Quaternary range and demographic expansion of *Liolaemus darwinii* (Squamata: Liolaemidae) in the Monte Desert of Central Argentina using Bayesian phylogeography and ecological niche modelling

ARLEY CAMARGO,* FERNANDA P. WERNECK,†‡ MARIANA MORANDO,* JACK W. SITES JR† and LUCIANO J. AVILA*

*Unitad de Diversidad, Sistemática y Evolución, Centro Nacional Patagónico, Consejo Nacional de Investigaciones Científicas y Técnicas, Boulevard Almirante Brown 2915, U9120ACD, Puerto Madryn, Chubut, Argentina, †Department of Biology & Bean Life Science Museum, Brigham Young University, Provo, UT 84602, USA, ‡Departamento de Zoología, Universidade de Brasília, Brasília, DF, CEP 70910-900, Brazil

Abstract

Until recently, most phylogeographic approaches have been unable to distinguish between demographic and range expansion processes, making it difficult to test for the possibility of range expansion without population growth and vice versa. In this study, we applied a Bayesian phylogeographic approach to reconstruct both demographic and range expansion in the lizard *Liolaemus darwinii* of the Monte Desert in Central Argentina, during the Late Quaternary. Based on analysis of 14 anonymous nuclear loci and the cytochrome *b* mitochondrial DNA gene, we detected signals of demographic expansion starting at ~55 ka based on Bayesian Skyline and Skyride Plots. In contrast, Bayesian relaxed models of spatial diffusion suggested that range expansion occurred only between ~95 and 55 ka, and more recently, diffusion rates were very low during demographic expansion. The possibility of population growth without substantial range expansion could account for the shared patterns of demographic expansion during the Last Glacial Maxima (OIS 2 and 4) in fish, small mammals and other lizards of the Monte Desert. We found substantial variation in diffusion rates over time, and very high rates during the range expansion phase, consistent with a rapidly advancing expansion front towards the southeast shown by palaeo-distribution models. Furthermore, the estimated diffusion rates are congruent with observed dispersal rates of lizards in field conditions and therefore provide additional confidence to the temporal scale of inferred phylogeographic patterns. Our study highlights how the integration of phylogeography with palaeo-distribution models can shed light on both demographic and range expansion processes and their potential causes.

Keywords: Bayesian phylogeography, demographic expansion, dispersal, *Liolaemus*, Monte Desert, range expansion

Received 10 October 2012; revision received 25 April 2013; accepted 25 April 2013

Introduction

A synthesis of phylogeographic studies of multiple taxa from the Northern Hemisphere has revealed repeated cycles of range retraction towards southern refugia during glacial periods and range expansion towards higher (northern) latitudes during interglacial periods (Hewitt 2000, 2004; Hewitt & Ibrahim 2001). Comparative studies in the Southern Hemisphere have found similar, but reversed, cyclical patterns of persistence in low-latitude refugia and postglacial expansion to higher (southern) latitudes in multiple taxa (Lessa *et al.* 2010; Sésic *et al.* 2011; Fraser *et al.* 2012). In Patagonia, other
taxa also show patterns of displacements towards the east and persistence in situ in the west (Breitman et al. 2012; Cosacov et al. 2012; Sede et al. 2012). These generalizations about shared phylogeographic patterns among species worldwide have been derived from multiple inferences of demographic history using genetic data and geographic information to elucidate, qualitatively or quantitatively, changes in population size (demographic expansion) and geographic range (spatial expansion). However, until recently, phylogeographic methods have failed to distinguish between demographic and spatial processes within a single statistical framework, making it difficult to test for the possibility of demographic expansion without range expansion, and vice versa.

Traditionally, inferences of population growth have been obtained with test statistics such as Tajima’s D (Tajima 1989), Fu’s Fs, (Fu 1997), Harpending’s raggedness index (Harpending 1994) and Ramos-Orsins and Rozas’ R² (Ramos-Orsins & Rozas 2002). In addition, several Bayesian Skyline Plot (BSP) methods are able to infer changes in effective population size over time using flexible demographic models (Ho & Shapiro 2011). In contrast, assessments of range expansion have relied upon model-free inferences about the association between haplotype clades and their geographic distribution using Nested Clade Phylogeographic Analysis (NCPA) (Templeton 2004, 2008). NCPA has been used routinely to detect range expansions in empirical studies although inferences based on this approach are not statistically robust due to its inability to distinguish between alternative scenarios and high error rates (Beaumont & Panchal 2004, 2008). NCPA has been used to detect range expansions in empirical studies although inferences based on this approach are not statistically robust due to its inability to distinguish between alternative scenarios and high error rates (Beaumont & Panchal 2004, 2008). Other methods have generally lacked a spatially explicit component such as the spatial expansion test implemented in Arlequin 3.5 (Excoffier & Lischer 2010) based on Excoffier’s range expansion model (Excoffier 2004). A fully integrated analysis of genetic and geographic information was achieved in the maximum-likelihood-based program Phylomapper (Leemon & Leemon 2008), which estimates ancestral distributions and assumes a constant rate of spatial diffusion, but does not incorporate either genealogical uncertainty (i.e. it uses a fixed tree topology) or a model of demographic expansion. More recently, a Bayesian method developed for area biogeography could potentially be applied to reconstruct ancestral distributions and dispersal events among discrete locations (Yu et al. 2011).

