

From climate change predictions to actions – conserving vulnerable animal groups in hotspots at a regional scale.

Silvia Benoliel Carvalho, José Carlos Brito, Eduardo J. Crespo, Hugh

Possingham

▶ To cite this version:

Silvia Benoliel Carvalho, José Carlos Brito, Eduardo J. Crespo, Hugh Possingham. From climate change predictions to actions – conserving vulnerable animal groups in hotspots at a regional scale.. Global Change Biology, Wiley, 2010, 16 (12), pp.3257. 10.1111/j.1365-2486.2010.02212.x . hal-00552617

HAL Id: hal-00552617 https://hal.archives-ouvertes.fr/hal-00552617

Submitted on 6 Jan 2011

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers. L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés. **Global Change Biology**

Global Change Biology

From climate change predictions to actions – conserving vulnerable animal groups in hotspots at a regional scale.

Journal:	Global Change Biology
Manuscript ID:	GCB-09-0856.R1
Wiley - Manuscript type:	Primary Research Articles
Date Submitted by the Author:	19-Jan-2010
Complete List of Authors:	Carvalho, Silvia; CIBIO - Centro de Investigação em Biodiversidade e Recursos Genéticos da Universidade do Porto; The University of Queensland, The Ecology Centre Brito, José; CIBIO - Centro de Investigação em Biodiversidade e Recursos Genéticos da Universidade do Porto Crespo, Eduardo; Faculdade de Ciências da Universidade de Lisboa, Departamento de Biologia Animal Possingham, Hugh; University of Queensland, The Ecology Centre
Keywords:	climate change, Iberian Peninsula, ensemble modelling, species distribution model, vulnerability, herptiles
Abstract:	Current climate change is a major threat to biodiversity. Species unable to adapt or move will face local or global extinction and this is more likely to happen to species with narrow climatic and habitat requirements and limited dispersal abilities, such as amphibians and reptiles. Biodiversity losses are likely to be greatest in global biodiversity hotspots where climate change is fast, such as the Iberian Peninsula. Here we assess the impact of climate change on 37 endemic and nearly endemic herptiles of the Iberian Peninsula by predicting species distributions for three different times into the future (2020, 2050 and 2080) using an ensemble of bioclimatic models and different combinations of species dispersal ability, emission levels and global circulation models. Our results show that species with Atlantic affinities that occur mainly in the North- western Iberian Peninsula have severely reduced future distributions. Up to 13 species may lose their entire potential distribution by 2080. Furthermore, our analysis indicates that the most critical period for the majority of these species will be the next decade. While there is considerable variability between the scenarios we believe that our results provide a robust relative evaluation of climate change impacts among different species. Future evaluation of the vulnerability of individual species to climate change should account for their adaptive capacity to climate change, including factors such as physiological climate tolerance,

geographical range size, local abundance, life cycle, behavioural and phenological adaptability, evolutionary potential and dispersal

1 2 3 4 5	geographical range size, local abundance, life cycle, be and phenological adaptability, evolutionary potential a
5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23	ability.
$\begin{array}{c} 23\\ 24\\ 25\\ 26\\ 27\\ 28\\ 29\\ 30\\ 31\\ 32\\ 33\\ 34\\ 35\\ 36\\ 37\\ 38\\ 39\\ 40\\ 41\\ 42\\ 43\\ 34\\ 45\\ 46\\ 47\\ 48\\ 49\\ 50\\ 51\\ 52\\ 53\\ 54\\ 55\\ 56\\ 57\end{array}$	

3	1	Title	
5 6 7	2	From climate change predictions to actions - co	onserving vulnerable animal groups in
8 9	3	hotspots at a regional scale.	
10 11 12	4	Running title	
13 14 15	5	From climate change predictions to actions	
16 17	6	Authors	
18 19	7	Sílvia B. Carvalho ^{1,2,3} José C. Brito ¹ , Eduardo J. Cres	po ^{2,5} & Hugh P. Possingham ^{3,4}
20 21 22	8		
23 24	9	1 – CIBIO - Centro de Investigação em Biodiversidad	de e Recursos Genéticos da
25 26 27	10	Universidade do Porto, R. Padre Armando Quintas, 4	485-661 Vairão, Portugal
27 28 29	11	2 – Departamento de Biologia Animal, Faculdade de	Ciências da Universidade de Lisboa,
30 31	12	Campo Grande, 1749-016 Lisboa, Portugal.	
32 33	13	3 - The University of Queensland, the Ecology Centre	e, St Lucia QLD 4072, Australia
34 35 36	14	4 - The University of Queensland, the School of Matl	hs and Physics, St Lucia QLD 4072,
37 38	15	Australia	
39 40 41	16	5 - CBA, Centro de Biologia Ambiental da Universid	lade de Lisboa, Campo Grande, 1749-
42 43	17	016 Lisboa, Portugal	
44 45	18		
46 47 48	19	Corresponding author : Silvia B. Carvalho	Other authors email address:
49 50	20	email: <u>silviacarvalho@mail.icav.up.pt</u> J	osé Carlos Brito: jcbrito@mail.icav.up.pt
51 52 53	21	Telephone: +351 252660416 E	Eduardo J. Crespo: ejcrespo@fc.ul.pt
53 54 55	22	Fax: +351 252661780	Hugh P. Possingham: <u>h.possingham@uq.edu.au</u>
56 57	23		
58 59 60	24	Key words: amphibians, reptiles, climate change, Ibe	erian Peninsula, ensemble modelling,
	25	species distribution model, vulnerability, adaptive ma	anagement

26 Abstract

Current climate change is a major threat to biodiversity. Species unable to adapt or move will face local or global extinction and this is more likely to happen to species with narrow climatic and habitat requirements and limited dispersal abilities, such as amphibians and reptiles. Biodiversity losses are likely to be greatest in global biodiversity hotspots where climate change is fast, such as the Iberian Peninsula. Here we assess the impact of climate change on 37 endemic and nearly endemic herptiles of the Iberian Peninsula by predicting species distributions for three different times into the future (2020, 2050 and 2080) using an ensemble of bioclimatic models and different combinations of species dispersal ability, emission levels and global circulation models. Our results show that species with Atlantic affinities that occur mainly in the North-western Iberian Peninsula have severely reduced future distributions. Up to 13 species may lose their entire potential distribution by 2080. Furthermore, our analysis indicates that the most critical period for the majority of these species will be the next decade. While there is considerable variability between the scenarios we believe that our results provide a robust relative evaluation of climate change impacts among different species. Future evaluation of the vulnerability of individual species to climate change should account for their adaptive capacity to climate change, including factors such as physiological climate tolerance, geographical range size, local abundance, life cycle, behavioural and phenological adaptability, evolutionary potential and dispersal ability.

45 Introduction

Anthropogenic driven climate change is evident and for the next two decades a warming of at least 0.2°C per decade is projected (IPCC, 2007) with associated changes in precipitation patterns. Throughout the history of Earth, climate has changed and species have coped and adapted to these changes, but current climate change is threatening biodiversity because it is fast compared to most past changes (Thomas *et al.*, 2004). Current climate warming was considered the second most important threat to terrestrial biodiversity, only exceeded by land-use change (Sala *et al.*, 2000).

Climate change have been reported to affect many aspects of populations: physiology, distribution, phenology, behaviour and propensity for local extinction (Hughes, 2000, Walther et al., 2002, McCarty, 2002, Root et al., 2003, Parmesan, 2006). Species may be able to adapt to climatic changes via ecological (Root et al., 2005) or evolutionary processes (Bradshaw et al., 2006, Skelly et al., 2007). However, species unable to achieve a sufficient level of adaptation will likely face local or global extinction and this is more likely to happen to species with restricted climate and habitat requirements, limited dispersal abilities and ectothermal physiology (Walther et al., 2002, Thomas et al., 2004, Massot et al., 2008).

Amphibians and reptiles are considered one of the most vulnerable taxonomic groups to climate change (Gibbons et al., 2000, Carey et al., 2003, Araújo et al., 2006, Wake, 2007). Climate warming is projected to induce a) changes in abundance; b) fragmentation of suitable habitats; c) changes in the timing of life-cycle events, such as hibernation, aestivation and breeding (Blaustein et al., 2001, Chadwick et al., 2006) and d) the spread of agents of infectious diseases such as the chytridiomycete fungus (Pounds et al., 2006, Bosch et al., 2007, Wake, 2007). The interaction of these impacts causes disruptions in population and metapopulations dynamics, which ultimately may lead to changes in distribution.

Global Change Biology

The impact of global warming on biodiversity is likely to be more severe in regions rich in endemic species that are also predicted to be affected by dramatic climatic changes. The Mediterranean Basin is a particularly susceptible region: it is a biodiversity hotspot holding many endemic species (Médail et al., 1999, Myers et al., 2000, Mittermeier et al. 2005) and climate predictions for this region include a substantial rise in temperature and a drastic drop in rainfall, contributing to desertification (MIO-ECSDE, 2003). The synergistic effect of climate change with other threats to biodiversity makes this region one of the most vulnerable in the world (Sala et al., 2000).

The Iberian Peninsula, in particular, is a Mediterranean sub-region with many endemic species. The high biodiversity derives from the fact that this area was one of the major glacial refugia in Europe during the Pleistocene (Hewitt, 1996) and a diversification centre afterwards (Gómez *et al.*, 2006, Pinho *et al.*, 2007). Biodiversity richness in this area is also influenced by the climatic heterogeneity, since the climate transition between Atlantic and Mediterranean allows the co-occurrence of species with African and Euro-Siberian affinities, which means that the Iberian Peninsula is a biogeographic crossroad (Spector, 2002).

In face of climate change challenges, conservation organizations are being asked to take proactive measures to mitigate impacts on biodiversity. To select appropriate measures we need to be able to predict the impact of climate change on biodiversity and evaluate the ability of biodiversity to adapt to those impacts. The combination of this information can be used to determine biodiversity vulnerability, which in turn becomes the basis for prioritizing species and defining management strategies (Kareiva *et al.*, 2008).

