1. Introduction– 9 years since Avise, 2000

Interpreted literally, the term phylogeography simply means the phylogenetic analysis of organismal data in the context of the geographic distribution of the organism. Even a cursory look at the literature reveals, however, that the meaning of “phylogeography” and the field that it describes is considerably more subtle and far-reaching. The term that launched the field arrived with a flourish in 1987 with the seminal review paper by Avise and colleagues (Avise et al., 1987), which aimed to unite evolutionary biologists in the disparate fields of phylogenetics and population genetics. Reaching back to the work of Dobzhansky (1937) and Mayr (1963), Avise et al. (1987) asked for a formal acknowledgement that “microevolutionary processes operating within species can be extrapolated to explain macroevolutionary differences among species and higher taxa” (p. 489). From our present perspective this connection appears obvious, but at the time of the paper—publication the application of explicitly historical terms and methods was unknown to many empiricists working at or below the species level. This dearth of historical perspective for population genetic analysis has changed radically in intervening years, due in part to the publication of Avise’s monumental book Phylogeography (2000) and the rise of coalescent theory.

As Avise conceived it, phylogeography is the phylogenetic analysis of geographically contextualized genetic data for testing hypotheses regarding the causal relationship among geographic
phenomena, species distributions, and the mechanisms driving speciation. Avise et al. (1987) also implored that genetic data from multiple codistributed taxa could augment investigation of deep-seated questions about the geographic, geological, and/or climatological phenomena that have generated the observed distribution of biodiversity (i.e., comparative phylogeography). This approach offers the opportunity of a natural experiment where focal objects (codistributed populations), have been independently submitted to the same “natural” evolutionary treatments (geologic and climate-change scenarios).

Fundamental to the empirical development of phylogeography was analysis of mitochondrial data at the species level. Mitochondrial DNA (mtDNA) in animals was promoted as the molecular marker of choice due to its lack of recombination, putative neutrality, and smaller effective population size, and consequently a shorter expected time to reciprocal monophyly between geographic regions. But the most salient and revolutionary aspect of graphic regions. Although this direct link between spatial patterning of different clades within a gene genealogy and organismal lineages within species might be conceptually misleading under some demographic histories (Irwin, 2002), it gave phylogeographic analysis its perceived power and appeal. In a practical sense, clades within species were often assumed to reflect the boundaries of populations without accounting for statistical uncertainty under an appropriate statistical model, and phylogeography began to thereby implicitly explore the history of clade-defined lineages within species.

While Avise and colleagues catalyzed the growth of phylogeography, mathematicians and theoretical population geneticists turned population genetics upside down by introducing the mathematical formulation of coalescent theory (Hudson, 1983; Kingman, 1982a,b; Tajima, 1983). Traditional population genetics theory was typically based on forward-looking diffusion equations that allowed one to predict frequencies of alleles in future generations based on modeling entire populations (Ewens, 1979). However, coalescent theory elegantly formalized a powerful way of using only the sample of alleles such that genealogies are modeled backwards in time under virtually any complex demographic history in order to estimate phylogeographic parameters such as historical population sizes, divergence times, and migration rates given the stochastic timing of coalescent events (Wakeley, 2008). Even so, it took some time before statistical coalescent model-based methods percolated into the empirical phylogeography literature. Until statistical phylogeographic methods became more commonly used (Knowles and Maddison, 2002), empiricists generally relied on equating geographical patterns of the gene genealogy with the demographic history, as implied by the word “phylogeography”. Yet in the context of the coalescent, a single genealogy is but one realization of a stochastic process and estimating a single geographically contextualized phylogeny of multiple unlinked loci is often a misleading exercise (Edwards, 2009; Maddison, 1997). Instead of using estimated gene genealogies to directly infer the demographic history, coalescent methods generally treat these genealogies as a transition parameter to obtain estimates of biogeographically informative demographic parameters such as divergence times and migration rates (Hey and Machado, 2003).

To move beyond equating genealogical patterns with biogeographic processes, Avise (2000) acknowledged coalescent theory to be the appropriate statistical and methodological framework for testing phylogeographic hypotheses. One of the insights from post-coalescent phylogeography was that the standard single locus mtDNA or cpDNA data sets that were most commonly collected in the 1990s were often insufficient for obtaining precise parameter estimates and that multi-locus data could dramatically improve the performance of analytical methods derived from the coalescent (Carling and Brumfield, 2007; Edwards and Beerli, 2000; Felsenstein, 2006; Hickerson et al., 2006a). Phylogeography has therefore expanded its focus to nuclear markers (Dolman and Moritz, 2006; Hare and Avise, 1998; Harlin-Cognato et al., 2007; Hurt et al., 2009; Ingvarsson, 2008; Lee and Edwards, 2008; Peters et al., 2008; Rosenblum et al., 2007) with careful attention paid to the resolution of individual nuclear haplotypes (Salem et al., 2005). The coalescent has also become relevant and necessary for phylogeographers to estimate species-level phylogeny at low levels of divergence given data from multiple loci and multiple individuals per subspecies or populations (Edwards et al., 2007; Kubatko et al., 2009; Oliver, 2008).