A Bayesian approach for spatio-temporal phylogeographic reconstruction has been implemented recently in the software BEAST 1.7 (Drummond et al. 2012). This approach allows inference about the geographic location of ancestors while accommodating genealogical uncertainty, and using models of either discrete or continuous diffusion over space and time. While models of homogenous diffusion assume a Brownian constant rate among discrete locations (Lemey et al. 2009a), an extension of this approach that models diffusion on a continuous space allows for variation in the diffusion rate across branches of the genealogy (Lemey et al. 2010), based on analogous models to those implemented for relaxed-molecular clocks (Drummond et al. 2006). Because the analysis is embedded within the BEAST program, the spatial diffusion process can be coupled with the estimation of demographic models (e.g. constant size, exponential growth, BSP, etc.), and model-selection based on marginal likelihoods can be used to test for alternative diffusion and demographic models (Baele et al. 2012). This flexible phylogeographic approach that accomplishes both pattern inference and model testing has resulted in novel insights about the spread of several viral epidemics (Lemey et al. 2009b; Allicock et al. 2012; Pybus et al. 2012), and even the origin of human languages using vocabulary instead of molecular data (Walker & Ribeiro 2011; Bouckaert et al. 2012).

Ecological niche modelling (ENM) has become a popular tool in phylogeography, evolutionary biology and conservation biology for inference of potential geographic distributions of species in past, present and future climatic conditions (Chan et al. 2011). Among other applications, ENM has been used for delimiting species (Raxworthy et al. 2007; Rissler & Apodaca 2007), locating Pleistocene refugia in the Northern (Waltari et al. 2007) and Southern (Carnaval & Moritz 2008; Werneck et al. 2011, 2012b; Fontanella et al. 2012) Hemispheres, developing models of population history (Carstens & Richards 2007; Richards et al. 2007), and studying niche evolution (Warren et al. 2008). Despite several drawbacks of species-distribution models in general, and of ENM in particular (Pearson et al. 2006; Elith & Graham 2009; Elith & Leathwick 2009; Araujo & Peterson 2012), this approach offers the possibility to build a finite set of phylogeographic hypotheses for evaluation with independent genetic data using model-based statistical approaches (Knowles 2009; Bloomquist et al. 2010; Hickerson et al. 2010). Expanded phylogeographic approaches that integrate geospatial methods, such as ENM, can help elucidate the ecological and evolutionary mechanisms underlying population history both when spatio-temporal changes in climate and habitat suitability match demographic inferences (Chan et al. 2011), or when they do not explicitly agree (Thomé et al. 2010; Werneck et al. 2012a).

In this study, we reconstruct an explicit spatio-temporal phylogeographic model for Liolaemus darwinii (Squamata: Liolaemidae), a lizard that occupies sandy habitats in the southern and central portions of the Monte Desert in Argentina (Etheridge 1993). A recent species-delimitation
study based on coalescent methods and multilocus data, found that a southern lineage of *L. darwinii* is sister to a new species distributed further north (Camargo et al. 2012). The two lineages have a distributional gap in central Mendoza Province and differ in scale counts and colouration (Etheridge 2001; Abdala 2007). A previous study found significant signal for demographic and range expansion in southern *L. darwinii* based on cytochrome *b* (cyt *b*) analyses using NCPA and several test statistics, and proposed that Pleistocene climatic cycles could have been associated with this population history (Morando et al. 2004).

Herein, we increased sample sizes and included nuclear loci in order to test the hypothesis of associated population growth and range expansion of *L. darwinii* during the Late Quaternary, using spatio-temporal phylogeographic reconstruction and demographic analyses. Further, we investigated the possible role of the Quaternary climate change in this expansion event via palaeo-distribution modelling since the Last Interglacial (120 ka). We predicted that if population expansion was due to recent climatic change, we would be able to find: (i) a temporal association between the population growth and the range expansion; and (ii) a spatial association between the range expansion and palaeo-distribution models. Specifically, we would expect that the geographic range of the species inferred with both phylogeographic and palaeo-distribution reconstructions should match spatially during the progression of the expansion event.

**Material and methods**

We sampled 404 individuals of *L. darwinii* from 105 localities including the distribution of southern lineages (Fig. 1, Table S1, Supporting information). Tissue samples from liver and/or tail muscle were preserved in absolute ethanol and stored at −20 °C. Specimens were
fixed in 10-20% formalin, later transferred to 70% ethanol and deposited in the herpetological collection of the Centro Nacional Patagón (LJAMM-CNP, CENPAT–CONICET; http://www.cenpat.edu.ar/colecciones03.html) and the Bean Life Science Museum, Brigham Young University (http://mlbean.byu.edu/researchcollections/collections/reptilesandamphibians.aspx). Genomic DNA was extracted with the DNAEasy Qiagen kit (Qiagen).

We used the Green Go Taq PCR kit (Promega) for all PCRs in PTC-200 DNA Engine (MJ Research) or GeneAmp PCR 9700 thermal cyclers (Applied Biosystems, Inc.). Sequencing reactions used the Big-Dye Terminator v3.1 Cycle Sequencing Kit (Applied Biosystems, Inc.) in a GeneAmp PCR 9700 thermal cycler (Applied Biosystems, Inc.). Sequencing products were cleaned with Sephadex G-50 Fine (GE Healthcare Bio-Sciences AB) and sequenced in an ABI 3730xl DNA Analyzer (Applied Biosystems, Inc.). We sequenced the cyt b mtDNA gene for all individuals following protocols in Morando et al. (2004) and subsampled 29 individuals across the geographic range representing the variation found in mtDNA haplotypes for screening anonymous nuclear loci (ANL). We sequenced 14 ANL that were developed from the genomic DNA of a L. darwinii individual (LJAMM-CNP 7097) following protocols in Noonan & Yoder (2009) (Table S2, Supporting information). We generated ~200 random fragments, cloned, sequenced and BLAST searched these to confirm they were anonymous, then designed primers for fragments with confirmed anonymity and used the PCR temperature profile of Noonan & Yoder (2009) to screen ANL in all subsampled individuals.