90 Species distribution models (SDMs) are frequently used to assess the impacts of climate 91 change on species distributions (*e.g.* Araújo *et al.*, 2006). These statistical tools relate present 92 day distributions with current environmental conditions and then use future potential climate 93 conditions to predict future species distributions (Pearson *et al.*, 2003). Predicting the impacts

Global Change Biology

of climate change on species is a challenging task because SDMs include parameters with many sources of uncertainty (Webster et al., 2002) mostly related to: a) the statistical tool used for modelling species distributions; b) the global circulation models used to predict future climate conditions (Thuiller, 2004, Pearson *et al.*, 2006, Beaumont *et al.*, 2008); and c) uncertainty derived from scale effects (Seo et al., 2009). Recently, advances in SDMs have made it possible to significantly reduce prediction uncertainties. For example, several robust statistical modelling methods have been developed to predict species distributions (see Elith et al., 2006 for review), while advanced methodologies in ensemble forecasting allow us to overcome the problem of variability in predictions made by different modelling techniques or different global circulation models (Pearson et al., 2006, Araújo et al., 2007, Marmion et al., 2009). Additionally, climate predictions have been improved at smaller spatial scales (Hijmans et al., 2005).

In this study, we will evaluate potential impacts of climate change on the distribution of amphibians and reptiles in the Iberian Peninsula. We focus on endemic and nearly endemic species, the later defined as species that have aproximatelly more than two thirds of their entire range in the Iberian Peninsula. We then use information about species adaptability and vulnerability to make conservation recommendations.

Araújo et al. (2006) identified the Iberian Peninsula as one of the areas in Europe where amphibians and reptiles are likely to undergo major contractions in their ranges. Therefore, a detailed analysis of the Iberian Peninsula at a scale appropriate for establishing management strategies is urgently needed. We complement Araújo et al.'s (2006) European-wide research by providing a more detailed analysis of the impact of climate change on amphibian and reptile's distribution within the Iberian Peninsula. In particular, we used the most recent distribution data which are geographically more detailed than the one used in Araújo et al. (2006) (10x10 km in opposition to 50x50km) and incorporate recent taxonomic discoveries,

Global Change Biology

that significantly increases the number of endemics (21 species vs. 12 species). Additionally,
we project species ranges to three times in the future (2020, 2050, 2080 instead of 2050
only), which may assist conservation decision making in terms of prioritizing the allocation
of conservation funds through time.

We provide specific recommendations to conservation practitioners for enhancing the probability of species persistence by answering the following questions: In the Iberian Peninsula, which endemic and nearly endemic amphibian and reptile species are predicted to lose and gain suitable habitat in the future? For species predicted to lose suitable habitat, is the loss rate constant throughout time or are there more critical periods? Which areas will be more impacted by species loss? Which species should be under priority conservation action? Which conservation measures are most likely to increase the probability of species persistence?

132 Methods

133 Study area

The study region is the continental Iberian Peninsula, situated in the extreme southwest of Europe (bounded by $9^{\circ}32$ ' to $3^{\circ}20$ 'E and $35^{\circ}56$ ' to $43^{\circ}55$ 'N). With an area of 582 860 km², it includes the continental territories of Portugal and Spain. It is bordered to the south and east by the Mediterranean Sea and to the north and west by the Atlantic Ocean. The Pyrenees and the Strait of Gibraltar separate most of the region from the remainder of Europe and Africa, respectively (Figure 1). The dominant climate type of the region is Mediterranean, but the north and northwest of the Iberian Peninsula and the major mountain systems are characterised by an Atlantic climate.

143 Species and distribution data

Global Change Biology

Distribution data for 37 endemic and nearly endemic species (15 amphibians and 22 reptiles) were collected from the most recent herpetological atlases of Portugal (Loureiro et al., 2008) and Spain (Pleguezuelos et al., 2002), which are referenced to the UTM grid of 10x10 km. Taxonomy was defined according to the most recent revision of the taxonomic list in Carretero et al. (2009). We excluded from our analysis records from the Portuguese atlas for Triturus marmoratus because that data does not distinguish between Triturus marmoratus and T. pygmaeus. We followed the same method for Portuguese records of Pelodytes punctatus because they represent two species: Pelodytes punctatus and P. ibericus. Our data contained only two records of *Iberolacerta martinezricai* so we were unable to develop a plausible predictive model.

Climate data

Current bioclimatic data were downloaded from WorldClim database (Hijmans et al., 2005) which is a set of global climate layers generated through interpolation of climate data from weather stations. We used nine variables that were not tightly correlated with each other (with a Pearson correlation coefficient between them lower than 0.75): temperature seasonality, maximum temperature of warmest month, minimum temperature of coldest month, temperature annual range, annual precipitation, precipitation of wettest month, precipitation of driest month, precipitation seasonality and altitude. All variables were downloaded in a 2.5 arc-minute resolution. We converted these data to match with the same grid format as the species distribution data by averaging the variable's values inside each grid cell.

We used future climate data for three Global Circulation Models (GCM) (CCCMA,
HADCM3 and CSIRO) and two IPPC 3rd assessment storylines (A2 and B2). The IPCC
storylines describe the relationships between the forces driving greenhouse gas and aerosol

Global Change Biology

emissions such as demographic, social, economic, technological, and environmental developments (IPCC-TGICA, 2007). The two storylines used for this study assume regionally oriented economic growth, with population and economic growth being higher in A2 than B2. Future climate data were downloaded from WorldClim, for three different future years (2020, 2050 and 2080) creating six storyline-GCM combinations for each future year. We downloaded monthly averages of maximum and minimum temperatures and total precipitation and calculated the bioclimatic variables according to the same methodology used to calculate them for current climate conditions (Hijmans et al., 2005).

178 Species distributions: current predictions and future projections

We used a set of nine modelling techniques to predict the distribution of each of the 37
species in four different times (current, 2020, 2050 and 2080): Maximum Entropy (MXT),
Generalised Linear Models (GLM), Generalised Additive Models (GAM), Classification
Tree Analysis (CTE), Artificial Neural Networks (ANN), Generalised Boosting Model
(GBM), Breiman and Cutler's random forest for classification and regression (RF), Mixture
Discriminant Analysis (MDS) and Multiple Adaptive Regression Splines (MARS).

To produce Maximum Entropy models we used Maxent software (Phillips *et al.*, 2004). Maxent estimates the range of a species with the constraint that the expected value of each variable (or its transform and/or interactions) should match its empirical average, *i.e.* the average value for a set of sample points taken from the species-target distribution (Phillips *et al.*, 2006, Phillips *et al.*, 2008). We used the default "auto features" option, logistic output, the recommended default values for the convergence threshold (10⁻⁵) and the maximum number of iterations (500).

 The remaining eight models were built within BIOMOD (Thuiller et al., 2009). BIOMOD is a collection of functions running within the R software v. 2.8.1 (R Development Core Team 2008) for ensemble forecasting of species distributions and a summary of model statistics is described in Thuiller et al., (2009). The majority of model-techniques requires data about presences and absences, thus we determined pseudo-absences for each species by using the "random strategy" in BIOMOD. The number of selected pseudo-absences for each species was equal to the number of its occurrences, whenever possible. In cases where the number of locations where species was not recorded was less than the number of occurrences, we selected all non-occurrences as pseudo-absences. All models were produced using default BIOMOD parameters where possible (Thuiller et al., 2009). Further parameters were as following: GLMs were generated using quadratic terms and a stepwise procedure with the AIC criteria. GAMs were generated with a spline function with 4 degrees of smoothing. ANNs were produced with two cross-validations. BIOMOD allows evaluation of model performance on different data split runs and then allows using 100% of the data to make a final calibration of the models for prediction. Thus we randomly assigned 80% of occurrence data to train the model with the remaining 20% for testing. Each model was run ten times to avoid bias resulting from randomly splitting the data into training and testing.

To evaluate the performance of individual models from each technique, we calculated the area under the receiver operating characteristic curve (AUC) (Zweig et al., 1993, Fielding et al., 1997). For each of the 37 species we produced nine models for the current time and climate (with different modelling techniques) and 54 models (combination of nine modelling techniques, three GCM and two storylines) for each of the three times in the future (2020, 2050 and 2080), in a total of 171 models per species.

For each year in the future (2020, 2050 and 2080) we calculated an ensemble forecast for
current time and for each IPCC scenario for the future (A2, and B2). For this purpose, we

used weighted average consensus method based on AUC values, because this method is considered to significantly improve the predictive accuracy of single models (Marmion *et al.*, 2009). Thus, AUCs values of each model technique were assigned the weights of the weighted average in order to enhance contributions of those models with higher model performance values (equation 1).

$$WA_{i} = \frac{\sum_{m} (AUC_{m} \times p_{mi})}{\sum_{m} AUC_{m}}$$
(equation 1)

where *i* is the index of the grid cell, *m* is the model technique and *p* is the probability of occurrence of the species (according to model *m* in grid cell *i*).

To investigate if species ranges were predicted to contract or expand, we needed to convert the consensus forecasted probabilities of occurrence in each year to a binary value of predicted presence/absence. For each model technique, we calculated the threshold of the receiver operating characteristic curve that maximizes both correctly predicted presences and absences, relative to the evaluation data (Liu *et al.*, 2005). Subsequently, we determined a consensus threshold (CT) by calculating the weighted average threshold, assigning the AUC value of each model as a weight (equation 2).

$$CT = \frac{\sum_{m} (AUC_{m} \times t_{m})}{\sum_{m} AUC_{m}}$$
(equation 2)

235 where *m* is the model technique and t_m is the optimised threshold of model *m*.

