## Selection and demography shape genomic variation in a 'Sky Island' species

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## 1 Abstract

2 Over time, populations of species can expand, contract, and become isolated, creating subpopulations that 3 can adapt to local conditions. Understanding how species adapt following these changes is of great interest, 4 especially as the current climate crisis has caused range shifts for many species. Here, we characterize how 5 Drosophila innubila came to inhabit and adapt to its current range: mountain forests in southwestern USA 6 separated by large expanses of desert. Using population genomic data from more than 300 wild-caught 7 individuals, we examine four distinct populations to determine their population history in these mountain-8 forests, looking for signatures of local adaptation to establish a genomic model for this spatially-distributed 9 system with a well understood ecology. We find D. innubila spread northwards during the previous 10 glaciation period (30-100 KYA), and has recently expanded even further (0.2-2 KYA). Surprisingly, D. 11 innubila shows little evidence of population structure, though consistent with a recent migration, we find 12 signatures of a population contraction following this migration, and signatures of recent local adaptation 13 and selective sweeps in cuticle development and antifungal immunity. However, we find little support for 14 recurrent selection in these genes suggesting recent local adaptation. In contrast, we find evidence of 15 recurrent positive selection in the Toll-signaling system and the Toll-regulated antimicrobial peptides.

#### 16 Introduction

17 In the past 25,000 years, the earth has undergone substantial environmental changes due to both human-18 mediated events (anthropogenic environment destruction, desert expansion, extreme weather and the 19 growing anthropogenic climate crisis) (CLOUDSLEY-THOMPSON 1978; ROSENZWEIG et al. 2008) and 20 events unrelated to humans (glaciation and tectonic shifts) (HEWITT 2000; HOLMGREN et al. 2003; SURVEY 21 2005). These environmental shifts can fundamentally reorganize habitats, influence organism fitness, rates 22 of migration between locations, and population ranges (SMITH et al. 1995; ASTANEI et al. 2005; 23 ROSENZWEIG et al. 2008; SEARLE et al. 2009; CINI et al. 2012; PORRETTA et al. 2012; ANTUNES et al. 24 2015). Signatures of the way organisms adapt to these events are often left in patterns of molecular variation 25 within and between species (CHARLESWORTH et al. 2003; WRIGHT et al. 2003; EXCOFFIER et al. 2009).

26 When a population migrates to a new location it first goes through a population bottleneck (as only 27 a small proportion of the population will establish in the new location) (CHARLESWORTH et al. 2003; 28 EXCOFFIER et al. 2009; LI AND DURBIN 2011). These bottlenecks result in the loss of rare alleles in the 29 population (TAJIMA 1989; GILLESPIE 2004). After the bottleneck, the population will grow to fill the 30 carrying capacity of the new niche and adapt to the unique challenges in the new environment, both signaled 31 by an excess of rare alleles (EXCOFFIER et al. 2009; WHITE et al. 2013). This adaptation can involve 32 selective sweeps from new mutations or standing genetic variation, and signatures of adaptive evolution 33 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).

38 Signatures of demographic change are frequently detected in species that have recently undergone 39 range expansion due to human introduction (ASTANEI et al. 2005; EXCOFFIER et al. 2009) or the changing 40 climate (HEWITT 2000; PARMESAN AND YOHE 2003; GUINDON et al. 2010; WALSH et al. 2011; CINI et al. 41 2012). Other hallmarks of invasive species population genomics include signatures of bottlenecks visible 42 in the site frequency spectrum, and differentiation between populations (CHARLESWORTH et al. 2003; LI 43 AND DURBIN 2011). This can be detected by a deficit of rare variants, a decrease in population pairwise 44 diversity and an increase in the statistic, Tajima's D (TAJIMA 1989). Following the establishment and 45 expansion of a population, there is an excess of rare variants and local adaptation results in divergence 46 between the invading population and the original population. These signatures are also frequently utilized 47 in human populations to identify traits which have fixed upon the establishment of a humans in a new 48 location, or to identify how our human ancestors spread globally (LI AND DURBIN 2011).

49 The Madrean archipelago, located in southwestern USA and northwestern Mexico, contains 50 numerous forested mountains known as 'Sky islands', separated by large expanses of desert (MCCORMACK 51 et al. 2009; COE et al. 2012). These 'islands' were connected by lush forests during the previous glacial 52 maximum which then retreated, leaving forest habitat separated by hundreds of miles of desert, presumably 53 limiting migration between locations for most species (SURVEY 2005; MCCORMACK et al. 2009). The 54 islands are hotbeds of ecological diversity. However, due to the changing climate in the past 100 years, they 55 have become more arid, which may drive migration and adaptation (MCCORMACK et al. 2009; COE et al. 56 2012).

57 Drosophila innubila is a mycophageous Drosophila species found throughout these Sky islands 58 and thought to have arrived during the last glacial maximum (DYER AND JAENIKE 2005; DYER *et al.* 2005). 59 Unlike the lab model *D. melanogaster*, *D. innubila* has a well-studied ecology (LACHAISE AND SILVAIN 60 2004; DYER AND JAENIKE 2005; DYER *et al.* 2005; JAENIKE AND DYER 2008; UNCKLESS 2011a; UNCKLESS 61 AND JAENIKE 2011; COE *et al.* 2012). In fact, in many ways the 'island' endemic, mushroom-feeding 62 ecological model *D. innubila* represent a counterpoint to the human commensal, cosmopolitan, genetic 63 workhorse *D. melanogaster*.

