

# 1 Evidence of climate change impacts on the iconic *Welwitschia mirabilis* in the Namib Desert

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14

## 15 Abstract

16 Climate change represents an important threat to global biodiversity and African ecosystems are particularly  
 17 vulnerable. Recent studies predicted substantial variations of climatic suitability for *Welwitschia mirabilis*  
 18 under future conditions. Latitudinal/altitudinal range shifts are well-known responses to climate change but  
 19 not coherent patterns were documented. This study aims to verify whether welwitschia populations are  
 20 responding to climate change and if the assumption of a latitudinal/altitudinal shift is applicable. We  
 21 collected field data on welwitschia distribution, health condition, reproductive status, and plant size in  
 22 northern Namibia. We used ecological niche models to predict the expected geographic shift of climatic  
 23 suitability under future scenarios. For each variable, we compared the observed pattern with the expected  
 24 responses. Finally, we tested the presence of simple geographical gradients in the observed patterns. The  
 25 realized thermal niche of welwitschia will be almost completely unavailable in the next 30 years in northern  
 26 Namibia. Expected reductions of climatic suitability in the stand sites are strongly associated with indicators  
 27 of negative population conditions. The same population conditions does not fit any simple latitudinal or  
 28 altitudinal gradient. The observed pattern of population conditions mirrors the expected pattern of climate  
 29 change effect but no simple geographical gradient was relieved. Overall, we observed negative population  
 30 conditions in areas with stronger reductions of suitability. This makes welwitschia a suitable sentinel for  
 31 climate change effect in the Namib Desert ecosystems. Our approach to detect population responses to  
 32 climate change could be extensively adopted for selecting sentinel species in other regions and ecosystems.

33

## 34 Keywords

35 Global warming, climatic suitability, range shift, niche modeling, southern Africa

## 1. Introduction

Climate change is one of the strongest threats for ecosystems worldwide. Variations in the density of species, range shifts, and extinction events have been documented at local and global level (Cristofari et al., 2018; Parmesan and Yohe, 2003; Walther et al., 2002). Furthermore, changes in species diversity, ecosystem functioning, and service provision are expected for the future as a consequence of additional pressures on natural populations (Ge et al., 2015; e.g. Hole et al., 2009; Moritz and Agudo, 2013). In Africa, deep impacts by climate change were forecasted for animals (e.g. Garcia et al., 2012; Huntley and Barnard, 2012; Kirchhof et al., 2017), plants (Blach-Overgaard et al., 2015; e.g. Midgley et al., 2003; Midgley and Bond, 2015), and biodiversity conservation in general (Hole et al., 2009; Revermann et al., 2018). Arid regions of southern Africa seem to be particularly exposed to the effects of climate change (Midgley and Thuiller, 2011). For the quiver tree (*Aloidendron dichotomum* Klopper & Gideon 2013), climate-linked increases of mortality were observed, although this evidence is still controversial (Foden, 2002; Foden et al., 2007; Guo et al., 2011; Jack et al., 2016).

Recently, Bombi (2018) highlighted potential effects of climate change on welwitschia trees (*Welwitschia mirabilis* Hooker 1863; welwitschia) were highlighted. *Welwitschia mirabilis* is regarded as a living fossil, representing an ancient lineage of gymnosperm plants and it is recognized as a symbol of the Namib Desert biodiversity. This species has a peculiar morphology, being a long-living dwarf tree with only two leaves growing throughout the entire plant life (Roskov et al., 2019). This is also a key species in the Namib ecosystems, where it provides food, water, and refuge for many animal species (Henschel and Seely, 2000). The distribution of *W. mirabilis* is restricted to the central and northern Namib Desert, extending from the Kuiseb River in Namibia to the Nicolau River, north of Namibe, in Angola (Giess, 1969; Kers, 1967). In this area, welwitschia trees occur in four separated sub-ranges, three in western Namibia (Fig. 1A) (Bubenzer et al., 2004) and one in south-western Angola. Bombi (2018) showed that populations living in three Namibian subranges have experienced and will face rather different climatic conditions, and a significant reduction of climatic suitability is expected in the northernmost Namibian subrange under current climate change. In particular, the ongoing rise of temperature can drive the local climate out of the realized niche for the northern populations, thus increasing their extinction risk (Bombi, 2018). Although these findings were

63 potentially important for conservation planning, the study was based on low spatial resolution data available  
64 at the national scale, thus limiting its utility at a finer scale for such purposes.

65 Distribution ranges shifts are well-known responses of species to climate change (Araújo and Rahbek, 2006;  
66 Garcia et al., 2014; Parmesan, 2006). These shifts have been generally described as poleward and upward  
67 movements of species to track suitable temperature conditions along latitudinal and altitudinal gradients  
68 (Hickling et al., 2006; Parmesan et al., 1999; Thomas, 2010). However, in many cases documented  
69 geographic patterns of response are complex and do not align with simple latitudinal and altitudinal shifts  
70 (Fei et al., 2017). Indeed, the assumption of simple, uni-directional distribution shifts does not account for  
71 intricate interactions among temperature, precipitation, and species-specific tolerances and can drive to  
72 substantially underestimate the effect of climate change on species distributions (VanDerWal et al., 2012).  
73 To overcome these drawbacks, one promising approach is based on the comparison of the species-specific  
74 spatial pattern of expected responses, generated by predictive models, with the observed pattern of species  
75 response measured in the field with appropriate metrics of population trends (Bombi et al., 2017). This  
76 approach can increase our capacity of identifying the footprint of climate change on species dynamics.

