Adaptation, ancestral variation and gene flow in a 'Sky Island' *Drosophila* species

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Abstract

Over time, populations of species can expand, contract, fragment and become isolated, creating

subpopulations that must adapt to local conditions. Understanding how species maintain variation after

divergence as well as adapt to these changes in the face of gene flow, is of great interest, especially as the

current climate crisis has caused range shifts and frequent migrations for many species. Here, we

characterize how a mycophageous fly species, *Drosophila innubila,* came to inhabit and adapt to its current

range which includes mountain forests in southwestern USA separated by large expanses of desert. Using

population genomic data from more than 300 wild-caught individuals, we examine four populations to

determine their population history in these mountain forests, looking for signatures of local adaptation. We

find *D. innubila* spread northwards during the previous glaciation period (30-100 KYA), and has recently

expanded even further (0.2-2 KYA). *D. innubila* shows little evidence of population structure, consistent

with a recent establishment and genetic variation maintained since before geographic stratification. We also

find some signatures of recent selective sweeps in chorion proteins and population differentiation in

antifungal immune genes suggesting differences in the environments to which flies are adapting. However,

we find little support for long-term recurrent selection in these genes. In contrast, we find evidence of long-

term recurrent positive selection in immune pathways such as the Toll-signaling system and the Toll-

regulated antimicrobial peptides.

Introduction

In the past 25,000 years, the earth has undergone substantial environmental changes due to both human-

mediated events (anthropogenic environment destruction, desert expansion, extreme weather and the

growing anthropogenic climate crisis) (CLOUDSLEY-THOMPSON 1978; ROSENZWEIG *et al.* 2008) and

events unrelated to humans (glaciation and tectonic shifts) (HEWITT 2000; HOLMGREN *et al.* 2003;

ARIZONA-GEOLOGICAL-SURVEY 2005). These environmental shifts can fundamentally reorganize habitats,

and influence organism fitness, rates of migration between locations, and population ranges (SMITH *et al.*

1995; ASTANEI *et al.* 2005; ROSENZWEIG *et al.* 2008; SEARLE *et al.* 2009; CINI *et al.* 2012; PORRETTA *et*

al. 2012; ANTUNES *et al.* 2015). Signatures of the how organisms adapt to these events, or to population

migration, are often etched in the patterns of molecular variation within and between species

(CHARLESWORTH *et al.* 2003; WRIGHT *et al.* 2003; EXCOFFIER *et al.* 2009).

The Madrean archipelago is an environment currently undergoing extensive environmental change

due to the climate crisis, and presents a good model environment to study the genomic consequences of

local adaptation and gene flow in a changing environment (SMITH AND FARRELL 2005; COE *et al.* 2012;

MANTHEY AND MOYLE 2015; WIENS *et al.* 2019). This range, located in southwestern USA and

northwestern Mexico, contains numerous forested mountains known as 'Sky islands', separated by large

expanses of desert (MCCORMACK *et al.* 2009; COE *et al.* 2012). These 'islands' were connected by

woodland forests during the previous glacial maximum which then retreated, leaving montane forest habitat

separated by hundreds of miles of desert, presumably limiting migration between locations for most species

(ARIZONA-GEOLOGICAL-SURVEY 2005; SMITH AND FARRELL 2005; MCCORMACK *et al.* 2009). The

islands are hotbeds of ecological diversity. However, due to the changing climate in the past 100 years, they

have become more arid and prone to wild fires, which may drive migration, adaptation and even extinction

events (MCCORMACK *et al.* 2009; COE *et al.* 2012; FAVE *et al.* 2015; MANTHEY AND MOYLE 2015).

Several studies of genetic diversity in Sky Island populations have explored the how species persist

in these environments and the extent that gene flow occurs between locations. These studies primarily find

that while birds are mostly unstructured (MCCORMACK *et al.* 2009; MANTHEY AND MOYLE 2015),

flightless species are highly structured (SMITH AND FARRELL 2005; MCCORMACK *et al.* 2009; FAVE *et al.*

2015; WIENS *et al.* 2019), due to the difficulty in migrating between these mountains separated by hundreds

of miles of desert. These studies also highlight that many species are adapting to the increasingly hot and

arid conditions of the Sky Islands, caused by climate change (MCCORMACK *et al.* 2009; COE *et al.* 2012;

MISZTAL *et al.* 2013). Specifically, several animal species are changing their population distributions in the

Sky Islands and adapting to increasingly hostile conditions (FAVE *et al.* 2015; WIENS *et al.* 2019).

Additionally, the delay in the monsoon period has resulted in several plant species pushing back their

flowering period, until adequate water is available (MCCORMACK *et al.* 2009; CRIMMINS *et al.* 2011).

To date, no studies have considered demographics and local adaptation in the Sky Islands from a

whole genome perspective and few have considered demography and local adaptation in insects (FAVE *et*

al. 2015). Luckily, the Madrean archipelago is inhabited by the ecological *Drosophila* model species,

Drosophila innubila (JAENIKE et al. 2003; DYER AND JAENIKE 2005). *Drosophila innubila* is a

mycophageous species found throughout these Sky islands and thought to have arrived during the last

glacial maximum (DYER AND JAENIKE 2005; DYER *et al.* 2005). This 'island' endemic mushroom-feeder

emerges during the rainy season in the Sky Islands for 10-12 weeks in late summer and early fall, before

returning to diapause for the winter and dry seasons (PATTERSON 1954). *D. innubila* and other

mycophageous *Drosophila* are well-studied ecologically, giving us a thorough understanding of their life

history, environment and pathogens (SHOEMAKER *et al.* 1999; PERLMAN *et al.* 2003; DYER AND JAENIKE

2005; DYER *et al.* 2005; JAENIKE AND DYER 2008; UNCKLESS 2011; UNCKLESS AND JAENIKE 2011). For

example, *D. innubila* is infected with a male-killing, maternally transmitted pathogen, called *Wolbachia*

which has been extensively studied (JAENIKE *et al.* 2003; DYER 2004; DYER AND JAENIKE 2005; DYER *et*

al. 2005). *D. innubila* is also frequently exposed to a highly virulent DNA virus (UNCKLESS 2011), as well

as toxic mushrooms (SCOTT CHIALVO AND WERNER 2018). Given these environmental and pathogenic

factors, we expect to identify recent signatures of evolution in the *D. innubila* genome on establishment in
the Sky Islands, as well as signatures of a recent population migration.

Newly established populations are almost always small because they usually consist of a few
founders (CHARLESWORTH *et al.* 2003; EXCOFFIER *et al.* 2009; LI AND DURBIN 2011). This results in a loss
of rare alleles in the population (TAJIMA 1989; GILLESPIE 2004). The population will then grow to fill the
carrying capacity of the new niche and adapt to the unique challenges in the new environment, both signaled
by an excess of rare alleles (EXCOFFIER *et al.* 2009; WHITE *et al.* 2013). This adaptation can involve
selective sweeps from new mutations or standing genetic variation, and signatures of adaptive evolution
and local adaptation in genes key to the success of the population in this new location (CHARLESWORTH *et*
al. 2003; HERMISSON AND PENNINGS 2005; MCVEAN 2007; MESSER AND PETROV 2013). However, these
signals can confound each other making inference of population history difficult. For example, both
population expansions and adaptation lead to an excess of rare alleles, meaning more thorough analysis is
required to identify the true cause of the signal (WRIGHT *et al.* 2003). Additionally, continual migration
between populations can alter the allele frequencies within populations, possibly removing adaptive alleles,
or supplying beneficial alleles from other subpopulations (TIGANO AND FRIESEN 2016). Signatures of
demographic change are frequently detected in species that have recently undergone range expansion due
to human introduction (ASTANEI *et al.* 2005; EXCOFFIER *et al.* 2009) or the changing climate (HEWITT
2000; PARMESAN AND YOHE 2003; GUINDON *et al.* 2010; WALSH *et al.* 2011; CINI *et al.* 2012). Other
hallmarks of a range expansion include signatures of bottlenecks visible in the site frequency spectrum, and
differentiation between populations (CHARLESWORTH *et al.* 2003; LI AND DURBIN 2011). This can be
detected by a deficit of rare variants, a decrease in population pairwise diversity and an increase in the
statistic, Tajima's D (TAJIMA 1989). Following the establishment and expansion of a population, there is
an excess of rare variants and local adaptation results in divergence between the invading population and
the original population. These signatures are also frequently utilized in human populations to identify traits
which have fixed upon the establishment of human populations in a new location, or to identify how our
human ancestors spread globally (LI AND DURBIN 2011).

