Evolutionary Rates of Bumblebee Genomes Are Faster at Lower Elevations

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Abstract

The importance of climate in determining biodiversity patterns has been well documented. However, the relationship between climate and rates of genetic evolution remains controversial. Latitude and elevation have been associated with rates of change in genetic markers such as cytochrome b. What is not known, however, is the strength of such associations and whether patterns found among these genes apply across entire genomes. Here, using bumblebee genetic data from seven subgenera of Bombus, we demonstrate that all species occupying warmer elevations have undergone faster genome-wide evolution than those in the same subgenera occupying cooler elevations. Our findings point to a critical biogeographic role in the relative rates of whole species evolution, potentially influencing global biodiversity patterns.

Key words: Bombus, diversity, genetic evolution, integrated evolutionary speed, genome, mitochondrial, nuclear.

Species diversity patterns are correlated with environmental variables such as temperature, primary productivity, and biome area (McBride et al. 2014; Gillman et al. 2015; Jonathan and Walter 2015), and declining species richness among bees with increasing elevation has been associated with temperature (Alice et al. 2015). Rensch (1959) suggested that high diversity may derive from greater evolutionary speed in warmer climates, possibly due to shorter generation times and/or higher mutation rates (Rohde 1992). The strength and importance of relationships among environmental and geographical variables and rates of genetic evolution is, however, controversial (Gillman et al. 2011; Weir and Schluter 2011; Rolland et al. 2016). Faster rates of genetic evolution have been found in warmer climates, such as at lower elevations, for a range of taxa including endotherms and ectotherms (Gillman and Wright 2014; Dugo-Cota et al. 2015), and greater intraspecific divergence has been reported from populations occupying warmer environments (Oppold et al. 2016). Life-history traits, including generation time, longevity, metabolic rate, and body mass have also been related to rates of substitution (Nabholz et al. 2008; Santos 2012; Lehtonen and Lanfear 2014; Bromham et al. 2015) and may underpin the geographic and climatic associations with rates of evolution. However, these studies have all relied on relatively small genetic markers from mitochondria and, less often, nuclear DNA, to characterize species-level rates of evolution, and it is unknown whether the results are representative of genome-wide evolution. Therefore, the importance of environment in mediating the rate of evolution remains unclear. Here, we address this issue by comparing rates of genetic evolution across whole nuclear and mitochondrial genomes for seven pairs of bumblebee species that occupy contrasting elevational distributions.

The genus Bombus (Hymenoptera: Apidae) includes ~250 species within 15 subgenera (Cameron et al. 2007; Williams et al. 2008). Elevational segregation of closely related species enabled us to select seven pairs of species, within seven subgenera, with elevations differing between species within each pair by an average of 1,867 m (two-tailed paired t-test, \( P = 9.5 \times 10^{-4} \)) (supplementary table S1, Supplementary Material online). We sequenced 12 of the 14 species (SRA accession No. PRJNAS058540; supplementary table S2, Supplementary Material online) and obtained published genome sequences (Sadd et al. 2015) for the remaining two (B. impatiens and B. terrestris). Nuclear genome sequences (NUC) and mitochondrial genome sequences (MIT) were each concatenated for phylogenetic analysis. A total of 7,738 nucleotide genes with 4,641,249 bp and 13 mitochondrial genes with 10,842 bp were used in our analyses.

Maximum likelihood with CODON models was used for phylogenetic analysis. The best-fit model for NUC and MIT was GY+F+R4 and GY+F+R5, respectively. The topologies
of the two phylogenetic trees (fig. 1) were consistent with each other, and with previous studies (Williams et al. 2008). Branch lengths for all lowland species were longer than their high-elevation counterparts for both NUC and MIT (fig. 1, table 1). Consequently, mean log transformed branch lengths of lowland species were longer than those for high-elevation species (table 2). We also calculated the tip age (total branch length from each tip to the root) of each species (supplementary table S2, Supplementary Material online) and performed correlation analyses between log transformed tip ages and their corresponding mean elevations. Both nuclearistime \( r = -0.653, P = 0.011, n = 14 \) and mitochondrial \( r = -0.736, P = 0.003, n = 14 \) tip ages were negatively correlated with mean elevations.

Thus, all comparisons conformed to the pattern of greater genome evolution in lower-elevation species. Our results confirm a pattern previously found among both ectotherms and endotherms using limited genetic data from marker genes (Gillman and Wright 2014; Dugo-Cota et al. 2015). However, the results from previous studies were less consistent, with approximately one-third of comparisons in each of these studies producing a result contrary to the overall pattern. Here, we show a stronger, indeed universal pattern across our data set derived from full genomes.