As intended by the architects of the field, phylogeography is rapidly becoming one of the most integrative fields in evolutionary biology, as different analytical tools have been developed, and of necessity, as the complexity of the hypotheses being addressed has intensified. For example, phylogeographic parameter estimates and model testing can be potentially combined with ecological niche models (Peterson et al., 2002), studies of ecological speciation and radiation, tests of community assembly models, as well as spatial analysis of quantitative trait variation and natural selection. Thus, the field that was originally conceived to bridge two disciplines—phylogenetics and population genetics—has evolved into one of the most integrative disciplines in all of biology. In addition to the initial integration of historical and contemporary genetic analysis, phylogeography borrows from fields as diverse as geospatial analysis, geology, climatology, and computer science, just to name a few. In this review, we will highlight the field’s initial focal areas, touch on analytical advances, and point to exciting future directions integrating the analytical advances with emerging tools to distinguish between different complex historical demographic models while uncovering spatial patterns of adaptation, selection, and community membership. Throughout, we suspect that the reader will observe the many ways that phylogeographic theory and methods are woven together by the conceptual thread first developed by Avise (2000) and Avise et al. (1987).

2. What kinds of questions has phylogeography addressed?

2.1. Single species phylogeographic studies

Perhaps the greatest impact of phylogeographic approaches has been on the most basic of biological questions—what is a species? Species definitions are many and varied, but many of those in use today have some phylogeographic aspects either explicit or implicit in their definitions (Avise and Ball, 1990; De Queiroz, 2007; Templeton, 2001). These can be incorporated into phylogeographic methods of species delimitation and provide operational practicality in the face of real data (Leavitt et al., 2007; Morando et al., 2003; Sites and Marshall, 2004). However, species delimitation and/or DNA-barcoding methods will be most effective when incorporating coalescent stochasticity (Hickerson et al., 2006b; Hudson and Coyne, 2002; Knowles and Carstens, 2007; Pons et al., 2006; Rosenberg, 2003, 2007) or using robust non-coalescent models (Meyer and Paulay, 2005).

Phylogeographic approaches can also identify historical hybridization events, hybrid zones, occurrences of introgression (Gonzalez-Rodriguez et al., 2004; Hewitt, 2001; Swenson and Howard, 2005) and the geographic determinants of isolation. Such insights can be used to generate allopatric speciation hypotheses that are subsequently tested with genetic data in taxa that span the putative isolating barrier. When allopatric speciation is strongly
supported, new methods have been developed to test alternative allopatric speciation models such as peripatric speciation, colonization or vicariance (Hickerson and Meyer, 2008) or vicariance with recurrent migration (Hey and Nielsen, 2004).

2.2. Multi-species phylogeographic studies

By comparing historical patterns of gene flow and divergence among species that overlap in time and space, comparative phylogeography has helped elucidate the relative influence of shared earth history events on current patterns of biodiversity. Although Pleistocene climatic oscillations have been implicated as important contributors to allopatric divergence (Jordan, 1905) and range expansion (Adams, 1905) since the turn of the century, this was also one of the first results in the nascent stages of comparative phylogeographic research (Avise et al., 1987). Investigations have helped predict range shifts arising from future climate change by allowing insights into the dynamics and persistence of communities during Pleistocene climate cycling (Hewitt, 2000, 1996). While some tropical communities appear stable through historical climate-change cycles (Moritz et al., 2000), other studies suggest that temperate and boreal communities experienced severe range fluctuations where codistributed taxa ranges shift congruently (Alsos et al., 2007; Hewitt, 1999) or independently (Carstens et al., 2005a; McLachlan et al., 2005; Wares and Cunningham, 2001) in response to Pleistocene climate cycling.

Some well-developed regional study systems include regions such as the Australian wet tropics (Schneider et al., 1998), southeastern North America (Avise et al., 1987; Solits et al., 2006), Baja California (Leaché et al., 2007; Riddle et al., 2000; Rigosino, 2005), the North Atlantic (Maggs et al., 2008; Olsen et al., 2004), the Indo-Pacific coral triangle (Barber et al., 2000), the Pacific Northwest of North America (Brunsfeld et al., 2001; Carstens et al., 2005a), California (Lapointe and Rissler, 2005), the Pacific Coast of North America (Cook et al., 2001; Jacobs et al., 2004), the neotropical rainforests (Burney and Brumfield, 2009), East Africa (Fjeldsa and Bowie, 2008) and Europe (Hewitt, 2000; Weiss and Nuno, 2006). Notably, southern hemispheric regions have received much less attention (Beheregaray, 2008; Heckman et al., 2007; Victoriano et al., 2008; Yoder et al., 2005). However, a key challenge for comparative phylogeography is the need for developing analytical tools that can be used to evaluate spatial and temporal congruence or incongruence in phylogeographic patterns across multiple species, regardless of the system that these tools are applied to.

