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▶ To cite this version:

K.E. Atkins, J.M.J. Travis. Local adaptation and the evolution of species ranges under climate change. Journal of Theoretical Biology, 2010, 266 (3), pp.449. 10.1016/j.jtbi.2010.07.014 . hal-00616258

HAL Id: hal-00616258 https://hal.science/hal-00616258v1

Submitted on 21 Aug 2011

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Author's Accepted Manuscript

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PII:S0022-5193(10)00363-2DOI:doi:10.1016/j.jtbi.2010.07.014Reference:YJTBI6076

To appear in: Journal of Theoretical Biology

Received date:5 August 2009Revised date:13 July 2010Accepted date:14 July 2010



www.elsevier.com/locate/yjtbi

Cite this article as: K.E. Atkins and J.M.J. Travis, Local adaptation and the evolution of species ranges under climate change, *Journal of Theoretical Biology*, doi:10.1016/j.jtbi.2010.07.014

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LOCAL ADAPTATION AND THE EVOLUTION OF SPECIES RANGES UNDER CLIMATE CHANGE 2

Atkins, K.E. *^{†‡} Travis J.M.J.[§]

Corresponding Author: katherine.atkins@yale.edu

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.ed for the second sec *Present Address: Department of Epidemiology and Public Health, Yale University School of Medicine, New Haven, CT 06520

[†]Centre for Infectious Diseases, University Of Edinburgh, West Mains Road, EH9 3JT, UK

[‡]Depts. of Biology and Mathematics, University of York, Heslington, York, YO10 5DD, UK

[§]Institute of Biological Sciences, Zoology Building, Tillydrone Avenue, University of Aberdeen, Aberdeen, AB24 2TZ, UK

Abstract

6	The potential impact of climate change on biodiversity is well documented. A well
	developed range of statistical methods currently exists that projects the possible fu-
8	ture habitat of a species directly from the current climate and a species distribution.
	However, studies incorporating ecological and evolutionary processes remain limited.
10	Here, we focus on the potential role that local adaptation to climate may play in
	driving the range dynamics of sessile organisms. Incorporating environmental adap-
12	tation into a stochastic simulation yields several new insights. Counter-intuitively,
	our simulation results suggest that species with broader ranges are not necessarily
14	more robust to climate change. Instead, species with broader ranges can be more
	susceptible to extinction as locally adapted genotypes are often blocked from range
16	shifting by the presence of cooler adapted genotypes that persist even when their
	optimum climate has left them behind. Interestingly, our results also suggest that it
18	will not always be the cold-adapted phenotypes that drive polewards range expan-
	sion. Instead, range shifts may be driven by phenotypes conferring adaptation to
20	conditions prevalent towards the centre of a species' equilibrium distribution. This
	may have important consequences for the conservation method termed predictive
22	provenancing. These initial results highlight the potential importance of local adap-
	tation in determining how species will respond to climate change and we argue that
24	this is an area requiring urgent theoretical and empirical attention.

Keywords: local density dependence; environmental gradient; environmental 26 niche; extinction, species' range.

Introduction

28The threat that climate change poses to biodiversity is well documented (Parmesan and Yohe, 2003; Pearson et al., 2002; Peterson et al., 2002; Thomas et al., 2004; 30 Walther et al., 2002). Many species have already exhibited responses to the changing environment; in a meta-analysis Parmesan and Yohe (2003) found that 434 species 32out of 893 had shifted in distribution and/or abundance. A key question is whether climate change increases the risk that species will become extinct, and initial results 34from a study by Thomas et al. (2004) are alarming. Even for a best-case scenario. using the minimum expected climate change and ignoring the potential for dispersal 36 limitation, the authors predict a 9-13% loss in species abundance. Using the worst climate scenario and with extreme dispersal limitation this figure rises to a staggering 38 38-52%. Whilst this study represents a useful exercise, not least in terms of focussing minds on the problem at hand, it simultaneously highlights some important deficien-40 cies that it shares with the many bioclimate envelope models (BEMs), widely used to predict future ranges (Hampe, 2004). There is an urgent need for the development 42 of a new generation of models to make this type of prediction. However, this goal is hindered by a basic lack of understanding of the likely consequences of key ecological 44 and evolutionary processes for range dynamics under climate change. Local adaptation is a good example of one of these processes: many species comprise collections of 46 populations each adapted to local climatic conditions. However, there has been little formal consideration of the impact the degree of local adaptation present within a 48species might have on range dynamics (Polechová et al., 2009), and no attempt to