Faster rates of genetic evolution at lower elevations may be due to a population size effect. Nearly neutral theory predicts elevated rates of substitution in small populations due to relaxed purifying selection of slightly deleterious mutations (Ohta 1972) thereby elevating nonsynonymous substitutions (dN) relative to synonymous substitutions (dS). We therefore examined dN and dS. Lower-elevation species in all cases exhibited higher dN and dS in both mitochondrial and nuclear DNA than the contrasted higher-elevation species (table 1). The log transformed means for dN and dS of both NUC and MIT were all significantly larger in low-elevation species (table 2). By contrast, there was no indication of a mean difference in dN/dS between high- and low-elevation species for either NUC or MIT (paired-samples t-test for log transformed values, \( P = 0.196 \) and 0.337, respectively) (table 2). Relaxed selection due to small populations cannot therefore explain our results.

The opposite population effect, whereby total mutations increase with population size and thus increase rates of substitution, has been previously modeled (Kimura 1979), and the integrated evolutionary speed hypothesis predicts elevated rates of genetic evolution in larger populations (Gillman and Wright 2014). Longer branch lengths have been reported for mainland bird species with larger populations relative to island species (Wright et al. 2009). The low-elevation species in our study occur in higher densities and are more widely distributed than the high-elevation species (An et al. 2014) and therefore this mechanism provides a potential explanation for our results.

Life-history traits such as body size, metabolic rate, longevity, and generation time have been negatively associated with temperature and with substitution rates (Martin and Palumbi 1993; Welch et al. 2008; Bromham et al. 2015). Life history therefore provides plausible explanations for lower rates in cooler environments. However, in many cases life-history traits, such as longevity and body size, have failed to explain latitudinal or elevational relationships (Gillman et al. 2012; Gillman and Wright 2013; Lourenço et al. 2013). Neither generation time nor longevity is a tenable explanation for the divergent substitution rates we measured because all of the study species have an annual life cycle (Goulson 2010). Likewise, larger bodies at higher elevations are an unlikely explanation for slower substitution rates because the high-elevation species in our study are generally smaller, not larger, than their low-elevation counterparts (Williams et al. 2009).

Lower basal metabolic rates among the high-elevation species are another potential explanation for our results. However, empirical testing with large data sets has previously failed to find significant relationships between basal metabolic rate and substitution rate (Bromham et al. 1996; Lanfear et al. 2007). Active, rather than basal, metabolic rates have, by contrast, been positively associated with rates of genetic evolution (Santos 2012; Gillman and Wright 2013) and may be suppressed at higher elevations due to colder temperatures. A low substitution rate in two thermophiles is thought to result from internal control mechanisms that counterbalance potentially deleterious effects of elevated mutations at extreme high temperatures (Drake 2009; Swami 2009). Similarly, a conserved genetic structure might result due to hypoxia stress among bumblebees occupying cold environments (Cai et al. 2013; Zhang et al. 2013). As potential explanations for variable rates of evolution, these mechanisms deserve further empirical investigation.
We clearly demonstrate faster molecular evolution among low-elevation bumblebee species than among high-elevation species. Further work is needed to identify the mechanisms producing this pattern. Our genomic data also provide fertile ground for investigation into functional implications of genetic divergence in contrasting environments. The unequivocal pattern we reveal among our study species suggests a critical biogeographic role in the relative rates of whole species evolution. Tests for an association between substitution rates for log transformed mean differences.

**Materials and Methods**

Bumblebees (adult workers) were live-trapped using sweep nets and stored in a refrigerator (supplementary table S3, Supplementary Material online). DNA was extracted from each bumblebee and a 350-bp library constructed. The paired-end library was sequenced (Illumina HiSeq2000) with both directions of 150-bp reads representing $\sim 100 \times$ coverage of the genome. Using FASTX-Toolkit (http://hannonlab.cshl.edu/fastx_toolkit; last accessed March 14, 2019) the sequenced data were filtered by removal of adaptors, low-quality reads, and ambiguous reads. The clean reads were used for de novo assembly with IDBA-UD (Peng et al. 2012). The 200-bp longer contigs were used for further analyses. In addition to the 12 species sequenced by us, previously published genomes (Sadd et al. 2015) of *B. terrestris* (RefSeq assembly accession: GCF_000214255.1) and *B. impatiens* (RefSeq assembly accession: GCF_000188095.2) were used.