Inferences from such regional studies have subsequently helped inform conservation priorities by identifying and delimiting areas with codistributed singular evolutionary histories in order to prioritize geographical units for biological conservation. To date, the operative criteria for defining conservation areas are richness, endemism and phylogenetic diversity (Brooks et al., 2002; Lamoreux et al., 2006; Orme et al., 2005; Spathelfa and Waite, 2007). However, knowledge about where unique evolutionary events or processes occurred have complemented diversity information thereby allowing us to conserve not only extant biodiversity but also the processes that generate this diversity (Moritz and Faith, 1998). Comparative phylogeography thus allows reconstruction of concerted evolutionary changes in codistributed species that have been affected by past climate changes (e.g., glacial and precipitation cycling) in addition to informing conservation planning. This approach will be useful for predicting how climate change will genetically, demographically, and spatially influence regional biodiversity (Ferrier and Guisan, 2006; Taberlet and Cheddadi, 2002; Williams et al., 2007). So far the applications of comparative phylogeography provide a powerful battery for understanding evolutionary history and strengthening conservation efforts, yet this exciting next phase will require the deployment of powerful coalescent model-based statistical methods.

3. The emergence of model-based methods in phylogeography

3.1. Descriptive phylogeographic inference

In the first wave of phylogeography that coincided with the increased use of the polymerase chain reaction, study conclusions were often directly based on qualitative interpretations of each taxon’s single locus genealogy such that the shape of phylogenies, the geographic distribution of lineages, and estimated dates of gene tree branching events, could be used directly to infer the demographic history of each taxon. Under this approach, direct interpretations from geographically contextualized gene genealogies (i.e., geographically restricted monophyly = isolation) were often combined with reporting geographically defined summary statistics (i.e., hierarchically partitioned $F$ statistics). For example, estimates of gene flow or divergence times can be obtained from the geographic patterns in a gene genealogy (Slatkin and Maddison, 1989) or summary statistics that partition genetic differences among and within populations (Nei and Li, 1979; Wright, 1969). Permutation tests on various summary statistics were also commonly used to test for demographic expansion and/or geographic structuring of genetic variation (Excoffier et al., 1992; Fu and Li, 1993; Tajima, 1989). In combining inference from estimated gene genealogies or gene networks with permutations of spatially-explicit summary statistics, the nested clad phylogeographic analysis (NCPA) became the paradigmatic method associated with this era of phylogeography. By testing for an association of geography and patterns in the gene genealogy, followed by using an inference key to interpret the patterns as being linked with particular historical scenarios (Templeton, 1998, 2004), NCPA provided a statistical framework for using gene genealogies to directly infer demographic history. However, coalescent simulation studies from independent researchers have revealed the single locus implementation of this method to be potentially problematic (Knowles and Maddison, 2002; Panchal and Beaumont, 2007; Petit and Grivet, 2002).

While such simulation testing of NCPA has provoked debate (Beaumont, 2008b; Garrick et al., 2008; Knowles, 2008; Templeton, 2004, 2009a,b), it is also a sign that the field is becoming a more statistically rigorous endeavor and that empiricists are coming to recognize that estimating geographical pattern with demographic and evolutionary processes can lead to over interpretation when ignoring coalescent stochasticity in the data (Arbogast et al., 2002; Edwards and Beerli, 2000; Irwin, 2002). Along these lines, phylogeographic studies are increasingly using simulation-based statistical methods that employ an explicit parameterized coalescent model to estimate parameters as well as test alternative a priori historical hypotheses. In this new wave of studies, the demographic history is not directly interpreted from the gene genealogy and therefore the gene genealogy is not the central point of a phylogeographic analysis. Instead the gene genealogy is a transition variable for connecting data to demographic parameters under an explicit statistical coalescent model (Hey and Machado, 2003).

3.2. Model-based statistical phylogeographic inference

Using statistical approaches based on coalescent models for parameter estimation and hypothesis testing has been described as statistical phylogeography (Knowles and Maddison, 2002) and is philosophically consistent with a methodological approach first described by Chamberlin (1890) (reprinted 1965). In scientific disciplines that investigate historical events that cannot be
observed directly or replicated experimentally, scientists have long proceeded by considering competing hypotheses that serve as separate but plausible explanations for a given phenomenon (Chamberlin, 1890 (reprinted 1965)). Under this approach, one hypothesis is identified as more probable relative to others when all available data are more likely under the corresponding model. Chamberlin's epistemological strategy fortuitously dovetails with the statistical phylogeographic approach whereby coalescent theory is used to build statistical models for hypothesis testing under a Bayesian and/or likelihood-based framework. Indeed, under the Bayesian/likelihood-based stratagem each competing hypotheses can be evaluated by fitting the data to each model relative to other models, by way of different decision theoretical methods (e.g., Bayes factors or likelihood ratio tests). Alternatively, under a null hypothesis framework, a model is treated as a null hypothesis whereby the probability of observing data more extreme than the observed data is calculated assuming the null hypothesis is true. Under either approach, any one competing model is never "true" but instead is a useful approximation that should capture the essential features of a demographic history and is most useful if it is somewhat robust to violation of model assumptions (Anderson, 2007; Wakeley, 2004).