Chromatograms were checked by eye, and ambiguity codes were used to represent polymorphisms of heterozygote individuals in Sequencher 4.10.1 (Gene Codes Corporation). Gametic phase of heterozygotes was resolved with the program PHASE 2.1.1 (Stephens et al. 2001) based on the input files prepared with SeqPHASE (http://www.mnhn.fr/jflot/seqphase/) (Flot 2010). The heterozygous indels (alleles with length variation) found in 8 ANL were phased along with base polymorphisms by specifying them as unknown phases in a separate input file (known). We ran two independent runs consisting of 10 000 iterations, a thinning interval of 1 and a burn-in of 10 000. To ensure convergence between runs, we checked for identical best haplotype pairs estimated for each individual between runs. Input and output files of PHASE analyses are provided as supplemental files in Dryad (see Data Accessibility below). Subsequently, the output files from PHASE were converted back into FASTA alignments of phased sequences using SeqPHASE. Sequences were inspected by eye to check that there were no fixed heterozygotes at any polymorphic site and to ensure we were not using multiple-copy markers (Thomson et al. 2010). Alignments were tested for recombination with the program RDP3 beta35 (Martin et al. 2010), and best-substitution models for each locus were selected with jModeltest 2 (Darriba et al. 2012).

Population structure

We estimated haplotype networks for all loci using TCS 1.21 (Clement et al. 2000) using a 95%-connection limit and considering gaps as a fifth state for the ANL. In the case of cyt b, we first eliminated positions with missing nucleotides to minimize ambiguous connections among haplotypes, reaching a final alignment length of 357 bp. We also estimated the optimal number of populations using the 15-locus data set representing 29 individuals with the Bayesian program Geneland 4.0.3 (Guillot et al. 2005a) under the assumption of Hardy–Weinberg equilibrium among populations and linkage equilibrium between loci (Guillot et al. 2005b). We ran five independent replicate runs with the number of populations ranging between 1 and 10, and assuming a correlated allele frequencies model and a spatial model without uncertainty on coordinates. Each run consisted of one million iterations, a thinning interval of 1000 and a burn-in phase of 200 iterations. To assess convergence among runs, we plotted burn-in plots of the posterior probability parameter and compared the posterior estimates of the number of populations. We created probability maps of population membership on a spatial domain defined by 100 pixels in both X- and Y-axes following the authors’ recommendations.

Demographic expansion

We reconstructed the change in effective population size over time with a multilocus Extended Bayesian Skyride Plot (EBSP, Heled & Drummond 2008) in BEAST 1.7.4 (Drummond et al. 2012). We unlinked substitution, clock and tree models for all loci, and specified a linear model of population size, instead of the less realistic stepwise model. We applied a strict clock model for cyt b with the substitution rate sampled from a normal distribution with mean of $1.94 \times 10^{-9}$ (standard deviation $= 3.46 \times 10^{-5}$) to match the calibrated values of Olave et al. (in review). Olave et al. used a recently dated fossil of Liolaemus to calibrate a divergence event within the genus and to obtain an estimate of the substitution rate of multiple loci (including cyt b) based on a representative sample of species that were analysed with the *BEAST species-tree method. We applied a strict clock model because a likelihood ratio test could not reject the hypothesis of a molecular clock based on the
Reconstructing the spatial diffusion pattern and produce an animation for visualization in Google Earth, using spread 1.0.6 (Bielejec et al. 2011). We also used TimeSlicer (kindly provided by P. Lemey) to summarize the variation in diffusion rate over time based on the posterior sample of trees at multiple time slices.

All BEAST analyses were run with the Beagle high-performance library (Ayres et al. 2012) in the m6 cluster of the Fulton Supercomputing Lab at Brigham Young University (https://marylou.byu.edu/). All input xml and output log files from these analyses are available in the Dryad digital repository (see Data Accessibility below).

Palaeo-distribution modelling

We used the maximum entropy machine-learning algorithm implemented by Maxent 3.3.3k (Phillips et al. 2006) to infer the potential geographic range of L. darwinii at the present and at two time periods during the Quaternary (120 and 21 ka) based on palaeo-climatic layers downloaded from the WorldClim database (Hijmans et al. 2005). Palaeo-climatic models for the Last Inter glacial (120 ka, LIG) followed Otto-Bliesner et al. (2006), while we used two models currently available for download for the Last Glacial Maximum (21 ka, LGM): CCSM and MIROC, both developed originally from the Paleoclimate Modelling Intercomparison Project Phase II (PMIP2).

After downloading, all bioclimatic layers (current and past) were cropped to span from 30°41.28’S to 59°18.78’S and from 76°24.39’W to 55°31.89’W, a larger spatial range than the current Monte Desert and Patagonia distributions. We then generated correlation matrices between the climatic variables with ArcGIS version 10 (ESRI) to remove highly correlated variables from the models (r > 0.85). We ultimately selected nine of the 19 initial climatic variables (bio2 – Mean Diurnal Range, bio3 – Isothermality, bio4 – Temperature Seasonality, bio6 – Min Temperature of Coldest Month, bio8 – Mean Temperature of Wettest Quarter, bio9 – Mean Temperature of Driest Quarter, bio15 – Precipitation Seasonality, bio16 – Precipitation of Wettest Quarter, and bio17 – Precipitation of Driest Quarter). To statistically evaluate model performance, we used the area under the curve (AUC) of the receiver operating characteristic (ROC) plot, a threshold-independent measure of model performance as compared to null expectations.

Because ENMs can produce over-fitted models (often represented by highly complex and irregular variable response curves) and affect model generality and transferability, we did species-specific tuning under the current climatic conditions to increase model performance (Anderson & Gonzales 2011). In the tuning phase, we
explored the strength of protection against over fitting (regularization), which applies a penalty for each term that is included in the model (Phillips et al. 2006). We investigated 12 values for the regularization multiplier (from 0.5 to 6.0, in increments of 0.5) under the autofeatures that allows all the set of feature classes (linear, quadratic, hinge, product and threshold) to depend on the number of presence records. We then retained for the past climate projections the best parameter set that produced smoother distributions of the variable response curves, making them more regular and achieving the highest model performance (lowest values of average differences between training and test AUCs) (Anderson & Gonzales 2011). From the investigated values, the intermediate regularization multipliers led to the best models, with reduced performances at lower and higher regularization (regularization multiplier selected = 4; results shown in Table S3 and Fig. S1, Supporting information). As reported in other studies (Elith et al. 2010), the species-specific smoothing increased model quality and provided more confidence for our projections to the past climate scenarios.