77 The main aim of this study is to verify whether the observed geographic pattern of population conditions of  
78 *welwitschia* trees in northwestern Namibia can be associated to the ongoing climate change. Secondly, we  
79 tested if the same pattern follows a latitudinal or altitudinal gradient in agreement with the assumption of a  
80 poleward or upward range shift. More specifically, we want first to validate with field-based data the  
81 predictions of potential impacts of climate change on *W. mirabilis*. Second, we want to assess whether the  
82 simple assumption of a poleward/upward range shift is suitable for detecting climate change footprints in this  
83 case. To do this, we compared the geographic pattern of population conditions, measured in the field, with  
84 the expected pattern of response, predicted by ecological niche models. If climate change affects *welwitschia*  
85 populations, we expect worst population conditions in sites where climatic suitability will decrease than in  
86 sites where climatic suitability will increase. Moreover, if a poleward/upward range shift is the major  
87 response to climate change we can expect a latitudinal or altitudinal trend in the observed patterns of  
88 response. Since potential divergent responses to climate change by intraspecific lineages were evidenced  
89 (Pearman et al., 2010) and different realized niche were described for each distinct Namibian subranges  
90 (Bombi et al., 2017), we focused on populations in the northern subrange and considered them as an

independent ecological unit, with its own climatic niche and with its (sub)specific expected response. By verifying our main and secondary hypotheses, we provide documented information to the long-term conservation of *W. mirabilis* and further contribute to the scientific debate on the climate change impacts on biodiversity.

## 2. Materials and methods

### 2.1. Field data collection

During May 2019, we carried out a field expedition in the northernmost Namibian subrange of *W. mirabilis*, as defined by the 'Digital Atlas of Namibia' (Bubenzer et al., 2004), in order to obtain information relevant for the species conservation. During the expedition, we spent 10 full days searching for *welwitschia* trees across the subrange by (1) driving at low speed along the available tracks (more than 330 km) while recording the presence of plants in a ~30 m wide transect on each side of the vehicle, and (2) walking across potentially suitable habitats (more than 65 km). The starting points and spatial extent of our walking transect-based searches were informed by the knowledge of our local team members, who have an intimate knowledge of the area. We are confident that the combination of local knowledge-informed searches and systematic transects extending beyond the known range have allowed us to establish the extent and characteristics of the majority of this sub-range.

During our transects, we collected detailed data on plant location, health condition, reproductive status, and body size. Specifically, we recorded the precise coordinates (using a handheld GPS), the gender, and the presence/absence of cones for almost all the individual plants we observed (just few, unreachable plants were excluded). In sites with a sufficient number of plants, we also measured the stem diameters (minimum and maximum along the two main axes of the stem) and the leaf length, and we recorded the health condition (ranked as dead, poor, average, or good) for a random subset of ~60 plants. We ranked health condition on a four-point scale (dead, poor, average, good) based on leaf color and the general aspect of the plant. Although this is a relatively coarse scale, the brightness of the green color and the ratio of red/brown to green together are a remarkably consistent and accurate indicator of good health condition as measured by photosynthesis efficiency (Shuuya, 2016). The green color of the leaf is associated to the chlorophyll content and the

photosynthesis efficiency of the tissues (Menzies et al., 2016; Terashima et al., 2009), which is influenced by environmental stress (Chaves, 1991; Munns et al., 2006). An estimate of health condition such as the above is both a direct reflection of the environmental (including climatic) stress that the plant experiences and an index of the likelihood that its resistance to parasites might be compromised (Mattson and Haack, 1987; Schoeneweiss, 1978). We expected that changes in local climate will be visible in its leaf colour as a quick proxy of plant health.

## 2.2. Observed pattern of response

For each welwitschia stand (defined *a posteriori*, through a GIS-based analysis, as a group of plants separated from the other groups by a distance larger than the intra-group mean distance), we calculated three categories of indicators of population response (derived from plant health, reproductive status, and size) from the field-measured data. For each stand, we calculated the proportion of plants that were dead or in poor, average and good condition. We also calculated the reproductive status (the proportion of plants in the stand that had cones) and the plant size (average stem length, stem width, and leaf length).