Populations with limited migration provide a rare opportunity to observe replicate bouts of
evolutionary change and this is particularly interesting regarding coevolution with pathogens, as could be
examined in *D. innubila* (DYER AND JAENIKE 2005; UNCKLESS 2011).We sought to reconstruct the
demographic and migratory history of *D. innubila* inhabiting the Sky islands to understand if migration is
occurring between Sky Islands, and how the species is adapting to its new environment, based on signatures
of selection in the genome (TIGANO AND FRIESEN 2016). Additionally, we hoped to determine if
populations are locally adapting to their recent environment or if signatures of local adaptation are swamped
out by species-level adaptation occurring before population divergence. We resequenced whole genomes

of wild-caught individuals from four populations of *D. innubila* in four different Sky island mountain

ranges. This differs from most population genomic studies of *Drosophila*, which rely on lab maintained

inbred strains. Interestingly, we find little evidence of population structure by location, with structure

limited to the mitochondria and a single autosome. However, we do find some signatures of local

adaptation, such as for cuticle development and fungal pathogen resistance. We also find evidence of

mitochondrial translocations into the nuclear genome, with strong evidence of local adaptation of these

translocations, suggesting potential adaptation to changes in metabolic process of the host between location

(JAENIKE AND DYER 2008). Finally, we find signatures of long-term selection of the Toll immune pathway,

across all populations.

Results

Drosophila innubila has recently expanded its geographic range and shows little divergence between

geographically isolated populations

To characterize how *D. innubila* came to inhabit its current range, we collected flies from four Sky island

locations across Arizona in September 2017: Chiricahuas (CH, 81 flies), Huachucas (HU, 48 flies), Prescott

(PR, 84 flies) and Santa Ritas (SR, 67 flies) (Locations shown in Figure 1). Interestingly, while previous

surveys mostly failed to collect *D. innubila* north of the Madrean archipelago in Prescott (DYER AND

JAENIKE 2005), we easily sampled from that location, suggesting a possible recent colonization (though we

were also unable to collect *D. innubila* in the exact locations previously sampled) (DYER AND JAENIKE

2005). If this was a recent colonization event, it could be associated with the changing climate of the area

leading to conditions more accommodating to *D. innubila*.

To determine when *D. innubila* established in each location and rates of migration between

locations, we isolated and sequenced the DNA from our sampled *D. innubila* populations and characterized

genomic variation. We then examined the population structure and changes in demographic history of *D.*

innubila using silent polymorphism in StairwayPlot (LIU AND FU 2015). We found all sampled populations

have a current estimated effective population size (N_e) of ~1 million individuals and an ancestral N_e of ~4

million individuals, and all experience a bottleneck between 70 and 100 thousand years ago to an N_e of 10-

20 thousand (Figure 1, Supplementary Figure 1A & B). This bottleneck coincides with a known glaciation

period occurring in Arizona (ARIZONA-GEOLOGICAL-SURVEY 2005). Each surveyed population then

appears to go through a population expansion between one and thirty thousand years ago, with populations

settling from south to north (Figure 1A, Supplementary Figure 1A & B). Specifically, while the Huachucas

population appears to have been colonized first (10-30 thousand years ago), the Prescott population was

colonized much more recently (200-2000 years ago). This, and the absence of *D. innubila* in Prescott

sampling until ~2016 sampling suggests very recent northern expansion of *D. innubila* (Figure 1). Note,

however, that StairwayPlot (Liu and Fu 2015) has estimated large error windows for Prescott, meaning
the colonization could be more recent or ancient than the 200-2000 year estimate.

The *Drosophila* genome is organized into Muller elements, these are 6 chromosome arms which
are generally conserved across all *Drosophila* species, named A to F (with A representing the ancestral
Drosophila X chromosome) (Patterson and Stone 1949; Markow and O'Grady 2006; Vicoso and
Bachtrog 2015). While the fusions of these elements differ between *Drosophila* species, the broad
synteny is conserved. In our analysis we refer to the chromosomes by their Muller element names, instead
of any species-specific chromosome name. Given the geographic distance between populations, we
expected to find a corresponding signature of population differentiation across the populations for each
Muller element. Using Structure (Falush *et al.* 2003), we find surprisingly little population differentiation
between locations for each Muller element (Supplementary Figure 1C) but some structure by location for
the mitochondrial genome (Supplementary Figure 1D), consistent with previous findings (Dyer 2004;
Dyer and Jaenike 2005). Together these suggest that there is either still a large amount of standing
ancestral variation shared between populations, or that there is gene flow between populations mostly via
males.

Figure 1: A. Schematic of the route of colonization of *D. innubila*, inferred from population size history
based on StairwayPlot results across the four sample locations in Arizona (AZ), Chiracahua's (CH, red),
Huachucas (HU, blue), Prescott (PR, green) and Santa Ritas (SR, purple). Stairway plot results used to
infer this history are shown in Supplementary Figure 1. **B.** Median pairwise F_{ST} between populations,
comparing the autosomal nuclear genome (below diagonal) and the mitochondrial genome (above
diagonal).

Given that our populations are either recently established or have high levels of migration, we
sought to determine which demographic model best explains the pattern of shared nuclear and

mitochondrial variation seen in *D. innubila*. The relatively recent establishment of these populations

suggests they may not be at migration-drift balance meaning that F_{ST} may not be an appropriate statistic to

determine the extent of gene flow (GUTENKUNST *et al.* 2010; PUZEY *et al.* 2017). Using the silent nuclear

polymorphism of each population in δaδi (GUTENKUNST *et al.* 2010), we found the best fitting model for

the nuclear variation in each pair of populations. In nearly all cases, the best model was one considering all

populations as a single population following a population bottleneck and expansion, as opposed to separate

populations with or without migration. This was true whether we considered all variation together or

separately for each Muller element (Supplementary Table 3, *p*-value > 0.94). This suggests that the shared

variation between populations is due to ancestrally maintained variation as opposed to elevated gene flow

homogenizing new mutations arising in particular populations. Contrasting this, we find that the variation

in mitochondrial genome fits a model of separate populations with some migration (mean migration rate =

0.0525, *p*-value < 0.00794) for all but one pair of populations (HU and SR best fit the model of a single

population following a bottleneck, *p*-value = 0.423). This suggests that these populations are recently

diverged, and HU and SR should be considered as a single population in all regards (Figure 1). The smaller

effective population size of the mitochondrial genome may have resulted in it reaching migration-drift

equilibrium faster than the nuclear genome.