Potential range shifts of each species and time period were measured under two extremedispersion scenarios: unlimited dispersal (scenario D), assuming that species can disperse to

any grid cell with suitable habitat, and no dispersal (scenario ND), assuming that species are not capable to disperse even if suitable habitat is available. To predict species presence under the D scenario, we transformed the consensus probability value in each grid cell into 0 if it was lower than CT and to 1 if it was equal to or higher than CT. To predict species presences under the ND scenario, we transformed predicted presences to absences in grid cells where species is not presently predicted to occur. We evaluated the degree of uncertainty of projections for each year and scenario by calculating the standard deviation of predicted occurrence of species by grid cell.

Subsequently, we calculated predicted species richness in each grid cell for the three future periods. Future species richness was estimated by summing the number of species predicted to occur in each grid cell under each of the storylines and the dispersion scenarios. Species turnover (T) is the dissimilarity index between the present and future species composition of a given area. It accounts both for species gain and losses and its relation to the overall number of species occurring in the grid cell in present and future. Thus, a turnover value of 0 indicates that the predicted assemblage in the future would be the same as the current assemblage, whereas a turnover value of 100 indicates that the assemblage would be completely different under climate change. To determine it, we first calculated the number of species lost (L), *i.e.*, the number of species predicted to lose suitable habitat in each grid cell, and the species gain (G), *i.e.*, the number of species predicted to gain suitable habitat in each grid cell. Percentage of species turnover by grid cell was then calculated according to equation 3.

259
$$T = 100x \frac{L+G}{SR+G},$$
 (equation 3)

260 where SR is the current predicted species richness.

Results

Model predictions within different modelling techniques, GCMs, and storylines showed high variability in the projection of range shifts, with most species being projected to both lose and gain suitable habitat, depending on the scenario (supplementary material A.1 and A.2). However, when analysing the ensemble model results (Tables 1 and 2), we found that 46% of the species are consistently predicted to have a smaller distribution in both storylines (9 amphibian species and 8 reptile species), 28% of the species are consistently predicted to have a larger distribution (3 amphibian species and 8 reptile species) and the remaining species show increases and decreases depending in the time period and storyline. For example Algyroides marchi is predicted to have a smaller distribution until 2050 and then increase its distribution by 2080, while, Alytes cisternasii and Lissotriton boscai are predicted to have a larger distribution until 2020 and then have a smaller distribution.

Several species are predicted to lose a significant fraction of their current range (Figure 2) and 10 species are predicted to lose their entire suitable distribution at some time in both storylines. Three other species are predicted to lose all suitable range in one of the storylines only (Table 3). Surprisingly 5 to 10 of the 13 species that are predicted to lose all their suitable distribution lose it by 2020 (Table 3). Additionally, 9 of these species are also threatened by other causes, and currently considered "critically endangered", "endangered" or "vulnerable", in Portuguese or Spanish red data books (Tables 1 and 2).

If we assume that species have no ability to disperse, 34 species are predicted to contract their
distribution and 10 species are predicted to completely lose their entire distribution in both
storylines.

It is noteworthy that for species predicted to have a smaller distribution in the future, both the magnitude of contraction and the rate of contraction differ between species (Figure 3). While for most contracting species, the rate of range contraction is greater now (until 2020), for

some species there is almost a constant rate of predicted distribution loss (*e.g. Rana iberica*),
while for a minority of species the period of greatest contraction is later (*e.g Psammodromus algirus*).

Predicted patterns of species richness are substantially different between the two dispersion scenarios, but higher species richness was commonly predicted in south-western Iberian Peninsula, the Central System mountain range and Morena Mountains (Figures 4 and 5). Major species gain, in the unlimited dispersion scenario, was predicted for the central plateaus, the central-western coast of the Iberian Peninsula, and the Cantabrian Mountains (Figure.6). Major loss of species ranges is predicted for the Atlantic climate regions, mostly along the northwest of the Iberian Peninsula and the main mountain ranges, such as the Central System and the Morena Mountains (Figure. 6). Spatial patterns of predicted species richness are similar in both storylines, although storyline B2 predicted slightly higher species losses for 2020 and storyline A2 predicted higher species loss for 2080 (Figures 4 and 5).

301 Discussion

302 Main findings and relation with previous projections

Our results suggest that climate change might have serious impacts on the distribution patterns of the endemic and nearly endemic amphibians and reptiles of the Iberian Peninsula, particularly for species with Atlantic climate affinities such as Chioglossa lusitanica, Rana iberica and Vipera seoanei and particularly high altitude species with Atlantic climate affinities such as, Rana pyrenaica, Iberolacerta monticola, I. aranica, I. aurelioi, and I. *bonnali*. It is also possible that other species may be more affected by climate change than predicted by the models, such as Iberolacerta cyreni. This species is an endemic with a current range restricted to the Gredos Mountains in Spain. However, the ensemble model for

Global Change Biology

this species predicted a current range much larger than probably the one where the species actually occurs. This indicates that predicted future range may also be overestimated. On the other side, models for *Pelodytes ibericus, Triturus marmoratus and T. pygmaeus* were built only based on Spanish records, although they also occur in Portugal. This methodological caveat may have conduced to an under estimation of current and future ranges of those species.

317 Our analysis indicates that the most critical period for the majority of these species will be the 318 next decade. This result emphasises the need for immediate conservation action in the Iberian 319 Peninsula to ameliorate the impact of climate change.

Previous macroscale studies have called attention to the vulnerability of biodiversity in the Mediterranean basin. This region was considered one of the most vulnerable in the world (Sala et al., 2000) and one of the biodiversity hotspots likely to undergo major losses due to climate change, along with the Cape Floristic Region, the Caribbean, Indo-Burma, Southwest Australia, and the Tropical Andes (Malcolm et al., 2006). The Iberian Peninsula is considered the Mediterranean sub-region most likely to be affected by future climate change (EEA, 2004). Our analysis confirm the vulnerability of species occurring in the Iberian Peninsula to climate change, but curiously, species with Atlantic climate affinities were predicted to be more affected by climate change then the Mediterranean ones. The Atlantic climate region is much more restricted in the Iberian Peninsula then the Mediterranean. Due to predicted drops in precipitation and temperature raise it is likely that the all Iberia became dominated by a Mediterranean climate. Thus, it seems logical that Atlantic species become more affected by climate change then the Mediterranean ones. Climate change impacts on Atlantic species was exacerbated by the region being a peninsula so dispersal out of the region to other Atlantic climate regions is restricted. Similar patterns should be found in other Mediterranean Peninsulas that present different climatic types (*e.g.* the Italian Peninsula and the Balkans).

Our results are consistent with those found by Araújo et al. (2006), in the sense that climate change is predicted to cause a major contraction in the distribution of a considerable number of amphibian and reptile species in the Iberian Peninsula. However, our results revealed that major losses for endemic and nearly endemic species will occur in substantially different areas than the ones predicted by Araújo et al. (2006) for the all amphibians and reptiles. We predict that the north-west of the region, the Central System and the Morena Mountains will lose many species in contrast to their results where there is expected to be species gain. The reasons for this difference may be our focus on endemic and nearly endemic species rather than all species, but it could also be differences in spatial scale or methods used to construct the ensemble models. Dissimilarities in areas predicted to gain species may also be related to the colonization of non endemic or nearly endemic species and non Iberian species not included in our analysis. These divergent results highlight the importance of finer-scale analyses for areas identified as vulnerable in broader continental-scale studies.

Interestingly, a previous study of climate change threats to plant diversity in Europe (Thuiller et al., 2005) also identified the Iberian Peninsula as one of the European regions likely to undergo major species loss. Garzón et al. (2008) identified the north and northwest of the Iberian Peninsula and the main mountain ranges, as the Iberian areas likely to have the highest tree species loss, which is partially coincident with our results. Thus, if changes in vegetation communities are also predicted, it means that amphibians and reptiles will face a synergistic impact of climate and habitat change, which ultimately will also be coupled with the impact of changes in biotic interactions resulting from changes in the community of species. These results also suggest that the impacts of climate change on species ranges might be similar across different biodiversity groups that occur in the same area, which calls for the need to evaluate possible climate change impacts on those groups.

362

1

Global Change Biology

2	
2 3	
4	
5	
6	
6 7	
8	
9	
9 4 0	
10 11 12	
11	
12	
13	
14	
15	
16	
17	
18	
19	
20 21 22 23 24 25 27 28 20 31 32 33 34 35 36 37 38 39	
21	
22	
23	
24	
25	
20	
20	
21	
28	
29	
30	
31	
32	
33	
34	
35	
36	
37	
38	
39	
40	
41	
42	
43	
43 44	
44 45	
46 47	
48	
49	
50	
51	
52	
53	
54	
55	
56	
57	
58	
59	
60	

Our dramatic predictions should be interpreted with caution given all the uncertainties in the process: the statistical methods used for modelling species distributions, the global circulation models used to predict future climate conditions (*e.g.* Araújo *et al.*, 2007, Beaumont *et al.*, 2008), the scale of the analysis (Seo *et al.*, 2009), and specific dispersal abilities. Additionally species distribution models disregard important biological parameters that ultimately are the determinants of species capability to adapt to climate change.

From predicted impacts (models) to vulnerability

369 Different modelling techniques, GCM and storylines provided fairly different results for the 370 number of predicted occurrences of a species. The ensemble methodology allows us to 371 distinguish the strongest signal emerging from the noise associated with different model 372 outputs but it is not immune from uncertainty. Taking uncertainty into account, for example 373 by using as an uncertainty measure the standard deviation of the total number of occurrences 374 predicted by the different models for each species (appendix 1A and A2), the most likely 375 impact of climate change will be on species whose ranges are predicted to contract more 376 (relatively to their current extent), or to become more restricted (considering their future 377 absolute extent) with less uncertainty. For example, analysing only the year 2020 projections 378 for storyline A2, species that fulfil both criteria are *Discoglossus galganoi* (considering the 379 species whose ranges are predicted to contact more) and Iberolacerta aranica, (considering 380 the species whose ranges are predicted to become more restricted or disappear).