We sought to reconstruct the demographic and migratory history of *D. innubila* inhabiting the Sky islands. Isolated populations with limited migration provide a rare opportunity to observe replicate bouts of evolutionary change and this is particularly interesting regarding the coevolution with pathogens (DYER AND JAENIKE 2005; UNCKLESS 2011a). We also wanted to understand how *D. innubila* adapt to their local

68 climate and if this adaptation is recurrent or recent and specific to local population. We resequenced whole 69 genomes of wild-caught individuals from four populations of D. innubila in four different Sky island 70 mountain ranges. Surprisingly, we find little evidence of population structure by location, with structure 71 limited to the mitochondria and a single chromosome (Muller element B which is syntenic with 2L in D. 72 melanogaster) (MARKOW AND O'GRADY 2006). However, we find some signatures of local adaptation, 73 such as for cuticle development and fungal pathogen resistance, suggesting potentially a difference in fungal 74 pathogens and toxins between locations. We also find evidence of mitochondrial translocations into the 75 nuclear genome, with strong evidence of local adaptation of these translocations, suggesting potential 76 adaptation to changes in metabolic process of the host between location, and possibly even as a means of 77 compensating for reduced efficacy of selection due to Wolbachia infection (JAENIKE AND DYER 2008).

#### 78 **Results**

Drosophila innubila has recently expanded its geographic range and shows high levels of gene flow
 between geographically isolated populations

81 To characterize how *D. innubila* came to inhabit its current range, we collected flies from four Sky island 82 locations across Arizona: Chiricahuas (CH, 81 flies), Huachucas (HU, 48 flies), Prescott (PR, 84 flies) and 83 Santa Ritas (SR, 67 flies) (Figure 1). Interestingly, previous surveys mostly failed to collect D. innubila 84 north of the Madrean archipelago in Prescott (DYER AND JAENIKE 2005). We easily sampled from that 85 location, suggesting a possible recent invasion (though we were also unable to collect D. innubila in the 86 exact locations previously sampled) (DYER AND JAENIKE 2005). If this was a recent colonization event, it 87 could be associated with the changing climate of the area leading to conditions more accommodating to D. 88 innubila, despite over 300 kilometers of geographic isolation (Figure 1).

89 To determine when D. innubila established each population and rates of migration between 90 locations, we isolated and sequenced the DNA from our sampled D. innubila populations and characterized 91 genomic variation. We then examined the population structure and changes in demographic history of D. 92 innubila using silent polymorphism in Structure (FALUSH et al. 2003) and StairwayPlot (LIU AND FU 2015). 93 We find all sampled populations have a current estimated effective population size (N<sub>e</sub>) of  $\sim$ 1,000,000 94 individuals and an ancestral N<sub>c</sub> of  $\sim$ 4.000,000 individuals, though all experience a bottleneck about 100,000 95 years ago to an Ne of 10,000-20,000 (Figure 1, Supplementary Figure 1A & B). This bottleneck coincides 96 with a known glaciation period occurring in Arizona (SURVEY 2005). Each surveyed population then 97 appears to go through separate population expansions between one and thirty thousand years ago, with 98 populations settling from south to north (Figure 1A, Supplementary Figure 1A & B). Specifically, while 99 the HU population appears to have settled first (10-30 thousand years ago), the PR population was settled 100 much more recently (200-2000 years ago). This, in combination of the absence of D. innubila in PR until

~2016 sampling suggests recent northern expansion of *D. innubila* (Figure 1). Also note that StairwayPlot
 (LIU AND FU 2015) has estimated large error windows for PR, meaning the invasion could be more recent
 or ancient than the 200-2000 year estimate.

Given the geographic isolation between populations, we expected to find a corresponding signature of population differentiation among the populations. Using Structure (FALUSH *et al.* 2003), we find surprisingly little population differentiation between locations for the nuclear genome (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 still consistent gene flow between populations potentially via males.

110 **Figure 1:** A. Schematic of the range expansion of *D. innubila* and DiNV based on StairwayPlot results

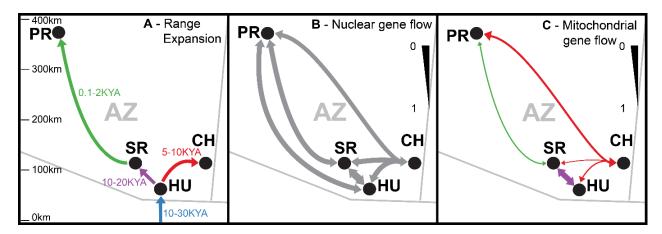
111 across the four sample locations in Arizona (AZ), Chiracahua's (CH), Huachucas (HU), Prescott (PR) and

112 Santa Ritas (SR). **B** and **C**. Summary of Structure/Fst results for **B**. autosomal polymorphism and **C**.

113 mitochondrial polymorphism. Thickness of arrows in **B** and **C** indicates the median of F<sub>ST</sub> for genes in

each category, with 1 indicating completely isolated populations and 0 indicating complete gene flow.