incorporate these processes into predictive modelling. It is likely that two aspects of
a species' adaptation will be important. First, the steepness with which an individual's fitness declines as it is moved from its optimal climate and second, the total
range of environmental conditions under which a species, as a whole, can survive.
In this contribution, we take a first step and develop a relatively simple simulation
model to explore the issue of how local adaptation influences a species ability to track
changing climate.

A plethora of studies have used bioclimate envelope modelling to predict the 56future biogeographic ranges of species (Bakkenes et al., 2002; Broennimann et al., 582006; Peterson et al., 2002; Schwarz et al., 2006; Thuiller et al., 2005). Refined methods are being developed that entail the use of increasingly sophisticated statistics 60 (e.g. Gavin and Hu (2005)), genetic algorithms (e.g. Termansen et al. (2006)) and neural networks (e.g. Pearson and Dawson (2003)), and it has even been argued 62that the optimal approach should use a combination of these approaches in order to look for concordance between them (Araujo et al., 2005). Intense activity in this 64 field has provided a suite of methods that obtain a relationship between the climate and a species range prior to environmental change: thus defining the 'climate space' 66 occupied by the species. Utilising future climate scenarios, these methods predict the future climate space for the species. However, there is a growing realisation that this 68 correlational approach alone is ill-equipped to predict future species ranges (Davis et al., 2005; Hampe, 2004; Lawton, 2000; Woodward and Beerling, 1997).

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The dynamics of a species' range during a period of climate change will be dependent upon the existence and interplay of various ecological and evolutionary pro-

72 cesses. Dispersal is perhaps the most obvious of these and it is anticipated that species with greater dispersal ability will be better able to track a changing climate
74 (Travis and Dytham, 2002; Travis et al., 2009). Midgley et al. (2002) incorporate simple dispersal functions into a model predicting future ranges of Cape Proteaceae.
76 The description of the spatial population dynamics used in this model is relatively crude but, in linking a model of ecological dynamics with a BEM, Midgley et al. (2002) take an important step in the right direction.

Theoretical studies, often focussing on invasion biology rather than range dynam-80 ics under climate change, have considered a range of evolutionary dynamics during expansions: Garcia-Ramos and Rodriguez (2002) demonstrate the interacting roles 82 played by local adaptation and habitat heterogeneity in determining spread rate; several theoretical studies have now demonstrated that range expansion might be 84 accelerated by the evolution of increased dispersal propensity at expanding fronts (e.g. Travis and Dytham, 2002; Burton et al., 2010) and recent empirical evidence 86 confirms that this is an important effect (Phillips et al., 2006); Klopfstein et al. (2006) demonstrate that neutral mutations arising on the edge of a range expansion some-88 times 'surf' on the wave of advance and can thus reach a wider spatial distribution and higher frequency than would be expected in a stationary population. Klopfstein 90 et al. (2006) suggest that this surfing phenomenon may increase the rate of evolution of spatially expanding populations. More recent extensions to this work have demon-92strated that this surfing dynamic can be important for non-neutral mutations. Even deleterious mutations can sometimes attain high abundance at an expanding front 94(Burton and Travis, 2008a; Excoffier and Ray, 2008; Travis et al., 2007) and this effect

can substantially modify evolutionary dynamics where fitness landscapes are rugged
 (Burton and Travis, 2008b). In this paper we concentrate on local adaptation, an
 important evolutionary process that, to date, has rather surprisingly received little
 attention in terms of its potential role in range expansions and determining a species
 dynamics during an episode of climate change.