The scale of analysis may also be a relevant limitation in determining the impact of climate change, because the scale used does not account for microhabitat variability within sites. This constraint is quite relevant in the case of amphibians and reptiles because they might find suitable habitat, for instance, in small ponds, water tanks, below stones or underground (Kearney *et al.*, 2009). Also, the scale of the analysis probably obscures altitudinal 386 microclimatic gradients which might allow species to move towards future suitable habitats387 without the need to disperse long distances.

 Specific dispersal abilities may strongly determine the impacts of climate change on the future distribution of species. However, given that this parameter is difficult to determine for every species, we predicted future distributions under the assumptions of maximal and minimal possible dispersal ability for each species. We recognize that both of these assumptions are unrealistic, but they allow us to predict the largest and smallest possible future ranges. A more realistic prediction would probably be somewhere in the middle of these two extremes, but we cannot make an exact prediction with high level of certainty.

Assuming these limitations we do not argue that species predicted to completely lose their distribution under our analysis will go extinct. Rather, we consider that the degree, the certainty and the time of predicted range contractions provides a relative measure of the magnitude of the impact of climate change on each species. Ultimately, the vulnerability of a species to a given impact will depend also on the species resistance, *i.e.* the ability of a species to withstand an environmental perturbation, and resilience, *i.e.* the ability of a species to adapt and recover from a perturbation (Isaac et al., 2008, Williams et al., 2008,). Thus, prioritizing species for management actions would require a further analysis of the specific factors that determine resistance and resilience. In general, the species traits that are predicted to promote species resistance to climate change are physiologic climatic tolerance (Calosi et al., 2008), geographical range size and local abundance.

Overall, reptiles are usually considered more resistant to global warming effects than amphibians because they have evolved a set of adaptations to water scarcity, such as eggs with calcareous shells (while amphibian eggs are enclosed by simple gelatinous membranes), and the excretion of metabolic wastes in the form of urea or uric acid, which is an adaptation to retain body-water (Gibbons et al., 2000). Additionally, most reptiles are not dependent on

Global Change Biology

411 water for reproduction. Although more specific climatic tolerances might be difficult to 412 determine, they may be inferred by morphological traits or by the complete climatic envelope 413 of species. Also, the climatic envelope of phylogenetic related species might provide some 414 clues, if one considers niche conservatism (Hawkins *et al.*, 2007). For example, species 415 belonging to clades that evolved under warm environments may be more tolerant to global 416 warming.

417 Species with restricted ranges are relatively more vulnerable to climate change because 418 contractions of their small range will not be balanced elsewhere. From this point of view, 419 species with very restricted distributions predicted to completely lose suitable habitat in the 420 Iberian Peninsula such as *Iberolacerta aranica*, *I. aurelioi*, and *Rana pyrenaica* may be more 421 impacted by climate change than more widespread Iberian species such as *Lissotriton boscai* 422 or *Chalcides bedriagai*.

Demographic parameters are also important factors in determining resistance to climate change (Keith *et al.*, 2008). Species with restricted ranges but with high local abundance may have more chances to adapt because genetic variation and potential response to selection pressures are positively correlated with population size. Thus, biological traits that regulate abundance, such as reproductive rates, age of female sexual maturity and life span length are also relevant. Discoglossus galganoi and Podarcis bocagei are species predicted to completely lose their range in the Iberian Peninsula but have high abundance and reproductive rate, and consequently will probably be less impacted by climate change than the others will.

432 The adaptive capacity (or plasticity) of a species describes the intrinsic ability of a species to
433 adapt to changing conditions. Species might be able to adjust their behaviour and phenology
434 by switching periods of daily activity, aestivation and hibernation towards more favourable
435 climatic conditions (Parmesan, 2007). Species might also be able to evolve traits that allow

them to adapt to different climatic conditions (Harte *et al.*, 2004, Bradshaw *et al.*, 2006).
Adaptation ability may be species-specific, however, the current knowledge is insufficient to
determine which species are more able to adapt and further studies and monitoring are
required to fully understand it.

Species might also adapt to novel climate conditions by dispersing to other areas (Thuiller, 2004, Massot et al., 2008). Dispersion ability has been identified as one of the most decisive parameters in determining species resilience to climate change as species with greater dispersion ability may be able to track climate transitions. However this parameter is also one of the most difficult to determine. A growing body of literature proposes a vast collection of complex models to predict dispersion ability (reviewed by Thuiller et al., 2008) but these require detailed data on an array of ecological processes that usually are unavailable for large numbers of species.

A preliminary analysis of the model results and previous considerations indicate that species requiring more conservation attention under a climate warming perspective may be *Rana* pyrenaica, Rana iberica and Calotriton asper, within the amphibians, and Iberolacerta aranica, Iberolacerta aurelioi, Iberolacerta bonnali and Iberolacerta monticola within the reptiles. However, this does not mean that these are the most vulnerable species to extinction, because this evaluation is only based on vulnerability to climate change, and does not account for other threats or the interaction of climate change with other threats (Brook et al., 2008). Particularly, climate change is likely to induce further habitat changes and fragmentation (due to shifts in plant species distribution and an increase in fire incidence) and the spread of agents of infectious diseases such as the chytridiomycete fungus which is already across the Iberian Peninsula (Garner et al., 2005). These factors are already considered the primary threats to amphibians and reptiles, along with pollution, invasive species, road kills and genetic depression.

2		
3 4 5	461	
2 3 4 5 6 7	462	From vulnerability to conservation actions
8 9 10	463	Recommendations to address climate change impacts on biodiversity include a wide variety
11 12	464	of measures, with the most popular being monitoring species (with emphasis on the
13 14	465	physiological, behavioural and demographic response), restoring habitats and system
15 16 17	466	dynamics, expanding reserve networks, performing assisted dispersal Hoegh-Guldberg et al.
18 19	467	(2009), reducing other threats and increasing connectivity between suitable habitats (Heller et
20 21	468	al., 2009, Lawler, 2009). Probably the most important questions for management are whether
22 23 24	469	species will be able to adapt to future climate conditions without the need to disperse or, if
24 25 26	470	not, if they will be able to disperse. Because these questions are difficult to answer with
27 28	471	current knowledge, novel management tools that promote flexible decision-making are
29 30 31	472	emerging, such as adaptative management (Kareiva et al., 2008). Following this
31 32 33	473	methodology, we recommend that monitoring should be directed at the most vulnerable
34 35	474	species identified by our research. To infer species-specific conservation measures, the
36 37	475	monitoring parameters should include: a) physiological changes in thermal tolerances; b)
38 39 40	476	phenological adjustments, such as changes in aestivation and hibernation periods along the
40 41 42	477	year; c) behavioural thermoregulation changes, such as burrowing or adjustments in daily
43 44	478	activity periods; d) quantification of dispersal rates; e) changes in population parameters,
45 46		
47 48 49	479	such as abundance, fertility and mortality f) incidence of infectious diseases and g) species
50	480	interactions.
51 52	481	If species are not able to disperse, then management measures will be needed in order to

If species are not able to disperse, then management measures will be needed in order to assist local adaptation. Habitat restoration has been proposed as a proactive measure to enhance amphibian resilience to climate change. In particular, a denser network of ponds and water tanks has been recommended for amphibians (Blaustein et al., 2001) in Mediterranean areas. This measure, along with a careful control of river flow and water quality may increase

Global Change Biology

the probability of amphibians to find suitable habitats during climate warming. Availability of freshwater habitats will be particularly important in the regions predicted to have high species loss, the northern Iberian Peninsula and main mountain systems, but also in extreme southern regions (Beja *et al.*, 2003).

Assisted colonisation has been suggested, as a measure to assist climate change adaptation (Mueller et al., 2008; Cheddadi et al., 2009). This triggered intense debate (McLachlan et al., 2007) because translocation of species originated catastrophic impacts in many existing cases. However, assisted dispersal is particularly relevant for amphibians and reptiles due to their low dispersal ability, and therefore we cannot disregard this management possibility under an extreme probability of extinction scenario. However, we agree with (Hoegh-Guldberg et al., 2008) in that assisted dispersal should only be considered for a given species if at least one of a set of assumptions is met, namely: a) the species should be in immediate risk of extinction; b) species should have low dispersal ability and c) the species range should be highly fragmented. Moreover, those authors suggest that translocations should only be undertaken within biogeographic regions, *i.e.*, regions that share similar species composition, and an assessment of translocation risks should be performed previously, including ecological and socio-economical risks. Translocation risk should subsequently be balanced against those of extinction and safeguarded by detailed scientific understanding.

Protected areas have long been considered one of the most effective tools to conserve biodiversity (Rodrigues *et al.*, 2004), but their effectiveness in securing species under rapid climate change is uncertain (Araújo *et al.*, 2004). However they contribute to minimising threats such as, habitat destruction and fragmentation, road kills and pollution, which ultimately may assist to protect species threatened by climate warming. Therefore, it is important to evaluate if present protected areas would be effective in securing species given their predicted range shifts and whereas there is habitat connectivity between current and

Global Change Biology

future species distributions. This could be achieved using reserve selection algorithms (*e.g.* Moilanen, 2009), which allow the identification the minimum set of areas necessary to represent all species at a given target. Ultimately, the ability of species to adapt to climate change within nature reserves will depend on the management actions undertaken in each of them. Reserve selection algorithms may also assist in identifying areas within nature reserves where the return of the management investment will be greater for a higher number of species.

518 If species are able to disperse to new habitats, than management actions will be needed to 519 facilitate dispersal, including the increase of habitat connectivity, in particular of freshwater 520 habitats.