Though not an appropriate model to determine population divergence, we used F_{ST} as a measure of

shared variation (WEIR AND COCKERHAM 1984), and consistent with the δaδi models we find little

differentiation between nuclear genomes (Figure 1B). We only find 65,899 nuclear SNPs (~2% of called

nuclear SNPs) are exclusive to populations, while most SNPs are shared between at least two populations

(Supplementary Figure 2), suggesting most variation is maintained from before each population established

separately. In contrast, there is higher F_{ST} between mitochondrial genomes in pairwise comparisons (Figure

1B). Both nuclear and mitochondrial δaδi and F_{ST} results are consistent with the Structure/StairwayPlot

results (Supplementary Figure 1).

We next determined whether the δaδi results were consistent with our expectations about a more

recent expansion of *D. innubila* into Prescott. We find the time between population split for Prescott is

significantly lower than all other comparisons (GLM t-value = -2.613, *p*-value = 0.0137), and the time since

population expansion is also significantly lower for PR than other populations (GLM t-value = -2.313, *p*-

value = 0.0275). F_{ST} of the nuclear genome is also significantly higher in Prescott comparisons (Figure 2A,

GLM t-value = 93.728, *p*-value = 2.73e-102), though is still extremely low genome-wide in all pairwise

comparisons involving PR (PR median = 0.0105), with some outliers on Muller element B like other

populations (Supplementary Figures 3 & 4). We also calculated the population genetic statistics pairwise

diversity and Tajima's D for each gene using total polymorphism (TAJIMA 1989). As expected with a recent

population contraction in Prescott (suggesting recent migration and establishment in a new location),

pairwise diversity is significantly lower (Figure 2B, GLM t-value = -19.728, *p*-value = 2.33e-86, Supplementary Table 2) and Tajima's D is significantly higher than all other populations (Figure 2C, GLM t-value = 4.39, *p*-value = 1.15e-05, Supplementary Table 2). This suggests that there is also a deficit of polymorphism in general in Prescott, consistent with a more recent population bottleneck, removing rare alleles from the population (Figure 2C, Supplementary Figure 5). Conversely, the other populations show a genome wide negative Tajima's D, consistent with a recent demographic expansion (Supplementary Figure 5), suggesting a more recent bottleneck in Prescott relative to other populations.

Figure 2: Summary statistics for each population. A. Distribution of F_{ST} per gene for each population versus all other populations. **B.** Distribution of genic pairwise diversity for each population. **C.** Distribution of genic Tajima's D for each population. **D.** F_{ST} distribution for Antifungal associated genes for each population. **E.** F_{ST} distribution for cuticular proteins for each population. **F.** F_{ST} distribution for all immune genes (excluding antifungal genes). In A, B & C all cases significant differences from CH are marked with an * and outliers are removed for ease of visualization. In D, E & F, significant differences from the genome background in each population are marked with a + and white diamond mark the whole genome average of F_{ST} for each population.

Population structure in the D. innubila genome is associated with segregating inversions

Though populations are panmictic for Muller element B (Using δaδi, Supplementary Table 3), F_{ST} is significantly higher on Muller element B compared to all other elements in all populations (Supplementary Figure 3, GLM t-value = 30.02, *p*-value = 3.567e-56) and is enriched for windows in the upper 97.5th percentile of F_{ST} (Supplementary Figure 3, χ^2 test: χ^2 = 79.66, d.f. = 1, *p*-value = 1.06e-16). Additionally, we identified 10kbp windows across the genome in the upper 97.5th percentile for the proportion of population-exclusive SNPs and found Muller B is significantly enriched for population-exclusive SNPs in every population (Supplementary Figure 3 & 4, χ^2 test: χ^2 = 60.28, d.f. = 1, *p*-value = 1.059e-12). All other chromosomes are significantly depleted for population-exclusive SNPs (χ^2 test: χ^2 = 31.90, d.f. = 1, *p*-value = 1.33e-6). On Muller element B, regions of elevated F_{ST} are consistent in all pairwise comparisons between populations (Supplementary Figure 4). Muller element B also has elevated Tajima's D compared to all other Muller elements (Supplementary Figure 5, GLM t-value = 10.402, *p*-value = 2.579e-25), suggesting some form of structured population unique to Muller element B. However, actual values of F_{ST} are still low on Muller element B and populations are considered panmictic when fitting models with δaδi for all 1Mbp windows across Muller element B (Supplementary Table 3) (GUTENKUNST *et al.* 2010), suggesting population structure by location is still minimal (Supplementary Figure 3, Supplementary Table 3, PR Muller B mean = 0.081, Muller C mean = 0.0087, Mitochondrial mean = 0.507).

We attempted to identify if this elevated structure is due to chromosomal inversions on Muller element B, comparing F_{ST} of a region to the presence or absence of inversions across windows (using only inversions called by both Delly and Pindel (YE *et al.* 2009; RAUSCH *et al.* 2012)). We find several putative inversions across the genome at appreciable frequencies (89 total above 1% frequency), of which, 37 are found spread evenly across Muller element B and 22 are found at the telomeric end of Muller element A (Figure 3A). The presence of an inversion over a region of Muller element B is associated with higher genic F_{ST} in these regions (Figure 3A, Wilcoxon Rank Sum test W = 740510, *p*-value = 0.0129), though, when calculating F_{ST} for inversions, we find these inversions are not unique or even at different frequencies in specific populations (Mean inversion F_{ST} = 0.024, Max inversion F_{ST} = 0.22, χ^2 test for enrichment in a specific population *p*-value > 0.361 for all inversions). Genes within 10kbp of an inversion breakpoint have significantly higher F_{ST} than outside the inverted regions, consistent with findings in other species (Figure 3C, GLM t-value = 7.702, *p*-value = 1.36e-14) (MACHADO *et al.* 2007; NOOR *et al.* 2007). However, genes inside inverted regions show no difference in F_{ST} compared to those outside (Figure 3C, GLM t-value = -0.178, *p*-value = 0.859). All regions of Muller element B have higher F_{ST} than the other Muller elements (Figure 3C, outside inversions Muller element B vs all other chromosomes: GLM t-value = 7.379, *p*-value = 1.614e-13), suggesting some chromosome-wide force drives the higher F_{ST} and Tajima's D, opposed to reduced recombination near inversion breakpoints (NOOR *et al.* 2007).

Given that calls for large inversions in short read data are often not well supported (CHAKRABORTY
et al. 2017) and the apparently complex nature of the Muller element B inversions (Figure 3A), we may not
have correctly identified the actual inversions and breakpoints on the chromosome. Despite this, our results
do suggest a link between the presence of inversions on Muller element B and elevated differentiation of
the entire Muller element B between *D. innubila* populations.

Figure 3: Summary of the inversions detected in the *Drosophila innubila* populations. **A.** Location and
frequency in the total population of segregating inversions at higher than 1% frequency and greater than
100kbp. Shaded bars highlight the beginning and end of each inversion (X-axis) on each chromosome, as
well as the inversion frequency (Y-axis). **B.** Tajima's D and **C.** F_{ST} for genes across Muller element B,
grouped by their presence under an inversion, outside of an inversion, near the inversion breakpoints (within
10kbp) or on a different Muller element. All inversions and frequencies inferred compared to the reference
D. innubila genome.