100Considerable empirical evidence indicates that locally adapted phenotypes are commonplace within many species (Lambrechts et al., 1996; McNeilly and Antonovic, 102 1968; Riihimaki et al., 2005; Santamaria et al., 2003). For example, it has long been known that many plant species show genetic differentiation in the timing of flowering McNeilly and Antonovic (1968); this variation is frequently clinal with 104northern populations consistently different from southern counterparts (Riihimaki 106et al., 2005). Occurrence of a genotype outside of its usual range may result in suboptimal flowering time and a corresponding reduction in fitness. Similar patterns 108 have been described for a bird species: across Europe, blue tits exhibit adaptive differences in photoresponsiveness that results in laying dates coinciding with local 110 caterpillar availability (Lambrechts et al., 1996). These two examples both involve the timing of key events being optimally aligned to seasonal environmental conditions, 112but there are other ways in which local adaptation may be manifest. It is likely that within many species there is variability in thermal tolerances with individuals 114towards warmer parts of the range adapted to function better in the heat while directional selection will have resulted in individuals found in cooler regions having 116 a physiology that equips them to function efficiently in the cold (e.g. fish (Feminella and Matthews, 1984), amphibians (Olsson and Uller, 2003) and insects (Ayres and

118 Scriber, 1994)).

There is now recognition that local adaptation is likely to be an important de-120terminant of species' responses to climate change and an interdisciplinary treatment of this issue, linking evolution and ecology, has already been called for (Pertoldi and 122Bach, 2007). We establish how local adaptation drives the dynamics of biogeographic ranges during and following a period of climate change. We utilise a spatially ex-124plicit individual based simulation to model the impact of climate change on a single species. This model uses both stochastic demographic parameters and allows inclusion of local density dependent rates for survival of individuals. In our discussion we 126emphasise the potential implications of our results for conservation biology and high-128light where we believe future effort is required to develop improved theory related to the evolutionary ecology of range shifting.

Accelet in a

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130 The Model

There has been considerable recent interest in modelling the ecological and evolutionary dynamics of spatially structured populations that live along environmental gradients (e.g. Alleaume-Benharira et al., 2006; Travis et al., 2006; Dytham, 2009) and some studies have extended this approach to consider the range shifting of populations structured in this way (e.g. Brooker et al., 2007; Mustin et al., 2009). We adopt a similar approach in this contribution and below we describe, in turn, the structure of the landscape, the ecological and evolutionary features of the model, and the simulation experiments that we conduct.

The Environment

140 Individuals of the single species inhabit a two-dimensional landscape that, unless specified otherwise, has dimensions 200 rows (x value) columns by 200 columns (y 142value), which are ordered from the bottom left hand corner (i.e. increasing an individual's x location would move the individual up the grid, whilst increasing its y144location would move it to the right of the grid). A value of (x,y) determines an individual's location. We assume that environmental conditions vary latitudonally (i.e. 146different rows have different conditions, for example due to variation in climate) and the parameter, $\theta(x)$, is used to denote this condition. We assume a linear gradient, 148b, in environmental conditions (i.e. for every value of y, the condition at x, $\theta(x)$ is bx). An optimal environmental location, J, is defined as the climate under which 150the highest potential population growth rates are obtained. Potential growth rate

declines with distance from this optimum location. J is defined as a row value, x, on 152the grid. In all our simulations, we allow populations to first establish under stable environmental conditions and in these periods the environmental conditions within 154each row remains constant through time. To simulate environmental change (e.g. climate warming), we increase $\theta(x)$ for each row, x and increase J by ν each year. The 156period of climate change lasts for T years. An example of an environmental condition changing latitudonally could be a warming climate, where the local temperature 158increases in all locations over time. Local adaptation to this condition would allow individuals to reside in the climate in which they are adapted. An optimal location, 160J, corresponds to a position where it is always favourable to be near, e.g. Pease et al. (1989) give the example of an advancing glacier, from which there exists an optimal 162place to reside. In the case of a warming climate, this optimal location could be any number of things e.g. a moving front of adverse weather conditions, the moving range 164of a quick-to-adapt parasite vector or food stuff in response to the changing climate. *Global* adaptation would allow individuals to reside near to this optimal location.

