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Genetic Variation in Past and Current Landscapes: Conservation Implications Based on Six Endemic Florida Scrub Plants

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If genetic variation is often positively correlated with population sizes and the presence of nearby populations and suitable habitats, landscape proxies could inform conservation decisions without genetic analyses. For six Florida scrub endemic plants (Dicerandra frutescens, Eryngium cuneifolium, Hypericum cumulicola, Liatris ohlingerae, Nolina brittoniana, and Warea carteri), we relate two measures of genetic variation, expected heterozygosity and alleles per polymorphic locus (APL), to population size and landscape variables. Presettlement areas were estimated based on soil preferences and GIS soils maps. Four species showed no genetic patterns related to population or landscape factors. The other two species showed significant but inconsistent patterns. For Liatris ohlingerae, APL was negatively related to population density and weakly, positively related to remaining presettlement habitat within 32 km. For Nolina brittoniana, APL increased with population size. The rather weak effects of population area/size and both past and current landscape structures suggest that genetic variation needs to be directly measured and not inferred for conservation planning.

1. Introduction

Protecting plant genetic variation is a conservation goal [1], as genetic variation reflects phylogenetic and population history, which is associated with fitness and evolutionary potential [2, 3], and can influence ecosystem processes [4, 5]. While neutral genetic variation is not directly linked to quantitative genetic traits [6], it may be significantly correlated to fitness [7]. If a large percentage of population-level genetic variation (as measured by values such as expected heterozygosity and alleles per polymorphic locus) could be predicted from such ecological and landscape measures as population size, isolation, and nearby suitable habitat, then conservation decisions intended to protect genetic variation could be made without the need for collection of costly and time-consuming molecular data.

This study focuses on several factors potentially affecting within-species genetic patterns. At the species level, genetic variation tends to be relatively low for endemics as compared to broadly distributed species [8]. Endemics of the southeastern U.S. tended to have higher genetic diversity at both species and population levels than endemics in general [9]. Within species, population-level genetic variation is affected by life history, being slightly higher in outcrossing species [8]. A recent meta-analysis shows that habitat fragmentation tends to reduce genetic variation, with similar effects sizes for several measures of genetic variation [10]. More specifically, some (but not all) individual studies have shown greater intraspecific genetic variation in larger [3, 11–15], denser [16], or less-isolated [17–20] populations. Other patterns include lower genetic variation in degraded and logged habitats [21], clines in allele frequencies [22, 23] discontinuities among geographically-separated portions of the range [18, 24] or between organisms separated by landscape barriers [25], higher variation in unglaciated portions of species’ ranges [26], higher genetic variation in core versus marginal populations [27–30], and differences in genetic variation between distinct habitats (prairie versus alvar) [31]. Studies
considering the effects of original habitat patch size, current
patch size, population area, and areas of nearby suitable
habitat within the landscape on local genetic variation are
less common but also show positive relationships [32, 33].
In general, habitat fragmentation tends to decrease genetic
variation of remaining populations [34].

For the endemic-rich Florida scrub [35, 36], such genetic
patterns have been investigated for animals [37, 38] but not
for plants. Florida scrub is a pyrogenic shrubland dominated
by long-lived woody plants, although most of its endemic
species are shorter-lived subshrubs and herbs [39]. Previous
genetic studies of a range of Florida scrub species have
focused on comparisons of species or varieties without
consideration of landscape structure [24, 40–47]. In this
paper, we consider the effects of landscape structure on
genetic variation for six species of plants in Florida scrub
with different spatial distributions and life histories.

Florida scrub species were probably never distributed in a
continuous fashion within their ranges due to specializations
for soil type [54], fire regime [39], canopy gaps [55], and
soil drainage [56], as well as the stochastic nature of their
metapopulation dynamics [57]. Habitat for these scrub
endemics has likely been naturally patchy to different degrees
due to low ocean levels during the Pleistocene [58]. However,
recent agricultural and urban developments have decreased
extent and connectivity of Florida scrub. Only 13% of the
original xeric uplands on the Lake Wales Ridge remain (for
2003, calculated from [59]), down from the estimate of 36% made roughly 20 years earlier (calculated from [60]). Despite
this ongoing loss, effects of habitat loss and fragmentation on
genetic variation of Florida scrub plants have not previously
been examined.