Evidence for increased population differentiation in antifungal and cuticle development genes

Though differentiation is low across most of the genome in each population as populations are recently established and most variants are shared, we still find several genomic regions with relatively elevated differentiation, using F_{ST} as a proxy for divergence. In addition to the entirety of Muller element B, there are narrow windows of high F_{ST} on Muller elements D and E (Figure 4, Supplementary Figure 3 & 4). We attempted to identify whether any gene ontology groups have significantly higher F_{ST} than the rest of the genome. We consider a peak of elevated F_{ST} to be genes in the upper 97.5^{th} percentile of F_{ST} on each chromosome. These peaks for F_{ST} are enriched for antifungal genes in all populations and all in pairwise comparisons (Figure 2D, Supplementary Table 4, GO enrichment = 16.414, *p*-value = 1.61e-10). These antifungal genes (including *Dif, cactus* and *Trx-2* in all populations, *grass* in PR, and *modSP* in CH) are distributed across the genome and so not all under one peak of elevated F_{ST}. Interestingly, this is the only immune category with elevated F_{ST} (Figure 2F), with most of the immune system showing no signatures of increased divergence between populations (Figure 2F, Supplementary Figure 6). This suggests that that fungal pathogens apply the strongest divergent selective pressure among populations.

Genes related to cuticle development also have significantly higher F_{ST}, when compared to all other genes (Figure 2E, Supplementary Table 4, GO enrichment = 5.03, *p*-value = 8.68e-08), which could be associated with differences in the environment between locations (toxin exposure, humidity, *etc*.). Consistent with this result, the peak of F_{ST} on Muller element D (Figure 5, Muller element D, 11.56-11.58Mb) is composed exclusively of genes involved in cuticle development (e.g. *Cpr65Au*, *Cpr65Av*, *Lcp65Ad*) with elevated F_{ST} in these genes in all pairwise comparisons involved SR and HU, and the one vs all SR and HU population comparisons as well as PR comparisons (Figure 5), suggesting that they may be adapting to differing local conditions in those populations.

Two other clear peaks of elevated F_{ST} on Muller element E are also composed of genes in similar genomic categories (e.g. cuticle development). There also appears to be three regions of the *D. innubila* genome with translocated mitochondrial genes (Figure 5). The first peak (Muller element E, 11.35-11.4Mb) is composed exclusively of one of these translocated mitochondrial regions with 3 mitochondrial genes (including *cytochrome oxidase II*). The second peak (Muller element E, 23.60-23.62Mb) contains four other mitochondrial genes (including *cytochrome oxidase III* and *ND5*) as well as genes associated with nervous system activity (such as *Obp93a* and *Obp99c*). We find no correlation between coverage of these regions and mitochondrial copy number (Supplementary Table 1, Pearsons' correlation t-value = 0.065, *p*-value = 0.861), so this elevated F_{ST} is probably not an artefact of mis-mapping reads. However, we do find these regions have elevated copy number compared to the rest of the genome (Supplementary Figure 7, GLM t-value = 9.245, *p*-value = 3.081e-20), and so this elevated divergence may be due to collapsed paralogs. These insertions of mtDNA are also found in *D. falleni* and are diverged from the mitochondrial genome,

suggesting ancient transpositions (HILL *et al.* 2019). The nuclear insertions of mitochondrial genes are also

enriched in the upper 97.5^{th} percentile for F_{ST} in HU and PR, when looking at only autosomal genes

(Supplementary Table 4, GO enrichment = 4.53, *p*-value = 3.67e-04). Additionally, several other energy

metabolism categories are in the upper 97.5^{th} percentile for F_{ST} for all pairwise comparisons involving CH.

We find peaks of elevated pairwise diversity exclusively on the mitochondrial translocations

(Supplementary Figure 8), suggesting unaccounted for variation in these genes which is consistent with

duplications detected in these genes (RASTOGI AND LIBERLES 2005) (Supplementary Figure 7). This

supports the possibility that unaccounted for duplications may be causing the elevated F_{ST} pairwise diversity

in mitochondrial genes (Supplementary Figures 6-8). Overall these results suggest a potential divergence

in the metabolic needs of each population, and that several mitochondrial genes may have found a new

function in the *D. innubila* genome and may be diverging (and changing in copy number, Supplementary

Figure 7) due to differences in local conditions.

Figure 4: Comparison of estimated statistics across the *D. innubila* genome for the Prescott (PR)

population. Values are as follows: the average pairwise divergence per gene (D_{XY}, Prescott vs. all other

populations), the gene wise population fixation index (F_{ST}, Prescott vs. all other populations), within

population pairwise diversity per genes, Composite Likelihood Ratio (CLR) per SNP calculated using

Sweepfinder2, and gene wise within population Tajima's D. For D_{XY}, F_{ST}, Pairwise Diversity and Tajima's

D, Genes or windows in the upper 97.5^{th} percentile per chromosome are colored red. The lower 2.5^{th}

percentile per chromosome is also colored in red for Tajima's D. For Composite Likelihood Ratio the upper

99th percentile windows are colored red. Shaded regions on the F_{ST} plot are further examined in Figure 5.

There has been considerable discussion over the last several years about the influence of demographic processes and background selection on inference of local adaptation (CUTTER AND PAYSEUR 2013; CRUICKSHANK AND HAHN 2014; HOBAN *et al.* 2016; MATTHEY-DORET AND WHITLOCK 2018). In contrast to F_{ST} which is a relative measure of population differentiation, D_{XY} is an absolute measure that may be less sensitive to other population-level processes (NEI 1987; CRUICKSHANK AND HAHN 2014). In our data, the upper 97.5th percentile for F_{ST} also significantly overlaps with the upper 97.5th percentile for D_{XY} (Figure 4, Supplementary Figures 3, 4 & 8, χ^2 text: $\chi^2 = 62.61$, p-value = 2.54e-14). The upper 97.5th percentile for D_{XY} is enriched for chorion proteins in all pairwise comparisons and antifungal proteins for all pairwise comparisons involving PR (Supplementary Table 6, *p*-value < 0.05). Further, the high peak of F_{ST} over Muller element D cuticle proteins seen in all populations is also present when using D_{XY} (Figure 4). We find no evidence for duplications in the antifungal, cuticle or chorion proteins, suggesting the elevated F_{ST} and D_{XY} is likely due to local adaptation and not because of unaccounted for copy number variation (Figure 4, Supplementary Figure 7).

Figure 5: Gene-wise F_{ST} showing regions of elevated divergence between populations for each population.

Plot shows F_{ST} for each gene in these regions to identify the causal genes. Genes with noted functions

(cuticle development or mitochondrial translocations) are shown by point shape. Note the Y-axes are on

different scales for each plot. Genes in the upper 97.5th percentile of F_{ST} on each chromosome are colored

in red.

Evidence for recent selective sweeps in chorion genes in each population

Recent adaptation often leaves a signature of a selective sweep with reduced polymorphism near the site of

the selected variant. We attempted to identify selective sweeps in each population (and identify similar

selective sweeps in multiple populations) using Sweepfinder2 to calculate the composite likelihood of a

selective sweep (HUBER *et al.* 2016). We first assessed if percentile of F_{ST} and composite likelihood overlap,

suggesting that the elevated F_{ST} is due to a recent selective sweep (NIELSEN 2005). We find no evidence of

selective sweeps overlapping with genes with elevated F_{ST} (Supplementary Figure 9A, χ^2 test for overlap

of 97.5th percentile windows $\chi^2 = 1.33$ *p*-value $= 0.249$), suggesting that the divergence between populations

is likely not because of local adaptation. This is consistent with the high levels of ancestrally maintained

variation we find and the recent establishment of each population (Figure 1). The upper 97.5th percentile of

composite likelihood is equally distributed across the genome and are not enriched for any gene ontology

categories (*p*-value > 0.31), but if we limit our survey, we find the upper 99th percentile is enriched for

chorion/egg development genes in all populations (Enrichment $= 5.12$, *p*-value $= 0.00351$).

