Beyond isolation by distance: What best explains functional connectivity among populations of three sympatric plant species in an ancient terrestrial island system?

Paul Gerard Nevill1 | Todd P. Robinson2 | Giovanni Di Virgilio3 | Grant Wardell-Johnson1

Abstract

Aim: Understanding how landscape features affect gene flow is critical to connectivity conservation and restoration management. Here, we examined the relationship between functional connectivity (gene flow) and structural connectivity (area and spatial configuration of habitats) in three co-occurring short-range plant taxa in an ancient terrestrial island system.

Location: Low-altitude mountain range south-western Australia.

Methods: We analysed spatial patterns of genetic differentiation at nuclear microsatellite loci using Bayesian clustering. Circuit theory modelling was used to generate all possible pathways that connect populations as resistance distance matrices based on two surfaces for each taxon. The first surface assumes a flat terrain and tests whether genetic similarity declines only with distance—isolation by distance (IBD). The second surface is habitat suitability based on species distribution modelling (SDM), which tests whether genetic similarity is a function of connected and suitable habitat. Multiple matrix regression with randomization was used to test the significance of the resistance distance matrices at predicting two metrics of genetic differentiation (FST and D_{EST}). Variance explained was partitioned using redundancy analysis.

Results: Genetic structure for the insect-pollinated taxa—Acacia adinophylla—and Tetratheca aphylla subsp. aphylla was at similar spatial scales. Unexpectedly, a higher level of genetic structure was found in the wind-pollinated Lepidosperma bungalbin. IBD best explained the gene flow of A. adinophylla (R² = 0.41–0.43) with partial support provided by habitat suitability (R² = 0.04–0.07). In contrast, connectivity by habitat suitability was highest for T. aphylla subsp. aphylla (R² = 0.56–0.59). Drivers of L. bungalbin connectivity were inconsistent between the two measures of genetic differentiation.

Main conclusions: Gene flow is facilitated by different factors for the three taxa. Habitat fragmentation would most strongly impede gene flow for T. aphylla subsp. aphylla. Geographical distance cannot be assumed as the sole or best determinant of gene flow among populations, nor can findings be generalized to coexisting taxa.
1 | INTRODUCTION

Human activities affect most of the Earth’s terrestrial systems (Hobbs, Higgs, & Harris, 2009) and have degraded billions of hectares (Gibbs & Salmon, 2015; Nkonya et al., 2016). These activities can result in habitat fragmentation and a reduction in population sizes of the plant and animal species affected. Habitat fragmentation often has adverse effects including the disruption of gene flow that may ultimately lead to the loss of allelic diversity and increased genetic divergence among populations (Elstrand, 1992; Slatkin, 1987; Young, Boyle, & Brown, 1996). Fragmentation can also have adverse impacts on mating systems by a reduction in available mates and, in plants, increased selfing for self-compatible species. This leads to reduced levels of reproductive success and fitness costs to progeny via inbreeding depression (Aguilar, Ashworth, Galetto, & Aizen, 2006; Jacquemyn, DeMeester, Longejans, & Honnay, 2012). Therefore, information on connectivity among populations of impacted species is critical to mitigate disturbances. Such information should be collected prior to activities that cause habitat fragmentation but is rarely collected at all.

Several measures have been developed to estimate habitat connectivity. They include functional connectivity or gene flow (Tischendorf & Fahring, 2000), typically indirectly estimated by using population genetic data (Keller, Brodbeck, Floss, Vonwil, & Holderegger, 2010), and structural connectivity—the area and spatial configuration of habitats. Functional connectivity among plant populations is maintained by dispersal of propagules, either seed or pollen (Sork & Smouse, 2006; Templeton, Shaw, Routman, & Davis, 1990; Young et al., 1996), and has been shown to be complex. Several studies have found that landscape structure (forests, agricultural land, riparian zones), and their fragmentation can facilitate or impede functional connectivity in plant species (e.g., Gaddis, Thompson, & Sork, 2016; DiLeo, Rico, Boehmer, & Wagner, 2017).

Landscape genetics provides the basis for examining the relationship between structural and functional connectivity. This allows insight into the evolution of plant populations such as how they diverge and ultimately speciate (Manel, Schwartz, Luikart, & Taberlet, 2003; Storfer et al., 2007). Information on the drivers of genetic structure in species is also important from a conservation perspective because it increases understanding of dispersal and the impacts of habitat fragmentation. This information can inform the development of management strategies required to maintain (or restore) connectivity and population viability (Avise, 2000). However, studies of landscape effects on gene flow in plants are commonly conducted on a single species, and findings should not be generalized to other species in the same location (Richardson, Brady, Wang, & Spear, 2016).

We examine spatial genetic patterns in three co-occurring, conservation priority plant species associated with banded iron formations (BIF) of an ancient terrestrial island system in south-western Australia (SWA). Such discrete rocky habitats are found globally (Porembski & Barthlott, 2000; Jacobi & Carmo, 2008; Gibson, Yates, & Dillon, 2010; Gibson, Meissener, Markey, & Thompson, 2012) and cover around 3% of the Earth’s surface (Guillot & Hattori, 2013). Ironstone ranges, including the Canga of Brazil and the BIFs of SWA, are ancient features and hotspots for plant species diversity and endemism (Gibson et al., 2012, 2010; Jacobi, Carmo, Vincent, & Stehman, 2007; Yates, Gibson, Petit, Dillon, & Palmer, 2011). The high species endemism characteristic of many of the world’s terrestrial island systems provides a model for studying evolutionary patterns and processes (Byrne et al., 2019). However, there have been no studies on multiple plant species using a landscape genetics framework in these systems. Acquiring information on genetic connectivity is important because in both Brazil and SWA, ironstone ranges are mined for iron ore (Ye, 2008). This creates a challenge to conserve biodiversity in these areas while exploiting their ore reserves. Thus, information on dispersal could be used to define and assess mining impact, and assist restoration.

In this study, we have two specific aims and associated hypotheses:

1. To compare spatial genetic structure in three sympatric, short-range BIF endemic plant taxa with different pollination systems, using Bayesian clustering. We hypothesized that the three study species will have different genetic patterns despite their co-occurrence and narrow geographical ranges and that each will require different management recommendations; and
2. To determine whether the variation in gene flow of the plant taxa studied can best be explained by the distance between populations (homogenous) or by connected corridors of suitable habitat (heterogeneous). We hypothesized that functional connectivity would not be spatially homogenous and could thus be significantly better explained by suitable habitat between populations than by geographical distance alone.