The study six species Warpia carteri Small (Brassicaceae)
[52, 61], Eryngium cuneifolium Small (Apiaceae) [49],
Hypericum cumulicolata (Small) W.P. Adams (Clusiaceae) [50],
Dicerandra frutescens Shinners (Lamiaceae) [48], Nolina
brittoniana Nash (Agavaceae) [44], and Liatris ohlingerae
(S.F. Blake) B.L. Rob. (Asteraceae) [51] vary considerably in
demography, reproductive biology, and habitat requirements
(Table 1). Like many studied Florida scrub plants, the six
subjects of this paper show variation in edaphic specialization:
D. frutescens and E. cuneifolium are soil specialists, while
L. ohlingerae, H. cumulicolata, N. brittoniana, and W. carteri
have more generalized soil preferences [8]. Seed dispersal is
passive for most of our study species, and dispersal distances
tend to be short [44].

Our prior isozyme genetic studies have focused largely
on genetic differences among species. We found that five of
these six species (D. frutescens, E. cuneifolium, H. cumulicolata,
N. brittoniana, and W. carteri but not L. ohlingerae) had lower
genetic variation than that found in comparable endemic
species elsewhere. Values of \( H_e \) ranged from 0.12 for L.
ohlingerae to less than 0.06 for E. cuneifolium, D. frutescens,
W. carteri, and H. cumulicolata [44]. Population differentiation
\( (F_{ST}) \) was particularly high for H. cumulicolata (0.72) and
ranged from 0.03 to 0.44 for other species [44].

Among the six species, levels of genetic variation were
higher in species with greater demographic stability but
were not related to geographic range size, habitat specificity,
or seed dispersal distances [44]. Nolina brittoniana and W.
carteri exhibited clines in allele frequencies along the major
(north–south) axis of their ranges [47, 53].

In this paper, we test whether within-species population-
level genetic variation is related to variables such as popula-
tion size and nearby available habitat. We define available
habitat by suitable soils (see below). We use geographic
information systems (GIS) to summarize and analyze land-
scape patterns of current populations, and the locations
and sizes of presettlement and extant soil patches that
support, or could support, each of the six species. We then
relate these variables to patterns of population-level genetic
variation determined from allozyme electrophoresis, using
new analyses of previously published data [43, 44].

In developing the appropriate data for this suite of
species, we hope to both inform current conservation
programs for Florida scrub and develop more general
conclusions to support decisions for other suites of narrowly
endemic plant species that have received less examination.
While smaller scale processes undoubtedly contribute to
patterns in observed genetic variation [62], these landscape
scale features are often used as surrogates for identifying pop-
ulation targets for conservation efforts [63–65]. Additional
empirical studies of multiple species in the same landscape
are needed to confirm the validity of this approach [87].

We hypothesize that the six Florida scrub species under
study will have greater genetic variation in larger populations
on larger habitat patches (both extant and presettlement),
in populations with more past or extant suitable habitat in
the vicinity, and in presettlement patches that have retained
larger proportions of intact area. Finally, because current
genetic patterns may not yet be in equilibrium with land use
changes, we hypothesize that genetic variation will be more
closely related to presettlement than to extant landscape
structure.

2. Methods

2.1. Sites, Population Sizes, and Tissue Collections. Sites for
tissue collections and population size estimates were chosen
during distributional data from the Florida Natural Areas
Inventory and Archbold Biological Station. We chose sites
across the species ranges (Figure 1), collecting tissue from
populations separated by at least 1.7 km. Over a three year
period, we sampled 164 populations (13–48 per species;
Table 2) from sites across five counties in central Florida
(Highlands, Lake, Orange, Osceola, and Polk) and three
ridges (Lake Wales, Winter Haven, and Orlando). This
sample included the majority of the known occurrences of
most study species. Most samples come from the Lake Wales
Ridge [66] in Highlands and Polk counties where there are
more protected sites [59, 67].