We also explored the Explain tool from Maxent to investigate which variables are driving the reduction in suitability in the palaeo-distribution estimates (see results). Because the Explain tool can only be implemented for additive models, we performed a 10-fold run without the product features for the interactive exploration of predictions. All analyses (including the species-specific tuning) were run with 10 replicates, and in the pictures, we display the pointwise means of the 10 output grids. The occurrence data set included a total of 105 trusted geographic records, and we used a cross-validation testing procedure which divides the occurrence points into training and test points based on the number of replicates (e.g. for the 10-fold cross-validation a test percentage of 10% is used).

**Results**

The reduced cyt b data set (357 bp) for 404 individuals contained 65 haplotypes with a highly variable frequency: 73% of the individuals (295) share the same haplotype (haplotype H1 in cytochrome b network of Fig. S2), while 22 haplotypes have a frequency below 2%, and 42 haplotypes are singletons. We successfully amplified and sequenced between 21 and 28 individuals for each ANL locus (see Table S1, Supporting information). We found no evidence of fixed heterozygote positions or recombination in any ANL, and the percentage of heterozygous individuals ranged between 18% and 96%. The gametic phase of heterozygous individuals was resolved for ~90% of variable positions at the 90%-confidence level. The 10% of positions below this threshold had an average phase-call probability of ~65%. Thus, we used the best allele pairs inferred for each individual in subsequent analyses, which on average had a probability of ~90%, and were consistently inferred between independent runs. We decided to use these best genotype estimates as simulations have shown that omitting unresolved genotypes could underestimate and/or bias parameter estimates (Garrick et al. 2010). The number of ANL haplotypes ranged between 6 and 16 and the number of polymorphic sites between 9 and 26 (Table 1). All haplotype networks show a similar topology: one or two high-frequency haplotypes and multiple, rare haplotypes (Fig. S2, Supporting information).

**Population structure**

The five independent runs in Geneland showed convergence of posterior probability based in burn-in plots and also in the posterior estimates of the number of populations. The mode in the number of populations was five in all runs with an average frequency among MCMC iterations after burn-in of 31% (Fig. S3, Supporting information). The probability maps for the distribution of these five population clusters partition the range of *L. darwinii* into central (clusters 1 and 3), southern (cluster 2) and northern regions (clusters 4 and 5) (Fig. 2). Input and output files of Geneland are deposited in Dryad (see Data Accessibility below).

**Demographic expansion**

The EBSP analysis shows an initial period of stability between 95 and 55 ka followed by exponential growth in population size since ~55 ka until the present (Fig. 3). The final population size is about 50 times higher than the initial population size before the demographic expansion (Fig. 3). The Skyride reconstruction based on cyt b only resulted in a similar pattern of population growth that started ~55 ka and also a very similar final population size (Fig. 3). Five ANL plus cyt b had significant D and Fs values, except for a nonsignificant Fs value for A8E (Table 1). Two additional ANL also have negative, but nonsignificant, D and Fs values (A1D and A12F). Based on mismatch distributions and the raggedness index, the model of demographic expansion was rejected only for three ANL (A6D, A9C and A12C) while the model of spatial expansion was not rejected for any loci (Table 1).

**Range expansion**

The comparison of the SRW and RRW models based on marginal likelihood estimates indicates that the variable diffusion rate model fits the data better than the
constant rate model. The log, marginal likelihood values for the RRW runs were $-3855.8$ and $-3855.2$, and for the SRW runs were $-4275.4$ and $-4274.8$, for path and stepping-stone sampling, respectively. The resulting log Bayes Factor ($-\log_2$ RRW – log$_2$ SRW) $-420$ represents very strong evidence in favour of the RRW model (Kass & Raftery 1995). The selected model was used to reconstruct phylogeographic diffusion patterns based on the MCC tree, which shows the origin of the expansion located in northern Río Negro Province (37.716 S, 67.957 W) and dated at ~95 ka (see supplemental KML file in Dryad; see Data Accessibility below). However, the uncertainty associated with this ancestral location of *L. darwini* is large because the 95%-HPD area ranges between 36.794 and 39.980 ° S latitude and from 66.594 to 69.285 ° W longitude.

The reconstruction suggests three phases in the expansion with initial long-distance colonizations starting at ~85 ka: (i) northwestward, to the four-corner region among Neuquén, Mendoza, La Pampa, and Río Negro Provinces; and (ii) southeastward to central Río Negro (Fig. 4a). During the second phase, which was already completed around 55 ka, there were three additional long-distance colonizations towards: (iii) southwestern Neuquén; (iv) eastern Río Negro; and (v) northeastern Chubut Province (Fig. 4b). In the last phase, contiguous range expansions took place in all regions colonized in the previous phases up to the