## 2.3. Expected pattern of response

We used a spatially explicit approach based on ecological niche modeling to predict the geographic pattern of plant response expected as a consequence of climate change. To do this, we defined our study area as a bounding box three times larger than the latitudinal and longitudinal extent of the previously known subrange of welwitschia in northern Namibia (Bubenzer et al., 2004). Inside this study area, we fitted models on 1000 pseudo-presence/absence points by using climate data from the WorldClim databank (Hijmans et al., 2005) at the spatial resolution of 30 arcsec (about 1 km). In order to control the model-associated uncertainty, we adopted an ensemble forecasting approach (Araújo and New, 2007) in the R-based (R Core Team, 2018) *biomod2* Package (Thuiller et al., 2016). In particular, we used Generalized Linear Models (McCullagh and Nelder, 1989), Generalized Additive Models (Hastie and Tibshirani, 1986), Generalized

143 Boosting Models (Ridgeway, 1999), Classification Tree Analyses (Breiman et al., 1984), Artificial Neural  
144 Network (Ripley, 1996), and Random Forest (Breiman, 2001) methods.

145 Pseudo-presence/absence points were randomly generated across the study area and classified as presence or  
146 absence points based on their position inside or outside the species extent of occurrence, generated as a  
147 minimum convex polygon from our detailed distribution data. Multicollinearity among predictors was  
148 reduced by discarding those with variance inflation factor higher than five (Belsley, 1991). Each independent  
149 model was projected into the study area under current climatic conditions and three-fold cross-validated by  
150 calculating the true skill statistic (TSS) (Allouche et al., 2006). Finally, we generated a single consensus  
151 model of current suitability by calculating the TSS-weighted sum of the independent models. Future climate  
152 suitability was predicted by projecting the models into future climatic conditions across the study area. All  
153 the available scenarios of future (2050) climate from CMIP5 were utilized for projecting the models.  
154 Suitability variation over time was calculated as the difference between future and current suitability and  
155 assigned to the observed plant stands on the basis of their location.

156

#### 157 *2.4. Association between observed and expected patterns*

158 For each variable, we tested the linkage between the observed and the expected patterns of responses by  
159 adopting a null-model approach (Gotelli and Graves, 1996; Gotelli and Ulrich, 2012; Harvey et al., 1983).  
160 First, we quantified the observed correlation between measured values and expected suitability variation in  
161 the same sites ( $r_{obs}$ ) by calculating the Pearson  $r$ . Second, we generated in *R* (R Core Team, 2018) (as all the  
162 other analyses) 30,000 random permutations of the measured values and calculated the simulated correlation  
163 with the expected suitability variation for each permutation ( $r_{sim}$ ). Third, we calculated the probability of the  
164 null hypothesis that the observed correlation was drawn at random from the distribution of the simulated  
165 correlations (Gotelli, 2000). Finally, in order to control the familywise error rate due to multiple  
166 comparisons, we corrected our  $p$  values adopting the approach proposed by Benjamini & Hochberg (1995).  
167 These corrected  $p$  values ( $p_{corr}$ ) measure the level to which the suitability variation (corresponding to the  
168 expected response) due to climate change explains the actual responses observed in the different stands.

169 In addition, we tested whether the observed responses follows a general and simple geographic pattern. In  
 170 particular, we tested the hypothesis of a latitudinal (equator-to-pole) or altitudinal (low-to-high elevation)  
 171 range shift. To do this, we adopted the same approach used for testing the linkage between the observed and  
 172 the model-based expected patterns of responses. In particular, we contrasted each measured variable with the  
 173 stand latitudes and altitudes. We quantified the correlation between the measured variable and the  
 174 latitude/altitude, calculated the probability that the observed correlation comes randomly from the simulated  
 175 correlations after 30,000 random permutations, and corrected our  $p$  values with the Benjamini & Hochberg  
 176 (1995) approach. As a result, we obtained an estimation of the extent to which climate change effects can be  
 177 explained as a simple geographic gradient.

178

### 179 3. Results

180 Overall, we recorded 1330 plants within the known distribution of *W. mirabilis* in northern Namibia. These  
 181 plants are clustered in 12 distinct stands, which are scattered across the central part of the known range at  
 182 elevations between 806 and 991 m above sea level (Fig. 1B). Our local team members, who know the area  
 183 and the species intimately from years of herding goats and cattle, could not point out any more locations than  
 184 the ones we recorded during the current study. We are thus confident that the plants that we recorded or  
 185 observed, and the resulting extent of occurrence, represent the majority of plants in this northern-Namibian  
 186 sub-range. The surface area of each recorded stand varied from 2000 – 825,000 m<sup>2</sup> and the number of plants  
 187 per stand varied between four and ~400. The extent of occurrence of *welwitschia* in the area covers about  
 188 215 km<sup>2</sup> and the inter-stand distance varied from 1.8 to 30 km (Fig. 1B). This is a markedly smaller area than  
 189 the distribution map previously published for the species in northern Namibia (Bubenzer et al., 2004) but a  
 190 significant improvement of the existing, but unpublished knowledge of plant location in the area.

191 The available climatic models revealed that the realized thermal niche of *W. mirabilis* in northern Namibia is  
 192 expected to become completely unavailable within its current extent of occurrence (Fig. 2A). In particular,  
 193 annual mean temperature within the stands will rise about 1.5 – 2.5 °C, with strong variations among the  
 194 different scenarios. In contrast, the total annual precipitation will likely remain relatively stable (Fig. 2B),  
 195 with small reductions or increments forecasted by different scenarios.