These statistical phylogeographic approaches are greatly strengthened when a wide range of plausible models is considered. One way of informing the choice of plausible models is to generate competing models from external sources of data that provide clues about past populations, including where they were located, when they may have become isolated from other populations, or how they were affected by historical events such as glaciations, mountain orogeny, or habitat perturbation. External sources of data may come from packrat middens (Cognato et al., 2003), fossils (Brunhoff et al., 2003), paleo-environmental data (Tribusch and Schonswetter, 2003), palynological data (Brunsfeld et al., 2001) or even ancient DNA (Barnes et al., 2007; Hadly et al., 2004). However, these types of data are not widely available for all systems, and a more generalized approach to generating hypotheses is useful (Richards et al., 2007).

Some models are relatively basic. For example, one might want to test two models, model A that posits that extant populations in the focal taxon arose from a single population that persisted since before the last glacial maximum (LGM), and model B that posits that extant populations descended from two isolated populations that both persisted since before the LGM. For example, in South America, different sets of independent evidence argue for two alternative histories. In one hypothesized history, LGM refugia are distributed mainly in the Coastal Range close to the Pacific (Heusser et al., 1988). In a second hypothesized history suggested by pollen analyses (Markgraf et al., 1995), multiple persistent isolated refugia during the LGM were located on both slopes of the Andes, or in fragmented areas within the ice shield (Allnutt et al., 1999; Marchelli et al., 1998; Pastorino and Gallo, 2002; Premoli et al., 2000; Premoli, 1997).

In one commonly used model-based statistical approach, a summary statistic is calculated from simulated data sets under each model to obtain a distribution of the summary statistic under each respective model. In this scheme, the probabilities of both models are evaluated with respect to the summary statistic calculated from the empirical data (Knowles, 2001). This simulation-based approach has become a useful method in statistical model-based phylogeography (Alter et al., 2007; Becquet et al., 2007; Berthier et al., 2006; Boul et al., 2007; Burridge et al., 2008; Carstens et al., 2005a; DeChaine and Martin, 2005; Eckert et al., 2008; Epps et al., 2005; Hickerson and Cunningham, 2005; Knowles, 2001; Mardulyn and Milinkovitch, 2005; Milot et al., 2000; Moya et al., 2007; Nettle and Dodd, 2007; Pavoine and Bailly, 2007; Spellman and Klicka, 2006; Steele and Storfer, 2006; Thalmann et al., 2007; Vila et al., 2005; Wilson, 2006).

Another common statistical approach is to assume a single model and estimate parameters under the model using full likelihood/Bayesian approaches that make use of all of the data (Beerli and Felsenstein, 2001; Hey and Nielsen, 2004, 2007; Kuhner, 2006; Kuhner et al., 1998). However, these full likelihood-based methods become intractable for complex phylogeographic models that contain many demographic parameters. Fortunately, there are model-based statistical approaches that circumvent this problem, such as composite likelihood which treats polymorphic site as unlinked and calculates the likelihood function accordingly (Nielsen and Beaumont, 2009).

Another promising family of methods called approximate Bayesian computation (ABC) circumvents this problem by bypassing the computational difficulties of calculating the likelihood functions (Beaumont et al., 2002; Pritchard et al., 1999). These “likelihood-free” ABC methods can approximate the posterior distribution of parameters thereby accomplishing estimation of parameters under an array of complex biogeographic scenarios by simulating data from a coalescent model using parameter values that are randomly drawn from the prior distribution (Chan et al., 2006; Cornuet et al., 2008; Estoup et al., 2004; Excoffier et al., 2005; Jobin and Mountain, 2008). Here, the posterior is approximated from simulated data sets that most closely match the observed data set using sets of summary statistics that are identically calculated from each.

The added flexibility and power of ABC is that one can also use it for hypothesis testing in addition to parameter estimation by treating a set of models as a categorical discrete parameter that is estimated such that model testing and parameter estimation are achieved at the same time (Beaumont, 2008a; Carnaval et al., 2009; Fagundes et al., 2007; Francois et al., 2008; Hickerson and Meyer, 2008; Verdu et al., 2009). While the computational shortcut in ABC does not make total use of the data and usually requires choosing summary statistics that are informative about parameters of interest (Sousa et al., 2009) as well as choosing how to combine summary statistics (Hamilton et al., 2005), side-stepping the need for an explicit likelihood frees up the practitioner to use sufficiently complex historical demographic models.