We estimated population size using a variety of tech-
niques. Smaller populations (generally <200 individuals)
were fully censused. We sampled larger populations using
belt transects, 2–4 m wide, placed in stratified random
fashion to cover the entire area of the population. This
sample was then extrapolated to the total population size
Table 1: Biological information on our study species.

<table>
<thead>
<tr>
<th>Species</th>
<th>Life Span and Life Form*</th>
<th>Postfire Recovery</th>
<th>Habitat</th>
<th>Primary Pollinators (distance)</th>
<th>Mating System</th>
<th>Citations</th>
</tr>
</thead>
<tbody>
<tr>
<td>Dicerandra frutescens</td>
<td>SL, PSS</td>
<td>Seed Bank</td>
<td>Florida scrub</td>
<td>Bee-Fly (long)</td>
<td>Mixed</td>
<td>[48]</td>
</tr>
<tr>
<td>Eryngium cuneifolium</td>
<td>SL, PH</td>
<td>Seed Bank</td>
<td>Florida scrub</td>
<td>Generalist Insects (variable)</td>
<td>Mixed</td>
<td>[49]</td>
</tr>
<tr>
<td>Hypericum cumulicola</td>
<td>SL, PH</td>
<td>Seed Bank</td>
<td>Florida scrub</td>
<td>Bees (short)</td>
<td>Mixed</td>
<td>[50]</td>
</tr>
<tr>
<td>Liatris ohlingerai</td>
<td>LL, PH</td>
<td>Resprout</td>
<td>Florida scrub</td>
<td>Butterflies (long)</td>
<td>Outcrossing</td>
<td>[51]</td>
</tr>
<tr>
<td>Nolina brittoniana</td>
<td>LL, PH</td>
<td>Resprout</td>
<td>Sandhill, Florida Scrub</td>
<td>Generalist Insects (long)</td>
<td>Outcrossing</td>
<td>[44]</td>
</tr>
<tr>
<td>Warea carteri</td>
<td>A</td>
<td>Seed Bank</td>
<td>Sandhill, Florida Scrub</td>
<td>Generalist Insects (variable)</td>
<td>Mixed, High Self</td>
<td>[52, 53]</td>
</tr>
</tbody>
</table>

* SL: short lived (<10 years), LL: long lived (>10 years), PSS: perennial subshrub, and PH: perennial herb, A: annual.

Table 2: Variables significant (P < .05) in predicting two measures of genetic variation, Alleles per Polymorphic Locus (APL) and Expected Heterozygosity (Hₑ), in multiple regressions. Nₑ refers to number of populations, Nᵢ refers to mean number of individuals sampled and scored per population. Variables were variously transformed before correlation analysis (see text). Beta is the standardized regression coefficient, P refers to significance in model with all significant variables included.

<table>
<thead>
<tr>
<th>Species</th>
<th>Nₑ</th>
<th>Nᵢ</th>
<th>Fₑa</th>
<th>Genetic Variation Measure</th>
<th>Mean</th>
<th>Predictor Variable</th>
<th>Beta</th>
<th>P</th>
</tr>
</thead>
<tbody>
<tr>
<td>Dicerandra frutescens</td>
<td>13</td>
<td>29</td>
<td>0.030</td>
<td>APL, Hₑ</td>
<td>3.58</td>
<td>None</td>
<td>None</td>
<td>None</td>
</tr>
<tr>
<td>Eryngium cuneifolium</td>
<td>16</td>
<td>30</td>
<td>0.445</td>
<td>APL, Hₑ</td>
<td>2.00</td>
<td>None</td>
<td>None</td>
<td>None</td>
</tr>
<tr>
<td>Hypericum cumulicola</td>
<td>34</td>
<td>28</td>
<td>0.724</td>
<td>APL, Hₑ</td>
<td>2.02</td>
<td>None</td>
<td>None</td>
<td>None</td>
</tr>
<tr>
<td>Liatris ohlingerai</td>
<td>28</td>
<td>22</td>
<td>0.120</td>
<td>APL, Hₑ</td>
<td>2.43</td>
<td>Density</td>
<td>−0.087</td>
<td>.003</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Extant: 32 km</td>
<td>+0.250</td>
<td>.033</td>
</tr>
<tr>
<td>Nolina brittoniana</td>
<td>48</td>
<td>25</td>
<td>0.411</td>
<td>APL, Hₑ</td>
<td>2.53</td>
<td>Population Size</td>
<td>+0.488</td>
<td>.009</td>
</tr>
<tr>
<td>Warea carteri</td>
<td>23</td>
<td>25</td>
<td>0.304</td>
<td>APL, Hₑ</td>
<td>2.30</td>
<td>None</td>
<td>None</td>
<td>None</td>
</tr>
</tbody>
</table>

