

Unraveling the Genomic Landscape of *Campylorhynchus* Wrens along Western Ecuador's Precipitation Gradient: Insights into Isolation by Distance, Isolation by Environment, and Hybridization

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Abstract

Climate variability influences genetic and phenotypic diversity within species, impacting biodiversity's evolution. Gene flow and selection maintain changes in genetic and phenotypic variants along an environmental gradient. We investigated a hybrid zone in western Ecuador, involving two wren species (Aves: Troglodytidae), *Campylorhynchus zonatus* and *C. fasciatus*, and their admixed populations. We addressed two primary questions: (1) What is the relative contribution of Isolation by Distance (IBD) and Isolation by Environment (IBE) to genetic differentiation in these species along the western Ecuadorian environmental gradient? (2) Is there evidence of genetic admixture and introgression between these taxa in western Ecuador? We analyzed 4,409 SNPs from 112 blood samples sequenced using ddRadSeq. Clusters ranged from K=2-4, aligning with geographic origins, known phylogenetics, and physical or ecological constraints. IBD was evident across all models, while IBE was less pronounced but still significant for annual mean precipitation and precipitation seasonality. Genetic admixture between *C. f. pallescens* and *C. zonatus* gradually changed along the environmental gradient. Genetic differentiation in the two *C. f. pallescens* populations could be attributed to an unreported potential physical barrier in central western Ecuador. The proximity of the Andes to the coastline restricted lowland habitats, limiting dispersal and gene flow, especially among dry-habitat specialists. Taxonomic changes are not proposed, but the admixture in *C. f. pallescens* suggests it may be a hybrid between *C. z. brevirostris* and *C. fasciatus*, with varying degrees of admixture in western Ecuador and northwestern Peru. This study enhances our understanding of avian population genomics in tropical regions.

Genomic Landscape of *Campylorhynchus* .

Running Title: Unraveling the Genomic Landscape of *Campylorhynchus* Wrens along Western Ecuador's Precipitation Gradient: Insights into Isolation by Distance, Isolation by Environment, and Hybridization.

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Genetic admixture between *C. f. pallescens* and *C. zonatus* gradually changed along the environmental gradient. Genetic differentiation in the two *C. f. pallescens* populations could be attributed to an unreported potential physical barrier in central western Ecuador. The proximity of the Andes to the coastline restricted lowland habitats, limiting dispersal and gene flow, especially among dry-habitat specialists. Taxonomic changes are not proposed, but the admixture in *C. f. pallescens* suggests it may be a hybrid between *C. z. brevirostris* and *C. fasciatus*, with varying degrees of admixture in western Ecuador and northwestern Peru. This study enhances our understanding of avian population genomics in tropical regions.

Keywords: Isolation by distance, isolation by environment, hybridization, *Campylorhynchus*.

INTRODUCTION

Geographical barriers are well-established drivers of speciation in many organisms, including birds and mammals (Tobias et al., 2020; Wright, 1943; Yao et al., 2022). However, other factors, such as environmental heterogeneity, geographical distance, and hybridization can affect the genetic differentiation of populations (DuBay & Witt, 2014; Sexton et al., 2014). The isolation by distance (IBD) and isolation by environment (IBE) models are two common ways to explain how geographic distance or environmental variability affect the genetic structure of populations. IBD occurs when genetic exchange is inversely proportional to geographic distance (Wright, 1943). Conversely, IBE is associated with greater genetic similarity between populations similar in environmental conditions (Alberto et al., 2013; Sexton et al., 2014). High gene flow between populations from different environments may hinder local adaptation (Tigano & Friesen, 2016). This occurs because gene flow can introduce maladaptive alleles from other populations, disrupting locally adapted genotypes (Tigano & Friesen, 2016). However, moderate gene flow may augment local adaptation by providing genetic variation upon which selection can act (Hoeksema & Forde, 2008). Predicting gene flow patterns across spatial and environmental gradients may facilitate forecasts of species' resilience to intensifying anthropogenic environmental change (Hoeksema & Forde, 2008). Unfortunately, the effects of spatial and ecological processes on genetic differentiation are poorly understood in tropical species.

The tropics are known to harbor the highest levels of biodiversity, but the true extent of diversity remains underestimated due to incomplete sampling and the lack of detailed studies on many species (Bálint et al., 2011; Lohman et al., 2010). South American bird populations exhibit significant genetic divergence across geographic space, particularly in species complexes, resulting in high intraspecific differences that overlap with interspecific differences (Camargo et al., 2015; Céspedes-Arias et al., 2021; Del-Rio et al., 2022; Milá et al., 2009; Tavares et al., 2011). The Tumbes-Choco-Magdalena biodiversity hotspot in South America, which spans a moisture gradient from the humid Choco-Darien-Western Ecuador region to the dry Tumbesian

region, has a high-endemism and distinct biogeographical patterns (Dodson & Gentry, 1991; Mittermeier et al., 1998). Research suggests that climatic factors might play a role in driving biogeographical patterns, as closely related species' distributional boundaries match climate regimen boundaries at the transition zone between the two regions in Western Ecuador (Albuja et al., 2012; Amador et al., 2019; Escibano-Avila et al., 2017; Morrone, 2006; Prieto-Torres et al., 2019). This suggests that climate may be a major factor influencing the distribution of these species. Western Ecuador provides a unique system for studying genetic structure patterns in which elevation does not vary. However, the factors underlying genetic differentiation in the region remain unknown.

To address this gap, we investigated patterns of genetic structure and the potential contributions of IBD, IBE, and geographical barriers in shaping inter and intraspecific genetic differentiation of two closely related bird species, the Band-backed Wren (*Campylorhynchus zonatus brevirostris*) and Fasciated Wren (*Campylorhynchus fasciatus*). These species exhibit parapatric distributions across wet and dry regions respectively of Western Ecuador and Peru, with the subspecies *C. f. fasciatus* found in the Marañón valley of Northeastern Peru (Figure 1). We hypothesized that different ecological preferences and geographical distances result in limited dispersal between species, and genetic clusters with species along the precipitation gradient in western Ecuador. In the context of testing IBD and IBE shaping distributions of *C. zonatus*, *C. fasciatus*, we asked: What is the relative contribution of IBD and IBE on patterns of genetic differentiation of these species along the environmental gradient in Western Ecuador? If ecological conditions and geographical distance influence genetic differentiation and admixture patterns, then we would expect *C. zonatus* and *C. fasciatus* to show significant associations between genetics, geographical distance (IBD), or climate (IBE), leading to population structure that closely mirrors the known geographic distribution of species and populations (Brawn et al., 1996; Culumber et al., 2012).

Environmental heterogeneity not only promotes genetic differentiation of populations but also facilitates hybridization by providing a pathway for secondary interaction between related species that occur along environmental gradients (Carling & Thomassen, 2012; Randler, 2006; Runemark et al., 2018). Consequently, hybrid zones, regions where genetically distinct populations meet and produce offspring (Harrison & Larson, 2014), commonly occur along environmental gradients (Fritzsche & Kaltz, 2000; Kameyama et al., 2008; Yanchukov et al., 2006). Over 200 avian hybrid zones have been formally described, and environmental gradients have been identified as a key factor in maintaining these hybrid zones (Miller et al., 2014). However, the extent of hybridization can vary depending on the strength of reproductive barriers, ecological factors, and genetic compatibility between the species (Ottenburghs, 2018; Winker, 2021). Hybrid zones are fundamental for studying evolutionary processes between divergent populations (Minder et al., 2007; Sloop et al., 2011; Whitham et al., 1999).