3.3. Comparative phylogeographic inference

Achieving comparative phylogeography's central goals of testing hypotheses about how abiotic and/or ecological processes drive evolution within whole communities (Arbogast and Kenagy, 2001; Avise, 2000; Bermingham and Moritz, 1998), methods for analysis are being developed that go beyond interpreting results from multiple single taxon analyses. One method statistically estimates levels of topological congruence across taxa and then assembles the genetic datasets from different taxa into a single supertree depicting geographic linkages (Lapointe and Rissler, 2005). This tree can then be tested against potential drivers of lineage divergence (e.g., climate) to determine what factors are correlated with concordant genetic breaks.

Another method that is both coalescent model-based and combines intra-specific data sets into a single analysis is an ABC method that employs a hierarchical Bayesian model (Hickerson et al., 2006c). Hierarchical Bayesian models have been used with ABC to incorporate mutation rate variation across loci (Excoffier et al., 2005; Pritchard et al., 1999), and now this strategy has been used to allow for variability in demographic parameters across taxa with a hyper-prior. In this case, each taxon's demographic parameters are drawn from a prior that is itself conditional on a hyper-prior. This allows estimating the degree of congruence across taxa in these demographic sub-parameters (e.g., level of congruence in divergence times or effective population sizes) and/or estimating the degree of congruence in historical demographic models (e.g.,
vicariance vs dispersal) via posterior estimates of hyper-parameters (Hickerson and Meyer, 2008) and hyper-posterior Bayes factors (Kass and Raftery, 1995). Thus hyper-parameter estimates can inform us about biogeographic processes across taxa while demographic parameters are allowed to vary independently within each codistributed taxon.

Comparative phylogeographic ABC methods are in their infancy, yet have so far been used to test for simultaneous divergence times across codistributed taxa in a variety of biogeographic settings (Hickerson et al., 2006c; Leaché et al., 2007; Topp and Winker, 2008; Voje et al., 2009). They have also been used to test for congruence in biogeographic scenarios across taxa such as vicariance or colonization in multiple gastropod taxa that span large portions of the Pacific ocean (Hickerson and Meyer, 2008) as well as testing for congruence in forest refugia models across multiple frog taxa distributed along the South American coastal forest (Carnaval et al., 2009).

4. Future directions for integrative comparative phylogeography

4.1. Ecological niche models

Integrating phylogeography with species range distribution models (i.e., ecological niche models; ENM; see glossary) is showing enormous promise for elucidating how isolation, speciation, and selection are directly or indirectly linked to abiotic factors (Kozak et al., 2008). If environmental factors are implicated in divergence and speciation, such integration can aid in species delineation (Bond and Stockman, 2008; Paxworthy et al., 2007; Rissler and Apodaca, 2007) and testing models of niche conservatism and niche divergence (Losos, 2008; Warren et al., 2008). Another useful application has been to compare phylogeographic historical parameter estimates with ancestral area predictions obtained from contemporary and paleo-ENMs (Hugall et al., 2002; Peterson and Nyári, 2007; Richards et al., 2007; Ruegg and Hijuans, 2006).

Comparative phylogeographic studies are starting to use ENMs from multiple taxa. At first such studies compared paleo-ENMs with phylogeographic parameter estimates across codistributed taxa to evaluate the extent to which codistributed taxa are spatially and temporally concordant. In one of the first of these integrative studies, Carstens and colleagues found that divergence time estimates were similar in a vole and a willow, which the authors attributed to a similar historic distribution and a shared response to climatic change at the end of the Pleistocene (Carstens et al., 2005a). However, the paleo-ENM models for these two taxa were markedly different suggesting recent niche divergence (Carstens and Richards, 2007). Another such study compared paleo-ENMs with phylogeographic inferences from 20 intra-specific mammal and bird studies from North America and found that 14 out of 20 inferences of late Pleistocene refugia were spatially correlated with inferences from ENM models (Waltari et al., 2007). There is also promise in using hierarchical ABC models to explicitly combine comparative phylogeographic data sets with ENM predictions. One such study did this to test environmental niche model predictions that three codistributed frog taxa co-expanded from putative coastal forest refugia after the LGM (Carnaval et al., 2009). In this case, the paleo-ENMs were directly used to construct a priori model such that Bayesian hypothesis testing was used to compare the fit of the genetic data relative to an alternative historic demographic model across all three taxa. Another promising direction will be the development of spatially-explicit coalescent models for more direct integration between ENMs and phylogeographic inference (Barton et al., 2009; Currat et al., 2004).

Nonetheless, it important to acknowledge that paleo-ENMs only predict the potential ranges under the assumption of niche conservatism (Araújo and Guisan, 2006), and therefore model-based phylogeographic inference can be a way to test competing ancestral range predictions constructed under the alternative assumptions of niche stasis and niche liability. Although ENMs are a recently developed application with various sources of error (Araújo and Guisan, 2006; Lozier et al., 2009; Stockwell and Peterson, 2002), this enterprise is rapidly advancing and we are bound to see important advances in the near-future.