by estimating the area. Population edges were determined by inappropriate habitat for the species, or by a ≥ 50 m break between plants. For the clonal Nolina brittoniana, we considered individuals as rosettes that were at least 1 m apart, based on patterns of allozyme genotypes [47].

Leaves (or flower buds in the case of H. cumulicola) were collected for allozyme analysis at the same time population size estimates were made. For populations of fewer than 30 plants, leaves or buds were collected from all individuals. For larger populations, we collected leaves from 30 individuals randomly selected but stratified by spatial location. Thus, variation in sample size reflects variation in observed population size, not in sampling effort. Sample sizes for numbers of individuals sampled and scored per population were generally >20 (Table 2). Standard genetic parameters, including percent of polymorphic loci, number of alleles per locus, and expected heterozygosity, were calculated for each population, using the software package Biosys [68]. We also hand calculated the number of alleles per polymorphic locus. Details of allozyme procedures and results for individual species are given in several papers [43, 47, 69].

2.2 Defining Suitable Soil Types. Soil properties are an important dimension of the habitat for many species [70, 71]. For these Florida scrub species, soil requirements are well known and quite specific [54]. We defined suitable soil types based on the intersection of known species locations with digital (1:24,000 scale) soil maps prepared by the Soil Conservation Service [54]. All the suitable soils are
hyperthermic, uncoated Typic or Spodic Haplaquolls [72]: sandy, low-nutrient soils that vary in soil color and distance from the water table. *Eryngium cuneifolium, H. cumulicola,* and *L. ohlingerae* were found primarily on white (St. Lucie and Archbold sands) and gray (Satellite, Duette, Daytona, and Pomello sands) sands [54]; these soils were defined as suitable in the current study. The gray sands have a higher water table than the white sands. Likewise, *D. frutescens* is a specialist for yellow sands with a deep water table (Paola, Astatula, Tavares, and Orsino sands) [54] and these soils were considered suitable habitat. *Warea carteri* is a soil generalist [54], with yellow and gray sands suitable, and *N. brittoniana* occurs over a wide range of soil types [54], with white, gray, and yellow colored sands suitable. Less than 20% of

Figure 1: Minimum convex polygons for distributions of six study species on the Lake Wales Ridge in central Florida.
the known occurrences, by species, fall outside these soil classifications [54].

For this analysis, we defined suitable habitat entirely based on soil preferences. Weather data (temperature, precipitation) are available on a km scale, but soil divisions occur on a much finer (m) scale. In addition, temperature and precipitation variation across our study area is minimal. For example, maps of 17 hydrologic variables (rainfall, humidity, and evaporation) show less than 5% variation over the study area [73]. Similarly, this area shows very little variation in temperature variables [74]. Fire history also affects whether a habitat patch is suitable for individual scrub species [48, 54] but it was not available for all our study sites. Likewise, colonization history would affect which habitat patches would be out of dispersal range, but we lack these data as well.

2.3. Analytical Methods. We used ARCINFO 9.1 software and aerial photographs to create a number of spatial coverages for each species: the populations from which genetic collections were made, extant home polygons, and other polygons with suitable habitat. The home polygon was defined as the merger of the soil polygon from which the collection was made with all adjoining extant suitable soil polygons. For example, if the home polygon is a polygon of Archbold sand, we merged it with adjoining polygons of the similar St. Lucie white sand, both of which are suitable habitat for white sand species. Likewise, we merged adjoining soil polygons with similarly suitable habitat across the landscape. These mergers were made separately for the presettlement and extant landscapes. Presettlement areas were inferred from all suitable soils; extant areas of suitable soils were digitized from aerial photographs. Extant areas included all intact habitat patches, including disturbed habitats (considered potential habitat for the study species), but excluded patches converted to housing, citrus, industry, or pasture. These coverages show that individual populations have experienced a wide range of habitat loss and fragmentation of nearby habitats (Figure 2).