In addition to the well-established ecological preferences of *C. zonatus brevirostris* and *C. fasciatus*, field observations of plumage patterns, frequency, and group size suggest that hybridization may occur in the transition zone of the precipitation gradient in western Ecuador (LDM, pers. obs). In the transition zone, some individuals identified as *C. zonatus brevirostris* may lack the ochraceous belly that distinguishes this species (Ridgely and Greenfield 2001, Henry 2005), suggesting that they may be potential hybrids of *C. zonatus* and *C. fasciatus*. Additionally, the frequency and group size of *C. zonatus* increases significantly in this zone (LDM, pers. obs). This suggests that admixed individuals may be present in populations along transitional habitats where *C. zonatus* is locally common. These observations prompted us to ask: Is there evidence of genetic admixture and introgression between these taxa in Western Ecuador? Our study sheds light on the potential role of local adaptation in structuring regional populations. It discusses the implications of physical barriers for the conservation of dry-habitat specialists in Southwest Ecuador and possible hypotheses to explain introgression patterns among species.

MATERIALS AND METHODS

Study Region

The lowland in Western Ecuador and Peru is characterized by a moisture gradient from the humid Choco-

Darien-Western Ecuador region in the north to the dry Tumbesian region in the south. Rainfall in Western Ecuador ranges from 2000 to 7000 mm annually, whereas the southern region has an eight-month dry season with less than 1000 mm annually (Dodson & Gentry, 1991) (Figure 1).

Species System

The Band-backed Wren (*Campylorhynchus zonatus*) is found in eastern Mexico to northwest Ecuador, with seven subspecies and four disjunct populations occupying edges, open disturbed areas, and wet forests (Kroodsma & Brewer, 2020a). One of its subspecies, *C. z. brevirostris*, is present in two disjunct populations in northwestern Ecuador and northern Colombia. The Fasciated Wren (*Campylorhynchus fasciatus*) is a commonly found species in western Ecuador and northwest Peru (Figure 1), where it inhabits arid and semiarid areas and deciduous forests (Kroodsma & Brewer, 2020b; Ridgely & Greenfield, 2001). The subspecies *C. fasciatus pallescens* occurs in western Ecuador and northwestern Peru, while the nominal subspecies *C. f. fasciatus* is present in western Peru and the dry Marañón valley on the eastern side of the Andes (Kroodsma & Brewer, 2020b; Ridgely & Greenfield, 2001). *C. z. brevirostris* and *C. fasciatus* have parapatric distributions along the precipitation gradient in western Ecuador, with *C. z. brevirostris* restricted to the wet region and *C. fasciatus* to the dry region (Ridgely & Greenfield, 2001). Phylogenetic relationships of *Campylorhynchus* show that *C. z. brevirostris* and *C. albobrunneus* are sister species and share a common ancestor with *C. fasciatus* (Barker 2007, Burleigh et al. 2015, Vázquez-Miranda and Barker 2021). Recent studies estimated the time of divergence between *C. z. brevirostris-albobrunneus* and *C. fasciatus-pallescens* approximately 1.9 million years ago (1.575-2.415 Ma) (Vázquez-Miranda & Barker, 2021). In contrast, the average time to speciation for several phylogroups at the equator is around two million years (Weir & Schluter, 2007).

Samples Collection

We collected blood samples from the brachial vein of 48 putative *C. z. brevirostris* and 49 *C. f. pallescens* individuals at 12 sampling locations along Western Ecuador, mostly between July and December 2018, with two exceptions in August 2017. We stored blood samples in Eppendorf tubes with lysis buffer (Tris HCl, pH = 8.0 = 0.1M; NaCl = 0.01M; EDTA = 0.1M; SDS = 3%). Breeding groups were sampled at least 400 m apart to avoid related individuals. We also obtained tissue samples preserved in 95% ethanol from the Florida Museum of Natural History (FLMNH), including ten *C. f. pallescens* samples from Sullana in Northwestern Peru (collected in October 2011) and 17 *C. f. fasciatus* samples from three locations in the Marañón valley of northeastern Peru: Jaen (October 2010), Chachapoyas (November 2009), and Celedin (November 2009). DNA extraction was performed using Qiagen DNeasy blood & tissue kits and Qiagen extraction protocols, with DNA concentration determined using the Qubit dsDNA HS Assay Kit and Qubit® Fluorometer.

Library Construction and Sequencing

We followed the ddRAD-seq protocol by Peterson et al. (2012) as modified by Thrasher et al. (2018). Briefly, we digested 20ul of DNA between concentrations of 3.5ng/ul-40ng/ul with SbfI and MspI and ligated with one of the 20 P1 adapters (each containing a unique inline barcode) and a P2 adapter (P2-MspI). Samples with similar concentrations were pooled in groups of 20 (each with a unique P1 adapter) and purified using 1.5x volumes of homemade MagNA made with Sera-Mag Magnetic Speed-beads (FisherSci) as described by Rohland & Reich (2012). Fragments between 400 bp and 700 bp were selected using BluePippin (Sage Science) by the Cornell University Biotechnology Resource Center (BRC). Index groups and Illumina sequencing adapters were added by performing 11 PCR cycles with Phusion DNA Polymerase (NEB). We multiplexed samples in several index groups (19 and 20 individuals each). Multiplexing was performed in several index groups (19 and 20 individuals each), with sequencing on an Illumina NextSeq 500 (150bp single-end) incorporating a ~10% PhiX spike-in for library diversity.

Quality Filtering and Demultiplexing

After the quality of the reads was assessed using FASTQC version 0.11.5 (Andrews et al., 2016), we trimmed all sequences to 147bp using fastX-trimmer (FASTX-Toolkit) to exclude low-quality calls near the 3' of the reads. We removed reads containing at least a single base with a Phred quality score of less than 10

(using `fastq_quality_filter`). We removed sequences if more than 5% of the bases had a Phred quality score of less than 20. Using the `process_radtags` module from the STACKS version 2.3 (Catchen et al., 2013), we demultiplexed the reads to obtain files with specific sequences for each individual. After demultiplexing, we retained samples with more than 10^5 reads for the de novo assembly, removing 12 samples with low read numbers. We ended with a final data set of 112 samples for analysis.

De novo assembly

Because we do not have a sequenced genome for any species or a close relative, we assembled the sequences de novo using the STACKS pipeline (Catchen et al., 2013). First, we selected 12 samples with the highest number of reads and ran `denovo_map.pl` testing values from one to nine for `-M` (number of mismatches allowed between stacks within individuals) and `n` (number of mismatches permitted between stacks between individuals) following the `n=M` rule (Paris et al., 2017) while keeping `m=3` (stack depth/number of identical reads required to initiate a new allele). We kept `r=0.8` (the minimum percentage of individuals in a population required to process a locus for that population). It has been shown that at least 80% of the population should present a specific locus to be included, known as the 80% rule or `r80` loci (Paris et al., 2017). We set all samples to the same population (`p=1`) for the parameter testing assembly. We plotted the number of SNPs called against the `M` parameters to find the optimum `M`, after which no additional SNP calling was observed. We found an optimum value for `n=m=5` for the final de novo assemblies. After the parameters testing assembly, we performed de novo assembly with all the samples and the parameters described above and set `p=1`. When a RAD locus had more than one SNP, the data were restricted to the first (`-write_single_snp`) to avoid including SNPs in high linkage disequilibrium (LD). We required a minor allele frequency of 0.05 to process a nucleotide site (`-min_maf`).