Mean elevation data for all species we sequenced were obtained by An et al. (2014). For *B. terrestris* and *B. impatiens*, distributions were obtained from the GBIF database (https://www.gbif.org; last accessed March 14, 2019), and average elevations were determined using the ArcGIS platform and the SRTM30 digital elevation model (https://dds.cr.usgs.gov/srtm/version2_1/SRTM30/; last accessed March 14, 2019). Seven sister pairs of bumblebee species, belonging to seven subgenera, were compared with respect to rates of molecular evolution and elevation.

First, we edited the gff file of *B. terrestris* (GCF_000214255.1_Bter_1.0_genomic.gff) downloaded from the National Centre for Biotechnology Information (NCBI) RefSeq assembly database, only keeping the coding sequence (CDS) of the longest exon of each gene. We then extracted the CDSs (CDS-L) with the gffread program from the Cufflinks package (Trapnell et al. 2010) using the gff file and the genomic assembly (GCF_000214255.1_Bter_1.0_genomic.fna).
Second, we translated CDS-L into peptide sequences (PEP-L) using MEGA (Kumar et al. 2016). With PEP-L as query sequences, we used TBlastN from the BLAST program package (Altschul et al. 1997) and seqtk (https://github.com/lh3/seqtk; last accessed March 14, 2019) to determine and extract potential homologous sequences (plus 1,000 bp of upstream/downstream regions). Third, we extracted the CDS region of the potential homologous sequences using the Exonerate program (Slater and Birney 2005). Finally, we extracted CDS-L and PEP-L of each species and identified and aligned the orthologous genes successively using the programs InParanoid (Remm et al. 2001), MultiParanoid (Alexeyenko et al. 2006), MACSE (Ranwez et al. 2011), and PRANK (Löytynoja and Goldman 2005), as previously described (Lin et al. 2014).

The 13 coding genes in the mitochondrial genome were extracted and aligned manually. We downloaded a mitochondrial genome from GenBank (accession number: KT368150.1) and extracted the 13 coding genes. Using each coding gene as the query sequence, we identified and extracted homologous sequences from the other 13 species with the TBlastN program. We then aligned each of the 13 genes manually in MEGA.

All the aligned nuclear and mitochondrial sequences were concatenated into two alignment sets (supplementary tables S4 and SS, Supplementary Material online). The substitution saturation of each of NUC and MIT was tested manually using DAMBE (Xia 2013). We used maximum likelihood with CODEON models for phylogenetic tree reconstruction because they are thought to be biologically more realistic than other substitution models for protein-coding sequence evolution (Gil et al. 2013; Galinskaya et al. 2014). IQ-TREE (Nguyen et al. 2015) was used to reconstruct trees with 1,000 bootstrap replicates. The ModelFinder (Darriba et al. 2012; Kalyaanamoorthy et al. 2017) program was automatically invoked by IQ-TREE to select the best-fitting substitution model for each alignment according to the Bayesian information criterion. The root setting with outgroup taxa does not influence the topology of ingroup taxa. In order to avoid loss of information in the processes of orthologous gene identification between the ingroup and outgroup, we did not use outgroup taxa in the reconstruction. Finally, the branch length as well as the tip age for each species was directly read from the phylogenetic trees. The “several ω ratio” branch model (model = 2) in CODEML in the PAML package (version 4.9h) was used to calculate dN/dS (Yang et al. 2000). The 14 external branches corresponding to the species were viewed as different foregrounds, whereas all the internal branches were viewed as a common background (supplementary table S6, Supplementary Material online). The prior branch lengths generated by IQ-TREE were used with the fix_blength = 3 (proportional) set.

Statistical analyses were performed using SPSS. Several data sets deviated from normal (one-sample Kolmogorov–Smirnov test) and therefore, in order to be consistent, all branch lengths, tip ages, dN, dS, and dN/dS were natural logarithmic (Ln) transformed. The two-tailed paired-samples t-test was used to compare means of each index between the high- and low-elevation species. The Spearman correlation test was performed to test the relationship between the tip ages and their corresponding mean elevations.

### Supplementary Material

Supplementary data are available at Molecular Biology and Evolution online.

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### Author Contributions

G.L., L.W., T.Z., and Z.C. carried out bumblebee sampling. G.L., Z.H., L.N.G., and F.Z. wrote the paper. L.N.G. and F.Z. managed the project and the team. All authors read and approved the final manuscript.

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