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**Table 1** Locus information, samples sizes, summary statistics and GenBank Accession numbers of sequences used in this study

| Locus | N | bp | H | s | k | $\pi$ | Hd | D | Fs | DE | SE |
|-------|---|----|---|---|---|------|----|---|----|----|----|
| cyt b | 404 | 357 | 65 | 59 | 0.596 | 0.00167 ± 0.00014 | 0.466 ± 0.032 | $-2.66019$ | $-28.16556$ | 0.69 | 0.82 |
| A1D  | 56 | 707 | 12 | 23 | 3.589 | 0.00508 ± 0.00092 | 0.597 ± 0.074 | $-0.9025$ | $-0.62422$ | 0.08 | 0.71 |
| A6D  | 56 | 641 | 10 | 10 | 3.244 | 0.00506 ± 0.00021 | 0.655 ± 0.041 | $1.37301$ | $0.2098$ | 0 | 0.25 |
| A8E  | 56 | 625 | 5 | 14 | 1.464 | 0.00236 ± 0.00054 | 0.525 ± 0.053 | $-1.54731$ | $1.73821$ | 0.93 | 0.3 |
| A8F  | 48 | 604 | 8 | 19 | 3.186 | 0.00911 ± 0.00077 | 0.650 ± 0.059 | $0.86834$ | $7.3238$ | 0.12 | 0.5 |
| A9C  | 52 | 481 | 7 | 12 | 4.837 | 0.00932 ± 0.00079 | 0.679 ± 0.049 | $2.02204$ | $4.05001$ | 0.02 | 0.53 |
| A9E  | 44 | 622 | 8 | 9 | 0.914 | 0.00148 ± 0.00035 | 0.499 ± 0.090 | $-1.60038$ | $-8.41053$ | 0.89 | 0.97 |
| A12C | 48 | 682 | 8 | 18 | 4.837 | 0.00716 ± 0.00066 | 0.749 ± 0.036 | $0.6099$ | $2.68505$ | 0.03 | 0.19 |
| A12D | 58 | 687 | 6 | 9 | 2.404 | 0.00350 ± 0.00038 | 0.649 ± 0.043 | $0.64451$ | $2.07738$ | 0.06 | 0.49 |
| A12F | 44 | 715 | 10 | 19 | 3.186 | 0.00448 ± 0.00084 | 0.695 ± 0.060 | $-0.8753$ | $-0.80804$ | 0.15 | 0.59 |
| B5B  | 52 | 646 | 16 | 26 | 7.126 | 0.01110 ± 0.00100 | 0.868 ± 0.033 | $0.77841$ | $1.22323$ | 0.1 | 0.28 |
| B6B  | 54 | 415 | 10 | 20 | 2.106 | 0.00508 ± 0.00108 | 0.542 ± 0.077 | $-1.6391$ | $-1.5793$ | 0.99 | 0.67 |
| B6D  | 50 | 550 | 16 | 18 | 1.92 | 0.00349 ± 0.00058 | 0.765 ± 0.057 | $-1.64329$ | $-8.85054$ | 1 | 0.89 |
| B7D  | 52 | 532 | 12 | 22 | 6.81 | 0.01297 ± 0.00073 | 0.817 ± 0.036 | $1.27839$ | $-2.55038$ | 0.66 | 0.91 |
| B7E  | 54 | 685 | 12 | 13 | 1.636 | 0.00199 ± 0.00028 | 0.709 ± 0.062 | $-1.54735$ | $-6.17009$ | 0.92 | 0.86 |

N, sample size; bp, number of base pairs; H, number of haplotypes; s, number of segregating sites; k, average number of differences between sequences; $\pi$, nucleotide diversity ± standard deviation; Hd, haplotype diversity ± standard deviation; D, Tajima’s D; Fs, Fu’s Fs; DE, Arlequin’s demographic expansion statistic; SE, Arlequin’s spatial expansion statistic.

Values in bold correspond to significant values at the 0.05 probability level.
Fig. 2 Maps of cluster membership and posterior probability for each cluster based on Geneland analyses. The estimated cluster membership represents the modal cluster assignment of each pixel (a) and coloured contours show regions of high (light yellow) to low (red) posterior probability of membership to cluster 1 (b), cluster 2 (c), cluster 3 (d), cluster 4 (e) and cluster 5 (f).

Fig. 3 Variation through time in effective population size based on EBSP analyses (blue), Skyride analyses (green) and in diffusion rate based on RRW analyses (red). Continuous lines above and below the mean parameter values represent 95%-HPD. The vertical dotted lines A, B and C are time slices that correspond with those shown in phylogeographic reconstructions (Fig. 4) and ENM suitability maps (Fig. 5). The two most recent glacial episodes, the Oxygen Isotope Stage 2 (OIS2) and OIS4, are shown as grey bars.
LGM (Fig. 4c) and with little subsequent change in the distribution until today (Fig. 4d).

The mean diffusion rate for the analysis was \(~1108\) km per million years (\(=1.1\) m/year; 95%-HPD: 0.9–1.4), but there was considerable variation over time. Diffusion rates were as high as \(~17\) m/year at the start of rapid range expansion, but declined to <1 m/year during the phase of demographic expansion since 55 ka (Fig. 3). An animation of the geographic diffusion over time is provided as a KML file in the Dryad digital repository for visualization in Google Earth (see Data Accessibility below).

**Palaeo-distribution modelling**

The average test AUC for the 10 replicate runs conducted under each regularization multiplier value in the tuning phase varied very little (from 0.954 to 0.961; Table S3, Supporting information). The average test AUC was 0.961 (standard deviation of 0.013) for the 10 replicate runs that applied the selected regularization multiplier of 4 on the past climate projections, indicating overall high and consistent performance for the models. The clamping grids produced by Maxent have shown that variables were constrained to remain within the range of values in the training data only in the case of 21 ka based on the CCSM model. However, regions of uncertain predictions because of the method of extrapolation are located mostly to the northwest of the suitable areas in this time period and mostly do not affect the core of the prediction range (Fig. S4, Supporting information).

Palaeo-distribution modelling of *L. darwinii* at 120 ka suggests a very restricted distribution in the northwestern portion of the present distribution albeit with moderate predictive power (Fig. 5a). Predictions at the LGM based on the MIROC climatic model suggest an even
smaller range, but those based on the CCSM suggest a southeastern range expansion with high-suitability values (Fig. 5b, c). The ENM at the present time occupies a larger high-suitability area than at the LGM suggesting range expansions towards the north and south, reaching coastal Chubut Province (Fig. 5d).