We used the `dartR` package (Gruber et al., 2018) to measure pairwise population-based Linkage Disequilibrium (LD) across all loci. We used 0.5 as the threshold for testing SNPs in LD (C. S. Carlson et al., 2004). We retained the entire data set for further analyses, given that only 0.1% of loci showed R^2 values over 0.5 across all pairwise combinations.

Population Structure and Admixture Patterns

We explored the genetic structure of the data set with a Spatial Principal Component Analysis (sPCA) analyzed using the function `sPCA` from the R package `Adegenet` 2.13 (Jombart, 2008). We set the function to build a distance-based connection network with neighbors within a Euclidean distance between one and 26.4km based on the maximum dispersal distance recorded for Cactus Wren (Lynn et al., 2022). The components of the sPCA are separated into global (positive eigenvalues) and local (negative eigenvalues) structures. Large global scores reflect the presence of clines or other structures, while local scores represent the presence of strong genetic distances between neighbors. We assessed the significance of both patterns with a Monte Carlo procedure included in the functions `global.rtest` and `local.rtest` using 99,999 permutations.

We used the Bayesian clustering software `STRUCTURE` version 2.3.4; (Pritchard et al., 2000) to estimate the membership coefficients for each individual (Q -value). We ran a spatial (`LOCPRIOR=1`) model using sampling locations as prior population information. We used the admixture model with 20 independent replicates with a burn-in of 10^5 and a run length of 10^6 Monte Carlo iterations. Following the recommendations by Gilbert et al. (2012) — for a number of genetic clusters (K) from one to 12, allowing for admixture (`NOADMIX=0`). We used the method described by (Evanno et al., 2005) implemented in `STRUCTURE HARVESTER` version 0.6.94; (Earl & Vonholdt, 2012) to find the value of K that captures most of the structure in the data, and that seems biologically sensible (Pritchard et al., 2003). We used the software `CLUMPP` (Jakobsson & Rosenberg, 2007) with a `LargeKGreedy` model and 50000 random repeats to combine replicates accounting for potential “label switching” and “genuine multimodality” differences. We further calculated the posterior probability of assignment of individuals using Discriminant Analysis of Principal Components (DAPC) from the R package `Adegenet` 2.13 (Jombart, 2008). First, we used the function `find.clusters` to determine the most likely number of genetic clusters and the group membership for each individual using 100 principal components (PCs) and 10^6 iterations for $K=1-20$. We selected the number of clusters with the

lowest Bayesian Information Criterion (BIC) value as optimal. We used the estimated group membership to perform a preliminary DAPC retaining 100 PCs and two Discriminant Analysis axes (DAs). We used the preliminary DAPC to calculate the optimal number of PCs to keep using the `optim.a.score` function set with ten simulations. We performed the final DAPCs for $K=2-4$ using the optimal number of PCs previously estimated.

Parental and Hybrid Classification

We used the Q-values from STRUCTURE to group individuals as hybrids if $0.1 \leq Q\text{-value} \leq 0.9$ for $K=2$, and as parental otherwise. We further estimated the Maximum Likelihood of individual Hybrid Indexes (HI, proportion of alleles inherited from one of the parental species). We classified parental individuals if they belong to the northern-most populations surveyed of *C. z. brevirostris* from Ecuador (Las Golondrinas, Pedro Vicente Maldonado, and Pedro Carbo) or the southern-most populations surveyed of “nominally” ssp. *C. f. fasciatus* (Jaen, Chachapoyas, and Celedin) and have Q-values ≥ 0.90 for either parental population, based on STRUCTURE Q-values for $K=2$. These individuals were “parentals” to train the function `est.h` of the R package INTROGRESS 1.2.3 (Gompert & Alex Buerkle, 2010). HI ranged from 0 (pure parental *C. z. brevirostris*) to 1 (pure parental *C. f. fasciatus*).

Genetic Diversity

We analyzed genetic diversity using the genetic clusters defined by STRUCTURE when $K=4$. DAPC identified $K=4$ as the most likely number of clusters, while STRUCTURE regarded it as the third most likely. Employing $K=4$ not only enables a finer classification but also resulted in genetic clusters with geographical boundaries that correspond closely to those of biogeographical regions that have been previously reported (Amador et al., 2019; Escribano-Avila et al., 2017; Morrone, 2006; Prieto-Torres et al., 2019) enabling us to explore potential barriers within Western Ecuador.

We assigned each sample to the genetic clusters if their Q-value for that cluster was greater than 0.9. We estimated allele frequencies, inbreeding coefficient (F_{is}) per population, and the observed heterozygosity (H_o) per individual using the function `gl.report.heterozygosity` from the R package `dartR` (Gruber et al., 2018). We estimated expected heterozygosity (H_e) as $2pq$, where p and q are the proportions of each allele within an individual. We then obtained the 2.5%, 25%, 50%, 75%, and 97.5% percentiles of H_o and H_e across individuals within genetic clusters. We minimized the potential bias of related individuals in genetic diversity estimates (Jankovic et al., 2010) by simultaneously selecting samples with low kinship coefficients among birds captured in the same mist net. Nei’s F_{st} estimates and kinship coefficients were estimated with the `HierFstat` R package (Goudet, 2005).

We partitioned the total genotypic variance into components due to differences between genetic clusters and differences between individuals within clusters using analysis of molecular variance (AMOVA) with pairwise Nei’s F_{st} distances between individuals (Nei & Li, 1979). We used the function `gl.amova` of the `dartR` package (Gruber et al., 2018) and evaluated significance levels with 9,999 permutations. Nei’s F_{st} (Nei & Li, 1979) provided a more comprehensive understanding of the genetic differentiation among distinct genetic clusters.

Isolation by Distance and Environment

We used climate variables from CHELSA 1.2 (Karger et al., 2017). CHELSA is a high-resolution (30 arc sec, $\sim 1\text{km}$) free global climate data set. We performed a multiple correlation analysis to identify redundancies among the climatic variables using the `Hmisc` package for R (Harrell, 2014). We selected the climatic variables that we considered biologically relevant and had the lowest Pearson correlation coefficients with other selected variables to avoid collinearity. After this process, we selected Annual Mean Temperature (AMT), Annual Mean Precipitation (AMP), and Precipitation Seasonality (PS).

We explored patterns of isolation by distance (IBD) and isolation by environment (IBE) using partial Mantel tests and Generalized Dissimilarity Models (GDM), along with two distinct dissimilarity datasets. The dissimilarity data sets were based on the mean per sampling location of pairwise Nei’s F_{st} (Nei & Li, 1979) and

kinship coefficients among samples, normalized as $1 - \frac{1 - \min(x)}{\max(x) - \min(x)}$ where x is the kinship coefficient between two samples. We estimated pairwise Nei's F_{st} and kinship coefficients using the HierFstat R package (Goudet, 2005). Euclidean distances between coordinates and climate values of each sample were used as predictors of the kinship coefficient matrix. Coordinates and climate values were averaged per sampling location, and then Euclidean distances were estimated and used as predictors for Nei's F_{st} matrix. Geographic and environmental Euclidean distances among the samples were used as predictors for the normalized kinship coefficients. To account for the Andes as a physical barrier, we conducted the partial Mantel tests and GDM using a dataset that included all sampling sites, and another that excluded the eastern sampling sites (Jaen, Chachapoyas, and Celedin).

