, along with the North Atlantic Oscillation, predicted variation in selection across plant and animal populations throughout many terrestrial biomes, whereas temperature explained little variation. By showing that selection was influenced by climate variation, our results indicate that climate change may cause widespread alterations in selection regimes, potentially shifting evolutionary trajectories at a global scale.

Main text:

Climate affects organisms in ways that ultimately shape patterns of biodiversity (1). Consequently, the rapid changes in Earth's recent climate impose challenges for many organisms, often reducing population fitness (2-4). While some species may migrate and undergo range shifts to avoid climate-induced declines and potential extinction (5), an alternative outcome is adaptive evolution in response to selection imposed by climate (6). However, we lack a general understanding of whether local and global climatic factors such as temperature, precipitation, and water availability influence selection (2, 7). Understanding these effects is critical for predicting the consequences of increasing droughts, heat waves, and extreme precipitation events that are expected in many regions (8, 9).

To quantify how climate variation influences selection, we assembled a large database of standardized directional selection gradients and differentials from spatially (mean = 4.6 ± 5.4 [standard deviation, SD] populations, range = 2 - 59 populations) and temporally (mean = 5.2 ± 6.8 [SD] years, range = 2 - 45 years) replicated selection studies (N = 168) in plant and animal populations (Table 1, Database S1). We focused on directional selection (selection that can generate increases or decreases in trait values) because it is well-characterized and is likely to drive rapid evolution (*10*) in response to variation in climatic factors. However, selection acting on trait combinations and trait variance may also be affected by climate (*7*). Selection gradients estimate the strength and direction of selection acting directly on a trait, while differentials estimate 'total selection' on a trait via both direct and indirect selection because of trait correlations (*11*). These standardized selection coefficients describe selection in terms of the relationship between relative fitness and quantitative traits measured in standard deviations, thus facilitating cross-study comparisons (*11*, *12*).

Geographically, the database contains many estimates of selection from temperate, midlatitude regions centered at 40° N (Fig. 1A). The populations in this database span many terrestrial biomes on Earth, with the exception of tundra and tropical rainforests where selection has rarely been quantified (Fig. 1B). This exception is concerning because tundra and tropical rainforests are likely to face severe effects of climate change (1, 13). Spatially and temporally replicated studies of selection in aquatic environments are also uncommon (Table 1), so our results pertain mainly to terrestrial systems. Additionally, the majority of studies are from vertebrate and plant populations, use fecundity or survival as a fitness measure, and use morphological traits (Table 1).

These data allowed us to determine whether directional selection covaries with changes in climatic factors among populations or across time within a given population. For each set of selection estimates, we geo-referenced the population and cross-referenced each population and time point with corresponding values of both local and global climatic factors (Database S2). We then used a random effects Bayesian Markov chain Monte Carlo meta-analysis to estimate the proportion of variation in selection within spatially and temporally replicated studies that was associated with climatic factors (*14*). This analysis is a hierarchical model, which separates the observation process (accounting for statistical noise in inference of individual selection coefficients because of sampling error) from a process model (modelling variation in the selection coefficients in relation to climate variables) (*14*). Under this analytical framework, we used a random regression mixed model component to model the distribution of within-study variation in the dependence of selection on climatic factors (*14*). As a measure of effect size, we present the mean and 95% credible intervals of the proportion of within-study variation in selection explained by a given climatic factor.

To investigate the role of local (0.5 x 0.5 degree cells) climatic factors, we analyzed air temperature, precipitation, and potential evapotranspiration (PET). While there is likely climate variation within a 0.5 degree grid, the populations where selection was quantified will often be spread out over this grid area, and the scale of climate variation is typically at an even larger geographic scale. We analyzed the data in two ways: with spatially and temporally replicated selection estimates both included together and treated separately. We modeled how mean annual

values of climatic factors influenced directional selection, as well as variation (the standard deviation [SD]), and the influence of extremes (minimum and maximum monthly values for a year) in these climatic factors because climate extremes frequently determine fitness and are expected to increase with climate change (*15, 16*).