115 Light grey lines show the Arizona border.

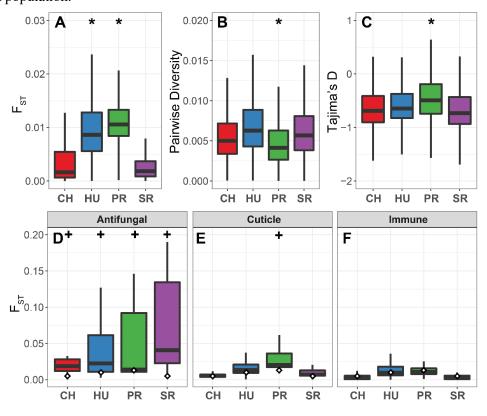


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117 We also calculated the amount of differentiation between each population and all other populations, 118 as the fixation index, F<sub>ST</sub>, using the total polymorphism across the Muller elements (Drosophila 119 chromosomes) (WEIR AND COCKERHAM 1984). F<sub>ST</sub> appears to be generally low across the genes in the D. 120 *innubila* genome (Figure 1B, total median = 0.00567), consistent with nuclear gene flow between 121 populations (WEIR AND COCKERHAM 1984). In contrast, there is higher  $F_{ST}$  between mitochondrial 122 genomes (Figure 1C). Both nuclear and mitochondrial results are consistent with the Structure/StairwayPlot 123 results. However, consistent with a more recent population contraction upon migration into PR, F<sub>ST</sub> of the 124 nuclear genome is significantly higher in PR (Figure 2A, GLM t-value = 93.728, p-value = 2.73e-102). 125 Though PR  $F_{ST}$  is still extremely low genome-wide (PR median = 0.0105), with some outliers on Muller

126 element B like other populations (Supplementary Figure 2). We also calculated the population genetic 127 statistics pairwise diversity and Tajima's D for each gene using total polymorphism (TAJIMA 1989). As 128 expected with a recent population contraction in PR (suggesting recent migration and establishment in a 129 new location), pairwise diversity is significantly lower (Figure 2B, GLM t-value = -19.728, p-value = 2.33e-130 86, Supplementary Table 2) and Tajima's D is significantly higher than all other populations (Figure 2C, 131 GLM t-value = 4.39, p-value = 1.15e-05, Supplementary Table 2). This suggests that there is also a deficit 132 of polymorphism in general in PR, consistent with a more recent population bottleneck, removing rare 133 alleles from the population (Figure 2C, Supplementary Figure 3). Conversely, the other populations show 134 a genome wide negative Tajima's D, consistent with a recent demographic expansion (Supplementary 135 Figure 3).

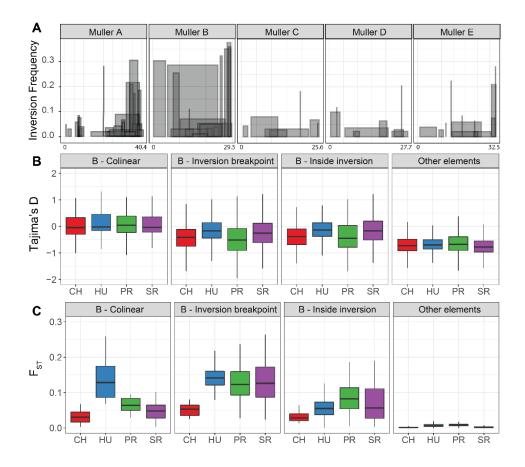
136 Figure 2: Summary statistics for each population. A. Distribution of  $F_{ST}$  across genes for each population 137 versus all other populations. **B.** Distribution of pairwise diversity for each population. **C.** Distribution of 138 Tajima's D for each population. **D.** F<sub>ST</sub> distribution for Antifungal associated genes for each population. **E.** 139  $F_{ST}$  distribution for cuticular proteins for each population. F.  $F_{ST}$  distribution for all immune genes 140 (excluding antifungal genes). In A, B & C all cases significant differences from CH are marked with an \* 141 and outliers are removed for ease of visualization. In D, E & F, significant differences from the genome 142 background in each population are marked with a + and white diamond mark the whole genome average of 143 F<sub>ST</sub> for each population.