Finally, biodiversity conservation can only be successful under climate change scenarios if socio-economical and environmental policies are integrated and if governments cooperate. Although climate change is a global issue, there is a tendency for actions to be taken by governments individually or by local administration, even with a common European environmental policy. It is recommended that the Portuguese and the Spanish governments embark on joint efforts to conserve Iberian biodiversity, particularly Iberian endemics.

528 Acknowledgements

529 SC was supported by a PhD grant (SFRH/BD/21896/2005) and JCB has a contract

530 (Programme Ciência 2007), both from Fundação para a Ciência e Tecnologia. HPP was

531 supported by Australian Research Council grants and an Australian Federal Government

532 Commonwealth Environmental Research Facility grant. We would like to thank Doctor

3 533 Wilfried Thuiller for assisting with BIOMOD procedures and Madeleine C. Bottrill and the

anonymous reviewers for useful comments on early versions of this manuscript.

1		
2 3	535	References
4	555	Kelefences
5 6 7	536	
8 9 10 11 12 13 14	537	Araújo MB, Cabeza M, Moilanen AM, Thuiller W, Hannah L, Williams PH (2004) Would
	538	climate change drive species out of reserves? An assessment of existing reserve-
	539	selection methods. Global Change Biology, 10, 1618-1626.
15 16 17	540	Araújo MB, New M (2007) Ensemble forecasting of species distributions. Trends in Ecology
18 19	541	and Evolution, 22 , 42-47.
20 21 22	542	Araújo MB, Thuiller W, Pearson RG (2006) Climate warming and the decline of amphibians
23 24	543	and reptiles in Europe. <i>Journal of Biogeography</i> , 33 , 1712 - 1728.
25 26	544	Beaumont LJ, Hughes L, Pitman AJ (2008) Why is the choice of future climate scenarios for
27 28 29	545	species distribution modelling important? <i>Ecology Letters</i> , 11 , 1135 - 1146.
30 31	546	Beja P, Alcazar R (2003) Conservation of Mediterranean temporary ponds under agricultural
32 33 34	547	intensification: an evaluation using amphibians. <i>Biological Conservation</i> , 114 , 317-
35 36	548	326.
37 38	549	Blaustein AR, Belden LK, Olson DH, Green DM, Root TL, Kiesecker JM (2001) Amphibian
39 40 41	550	breeding and climate change. <i>Conservation Biology</i> , 15 , 1804-1809.
42 43	551	Bosch J, Carrascal LM, Durán L, Walker S, Fisher MC (2007) Climate change and outbreaks
44 45 46 47 48 49 50 51 52 53	552	of amphibian chytridiomycosis in a montane area of Central Spain; is there a link?
	553	Proceedings of the Royal Society B: Biological Sciences, 274, 253-260.
	554	Bradshaw WE, Holzapfel CM (2006) Evolutionary Response to Rapid Climate Change.
	555	<i>Science</i> , 312 , 477-1478.
53 54 55	556	Brook BW, Sodhi NS, Bradshaw CJA (2008) Synergies among extinction drivers under
56 57	557	global change. Trends in Ecology & Evolution, 23, 453-460.
58 59 60	558	Calosi P, Bilton DT, Spicer JI (2008) Thermal tolerance, acclimatory capacity and
00	559	vulnerability to global climate change. <i>Biology Letters</i> , 4 , 99-102.

1		
2 3 4	560	Carey C, Alexander MA (2003) Climate change and amphibian declines: is there a link?
5 6	561	Diversity and Distributions, 9, 111 - 121.
7 8 9	562	Cheddadi R, Fady B, François L, Hajar L, Suc J-P, Huang K, Demarteau M, Vendramin GG,
10 11 12 13 14	563	Ortu E (2009) Putative glacial refugia of Cedrus atlantica deduced from Quaternary
	564	pollen records and modern genetic diversity. Journal of Biogeography, 36, 1361 -
15 16	565	1371.
17 18	566	Chadwick EA, Slater FM, Ormerod SJ (2006) Inter- and intraspecific differences in
19 20 21	567	climatically mediated phenological change in coexisting Triturus species. Global
22 23	568	<i>Change Biology</i> , 12 , 1069 - 1078.
24 25 26	569	Carretero, M.A., Ayllón, E. & Llorente, G. (eds) (2009) Lista patrón de los anfibios y reptiles
26 27 28 29 30 31 32 33	570	de España. Asociación Herpetológica Española, Barcelona.
	571	European Environmental Agency (2004) Impacts of Europe's changing climate - An
	572	indicator-based assessment. European Environmental Agency, Luxembourg: Office
34 35	573	for Official Publications of the European Communities, 107pp.
36 37	574	Elith J, Graham CH, Anderson RP, et al. (2006) Novel methods improve prediction of
38 39 40	575	species' distributions from occurrence data. <i>Ecography</i> , 29 , 129-151.
41 42	576	Fielding AH, Bell JF (1997) A review of methods for the assessment of prediction errors in
43 44	577	conservation presence/absence models. Environmental Conservation, 24, 38-49.
45 46 47 48 49 50 51	578	Garner TWJ, Walker S, Bosch J, Hyatt AD, Cunningham AA, Fisher MC (2005) Widespread
	579	European distribution of a global amphibian pathogen. Emerging Infenctious diseases,
	580	11 , 1639–1641.
52 53 54	581	Garzón MB, Dios RS, Ollero HS (2008) Effects of climate change on the distribution of
55 56	582	Iberian tree species. Applied Vegetation Science, 11, 169–178.
57 58	583	Gibbons JW, Scott DE, Ryan TJ, et al. (2000) The Global Decline of Reptiles, Déjà Vu
59 60	584	Amphibians. BioScience, 50, 653-666.

1 2		
3 4	585	Gómez A, Lunt DH (2006) Refugia within refugia: patterns of phylogeographic concordance
5 6	586	in the Iberian Peninsula. In Phylogeography of Southern European Refugia (eds
7 8 9	587	Weiss S, Ferrand N), pp. 155-188. Springer, Netherlands.
10 11 12 13 14	588	Harte J, Ostling A, Green JL, Kinzig A (2004) Biodiversity conservation: Climate change and
	589	extinction risk. <i>Nature</i> , 430 .
15 16	590	Hawkins BA, Diniz-Filho JAF, Jaramillo CA, Soeller SA (2007) Climate, Niche
17 18	591	Conservatism, and the Global Bird Diversity Gradient. <i>The American naturalist</i> , 170 ,
19 20 21	592	\$16–\$27.
22 23	593	Heller NE, Zavaleta ES (2009) Biodiversity management in the face of climate change: A
24 25	594	review of 22 years of recommendations. Biological Conservation, 142, 14-32.
26 27 28	595	Hewitt GM (1996) Some genetic consequences of ice ages, and their role in divergence and
29 30 31 32 33 34 35 36 37	596	speciation. Biological Journal of the Linnean Society, 58, 247–276.
	597	Hijmans RJ, Cameron SE, Parra JL, Jones PG, Jarvis A (2005) Very high resolution
	598	interpolated climate surfaces for global land areas. International Journal of
	599	<i>Climatology</i> , 25 , 1965-1978.
38 39 40	600	Hoegh-Guldberg O, Hughes L, McIntyre S, Lindenmayer DB, Parmesan C, Possingham HP,
40 41 42	601	Thomas CD (2008) Assisted Colonization and Rapid Climate Change. Science, 321,
43 44	602	345 - 346.
45 46 47	603	Hughes L (2000) Biological consequences of global warming: is the signal already apparent?
48 49 50 51	604	Trends in Ecology & Evolution, 15, 56-61.
	605	IPCC (2007) Climate Change 2007: Synthesis Report. Contribution of Working Groups I, II
52 53 54	606	and III to the Fourth Assessment Report of the Intergovernmental Panel on Climate
55 56	607	Change, Switzerland, 104pp.
57 58 50	608	IPCC-TGICA (2007) General Guidelines on the Use of Scenario Data for Climate Impact and
59 60	609	Adaptation Assessment. Version 2. Prepared by T.R. Carter on behalf of the

Page 27 of 42

1 2		
3 4	610	Intergovernmental Panel on Climate Change, Task Group on Data and Scenario
5 6	611	Support for Impact and Climate Assessment.
7 8 9	612	Isaac JL, Vanderwal J, Williams SE, Johnson CN (2008) Resistance and resilience:
10 11 12 13 14 15 16 17 18	613	quantifying relative extinction risk in a diverse assemblage of Australian tropical
	614	rainforest vertebrates. Diversity and Distributions, 15, 280-288.
	615	Kareiva P, C., Enquist C, Johnson A, et al. (2008) Synthesis and Conclusions. In Preliminary
	616	review of adaptation options for climate-sensitive ecosystems and resources. Report
19 20 21	617	by the U.S. Climate Change Science Program and the Subcommittee on Global
22 23	618	Change Research. (eds Julius SH, J.M. West), pp. 66. U.S. Environmental Protection
24 25	619	Agency. Washington, DC.
26 27 28	620	Kearney M, Porter W (2009) Mechanistic niche modelling: combining physiological and
29 30	621	spatial data to predict species' ranges. Ecology Letters, 2009, 334-350.
31 32	622	Keith DA, Akçakaya HR, Thuiller W, et al. (2008) Predicting extinction risks under climate
33 34 35	623	change: coupling stochastic population models with dynamic bioclimatic habitat
36 37	624	models. Biology Letters, 4, 560-563.
38 39 40	625	Lawler JJ (2009) Climate Change Adaptation Strategies for Resource Management and
40 41 42	626	Conservation Planning. Annals of the New York Academy of Sciences, 1162, 79-98.
43 44	627	Liu C, Berry PM, Dawson TP, Pearson RG (2005) Selecting thresholds of occurrence in the
45 46 47	628	prediction of species distributions. <i>Ecography</i> , 28 , 385-393.
48 49	629	Loureiro A, Ferrand de Almeida N, Carretero MA, Paulo OS (Eds.) (2008) Atlas dos Anfíbios
50 51 52 53 54 55 56	630	e Répteis de Portugal, Instituto da Conservação da Natureza e da Biodiversidade,
	631	Lisboa.
	632	Malcolm JR, Liu C, Neilson RP, Hansen L, Hannah L (2006) Global warming and
57 58	633	extinctions of endemic species from biodiversity hotspots. Conservation Biology, 20,
59 60	634	538-548.