When combining spatial and temporal studies, models that included temperature factors did not explain variation in selection (Fig. 2A, B). However, 20-40% of the variation in selection was associated with precipitation mean, maximum, and SD (Fig. 2C, D). Because precipitation factors are correlated (Table S1), our results collectively illustrate the potentially general importance of local precipitation as a selective force. In addition, minimum PET explained more than 20% of the variation in selection across the dataset (Fig. 2E, F). When we ran the analyses separately for spatial and temporal selection, the results largely mirrored the patterns in the combined analysis (Figs. S1-S2). However, we found that for selection gradients, but less so for differentials, precipitation factors were more strongly associated with temporal rather than spatial variation in selection (Figs. S1-S2). A multivariate model that included means and SDs of both precipitation and temperature together (*14*) supports the finding that variation in selection is most closely associated with precipitation factors (Table S2). However, given the low levels of replication typical of individual studies, we cannot unambiguously attribute a direct effect to any one of these four climate factors (Table S2).

We also explored whether within-study variation in selection associated with local climatic factors differed among subsets of major trait types, fitness components, and taxonomic groups (14). This analysis also indicated effects of precipitation and PET, although, there is substantial variation across the different subsets (Tables S3-S5). Among fitness components, no precipitation or PET climatic factors were consistently most associated with selection through mating success; however, selection through fecundity and survival were affected by precipitation, and survival alone was also affected by minimum PET (Table S3). Selection on morphological traits was most associated with precipitation factors, but not size or phenological

traits (Table S4). Precipitation also explained variation in selection on plants, whereas minimum PET consistently explained variation in selection among all major taxonomic groups (Table S5). While these findings are intriguing, it is important to note that the overall analysis revealed somewhat low precision in the estimates of the dependence of selection on climatic factors (Fig. 2, S1 and S2), and these subset analyses resulted in many estimates. With these important caveats in mind, we encourage a cautious interpretation of the above subset findings (*14*).

In addition to local climate variation, global climate cycles are known to be powerful agents of selection (*17*), but their capacity to operate as drivers of selection more broadly is unclear. To explore how annual global climate cycles may affect selection, we modeled the relationship between temporal variation in selection and the North Atlantic Oscillation (NAO) and the Oceanic Niño Index (ONI), which provide measures of inter-annual variability in atmospheric circulation for northern hemisphere and equatorial regions, respectively (*14*).

We found that the NAO explained between 10-30% of the variation in selection, whereas the ONI explained no appreciable variation (Fig. 3). The NAO was also most associated with selection through fecundity as a fitness component (Table S3), selection on morphological traits (Table S3), and on invertebrate and plant populations (Table S5). The overall stronger effect of the NAO (Fig. 3) relative to the ONI index is perhaps not unexpected because the ONI index would presumably be more important at equatorial latitudes (where studies of selection are rare), whereas the NAO index would be more important at northern latitudes (where selection is well documented; Fig. 1A). Indeed, although global in their reach, there are frequently correlations between large-scale climatic indices and local variation in climatic conditions that have subsequent effects on ecological and evolutionary processes (*18, 19*). Moreover, these global climate cycles are changing in response to climate change (*20*) and may therefore have cascading effects on selection at a global scale.

Previous studies have predicted the greatest fitness consequences associated with climate variation, especially related to precipitation, should occur at northern latitudes (2). Our results

add a further nuance to these potential climate effects and suggest that variation in fitness associated with precipitation may also influence selection (Fig. 2, S1-2). Increases in strong precipitation events that are predicted for the near future (21) could therefore result in considerable shifts in patterns of selection. Similarly, variation in selection was associated with variation in minimum PET—conditions when water deficits are low. While correlative, our findings do not support the idea that short-term moisture stress, as indicated by minimum precipitation or maximum PET, is a major driver of selection. Conversely, the effects of changes in mean precipitation could result from sustained drought conditions or changes in resource abundance related to water availability (17).

Whether climate-selection coupling will lead to local adaptation and reduce the risk of extinction is difficult to predict (3, 6), because adaptive evolution also depends on genetic variation in the traits under selection (3, 11). Moreover, if selection is strong relative to existing genetic variation, and if the rate of climate change is rapid, selection might result in population extinction faster than adaptation and evolutionary rescue (3, 22). Phenotypic plasticity might also therefore have a key role in promoting population persistence due to climate change (6, 7).