First, we correlated both dissimilarity matrices against environmental pairwise Euclidian distances controlled by geography using a partial Mantel set up at 9,999 permutations in the R package *vegan* (Oksanen, 2013). We used the log transformation of environmental and geographic distances — suggested for two-dimensional habitats — and $\frac{F_{st}}{(1 - F_{st})}$ for genetic distances following Rousset (1997). Because Mantel test tends to inflate type I error (Guillot & Rousset, 2013), we rejected the null hypothesis of no significant correlation if $p\text{-value} \leq 0.001$ (Diniz-Filho et al., 2013). Next, Generalized Dissimilarity Modeling (GDM; Ferrier and Guisan 2006, Ferrier et al. 2007, Manion 2009) was used to evaluate the association between both genetic dissimilarity datasets as the response variable, and environmental and geographic Euclidian distance as predictor variables. This statistical method uses matrix regression to investigate the relationships between dissimilarities in predictor and response variables, and it has been increasingly used in landscape genetic studies (Freedman et al., 2010; Geue et al., 2016; Thomassen et al., 2011). The GDM model combines multiple matrix regressions (I-splines) into a single non-linear function to analyze how dissimilarity between pairs of sampling locations responds to environmental gradients and geographical distance. In particular, the partial regressions of GDM take into account two important factors: (1) the non-stationary rate of change along an environmental gradient, and (2) the curvilinearity that characterizes the relationship between dissimilarity and environmental gradients (Ferrier et al., 2007; Fitzpatrick et al., 2013). We used the default of three I-splines per predictor. The significance of the model and predictors was tested with 9,999 permutations using the function `gdm.varImp` of the '*gdm*' package in R (Fitzpatrick et al., 2022). Significance is estimated using the bootstrapped p -value when the predictor has been permuted. The function also calculates the predictor importance measured as the percent decrease in deviance explained between the full model and the deviance explained by a model fit with that predictor permuted (Fitzpatrick et al., 2022).

Outliers Selection

Using two different approaches, we searched for possible effects of environmental variables controlled by latitude on potential local adaptations. First, we conducted a multivariate spatial analysis using Moran Spectral Outlier Detection (MSOD) (Wagner et al., 2017). This technique compares the squared correlation coefficients (power spectrum) — between alleles frequencies of each SNP and Moran eigenvectors maps (MEM) — with the average power spectrum of all SNPs (neutral reference). SNPs that deviate over a predetermined threshold from the average power spectrum are selected as outliers capturing the spatial signal of selection. Then, it uses Moran Spectral Randomization (MSR) to test the association between the outlier SNPs and environmental predictors, accounting for spatial autocorrelation (Wagner et al., 2017). Deviation of SNPs from the average power spectrum was measured using z -scores and p -values ≤ 0.01 . Significant associations with environmental predictors were tested using 9,999 permutations and $p\text{-value} \leq 0.01$. We obtained MEM and MSR using the R package *Adespatial* version 0.3-14 (Dray et al., 2017). Second, we searched for SNP-environmental associations using BAYESCENV version 1.1 (de Villemereuil & Gaggiotti, 2015). This Bayesian approach identifies outlier loci — those with large positive F_{st} values outside of a neutral F_{st} distribution — significantly correlated with environmental predictors. The model considers population and locus-specific F_{st} 's, described by a logistic model with a population-specific parameter β , that captures demographic effects; a locus-by-population interaction term γ that reflects the impact of selection on the variable of interest; and a locus-specific term α , that capture the effects of mutations and other forms of selection (de Villemereuil & Gaggiotti, 2015). We set the prior for the probability of moving away from the neutral model as $\pi=0.1$, and the preference for the parameter α over γ as $P=0.5$.

We used the default parameter settings for the MCMCs. Convergence was tested with Geweke's, and Heidelberg and Welch's convergence diagnostics included in the R package Coda (Plummer et al., 2006). Significant effects of the environmental variables were determined using q-values [?] 0.05 associated with the parameter γ . We regressed climate predictors and latitude to obtain the independent effects of the ecological variables from latitude. We used the average per sampling location of the absolute values of residuals as spatial-independent predictors for BAYESCENV. The environmental values in BAYESCENV should reflect the difference between the observed value in a local population and a reference value (de Villemereuil & Gaggiotti, 2015). As all variables are scaled, this already fulfills the requirement of a relative value. The residuals capture the deviations of the ecological variables from the expected values explained by latitude. The mean of absolute values of residuals never exceeded two, which makes them compatible with the default priors for the standard deviation of the model, and standardization was not necessary.

Gene Ontology

We additionally searched for biological processes (BP), molecular function (MF), and cellular components (CC) associated with the 17 candidate SNPs. We obtained fragment sequences around candidate SNPs from the population.loci.fa file from STACKS. We aligned these sequences to the genome assembly bTaeGut1.4.pri (project GCF_003957565.2) of the zebra finch (*Taeniopygia guttata*) (Formenti et al., 2021). We used the Blast+ command line software (Camacho & Madden, 2023) to perform the alignment, with the following parameters: eval=1 × 10⁻²⁰, word size = 5, and max target seqs = 5000.

We used the chromosome regions of the aligned sequences to obtain gene according to the HUGO Gene Nomenclature Committee (HGNC) and GO identification codes and using the Ensembl database (Cunningham et al., 2022) and the BioMart software (getBM function; Durinck et al. 2009). We then used the GO identification codes to obtain the gene ontology category and description of the GO term using the bioconductor annotation data package (GO.db; Carlson 2019) and the *Gallus gallus* annotated genome database (org.Gg.eg.db; Carlson 2022). Finally, we used the AnnotationDbi package (Pagès et al., 2023) for R to search for gene functions in GO.db associated with the sequences that included the candidate SNPs.

Demographic Scenarios

We used Momi2 to examine alternative two-population demographic models that differ in terms of the presence and timing of gene flow between *C. z. brevirostris* and *C. f. pallescens*: (i) pure isolation, (ii) isolation-with-migration, and (iii) isolation with secondary contact (bidirectional and unidirectional in either direction). Because Momi2 models gene flow as pulse events, we inserted four equally distant episodes of gene exchange as a function of divergence time for the isolation-with-migration model. We kept the effective population sizes (N_e) constant within each model. For the secondary contact model, we constrained the migration events to occur after the most recent time boundary of the Last Glacial Maximum (LGM; ~16,000 years ago) (Heine, 2000). We assumed a mutation rate of 4.9×10^{-9} substitutions per site per generation (Smeds et al., 2016) and a generation time of two years reported for *Campylorhynchus nuchalis* (Bird et al., 2020). We first performed ten optimizations for each model with ancestral N_e set to 1×10^5 and the stochastic_optimization function to set 100 mini-batches and ten iterations to get initial parameter estimates with reduced computational effort. Based on these results, we used the mean of N_e (4.7×10^5) and time since divergence (1.63×10^5) across runs and models as initial values for subsequent runs. Here, we report parameter values from models with 50 optimizations and initial values as mentioned above and the function stochastic_optimization set with 1000 SNPs per mini-batch and ten iterations. We used the parameter estimations of the 50 runs to generate the mean/median and 95% confidence interval for each estimated demographic parameter. For each model, we selected the optimization with the largest maximum likelihood value for model selection. We used the relative Akaike information criterion to select the best-fit model (Sakamoto et al., 1986). Finally, we assessed the effect of ancestral N_e on model selection by running each model twice as described in step two but with ten optimizations and ten different ancestral N_e ranging from 1×10^5 to 1×10^6 .

RESULTS

After filtering, we obtained 4,409 SNPs with an average coverage per individual of 35.4x (min=12.8x, max=77.9x) and an average genotype call rate of 93.5% from 112 individuals and 16 sampling locations.