1 2		
3 4 5 6 7 8 9	635	Marmion M, Parviainen M, Luoto M, Heikkinen RK, Thuiller W (2009) Evaluation of
	636	consensus methods in predictive species distribution modelling. Diversity and
	637	Distributions, 15, 59-69.
9 10 11	638	Massot M, Clobert J, Ferrière R (2008) Climate warming, dispersal inhibition and extinction
12 13	639	risk. Global Change Biology, 14, 461-469.
14 15 16	640	McCarty JP (2002) Ecological Consequences of Recent Climate Change. Conservation
17 18	641	<i>Biology</i> , 15 , 320 - 331.
19 20	642	McLachlan JS, Hellmann JJ, Schwartz MW (2007) A framework for debate of assisted
21 22 23	643	migration in an era of climate change. Conservation Biology, 21, 297-302.
23 24 25 26 27 28 29 30 31 32 33 34 25	644	Médail F, Quézel P (1999) Biodiversity hotspots in the Mediterranean Basin: Setting Global
	645	Conservation Priorities. Conservation Biology, 6, 1510 - 1513.
	646	MIO-ECSDE (2003) Climate change and the Mediterranean. Sustainable Mediterranean, 32,
	647	1-24.
	648	Mittermeier RA, Gil PR, Hoffman M, et al. (2005) Hotspots Revisited: Earth's Biologically
35 36	649	Richest and Most Endangered Terrestrial Ecoregions. University of Chicago Press,
37 38	650	392 pp.
39 40	651	
40 41		Moilanen A, Wilson KA, Possingham H (Eds.) (2009) Spatial Conservation Prioritization.
	652	Moilanen A, Wilson KA, Possingham H (Eds.) (2009) Spatial Conservation Prioritization. Quantitative Methods and Computational Tools Oxford University Press.
42 43	652 653	
42		Quantitative Methods and Computational Tools Oxford University Press.
42 43 44 45 46 47 48	653	<i>Quantitative Methods and Computational Tools</i> Oxford University Press. Mueller JM, Hellmann JJ (2008) An Assessment of Invasion Risk from Assisted Migration.
42 43 44 45 46 47 48 49 50	653 654	<i>Quantitative Methods and Computational Tools</i> Oxford University Press. Mueller JM, Hellmann JJ (2008) An Assessment of Invasion Risk from Assisted Migration. <i>Conservation Biology</i> , 22 , 562 - 567.
42 43 44 45 46 47 48 49	653 654 655	 Quantitative Methods and Computational Tools Oxford University Press. Mueller JM, Hellmann JJ (2008) An Assessment of Invasion Risk from Assisted Migration. Conservation Biology, 22, 562 - 567. Myers N, Mittermeler RA, Mittermeler CG, da Fonseca GAB, Kent J (2000) Biodiversity
42 43 44 45 46 47 48 49 50 51 52 53 54 55	653 654 655 656	 <i>Quantitative Methods and Computational Tools</i> Oxford University Press. Mueller JM, Hellmann JJ (2008) An Assessment of Invasion Risk from Assisted Migration. <i>Conservation Biology</i>, 22, 562 - 567. Myers N, Mittermeler RA, Mittermeler CG, da Fonseca GAB, Kent J (2000) Biodiversity hotspots for conservation priorities. <i>Nature</i>, 403, 853-858.
42 43 44 45 46 47 48 49 50 51 52 53 54	 653 654 655 656 657 	 <i>Quantitative Methods and Computational Tools</i> Oxford University Press. Mueller JM, Hellmann JJ (2008) An Assessment of Invasion Risk from Assisted Migration. <i>Conservation Biology</i>, 22, 562 - 567. Myers N, Mittermeler RA, Mittermeler CG, da Fonseca GAB, Kent J (2000) Biodiversity hotspots for conservation priorities. <i>Nature</i>, 403, 853-858. Parmesan C (2006) Ecological and Evolutionary Responses to Recent Climate Change.

Page 29 of 42

1 2		
3 4 5 6 7 8 9	661	Pearson RG, Dawson TP (2003) Predicting the impacts of climate change on the distribution
	662	of species: are bioclimate envelope models useful? Global Ecology and
	663	<i>Biogeography</i> , 12 , 361-371.
10 11	664	Pearson RG, Thuiller W, Araújo MB, et al. (2006) Model-based uncertainty in species range
12 13	665	prediction. Journal of Biogeography, 33, 1704-1711.
14 15 16 17 18 19 20	666	Phillips SJ, Anderson RP, Schapire RE (2006) Maximum entropy modeling of species
	667	geographic distributions. Ecological Modelling, 190, 231-259.
	668	Phillips SJ, Dudik M (2008) Modeling of species distributions with Maxent: new extensions
21 22 23	669	and a comprehensive evaluation. <i>Ecography</i> , 31 , 161-175.
24 25	670	Phillips SJ, Dudík M, Schapire RE (2004) A Maximum Entropy Approach to Species
26 27 28	671	Distribution Modeling. Proceedings of the Twenty-First International Conference on
28 29 30	672	Machine Learning, 655-662.
31 32	673	Pinho C, Harris DJ, Ferrand N (2007) Comparing patterns of nuclear and mitochondrial
33 34 35	674	divergence in a cryptic species complex: the case of Iberian and North African wall
36 37	675	lizards (Podarcis, Lacertidae). Biological Journal of the Linnean Society, 91, 121–
38 39	676	133.
40 41 42	677	Pleguezuelos JM, Márquez R, Lizana M (2002) Atlas y Libro Rojo de los Anfíbios y Reptiles
43 44	678	de España. Dirección General de Conservación de la Naturaleza - Asociación
45 46	679	Herpetologica Española: Madrid.
47 48 49	680	Pounds JA, Bustamante MR, Coloma LA, et al. (2006) Widespread amphibian extinctions
50 51	681	from epidemic disease driven by global warming. Nature, 439, 161-167.
52 53	682	R Development Core Team (2008). R: A language and environment for statistical computing.
54 55 56	683	Foundation for Statistical Computing, Vienna, Austria.
57 58	684	Rodrigues ASL, Andelman SJ, Bakarr MI, et al. (2004) Effectiveness of the global protected
59 60	685	area network in representing species diversity. Nature, 428, 640-643.

1 2		
3 4 5 6	686	Root TL, MacMynowski DP, Mastrandrea MD, Schneider SH (2005) Human-modified
	687	temperatures induce species changes: Joint attribution. Proceedings of the National
7 8 9	688	Academy of Science, 102 , 7465-7469.
10 11	689	Root TL, Price JT, Hall KR, Schneider SH, Rosenzweig C, Pounds JA (2003) Fingerprints of
12 13	690	global warming on wild animals and plants. <i>Nature</i> , 421 , 57-60.
14 15 16	691	Sala OE, Chapin FS, Armesto JJ, et al. (2000) Global Biodiversity Scenarios for the Year
17 18	692	2100. Science, 287 , 1770 - 1774.
19 20	693	Seo C, Thorne JH, Hannah L, Thuiller W (2009) Scale effects in species distribution models:
21 22 23	694	implications for conservation planning under climate change. Biology Letters, 5, 39-
24 25 26 27 28 29 30 31 32	695	43.
	696	Skelly DK, Joseph LN, Possingham HP, Freidenburg LK, Farrugia TJ, Kinnison MT, Hendry
	697	AP (2007) Evolutionary responses to climate change. Conservation Biology, 21,
	698	1353-1355.
33 34 35	699	Spector S (2002) Biogeographic crossroads as priority areas for biodiversity conservation.
36 37	700	Conservation Biology, 16, 1480-1487.
38 39	701	Thomas CD, Cameron A, Green RE, et al. (2004) Extinction risk from climate change.
40 41 42	702	Nature, 427 , 145-148.
43 44	703	Thuiller W (2004) Patterns and uncertainties of species' range shifts under climate change.
45 46	704	Global Change Biology, 10 , 2020 - 2027.
47 48 49	705	Thuiller W, Albert C, Araújo MB, et al. (2008) Predicting global change impacts on plant
50 51	706	species' distributions: Future challenges. Perspectives in plant ecology, evolution and
52 53	707	systematics, 9, 137–152.
54 55 56	708	Thuiller W, Lafourcade B, Engler R, Araujo MB (2009) BIOMOD - a platform for ensemble
57 58	709	forecasting of species distributions. <i>Ecography</i> , 32 , 369-373.
59 60		

3	
4	
5	
6	
7	
8	
q	
10	
10	
11	
12	
13	
14	
15	
16	
17	
18	
19	
20	
21	
22	
22	
20 24	
9 10 11 12 13 14 15 16 17 18 19 21 22 23 24 5 26 27 28 20 31 23 34 35 36 37 8 9 40	
25	
26	
27	
28	
29	
30	
31	
32	
33	
34	
25	
30	
30	
37	
38	
39	
40	
41	
42	
43	
44	
45	
46	
40 47	
47 48	
40 40	
49 50	
50	
51	
52	
53	
54	
55	
56	
57	
58	
59	
60	

710	Thuiller W, Lavorel S, Araújo MB, Sykes MT, Prentice IC (2005) Climate change threats to
711	plant diversity in Europe. Proceedings of the National Academy of Sciences of the
712	United States of America, 102 , 8245-8250.
713	Wake DB (2007) Climate change implicated in amphibian and lizard declines. PNAS, 104,
714	8201-8202.
715	Walther G-R, Post E, Convey P, et al. (2002) Ecological responses to recent climate change.
716	Nature, 416 , 389-395.
717	Webster M, Forest C, Reilly J, et al. (2002) In MIT Joint Program on the Science and Policy
718	of Global Change - report nº 95, pp. 23.
719	Williams SE, Shoo LP, Isaac JL, Hoffmann AA, Langham G (2008) Towards an Integrated
720	Framework for Assessing the Vulnerability of Species to Climate Change. PLOS
721	BIOLOGY, 6, 2621-2626.
722	Zweig MH, Campbell G (1993) Receiver-Operating Characteristic (ROC) plots: a
723	fundamental evaluation tool in clinical medicine. Clinical Chemistry, 39, 561-577.