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

146 As mentioned previously, F<sub>ST</sub> is significantly higher on Muller element B compared to all other elements 147 in all populations (Supplementary Figure 2, GLM t-value = 30.02, p-value = 3.567e-56). On Muller element 148 B, regions of elevated  $F_{ST}$  are similar in each population (Supplementary Figure 2). Additionally, Muller 149 element B has elevated Tajima's D compared to all other Muller elements (Supplementary Figure 3), 150 suggesting some form of structured population unique to Muller element B (GLM t-value = 10.402, p-value 151 = 2.579e-25). We attempted to identify if this elevated structure is due to chromosomal inversions, 152 comparing  $F_{ST}$  of a region to the presence or absence of inversions across windows (using only inversions 153 called by both Delly and Pindel (YE et al. 2009; RAUSCH et al. 2012)). We find several inversions across the genome at appreciable frequencies (89 total above 1% frequency), of which, 37 are found on Muller 154 155 element B (spread evenly across the entire chromosome) and 22 are found at the telomeric end of Muller 156 element A (Figure 3A). The presence of an inversion over a region of Muller element B is associated with 157 higher  $F_{ST}$  in these regions (Figure 3A, Wilcoxon Rank Sum test W = 740510, *p*-value = 0.0129), though 158 these inversions are not unique or even at different frequencies in specific populations ( $F_{ST} < 0.22$ ,  $\chi^2$  test 159 for enrichment in a specific population p-value > 0.361 for all inversions). Genes within 10kbp of an 160 inversion breakpoint have significantly higher F<sub>ST</sub> than outside the inverted regions consistent with findings 161 in other species (Figure 3C, GLM t-value = 7.702, p-value = 1.36e-14) (MACHADO et al. 2007; NOOR et 162 al. 2007), while inside inverted regions show no difference from outside (Figure 3C, GLM t-value = -0.178, 163 p-value = 0.859). However, all regions of Muller element B have higher F<sub>ST</sub> than the other Muller elements 164 (Figure 3C, outside inversions Muller element B vs all other chromosomes: GLM t-value = 7.379, p-value 165 = 1.614e-13), suggesting some chromosome-wide force drives the higher  $F_{ST}$  and Tajima's D. Given that 166 calls for large inversions in short read data are often not well supported (CHAKRABORTY et al. 2017) and 167 the apparently complex nature of the Muller element B inversions (Figure 3A), we may not have correctly 168 identified the actual inversions and breakpoints on the chromosome. Despite this, our results do suggest a 169 link between the presence of inversions on Muller element B and elevated differentiation in D. innubila and 170 that this may be associated with local adaptation.

171Figure 3: Summary of the inversions detected in the *Drosophila innubila* populations. A. Location and172frequency in the total population of segregating inversions at higher than 1% frequency and greater than173100kbp. B. Tajima's D and C.  $F_{ST}$  for genes across Muller element B, grouped by their presence under an174inversion, outside of an inversion, near the inversion breakpoints (within 10kbp) or on a different Muller175element.



#### 176

## 177 Evidence for local adaptation in each population

178 Though F<sub>ST</sub> is low across most of the genome in each population, there are several genomic regions with 179 elevated  $F_{ST}$ . In addition to the entirety of Muller element B, there are narrow chimneys of high  $F_{ST}$  on 180 Muller elements D and E (Figure 4, Supplementary Figure 2). We attempted to identify whether any gene 181 ontology groups have significantly higher F<sub>ST</sub> than the rest of the genome. We find that the genes in the 182 upper 2.5<sup>th</sup> percentile for F<sub>ST</sub> are enriched for antifungal genes in all populations, these genes are distributed 183 across the genome and so not all under one peak of elevated F<sub>ST</sub> (Supplementary Table 3, GO enrichment 184 = 16.414, p-value = 1.61e-10). Interestingly, this is the only immune category with elevated  $F_{ST}$  (Figure 185 2F), with most of the immune system showing no divergence between populations (Figure 2F, 186 Supplementary Figure 4). This might suggest that most pathogen pressures are consistent among 187 populations except for fungal pathogen pressure which may be more variable.

Another gene ontology category with significantly higher  $F_{ST}$  is cuticle genes (Figure 2E, Supplementary Table 3, 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 of exclusively cuticle development proteins (e.g. Cpr65Au, Cpr65Av, Lcp65Ad) with elevated  $F_{ST}$  in these

193 genes in both the SR and HU populations as well as PR (Figure 5), suggesting that they may be adapting to 194 differing local conditions in those populations.

195 Two other clear peaks on Muller element E are also composed related genes. Interestingly, there 196 appear to be three regions of the D. innubila genome with translocated mitochondrial genes (Figure 5). The 197 first peak (Muller element E, 11.35-11.4Mb) is composed exclusively of one of these translocated 198 mitochondrial regions with 3 mitochondrial genes (including cytochrome oxidase II). The second peak 199 (Muller element E, 23.60-23.62Mb) contains four other mitochondrial genes (including cytochrome oxidase 200 III and ND5) as well as genes associated with nervous system activity (such as *Obp93a* and *Obp99c*). We 201 find no correlation between coverage of these regions and mitochondrial copy number (Supplementary 202 Table 1, Pearsons' correlation t-value = 0.065, p-value = 0.861), so this elevated F<sub>ST</sub> is probably not an 203 artefact of mis-mapping reads. However, we do find these regions have elevated copy number compared to 204 the rest of the genome (Supplementary Figure 5, GLM t-value = 9.245, p-value = 3.081e-20), and so this 205 elevated divergence may be due to collapsed paralogs. These insertions of mtDNA are also found in D. 206 falleni and are diverged from the mitochondrial genome, suggesting ancient transpositions. The nuclear insertions of mitochondrial genes are also enriched in the 97.5th percentile for F<sub>ST</sub> in HU and PR, when 207 208 looking at only autosomal genes (Supplementary Table 3, GO enrichment = 4.53, p-value = 3.67e-04). 209 Additionally, several other energy metabolism categories are in the upper 97.5<sup>th</sup> percentile in CH. Overall 210 these results suggests a potential divergence in the metabolic needs of each population, and that several 211 mitochondrial genes may have found a new function in the D. innubila genome and may be diverging due 212 to differences in local conditions. Alternatively, given the male-killing Wolbachia parasitizing D. innubila 213 (DYER 2004), it is possible the mitochondrial translocations contain functional copies of mitochondrial 214 genes that can efficiently respond to selection unlike their still mtDNA-linked paralogs.