Population Structure and Admixture Patterns

According to the sPCA analysis, the first principal component explained 40% of the variance in the dataset, while the second principal component explained 8.7%. We found significant evidence for global structure in the dataset (p-value<0.001), but not for local structure (p-value=0.474). Upon visually inspecting the sPCA results, we observed two major clusters corresponding to the species-level distribution of *C. fasciatus* and the subspecies *C. z. brevirostris* in the northwest region of Ecuador (Figure 2).

STRUCTURE identified K=2 as the most likely number of genetic clusters (Delta K=26,357.04). The second most likely number of genetic clusters was K=3 (Delta K=1,880.47), followed by K=4 (Delta K=351.61) (Figure 3). The slope of log-likelihood values of the data against the number of clusters reaches the asymptote at K=2-4 (Figure S1). We also identified K=2-4 as having the most biological meaning, resembling the clusters observed in the sPCA. Samples from the study region showed a clinal transition from *C. z. brevirostris* from the North to *C. fasciatus* to the south, with admixed individuals falling out in the center (Figure 3). The additional DAPC identified four genetic clusters (Figure S2) as the optimal model (K=4, BIC=627) and cuts of genetic clusters resembled those from STRUCTURE. Further results from the DAPC showed a sharp decline of BIC for K=2 (641.77) and 3 (630.81).

At K=2, the distribution of assigned groups resembles the geographical boundary previously reported at the species level with Q values dropping under 0.9 at the sample sites of Chone and as far as Arenillas. When K=3, the assigned group of *C. fasciatus* on the west slope of the Andes matches the geographical distribution of the subspecies *C. f. pallescens*. The occurrence of admixture between *C. z. brevirostris* and *C. f. pallescens* were predominantly observed in the sampling sites spanning from Chone to Manglares Churute. Hereafter, we refer to this genetic cluster as *C. f. pallescens* North. Admixture between *C. f. pallescens* North and *C. f. fasciatus* was identified in the Southwest region of Ecuador and Northwest region of Peru. This genetic cluster is henceforth referred to as *C. f. pallescens* South (Figure S3).

Hybrid Index HI (proportion of alleles inherited from parental *C. f. fasciatus*) estimated by INTROGRESS showed a cline from the northern-most population of *C. z. brevirostris* (HI=0) to the southern-most population of *C. f. fasciatus* (HI=1) (Figure 4). HI estimates for individuals across sampling locations showed a gradual increase in the frequency of alleles from parental populations of *C. f. fasciatus*.

Genetic Diversity

The observed heterozygosity decreased whereas the expected heterozygosity increased when moving towards the southern genetic clusters of *C. f. fasciatus*. *C. f. pallescens* North was the only genetic cluster that showed higher H_o over H_e (Figure S4 and Table S1).

AMOVA showed that 85.4% ($\sigma^2=0.004$, p-value<0.001) of the genetic variation in our data set was within, while 14.6% ($\sigma^2=0.025$, p-value<0.001) was among genetic clusters (Table S2). Nei's F_{st} showed that the genetic differentiation among distinct genetic clusters was primarily attributed to the differentiation between *C. z. brevirostris* and other genetic clusters. *C. z. brevirostris* showed the highest Nei's F_{st} value (0.088) when compared to *C. f. pallescens* South, and the lowest (0.058) when compared to *C. f. fasciatus* (Table S3).

Isolation by Distance and Environment

Geographical distance was the main factor explaining genetic variation across all analyses, regardless of whether the eastern Andes sampling sites were included, the genetic distance metric used (Nei's F_{st} or kinship coefficient), or the statistical method (Mantel test or GDM). The only exception was the Mantel test using Nei's F_{st} , which did not reach significance at the α level of 0.001 ($r=0.638$, p-value=0.0013) (Table 1).

AMP was a significant predictor of genetic variation in all models using kinship coefficients, both with and without the eastern Andes sampling sites. It was also significant in the GDM using Nei's F_{st} distances, but

only when the eastern Andes sampling sites were excluded (predictor importance=9.088, p-value<0.05). PS was only significant in predicting kinship coefficients in the Mantel test without the eastern Andes sampling sites ($r=0.157$, p-value<0.001). AMT was not significant in any of the models (Table 1).

Outliers Selection

The MSOD analysis identified seven outlier SNPs (p-value [?] 0.01) putatively under selection for at least one environmental variable. MSOD identified latitude as the variable with the highest number of associated candidate SNPs ($n=6$), followed by AMP ($n=4$) and PS ($n=2$). MSOD did not detect any SNP associated with AMT. BAYESCENV analysis identified ten putative SNPs as significant for at least one environmental predictor (q-value [?] 0.01 for γ). Six SNPs were associated with AMT, three with PS and latitude, and two with AMP. No SNPs were identified with both techniques.

Gene Ontology

Searching of the 17 selected outliers yielded 7 genes and 48 GO terms. The GO terms were classified into 15 biological processes, 15 molecular functions, and 18 cellular components. The gene coding for SMG6 nonsense-mediated mRNA decay factor was associated with the most GO terms (Table S4).

Demographic Scenarios

The best-fit demographic model was isolation with secondary contact and asymmetric gene flow from *C. z. brevirostris* towards *C. f. pallescens* North (For the ancestral $N_e = 4.7 \times 10^5$, AIC = 23,800.64; Table S5). This model generally had the largest relative likelihood and lowest AIC across the two replicates and ten different ancestral N_e (Figure S5), except for one of the replicates with ancestral N_e of 2×10^5 , 3×10^5 , and 7×10^5 where three different models were selected. N_e at the time of divergence in the isolation model with secondary contact and North-South gene flow was 2.77×10^6 individuals (95% CI $2.77 \times 10^6 - 2.79 \times 10^6$) for *C. z. brevirostris* and 6.87×10^4 (95% CI $5.90 \times 10^4 - 6.96 \times 10^4$) for *C. f. pallescens* North. Gene flow from the north to the south was 33.38% of N_e (95% CI 30.01% – 44.21%). *C. z. brevirostris* and *C. f. pallescens* diverged around 0.93 million years ago (Ma) (95% CI $8.48 \times 10^4 - 1 \times 10^5$ years) (Table S6). Some parameter estimates, specifically migration rates, exhibited a pathological runaway behavior — where the inferred population sizes and epoch durations can degenerate to zero or diverge to infinity — common in SFS-based demographic inference algorithms (Rosen et al., 2018) and therefore should be interpreted with caution.

DISCUSSION

We show evidence of genetic structure in western Ecuador and northern Peru. The boundaries of identified groups correspond to the species replacement of *C. z. brevirostris* and *C. f. pallescens* and physical barriers to dispersal. The most likely clusters ranged from $K=2-4$, corresponding to categories defined by geographic origins, estimated phylogenies, and known physical or ecological constraints. Evidence for IBE due to AMP and IBD was strong across statistical analyses and datasets. We also observed admixture, with a gradual transition between *C. f. pallescens* and *C. z. brevirostris* along the environmental gradient. Genetic differentiation of the two populations of *C. f. pallescens* could be driven by a previously undescribed potential physical barrier near the center of western Ecuador. Lowland habitats in this region may be limited due to the proximity of the Andes to the coastline, limiting dispersal and gene flow, particularly among dry-habitat specialists since much of the habitat is mangrove, wetland, and wet forest.

Population Structure and Admixture Patterns

Although differentiation between western and eastern populations of *C. fasciatus* was expected due to the Andes as a barrier to gene flow, the differentiation of the two western populations of *C. f. pallescens* was not expected. Sampling sites assigned to *C. f. pallescens* in Southwest Ecuador and Northwest Peru formed a discrete group distinct from those in Midwest Ecuador. Heterozygosity patterns coupled with the cline of the ancestry probabilities from STRUCTURE (Figure 3), DAPC (Figure S2), and HI (Figure 4) suggested hybridization between *C. f. pallescens* and *C. z. brevirostris* in the contact zone in western Ecuador (Chone).