We find one peak of extremely high composite likelihood values present in all populations on

Muller element D, but most extreme in Prescott (Figure 4, Supplementary Figure 9B, Muller D, 18.75-

19Mb). The center of this peak is upstream of the cuticle protein *Cpr66D* and four chorion proteins (*Cp15,*

Cp16, Cp18, Cp19), with several cell organization proteins (*Zasp66, Pex7, hairy, Prm, Fhos*) within 10kbp

of the sweep center. These specific chorion proteins do not have elevated F_{ST} or D_{XY} compared to the rest

of the genome (Wilcoxon Rank Sum W = 399170, *p*-value = 0.273), but has significantly elevated D_{XY}

compared to other genes within 50kbp (Supplementary Figures 8 & 9, Wilcoxon Rank Sum W = 45637000,

p-value = 0.0158). However, this may be due to the reduced diversity seen in these chorion genes because

of the selective sweep (MCVEAN 2007).

Finally, we find evidence of several peaks of selective sweeps in the non-recombining telomere of

the X chromosome (Muller A, 39.5-40.5Mb), among several uncharacterized genes (Supplementary Figure

9). Consistent with this, we find F_{ST} is elevated between our 2001 and 2017 Chiricahua samples, suggesting

at least one recent sweep on the X telomere (Supplementary Figure 10A), resulting in different variants

fixing in the 16 years between samplings.

Toll-related immune show signatures of recurrent positive selection in D. innubila

Finally, we sought to if identify genes and functional categories showing strong signatures of selective

sweeps and divergence are also undergoing adaptive evolution, suggesting species-wide long-term

evolution as opposed to recent adaptation. We reasoned that if the elevated population differentiation seen

in antifungal genes and cuticle development proteins (Figure 2 & 3, Supplementary Figure 6) was due to

adaptation also acting over longer time periods, we would expect to see signatures of ancient adaptation in

those categories. Furthermore, Hill *et al.* used d_N/d_S-based statistics to show that genes involved in some

immune defense pathways were among the fastest evolving genes in the *D. innubila* genome (HILL *et al.*

2019). We also sought to identify what genes are evolving due to recurrent positive selection in *D. innubila*

in one or all populations. To this end we calculated the McDonald-Kreitman based statistic direction of

selection (DoS) (Stoletzki and Eyre-Walker, 2011) and SnIPRE selection effect (Eilertson et al., 2012) to

estimate the proportion of substitutions fixed by adaptation and identify genes under recurrent selection.

We then fit a linear model to identify gene ontology groups with significantly higher DoS or selection effect

than expected. In this survey we found cuticle genes and antifungal genes did have some signatures of

adaptive evolution (DoS > 0, Selection Effect > 0 and significant McDonald-Krietman test for 80% of genes

in these categories) but as a group showed no significant differences from the background (GLM t-value =

1.128, *p*-value = 0.259, Supplementary Table 5). In fact, we only found two functional groups significantly

higher than the background, Toll signaling proteins (GLM t-value = 2.581 *p*-value = 0.00986,

Supplementary Table 4) and antimicrobial/immune peptides (AMPs, GLM t-value = 3.66 *p*-value =

0.00025, Supplementary Table 4). In a previous survey we found that these categories also had significantly

elevated rates of amino acid divergence (HILL *et al.* 2019). These results suggest that this divergence is

indeed adaptive. We find no evidence that genes under recent selective sweeps are undergoing recurrent

evolution, suggesting these genes are only recent targets of selection (Supplementary Table 4).

379  Five immune peptides showed consistently positive DoS and selection effect values (which are also
380 among the highest in the genome): four *Bomanins* and *Listericin*. The *Bomanins* are immune peptides
381 regulated by Toll signaling, while JAK-STAT regulates *Listericin* (HOFFMANN 2003; TAKEDA AND AKIRA
382 2005). *Listericin* has been implicated in the response to viral infection due to its expression upon viral
383 infection (DOSTERT *et al.* 2005; ZAMBON *et al.* 2005; IMLER AND ELFTHERIANOS 2009; MERKLING AND
384 VAN RIJ 2013), additionally Toll-regulated AMPS have been shown to interact with the dsDNA Kallithea
385 virus in *Drosophila melanogaster* (PALMER et al. 2018). *D. innubila* is burdened by Drosophila innubila
386 nudivirus (DiNV), a Nudivirus closely related to Kallithea virus, that infects 40-80% of individuals in the
387 wild (UNCKLESS 2011; HILL AND UNCKLESS 2020). Consistent with the adaptation observed in Toll
388 signaling proteins, this suggests the Toll immune system is under long term selection to resist infection by
389 this DNA virus. However, Toll is also the major component of humoral defense against Gram-positive
390 bacteria and fungi (KIMBRELL AND BEUTLER 2001). Since *D. innubila* feed and breed in rotting mushrooms
391 (JAENIKE AND PERLMAN 2002), we can't rule out that the recurrent adaptation in Toll genes is related to
392 adaptation to the conditions of their pathogen-infested habitats. For all immune categories, as well as cuticle
393 proteins and antifungal proteins, we find no significant differences between populations for either MK-
394 based statistics, and no significant differences in the distribution of these statistics between populations
395 (GLM t-value < 0.211, p-value > 0.34 for all populations, Supplementary Table 4). Selection at these loci
396 is likely long term due to an ancient selective pressure and has been occurring since before the separation
397 of these populations. Mutation rates, efficacy of selection and population structure can vary across the
398 genome, which can confound scans for selection (CHARLESWORTH *et al.* 2003; STAJICH AND HAHN
399 2005).To work around this, we employed a control-gene resampling approach to identify the average
400 difference from the background for each immune category (CHAPMAN *et al.* 2019). We find no signatures
401 of ancient positive selection in antifungal genes (Supplementary Figure 11, 61% resamples > 0) or cuticle
402 genes (Figure 6, 54% resamples > 0) but do again find extremely high levels of positive selection in AMPs
403 (Figure 6, 100% resamples > 0) and Toll signaling genes (Figure 6, 99.1% resamples > 0). Segregating
404 slightly deleterious mutations can bias inference of selection using McDonald-Kreitman based tests
405 (MESSER AND PETROV 2012). To further account for this bias, we also calculated asymptotic α for all
406 functional categories across the genome (HALLER AND MESSER 2017). As before, while we find signals for
407 adaptation in antifungal and cuticle proteins (asymptotic α > 0), values in both categories do not differ from
408 the background (Supplementary Figure 11, permutation test Antifungal *p*-value = 0.243, Cuticle *p*-value =
409 0.137). As with the basic McDonald-Kreitman statistics (Figure 6), the only categories significantly higher
410 than the background are Toll signaling genes (Permutation test *p*-value = 0.033) and AMPs (Permutation
411 test *p*-value = 0.035). Together these results suggest that while genes involved in antifungal resistance and
412 cuticle development are slightly divergent between populations (Figure 5), this evolution has likely not

been occurring over long-time scales and so is restricted to their current environments. Alternatively, the adaptation may be too recent to detect a signal using polymorphism and divergence-based metrics. In either case, signals of longer-term adaptation in *D. innubila* appear to be driven by host-pathogen interactions (possibly with the long-term viral infection DiNV (HILL *et al.* 2019)) as opposed to local adaptation.

Figure 6: McDonald-Kreitman based statistics for immune categories in *D. innubila* and cuticle development. The left two plots show estimated statistics (Direction of Selection and Selection Effect) for each gene. The right two plots show the difference in mean statistic (Direction of Selection and Selection Effect) for each gene and a randomly sampled nearby gene.