2 | METHODS

2.1 | Study system and field sampling

The BIF ranges of the Yilgarn Craton in SWA are isolated, ancient ranges within a predominantly flat landscape. Our study site is the Helena and Aurora Ranges (HAR), which are topographically complex, low-altitude BIFs rising about 700 m above sea level (asl) and extending over ~13 km in length (surface area 52 km²) (Di Virgilio, Wardell-Johnson, Robinson, Temple-Smith, & Hesford, 2018). The ranges have high conservation values due to the presence of many

KEYWORDS
banded ironstone formations, circuit theory, gene flow, habitat conservation, IBD, inselbergs, landscape connectivity, landscape genetics, restoration planning
Threatened and Priority flora taxa, including endemic species, and restricted plant communities not represented in secure conservation reserves (Gibson et al., 2012, 2010). Our three study taxa are *Acacia adinophylla* (Figure 1a), *Tetratheca aphylla* subsp. *aphylla* (Figure 1b) and *Lepidosperma bungalbin* (Figure 1c). These species are all restricted to the HAR and due to their low population numbers, and proximity to mining activity, they are protected under the Wildlife Conservation Act 1950 (WA). *Acacia adinophylla* and *L. bungalbin* are Priority One (P1) taxa, while *T. aphylla* subsp. *aphylla* is vulnerable and further protected under the federal Environment Protection and Biodiversity Conservation Act 1999. These taxa inhabit different niches within the HAR; *A. adinophylla* is the most widely distributed and found on the slopes and adjacent plains surrounding ironstone ridges (Figure 1d; Maslin, 1999); *L. bungalbin* and *T. aphylla* subsp. *aphylla* are confined to cliff tops and steep stony slopes (Figure 1e; Barrett, 2007; Butcher, 2007), although the former prefers south facing, shady aspects. The presence of an elaiosome on seed of each of the three taxa suggests they are primarily dispersed by ants. Pollen of *Acacia adinophylla* is likely to be dispersed by insects, *L. bungalbin* is most likely wind-pollinated (Barrett, 2013; Butcher, Bradbury, & Krauss, 2011), and *T. aphylla* subsp. *aphylla* is predominantly buzz pollinated by a subsection of the bee fauna, mainly *Lasioglossum* species (Ladd, Yates, Dillon, & Palmer, 2019).

We sampled 300 (ca. 24 at each of 13 locations, labelled 1–13), 313 (ca. 24 at each of 13 locations, labelled 1–13) and 260 (ca. 24 at each of 11 locations, labelled 1–11) plants of *A. adinophylla*, *T. aphylla* subsp. *aphylla* and *L. bungalbin*, respectively. Sampling locations are henceforth referred to as populations. Where possible, plants were sampled using a regular grid, both within each population and for the whole area, as recommended by Richardson et al. (2016). We also ensured that we collected samples from across the range of each species and included outlier populations. When possible, selected plants were at least 5 m apart to avoid sampling closely related individuals and maximum sampling distance was ~ 400 m between individual plants in the same population. Sampling involved collection of fresh green leaf, phyllode or stem material, and storage in zip-locked bags. Location of each sample was determined and recorded by GPS. Sampling locations were based on a combination of spatially distinct plant groupings (i.e., where plants grouped together, separate from other plant groups) and by division of larger continuous plant groupings into areas that could be safely accessed...
due to topography. Collections were stored on ice in the field prior to storage at 4°C in the genetics facility at Curtin University until DNA extraction.

2.2 | Microsatellite genotyping and summary statistics

Genomic DNA was extracted following collection using the Nucleospin 96 Plant II method (Macherey-Nagel GmbH and Co.). Nuclear microsatellite markers were used to genotype individuals of each species. Nuclear microsatellite loci for A. adinophylla were aacur4, aacur5, aacur11, aacur19, aacur20, aacur21, aacur25, aacur26, aacur29, aacur32, aacur52 and aacur58 (Nevill & Wardell-Johnson, 2016a), and for L. bungalbin lbcur6, lbcur14, lbcur16, lbcur21, lbcur27, lbcur33, lbcur38, lbcur40, lbcur41, lbcur50, lbcur53 and lbcur57 (Nevill & Wardell-Johnson, 2016b). Microsatellite markers previously developed for Tetratheca species were used to genotype samples of T. aphylla subsp. aphylla (Butcher & Krauss, 2009; McPherson, Porter, Rymer, Crayn, & Rossetto, 2008; Krauss, 2014). Primer pairs used were te09bgt, te15bgt, te17bgt, b1, te03bgt, a106, c3, b11 and c131.

Each marker was amplified in a 6 µl reaction volume containing PCR buffer, Bioline Immolase DNA polymerase and dNTPs based on recommendations by Bioline. 1.5 mM MgCl₂, 0.06 µM of M13-labelled forward locus-specific primer, 0.13 µM of reverse locus-specific primer, 0.13 µM of fluorescently labelled M13 primer (as described by Schuelke, 2000) and 15 ng gDNA. The following PCR conditions were used: 94°C for 5 min followed by 11 cycles at 94°C for 30 s, 60°C for 45 s (dropping 0.5°C per cycle) and 72°C for 45 s; followed by 30 cycles at 94°C for 30 s, 55°C for 45 s and 72°C for 45 s; followed by 15 cycles at 94°C for 30 s, 53°C for 45 s and 72°C for 45 s; and a final elongation step at 72°C for 10 min. For a given panel, markers were pooled for each sample. Capillary electrophoresis of the product was performed by an Applied Biosystems (AB®) 3,730 DNA Analyser. Allele sizes were determined using geneious V.7.1 (Biomatters 2005–2014). Multiple replicate runs (2% of PCR products repeated) were performed to ensure the accuracy of the final data set.

Prior to analysis, data sets were checked for clonality using genclone v.2.0 (Arnaud-Haond & Belkhir, 2006), as per Millar, Byrne, and Coates (2010). Replicate multilocus genotypes found likely to have arisen by asexual reproduction were removed from subsequent analyses. We tested for linkage disequilibrium (LD) among loci using fstat v.2.9.3.2 (Goudet, 2002). Sequential Bonferroni corrections were applied to alpha values in the determination of significance to correct for multiple comparisons of LD (Rice, 1989). Departure from Hardy–Weinberg equilibrium was assessed for each locus and population by chi-square tests in genalex v.6.5 (Peakall & Smouse, 2012). The possibility of null alleles was checked using microchecker v.2.2.3 (Van Oosterhout, Hutchinson, Wills, & Shipley, 2004).

Standard measures of genetic variation including observed and expected heterozygosity (H₀, Hₑ) and private alleles (PA) (alleles found in only one population) were calculated using GenoDive (Meirmans & Van Tienderen, 2004). Allelic richness (AR) was also calculated and rarefied to the smallest population size using HP-Rare (Kalinowski, 2005) (n = 13 for A. adinophylla, n = 18 for L. bungalbin and n = 19 for T. aphylla subsp. aphylla).

2.3 | Population differentiation

Pairwise genetic differentiation Fₜₛ (Wright, 1965), a measure of past gene flow among populations, was calculated in fstat v.2.9.3.2 (Goudet, 2002) and used as a response variable. Gene flow among populations was also assessed by Dₑₛₕ estimated in smogd v.1.2.5 (Crawford, 2010). Dₑₛₕ is an alternative measure of allelic differentiation among populations not biased by the genetic diversity of the populations (Jost, 2008).

2.4 | Population size

Changes in population size can have a strong influence on genetic patterns, particularly on rates of genetic drift. To examine whether past changes in population sizes account for any genetic differences between species, populations were assessed for past reductions in population size (over 10s to 1000s of years) using bottleneck v.1.2.02 (Cornuet & Luikart, 1997). Of the three available tests, the Wilcoxon signed rank test was applied, because (a) the sign test has low statistical power; and (b) the standardized difference test requires data from 20 or more loci. We used the two-phase mutation model (TPM), which is intermediate between the stepwise mutation model (SMM) and the infinite allele model (IAM), because few microsatellite loci follow the strict (one-step) SMM (Di Rienzo et al., 1994). We ran the TPM simulation as 90% one-step mutations and 10% multistep changes.