We used GIS to calculate the areas of suitable habitat (either extant or presettlement) within 2, 8, and 32 km of the boundary of the extant or presettlement home polygon. We buffered in this way rather than buffering around the exact collection points (as in [75]) for two reasons. First, buffering the point would have included parts of the home polygon, which we intended as a separate variable. Second, we assumed that much of the polygon was occupied by the population (at one time or another) so that the landscape as seen from the habitat edges was most relevant in affecting landscape genetic structure.

We originally identified a total of 20 predictor variables. For both presettlement and extant landscapes, we calculated soil patch sizes and areas and proportions of remaining suitable soils within radii of 2 km, 8 km, and 32 km (Figure 2). We also calculated proportions of original patches remaining and current population size, area, and density and the shortest distance from the collection point to the patch edge. Finally, we calculated isolation indices for each presettlement and extant patch using the formula of Hanski and Thomas [76] for all patches within 32 km. Individual variables were transformed to improve their fit to assumptions of correlation analysis and multiple regression.

Within and among species, we noted many strong correlations among these transformed predictor variables. We therefore reduced the predictor variables to a set of seven relatively independent (absolute value of $r < 0.42$) variables: population size, population density, extant patch area, extant area within 32 km, distance to patch edge, presettlement isolation index, and extant isolation index.

For each population, we summarized genetic variation as expected heterozygosity and the number of alleles per polymorphic locus. Other measures of genetic variation (percent of polymorphic loci, observed heterozygosity) had strong correlations with these two measures for most of our study species.

To evaluate the relationship of the set of seven transformed predictor variables to expected heterozygosity and alleles per polymorphic locus, we performed forward stepwise multiple regressions for each species. Alleles per polymorphic locus did not vary among populations for Eryngium cuneifolium, so no analysis was done. All analyses were accomplished using SPSS version 11.5.0 [77].

3. Results

Genetic variation was generally poorly predicted by population and landscape variables (Table 2). For four of the six species (D. frutescens, E. cuneifolium, H. cumulicola, and W. carteri), we found no significant relationships between genetic variation and predictor variables (Table 2). For the other two species (L. ohlingerai and N. brittoniana), there were only two significant regressions, those predicting alleles per polymorphic locus. Although for one species (L. ohlingerai) the area within a buffer was a predictor of genetic variation, in no cases did isolation indices predict genetic variation (Table 2). No regressions successfully predicted expected heterozygosity from population and landscape variables for any of our six study species.

For L. ohlingerai, the number of alleles per polymorphic locus were lower in high density populations and higher in populations with larger areas of habitat remaining extant within 32 km (Table 2). The negative relationship between alleles per polymorphic locus and density (Figure 3) was opposite to that predicted. Conversely, the weak but positive relationship between area within 32 km in alleles per polymorphic locus (Figure 4) was consistent with our prediction. The two variables together predicted about one third of the variation in alleles per polymorphic locus ($r^2 = 0.34$).

For N. brittoniana, the species with the largest number of populations, alleles per polymorphic locus was higher for larger extant populations (Table 2, Figure 5), although this relationship explained less than 15% of the variation ($r^2 = 0.14$).
Figure 2: Landscape structure around three representative populations sampled for genetic variation. In each case, the location of the genetic sample is shown by the filled circle, the dark polygon represents the home polygon of the sample (with the home polygon being the soil polygon from which the genetic collection was made merged with all adjoining suitable soil polygons), and the light shading represents suitable soil polygons within the 2 km buffer (shown as a dark line). (a, b) presettlement and extant landscapes, respectively, for *Liatris ohlingeriae* at Archbold Biological Station West. The population is protected and nearby habitat loss and fragmentation is moderate. (c, d) presettlement and extant landscapes, respectively, for *Hypericum cumulicola* at Lizzie Lake. The population is unprotected and nearby habitat loss and fragmentation is high. (e, f) presettlement and extant landscapes, respectively, for *Dicerandra frutescens* at Camp Florida. The population is unprotected and nearby habitat loss and fragmentation is extreme.
4. Discussion