Global Change Biology

Table 1 – Amphibian species list, conservation status under IUCN criteria (NE – not evaluated, LC – least concern, NT – near threatened, VU – vulnerable, EN – endangered), number of grid cells where each species is currently predicted to occur (current), number of grid cells where habitat is predicted to become unsuitable (negative values) or become suitable (positive values) in the future and overall tendency under storylines A2 and B2 and unlimited dispersal assumption. * indicates species strictly endemic to the Iberian Peninsula.

Species	IUCN	Current	2020	2050	2080	Tendency	2020	2050	2080	Tendency
				-	-				-	
Alytes cisternasii*	NT	2123	2460	1263	2114	Expansion/ Contraction	1496	760	1164	Expansion/ Contraction
Alytes dickhilleni*	VU	415	-415	-414	-415	Contraction	-415	-415	-415	Contraction
Calotriton asper**	NT	385	-200	-384	-385	Contraction	-173	-316	-385	Contraction
Chioglossa lusitanica*	VU	790	-727	-750	-788	Contraction	-790	-734	-774	Contraction
			-			A		-	-	
Discoglossus galganoi*	NT	2472	2286	2472	2472	Contraction	-2337	2428	2416	Contraction
Discoglossus jeanneae*	NT	1474	1991	3922	4684	Expansion	2805	2439	5052	Expansion
Lissotriton boscai*	LC	2371	3571	-40	- 1776	Expansion/ Contraction	3148	1177	553	Expansion
Pelobates cultripes	NT	3087	3988	4232	3936	Expansion	3998	4097	3933	Expansion
Pelodytes ibericus*	LC	927	-2	487	-648	Contraction/ Expansion/Contaction	86	228	210	Expansion
Pelophylax perezi	LC	5607	-864	-47	- 1849	Contraction	-617	-752	-526	Contraction
Pleurodeles waltl	NT	2826	-689	- 1751	-284	Contraction	-974	- 1710	1	Contraction
Rana iberica*	VU	1324	-370	- 1115	- 1324	Contraction	-673	-946	- 1132	Contraction
Rana pyrenaica**	VU	89	-89	-89	-89	Contraction	-89	-89	-89	Contraction
Triturus marmoratus	LC	2080	۔ 2075	- 2080	- 2080	Contraction	-2080	- 2080	- 2080	Contraction
Triturus pygmaeus*	NT	1358	1277	1849	1585	Expansion	1649	1674	1849	Expansion

Global Change Biology

 Table 2 – Reptile species list, conservation status under IUCN criteria (NE – not evaluated, LC – least concern, NT – near threatened, VU – vulnerable, EN

 – endangered), number of grid cells where each species is currently predicted to occur (current), number of grid cells where habitat is predicted to become

 unsuitable (negative values) or become suitable (positive values) in the future and overall tendency under storylines A2 and B2 and unlimited dispersal

 assumption. * indicates species strictly endemic to the Iberian Peninsula.** indicates species endemic to the Pyrenean Mountains.

Species	IUCN	Current	2020	2050	2080	Tendency	2020	2050	2080	Tendency
Acanthodactylus erythrurus	NT	2545	4004	4354	4627	Expansion	3753	4065	4341	Expansion
Algyroides marchi*	VU	204	-155	292	305	Contraction/ Expansion	49	257	-65	Expansion/ Contraction
Blanus cinereus*	LC	2778	1725	2152	2398	Contraction	1787	1620	2416	Expansion
Chalcides bedriagai*	NT	2344	2891	3553	3510	Expansion	2432	2679	4110	Expansion
Chalcides striatus	LC	2937	1108	-263	-249	Expansion/ Contraction	714	-702	741	Expansion
Coronella girondica	LC	2899	4606	2082	4572	Expansion	2484	2432	4689	Expansion
Hemorrhois hippocrepis	LC	2583	1313	2325	2326	Expansion	1275	1962	2720	Expansion
Iberolacerta aranica**	CR	23	-23	-23	-23	Contraction	-23	-23	-23	Contraction
Iberolacerta aurelioi*	EN	19	-19	-19	-19	Contraction	-19	-19	-19	Contraction
Iberolacerta bonnali**	VU	64	-64	-64	-64	Contraction	-64	-64	-64	Contraction
Iberolacerta cyreni*	EN	867	-427	-588	-750	Contraction	-426	-538	-742	Contraction
Iberolacerta monticola*	VU	524	-469	-524	-524	Contraction	-524	-524	-524	Contraction
Lacerta schreiberi*	NT	1560	778	-367	-875	Expansion/ Contraction	-95	-54	-401	Contraction
Podarcis bocagei*	LC	807	-761	-806	-807	Contraction	-807	-807	-806	Contraction
Podarcis carbonelli*	VU	519	212	-226	-499	Expansion/ Contraction	209	-193	-7	Expansion/ Contraction
Podarcis hispanica	LC	4522	2233	2367	2926	Expansion	2235	2294	3033	Expansion
Psammodromus algirus	LC	4311	2643	1170	- 1220	Expansion/ Contraction	1842	1349	1470	Expansion
Psammodromus hispanicus	LC	2882	1705	2403	2524	Expansion	1959	2416	3736	Expansion
Rhinechis scalaris	LC	2674	4742	4915	4927	Expansion	4832	4911	4922	Expansion
Timon lepida	LC	4080	۔ 1325	- 1219	- 1950	Contraction	-1974	۔ 1310	-20	Contraction

Page	34	of	42
------	----	----	----

Vipera latastei	VU	2688	3865	3873	3630	Expansion	3344	3890	3139	1	
Vipera seoanei*	EN	873	-867	-873	-873	Contraction	-873	-873	-873	Contraction	
											33

Page 35 of 42

Global Change Biology

Table 3 – Species predicted to completely lose distribution range in the Iberian Peninsula until each of the
periods (2020, 2050 and 2080) according to storylines (A2, B2 or both) and under the unlimited dispersal
scenario. * indicates species strictly endemic to the Iberian Peninsula.** indicates species endemic to the
Pyrenean Mountains.

Species	2020	2050	2080
Alytes dickhilleni*	A2/B2	B2	A2/B2
Calotriton asper**	-	A2	A2/B2
Chioglossa lusitanica*	B2	-	-
Discoglossus galganoi**	-	-	A2
Iberolacerta aranica**	A2/B2	A2/B2	A2/B2
Iberolacerta aurelioi*	A2/B2	A2/B2	A2/B2
Iberolacerta bonnali**	A2/B2	A2/B2	A2/B2
Iberolacerta monticola*	B2	A2/B2	A2/B2
Podarcis bocagei*	B2	B2	A2
Rana iberica*	-	-	A2
Rana pyrenaica**	A2/B2	A2/B2	A2/B2
Triturus marmoratus	B2	A2/B2	A2/B2
Vipera seoanei*	B2	A2/B2	A2/B2

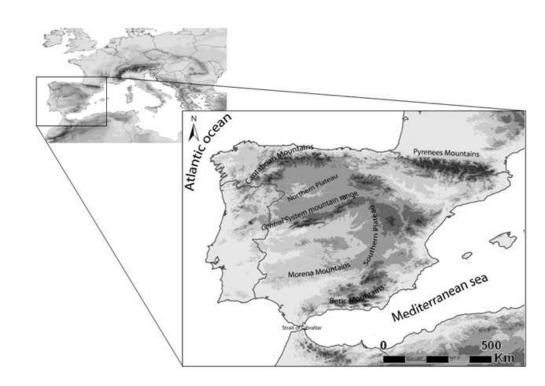
Global Change Biology

species in each year and in each storyline, within the 9 modelling methods and the 3 global circulation models.