2.5 | Bayesian clustering

We used a Bayesian clustering model to examine the level of population structure in each species. STRUCTURE v.2.3.4 software (Pritchard, Stephens, & Donnelly, 2000) assigns individuals probabilistically to user-defined K populations to achieve Hardy–Weinberg and linkage disequilibrium within populations. STRUCTURE was run using the admixture model, assuming correlated allele frequencies, and population information was used for all individuals. STRUCTURE was run with 250,000 Markov Chain Monte Carlo iterations after a burn-in period of 100,000 iterations. Preliminary data analysis using PCA of pairwise Fₜₛ values suggested <7 genetic clusters for each species. Therefore, we modelled with K = 1 to K = 8, with 10 iterations of each K. Structure Harvester (Earl & vonHoldt, 2012) was used to infer an optimal K based on the method of Evanno, Regnaut, and Goudet (2005). The 10 runs of the optimal value of K were summarized using CLUMPP (Jakobsson & Rosenberg, 2007) with the Greedy algorithm and graphically displayed.
2.6 | Landscape genetic analysis

2.6.1 | Landscape variables

We sought to examine the influence of distance between plant populations (IBD) and habitat suitability on the functional connectivity (genetic variation) of our taxa. The IBD model was developed as a raster where every pixel had a cost of one to represent a completely flat landscape and thus ignores the effect of terrain on gene flow (Noguerales, Cordero, & Ortego, 2016). Species distribution models (SDM; habitat suitability) were generated using MaxEnt species distribution modelling software (Phillips, Anderson, & Schapire, 2006) using default parameters.

Inputs into the SDM were derived from a 2-m digital elevation model, constructed from the last returns of LiDAR data, comprising measures of morphometry (topographic position index—TPI; topographic roughness index—TRI), hydrology (topographic wetness index—TWI; SAGA wetness index—SWI) and annual solar radiation (ASR). All metrics were uncorrelated. The TPI compares the elevation of each cell in a digital elevation model to the mean elevation of a specified neighbourhood of 10 × 10m around that cell. Positive TPI values represent locations that are higher than the average of their neighbourhood window (e.g., ridges), negative values are lower (e.g., valleys), and flat areas are close to zero (Guisan, Weiss, & Weiss, 1999). TRI was used to quantify relief heterogeneity, with higher values indicating greater landscape complexity (Riley, Gloria, & Elliot, 1999). The hydrological indices used were the topographic wetness index (TWI) and the SAGA wetness index (SWI). TWI is based on the equation in Gessler, Moore, McKenzie, and Ryan (1995) and assumes steady-state and spatially invariant conditions for infiltration and transmissivity (Gruber and Peckham 2009). The SWI supplemented TWI because it has predicted potential of soil moisture for cells with small vertical distance to a channel in valley floors more realistically than TWI. Annual solar radiation estimates yearly insolation in watt-hours per square metre and was calculated in ARCGIS v 10.4 (ESRI, 2015) based on Fu and Rich (2002). All variables were uncorrelated based on a correlation threshold of 0.7.

Model accuracy was determined using a 20% subset where absence data were generated in equal proportions for each species using a random point generator and summarized by calculating the area under the curve (AUC) of respective receiver operating characteristic (ROC) curves (Fielding & Bell, 1997). We used three summary statistics to compare SDMs. Niche overlap, which considers the pairwise similarity of the suitability values, was computed using Schoener’s D (Schoener, 1968; Warren, Glor, & Turelli, 2008). It ranges from 0 (no overlap) to 1 (complete niche overlap). Range overlap was calculated as the number of grid cells in which both species are predicted to occur above a suitability of 0.5, divided by the minimum number of grid cells in which either species in predicted to be present. Area overlap was calculated as the amount of suitable habitat from one species that is present in the suitable habitat of a second species.

2.6.2 | Circuitscape modelling

The two models (IBD and SDM) for each taxa were used as input into Circuitscape 4.0 (McRae, Shah, & Mohapatra, 2013) to model connectivity of multiple pathways and thus represent the empirical costs of gene flow through a flat landscape and one that incorporates the landscape features of the study area. All inputs to Circuitscape were set to represent resistance, which required inversion of the SDM, as high habitat suitability was assumed to have low resistance (Nowakowski, DeWoody, Fagan, Willoughby, & Donnelly, 2015). Current was set to flow in any direction (eight neighbours). Models were output as ASCII grids for visualization, and pairwise resistance distance matrices for all combinations of all focal nodes (population centroids) were extracted for statistical analyses.

2.7 | Statistical analyses: Relationship between resistance distance and genetic distance

The explanatory power of the two pairwise distance matrices (IBD and SDM) was tested against both response variables ($F_{ST}$ and $D_{EST}$) using multiple matrix regressions with randomization (MMRR) with the ecodist 2.0.1 package (Goslee & Urban, 2007) in R 3.3.0 (R Core Team, 2016). The response matrix was permuted 100,000 times to determine the significance of regression coefficients (Lichstein, 2006). Noting that all variables were significant at α = 0.05, we subsequently partitioned the variance into its two components (IBD, SDM) using the varpart function in VEGAN (Oksanen et al., 2013).

3 | RESULTS

3.1 | Genetic structure

We did not find any identical multilocus genotypes, and there was no evidence of linked loci after Bonferroni corrections. All loci were polymorphic in all populations, and there was no consistent departure from Hardy–Weinberg equilibrium for any locus across all populations (see Table S2.1–3 in Appendix S2). Replicate PCRs produced the same allele scores as the originals in ~ 99% of comparisons.

From MICROCHECKER, the average frequency of null alleles in A. adinophylla was < 5% across the whole data set but two loci (accur26 and accur5) had higher frequencies (10 and 8%, respectively). To determine the overall affect that these loci had on the results, all analyses were performed with these loci both present and absent. Excluding these loci reduced the overall average null frequency to only 1.7%, and neither $F_{ST}$ nor diversity metrics changed greatly. As such, all subsequent analyses were performed on the observed allele frequencies. For L. bungalbin and T. ophyllo subsp. ophylly, the average frequency of null alleles was < 5% across the whole data set.

Genetic diversity was high for the three species (see Table S2.4–6 in Appendix S2) but varied among populations. However, there was
FIGURE 2  Gene flow potential across the Helena and Aurora Ranges using circuit theory. Relationships between genetic differentiation and its most explanatory variable are paired with maps of connectivity between populations. These were isolation by distance for *Acacia adinophylla* (a, b); and habitat suitability for *Tetratheca aphylla* subsp. *aphylla* (c, d). The most explanatory variable for *Lepidosperma bungalbin* was inconclusive. We have chosen to show isolation by distance (e, f) as it is the most visually intuitive. Sampled population centroids are shown as white-filled triangles. Pie charts show the proportion of membership values from *structure* analysis for each population. Colours in pie charts indicate the mean proportion of assignment to clusters for all individuals within a population, where optimal number of clusters was defined by the Evanno et al. (2005) method.
no spatial pattern in relation to high or low diversity populations. Allelic richness was similar among populations, and private alleles were found in very low frequencies at multiple populations. However, again there was no spatial pattern. Evidence of genetic bottlenecks was detected in two populations of *A. adinophylla* (4 and 11) (*p* < 0.05) and four populations of *T. aphylla* subsp. *aphylla* (mode shift) (4, 7, 11 and 12) although there was no geographical pattern to the distribution in either species. For *L. bungalbin*, no evidence of genetic bottlenecks was detected in any population (*p* > 0.05 and no mode shifts were detected).