166 The Ecology

We assume that each cell on the lattice can support no more than one individual (i.e. we are using a patch occupancy model). In each year we simulate, in turn, reproduction, dispersal of offspring and mortality. All individuals first have the opportunity to reproduce. To avoid potential biases towards particular locations or individuals, the order in which individuals are given the opportunity for reproduction is randomised. The probability that an individual reproduces depends upon its fitness (see

Local Adaptation and Evolution, below). A single offspring is born to a reproduc-174ing adult and this offspring immediately disperses. This once-in-a-lifetime dispersal event occurs via a uniformly distributed Moore neighbourhood (nearest eight cells). 176Offspring that disperse to an unoccupied cell automatically establish. In most simulations, dispersal to an occupied cell results in automatic death and there is thus a 178strong priority effect in relation to space occupancy. However, in some simulations we relax this assumption and allow offspring to displace established individuals with 180 probability, p. We always assume reflective boundary conditions as a way of maintaining the current population size if the individuals are able to find space. After all 182individuals have had an opportunity to reproduce, and offspring have dispersed, we simulate mortality. All individuals are subject to stochastic mortality with proba-184bility, μ . Mortality is independent of both the condition of the environment and an individual's phenotype.

186 Local Adaptation and Evolution

Each individual carries two alleles. These are represented as integer values and their mean determines the environmental condition to which the individual is adapted. For example, an individual possessing two alleles with values 24 and 28, has phenotype, z=26 (and is optimally adapted to the conditions in a row x such that $\theta(x) = 26$). The probability that an individual reproduces in any given year depends upon both where the individual is located in relation to the environmental location with highest potential productivity, J, and the degree to which the individual is adapted to its local environment, $\theta(x)$. Formally, the probability of reproducing is given by the

following expression:

 $\exp\left[((\theta(x) - z)^2/2V_s) + ((J - x)^2/2W_s)\right]$

196 where, the first term accounts for the degree to which the individual is locally adapted and the second accounts for the proximity of the individual to the location 198 with highest potential productivity. V_s determines the steepness of decline in an individual's fitness as it is moved further from the conditions to which it is best 200 adapted, while W_s determines the rate at which maximum potential reproduction declines with distance from J. Note that W_s is akin to the notion of environmental 202 tolerance as described in Pease et al. (1989).

It is worth emphasising the distinction between W_s and V_s and the need for both 204parameters in our model using a simple illustrative example. Consider first two individuals, one of which is perfectly adapted to the conditions at location J while 206the other is perfectly adapted to conditions found 10 rows higher on the lattice than J. Initially we will assume both are located in the environments to which they are adapted. Despite the fact that both are optimally adapted, their reproductive fitness 208will not be the same as the first individual (at location J) is in the inherently more 210 productive environment. Our parameter W_s determines just how much lower the reproductive fitness is of the individual found 10 rows away from J. Next consider 212the case that these same two individuals are both located at the position 10 rows away from J. Now the first individual will have the lower reproductive fitness because, 214 while both are in the same environment, the first is not optimally adapted to that

environment. The amount that the first individual's fitness is reduced due to it being less well-adapted is determined by V_s .

When a birth event occurs the offspring inherits two alleles: a copy of one of its parent's alleles (chosen randomly from the set of 2 with equal probability) and a copy from an individual, selected at random, in the vicinity of its parent. This vicinity can be considered as the pollen dispersal neighbourhood and we term it h. If the vicinity contains no individuals then the offspring inherits both alleles from its parent (this is akin to 'selfing'). During the copying of alleles that ensues, mutation occurs at each locus with a probability of β . When a mutation occurs, the value of the allele is modified by an amount randomly drawn from the discrete uniform distribution, (-10, 10).

226 Simulation Experiments

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We have conducted a number of simulation experiments to explore the dynamics of the model. In the first set, we run the model under a stable climate (i.e. maintaining a constant position of maximum potential reproduction, J). We run the model for a range of parameter values and, in particular, vary both V_s and W_s . We output population abundance and mean phenotype across the environmental gradient. Subsequently, we run a large number of simulations where we initially allow the population to obtain quasiequilibrium under stable environmental conditions before simulating a period of climate change. In these simulations we are most interested in the probability that a population survives an episode of climate change. We run sensitivity analyses to ask, in turn, how this depends upon V_s and W_s , the strength

of the priority effect (p), the rate (ν) and duration of climate change (T), and the
rate and size of mutations. In the final set of simulations we vary the probability of mortality (β), the landscape width (by altering the number of columns on the grid),
and the dispersal neighbourhood for pollen (h). In all the sensitivity analyses, we run 50 replicate simulations for each parameter combination.