196

### 197 *3.1. Observed variation of plant parameters*

198 The most common class of health condition was ‘average’, with 50% of all the plants and a range between  
 199 32% and 74% across individual stands being found in this status. Plants in ‘poor’ condition were 32%  
 200 (range: 11-50%), but only 10% of all plants were in a ‘good’ condition (range: 0-30%). Seven percent of all  
 201 plants were dead (range: 0-30%) and 56% (range: 10-90%) had cones. Not all individuals could be sexed, but  
 202 among those that were, 56% were males, with a sex ratio (males/females) ranging between 0.6 – 1.7 across  
 203 stands. Stem length and stem width were highly variable, ranging from 2 to 100 cm ( $18.8 \pm 14.1$  cm; range:  
 204 10-33 cm) and from 0.3 and 55 cm ( $10.3 \pm 9.8$  cm; range 4.6-22 cm), respectively. Leaf length varied from  
 205 almost 0 cm (completely browsed plants) up to 93 cm ( $18.7 \pm 13.4$ ; range 11-40 cm).

206

### 207 *3.2. Expected pattern of species response*

208 The current climatic suitability for *W. mirabilis* is especially high in the eastern half of its extent of  
 209 occurrence (Fig. 3A). Some areas to the population’s south, as well as a northwest-trending band to the north  
 210 are also predicted to be highly suitable (Fig. 3A), although plants have never been recorded from these areas  
 211 before, nor did we find any. Our models further predict that, in the future, the most suitable areas will occur  
 212 to the northwest of the current extent of occurrence (Fig. 3B). As a result, the plants within in the current  
 213 extent of occurrence would all experience a reduction of climatic suitability (Fig. 3C) and may thus respond  
 214 negatively. All the recorded stands are expected to face suitability reductions in the years to come, with  
 215 variable intensities between almost no reduction and complete reduction. Even those areas that are currently  
 216 suitable but where the species has not been recorded (Fig. 3A) will similarly experience reductions in  
 217 suitability (Fig. 3C). In contrast, positive responses may occur in the future suitable area to the northwest of  
 218 the current extent of occurrence (Fig. 3C).

219

### 220 *3.3. Observed vs expected patterns*

Stronger predicted reductions of climatic suitability in the stand sites are associated with lower plant health condition, fewer plants with cones, and an increased number of dead plants. More specifically, the proportion of plants in poor condition in each stand increases with the reduction of suitability (Fig. 4A). In contrast, the proportion of plants in average and, marginally, of plants in good condition decreases as suitability variation decreases (Fig. 4B and 4C). The proportion of plants with cones (i.e. a proxy of the potential population recruitment) is lower in stands where stronger reductions of climatic suitability are expected (Fig. 4D). At the same time, the proportion of dead plants (i.e. population mortality) is negatively correlated with the expected variation of climatic suitability (Fig. 4E). However, neither the number of plants per stand (i.e. population size) (Fig. 4F) nor plant body size (Fig. 4G, 4H, and 4I) is correlated with the suitability variation. The observed geographic pattern of species response does not follow any simple geographic gradient. Indeed, the latitude of *welwitschia* stands is not correlated with any measured variable (Table 1). Similarly, altitude and the measured variables are not correlated (Table 1). Overall, no latitudinal or altitudinal variation is occurring as a response to climate change.

#### 4. Discussion

The results we obtained are coherent with our main hypothesis that the observed pattern of population conditions of *welwitschia* trees in northern Namibia can be explained as consequences of climate change. Differently, the secondary hypothesis that the geographic pattern of this response to climate change follows a latitudinal or altitudinal gradient is not verified. These outcomes strongly suggest that the ongoing climate change can cause significant alterations to *welwitschia* populations in the area, can produce important changes (i.e. shifts, contraction) in the local species distribution, and can represent a not negligible threat for the long term conservation of the species. On the other hand, these outcomes also evidence that the potentially serious impact of climate change on this species would be undetected if searched with an approach based on the simple assumption of poleward/upwards range shift.

Inter-stand variations of different population parameters are correlated to change in climatic suitability and can be interpreted as effects of climate change. In this light, the high correlation between the variation of climatic suitability and plant conditions can support a link among climate change, the distribution of plants and the variation of plant health with observed increment of individuals in poor conditions and the reduction

248 of plants in average or good conditions. The loss of climatic suitability can be also put in relationship to the  
 249 population trends, by affecting recruitment (as suggested by the observed reduction of plants with cones) and  
 250 mortality (as suggested by the observed increase of dead plants). Even if we measured static parameters of  
 251 population condition, the pattern of these static measures is coherent with the dynamism of an ongoing range  
 252 shift from areas currently suitable to areas that will be suitable in the future. Indeed, negative population  
 253 dynamics, which can be detected as bad population conditions (i.e. scarce plant health, low reproduction  
 254 potential, high mortality), are typically associated with the trailing edge of a shifting species distribution. The  
 255 surprisingly clear relationships between observed and expected patterns could indicate one of the very first  
 256 cases of documented effects of climate change on austral species (Jack et al., 2016).