215 There has been considerable discussion over the last several years about the influence of 216 demographic processes and background selection on inference of local adaptation (CUTTER AND PAYSEUR 217 2013; CRUICKSHANK AND HAHN 2014; HOBAN et al. 2016; MATTHEY-DORET AND WHITLOCK 2018). In 218 contrast to F<sub>ST</sub> which is a relative measure of population differentiation, D<sub>XY</sub> is an absolute measure that 219 may be less sensitive to other population-level processes (NEI 1987; CRUICKSHANK AND HAHN 2014). In 220 our data, windows with peaks of elevated  $F_{ST}$  also have peaks of  $D_{XY}$  in all pairwise comparisons (Figure 4, Supplementary Figure 6), and  $F_{ST}$  and  $D_{XY}$  are significantly correlated (GLM  $R^2 = 0.823$ , t-value = 221 222 11.371, p-value = 6.33e-30), consistent with local adaptation. The upper 97.5<sup>th</sup> percentile for  $D_{XY}$  is 223 enriched for chorion proteins in all pairwise comparisons and antifungal proteins for all comparisons 224 involving PR (Supplementary Table 5). We also find peaks of elevated pairwise diversity exclusively on 225 the mitochondrial translocations (Supplementary Figure 6), suggesting unaccounted for variation in these 226 genes which is consistent with duplications detected in these genes (RASTOGI AND LIBERLES 2005). This

supports the possibility that unaccounted for duplications may be causing the elevated  $F_{ST}$ ,  $D_{XY}$  and pairwise diversity (Supplementary Figure 5 & 6). 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 (Figure 4, Supplementary Figure 5).

231 Recent adaptation often leaves a signature of a selective sweep with reduced polymorphism near 232 the site of the selected variant. We attempted to identify selective sweeps in each population using 233 Sweepfinder2 (HUBER et al. 2016). There was no evidence of selective sweeps overlapping with genes with 234 elevated F<sub>ST</sub> (Supplementary Figure 7A,  $\chi^2$  test for overlap of 97.5<sup>th</sup> percentile windows  $\chi^2 = 1.33 p$ -value 235 = 0.249) but there was one extreme peak in PR on Muller D (Supplementary Figure 7B, Muller D, 18.75-236 19Mb). This peak was also found in all other populations though not on the same scale. The center of this 237 peak is just upstream of the cuticle protein Cpr66D, in keeping with the suggestion of local adaptation of 238 the cuticle in all populations, with the strongest signal in PR. This sweep is also upstream of four chorion 239 proteins (Cp15, Cp16, Cp18, Cp19) and covers (within 10kbp of the sweep center) several cell organization 240 proteins (Zasp66, Pex7, hairy, Prm, Fhos). These chorion proteins also have significantly elevated D<sub>XY</sub> 241 compared to other genes within 50kbp (Wilcoxon Rank Sum W = 45637000, p-value = 0.0158) and are 242 under a chimney of elevated D<sub>XY</sub> in all comparisons (Supplementary Figure 6 & 7), consistent with recent 243 selection of population specific variants. We also find evidence of several selective sweeps in the telomere of the X chromosome (Muller A, 39.5-40.5Mb), among several uncharacterized genes. Given the 244 245 suppression of recombination in the heterochromatic portions of chromosomes, we would expect evidence 246 for several selective sweeps for even weakly positively selected variants, as is also seen in the non-247 recombining Muller F (Supplementary Figure 7).

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

249 population. Values are as follows: the average pairwise divergence per gene  $(D_{XY})$ , the population fixation

250 index per genes (F<sub>ST</sub>), within population pairwise diversity per genes, Compositive Likelihood Ratio (CLR)

251 per SNP calculated using Sweepfinder2 and within population average Tajima's D per gene.

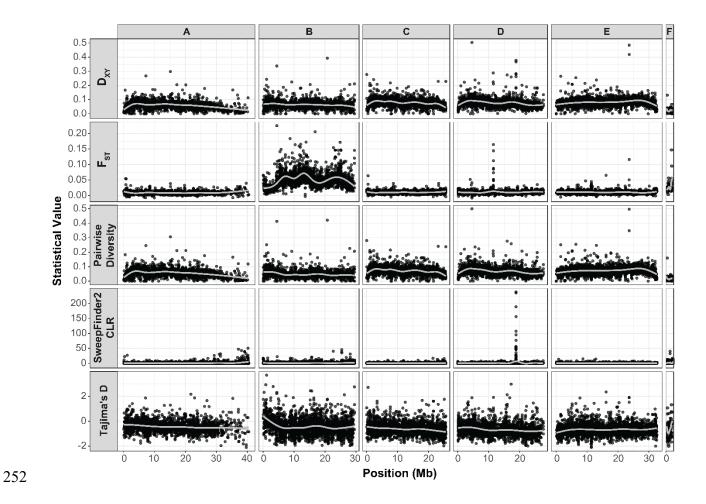
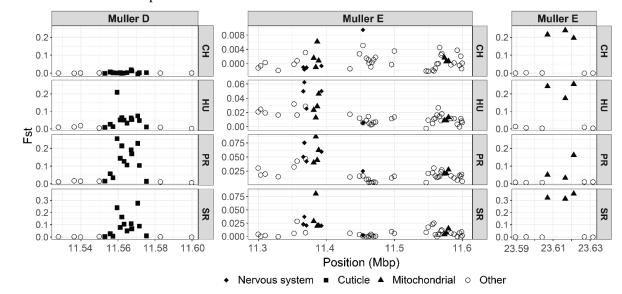


Figure 5: Gene-wise  $F_{ST}$  showing regions of elevated divergence between populations for each population. Plot shows  $F_{ST}$  for each gene in these regions to identify the causal genes. Genes with noted functions (cuticle development or mitochondrial translocations) are shown by point shape. Note the Y-axes are on different scales for each plot.