Mean pairwise *F*<sub>ST</sub> estimates for *A. adinophylla* were 0.030 ± 0.005 (mean and SD), and mean estimates of *D*<sub>EST</sub> were 0.042 ± 0.006. For *L. bungalbin*, mean *F*<sub>ST</sub> was 0.068 ± 0.006 and comparisons ranged an order of magnitude from 0.005 to 0.11 (see Table S2.7–9 in Appendix S2). Mean *D*<sub>EST</sub> was 0.187 ± 0.045. For *T. aphylla* subsp. *aphylla*, mean *F*<sub>ST</sub> was 0.031 ± 0.005 and mean *D*<sub>EST</sub> was 0.078 ± 0.027.

The STRUCTURE analysis and the method of Evanno et al. (2005) identified *K* = 2 as the optimal number of clusters to the data for *A. adinophylla* and *T. aphylla* subsp. *aphylla* (Figure 2). Geographical patterns of population clustering were generally similar. Populations in the south-west and north-east formed distinct clusters, with more admixed ancestry for centrally located populations. In contrast, analyses supported *K* = 3 as the optimal number of clusters in *L. bungalbin* and individuals were strongly assigned to each of the three multi-population genetic clusters (Figure 2). Populations located in the south-west corner of the HAR formed a distinct cluster, populations in the central part of the range formed a second cluster and populations in the north-east of the species range formed a third cluster.

**TABLE 1** Niche and range overlap between pairs of SDMs. Schoener’s D ranges from 0 (completely discordant habitat models) to 1 (identical habitat models). Range overlap is the number of grid cells in which both species are predicted to occur, divided by the minimum number of grid cells in which either species are predicted to be present. Area overlap is the amount of suitable habitat from one species that is present in the suitable habitat of a second species. When compared against itself, it is the total area of suitable habitat for that species.

| Species                  | Schoener’s D | Range Overlap | Area Overlap (ha) |
|--------------------------|--------------|---------------|-------------------|
|                          | AA           | TA  | LB   | AA | TA  | LB  | AA | TA  | LB  | AA | TA  | LB  |
| *Acacia adinophylla* (AA) | 1.00         | 0.35 | 0.27 | 1.00 | 0.10 | 0.07 | 2,332 | 107 | 17  |
| *Tetratheca aphylla* subsp. *aphylla* (TA) | 0.35         | 1.00 | 0.53 | 0.10 | 1.00 | 0.92 | 107  | 1,029 | 251 |
| *Lepidosperma bungalbin* (LB) | 0.27         | 0.53 | 1.00 | 0.07 | 0.92 | 1.00 | 17   | 251  | 272 |

**FIGURE 3** Composite map showing the modelled distribution of our three study species with the red channel assigned to the SDM of *Tetratheca aphylla* subsp. *aphylla*, green to *Lepidosperma bungalbin* and blue to *Acacia adinophylla*. Yellow indicates the distribution of *L. bungalbin* (green) almost entirely overlaps with *T. aphylla* subsp. *aphylla* (red).
3.2 | Landscape genetic analysis

SDMs of the three species are shown in Supporting Information (see Figure S1.1 in Appendix S1), and a composite model is shown in Figure 3. All models were highly accurate (AUC = 0.96 – 0.99). We found that topographic control on hydrology (the SWI) was the most important variable for L. bungalbin (82%) and T. aphylla subsp. aphylla (92%) and second most important for A. adinophylla (42%). The TRI was most important for A. adinophylla (55%). Solar radiation had some importance on the distribution of L. bungalbin (11%), which showed a preference for shade. Species response curves were presented in Robinson, Virgilio, Temple-Smith, Hesford, and Wardell-Johnson (2019).

A. adinophylla occupied the largest niche space of 2,332 ha (Table 1; Figure 3) and was positioned lower on the ranges than the other two species, exhibiting minimal niche (D = 0.27-0.35) or range overlap (0.07-0.10) between them (Table 1). Suitable habitat for T. aphylla subsp. aphylla was predicted to be 1,029 ha, which was distributed from the ridges down to the upper extent of A. adinophylla (Table 1; Figure 3). Although considerably smaller in area (272 ha), and far more fragmented, the range of L. bungalbin overlapped the range of A. adinophylla on the southern side of the HAR (Figure 3), as reflected from the range overlap coefficient of 0.92 (251 ha; Table 1).

Coefficients of determination were significant for both variables for all three taxa using either single or multiple regression and required constrained ordination to examine the shared variance. Variance partitioning identified that, except for a small unique contribution ($R^2 = 0.07$), the cost surface derived from the SDM was largely nested within IBD for A. adinophylla (Figure 4a-b), indicating geographical distance ($R^2 = 0.43$) to be the best descriptor of genetic differentiation for that taxa. In contrast, IBD made no unique contribution for explaining gene flow patterns of T. aphylla subsp. aphylla.

![Figure 4](https://example.com/figure4.png)

**FIGURE 4** Partitioning of model variance explained by isolation by distance (IBD; pink ellipses), and species distribution models (SDM; hatched green ellipse) for three plant species based on two measures of genetic divergence—$F_{ST}$ and $D_{EST}$. (a) Acacia adinophylla where $F_{ST}$ is the dependent variable; (b) A. adinophylla where $D_{EST}$ is the dependent variable; (c) Tetratheca aphylla subsp. aphylla where $F_{ST}$ is the dependent variable; and (d) T. aphylla subsp. aphylla where $D_{EST}$ is the dependent variable; (e) Lepidosperma bungalbin where $F_{ST}$ is the dependent variable; (f) L. bungalbin where $D_{EST}$ is the dependent variable. The Venn diagrams were drawn in eulerAPE 3 (Micallef & Rodgers, 2014)
which was best explained by habitat suitability ($R^2 = 0.56–0.59$; Figure 4c-d), indicating that genetic differentiation is more strongly structured by landscape resistance (Figure 2c-d), than by geographical distance between populations. Although statistically significant, the best explanatory variables of genetic distance changed between SDM ($F_{ST}$) and IBD ($D_{EST}$) for *L. bungalbin*, making it difficult to determine the model with the best support. The most explanatory measure of gene flow potential across the Helena and Aurora Ranges is mapped for each taxon and paired with graphs showing their relationship with the two measures of genetic differentiation in Figure 2.