257 The higher proportion of dead plants, which suggests an increased mortality, coupled with the lower  
 258 proportion of plants with cones, which suggests a reduced population recruitment, in areas that suffer  
 259 stronger effects of climate change, is particularly worrisome because can imply a global negative trend in  
 260 population size, potentially driving to the local extinction of entire *welwitschia* stands. In addition, it is worth  
 261 noting that the linkage between mortality and suitability reduction could be even underestimated by our  
 262 approach because there are no ways to discriminate plants dead long time ago (i.e. not related to climate  
 263 change), from recently dead plants. Indeed, the inclusion of old deaths in our dataset could attenuate the  
 264 climate-related signal. The lack of linkage between suitability variation and plant number could mean that  
 265 the combination of increased mortality and reduced recruitment has not caused a reduction of population size  
 266 until now. Similarly, the absence of association between the plant body size and climate change suggests that  
 267 the altered rates of recruitment and mortality did not modified the age class structure of the stands.  
 268 Nevertheless, population dynamics of long living organisms can be slow and become evident only in the  
 269 long term.

270 As mentioned above, the visual estimation of the plant condition can be considered a rough estimation of  
 271 chlorophyll content of the leaves and thus of the plant photosynthetic efficiency. Alterations of  
 272 photosynthesis is a well-known effect of environmental stress (Chaves, 1991; Munns et al., 2006). Heat  
 273 stress in particular inhibits photosynthesis in tropical and subtropical plants (Larcher, 1995; Salvucci and  
 274 Crafts-Brandner, 2004). This effect can be stronger in arid environments, where the water shortage can  
 275 hamper the leaf temperature mitigation (Idso et al., 1982). On the other hand, other studies (Shuuya, 2016)

276 evidenced that, in other populations of *W. mirabilis*, rainfall is followed by an increase in condition. As a  
 277 result, we can hypothesize that the observed worsening of plant condition is associated to the complex  
 278 interaction between the significant increment of temperature, which is the main climate alteration expected in  
 279 the area (Fig. 2), and the constant but limited water availability in the desert environment. Anyway,  
 280 specifically designed experiments would be needed to tease apart the different possible forces that could  
 281 cause the observed responses.

282 Our results confirm the expectation of previous works on the potential impacts of climate change on  
 283 *welwitschia* populations in northern Namibia. The study of Bombi (2018), carried out at the national level  
 284 and at a much coarser spatial resolution, predicted a general reduction of climatic suitability for *W. mirabilis*  
 285 and suggested potential effects on population recruitment with consequent influences on population  
 286 structure. At the same time, the author evidenced that living *welwitschia* trees would have been probably  
 287 able to cope with the expected climate suitability reduction. On this point, the correlation we observed  
 288 between increased mortality and predicted influence of climate change would indicate a more worrisome  
 289 scenario, with a progressive degradation of the plant health and the potential long-term reduction of the  
 290 population size. This evidence should encourage specific management plans for northern Namibian  
 291 populations and take into consideration climate change among the conservation issues.

292 Quantitative data on the plant physiological performances (e.g. leaf growth rate, photosynthesis level, water  
 293 use efficiency) are required to obtain a more detailed picture of the occurring alterations and to clarify the  
 294 possible mechanistic linkage with climatic stress. Repeated measures of physiological parameters in different  
 295 sites would make possible following plant responses over time. The activation of programs for the long-term  
 296 monitoring of the species in the region would be particularly helpful, allowing critical situations to be  
 297 detected at early stages and planning effective recovery measures. Obviously, activities of long-term  
 298 monitoring in this remote area would be difficult and would require the involvement of local communities as  
 299 well as the provision of significant resources by local and international agencies aimed at the conservation of  
 300 desert ecosystems in Namibia.

301 Despite the great interest on *W. mirabilis*, which is considered a living fossil, for its morphological and  
 302 evolutionary uniqueness (Khoshoo and Ahuja, 1963), and an iconic species of the Namib desert, for its key-

303 role in this ecosystem, several aspects of the species distribution and biology are still to be clarified for a  
 304 science-based conservation strategy. First, the real level of geographic and genetic isolation of the different  
 305 subranges should be verified in order to identify intra-specific evolutionary and conservation units. Second,  
 306 an effort to census and make available the current knowledge on species distribution, demography, and  
 307 conservation should be undertaken. Indeed, a significant amount of this information is probably pulverized  
 308 into a plethora of unpublished datasets and field observations. Third, an analysis of the climate change  
 309 impacts should be extended to the other subranges and a science-based assessment of the conservation status  
 310 should be made at local and global level. This set of measures could significantly contribute at planning  
 311 conservation measures for the species that are effective on the long term.

312 The geographic pattern of response we observed in *welwitschia* is more complex than the simple  
 313 poleward/upward shift that was often observed for other species (Parmesan et al., 1999; Root et al., 2003;  
 314 Thomas, 2010). In the case of *W. mirabilis* populations of northern Namibia, the observed pattern of  
 315 population conditions, which can represent a response to climate change, follows local contingencies rather  
 316 than a simple latitudinal or altitudinal trend (Table 1). This could be associated with the small scale of the  
 317 study but is also in agreement with previous large-scale studies. These studies pointed out that specific  
 318 responses to climate change can be divergent (Fei et al., 2017) and that assuming a simplified  
 319 poleward/upward species movement can bring to underestimate climate change impacts (VanDerWal et al.,  
 320 2012). In our specific case, the linkage between climate change and population conditions, which is  
 321 suggested by our results, would have been completely undetected with a simplified, but frequently used  
 322 approach based on the assumption of poleward/upward shifts.