242 **Results**

Under stable environmental conditions, population distributions establish as one would expect. For smaller W_s (corresponding to a steeper decline in maximum potential reproductive rate with distance from J), smaller ranges emerge (Figures 1a,1b and 1c). The corresponding mean phenotype is shown in Figures 1d, 1e and 1f. For smaller W_s , there is a greater mismatch between the local environment and the mean phenotype away from J (in these figures J=140).

Most of our results relate to range dynamics under climate change. Figure 2 il-250lustrates typical dynamics showing snapshots at various stages in the climate change period. Initially, we allow the population to obtain quasiequilibrium under stable 252climate and this is what is shown in Figure 2a. At this stage, we define a 'climate envelope' for our simulated species. We establish the range of environmental condi-254tions within which 99% of the individuals are found. We show the location of this climate envelope during the period of environmental change to illustrate how well 256the population's range responds. In the example given in Figure 2, as the environment changes, the population's range lags behind its climate envelope. This is most 258marked at the expanding front, with the leading edge of the range advance falling a substantial distance behind the shifting climate window. However, it is also apparent 260that at the trailing edge some individuals are found well behind what we defined as the species' climate envelope.

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Running the simulation model for a wide range of parameter combinations revealed a rich range of behaviours. We illustrate some of the most important features

264in Figure 3a which shows mean phenotypes across the environmental gradient on one axis, and how this distribution of phenotypes changes during and subsequent to 266a period of climate change (see Figure 3). For most runs of the model where the population survives the episode of climate change we observe that the range width 268following climate change is narrower (Figures 3a, 3b and 3d). This is a consequence of the loss of phenotypes adapted to the warmest and coolest extremes that were 270occupied by the population prior to the onset environmental change. Ultimately phenotypic variation is recovered and the range expands to reoccupy its original 272niche but this can take many thousands of generations. A common feature of the results is the initial expansion of the cold-adapted phenotypes (shown in green in 274Figure 3a) after the onset of change, but this is followed by a rapid collapse of these phenotypes (Figures 3b, 3c and 3d) and, in some cases, a rapid collapse of the whole 276population (Figure 3c). Figure 3c illustrates an important result: even when there is a considerable overlap between the range of a species prior to climate change and 278the climate envelope of the species following change (shown by the white lines in Figure 3) extinction can still occur. Finally, we sometimes observe dynamics where 280the population splits into distinct subpopulations comprising individuals with quite distinct phenotypes. Because they are adapted to quite different climates, these sub-282populations can remain spatially disjunct for quite lengthy periods following climate change (see Figure 3d). 284

Figures 4 to 6 show the results of our sensitivity analyses, conducted to establish to roles of the different model parameters in determining the likelihood that a population is able to survive a period of climate change. Unsurprisingly, steeper declines in

individuals' reproductive fitness as they are moved away from their own optima (i.e. small V_s), lead to reduced probabilities of populations surviving climate change (Figure 4a). However, for the sensitivity analysis around W_s , the results are a little less expected. Higher probabilities of extinction are observed when the fundamental environment niche of the species is broader (i.e. the chance of extinction is greater for higher W_s - see Figure 4b).

Perhaps our most important result is illustrated in Figure 5. This figure shows 294the results of simulations conducted to establish the role of p in determining the likelihood that species survive an episode of climate change. Recall that p determines 296the probability that an offspring is able to displace an already established individual from the site to which it disperses. The counterintuitive result shown in Figure 4b 298was obtained for p = 0.0 (i.e. there is an absolute priority effect). When the opposite is assumed, and offspring always displace existing occupants (i.e. p = 1.0), we find 300 that species that have broader fundamental niches (higher W_s) are now more likely to survive than those with narrower niches (lower W_s). Interestingly, when we assume 302 an offspring has a 50% chance of displacing an existing occupant (p = 0.5), we find that the initial result holds and extinction risk is higher for higher W_s .