We believe that admixture events between *C. z. brevisrostris* and *C. f. pallescens* North are suggested to have played a role in the much higher H_o estimated for *C. f. pallescens* North than found in other clusters (Table S1). The contact zone between hybridizing taxa is expected to exhibit higher levels of heterozygosity (Boca et al., 2020). We hypothesize that reduced gene flow across the Andes may have contributed to the low H_o in *C. f. fasciatus*. Nonetheless, it is important to note that small sample sizes may inflate heterozygosity levels (Schmidt et al., 2021), though utilizing appropriate estimators and a substantial number of bi-allelic markers (>1000), it may be possible to use as few as four individuals (Willing et al., 2012). Thus, it is noteworthy that *C. f. fasciatus* had the lowest sample size ($n=12$) and H_o values (Table S1). Furthermore, differences in heterozygosity between *C. z. brevisrostris* and *C. f. fasciatus* could also be explained by a Wahlund effect, characterized by a decrease in heterozygosity due to a fine-scale population subdivision not accounted for in the sampling (Freeland et al., 2011). Nonetheless, no strong evidence was found to support further population subdivision in our population structure analyses that could lead to the Wahlund effect.

Incomplete lineage sorting (ILS) can also generate genetic diversity patterns comparable to those caused by hybridization (Huerta-Sánchez et al., 2014). ILS refers to the retention and stochastic sorting of ancestral polymorphisms (Maddison et al., 2006). ILS and secondary gene flow can be distinguished when geographic distribution information is available by comparing patterns of genetic diversity between pairs of neighboring and distantly located populations of the different species (Muir & Schlötterer, 2005). Gene flow is expected to occur preferentially between neighboring populations, resulting in higher intraspecific genetic diversity and lower interspecific genetic differentiation than between distantly located populations (Petit & Excoffier, 2009). In contrast, shared polymorphisms are expected to be distributed evenly across all populations under the ILS scenario (Petit & Excoffier, 2009; Zhou et al., 2017). In this study, the progressive increase in the frequencies of alleles as shown by the Hybrid Index (HI, proportion of alleles inherited from parental *C. fasciatus*) differs from the expected allele frequency pattern randomly distributed across two species in ILS. Furthermore, the coalescent-based demographic analysis would identify isolation-with-migration as the best-fit model under ILS (S. Wang et al., 2019). In contrast, the best model was isolation with secondary contact and asymmetrical gene flow (Table S5).

Manglares Churute as a Barrier to Gene Flow

The lowlands (from 0 to 800 masl) between the Andes and the coastline constitute a 16km wide corridor near Manglares Churute (a.k.a. Maglares Churute Corridor, MCC). Habitats in the MCC are discontinuous and consist primarily of lentic bodies of water, wetland, second-growth, and evergreen forest with few deciduous and semideciduous remnants (Alava et al., 2007; BirdLife International, 2022). In such geographical settings, dispersal between adjacent sites in a one-dimensional stepping-stone model may be limited (Kimura & Weiss, 1964). Higher precipitation of the Andean slopes breaks the continuity of arid habitats along the MCC, so dispersal and gene flow may be more difficult for dry-habitat specialists in this region. The effect of IBD restricting gene flow is intensified along narrow corridors, particularly for short-range dispersal species (Kimura & Weiss, 1964; Wright, 1943). Typical dispersal distances for most Troglodytidae remain poorly understood. Current knowledge indicates that Cactus Wrens, for instance, can disperse up to a maximum distance of 26 km and an average of two km (Lynn et al., 2022). Additionally, it is known that cooperative breeding systems may impose constraints on dispersal (Hatchwell, 2009). We hypothesize that the *Campylorhynchus* dispersal characteristics and environmental and geographical factors likely contribute to the restricted gene flow along the MCC.

The barrier to gene flow that the MCC imposes on terrestrial lowland species — coupled with anthropogenic threats — might have significant consequences for conservation and evolution (Wagner & Fortin, 2013). Other species that show morphometric and plumage differentiation across the MCC could have similar genetic patterns. For example, Necklaced Spinetail (*Synallaxis stictothorax*) has two races: the nominal *stictothorax* occurs north of MCC whereas *maculata* occurs south of the MCC (Ridgely & Greenfield, 2001). The same pattern is observed for the nominal race of Collared Antshrike (*Thamnophilus bernardi*), which occurs north of MCC, while *piurae* occurs south of the MCC (Ridgely & Greenfield, 2001). We suspect species such as Blackish-headed Spinetail (*Synallaxis tithys*) might exhibit similar genetic differentiation across MCC. If

this is correct, it would mean that cryptic biodiversity in the dry forest of west-central Ecuador might need additional conservation attention. We propose that MCC may be an essential barrier to gene flow for lowland dry-habitat specialists and should be explored in future studies.

Isolation by Distance and Environment

We found potential evidence for IBD and IBE, supported by the significant positive relationship between geographic and AMP and genetic distances. Additionally, we detected 17 SNPs associated with climatic factors suggesting potential local adaptations and selection independent of geographic distance. The extent, pattern, and consistency of gene exchange in transitional zones can be explained by both environment-independent (endogenous) and environment-dependent (exogenous) selection (exogenous) (Pyrón & Burbrink, 2013). Thus, if the selection is exogenous, a clinal genetic pattern like the one reported in this study (Figures 3A and 4) may be maintained through differential selection across an environmental gradient, such as a climatic boundary (Haldane, 1948; Harrison, 2012).

It is important to recognize that although patterns of IBE help identify potential systems for adaptive divergence (I. J. Wang & Bradburd, 2014), evidence for IBE does not necessarily imply that local adaptations are involved. IBE can arise from several mechanisms other than selection and can be confounded with incipient ecological speciation (I. J. Wang & Bradburd, 2014). Discerning the relative contribution of geography and environment in shaping genetic diversity remains challenging (Saenz-Agudelo et al. 2015). One major complication in discriminating between these two factors in evolution is that geographical distance and environmental differences are often correlated (Saenz-Agudelo et al., 2015; I. J. Wang & Bradburd, 2014). Although efforts were made to account for collinearity among predictors in the statistical analyses, it cannot be completely ruled out that collinearity may have affected the reported association in this study.

Demographic Scenarios and Mechanisms of Introgression

The gradual change of the genomic composition (Figures. 3A and 4) and the best-fit demographic model (Table S5) suggest introgression from *C. z. brevirostris* into *C. f. pallescens* North. The second best-fit model was isolation with secondary contact and south-north gene flow, implying that gene flow in the opposite direction may also be possible (Table S6). As far as we know, no hybridization involving *C. fasciatus* has been reported previously, but hybridization between *C. albobrunneus* (White-headed Wren) of western Colombia and Panama and *C. z. brevirostris* of Ecuador has been suggested in the north of Ecuador (Ridgely & Greenfield, 2001).

One driver of asymmetrical introgression, as we observed here, can be caused by sexual selection through a combination of female choice or male-male interactions (Martin & Mendelson, 2016; Stein & Uy, 2006). In some circumstances, the direction of the introgression is likely to be driven by the sex that determines reproductive choices. For instance, heterospecific female pairing preference for the aggressive golden-collared males in a manakin hybrid zone caused asymmetric introgression of plumage traits into the less aggressive white-collared manakin (Stein & Uy, 2006). The lack of female preference for either hetero or conspecifics could produce similar patterns. For example, introgression skewed toward the Small Tree-Finch (*Camarhynchus parvulus*) in the Galápagos Archipelago was associated with the lack of assortative preference of females of the rarer Medium Tree-Finch (*Camarhynchus pauper*) (Peters et al., 2017). Data on female mating preferences are needed to determine whether female choice drives the introgression of *C. z. brevirostris* into *C. f. pallescens* North.