4.2. Studies of natural selection

Although studies of natural selection could be integral to phylogeographic inference, they are usually omitted, perhaps due to the field’s early emphasis on neutral genetic variation and many of the subsequent analytical approaches that were developed around such assumptions. However, it is clear that selection plays a central role in generating biodiversity (Funk et al., 2006), and thus should be an important integrative component in future comparative phylogeographic studies. As genomic data become available for non-model organisms (Vera et al., 2008), comparative phylogeographic studies will allow identification of different locus-specific divergent selection patterns (Ellegren and Sheldon, 2008; Luikart et al., 2003; MacCallum and Hill, 2006) between pairs of codistributed taxa or taxa that co-occur along the same geographic gradient (Bonin et al., 2006; Joost et al., 2007, 2008) while also testing various multi-taxa demographic historical scenarios. This could involve identifying cases of simultaneous expansion into novel environments accompanied by congruent or incongruent patterns of selection or drift across suites of loci (Gavrilo, 2004). Such large-scale analysis involving multiple taxa, multiple individuals and 100’s of loci will indeed bring serious analytical challenges ranging from data management to testing complex highly parameterized models of selection and demography, yet methodological advances such as ABC or composite likelihood should allow such studies to be tractable in the near-future (Nielson and Beaumont, 2009).

There are also now substantial opportunities to integrate studies of natural selection via population genomics and multi-species phylogeography with studies of quantitative genetics (Stinchcombe and Hoekstra, 2008). By comparing variation in quantitative traits with neutral genetic variation (Falconer and Mackay, 1996), and incorporating classic quantitative genetics experiments (Wright, 1943) drawing from populations across their geographic ranges and environmental gradients, investigators have another means of detecting selection in a biogeographic context. This could not only include using direct comparisons of quantitative trait variation with neutral genetic variation using Qst—Fst comparisons (McKay and Latta, 2002), but also include looking at eco-physiological trait differentiation and selection across geographic or climatic gradients (Cavender-Bares, 2007; Dudley, 1996; Etter, 2008, 2004; Wright and Stanton, 2007). Looking further ahead, model-based phylogeographic inference tools can be used to relate divergence times and migration with levels of ecologically driven selection at genetic loci and quantitative traits. Furthermore, ENMs can be used to quantify and test for environmental divergence between pairs of taxa that are hypothesized to have undergone divergent selection at quantitative traits, genetic loci as well as geographic isolation (Nakazato et al., 2008; Warren et al., 2008).

4.3. Ecological speciation

Not only should researchers be able to uncover the genetic determinants of species boundaries (Orr et al., 2004) using genetic markers believed to be associated with reproductive isolation (Palopoli and Wu, 1994; Wu and Davis, 1993), there is now great promise for comparative phylogeographic studies to unravel pro-
cesses behind ecologically driven speciation (Funk et al., 2006; Mendelson and Shaw, 2005; Rundle and Nosil, 2005; Schluter, 2000) and the genomic basis of selection driving these patterns of ecological speciation (Egan et al., 2008; Rogers and Bernatchez, 2007; Scluter, 2009; Wood et al., 2008). A ‘unified framework’ for testing ecological speciation hypotheses can be deployed using divergence time estimates across sister taxon-pairs (Funk, 1998; Nosil et al., 2009). Ecologically driven selection is widely found in nature, yet cases that entail full reproductive isolation are less common and depend on either the strength of the divergent selection at single traits or the number of traits at which selection is acting (Nosil et al., 2009). When traits under divergent selection can be identified in multiple sister taxon-pairs that are co-occur along an environmental gradient or have co-expanded into a novel environment, one could use model-based methods such as hierarchical ABC to initially subdivide these sister-pairs across different temporal waves of isolation via colonization and/or geographic expansion (Leaché et al., 2007). The resulting groupings of taxon-pairs defined by isolation times could then be used within a multiple regression framework to estimate strength of selection at genetic loci, phenotypic traits as well as estimate divergence or stasis at ecological niche dimensions between the sister taxon-pairs.

If timing of isolation is positively correlated with reproductive isolation (Funk et al., 2006), one could then test if timing of isolation is also correlated with the multifarious selection hypothesis where the completeness of reproductive isolation is positively correlated with the numbers of loci subject to selection (Seehausen et al., 2008). Alternatively, one could test if the time since isolation or subsequent gene flow is correlated with the magnitude of selection at single genetic loci across taxon-pairs and thereby more consistent with the “stronger selection” hypothesis (Nosil et al., 2009). Furthermore, if subsequent gene flow rather then time is positively correlated with the magnitude of selection, then stronger selection with reinforcement can be implicated across a subset of taxon-pairs (Hoskin et al., 2005; Howard, 1993). These inferences of ecological speciation, selection and timing of isolation could then be compared to estimates of divergence along an array of environmental gradients or ecological niche dimensions using ENMs (Kozak and Wiens, 2007; Seehausen et al., 2008; Warren et al., 2008).