323 The comparison of the expected pattern of response to climate change, as predicted by suitability modeling,  
 324 with the observed patterns of population conditions, as measured in the field, appeared as a powerful approach  
 325 for detecting impacts on wild species. This approach, proposed by Bombi et al. (2017), allowed to indicate  
 326 climate change as one of the most probable drivers of the geographical variation of population features we  
 327 observed in the field. This study underlines the importance of considering species responses to climate change  
 328 as the emergent property of the different effects on individual populations. At a higher biodiversity level,  
 329 ecosystem responses to climate change can be considered as the emergent property of the effects on individual  
 330 species. Such a hierarchical relationship provides direction for the application of spatial explicit approaches

such as the one used in this study, to multiple species and across diverse ecosystems. In this light, it can be advocated the setting of a large scale program for the identification of sentinel species of climate change effects, which allows to detect, estimate, and follow the climate change impacts on biodiversity and to improve the long-term conservation of species at the ecosystem level.

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491 Table 1. Correlation between the measured variables and the stand latitude and altitude.

492

	Latitude		Altitude	
	<i>r<sub>obs</sub></i>	<i>p<sub>corr</sub></i>	<i>r<sub>obs</sub></i>	<i>p<sub>corr</sub></i>
Proportion of plants in poor conditions	-0.342	0.496	0.293	0.332
Proportion of plants in average conditions	-0.083	0.375	0.039	0.33
Proportion of plants in good conditions	0.258	0.425	-0.418	0.463
Proportion of plants with cones	0.425	0.151	-0.492	0.201
Proportion of dead plants	0.275	0.356	-0.048	0.066
Number of plants	-0.669	0.06	0.096	0.461
Stem length	0.521	0.26	-0.131	0.461
Stem width	0.542	0.177	-0.279	0.225
Leaf length	-0.243	0.102	0.573	0.3765