4 | DISCUSSION

The persistence of disjunct populations of rare plant species impacted by anthropogenic activity may depend on gene flow between other populations to limit the loss of genetic variation through drift (Young et al., 1996). Thus, the development of strategies to maintain gene flow should identify structural features (e.g., habitat suitability) that facilitate functional connectivity. We studied the spatial genetic patterns and effect of structural connectivity on gene flow among populations of three short-range BIF endemic plant taxa in an ancient terrestrial island system in SWA. For each, we asked whether isolation by distance (IBD) and/or habitat suitability (SDM) best explained gene flow. We found different levels of population structure in these species and that they differed in response to these two explanatory variables.

4.1 | Population structure

STRUCTURE analysis suggested that the number and location of genetic clusters were similar for *A. adinophylla* and *T. aphylla* subsp. *aphylla*. In these taxa, geographically proximate populations were more similar than geographically distant ones and there were more mixed proportions of membership for centrally located populations, suggesting some influence of IBD. In contrast, STRUCTURE clearly segregated populations of *L. bungalbin* in central, south-western and north-eastern sections of the HAR, and individuals were relatively strongly allocated to one of the three genetic clusters. In agreement, all of the STRUCTURE models showed a divide between the north-eastern and south-western populations. IBD is the most likely explanation but another possibility is isolation by adaptation (IBA) whereby gene flow among populations is reduced by local genetic adaptation to different ecological characteristics (Orsini, Vanoverbeke, Swollen, Mergeay, & Meester, 2013). Nonetheless, we found no habitat differences between sites on either side of this divide or current disjunctions in the species distribution that could also explain the location of geographical clusters.

There are no direct estimates of pollen or seed dispersal distances for *A. adinophylla*, *L. bungalbin* and *T. aphylla* subsp. *aphylla*. However, studies of congeners on BIF ranges in the region found genetic structure (examined using $F_{ST}$ and STRUCTURE) at similar spatial scales to our study in *A. woodmaniorum* (Millar, Coates, & Byrne, 2013), *A. karina* (Funnekotter, Millar, Krauss, & Nevill, 2019) and *L. sp. Mt Caudan* (Binks, Millar, & Byrne, 2015), but much higher structure in *T. paynterae* subsp. *paynterae* (Butcher et al., 2011; Butcher, McNee, & Krauss, 2009). The three study species exhibited significant divergence among most populations according to $F_{ST}$ estimates. High $H_e$ restricts maximum $F_{ST}$ values (Meirmans & Hedrick, 2011), and an alternative measure of genetic divergence, $D_{EST}$, which is unaffected by genetic diversity levels (Jost, 2008; Meirmans & Hedrick, 2011), was also estimated. $D_{EST}$ indicated higher levels of differentiation for the study species than $F_{ST}$, and a greater range of values, particularly for *L. bungalbin*, which also had the highest $H_e$. Levels of population differentiation are comparable to the few studies of short-range *Acacia* and *Lepidopserma* species on BIFs (e.g., Millar et al., 2013; Binks et al., 2015; Funnekotter et al., 2019). However, *T. aphylla* subsp. *aphylla* had a much lower level of differentiation than *T. paynterae* subsp. *paynterae* (Butcher et al., 2011, 2009). Why spatial genetic structure was lower than expected in *T. aphylla* subsp. *aphylla* is difficult to explain given the lack of direct pollinator observations in this taxon. However, seed and most pollen dispersal is likely to be far more extensive than in *T. paynterae* subsp. *paynterae*. This finding indicates differences in processes that structure genetic variation in congeners with limited ranges and similar habitats. Furthermore, it supports the conclusions of previous studies in the region which have shown that species on or associated with BIFs show a broad range of spatial genetic structure that is not predictable based on life history traits alone (reviewed in Byrne et al., 2019).

Spatial genetic structure and population divergence are difficult to predict because they can arise from a variety of factors, including adaptation and range disjunctions, but gene flow through pollen and seed dispersal are key determinants in their establishment (Wright, 1943). However, seed dispersal is unlikely to be the major factor in the maintenance of connectivity among populations of our study species, many of which are separated by at least 100s of metres from the closest neighbouring population. All three species have seeds adapted for ant-mediated seed dispersal and ants are thought to disperse seeds over relatively small distances compared to other dispersal modes (Davidson & Morton, 1981; Gomez & Espadaler, 2013), although a recent study in the region suggests that seeds are often dispersed 10s of metres by ants (Pasco et al., 2015).

It is possible that some seeds may be dispersed by other modes (water, wind or other animals), but this is unlikely. Firstly, given a semi-arid climate and complex topography, it is questionable how seed could be dispersed long distances by water across ridges on the HAR. Secondly, seed morphology suggests that wind is unlikely to play a role in seed dispersal. Thirdly, seeds are likely to be quickly collected by ants, often within 24 hr of gravity dispersal (Majer, 1980, 1984). Passerine species may be responsible for low levels of long-distance dispersal. However, other potential vectors including emus (*Dromaius novaehollandiae*, Casuariidae) and macropods (Macropodidae) generally do not occur in the cliff top environment where *L. bungalbin* and *T. aphylla* subsp. *aphylla* are found. Therefore,
pollinator movements, and in the case of *L. bungalbin*, prevailing wind direction and speed, are likely to be determinants of connectivity among populations and are avenues for future research.

### 4.2 Landscape genetic analysis

We hypothesized that functional connectivity would be better explained by suitable habitat between populations than by geographical distance alone. This was an accurate postulate for *T. aphylla* subsp. *aphylla* where connectivity derived from suitable habitat, which was almost entirely predicted from modelled water flow (SWI), explained up to 59% of the variation and geographical distance made no unique contribution. Accordingly, connectivity among populations of *T. aphylla* subsp. *aphylla* would most effectively be maintained through the retention of corridors of suitable habitat.

In contrast, gene flow among the study populations of *A. adinophylla* was best explained by geographical distance with a minor contribution from habitat suitability, which was otherwise nested within IBD. This species distribution was predicted predominantly from terrain ruggedness and water flow and occupies the greatest amount of suitable habitat of the three studied (2,332 ha vs. 272-1029 ha). The relative abundance of unfragmented suitable habitat appears to have resulted in an overestimation of connected pathways, consequently reducing its explanatory value in the linear regression models. Furthermore, whilst habitat may be suitable for *A. adinophylla*, it does not consider impedance from competing species, especially given that rugged habitats often exhibit higher species richness than physically simpler ones (Di Virgilio et al., 2018). Nonetheless, given that IBD was the most explanatory variable, modification of suitable habitat may therefore have less of an effect on gene flow for this species.

Relationships between the explanatory variables and *L. bungalbin* were poorer than for the other taxa and inconclusive. Suitable habitat was mainly derived from water flow and solar radiation (shade), and its range was nested within that of *T. aphylla* subsp. *aphylla*. Given the strong response of genetic distance to SDM resistance for *T. aphylla* subsp. *aphylla*, along with its relatively small range (272 ha), we expected a similar response for *L. bungalbin*. This proved to be the case when regressed against $F_{ST}$ ($R^2 = 0.43$) but converted to IBD when regressed against $D_{est}$ ($R^2 = 0.42$), confounding conclusions. We know little about the pollinators of most plant species on BIFs in SWA even though their foraging characteristics and preferences for habitat are likely to be important determinants of connectivity. Whilst there are no studies of the mating system of *L. bungalbin*, the species is likely to be wind-pollinated (Barrett, 2013). Therefore, pollinator behaviour is not important for connectivity. Rather, the distance and direction that wind can disperse pollen is relevant. Therefore, we suggest that further research to model wind direction during flowering would provide useful insight.