We find that the probability that a species' range survives a period of climate change is far more sensitive to some of the model's parameters than others. In general we find that the results are largely insensitive to the mutation parameters, neither the mutation rate nor the effect size of a single mutation have noticeable impacts (results not shown). However, the scale of pollen dispersal, the length of the period of climate change, the extrinsic probability of mortality and the width of the

landscape all have major effects. In general, extinction risk is higher when pollen disperses shorter distances (Figure 6a), the period of climate change is longer (Figure 312
6b), the probability of extrinsic mortality is lower (Figure 6c) and the landscape is broader (Figure 6d).



314 Discussion

In this contribution, we have presented results from a simulation model that inte-316 grates key features from similar models exploring species' dynamics in response to environmental change (Brooker et al., 2007; Dytham, 2009; Mustin et al., 2009, e.g.) 318 and from previous analytical theory exploring local adaptation on environmental gradients (Kirkpatrick and Barton, 1997; Pease et al., 1989). We establish that local 320 adaptation can play a major role in determining the dynamics of range shifting under climate change, and suggest that this is likely to have important implications for 322 predictions of the impact of environmental change on both future species ranges and the rates of extinction.

324 Two results from our work clearly highlight how local adaptation can alter the nature of range shifting. First, a species may fail to survive a period of climate change, 326 even when there is an overlap between its range prior to climate change and the area where climate is predicted to be suitable following climate change. Second, we find 328 that often a range begins to expand in the direction of climate change before rapidly collapsing to extinction. Both these results have important applied implications. 330 The first suggests that any predictions based upon treating the species as a single homogenous unit may be flawed and that, for this reason, simple extensions of the 332 bioclimate envelope approach are likely to be inadequate. The second implies that we need to be extremely cautious before inferring that species that are already expanding 334their ranges in the direction of climate change will continue to do so.

In our initial simulations, we found that species occupying broader niches more

336 at risk of extinction due to climate change. This counterintuitive result required further investigation, and the key assumption driving this difference turned out to 338 be local density dependence in the form of space occupancy. In the initial simulation model, individuals occupy a patch until they die and are never displaced by a 340 newborn individual. When we alter the priority effect within the simulations, such that established individuals are no longer invulnerable to displacement by newborn 342 individuals, we find that species with broader ranges in a stable climate are better able to survive a period of environmental change. Pease et al. (1989) concluded that 344without local density dependence, a species with a broader range was less likely to become extinct during an episode of climate change. This is agreement with our 346 findings.

Examining the original distributions of the population before climate change in 348 the simulation, it becomes clear why the result has been obtained. The individuals at the periphery of the range (those lying on the lines at the extrema of the current 350climate) are lower in density and will necessarily have a lower birth rate than those towards the centre of the distribution (given the same local adaptation). The offspring 352of these individuals will however be able to establish more readily since the density of individuals decreases away from the globally optimal location. Thus, we note that 354the absolute gradient of the density of individuals away from this global optimum increases with decreasing W_s . As a result, we see that decreasing W_s increases the 356 availability of patches surrounding the optimum location, into which offpsring can be born. Therefore as the climate starts to shift and the optimal location moves upwards, 358 the high frequency of better adapted individuals with higher birth rates around the

centre of the distribution have more space to move forward in the range and the individuals at the front of the distribution are soon optimally located, making them more fecund (despite being less locally adapted) but with fewer density dependent effects limiting their birth rate. This then creates a faster turnover of new individuals, on which selection can act. When W_s is large, this movement and selection of phenotypes is much impaired leading to an overall reduction in the adaptability of the population and an increased chance of extinction. The local density-dependence effects have a much more striking effect on extinction probability than mutation rate or mutation size, which do not ameliorate extinction probability if density dependent effects are present.