## 258 Evidence for divergence in the X chromosome over time and between sexes

259 We next compared the samples from the 35 CH males in 2017 to those we sequenced from a CH collection 260 of 38 males in 2001 to identify changes over time between populations (due to elevated F<sub>ST</sub> because of 261 differences in allele frequencies between populations). We find little differentiation between the two 262 timepoints (median  $F_{ST} = 0.0004$ , 99<sup>th</sup> percentile = 0.0143), and find no significant enrichments (GO *p*-263 value < 0.05) in the upper 97.5<sup>th</sup> percentile of F<sub>ST</sub>. However, we do find divergence in the genes at the 264 telomere of the X chromosome (Supplementary Figure 8A, Muller element A, 35-40.5Mb, median  $F_{ST}$  = 265 0.0029). Looking at actual allele frequency differences between time points, the minor allele frequency 266 increases between 2001 and 2017 at the X telomere while most other chromosomes appear to show little 267 change. Interestingly, there is also evidence of recent selective sweeps in the telomere of X (Supplementary 268 Figure 7A). The minor allele frequency has decreased on average on Muller element B between 2001 and 269 2017 (Supplementary Figure 9A). This suggests something else may be influencing allele frequency change 270 on Muller B compared to other autosomes.

271 We also compared the allele frequencies between 2017 male samples versus 2017 female samples. 272 Again,  $F_{ST}$  is extremely low genome wide (median  $F_{ST} = 0.0004$ , 99<sup>th</sup> percentile = 0.0501), but we again 273 find a peak of F<sub>ST</sub> at the telomere of the X chromosome (Supplementary Figure 8B), which we find when 274 comparing all populations sexes and in a total population male versus female comparison. Again, we find 275 no significant enrichments in the 97.5th percentile for F<sub>ST</sub>, as most of the divergent genes currently have no 276 functional annotation. We also compared the raw allele frequency change of synonymous variants. 277 Strangely, the population minor allele frequency of euchromatic SNPs on the X chromosome are found at 278 higher frequencies in females (Supplementary Figure 9B), while the X telomere SNPs are overrepresented 279 in male samples. These results are consistent when examining each population separately, suggesting sex 280 specific biases in the X chromosome are found in every populations. It is possible that this signal is caused 281 by an ascertainment bias for SNP calling in females, resulting in more accurate SNP calls in one of the 282 sexes in the euchromatin which is not seen in the heterochromatin. Alternatively, the region of the X 283 chromosome with multiple overlapping inversions could be female-biased due to a female driver, resulting 284 in its overrepresentation in females (and an overrepresentation of the alternate variants in males) (BURT 285 AND TRIVERS 2006). Finally, the X chromosome may be adapting to the skewed sex-ratio associated with 286 D. innubila's male-killing Wolbachia (KAGEYAMA et al. 2009; UNCKLESS 2011b).

*Toll-related immune genes are evolving recurrently in D. innubila likely due to strong pathogen pressures*We next sought to identify genes and functional categories showing strong signatures of adaptive evolution,
suggesting recurrent evolution as opposed to recent local adaptation. We reasoned that if the population
differentiation seen in antifungal genes and cuticle development proteins (Figure 2 & 3, Supplementary)

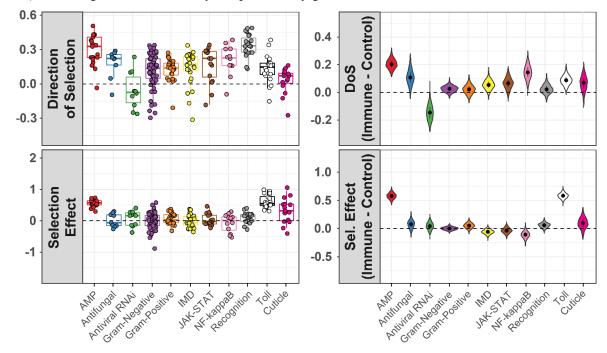
291 Figure 4) was due to local adaptation also acting over longer time periods, we would expect to see signatures 292 of adaptation in those categories. Furthermore, Hill et al. used dN/dS-based statistics to show that genes 293 involved in some immune defense pathways were among the fastest evolving genes in the D. innubila 294 genome (HILL et al. 2019). We also sought to identify what genes are evolving due to recurrent positive 295 selection in D. innubila in one or all populations, and if this is associated with environmental factors. To 296 this end we calculated the McDonald-Kreitman based statistic direction of selection (DoS) (Stoletzki and 297 Eyre-Walker, 2011) and SnIPRE selection effect (Eilertson et al., 2012) to identify an excess of selection. 298 We then fit a linear model to identify gene ontology groups with significantly higher DoS or selection effect 299 than expected. In this survey we find cuticle genes and antifungal genes did have some signatures of 300 adaptive evolution (DoS > 0 and selection effect > 0 for 80% of genes in these categories) but as a group 301 showed no significant differences from the background (GLM t-value = 1.128, p-value = 0.259, 302 Supplementary Table 4). In fact, we only found two functional groups significantly higher than the 303 background, Toll signaling proteins (GLM t-value = 2.581 p-value = 0.00986, Supplementary Table 3) and 304 antimicrobial peptides (AMPs, GLM t-value = 3.66 p-value = 0.00025, Supplementary Table 3). In a 305 previous survey we found that these categories were also the only functional groups to have significantly 306 elevated rates of amino acid divergence (HILL et al. 2019). These results suggest that this divergence is 307 indeed adaptive.