Results from the analysis of variable contributions (Table S4, Supporting information) indicate that the variables that have higher contribution for the model are bio16 (precipitation of wettest quarter), bio4 (temperature seasonality), bio2 (mean diurnal range) and bio3 (isothermality). The interactive exploration of predictions (Explain tool) indicates that bio3 and bio4 had the most negative contribution to logit of the palaeodistribution estimates with reduced suitability, which means they are the most limiting variables for the presence of L. darwinii (Fig. S5, Supporting information). These results indicate that contrasting between environmental conditions (seasonality and temperature evenness) probably limited L. darwinii distributions during the different past time periods more than other variables that represent mean environmental conditions.

Discussion

Our study takes advantage of novel developments in Bayesian phylogeographic inference that have been applied successfully to many viral epidemics (Lemey et al. 2009b; Raghwani et al. 2011; Lam et al. 2012; Pybus et al. 2012), and a few animal and plant species (Chiari et al. 2012; Escobar-García et al. 2012) to infer the diffusion dynamics of L. darwinii over continuous space and time. In addition to reconstructing the range expansion pattern spatially and temporally, this
approach also allowed modelling the demographic expansion using BSP, effectively decoupling the two components of the expansion process, which were confounded in previous approaches. Instead of a past ‘averaged’ estimate of range distribution, this Bayesian approach provides a dynamic reconstruction that can be directly compared with similar reconstructions based on species-distribution modelling approaches (e.g. ENM). For example, Bayesian diffusion and palaeo-distribution models have been used before to identify potential locations of glacial refugia and range expansions in invertebrates and plants (Smith et al. 2011; Marske et al. 2012).

The combination of these approaches in our study supports the hypothesis that the demographic history of *L. darwinii* was influenced by climatic changes during the late Pleistocene and Holocene. First, both phylogeographic reconstruction and ENM suggest a restricted range of *L. darwinii* in northern Rio Negro Province in the mid LIG at about 95 ka. Despite substantial uncertainty in the estimated ancestral location of *L. darwinii*, which can be enclosed within a square of ~2° of latitude and longitude, it occurs within an area of predicted relative suitability for *L. darwinii* at 120 ka (see Fig. 5a). The high uncertainty is not unexpected because deep phylogeographic histories (i.e. thousands of years) and the lack of serially sampled data close to the root node hamper the ability of this Bayesian method to infer the geographic location of the root with high precision (Lemey et al. 2009a). Second, multiple long-distance colonizations via high diffusion rates increased substantially the distribution of *L. darwinii* to the west and southeast between 95 and 55 ka (Fig. 4b) but without population growth (Fig. 3, see time slices A and B). Subsequently, diffusion rates decreased, but population size began its expansion in a region of high suitability until the LGM (Figs 4c and 5b, c; see time slices B and C in Fig. 3).

Both Bayesian phylogeography and ENM models suggest only modest increases in distribution range after the LGM and up to the present time, but a large increase in population size (Figs 3, 4d, 5d). The direction and timing of range expansion coincided between both approaches as they suggest a major southeastern expansion during the termination of the LIG, but basically no further major increases in range size after the LGM. Independent evidence further corroborates the palaeo-distribution model at the LGM because a fossil of *L. darwinii* dated to the ‘late Pleistocene to early Holocene’ has been described from southern Buenos Aires Province as marginally predicted by the CCSM model (Albino 2005)(Fig. 5c). Consequently, fossil evidence supports the CCSM model (instead of the geographically more restricted MIROC prediction), which as discussed above, is consistent with the range distribution inferred with Bayesian phylogeography at the LGM.

After these phases of range and demographic expansions in the recent past, the current population structure of *L. darwinii* seems to be partitioned into five clusters arranged into a north-to-south axis (Fig. 2). This geographic partitioning probably has an historical component because it appears associated with the biogeographic subdivision of the Monte Desert into districts and subdistricts based on vegetation distribution (Roig et al. 2009). In addition, the altitude and the current climatic zonation of the Monte Desert, in particular, total annual rainfall (Labraga & Villalba 2009), also show gradients in a north-to-south direction that could partially limit gene flow and account for some of the differentiation between the inferred population clusters within *L. darwinii*.

**Demographic history during the Late Pleistocene**

A number of previous studies of lizards and other vertebrates, terrestrial and aquatic, have revealed demographic expansions that took place during the Late Pleistocene but before the LGM. Based on NCPA and test summary statistics, Yoke et al. (2006) found a strong signal of a recent demographic expansion in *Cnemidophorus longicauda*, a lizard that co-occurs with *L. darwinii* in the southernmost half of the Monte Desert. More recently, a time-calibrated BSP analysis showed that the population size of *L. gracilis*, another lizard codistributed with *L. darwinii*, remained relatively stable until ~100 ka when a demographic expansion started without subsequent changes until the Holocene (Olave et al. 2011). The onset of this expansion occurred in the middle of the LIG period known as the Oxygen Isotope Stage 5 (OIS 5, ~70–140 ka) and continued unaffected through the late glaciations of the Pleistocene (OIS 4, 60–70 ka; OIS 2, 15–35 ka), which were less severe than the penultimate glacial episode (OIS 6, ~140–180 ka) (Rabassa & Clapperton 1990; Rabassa et al. 2005, 2011; Rabassa 2008). In contrast, a clade of the lizard *L. petrophilus*, from northern Patagonia and distributed south of the *L. darwinii*’s range, showed demographic stability during the Pleistocene (Fontanella et al. 2012).