4.1. Genetic Patterns in Florida Landscapes. The conservation biology literature often predicts that population-level genetic variation will be positively correlated with population size [3, 14, 78] and the size and proximity of habitat patches [17, 18, 79]. These predictions were completely unsupported for our six Florida scrub species based on expected heterozygosity. Patterns based on alleles per polymorphic locus were weak.

Our only clonal study species, *N. brittoniana*, was one of the more genetically variable species and the only one that showed genetic patterns with population size. Although clonal spread does not provide for recombination and might be expected to suppress population-level genetic variation, some clonal species are genetically diverse [80, 81]. However, clonal species that cannot reproduce sexually may have very limited genetic variation [46, 82]. The genetic variation we observed in *N. brittoniana* suggests that occasional seedling recruitment must occur in this species, although seedlings are virtually unknown. If seedling recruitment and genetic diversity are mutually re-enforcing, this may explain the positive relationship of genetic variation and population size observed in this species.

For butterfly-pollinated *L. ohlingerae*, density was negatively related to genetic variation. One observation to explain this result may be that pollinators travel longer distances in less dense populations [83, 84], effectively increasing outcrossing. Density has effects on spatial genetic structure in other populations [85].

Why did we find such weak correlation between genetic variation and landscape variables? Because a great deal of habitat fragmentation in central Florida has occurred in the last few decades, the general lack of a population size effect on genetic variation in our study may be due to the lag time between population reductions due to habitat fragmentation and genetic deterioration [86, 87]. Habitat loss and degradation might be slow to impact genetic diversity (compared to impacts on inbreeding, reproductive output, and fitness), as found in an analysis of studies of neotropical tree species [87]. Fragmentation effects increase with the number of generations for plant populations [6, 10].

To better resolve the complex links between landscape variables and genetic variation [1, 88], we clearly separated population size variables and variables related to isolation in our study (as recommended by Ouborg et al. [13]). The complex habitat patch shapes for our species resulted in isolation indices for individual habitat patches were often idiosyncratic, being heavily weighted by the presence of nearby, large habitat fragments. Idiosyncratic relationships of genetic variation to land use history have been found in other studies [75] and may have affected our analyses.
Genetic patterns will also clearly be influenced by differences in the life histories, evolutionary trajectories, and ecological adaptations of each species. Fragmentation effects are greatest for common or recently rare (versus naturally rare) species and for outcrossing (versus selfing) plants [10]. Our previous analysis showed that the amount of genetic variation among these species (being highest in L. ohlingeræ and N. brittoniana) was consistent with species rankings for life span, demographic stability, an outcrossing breeding system, and longer pollinator dispersal distances [44]. Population differentiation was highest in H. cumulicola, which was attributed to its specialized pollination syndrome and the inferred modest movement patterns of its bee pollinators [44].

The lack of general patterns detected in our study may also reflect the stochastic and dynamic nature of most of these species. Current population size and distribution are likely not in equilibrium and may not reflect the past conditions that have influenced genetic patterns [43]. This is likely particularly true for short-lived species that undergo continual population fluctuations. Many of our study species show substantial fluctuations in relation to fire [48, 49], and lack of fire history data across our study populations may hamper our ability to understand geographic patterns of genetic variation. Past events such as bottlenecks, founder events, low effective population sizes, and local extinctions may also have affected patterns of genetic variation in idiosyncratic ways [89, 90].

Unmeasured effects, such as habitat and matrix quality [91], low sample size, and limited genetic variation, could have helped improve the statistical signals in our analyses. Limited genetic variation and low statistical power due to a small number of available populations for some of our species probably constrain our ability to detect patterns. Although we included an unusually large number of populations in our study, and in many cases all known populations, small numbers of populations for some species still hamper the analysis. Population numbers were particularly small for D. frutescens and E. cuneifolium, two species for which we found no link between geographic and genetic variables. In addition, low genetic variation is typical for narrow endemics [92, 93] including our study species [44] and reduces the power to detect trends. Therefore, we suggest that population size and landscape variables may be particularly unsuccessful in predicting genetic variation when species are narrowly endemic.