308 Interestingly, D. innubila is burdened by Drosophila innubila nudivirus (DiNV), a Nudivirus that 309 infects 40-50% of individuals in the wild (UNCKLESS 2011a). A close relative of the virus suppresses Toll-310 regulated AMPs in D. melanogaster (PALMER et al. 2018), which might explain why the Toll pathway and 311 AMPs are fast evolving in *D. innubila*. Five AMPs showed consistently positive DoS and selection effect 312 values (which are also among the highest in the genome): four Bomanins and Listericin. All are AMPs 313 regulated by Toll signaling (and additionally JAK-STAT in the case of *Listericin*) (HOFFMANN 2003; 314 TAKEDA AND AKIRA 2005). Listericin has been implicated in the response to viral infection due to its 315 expression upon viral infection (DOSTERT et al. 2005; ZAMBON et al. 2005; IMLER AND ELFTHERIANOS 316 2009; MERKLING AND VAN RIJ 2013). For all immune categories, as well as cuticle proteins and antifungal 317 proteins, we find no significant differences between populations for either MK-based statistics, and no 318 significant differences in the distribution of these statistics between populations (GLM t-value < 0.211, p-319 value > 0.34 for all populations, Supplementary Table 3). Thus, perhaps selection at these loci is ubiquitous 320 and genes flow between populations homogenizes that signature.

Mutation rates, efficacy of selection and population structure can vary across the genome, which can confound scans for selection (CHARLESWORTH *et al.* 2003; STAJICH AND HAHN 2005). To work around this, we employed a control-gene resampling approach to identify the average difference from the background for each immune category (CHAPMAN *et al.* 2019). Consistent with our results previous results,

325 we find no signatures of recurrent positive selection in antifungal genes (Supplementary Figure 10, 61%) 326 resamples > 0) or cuticle genes (Figure 6, 54% resamples > 0) but do again find extremely high levels of 327 positive selection in AMPs (Figure 6, 100% resamples > 0) and Toll signaling genes (Figure 6, 99.1%) 328 resamples > 0). Segregating slightly deleterious mutations can bias inference of selection using McDonald-329 Kreitman based tests (MESSER AND PETROV 2012). To account for this bias, we also calculated asymptotic 330  $\alpha$  for all functional categories across the genome (HALLER AND MESSER 2017). To this end we calculated 331 the asymptotic  $\alpha$  for all functional categories across the genome (HALLER AND MESSER 2017). As before, 332 while we find signals for adaptation in antifungal and cuticle proteins (asymptotic  $\alpha > 0$ ), we find no 333 evidence of higher rates of adaptation than the background (Supplementary Figure 10, permutation test 334 Antifungal *p*-value = 0.243, Cuticle *p*-value = 0.137). Again, the only categories significantly higher than 335 the background are Toll signaling genes (Permutation test p-value = 0.033) and AMPs (Permutation test p-336 value = 0.035). Together these results suggest that while genes involved in antifungal resistance and cuticle 337 development are evolving adaptively, it is not recurrent across the whole functional category, instead only 338 occurring in one or two specific genes. Alternatively, the adaptation may be too recent to detect signal using 339 these metrics. Long-term recurrent adaptation appears to be driven by host-pathogen interactions (likely 340 with DiNV (HILL et al. 2019)) as opposed to local adaptation.

341 Figure 6: McDonald-Kreitman based statistics for immune categories in *D. innubila* and cuticle 342 development. The left two plots show estimated statistics (Direction of Selection and Selection Effect) for 343 each gene. The right two plots show the difference in average statistic (Direction of Selection and Selection 344 Effect) for each gene and a randomly sampled nearby gene.



## 346 Discussion

347 Migration and environmental change can drive adaptation (RANKIN AND BURCHSTED 1992; 348 CHARLESWORTH et al. 2003; GILLESPIE 2004; EXCOFFIER et al. 2009; PORRETTA et al. 2012; WHITE et al. 349 2013). Species with somewhat isolated or divided populations are likely to adapt to their differing local 350 environments. Migration can both facilitate and hinder such adaptation, allowing new variation (including 351 potentially beneficial variants) to be spread between populations and preventing inbreeding depression. 352 Strong migration can also import locally nonadaptive variants and prevent the fixation of the most fit 353 variants in local populations. We sought to examine the extent that these processes take place in a species 354 of Drosophila found across four forests separated by large expanses of desert.