Very similar patterns of demographic expansions during the LIG with little or no subsequent effect of late glaciations, especially the LGM, have also been detected with BSP analyses in fish and rodents from the Monte Desert and the adjacent Patagonian steppe. Populations of the fish *Percichthys trucha* in this region started a demographic expansion ~80 ka, near the end of OIS 5, with a slow down during OIS 2 (Ruzzante et al. 2008). In addition, three rodents (*Phyllotis xanthopygus*, *Abrothrix olivaceus* and *Eligmodontia typus*) began...
demographic expansions around ~53–63 ka, during the interglacial OIS 3, and continued growing over the last glaciation OIS 2 (Lessa et al. 2010; Abdur 2011; Da Silva 2011; Riverón 2011). Therefore, the evidence from multiple codistributed taxa in the Monte Desert region suggests congruent demographic stability in the penultimate glaciation, population expansions starting in the LIG and continuing through the LGM, before stabilizing again during the Holocene.

Demographic stability in multiple codistributed species in the Monte Desert is consistent with palaeo-climatic modelling, geological dating and pollen records suggesting colder, drier and windier climates during the cooling phases of the Pleistocene (Clapperton 1993; Rabassa et al. 2005; Labraga & Villalba 2009). Pollen records indicate a dominance of typical Andean and Patagonian steppe, herbaceous taxa including Grammineae and Asteraceae, in the Monte Desert during the LGM (Labraga & Villalba 2009), suggesting that during cooling phases, ranges shifted towards glacial refugia at lower latitudes. The smaller and more northerly restricted range of Larrea at the onset of the LIG (Figs 4a and 5a) also supports the prediction that these climatic and biogeographic shifts occurred multiple times during the Pleistocene, especially during the longer and more intense penultimate glaciation (i.e. OIS 6). Further, the dominance of typical Monte Desert taxa, such as Poaceae, creosote (Larrea) bush and mesquite (Prosopis) trees, in warmer and more humid climates (Labraga & Villalba 2009), is consistent with demographic expansions of Monte Desert species during the relatively long and benign LIG. However, the continuous population growth over OIS 4 and 2 seems an unexpected pattern even though the last glaciations were less severe and shorter than the OIS 6 (Rabassa 2008). This apparent conflict between phylogeographic patterns in multiple species and palaeo-climatic predictions for the LGM could be derived from the assumption of simultaneous spatial and demographic expansions when using BSP and test statistics.

While previous studies have not been able to distinguish between spatial and demographic expansions, we could model and estimate relevant parameters of these processes separately with a Bayesian phylogeography method. In fact, the temporal reconstruction of both processes suggests that they did not occur simultaneously in L. darwinii, which can account for unexpected results seen in previous studies. We found that the demographic expansion in L. darwinii, even during the last glaciations, took place with modest increases in range distribution (Figs 3 and 4b,c). In contrast, L. darwinii expanded its range dramatically during the LIG in spite of a relatively stable population size (Figs 3 and 4a,b). Therefore, the counter-intuitive inference of demographic expansion during unfavourable glacial climates for several Monte Desert taxa could be accounted for a model of population growth without range expansion during the last glaciation. The Bayesian ‘phylodynamic’ approach applied in our study has given novel insights about the independence in terms of timing between range and demographic expansions and their association with Pleistocene climatic events, which has not been possible with other methods in recent studies.

**Diffusion vs. dispersal rates**

The historical estimates of diffusion rates from Bayesian phylogeography provide an opportunity for comparison with known field estimates of dispersal distance in other lizards. Even though the dispersal dynamics of stable vs. expanding populations can be very different, we were interested in exploring whether estimates of diffusion rates were realistic in comparison with the observed dispersal abilities of lizards in the field. In Liolaemus, most capture-mark-recapture studies have actually quantified the home range of adults (Rocha 1999; Frutos et al. 2007; Kacoliris et al. 2009), but at least one study in L. koslowskyi (Frutos & Belver 2007), a member of the L. darwinii group, also measured the shift of the home range between two consecutive breeding attempts (i.e. breeding dispersal; Clobert et al. 1994; Olsson & Shine 2003). Most individuals of L. koslowskyi maintained their home ranges from one year to the next, and males dispersed further than females, an average of ~12 m/year (Frutos & Belver 2007). However, long-term studies of home-range shifts including hatchling/juvenile stages (i.e. natal dispersal), which represent a significant portion of the total dispersal distance during a lifetime, have not been carried out in any species of Liolaemus.

Very few studies have quantified natal and/or breeding dispersal in other lizards, including Sceloporus occidentalis (Massot et al. 2003), Lacerta agilis (Olsson et al. 1997; Ryberg et al. 2004), Lacerta vivipara (Le Galliard et al. 2003), Niveoscincus microlepidotus (Olsson & Shine 2003) and Xantusia vigilis (Davis et al. 2011). Based on a multi-year study, Davis et al. (2011) found larger dispersal distances in juveniles (~50 m) in comparison with adults (~25 m) of X. vigilis, a relatively sedentary, desert-dwelling lizard. In Sceloporus undulatus, an oviparous semi-arboreal species, the mean dispersal distance of hatchlings was also ~50 m (Massot et al. 2003). In Lacerta agilis, a viviparous species, mean dispersal distance of adult females measured as the shift in home-range location was ~10 m (Ryberg et al. 2004). In general, dispersal distances vary among species co-occurring in the same habitats (Hokit et al. 1999), but also within species, due to body size, age, sex
and social interactions (Le Galliard et al. 2003; Massot et al. 2003; Cote & Clobert 2007). Based on the first phase of range expansion, between 95 and 55 ka, mean diffusion rates were as high as 20 m/year, but during the demographic expansion phase (from 55 ka to the present), rates were <1 m/year. Despite differences in habitat and dispersal abilities, our range of diffusion rates for L. darwinii is somewhat lower than the dispersal rates for Xantusia and Sceloporus (albeit of the same order of magnitude), similar to estimates for Lacerta, and especially close to the estimates for the congeneric and closely related L. koslowskyi. A good fit between dispersal rates estimated from current field data and phylogeographic estimates is not expected as dispersal dynamics generally differ between historical range expansions and ongoing range shifts (Phillips et al. 2008). Therefore, based on the comparisons made above, our Bayesian diffusion rates are probably realistic estimates of the historical dispersal dynamics and provide further support to the temporal scale of the phylogeographic reconstruction.