### 4.3 Management implications

We identified divergent population structures and different factors that explain functional connectivity in the three narrow range study species. There is long-standing debate on the suitability of using neutral markers for delineation of conservation units (e.g., Patkeau, 1999; Funk, McKay, Hohenlohe, & Allendorf, 2012; Hoelzel, Bruford, & Fleischer, 2019). However, it is acknowledged that, where there is significant genetic differentiation among populations or groups of populations, they should be managed as distinct entities. While the study species are all narrowly distributed and largely sympatric, our data show that they have different management requirements. For example, relatively higher levels of genetic differentiation combined with geographical isolation among different genetic clusters suggest three management units for in situ and ex situ conservation in *L. bungalbin*. In contrast, at most two units are more appropriate for *A. adinophylla* and *T. aphylla* subsp. *aphylla*, which both have lower levels of genetic differentiation than *L. bungalbin*.

Our findings emphasize that connectivity is complex and difficult to predict. Whenever possible, multispecies studies of genetic structure and gene flow should be conducted to allow a better understanding of the effects of proposed developments on the future viability of gene flow and populations, and accordingly, whether species require separate management plans (Rouget, Cowling, Lombard, Knight, & Kerley, 2006). Our findings also highlight that direct studies of both pollen and seed dispersal should be undertaken to inform conservation planning. Findings from this study could also be used to guide restoration planning and monitoring. For example, the estimates of historical gene flow could be set as baselines and targets for connectivity following restoration (Proft, Jones, Johnson, & Burridge, 2018; Ritchie, Dyer, Nevill, Sinclair, & Krauss, 2019).

Finally, this study suggests that geographical distance between populations should not necessarily be considered the sole or best determinant of gene flow levels among populations of plants. In this case, if anthropogenic disturbance increases in the study area, management actions that maintain suitable habitat between fragmented populations may improve connectivity by enhancing the potential for gene flow, particularly for *T. aphylla* subsp. *aphylla*. Further research could use the SDMs and conductance maps to design suitable habitat corridors.

The potential of any development activity to minimize impact on surrounding biodiversity will depend on the effort expended to inform the environmental impact assessment process, ideally through the collection of multiple sources of data (here, genetic, remote sensing and field-derived plant density) for the whole site, and for multiple species. Undertaking these activities in partnership with industry end users will inform stakeholder planning and help mitigate adverse effects on surrounding biodiversity.

**ACKNOWLEDGEMENTS**

This research was funded by Mineral Resources Ltd. We thank D. Temple-Smith and J. Hesford (Mineral Resources) for supporting this research. We are also grateful to M. Hay (Ecologia) and A. White (CAD Resources) for assistance with the project. PGN was supported partly by the Australian Government through the Australian...
Research Council Industrial Transformation Training Centre for Mine Site Restoration (project number ICI150100041). The views expressed herein are those of the authors and are not necessarily those of the Australian Government or Australian Research Council. Samples were taken under licences/permits # SW017143, CE004960 and 6-1516 issued by the Department of Parks and Wildlife.

DATA AVAILABILITY STATEMENT

All genotype data is available from the Dryad data repository. https://doi.org/10.5061/dryad.h0t0dk5

ORCID

Paul Gerard Nevill https://orcid.org/0000-0001-8238-0534

REFERENCES

Aguilar, R., Ashworth, L., Galetto, L., & Aizen, M. A. (2006). Plant reproductive susceptibility to habitat fragmentation: Review and synthesis through a meta-analysis. Ecology Letters, 9, 968–980. https://doi.org/10.1111/j.1461-0248.2006.00927.x

Arnaud-Haond, S., & Belkhir, K. (2006). Genclone: A computer program to analyse genotypic data, test for clonality and describe spatial clonal organization. Molecular Ecology Notes, 7, 15–17. https://doi.org/10.1111/j.1471-2126.2006.01522.x

Avise, J. C. (2000). Phylogeography: The history and formation of species. Cambridge, UK: Harvard University Press.

Barrett, R. L. (2007). New species of Lepidosperma (Cyperaceae) associated with banded ironstone in southern Western Australia. Nyctislia, 17, 37–60.

Barrett, R. L. (2013). Ecological importance of edges: A survey of the Australasian Cyperaceae genus Lepidosperma. Annals of Botany, 111, 499–529. https://doi.org/10.1093/aob/mct008

Binks, R. M., Millar, M. A., & Byrne, M. (2015). Not all rare species are the same: Contrasting patterns of genetic diversity and population structure in two narrow endemic sedges. Biological Journal of the Linnean Society, 114, 873–886.

Butcher, P. A., Bradbury, D., & Krauss, S. L. (2011). Limited pollen-mediated dispersal and partial self-incompatibility in the rare ironstone endemic Tetraphylla paynterae subsp. paynterae increase the risks associated with habitat loss. Conservation Genetics, 12, 1603–1618. https://doi.org/10.1007/s10592-011-0258-1

Butcher, P. A., & Krauss, S. L. (2009). Development of microsatellites from the rare ironstone endemic, Tetraphylla paynterae ssp. paynterae and across-species amplification. Molecular Ecology Resources, 9, 386–389.

Butcher, P. A., McNeill, S. A., & Krauss, S. L. (2009). Genetic impacts of habitat loss on the rare ironstone endemic Tetraphylla paynterae subsp. paynterae. Conservation Genetics, 10, 735–746. https://doi.org/10.1007/s10592-008-9775-y

Butcher, R. (2007). New taxa of “leafless” Tetraphylla (Elaeocarpaceae, formerly Tremandraceae) from Western Australia. Australian Systematic Botany, 139–160. https://doi.org/10.1071/SB06015

Butcher, R., Byrne, M., & Crayn, D. (2007). Evidence for convergent evolution among phylogenetically distant rare species Tetraphylla (Elaeocarpaceae, formerly Tremandraceae). Australian Systematic Botany, 20, 126–138.

Byrne, M., Krauss, S. L., Millar, M. A., Elliott, C. P., Coates, D. J., Yates, C., ... Gibson, N. (2019). Persistence and stochasticity are key determinants of genetic diversity in plants associated with banded iron formation inselbergs. Biological Reviews, 94, 753–772. https://doi.org/10.1111/brv.12477

Cornuet, J. M., & Luikart, G. (1997). Description and power analysis of two tests for detecting recent population bottlenecks from allele frequency data. Genetics, 144, 2001–2014.