Interspecific territoriality — a common type of interference competition in animals — is strongly associated with bird hybridization, implying that reproductive interference favors the maintenance of interspecific territoriality (Cowen et al., 2020; Drury et al., 2020). Interspecific territoriality leads to confrontation between competitors. As a result — regardless of their foraging efficiency — losers in these interactions are frequently excluded from all resources defended by the dominant individual (Grether et al., 2013; Gröning & Hochkirch, 2008). Interspecific territoriality could be driving the introgression of *C. z. brevirostris* towards *C. f. pallescens* North if the former is the dominant species. Introgression driven by a dominant species is a pattern found in other species (McDonald et al., 2001; Pearson & Rohwer, 2000). Aggressive interference studies are

needed to understand the dominant behavior between these species, and whether it is consistent with the directionality of introgression. While the underlying mechanisms behind male-male interactions and female choice differ, they are not mutually exclusive.

When hybrid fitness depends on ecological conditions, fitness consequences of hybridization may vary with environments or fitness components (Harrison, 2012; Parris, 2001). In such cases, alleles conferring greater fitness under specific ecological conditions may determine the direction of introgression (Coster et al., 2018). Ecological factors might be influencing the introgression of *C. z. brevirostris* into *C. f. pallescens* North if the former has alleles adapted to the climate along the transition zone. An experimental approach to study physiological adaptations or genome-environment association studies could help unravel climate as a driving force for introgression between these two species.

The direction of introgression may be influenced by demographic scenarios, such as the highly different population sizes of hybridizing species (Currat et al., 2008; Lepais et al., 2009). According to the “Hubbs principle” — also known as the “desperation theory” — (Hubbs, 1955; Randler, 2002), birds are more prone to hybridize when the number of individuals in one or both species is limited (Currat et al., 2008; Lepais et al., 2009). The Hubbs principle would predict introgression from *C. f. pallescens* North into *C. z. brevirostris*. In contrast, our results showed that gene flow tends to move towards the larger and more abundant species *C. f. pallescens* North.

CONCLUSION

Different habitat preferences of *Campylorhynchus z. brevirostris* and *C. f. pallescens*, as well as field observations of potentially admixed individuals, prompted us to search for evidence of isolation-by-distance (IBD) and isolation-by-environment (IBE), and to test hybridization between these species.

We found strong evidence of IBD and some evidence of IBE when we included annual mean precipitation (AMP) in the models. We also found some single-nucleotide polymorphisms (SNPs) highly associated with climate. These findings indicate that the genetic variation of *Campylorhynchus* wrens in Western Ecuador and Peru is driven by dispersal limitations and potential adaptation linked to variation in precipitation.

Geographical and bioclimatic pressures in Western Ecuador shape genetic variation. As the ranges of these and other species shift under global climate change, it is essential to understand how these pressures shape genetic diversity. By studying species’ genetic diversity across complex bioclimatic landscapes, we can gain insights into local adaptation and the factors that shape genetic variation.

Understanding the impact of climate on genetic diversity is essential for effective conservation strategies in the face of climate change. Genetic diversity is a fundamental objective of conservation biology because it facilitates rapid adaptation to environmental changes. By characterizing current ranges and assessing whether species harbor and exchange adaptive genetic variants, we can predict their responses to future climates and inform conservation strategies for wrens and other species with similar distributional patterns.

Populations at niche margins, such as those around the sampling sites of Chone and Manglares Churute, where species ranges approach current and future climate niche limits, likely hold genetic diversity critical for adaptation to changing climate. By overlaying genetic monitoring efforts with areas of niche marginality, we can identify where genetic monitoring coincides with anticipated climate change effects on biodiversity.

We do not propose taxonomic changes, but the admixture observed in *C. f. pallescens* suggests that this described subspecies could be a hybrid between *C. z. brevirostris* and *C. fasciatus*, with different degrees of admixture along western Ecuador and northwestern Peru. Further studies including more *Campylorhynchus* species could help to corroborate this hypothesis. Despite being a major conservation concern, hybridization is also a significant source of novel genetic and phenotypic variation. Hybrid zones provide valuable insights into evolutionary processes. By studying hybrid zones, we can better understand the mechanisms that underlie distribution changes, species interaction dynamics, and adaptive introgression. Furthermore, investigating how hybrid zones respond to climate change can provide a more comprehensive understanding of the influence

of both abiotic and biotic factors on range limits, as well as how interacting species respond to climate change in the Tumbes-Choco-Magdalena biodiversity hotspot in South America.

Genetic differentiation between populations of *C. f. pallescens* across Midwest Ecuador and the disjoint distributions of some dry-habitat specialists suggest the presence of a barrier between Manglares Churute and Arenillas sampling sites. We suggest that this barrier is formed by the proximity of the Andes to the coast and the spatial discontinuity of dry habitats. If this hypothesis is correct, poor dispersers and dry-habitat specialists like the Blackish-headed Spinetail (*Synallaxis tithys*) and the Slaty Becard (*Pachyrhamphus spodiurus*), which are threatened by habitat loss, may experience reduced gene flow across MCC. Reduced gene flow may amplify the synergistic effects of climate change and land use change on these species. These findings could be significant from an evolutionary standpoint and for conserving species that show similar distribution patterns. Further studies using whole-genome sequencing with greater coverage will allow a more holistic understanding of the influence of both abiotic and biotic factors on range limits and how interacting species respond to climate change.

Understanding the relationship between climate and genetic diversity is crucial for effective conservation strategies in the face of climate change. By characterizing current ranges, assessing adaptive genetic variants, and monitoring genetic diversity, we can predict species' responses to future climates and inform conservation efforts.

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Authors Contributions

Luis Daniel Montalvo was responsible for the study design, data collection, library construction, sequencing, statistical analyses, interpreting results, writing, and editing the manuscript. Rebecca T. Kimball contributed to the study design, interpreting results, and manuscript editing. James Austin contributed to the interpretation of results and edition of the document and provided feedback on the statistical analysis. Scott Robinson was involved in the study design, interpretation of results, writing, and editing of the manuscript.

Conflict of Interest

The authors state that there are no conflicts of interest that might affect the research presented in this paper.

Data Archiving

All sequencing data from this study have been deposited at NCBI Sequence Read Archive under Bioproject accession number PRJNA925654 at <https://submit.ncbi.nlm.nih.gov/subs/sra/SUB12536456/overview>. All bioinformatic pipelines, sequence alignments and analytical scripts are available on GitHub <https://github.com/ldmontalvo/Landscape-Genomic-Wrens>. Sample metadata and accession numbers are listed for individual samples in Table S7 of the Supplementary Material.