**Relaxed Bayesian diffusion models**

Our model comparisons based on Bayes factors suggest that a time-heterogeneous rate of dispersal fits the data better than a standard, time-homogeneous model. While the geographic area occupied by L. darwinii consists of a relatively homogenous landscape of Monte Desert habitat, the climatic changes during the late Quaternary suggest that favourable conditions for dispersal varied over time. The log-normal distribution used to model the diffusion rate variation is an extremely relaxed model, but it is possible that more restrictive models of over-dispersion, such as gamma-distributed or autocorrelated rates, can also fit better than the standard Brownian model (Lemey et al. 2010). This possibility becomes more plausible given the observed pattern of range expansion, which indicates high diffusion rates at deep branches presumably associated with sudden changes in habitat suitability during demographic history. In particular, autocorrelated models might applied well to the observed diffusion dynamics where some branches in specific parts of the distribution consistently have higher diffusion rates leading to a net southeastward range expansion. Furthermore, simulations and real data from invasive species suggest that dispersal rates in the expanding front are higher than near the origin of expansion due to strong spatial selection on dispersal ability (Phillips et al. 2008), which also makes appealing the implementation of autocorrelated models of diffusion rate in phylogeography. Finally, future developments of Bayesian phylogeographic models are directed towards incorporating spatial heterogeneity via dispersal constraints derived from ENM (Lemey et al. 2010). Indeed, our comparison of phylogeographic reconstructions and ENM further encourages a full integration of these approaches for achieving more biologically realistic models of range expansion and for disentangling this process from demographic expansion.

**Acknowledgements**

We want to dedicate this research work to our close colleague Dr. Natalia Feltrin, who recently passed away on March 7th, in a car accident. She developed her undergraduate honour’s thesis and her PhD project in our research group for the last 8 years; and during that time, we shared many moments, including travels to the US and all around Patagonia, as well as discussions about academia and life in general. Natalia was a happy, optimistic and hard-working person, and a good example of how much one can accomplish with perseverance and team work. Her passing is a profound loss for our group, as we all loved her very much and she will always be with us.

We thank two anonymous reviewers whose constructive comments and suggestions greatly improved an earlier version of the manuscript. We thank B. Noonan for his help in ANL development. A.C. acknowledges a postdoctoral fellowship from the Consejo Nacional de Investigaciones Científicas y Técnicas (CONICET, Argentina), financial support from the BYU Department of Biology and Graduate Studies Office, as well as support from the Society for Systematic Biologists and the Society for the Study of Amphibians and Reptiles. F.P.W. was supported by fellowships from Coordenação de Aperfeiçoamento de Pessoal de Nível Superior (CAPES/Fulbright no. 15073722-269/06-08) and Atração de Jovens Talentos from the Conselho Nacional de Desenvolvimento Científico e Tecnológico (CNPq: 374307/2012-1). L.J.A. and M.M. acknowledge several grants under programmes PIP & IBol (CONICET) and FONCYT PICT (ANPCYT) (Argentina).

Specimens were collected following local regulations and permits issued by Administración de Parques Nacionales, Direcciones Provinciales de Fauna de Río Negro, Chubut, Neuquén, La Pampa, Mendoza and Administración de Areas Protegidas de Neuquén and Mendoza (Argentina). JWS, Jr acknowledges laboratory support from the BYU Jessie Knight Fellowship. Funding for this study was provided by the NSF ‘Partnership for International Research and Education’ award (OISE 0530267) for collaborative research on Patagonian biodiversity, granted to the following institutions (listed alphabetically): BYU, Centro Nacional Patagónico, Dalhousie University, Darwin Botanical Institute, George Washington University, Universidad Nacional de Córdoba, Universidad Austral de Chile, Universidad Nacional del Comahue and the Universidad de Concepción.

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A.C. designed the study, performed phylogeographic analyses, prepared figures and wrote the manuscript. F.P.W. performed palaeo-distribution modelling, prepared figures and wrote the manuscript. M.M. collected field samples and sequenced DNA data. J.W.S., Jr provided laboratory space, supplies, equipment and facilitated BYU computing resources. L.J.A. collected field samples, and provided laboratory space and equipment.

Data accessibility
DNA sequences: GenBank Accession numbers (see Table 1). Voucher and locality information: uploaded as online supporting information.

Alignments of DNA sequences and input/outfiles files from analyses PHASE, TCS, GENELAND, BEAST and MAXENT analyses: Dryad entry doi: 10.5061/dryad.m81b5.

Supporting information
Additional supporting information may be found in the online version of this article.
Fig. S1 Individual variable response curves in Maxent analyses.
Fig. S2 Haplotype network for cyt b, maximum credibility tree (MCC) of cyt b, and haplotype networks of nuclear ANL.
Fig. S3 Burn-in plot of log-posterior probability and the relative frequency of the number of clusters in Geneland analyses.
Fig. S4 Clamping grids of Maxent analyses.
Fig. S5 Interactive exploration of predictions using the 'Explore' tool in Maxent analyses.

Table S1 Voucher information of specimens used in this study, including locality data, population clusters assignments estimated with Structure using cyt b, and haplotype codes for cyt b and ANL used in Figures 1 and S2.
Table S2 Primer sequences of ANL used in this study.
Table S3 Summary of the results from the species-specific tuning in Maxent analyses.
Table S4 Analysis of variable contributions in Maxent analyses.