Our analysis benefits from drawing on decades of accumulated inferences about natural selection. However, we acknowledge a potential limitation: annual measures of local climate factors may not always reflect the most relevant scale underpinning selection in a population (19). Although annual variation at even larger geographic scales such as the NAO (Fig. 3) often have considerable predictive power for explaining variation in demographic rates (18, 19), short-term climatic and extreme weather events, including winter storms and heat waves, can also generate strong selection (23). Our finding of no effect of temperature on selection, despite case studies showing an influence of temperature (24), suggests that such selection may be occasionally driven by shorter-term thermal variation. The association between selection and PET is consistent with this interpretation because PET is calculated from temperature, but reflects temperature during the growing season when selection is most often studied. In contrast,

the observed relationship between precipitation and selection at the annual scale makes sense because moisture availability is determined by precipitation over longer periods. Ultimately, to more fully understand and predict the consequence of climate variation on selection we also need replicated transplant experiments across broad climate gradients in diverse systems (*6*). Transplant experiments would be especially beneficial because past selection may have eroded trait variation as populations locally adapted to a given climate regime, and such experiments would force populations to experience potentially stronger selective climate conditions, much like they could under climate change.

We have identified a signature of the effects of climate on selection in a phylogenetically diverse dataset across multiple environments. This provides evidence that local and global climate cycles are likely important drivers of selection in the wild. Thus, rather than selection being driven entirely by the local idiosyncrasies of each system, selection is partly predictable based on shared environmental features. Although ecologists and biogeographers have long recognized the importance of climate for explaining major ecological patterns, our analyses reveal a role for climate in explaining a key evolutionary process. In this era of unprecedented change to Earth's climate (*8*, *9*), and as future climatic conditions are expected to become increasingly more variable (*15*), natural populations will likely have to contend with greater climate variation than they have in the recent past. Such shifting climatic conditions, particularly changing precipitation patterns (*2*, *21*), may present a challenge for many organisms (*7*, *16*).

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Acknowledgments: We thank the field biologists whose efforts allowed us to perform this analysis, and C. Benkman, A. Hendry, and M. McPeek for comments. This work originated from a NESCent working group (NSF grant EF-0905606). AMS acknowledges NSF (DEB1620046); BCS is a Wolfson Research Merit Award holder; RAM was supported by NSF (DBI-1300426); KMG was supported by NSERC; LEBK was supported by the Australian Research Council. The data reported in this paper are available at Datadryad.org and as Supporting Online Material.

Supplementary Materials:

Materials and Methods

Figs. S1 and S2

Table S1-S5

Databases S1 and S2

References (25-41)

Figure legends:

Fig. 1. Selection estimates included in this study are broadly distributed geographically and

in climate space. (**A**) Red circles denote individual study locations of natural selection. (**B**) Shown are individual studies overlaid on Whittaker's terrestrial biome plot, which demarks biomes as a function of mean annual precipitation and temperature (*14*). Points represent mean annual temperature and precipitation across the years of investigation for each study and lines denote the minimum and maximum across the time period of each study. **Fig. 2. Variation in selection is explained by local climate factors.** Shown are mean and 95% credible intervals of the proportion of within-study variation in selection (combining temporal and spatial variation; see fig. S1 and S2 for temporal and spatial variation analyzed separately, respectively) explained by a given climatic factor. Little variation in selection gradients (**A**) and differentials (**B**) is accounted for by temperature, whereas considerable variation in gradients (**C**) and differentials (**D**) is accounted for by precipitation. Likewise, minimum PET also consistently explains variation in selection for both selection gradients (**E**) and differentials (**F**).

Figure 3. Variation in selection is explained by global climate indices. Shown are mean and 95% credible intervals of the proportion of within-study variation in selection gradients (black circles) and differentials (grey circles) explained by the North Atlantic Oscillation (NAO) index and the Oceanic Niño Index (ONI).

Table 1. Summary of records in the selection database.

Numbers refer to the number of items in the database. Only those records with SE's were used in analyses (14).

	Replicate type	
Item	Spatial	Temporal
Studies	84	120
Selection coefficients		
Linear differentials	1608	2539
Linear gradients	2658	3120
Species	70	97
Habitat		
Terrestrial	3098	4409
Freshwater	527	713
Marine	8	73
Taxon type		
Invertebrates	1050	627
Plants	1381	1046
Vertebrates	1202	3522
Trait Type		
Behavioral	21	54
Other	126	286
Morphological	2298	1818
Life History	334	542
Principal Components	158	307
Phenology	327	1154
Size	369	1034
Fitness Component		
Fecundity/Fertility	1848	1758
Mating Success	847	863
Other	227	35
Survival	656	2481
Survival and Fecundity	16	0
Total Fitness	39	58