Discussion

Interpopulation divergence is a time-dependent process where pro-divergence factors of genetic drift and local adaptation oppose ancestral variation and migration (RANKIN AND BURCHSTED 1992; GILLESPIE 2004; EXCOFFIER *et al.* 2009; WHITE *et al.* 2013). We sought to examine the extent that these factors drive divergence, and the genomic basis of colonization, divergence and local adaptation in *Drosophila innubila,* a geographically structured species found across four forests separated by large expanses of desert. We characterized the phylogeographic history of four populations of *Drosophila innubila* (Figure 7), a mycophagous species endemic to the Arizonan Sky islands using whole genome resequencing of 318 wild-caught individuals. *D. innubila* expanded into its current range during or following the previous glacial maximum (Figure 1, Supplementary Figure 1). Interestingly, there is very little support for population structure across the nuclear genome (Figures 1 & 2, Supplementary Figures 1-4), including in the repetitive

content (Supplementary Figure 12), but some evidence of population structure in the mitochondria, as found previously in *D. innubila* (DYER AND JAENIKE 2005). There are two models which could explain this pattern: the first is extremely high migration, but of mostly males (as seen in other disparate species (RANKIN AND BURCHSTED 1992; SEARLE *et al.* 2009; MA *et al.* 2013; AVGAR AND FRYXELL 2014)), resulting in a panmictic nuclear genome but diverged mitochondrial genome (as shown in Figure 1B). The second model (which we found better support for using δaδi, Supplementary Table 3) is that ancestral variation is maintained due to the recent isolation of populations (less than $4N_e$ generations ago) and recent coalescent times (Figure 1B, Supplementary Figure 3), though the elevated mutation rate and reduced N_e of the mitochondria results in an increased rate of differentiation by drift between mitochondrial genomes (GILLESPIE 2004). Similarly, Drosophila innubila Nudivirus, a DNA virus infecting *Drosophila innubila,* is also highly structured, possibly due to the smaller N_e, elevated mutation rate and high levels of adaptation (HILL AND UNCKLESS 2020). *D. innubila* is also parasitized by a male-killing strain of the maternally transmitted, intracellular bacterium, *Wolbachia.* This alphaproteobacterium infects about 65% of arthropods (WERREN *et al.* 2008), and has evolved several strategies to improve their chances of transmission between generations, including killing male offspring in *D. innubila* (DYER AND JAENIKE 2005; DYER *et al.* 2005; JAENIKE AND DYER 2008; WERREN *et al.* 2008). The mitochondrial divergence could be driven by linkage to the *Wolbachia*: when *Wolbachia* variants sweep (such as variants that improve parasitism), mitochondrial variants will hitchhike to fixation (JAENIKE *et al.* 2003; DYER AND JAENIKE 2005). This will then further decrease the effective population size of the mitochondria, reaching migration-drift equilibrium faster than expected for a haploid genome under drift.

Figure 7: Summary of signatures of selection (Selective sweeps) and population differentiation (F_{ST} & D_{XY}) shown on the phylogeny of *Drosophila innubila* populations. Signals seen as divergence between all populations (F_{ST}) or recurrent/long-term evolution (McDonald-Kreitman, MK) is shown in black and grey respectively on the left-hand side. Signals seen on each branch are colored by the branches that the signals are seen on, shown on the right-hand side.

Most Sky Island species are highly diverged between locations, suggesting that *D. innubila* may be

much more recently established than other species (SMITH AND FARRELL 2005; FAVE *et al.* 2015; WIENS

et al. 2019). Similar patterns of limited nuclear divergence between species are observed in a Sky Island

bird species, *S. carolinensis* (MANTHEY AND MOYLE 2015), however this flying species can easily migrate

between Sky Islands (MANTHEY AND MOYLE 2015). In contrast, the patterns observed in *D. innubila*

populations are more likely ancestrally maintained and not due to gene flow (Figure 1 & 2, Supplementary

Figure 2). In *S. carolinensis,* differential selection is restricted to loci associated with recent changes in

environmental extremes (MANTHEY AND MOYLE 2015). This may also be the case in *D. innubila*, as we

find some evidence of recent selection, limited to genes associated with environmental factors, chorion

genes and antifungal immune genes. (Figure 7, Supplementary Figure 2). Most other insect and non-flying

species found across the Sky Islands are highly structured between locations (SMITH AND FARRELL 2005;

FAVE *et al.* 2015; WIENS *et al.* 2019). If the lack of geographic divergence is due to the maintenance of

ancestral variation, we expect *D. innubila* populations to become as structured as these other species over

time.

Segregating inversions are often associated with population structure (DOBZHANSKY AND

STURTEVANT 1937; FULLER *et al.* 2016; PUZEY *et al.* 2017) and could explain the excess interpopulation

divergence seen on Muller element B here (Supplementary Figures 2-5). Our detection of several putative

segregating inversions on Muller element B relative to all other chromosomes (Figure 3A) supports this

assertion. However, all are large and common inversions characterized in all populations, suggesting the

inversions are not driving the elevated F_{ST}. We suspect that the actual causal inversions may not have been

correctly characterized for two reasons. First there are limitations of detecting inversions in repetitive

regions with short read data (MARZO *et al.* 2008; CHAKRABORTY *et al.* 2017). Second, and confounding

the first, the likely existence of several overlapping segregating inversions complicates correctly calling

inversion breakpoints further. The elevated F_{ST} could also be caused by other factors, such as extensive

duplication and divergence on Muller element B being misanalysed as just divergence. In fact, the broken

and split read pairs used to detect inversions are very similar to the signal used to detect duplications (YE

et al. 2009; RAUSCH *et al.* 2012; CHEN *et al.* 2016), suggesting some misidentification may have occurred.

If a large proportion of Muller B was duplicated, we would see elevated mean coverage of Muller element

B in all strains compared to other autosomes, which is not the case (Supplementary Table 1). There may

also be seasonal variation in inversions which causes the elevated Tajima's D and F_{ST} seen here

(Supplementary Figures 3-5) and may be the cause of the difference in minor allele frequency between

2001 and 2017 on Muller element B (Supplementary Figure 10B). In fact, in several species, inversions can

change in frequency as the multiple linked variants they contain change in fitness over time (RODRIGUEZ-

TRELLES *et al.* 1996; ONEAL *et al.* 2014; KAPUN *et al.* 2016; PUZEY *et al.* 2017).

While we find no differences in inversion frequencies between time points (2001 and 2017) in *D.*
innubila (Wilcoxon Rank Sum Test W = 561, *p*-value = 0.352), we again may not have correctly identified
the inversions and so cannot properly infer frequency changes. Segregating inversions, such as those on
Muller element B, are frequently associated with local or temporal variation (DOBZHANSKY AND
STURTEVANT 1937; RODRIGUEZ-TRELLES *et al.* 1996; ONEAL *et al.* 2014), and consistent with this find
differences in allele frequencies between time points (Supplementary Figure 10B). There was an extensive
forest fire in the Sky Islands in 2011 which could plausibly have been a strong selective force driving a
change in allele frequency on Muller B between time points (Supplementary Figure 10B) (ARECHEDERRA-
ROMERO 2012). It is also worth noting that *D. pseudoobscura* segregates for inversions on Muller element
C and these segregate by population in the same Sky island populations (and beyond) as the populations
described here (DOBZHANSKY AND STURTEVANT 1937; DOBZHANSKY *et al.* 1963; FULLER *et al.* 2016).
The inversion polymorphism among populations is a plausible area for local adaptation and may provide
an interesting contrast to the well-studied *D. pseudoobscura* inversions. (HERMISSON AND PENNINGS 2005;
MCVEAN 2007; PRZEWORSKI *et al.* 2011).