The important role played by local density dependence, mediated through space 370 occupancy effects, has recently been demonstrated for species assemblages and our results clearly demonstrate that it can also be important within a species. Takenaka 372 (2004) uses simulation modelling to explore the distributions of coexisting tree species as they shift their ranges due to climate change. Takenaka's work shows that the 374replacement of one species by another better adapted to the new climatic conditions can be dramatically slowed if mature individuals continue to block space for lengthy 376 periods even though the climate no longer suits them. Brooker et al. (2007) also described a similar space blocking effect in a simulation model describing the range 378 dynamics of an assemblage comprising species arranged along an environmental stress gradient. The authors highlighted that the longevity of an organism would determine 380 the strength of the effect. Our work highlights that these effects are not limited to interspecific interactions; prolonged space occupancy beyond the point at which an

382 individual is well adapted to local conditions can similarly limit the range shifting of locally adapted phenotypes within a single species. An important determinant 384 of species' extinction, both within and between species, will be the relative rates at which fecundity and mortality are impacted as the climate shifts. If fecundity 386 declines relatively rapidly, space occupancy effects will be greater, while if mortality increases more rapidly, when environmental conditions are displaced from those to 388 which an individual is adapted, the effect will be weaker. Additionally, higher rates of fitness-independent mortality will reduce the strength of the space blocking effect 390 and allow more successful range shifting (see Figure 6c for an example). In future work, it will be interesting to explore how the partitioning of local adaptation within 392 and between species may influence the spatial dynamics of a species guild under climate change scenarios.

394 An interesting feature of our results is that a species can become extinct even when the climate envelopes before and after climate change overlap. The realisa-396 tion of the simulation model illustrated in Figure 3c shows this effect very clearly. Representation of a species as a single unit, as is the case in the statistical pattern 398 matching methods used in bioclimate envelope modelling (Bakkenes et al., 2002; Broennimann et al., 2006; Schwarz et al., 2006; Thuiller et al., 2005) neglects this 400potentially crucial effect. There is an urgent need for a careful consideration of the taxanomic resolution at which we should be modelling the response of biodiversity 402 to climate change. As further information becomes available on the extent to which populations within species are differentially adapted to local conditions, a critical 404appraisal of the importance of this evolutionary process will be required.

This question of the organisational level at which species are represented also 406needs addressing within the community using dynamic vegetation models to predict both changes in ecological and biogeochemical processes during climate change 408 (Cramer et al., 2001; Morales et al., 2005). Many of these models do not even work at the resolution of the species, instead they group species into functional types and 410assume that all members of a functional type share an identical environmental niche. In reality, species (or populations) towards the cooler end of the environmental space 412 that can be occupied by the functional type will not function in warmer conditions as effectively as those species (or populations) that have adapted to the heat, and vice 414 versa. This may have important consequences not only for the projections of future patterns of biodiversity made by this type of model but also for their projections of 416 future biogeochemical cycles and balances.

A new generation of models is required that takes the best parts from bioclimate 418 modelling and incorporates ecological and evolutionary dynamics. The results presented in this paper make a strong case for the inclusion of local adaptation as a 420 matter of urgency. One conservation management option to mitigate some of the effects described in this paper would be the translocation of genotypes within a species 422 range. Our simulations tend to show that the phenotypes adapted to the local conditions found in the core of the range tend to be the most important for range shifting. 424 and from this we would suggest that a potential strategy would be to translocate individuals (for plants, pollen might suffice) from core areas to populations located 426 towards the expanding front. A model paramaterised with species-specific information on dispersal distance and the genetics of local adaptation would be an invaluable

- 428 aid in developing this type of strategy, and could be used to provide recommendations on the optimal amount and frequency of translocation.
- 430 The model considered in this paper has much in common with a model used to explore the dynamics of speciation along environmental gradients (Doebeli and 432 Dieckmann, 2003). Interestingly, we have observed that in some regions of parameter space we find that disruptive selection can result in two (or more) spatially distinct 434species arising in our simulation model. This is not the main topic of this paper, however we note that both during and following a period of climate change a species 436 range tends to become fragmented (see Figures 2 and 3d). In some cases this spatial separation can last for a considerable period following the stabilisation of climate 438(Figure 3d) and it is interesting to speculate that in some cases this may act to promote speciation. It would certainly be worthwhile extending recent speciation 440models (Artzy-Randrup and Kondrashov, 2006; Bolnick, 2006; Yukilevich and True, 2006) to consider the potential impacts of environmental change.
- 442 Understanding the response of biodiversity at different organisational levels under a period of rapid environmental change represents the greatest challenge faced
 444 by ecologists today. Here, we have considered how local adaptation within a sessile species' range can impact on that species dynamics during climate change. By apply446 ing ideas and methods that are already well developed in the ecological literature we have obtained several new insights that we believe have important consequences for
 448 how we should go about both predicting and also managing changes in biodiversity under climate change. We hope that the results from theoretical studies such as this
 450 one make a persuasive case for a closer integration between the statistical modelling

that currently dominates the climate impacts literature and dynamic spatial models incorporating biological processes.