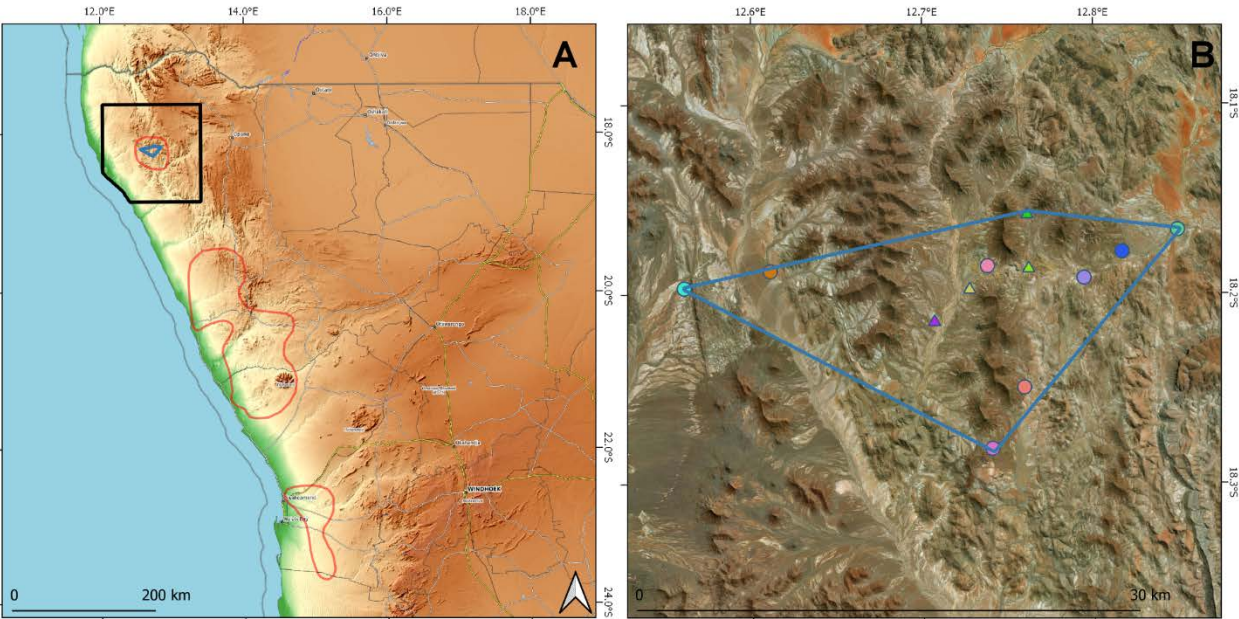
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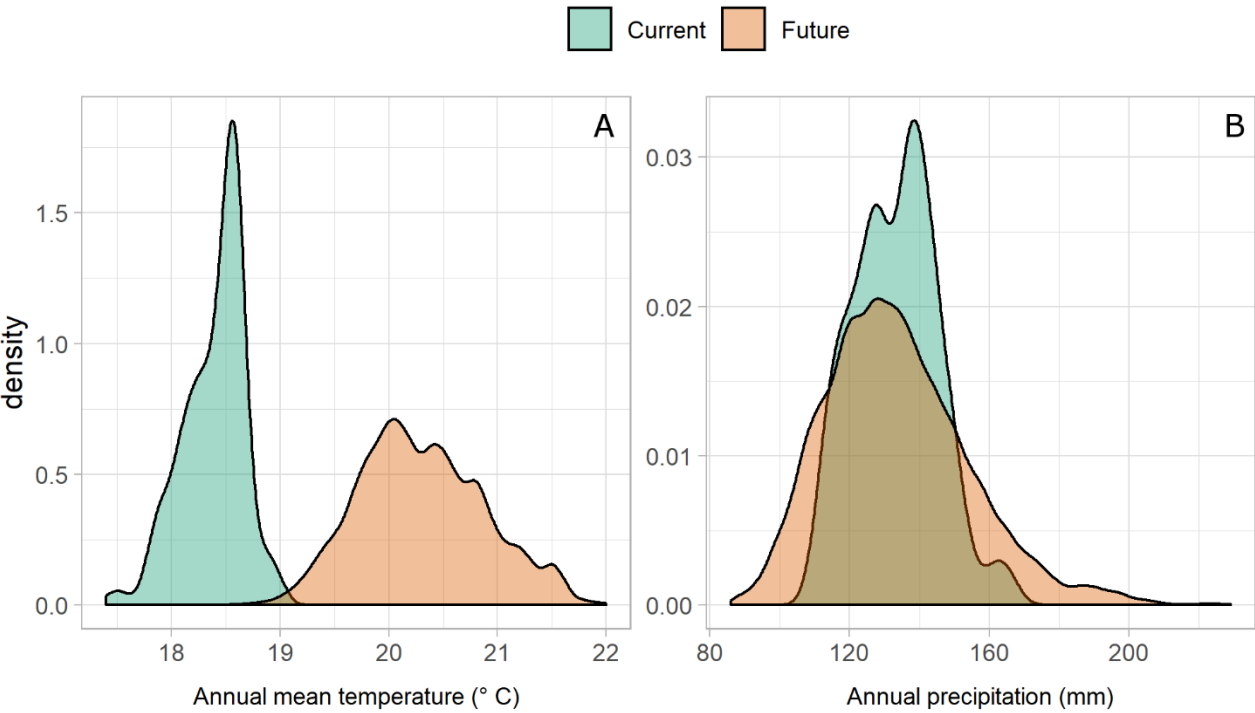
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500 Figure 1. Distribution of *Welwitschia mirabilis* in Namibia and location of the study area (A) and position of  
501 detected plant stands inside the observed extent of occurrence (B). In A, the black polygon indicates the  
502 study area and the red polygons show the known species distribution. In B, the colored triangles are the plant  
503 locations that were known before our study and the colored circles are the new occurrences. In both the  
504 maps, the blue polygons represent the boundaries of the observed extent of occurrence in Northern Namibia.

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508 Figure 2. Distribution of current climatic data (i.e. realized climatic niche of *Welwitschia mirabilis*) in the  
509 extent of occurrence (in green) and expected future values (in orange) for annual mean temperature (A) and  
510 annual precipitation (B).