4.4. Integrating comparative phylogeography with studies of community assembly

One of the original objectives of comparative phylogeography was to resolve deep-seated questions about how climate change drives community assembly and evolution within whole biotas (Avise et al., 1987). Although inter-specific phylogenetic data is increasingly being used to address questions of community assembly (Cavender-Bares et al., 2009; Emerson and Gillespie, 2008; Jabot and Chave, 2009; McPeek, 2008; Vamosi et al., 2009; Webb et al., 2002), using comparative phylogeographic data for such purposes has so far been handicapped because such studies rarely involve more than a handful of codistributed taxa. However, comparative phylogeographic datasets are bound to have explosive growth as collecting DNA sequence data across a wide diversity of taxa will soon scale up to the level of comprehensive ecosystem sampling. Such “community-scale” comparative phylogeographic data sets could potentially test classic biogeographic hypotheses (e.g., vicariance versus dispersal) at the community level (Carlquist, 1966; Rosen, 1978), as well as test controversial and fundamental hypotheses of community assembly such as Hubbell’s neutral theory (Hubbell, 2001), Tilman’s stochastic competitive assembly model (Tilman, 2004), and Diamond’s niche assembly rules (Diamond, 1975; Gotelli and McCabe, 2002). As comparative phylogeographic datasets grow to include >100 codistributed taxon-pairs, the computational advantages of the hierarchical ABC approach will be well suited to test community assembly models that are explicitly defined by temporal patterns of dispersal and speciation. Although species interactions may not be explicitly incorporated into such tests of community assembly models, hierarchical ABC could allow combining summary statistics from simulated phylogeographic data, phylogenetic data and species abundance data (Jabot and Chave, 2009) to estimate parameters under various community assembly models as well as compare the relative likelihoods of these models. Another promising approach for potentially integrating comparative phylogeography with studies of community assembly is using fossil pollen data. One study used this approach to infer historical range expansion of major tree species after the last North American glaciation to understand the rate at which species can move in response to climate change and the extent to which communities reassemble after disturbance (McLachlan et al., 2005). Finally, researchers will be able to combine these inferences with studies of selection, reproductive isolation, phenotypic evolution as well as inferences of niche conservatism and niche evolution using ENMs.

4.5. Empirical systems

Given this exciting opportunity to co-estimate divergent selection patterns and demographic histories across codistributed non-model taxa, what are some candidate empirical systems? In the same way that “model organisms” were carefully chosen for whole genomes, there are a plethora of interesting “model communities” that stand to benefit from this new wave in comparative phylogeography. At the broadest scale, such techniques could be deployed to investigate how latitudinal patterns in biodiversity arise. Selection driven by environmental gradients has been implicated as the cause of biogeographic diversity patterns in the tropics (Moritz et al., 2000) and using emerging genomic tools for comparative phylogeography will provide a means to better test this controversial hypothesis as well as examine how selection drives biodiversity patterns in temperate regions that have experience cyclical range expansions throughout the Pleistocene. Such studies will be critical for identifying and preserving geographic patterns of endemism (Crandall et al., 2000; Moritz, 2002), just as determining whether populations are differentially adapted is an important issue in conservation biology and restoration ecology (McKay et al., 2005; Wilkinson, 2001). For instance, the geography of adaptive genetic variation may be more relevant for the conservation of locally adapted ecotypes than that of neutral variation (Conover et al., 2006).

These new model-based comparative phylogeographic techniques could also be used to test patterns of co-speciation in multiple taxon-pairs of plant hosts and their insect mutualists or parasites. For example, phylogenetic data from multiple western Palearctic oak and gallwasp species suggests that host parasite relations are persistent and ancient (Stone et al., 2009), and hierarchical ABC methods could be used to estimate temporal patterns of Pleistocene co-expansion from Asia into Europe. Subsequently, one could further investigate how timing of expansion in the host and parasite taxon-pairs relates to levels of selection, numbers of loci under selection, phenotypic divergence, and divergence along ecological niche dimensions. In studies of ecological speciation, excellent empirical systems for deploying hierarchical ABC methods to control for time when uncovering processes behind the ecological, physiological and environmental factors behind adaptive divergence and reproductive isolation include Andean tomato species (Nakazato et al., 2008), stickleback stream-lake and ocean freshwater taxon-pairs (Hendry and Taylor, 2004), Trinidadian guppies (Crispo et al., 2006), sister-pairs of walking-sticks living on different host plants (Nosil, 2007), African cichlids (Seehausen et al., 2002), and the evolution of Palearctic oak and gallwasp species (Cavender-Bares et al., 2009; Emerson and Gillespie, 2008; Jabot and Chave, 2009).
et al., 2008), Gambusia fishes (Langerhans et al., 2007) and Heliconius butterflies (Jiggins et al., 2004).