Acknowledgements

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454Thanks to Zaq Coelho and Gustav Delius for continued support. This research was Accepted manuscrit funded by NERC and ALTERNET.

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Figure 1: Frequency distributions for population size and trait value for the equilibrium in a constant environment, where the environment changes with x. The results were averaged over 50 realisations of the model. A set of simulations were run for a different values of $2lnW_s$ given on the graph by different colours (red-2, blue-3, green-4, yellow-5, black-6). An increasing value of W_s relates to an increased environmental niche range, seen in Figures (a), (b) and (c). In Figures (d), (e) and (f) the mean phenotype is an average of all allele values (2 per individual) existing on the environmentally invariant line y = const. In the case of zero density, the mean phenotype is given a value 0.



Figure 2: Snapshots of one realisation of the model showing population inhabiting the 2d lattice landscape. The grid corresponds to the individuals on the grid with the invariant environment of each row and changing environmental gradient through each column. The colour grid shows the optimal trait value for each location, $\theta(x)$, running from hot (red) to cool (blue). Individual patch colour shows trait value to which the individual is adapted, z. The climate envelope (shown by black lines) indicates the area holding approximately 98% of the population at pre-climate change equilibrium, with the median at the optimal. (a) shows the equilibrium in a constant environment before any climate change has occurred; (b), (c), (d) and (e) show the population at a specified time after change onset and (f) shows final distribution after climate change. Illustrative parameters are chosen as $W_s = \exp(5)/2$ and $V_s = 25$, under a climate change rate of $\nu = 0.3$ over T = 200 years to emphasise the range shift. During the climate shift this envelope tracks the environmental optimal for the population. The population splits into clusters before combining once the shift ends.



Figure 3: Four realisations of the model through a period of climate change. The colours show the population and its local adapation, z (blue through to red). (a) shows population persistence with the majority of individuals tracking the climate; $T=300, V_s=100, W_s=500, \nu=25$ leading to the subpopulation at the rear and very front of the range being lost but population persisting due to the middle third of the pre shift population moving. (b) shows population persistence through climate change with only a small subpopulation tracking the climate; $T=300, V_s=10, W_s=500, \nu=23$ allowing population movement due to the middle of the front half of the range shifting, although front margin does show signs of initial movement. (c) shows population extinction during climate change; T=100, $V_s=50$, $W_s=250$, $\nu=40$ produces a climate shift too fast for adaptation. (d) shows population persistence with two small subpopulations tracking the climate; $T=300, V_s=7.5, W_s=500, \nu=23$ produces similar subpopulation shifting its range as seen in (b) but shift occurs in two geographically distinct subpopulations, which eventually rejoin for good on reaching the new equilibrium.



Figure 4: Sensitivity analysis for variance revealing both population variances $(V_s \text{ and } W_s)$ having opposite effects on extinction probability in the model with (a) showing the effect of variance around local optimal adaptation and (b) the effect of variance around the optimal location. The extinction probability for any population can be noted as a non-decreasing function of the rate of climate change, ν . Notice that an increase in the environmental niche (an increase in W_s) leads to a increase, or no change, in the extinction probability.



Figure 5: The effect of decreased density dependent effects on proportion of populations extinct under climate change. Red to Blue scale corresponds as before to 100% to 0% of populations extinct after climate shift. Note the qualitative difference in the two trends: in (a) where p=0.5, a large range width corresponds to a higher extinction probability, whereas in (b) a large range width allows better population persistence.



Figure 6: The effect of different parameters on species extinction. The plots show the proportion of realisations (out of a total of 50) where the species became extinct with varying rates of climate change. In all cases, $W_s = 150$ and unless otherwise varied, parameter values are kept constant at N = 200, T = 200, $V_s = 50$, $\mu = 0.005$, $\sigma = 1.207$ and maximum mutation size is set at 10.