Ours is one of few studies that sequences whole genomes from individual wild-caught *Drosophila*
and therefore avoids several generations of inbreeding that would purge recessive deleterious alleles
(GILLESPIE 2004; MACKAY *et al.* 2012; POOL *et al.* 2012). During inbreeding (or due to unintentional
inbreeding over generations as lines are maintained) a large proportion of variation found in the population
will be removed by background selection (CHARLESWORTH *et al.* 1993). This limits the scope of studies,
and the ability to infer the history of the population. Additionally, when attempting to calculate the extent
of adaptation occurring in populations, inbred panels will have removed a large proportion of segregating
(likely deleterious) variation, which may lead to misestimations of McDonald-Kreitman based statistics,
which normalize divergence to polymorphism (MCDONALD AND KREITMAN 1991A; EILERTSON *ET AL.*
2012; MESSER AND PETROV 2012). This could also explain the excessive rates of adaptation in *Drosophila*
melanogaster compared to other species (CAI *et al.* 2009; SELLA *et al.* 2009). The excess of putatively
deleterious alleles which may be present in our study harkens back to early studies of segregating lethal
mutations in populations as well as recent work on humans (DOBZHANSKY *et al.* 1963; MARINKOVIC 1967;
DOBZHANSKY AND SPASSKY 1968; WATANABE *et al.* 1974; GAO *et al.* 2015).

To date, most of the genomic work concerning the phylogeographic distribution and dispersal of
different *Drosophila* species has been limited to the *melanogaster* supergroup (POOL *et al.* 2012; POOL AND
LANGLEY 2013; BEHRMAN *et al.* 2015; LACK *et al.* 2015; MACHADO *et al.* 2015), with some work in other
Sophophora species (FULLER *et al.* 2016). This limits our understanding of how non-commensal species
disperse and behave, and what factors seem to drive population demography over time. Here we have
glimpsed into the dispersal and history of a species of mycophageous *Drosophila* and found evidence of

changes in population distributions potentially due to the changing climate (Figure 7) (ARIZONA-
GEOLOGICAL-SURVEY 2005) and population structure possibly driven by segregating inversions. Because
many species have recently undergone range changes or expansions (EXCOFFIER *et al.* 2009; PORRETTA *et*
al. 2012; WHITE *et al.* 2013), we believe examining how this has affected genomic variation is important
for population modelling and even for future conservation efforts (EXCOFFIER *et al.* 2009; COE *et al.* 2012).

Methods
Fly collection, DNA isolation and sequencing
We collected wild *Drosophila* at the four mountainous locations across Arizona between the 22nd of August
and the 11th of September 2017: the Southwest research station in the Chiricahua mountains (CH, ~5,400
feet elevation, 31.871 latitude -109.237 longitude, 96 flies), in Prescott National Forest (PR, ~7,900 feet
elevation, 34.586 latitude -112.559 longitude, 96 flies), Madera Canyon in the Santa Rita mountains (SR,
~4,900 feet elevation, 31.729 latitude -110.881 longitude, 96 flies) and Miller Peak in the Huachuca
mountains (HU, ~5,900 feet elevation, 31.632 latitude -110.340 longitude, 53 flies) (COE *et al.* 2012).
Drosophila innubila only emerge for a short period of time in the Arizona wet season (~2 months), limiting
dates of collection to this period of time each year (PATTERSON AND STONE 1949). Baits consisted of store-
bought white button mushrooms (*Agaricus bisporus*) placed in large piles about 30cm in diameter, with at
least 5 baits per location. We used a sweep net to collect flies over the baits in either the early morning or
late afternoon between one and three days after the bait was set. We sorted flies by sex and species at the
University of Arizona in Tucson, AZ and flash frozen at -80°C before shipping on dry ice to the University
of Kansas in Lawrence KS.
We sorted 343 flies (172 females and 171 males) which phenotypically matched *D. innubila*. We
then homogenized and extracted DNA using the Qiagen Gentra Puregene Tissue kit (USA Qiagen Inc.,
Germantown, MD, USA), isolating DNA for each fly individually. We also isolated the DNA of 40 *D.*
innubila samples individually, collected in early September in 2001 from CH. We prepared a genomic DNA
library of these 383 DNA samples using a modified version of the Nextera DNA library prep kit (~ 350bp
insert size) meant to conserve reagents, during this process we barcoded each sample to distinguish
individuals during sequencing. We sequenced the libraries on four lanes of an Illumina HiSeq 4000 (150bp
paired end) (Supplementary Table 1, Data deposited in the NCBI SRA: SRP187240).

Sample filtering, mapping and alignment
We removed adapter sequences using Scythe (BUFFALO 2018), trimmed all data using cutadapt to remove
barcodes (MARTIN 2011) and removed low quality sequences using Sickle (parameters: -t sanger -q 20 -l
50) (JOSHI AND FASS 2011). We masked the *D. innubila* reference genome, using *D. innubila* TE sequences
generated previously and RepeatMasker version 4.0 (parameters: -s -gccalc -gff -lib customLibrary) (SMIT

AND HUBLEY 2013-2015; HILL *et al.* 2019). We then mapped the short reads to the masked *D. innubila*

genome using BWA MEM Version 0.7.17 (LI AND DURBIN 2009), and sorted and indexed using SAMTools

Version 1.9 (LI *et al.* 2009). Following mapping, we added read groups, marked and removed sequencing

and optical duplicates, and realigned around indels in each mapped BAM file using Picard Version 2.23.4

and GATK Version 4.0.0.0 (HTTP://BROADINSTITUTE.GITHUB.IO/PICARD ; MCKENNA *et al.* 2010;

DEPRISTO *et al.* 2011). We then removed individuals with low coverage of the *D. innubila* genome (less

than 5x coverage for 80% of the non-repetitive genome), and individuals we suspected of being

misidentified as *D. innubila* following collection due to anomalous mapping. This left us with 280 *D.*

innubila wild flies (48 - 84 flies per populations) from 2017 and 38 wild flies from 2001 with at least 5x

coverage across at least 80% of the euchromatic genome (Supplementary Table 1).

Nucleotide polymorphisms across the population samples

For the 318 sequenced samples with reasonable coverage, we called SNPs using GATK HaplotypeCaller

Version bcf4.0.0.0 (MCKENNA *et al.* 2010; DEPRISTO *et al.* 2011) which generated a multiple strain VCF

file. We then used BCFtools Version 1.7 (NARASIMHAN *et al.* 2016) to remove sites with a GATK quality

score (a composite PHRED score for multiple samples per site) lower than 950 and sites absent (e.g. sites

of low quality, or with 0 coverage) from over 5% of individuals. This filtering left us with 4,522,699 SNPs

and small indels across the 168Mbp genome of *D. innubila*. We then removed SNPs found as a singleton

in a single population (as possible errors), leaving us with 3,240,198 SNPs. We used the annotation of *D.*

innubila and SNPeff Version 5.0 (CINGOLANI *et al.* 2012) to identify SNPs as synonymous, non-

synonymous, non-coding or another annotation. Simultaneous to the *D. innubila* population samples, we

also mapped genomic information from outgroup species *D. falleni* (SRA: SRR8651761) and *D. phalerata*

(SRA: SRR8651760) to the *D. innubila* genome and called divergence using the GATK variation calling

pipeline to identify derived polymorphisms and fixed differences in *D. innubila*.