Crawford, N. G. (2010). Smogd: Software for the measurement of genetic diversity. Molecular Ecology Resources, 10, 556–557. https://doi.org/10.1111/j.1755-0998.2009.02801.x

Davidson, D. W., & Morton, S. R. (1981). Myrmecochory in some plants (F. Chenopodiaceae) of the Australian arid zone. Oecologia, 50, 357–366. https://doi.org/10.1007/BF00344976

Di Rienzo, A., Peterson, A. C., Garza, J. C., Valdes, A. M., Slatkin, M., & Freimer, N. B. (1994). Mutational processes of simple-sequence repeat loci in human populations. Proceedings of the National Academy of Sciences of the USA, 91, 3166–3170. https://doi.org/10.1073/pnas.91.8.3166

Di Robinson, T. P., Virgilio, G., Temple-Smith, D., Harf, J., & Wardell-Johnson, G. W. (2019). Characterisation of range restriction amongst the rare flora of banded ironstone formation ranges in semi-arid south-western Australia. Australian Journal of Botany, 67, 234–247. https://doi.org/10.1071/BT18111

Di Virgilio, G., Wardell-Johnson, G. W., Robinson, T. P., Temple-Smith, D., & Harf, J. (2018). Characterising fine-scale variation in plant species richness and endemism across topographically complex, semiarid landscapes. Journal of Arid Environments, 156, 59–68. https://doi.org/10.1016/j.jaridenv.2018.04.005

DiLeo, M. F., Rico, Y., Boehmer, H. J., & Wagner, H. H. (2017). An ecological connectivity network maintains genetic diversity of a flagship wildflower, Pulsatilla vulgaris. Biological Conservation, 212, 12–21. https://doi.org/10.1016/j.biocon.2017.05.026

Earl, D. A., & vonHoldt, B. M. (2012). STRUCTURE HARVESTER: A website and program for visualizing STRUCTURE output and implementing the Evanno method. Conservation Genetics Resources, 4, 359–361. https://doi.org/10.1007/s12686-011-9548-7

Ellstrand, N. C. (1992). Gene flow by pollen: Implications for plant conservation genetics. Oikos, 63, 77–86. https://doi.org/10.2307/3545517

ESRI (2015). ArcGIS desktop: Release 10.4. Redlands, CA: Environmental Systems Research Institute.

Evanno, G., Regnaut, S., & Goudet, J. (2005). Detecting the number of clusters of individuals using the software STRUCTURE: A simulation study. Molecular Ecology, 14, 2611–2620. https://doi.org/10.1111/j.1365-294X.2005.02553.x

Fielding, A. H., & Bell, J. F. (1997). A review of methods for the assessment of prediction errors in conservation presence/absence models. Environmental Conservation, 24, 38–49.

Fu, P., & Rich, P. M. (2002). A geometric solar radiation model with applications in agriculture and forestry. Computers and Electronics in Agriculture, 37, 25–35. https://doi.org/10.1016/S0168-1699(02)00115-1

Funk, W. C., McKay, J. K., Hohenlohe, P. A., & Allendorf, F. W. (2012). Harnessing genomics for delineating conservation units. Trends in Ecology & Evolution, 27, 489–496. https://doi.org/10.1016/j.tree.2012.05.012

Funnellker, A., Millar, M., Krauss, S., & Nevill, P. G. (2019). Phylogeographic analyses of Acacia karina (Fabaceae) support long-term persistence of populations both on and off banded iron formations. Australian Journal of Botany, 67, 194. https://doi.org/10.1071/BT18045

Gaddis, K. D., Thompson, P. G., & Sork, V. L. (2016). Dry-washes determine gene flow and genetic diversity in a common desert shrub. Landscape Ecology, 31, 2215–2229. https://doi.org/10.1007/s10109-016-0393-7

Gessler, P. E., Moore, I. D., McKenzie, N. J., & Ryan, P. J. (1995). Soil-landscape modelling and spatial prediction of soil attributes. International Journal of Geographical Information Systems, 9, 421–432.
Majer, J. D. (1980). The influence of ants on broadcast and naturally spread seed in rehabilitated bauxite mined areas. Reclamation and Revegetation Review, 3, 3–9.

Majer, J. D. (1984). The influence of ants on seeding operations in northern Australian mined areas. Reclamation and Revegetation Review, 2, 299–313.

Mannel, S., Schwartz, M. K., Luijhart, G., & Taberlet, P. (2003). Landscape genetics: Combining landscape ecology and population genetics. Trends in Ecology and Evolution, 18, 189–197. https://doi.org/10.1016/S0169-5347(03)00089-9

Maslin, B. (1999). Accacia adinophylla Maslin. Nuytsia, 12, 318–320.

McPherson, H., Porter, C., Rymer, P. D., Crayn, D. M., & Rossetto, M. (2008). Isolation and characterization of polymorphic microsatellite loci from Tetrasheca ericifolia (Elaeocarpaceae). Molecular Ecology Resources, 8, 867–869.

McRae, B. H., Shah, V. B. H., & Mohapatra, T. K. (2013). Circuitscape 4 user guide. Arlington, VA: The Nature Conservancy. https://www.circuitscape.org

Meirmans, P. G., & Hedrick, P. W. (2011). Assessing population structure: FST and related measures. Molecular Ecology Resources, 11, 5–18. https://doi.org/10.1111/j.1755-0998.2010.02927.x

Meirmans, P. G., & Van Tienderen, P. H. (2004). genotype and genodive: Two programs for the analysis of genetic diversity of sexual organisms. Molecular Ecology Notes, 4, 792–794. https://doi.org/10.1111/j.1471-8286.2004.00770.x

Miclea, L., & Rodgers, P. (2014). eulerAPE: Drawing area-proportional 3-Venn diagrams using ellipses. PloS one, 9(7), e101717.

Millar, M. A., Byrne, M., & Coates, D. J. (2010). The maintenance of disparate levels of clonality, genetic diversity and genetic differentiation in disjunct subspecies of the rare Banksia ilicifolia. Molecular Ecology, 19, 4217–4227. https://doi.org/10.1111/j.1365-294X.2010.04817.x

Millar, M. A., Coates, D. J., & Byrne, M. (2013). Genetic connectivity and diversity in inselberg populations of Acacia woodsiiflora, a rare endemic plant of the Yilgarn Craton Banded Iron Formations. Hereditas, 111, 437–444.

Nevill, P. G., & Wardell-Johnson, G. (2016a). Microsatellite primers for the rare shrub Acacia adinophylla (Fabaceae). Applications in Plant Sciences, 4, 1600084. https://doi.org/10.3732/apps.1600084

Nevill, P. G., & Wardell-Johnson, G. (2016b). Microsatellite primers for the rare sedge Lepidopteris bunguloides (Cyperaceae). Applications in Plant Sciences, 4, 1600083. https://doi.org/10.3732/apps.1600083

Nkonya, E., Anderson, W., Kato, E., Koo, J., Mirzaeaa, A., von Braun, J., & Meyer, S. (2016). Global cost of land degradation. Economics of land degradation and improvement—A global assessment for sustainable development (pp. 117–165). Cham, Switzerland: Springer.

Nogueiras, V., Cordero, P. J., & Ortego, J. (2016). Hierarchical genetic structure shaped by topography in a narrow-endemic montane grass-hopper. BMC Evolutionary Biology, 16, 96. https://doi.org/10.1186/s12862-016-0663-7

Nowakowski, A. J., DeWeedy, J. A., Fagan, M. E., Willoughby, J. R., & Donnelly, M. A. (2015). Mechanistic insights into landscape genetic structure of two tropical amphibians using field-derived resistance surfaces. Molecular Ecology, 24, 580–595. https://doi.org/10.1111/mec.13052

Oksanen, J., Blanchet, F. G., Kindt, R., Legendre, P., Minchin, P. R., O’hara, R. B., ... Oksanen, M. J. (2013). Package ‘vegan’. Community ecology package, version, 2.9.