4.2. Past and Current Landscape Effects on Genetic Variation. We predicted that presettlement landscape structure for our suite of study species would have a stronger influence on genetic variation than extant landscape structure. However, we found little effect of landscape structure at either time on genetic variation.

Few studies have explicitly compared past and extant landscapes for their effects on landscape genetic structure, as we attempted in this study. Studies on bush crickets and butterflies have found that past landscape configuration and metapopulation structure, respectively, affected pairwise genetic similarities [94, 95]. In alpine meadow-dwelling butterflies, both contemporary and past (40 year) forest covers had effects on genetic patterns, with past cover affecting expected heterozygosity [96]. In wood frogs, populations underwent rapid genetic shifts in response to shifts in recent landscape structure [97]. However, in a ground beetle, habitat fragmentation did not explain current genetic structure [98]. In plants, the signature of historical habitat connectivity has been detected in several studies. For example, genetic diversity in Anhythmis vulneraria was more closely related to current fragment area and population size than to historical landscape configuration [33]. For Globularia bisnagarica, historical landscape structure and its influence on gene flow are thought responsible for the genetic similarity of currently fragmented populations [20]. Likewise, genetic structure of remnant populations of the endangered shrub Grevillea calceyi continued to reflect fine-scale genetic structure present before fragmentation [99]. These two studies attribute the continued influence of historical landscapes to a lag in the influence of contemporary mutation and drift in shaping population structure. This lag is brought about through long generation time, overlapping generations, and dormant seed banks.

4.3. Conclusions/Conservation Implications. Genetic variation is an increasingly important criterion in reserve design [100]. In our case, conservation strategies for most of these species are not well informed by our genetic data. Neither population size nor landscape factors are generally useful to identify the populations most critical for conservation of genetic variation. However, the best candidates for this inference may be long-lived, resprouting species such as L. ohlingeræ and N. brittoniana, for which relatively stable population sizes provide more predictive power. In this study these species showed some relationship of genetic variation to the current and past landscape.

If direct identification of critical sites for conservation of genetic variation is not possible, alternative approaches for identifying sites are necessary. One alternative approach for prioritizing Lake Wales Ridge scrub habitat for conservation is based on the cumulative gain for multiple species and on the irreplaceability of sites for the rarest species [59, 70]. Acquisition of the top eight unprotected sites recommended by these authors for conservation would result in the additional protection of five populations of N. brittoniana, four of L. ohlingeræ, three of H. cumulicola, two of W. carteri, and one of D. frutescens. Only E. cuneifolium does not occur in any of these sites [59]. Although the populations so identified [60] are worthy of protection, our allozyme data do not specifically support the selection of these particular sites if increasing genetic variation of protected populations is a conservation goal. None have unique alleles and, in all cases, alleles per polymorphic locus and expected heterozygosity are within the 90% confidence intervals for these values across all populations. Prioritizing sites to maximize the number of species captured does not seem to also maximize capture of genetic variation in this analysis.
Our results are consistent with much of the literature in suggesting that generalizations about patterns in genetic (allozyme) variability cannot always be predicted from landscape variables. In our case, we assumed that pre-settlement distributions included occupation of all sites with suitable soils. More complex, realistic landscape variables (e.g., detailed information on fire histories, not available for this landscape) might contribute to better prediction of genetic variation. Likewise, alternative types of genetic data (microsatellite or quantitative traits) might show stronger patterns in relation to the landscape. Nonetheless, as others have found, we have shown that ecological and landscape data are not necessarily consistent predictors of genetic variation. Genetic variation is also likely to be affected by climate change, for example, in the ways that climate change and habitat fragmentation may interact to affect phenology and demography [101]. As landscapes and the genetics of populations are still in considerable flux, new approaches to investigate of their linkages will be needed to improve understanding and predictability.

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