Population genetic summary statistics and structure

Using the generated total VCF file with SNPeff annotations, we created a second VCF containing only

synonymous polymorphism using BCFtools Version 1.7 (NARASIMHAN *et al.* 2016). We calculated

pairwise diversity per base, Watterson's theta, Tajima's D (TAJIMA 1989) and F_{ST} (WEIR AND COCKERHAM

1984) (versus all other populations) across the genome for each gene in each population using VCFtools

Version 0.1.13 (DANECEK *et al.* 2011) and the VCF containing all variants. Using ANGSD Version 0.9.11

to parse the synonymous polymorphism VCF (KORNELIUSSEN *et al.* 2014), we generated synonymous

unfolded site frequency spectra for the *D. innubila* autosomes for each population, using the *D. falleni* and

D. phalerata genomes as outgroups to the *D. innubila* genome (HILL *et al.* 2019).

We used the population silent SFS with previously estimated mutation rates of *Drosophila*
(SCHRIDER et al. 2013), as inputs in StairwayPlot Version 2 (LIU AND FU 2015), to estimate the effective
population size backwards in time for each location.

We also estimated the extent of population structure across samples using Structure Version 2.3.4
(FALUSH *et al.* 2003), repeating the population assignment for each chromosome separately using only
silent polymorphism, for between one and ten populations (k = 1-10, 100000 iterations burn-in, 400000
iterations sampling). Following (FRICHOT *et al.* 2014), we manually assessed which number of
subpopulations best fits the data for each *D. innubila* chromosome and DiNV to minimize entropy.

As our populations have recently established, we used δaδi Version 1.6.7 to determine which
model of population dynamics best fits for each *D. innubila* chromosome (GUTENKUNST *et al.* 2010).For
pairs of populations, we compared models to determine if populations are behaving as one population
having gone through a bottleneck (*bottlegrowth* model), as split populations with migration
(*bottlegrowth_split_mig* model) or recently split populations with no migration (*bottlegrowth_split* model).
We generated a site frequency spectrum for each chromosome for each population from our VCF of silent
polymorphism in δaδi (though using the total polymorphism for the mitochondria), using the
Spectrum.from_data_dict function. For each pair of populations and each chromosome, we estimated the
optimal parameters for each model using the *Inference.optimize_log* function. We then fit each model
(using the *Inference.ll_multinom* function) and compared the fit of each model using a likelihood-ratio test,
to determine what demographic model best fits each Muller element and the mitochondria.

Signatures of local adaptive divergence across D. innubila populations
We downloaded gene ontology groups from Flybase (GRAMATES *et al.* 2017). We then used a gene
enrichment analysis to identify enrichments for particular gene categories among genes in the upper and
lower 2.5th percentile for F_{ST}, Tajima's D and Pairwise Diversity versus all other genes (SUBRAMANIAN *et*
al. 2005). Due to differences on the chromosomes Muller A and B versus other chromosomes in some
cases, we also repeated this analysis chromosome by chromosome, taking the upper 97.5th percentile of
each chromosome.

We next attempted to look for selective sweeps in each population using Sweepfinder2 (HUBER *et*
al. 2016). We reformatted the polarized VCF file to a folded allele frequency file, showing allele counts for
each base. We then used Sweepfinder2 Version 1.0 on the total called polymorphism in each population to
detect selective sweeps in 1kbp windows (HUBER *et al.* 2016). We reformatted the results and looked for
genes neighboring or overlapping with regions where selective sweeps have occurred with a high
confidence, shown as peaks above the genomic background. We surveyed for peaks by identifying 1kbp
windows in the 97.5th percentile and 99th percentile for composite likelihood ratio per chromosome.

Using the total VCF with outgroup information, we next calculated D_{XY} per SNP for all pairwise

population comparisons (NEI AND MILLER 1990), as well as within population pairwise diversity and d_S

from the outgroups, using a custom python script. We then found the average D_{XY} and d_S per gene and

looked for gene enrichments in the upper 97.5th percentile, versus all other genes.

Inversions

For each sample, we used Delly Version 2.0 (RAUSCH *et al.* 2012) to generate a multiple sample VCF file

identifying regions in the genome which are potentially duplicated, deleted or inverted compared to the

reference genome. Then we filtered and removed inversions found in fewer than 1% of individuals and with

a GATK VCF quality score lower than 200. We also called inversions using Pindel Version 0.2.5 (YE *et*

al. 2009) in these same samples and again removed low quality inversion calls. We next manually filtered

samples and merged inversions with breakpoints within 1000bp at both ends and significantly overlapping

in the presence/absence of these inversions across strains (using a χ^2 test, *p*-value < 0.05). We also filtered

and removed large inversions which were only found with one of the two tools. Using the remaining filtered

and merged inversions we estimated the frequency of each inversion within the total population.

Signatures of recurrent selection

We filtered the total VCF with annotations by SNPeff and retained only non-synonymous (replacement) or

synonymous (silent) SNPs (CINGOLANI *et al.* 2012). We then compared these polymorphisms to the

differences identified to *D. falleni* and *D. phalerata* to polarize changes to specific branches. Specifically,

we sought to determine sites which are polymorphic in our *D. innubila* populations or are substitutions

which fixed along the *D. innubila* branch of the phylogeny. We used the counts of fixed and polymorphic

silent and replacement sites per gene to estimate McDonald-Kreitman-based statistics, specifically direction

of selection (DoS) (MCDONALD AND KREITMAN 1991b; SMITH AND EYRE-WALKER 2002; STOLETZKI AND

EYRE-WALKER 2011). We also used these values in SnIPRE (EILERTSON *et al.* 2012), which reframes

McDonald-Kreitman based statistics as a linear model, taking into account the total number of non-

synonymous and synonymous mutations occurring in user defined categories to predict the expected

number of these substitutions and calculate a selection effect relative to the observed and expected number

of mutations (EILERTSON *et al.* 2012). We calculated the SnIPRE selection effect for each gene using the

total number of mutations on the chromosome of the focal gene. Using FlyBase gene ontologies

(GRAMATES *et al.* 2017), we sorted each gene into a category of immune gene or classed it as a background

gene, allowing a gene to be classed in multiple immune categories. We fit a GLM to identify functional

categories with excessively high estimates of adaptation, considering multiple covariates:

656   *Statistic $\sim$ Population + Gene group + (Gene group $*$ Population) + Chromosome*

657   *+ Chromosome: Position*

We then calculated the difference in each statistic between our focal immune genes and a randomly sampled

nearby (within 100kbp) background gene, finding the average of these differences for each immune

category over 10000 replicates, based on (CHAPMAN *et al.* 2019).

To confirm these results, we also used AsymptoticMK (HALLER AND MESSER 2017) to calculate

asymptotic α for each gene category. We generated the non-synonymous and synonymous site frequency

spectrum for each gene category, which we then used in AsymptoticMK to calculate asymptotic α and a

95% confidence interval. We then used a permutation test to assess if functional categories of interest

showed a significant difference in asymptotic α from the rest of categories.

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Data Availability

All sequencing information is available on the NCBI SRA (SRA: SRP187240), while all genomes used in

this study are available on the NCBI genome database (SRA: GCA_004354385.1, GCA_005876895.1).

All supplementary data used in this manuscript is available at:

https://figshare.com/projects/innubila_population_genomics/87662

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980

Divergent Selection (F$_{ST}$) in:
Antifungal genes, mt translocations
Muller element B (inversion associated)

**Long-term
selection (MK) in:**
Toll-signalling, AMP

HU

Divergence in Cuticle
Sweeps in Chorion genes

SR

PR

Global elevated F$_{ST}$
Global decreased diversity
Global elevated Tajima's D
Divergence in Cuticle
Sweeps in Chorion genes

CH

Divergence in TE content
Sweeps in Chorion genes

D. falleni