Orsini, L., Vanoverbeke, J., Swilling, L., Mergeay, J., & De Meester, L. (2013). Drivers of population genetic differentiation in the wild: Isolation by dispersal limitation, isolation by adaptation and isolation by colonization. Molecular Ecology, 22, 5983–5999. https://doi.org/10.1111/mec.12561

Patkeau, D. (1999). Using genetics to identify intraspecific conservation units: A critique of current methods. Conservation Biology, 13, 1507–1509. https://doi.org/10.1046/j.1523-1739.1999.89507.x
NEVILL ET AL.

Pasco, C. M., Nevill, P. G., Elliott, C. P., Majer, J. D., Anthony, J. M., & Krauss, S. L. (2015). The critical role of ants in the extensive dispersal of Acacia seeds revealed by genetic parentage assignment. Oecologia, 179, 1123–1134. https://doi.org/10.1007/s00442-015-3400-9

Peakall, R., & Smouse, P. E. (2012). GenAlEx 6.5: Genetic analysis in Excel. Population genetic software for teaching and research—an update. Bioinformatics, 28, 2537-2539. https://doi.org/10.1093/bioinformatics/bts460

Phillips, S., Anderson, R., & Schapire, R. (2006). Maximum entropy modeling of species geographic distributions. Ecological Modelling, 190, 231–259.

Porembski, S., & Barthlott, W. (2000). Inselbergs. Biotic diversity of isolated rock outcrops in tropical and temperate regions. Berlin, Heidelberg, New York: Springer Verlag.

Pritchard, J. K., Stephens, M., & Donnelly, P. (2000). Inference of population structure using multilocus genotype data. Genetics, 155, 945–955.

Proft, K. M., Jones, M. E., Johnson, C. N., & Burridge, C. P. (2018). Making the connection: Expanding the role of restoration genetics in restoring and evaluating connectivity. Restoration Ecology, 26, 411–418. https://doi.org/10.1111/rec.12692

R Core Team (2016). R: a language and environment for statistical computing. Vienna, Austria: R Foundation for Statistical Computing.

Rice, W. R. (1989). Analysing tables of statistical tests. Evolution, 43, 223.

Richardson, J. L., Brady, S. P., Wang, I. J., & Spear, S. F. (2016). Navigating the pitfalls and promise of landscape genetics. Molecular Ecology, 25, 849–863. https://doi.org/10.1111/mec.13527

Riley, S. J., De Gloria, S. D., & Elliot, R. (1999). A Terrain ruggedness that quantifies topographic heterogeneity. Intermountain Journal of Science, 5(1–4), 23–27.

Ritchie, A. L., Dyer, R. J., Nevill, P. G., Sinclair, E. A., & Krauss, S. L. (2019). Wide outcrossing provides functional connectivity for new and old Banksia populations within a fragmented landscape. Oecologia, 190, 255–268. https://doi.org/10.1007/s00442-019-04387-z

Rouget, M., Cowling, R. M., Lombard, A. T., Knight, A. T., & Kerley, G. I. H. (2006). Designing large-scale conservation corridors for pattern and process. Conservation Biology, 20, 549–561. https://doi.org/10.1111/j.1523-1739.2006.00297.x

Schoener, T. W. (1968). Anolis lizards of Bimini: Resource partitioning in a complex fauna. Ecology, 49, 704–726.

Schuelke, M. (2000). An economic method for the fluorescent labelling of PCR fragments. Nature Biotechnology, 18, 233–234.

Slatkin, M. (1987). Gene flow and the geographic structure of natural populations. Science, 236, 787-792. https://doi.org/10.1126/science.3576198

Sork, V. L., & Smouse, P. E. (2006). Genetic analysis of landscape connectivity in tree populations. Landscape Ecology, 21, 821–836. https://doi.org/10.1007/s10180-005-5415-9

Storfer, A., Murphy, M. A., Evans, J. S., Goldberg, C. S., Robinson, S., Spear, S. F., ... Waits, L. P. (2007). Putting the “landscape” in landscape genetics. Heredity, 98, 128–142. https://doi.org/10.1038/sj.hdy.6800917

Templeton, A. R., Shaw, K., Routman, E., & Davis, S. K. (1990). The genetic consequences of habitat fragmentation. Annals of the Missouri Botanical Garden, 77, 13–27. https://doi.org/10.1002/23399621

Tischendorf, L., & Fahring, L. (2000). On the usage and measurement of landscape connectivity. Oikos, 90, 7–19. https://doi.org/10.1034/j.1600-0706.2000.900102.x

Van Oosterhout, C., Hutchinson, W. F., Wills, D. P. M., & Shipley, P. (2004). Micro-checker: Software for identifying and correcting genotyping errors in microsatellite data. Molecular Ecology Notes, 4, 535–538. https://doi.org/10.1111/j.1471-8286.2004.00684.x

Warren, D. L., Glor, R. E., & Turelli, M. (2008). Environmental niche equivalency versus conservatism: Quantitative approaches to niche evolution. Evolution, 62, 2868–2883. https://doi.org/10.1111/j.1558-5646.2008.00482.x

Wright, S. (1943). Isolation by distance. Genetics, 28, 114–138.

Wright, S. (1951). The interpretation of population structure by F-statistics with special regard to systems of mating. Evolution, 19, 395–420. https://doi.org/10.1111/j.1558-5646.1965.tb01731.x

Yates, C., Gibson, N., Petit, N. E., Dillon, R., & Palmer, R. (2011). The ecological relationships and demography of restricted ironstone endemic plant species: Implications for conservation. Australian Journal of Botany, 59, 692–700. https://doi.org/10.1071/BT11199

Ye, Q. (2008). Commodity booms and their impacts on the Western Australian economy: The iron ore case. Resources Policy, 33, 83–101. https://doi.org/10.1016/j.resourpol.2007.10.003

Young, A., Boyle, T., & Brown, T. (1996). The population genetic consequences of habitat fragmentation for plants. Trends in Ecology & Evolution, 11, 413–418. https://doi.org/10.1016/0169-5347(96)10045-8

Biosketch

P.G.N.’s research focuses on the use of molecular tools for the conservation and restoration of plant species. The research group is interested in combining genetic and geospatial data analytics for improved conservation outcomes.

Author contributions: P.G.N., T.P.R. and G.W.J. conceived and designed the study; P.G.N. conducted fieldwork and collected the data; P.G.N., T.P.R. and G.V.D. analysed the data; P.G.N wrote the manuscript with contributions and editorial comment from T.P.R., G.V.D. and G.W.J. All authors have reviewed and approved the final version of the manuscript.

Supporting Information

Additional supporting information may be found online in the Supporting Information section at the end of the article.

How to cite this article: Nevill PG, Robinson TP, Di Virgilio G, Wardell-Johnson G. Beyond isolation by distance: What best explains functional connectivity among populations of three sympatric plant species in an ancient terrestrial island system? Divers Distrib. 2019;25:1551–1563. https://doi.org/10.1111/ddi.12959