511

512

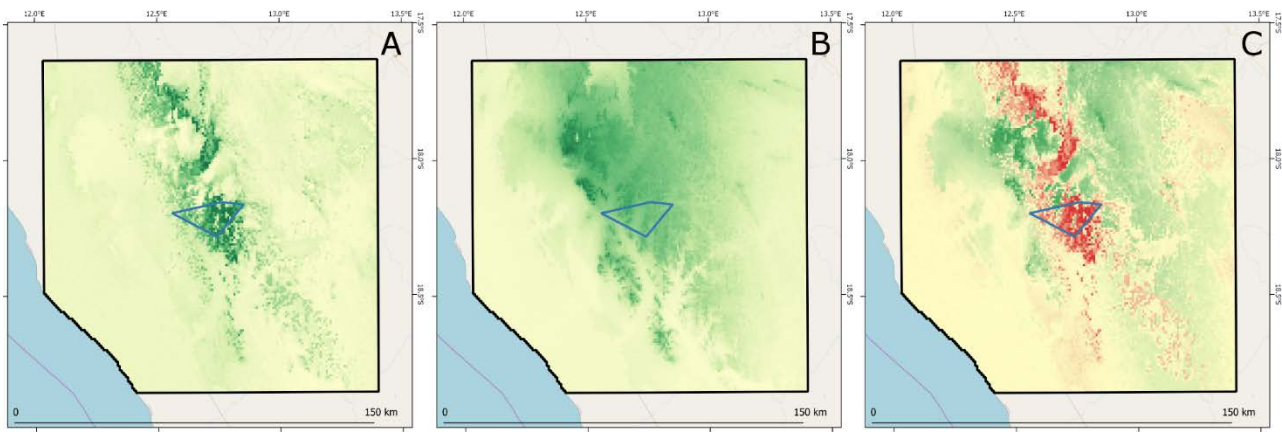
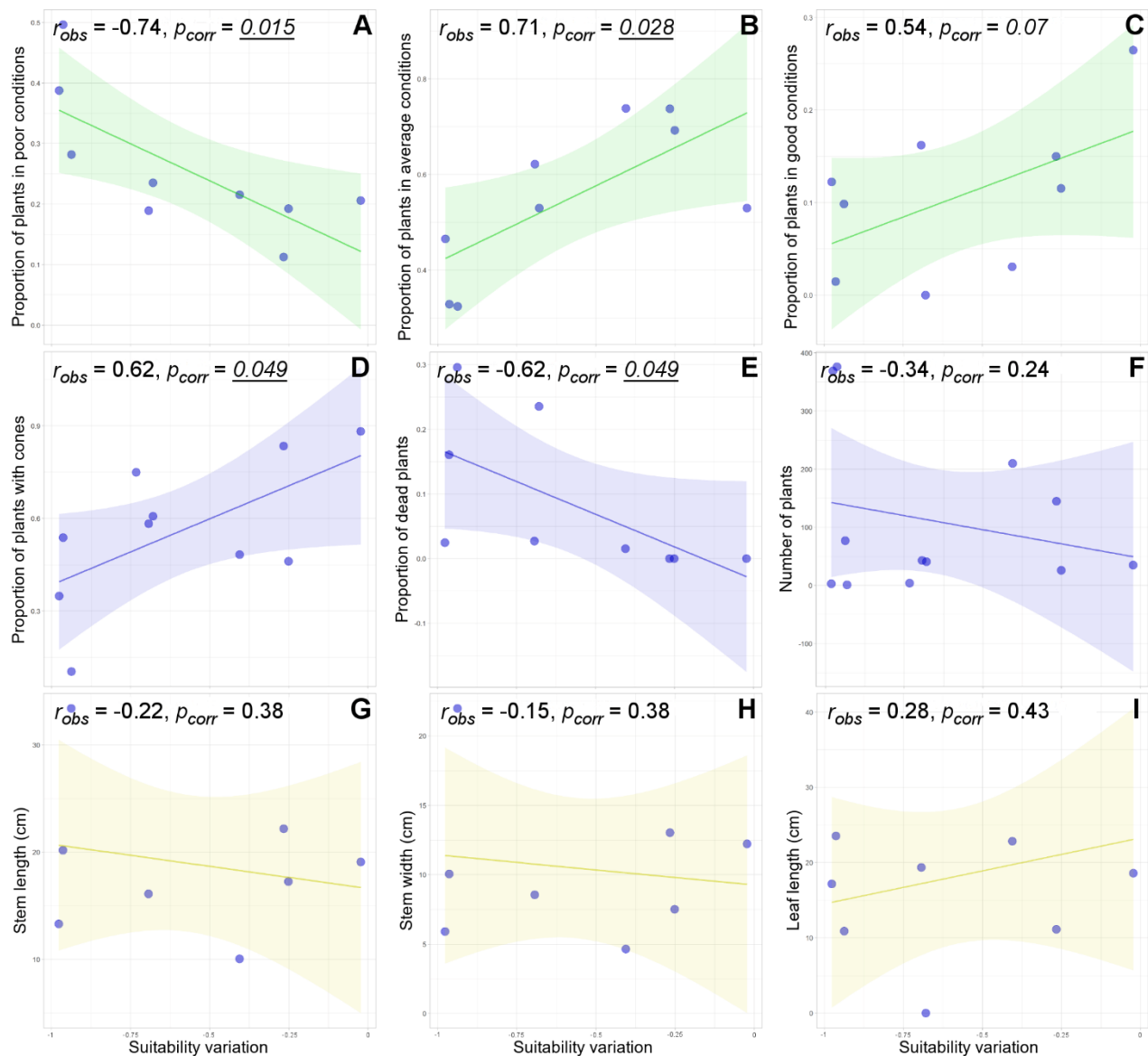


Figure 3. Climatic suitability for *W. mirabilis* in the study area under current (A) and future (B) climatic conditions (green shades indicate growing suitability). Expected suitability variation from climate change (C) (red and green shades indicate negative and positive variations respectively). In all the maps, black polygons indicate the study area and the blue polygons represent the boundaries of the observed extent of occurrence in Northern Namibia.



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525 Figure 4. Population features as functions of the expected suitability variation in the stands. In green (first  
526 row), features related to plant health condition: proportion of plants in poor (A), average (B), and good (C)  
527 conditions. In blue (second row), features related to potential population trends: proportion of plants with  
528 cones (D) and of dead plants (E), and number of plants per stand (F). In yellow (third row), features related  
529 to plant body size: stem length (G) and width (H), and leaf length (I). In all the plots, blue dots are values for  
530 plant stands. Note that suitability variation values (X-axis) are all negative; thus, the reduction of suitability  
531 increases from right to left.