5. Conclusion

Nearly 10 years after Avise (2000) and 20 years after Avise et al. (1987), phylogeography is entering a new and exciting phase. With regard to data analysis, the field is rapidly moving from descriptive methods and into using coalescent models for parameter estimation (Kuhner, 2008), a priori model testing (Beaumont, 2008a; Carstens et al., 2005b; Fagundes et al., 2007), the estimation of spatially-explicit demographic histories (Lemmon and Moriarty Lemmon, 2008), and testing for temporal and/or spatial congruence across codistributed taxa (Carnaval et al., 2009; Hickerson et al., 2006c). Concurrent with these methodological advances will be new ways to visualize phylogeographic data (Kidd and Ritchie, 2006), a rapid growth in genomic phylogeographic data (Townsend et al., 2008; Vera et al., 2008), expansion of taxa included in comparative phylogeographic data sets (Avise, 2008) as well as advances allowing analysis of ancient DNA (Hofreiter, 2008) such that researchers can potentially resolve if humans were the causal agents in the extinction of large mammals at the last glacial maximum rather than climate changes (Barnes et al., 2007).

Nonetheless, deciding which set of models to evaluate and use for parameter estimation remains to be a substantial challenge given the bewildering number of possible histories underlying any phylogeographic data set. Moreover, recent understanding of Pleistocene and Holocene climatic history suggests short-lived and extreme climate fluctuations with consequent cyclical changes of habitat suitability occurred with high frequency on timescales of centuries and millennia (Ditilvesen et al., 1996; Mayewski et al., 2004). Although phylogeographic models incorporating sufficient spatial complexity and frequency of these recurrent range expansions/contractions will require further analytical innovations, the rapid development of flexible methods should rise to the task. However, it should be recognized that tools for complex model-based inference that can be easily used by empiricists such as DIY-ABC (Cornuet et al., 2008) is presently limited by the paucity of software packages.

Coinciding with these methodological advances is the explosion of opportunities for integrating comparative phylogeographic data with other fields that are also rapidly advancing. This includes ENMs (Waltari et al., 2007), spatial analysis of genomic signatures in natural selection (Jost et al., 2007), spatial analysis of morphological and functional trait evolution, studies of ecological speciation (Schutler, 2009), and studies of community assembly that make use of ABC (Jabot and Chave, 2009). With this potential interdisciplinary synthesis, comparative phylogeography is poised to achieve what Avise and colleagues originally envisioned - resolving deep-seated puzzles about how climate, geography, and ecological interactions determine and interact with community composition and evolution.

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**Glossary**

**Approximate Bayesian computation:** This provides approximations of posterior probability estimates of model parameters and/or models based on the closest between summary statistics of the observed data and data simulated across the prior distribution of model parameter space.

**Coalescent theory:** Probabilistic model underlying gene genealogies within a population or populations. Coalescent theory provides the statistical framework to build virtually any arbitrary complex historical demographic model with parameters such as migration rates, population sizes, divergence times, recombination and selection.

**Community assembly:** The dynamics and processes underlying community composition via invasion, colonization, dispersal, speciation, and extinction.

**Ecological niche models:** Alternatively referred to as biomimetic modeling or species range distributional modeling, ecological niche models (ENMs) predict the occurrence of species on a landscape from geo-referenced site locality data and sets of spatially-explicit environmental data layers that are assumed to correlate with the species’ range (Peterson et al., 2002).

**Gene genealogy:** Inheritance relationship between sampled homologous gene copies that can be depicted as a branching tree that arises within a population or set of populations of individuals. Instead of using estimates of gene genealogies to directly infer a species phylogeny or species demographic history, coalescent model-based methods use gene genealogies as a transition variable for testing models and making model parameter estimates from data (Hey and Machado, 2003).

**Hierarchical Bayesian model:** A model in which a set of parameters are drawn from a prior distribution that is itself conditional on and drawn from a hyper-prior distribution.

**Model/hypothesis testing:** Decision theoretical approach to evaluate the fit of the data to a model. In Bayesian statistics the fit of the data to a model is evaluated relative to other models. In frequentist statistics, a decision is made about the null hypothesis; that is, if the null hypothesis is true, what is the probability of observing data that is more extreme than the observed data.

**Parameter estimation:** A method for approximating unknowable quantities (i.e., parameters) with data under a probabilistic model. In Bayesian statistics, the probabilistic model describes the prior distribution of the parameters. Bayesian posterior distribution estimates include the most likely values of parameters as well as the level of statistical uncertainty under the model. In population genetics and phylogeography, unobserved parameters often include effective population size, migration rates, recombination and magnitude of population size change.

**Phylogeographic model:** A probabilistic model that explicitly describes how demographic parameters affect DNA polymorphism data. Models are approximations of actual demographic history, but one must explicitly specify the connection between demography and genetic data through a model or set of models, in order to estimate parameters or statistically test alternative models differing in their parameters (i.e., hypothesis testing). Although any model will never be absolutely “true”, useful models contain the essential features of the demographic histories that are of interest.

**Summary statistic:** A mathematical function that condenses the observed data into a numerical metric. In population genetics and phylogeography, summary statistics quantify patterns in DNA polymorphism within and between populations and are most useful if they convey information about and correlate with unobserved parameters such as effective population size, mutation rates, migration rates, recombination rates and magnitude of population size change.