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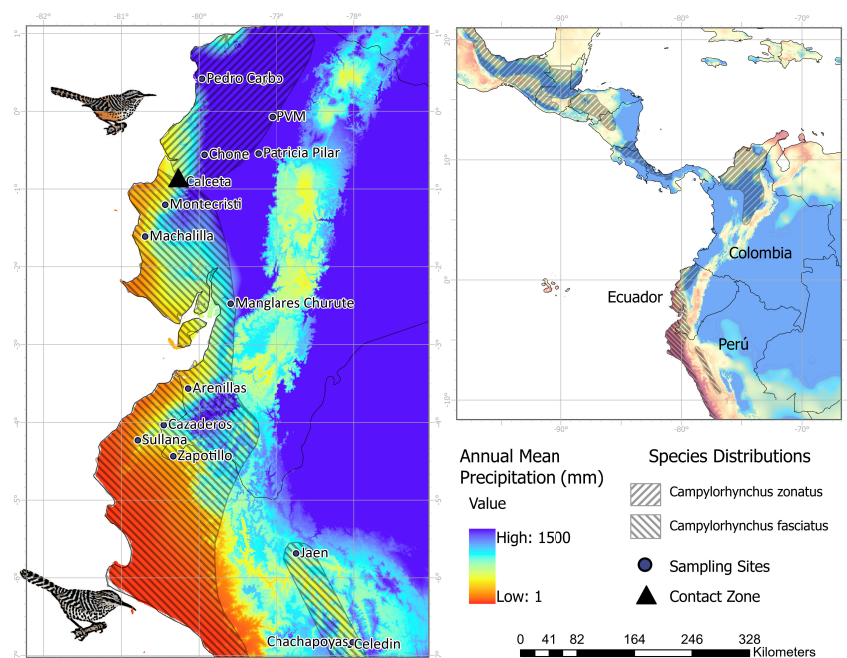
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- Figure 1. Sampling sites for genomic data and distribution ranges of *C. z. brevirostris* and *C. fasciatus*.

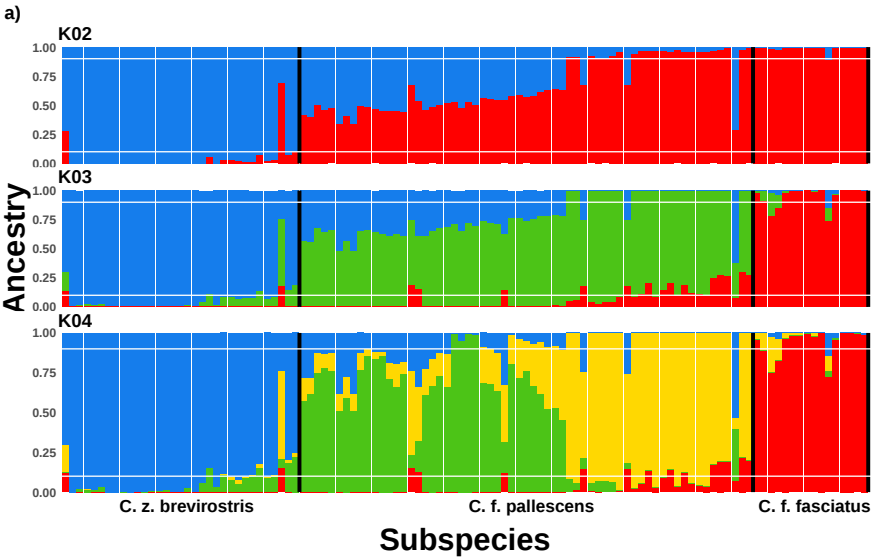
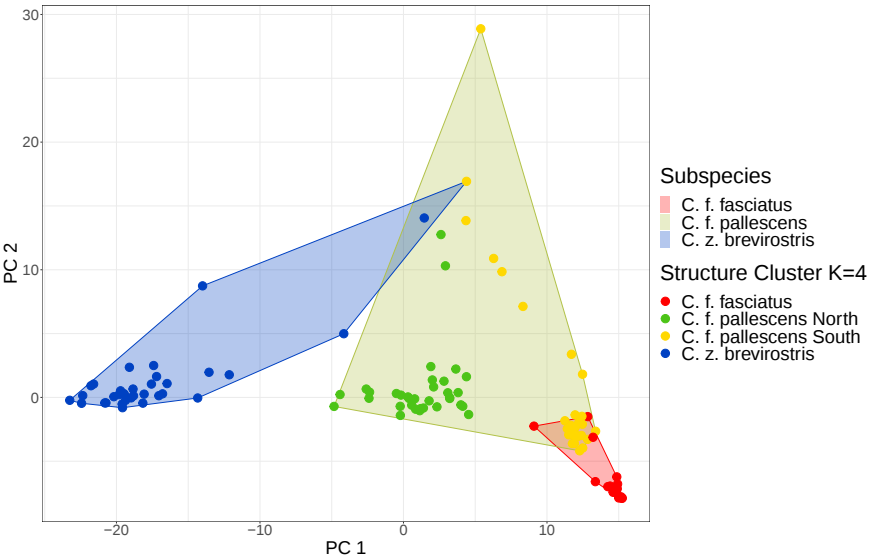
Figure 2. Spatial Principal Component Analyses. The polygons are colored according to the subspecies to which each sample belongs according to Ridgely and Greenfield (2001). The dot colors correspond to the genetic clusters assigned by the analysis with the software Structure when $K=4$.

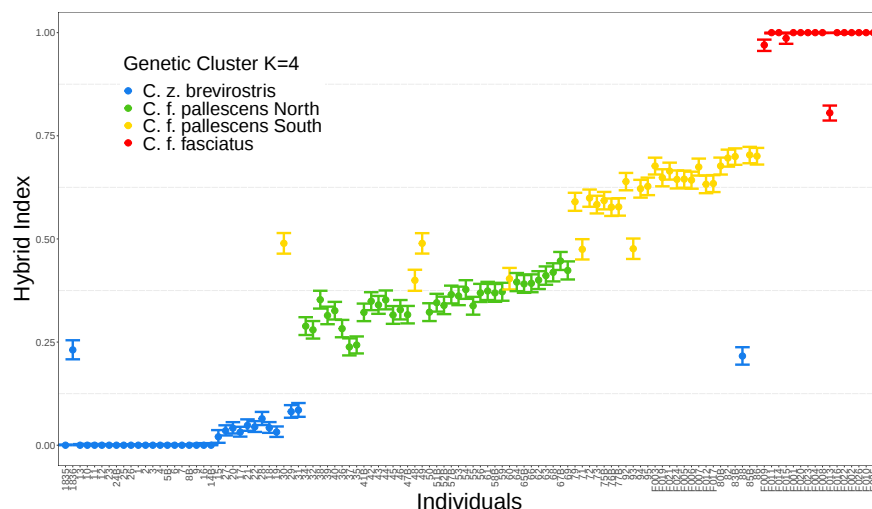
Figure 3. Population Structure. Cluster assignment probabilities as estimated by the software STRUCTURE. Samples are ordered by Latitude. Black vertical lines show approximate boundaries among genetic clusters. White horizontal lines show $Q=0.9$ and 0.1 ancestry probabilities.

Figure 4. Hybrid Index (HI) per individual as estimated by INTROGRESS R package. Individuals are ordered by Latitude. Colors show different genetic clusters. Samples from parental individuals scored zero at the bottom left and one at the upper right.

Table 1 Summary of generalized dissimilarity models (GDMs) and Mantel tests used to explore the effects of geographical and environmental predictors (AMT=annual mean temperature, AMP=annual mean precipitation, and PS=precipitation seasonality) on genetic dissimilarity, as measured by Nei's F_{st} distances among sampling sites and normalized kinship coefficient among samples. We performed two sets of analyses, one including and one excluding eastern Andes sampling sites. Predictor importance was measured as the percent decrease in deviance explained between the full model and a model fit with that predictor permuted. Significance is estimated using bootstrapped p-values, with ** indicating a p-value < 0.05 and *** indicating a p-value < 0.001 .







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