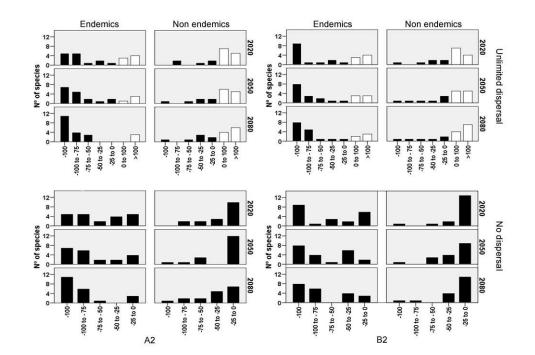
			20	50			2080											
		B2				A2			B2			A2		B2				
Species	Max	Min	STDV	Max	Min	STDV	Max	Min	STDV	Max	Min	STDV	Max	Min	STDV	Max	Min	STDV
Alytes cisternasii	7293	71	2242.1	7535	34	2370.0	7207	2	2740.5	7023	8	2479.8	7602	0	2647.8	7602	0	2653.5
Alytes dickhilleni	5358	0	1406.3	5822	0	1795.8	5529	0	1792.7	5598	0	1695.5	7602	0	2315.4	7602	0	2305.5
Calotriton asper	7599	0	2858.6	7600	0	1755.6	4292	0	1090.9	7581	27	2156.2	7434	0	2258.4	7587	0	2203.3
Chioglossa lusitanica	1064	0	350.9	886	0	295.2	1584	0	391.5	7385	0	1393.2	3156	0	709.1	2428	0	524.7
Discoglossus galganoi	7420	0	2234.4	6310	0	1748.5	4614	0	1809.2	4777	0	1737.1	7602	0	2906.3	7602	0	2683.6
Discoglossus jeanneae	7602	228	2569.2	7602	374	2461.3	7602	776	2443.3	7602	749	2360.1	7602	197	3010.9	7602	710	2608.1
Lissotriton boscai	7595	1076	2196.3	7345	167	2069.3	7199	31	2082.1	7212	490	1917.1	7602	0	2607.7	7602	0	2444.6
Pelobates cultripes	7602	52	2273.2	7602	162	2235.3	7569	148	2406.4	7521	122	2220.9	7602	0	2709.4	7602	0	2592.4
Pelodytes ibericus	7387	13	2051.0	7584	268	2059.7	7516	509	2088.5	6857	22	1866.0	7519	0	2373.3	7328	0	2256.2
Pelophylax perezi	7601	372	2572.3	7602	347	2523.7	7602	315	2561.0	7602	311	2424.0	7602	0	2789.7	7602	0	2794.3
Pleurodeles waltl	7170	7	2255.1	7552	291	2124.8	7308	73	1997.5	7391	154	1912.2	7602	0	2568.5	7602	212	2544.5
Rana iberica	4186	7	1271.4	7511	0	1873.9	7413	0	1834.1	7503	0	2230.5	7515	0	2163.2	7544	0	2186.6
Rana pyrenaica	7487	0	2225.4	7389	0	1954.3	7466	0	2385.7	7590	0	2394.7	7548	0	2077.4	7490	0	2383.3
Triturus marmoratus	4589	0	1458.7	6450	0	1729.7	7313	0	1841.9	3889	0	1193.0	4473	0	1266.2	4179	0	1244.7
Triturus pygmaeus	7602	248	2386.1	7590	651	2141.6	7106	133	2218.2	7034	90	2155.6	7602	0	2720.7	7602	406	2336.6

Appendix A.2 –Maximum (Max), minimum (Min), and standard deviation (STDV), of the number of occurrences predicted for each reptile species in each year and in each storyline, within the 9 modelling methods and the 3 global circulation models.

			20	20					20	50					20	80		
		A2			B2			A2			B2			A2			B2	
Species	Max	Min	STDV	Max	Min	STDV	Max	Min	STDV	Max	Min	STDV	Max	Min	STDV	Max	Min	STDV
Blanus cinereus	7322	591	1676.3	7524	636	1693.2	7369	850	1707.5	6521	374	1691.0	7602	34	2593.4	7602	73	2157.2
Chalcides bedriagai	7598	53	2481.0	7148	268	2297.7	7409	299	2470.3	7376	111	2394.4	7602	20	3096.2	7602	63	2818.1
Chalcides striatus	7602	328	2238.7	7466	47	2164.6	7509	157	2316.2	7307	396	2121.2	7602	59	2565.4	7602	515	2502.4
Coronella girondica	7602	182	2662.8	7602	52	2788.2	7602	157	2599.1	7602	262	2676.9	7602	74	2992.2	7602	130	2815.2
Hemorrhois hippocrepis	6982	1380	1545.2	6359	902	1468.2	7281	1241	1835.1	7121	1208	1719.0	7602	0	2510.3	7602	0	2137.0
Iberolacerta aranica	463	0	122.1	2553	0	489.7	122	0	28.3	7493	0	1441.0	7600	0	1461.8	7592	0	1458.5
Iberolacerta aurelioi	7456	0	2368.1	7527	0	2390.1	7531	0	2401.9	7527	0	2377.6	7586	0	2414.6	7533	0	2400.6
Iberolacerta bonnali	7543	0	2406.0	7570	0	2413.6	7579	0	2420.1	7573	0	2402.6	7567	0	1459.5	7577	0	2418.1
Iberolacerta cyreni	7598	0	2327.3	7598	9	2324.0	7598	0	2320.5	7597	0	2343.6	7582	0	2331.2	7596	0	2393.4
Iberolacerta monticola	7510	0	2059.9	7446	0	1973.2	5825	0	1154.0	5002	0	982.8	6755	0	2171.6	6166	0	1612.7
Lacerta schreiberi	7378	361	2302.8	7235	0	2509.0	7325	0	2491.8	7291	58	2565.3	7602	0	2564.4	7313	0	2553.4
Podarcis bocagei	1609	0	527.1	1880	0	545.7	1786	0	561.8	1289	0	474.5	7602	0	1515.4	7602	0	1938.3
Podarcis carbonelli	7522	152	2816.1	7524	15	2654.5	7487	45	2465.4	6223	43	2142.9	7602	0	2742.7	7602	0	2703.4
Podarcis hispanica	7602	380	2543.7	7602	79	2739.4	7602	565	2756.7	7602	351	2626.6	7602	0	3080.0	7602	0	2922.5
Psammodromus algirus	7602	140	2737.2	7602	317	2747.0	7602	178	2833.4	7602	383	2740.9	7602	0	3224.3	7602	0	2889.7
Psammodromus hispanicus	7417	274	2760.3	7072	164	2697.7	7439	231	2974.2	7434	271	2867.7	7602	0	3164.2	7602	0	3026.3
Rhinechis scalaris	7602	372	2233.7	7602	82	2419.2	7602	1275	2299.2	7602	1538	2096.5	7602	0	2596.5	7602	0	2587.9
Timon lepida	7602	237	2466.0	7602	82	2709.5	7602	255	2772.4	7602	591	2616.2	7602	0	3121.7	7602	0	3022.9
Vipera latastei	7594	63	2148.9	6928	71	2270.8	7222	198	2178.5	7244	556	1811.9	7570	0	2997.2	7232	0	2838.1
Vipera seoanei	3144	0	643.7	949	0	265.2	3314	0	656.8	967	0	291.6	4931	0	967.9	4811	0	940.0

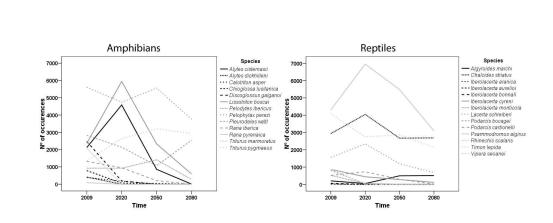


Location and map of the Iberian Peninsula depicting altitude and major geographic features. Altitude is represented in a range of grey colours where darker areas represent higher altitude. 168x123mm (96 x 96 DPI)

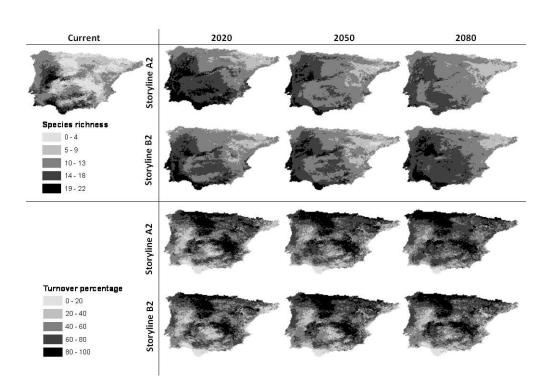


Number of species predicted to gain or lose suitable habitat under different dispersion assumptions (unlimited dispersal and no dispersal) and storylines (A2 and B2), categorised by the percentage of range contraction or expansion relatively to present (x axis). Endemics and non endemics species are represented separately, with endemics representing species strictly endemic to the Iberian Peninsula and species endemic to the Pyrenean Mountains. Black columns represent species predicted to lose suitable habitat while white columns represent species predicted to gain suitable habitat.

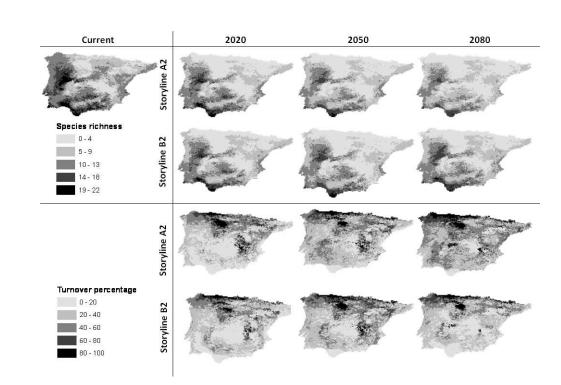
275x190mm (96 x 96 DPI)



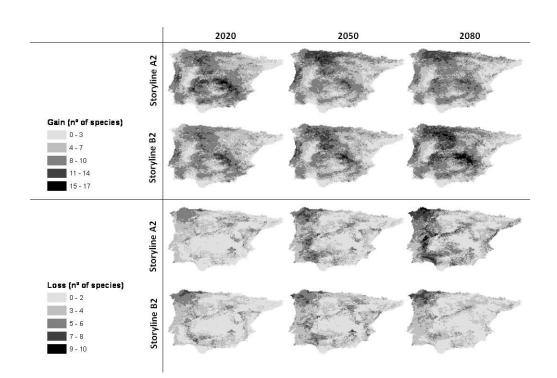
Variation in the predicted number of occurrences through time for species predicted to lose suitable habitat according to the unlimited dispersion scenario and storyline A2. 132x47mm (300 x 300 DPI)



Spatial distribution of predicted current and future (2020, 2050 and 2080) species richness and turnover percentage according to storylines A2 and B2 and under the unlimited dispersion scenario. $275 \times 190 \text{ mm}$ (96 x 96 DPI)



Spatial distribution of predicted current and future (2020, 2050 and 2080) species richness and turnover percentage according to storylines A2 and B2 and under the no dispersion scenario. $275 \times 190 \text{ mm}$ (96 x 96 DPI)



Spatial distribution of predicted gain (under the unlimited dispersion scenario) and loss (in both dispersion scenarios) of number of species in the future (2020, 2050 and 2080) according to storylines A2 and B2 and under the unlimited dispersion scenario. 275x190mm (96 x 96 DPI)