355 We characterized the phylogeographic history of four populations of *Drosophila innubila*, a 356 mycophagous species endemic to the Arizonan Sky islands using whole genome resequencing of wild-357 caught individuals. D. innubila expanded into its current range during or following the previous glacial 358 maximum (Figure 1, Supplementary Figure 1). We find some evidence of local adaptation, primarily in the 359 cuticle development genes and antifungal immune genes (Figure 2, Supplementary Figure 2). Interestingly, 360 there is very little support for population structure across the nuclear genome (Figures 1 & 2, Supplementary 361 Figures 1 & 2), including in the repetitive content (Supplementary Figure 11), but some evidence of 362 population structure in the mitochondria, as found previously in *D. innubila* (DYER AND JAENIKE 2005). 363 This suggests that if gene flow is occurring, it could be primarily males migrating, as is seen in other non-364 Drosophila species (RANKIN AND BURCHSTED 1992; SEARLE et al. 2009; MA et al. 2013; AVGAR AND 365 FRYXELL 2014). Based on the polymorphism data available, coalescent times are not deep, and given our 366 estimated population history, this suggests that variants aren't ancestrally maintained and are instead 367 transmitted through migration between locations (CHARLESWORTH et al. 2003).

368 Segregating inversions are often associated with population structure and could explain the 369 abnormalities seen on Muller element B here (Supplementary Figures 2-5). Our detection of several putative 370 segregating inversions on Muller element B relative to all other chromosomes (Figure 3A) supports this 371 assertion. However, few of the putative inversions support this hypothesis, in that all are large and common 372 inversions characterized in all populations, suggesting the inversions are not driving the elevated F<sub>ST</sub>. We 373 suspect that the actual causal inversions may not have been characterized due to the limitations of detecting 374 inversions in repetitive regions with short read data (MARZO et al. 2008; CHAKRABORTY et al. 2017). The 375 elevated F<sub>ST</sub> could also be caused by other factors, such as extensive duplication and divergence on Muller 376 element B being misanalysed as just divergence. In fact, the broken and split read pairs used to detect 377 inversions are very similar to the signal used to detect duplications (YE et al. 2009; RAUSCH et al. 2012; 378 CHEN et al. 2016), suggesting some misidentification may have occurred. If a large proportion of Muller B 379 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). Further study is necessary to disentangle if inversions or other factors are causing this elevated  $F_{ST}$  and the selective and/or demographic pressures driving this differentiation. However, it is 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). Thus, 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.

387 We find very few signatures of divergence between samples from 2001 and 2017 (Supplementary 388 Figure 8). Though the environment has changed in the past few decades, there may have been little impact 389 on the habitat of D. innubila in the Chiricahuas, resulting in few changes in selection pressures in this short 390 period of time, unlike most bird and mammal's species in the same area (COE et al. 2012). Interestingly, 391 there was an extensive forest fire in 2011 which could plausibly have been a strong selective force but we 392 see no genome-wide signature of such (ARECHEDERRA-ROMERO 2012). Alternatively, seasonal 393 fluctuations in allele frequencies may swamp out directional selection. Excessive allele frequency change 394 is limited to a few genes with no known association to each other, and little overlap with the diverging 395 genes between populations. Some of the genes with elevated F<sub>ST</sub> (and differing in allele frequency) between 396 time points overlap with divergent genes between sexes, primarily at the telomere of the X chromosome 397 (Muller element A, Supplementary Figure 8). In fact, F<sub>ST</sub> is significantly correlated on Muller element A 398 between the two surveys (Pearson's correlation t = 82.411, p-value = 1.2e-16), even with the 2001-2017 399 survey only considering male samples, supporting an association between the factors driving divergence 400 between sexes and over time. Given the sex bias of SNPs in this region, this could suggest that a selfish 401 factor with differential effects in the sexes is located on the X chromosome near the telomere (BURT AND 402 TRIVERS 2006). Often these selfish elements also accumulate inversions to prevent the breakdown of 403 synergistic genetic components (BURT AND TRIVERS 2006), and the Muller A telomere appears to have 404 accumulated several inversions (Figure 3A). However, populations of D. innubila are already female-biased 405 due to the male-killing Wolbachia infection found in 30-35% of females (DYER 2004; DYER AND JAENIKE 406 2005; JAENIKE AND DYER 2008). Thus D. innubila could be simultaneously parasitized by both the male-407 killing Wolbachia and a selfish X chromosome. Alternatively, the strong signals associated with the 408 telomere of the X could be a signature of selection related to the *Wolbachia* infection (UNCKLESS 2011b).

Ours is one of few studies that sequences 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). The excess of putatively deleterious alleles 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).

414 To date, most of the genomic work concerning the phylogeography and dispersal of different 415 Drosophila species has been limited to the melanogaster supergroup (POOL et al. 2012; POOL AND 416 LANGLEY 2013; BEHRMAN et al. 2015; LACK et al. 2015; MACHADO et al. 2015), with some work in other 417 Sophophora species (FULLER et al. 2016). This limits our understanding of how non-commensal species 418 disperse and behave, and what factors seem to drive population demography over time. Here we have 419 glimpsed into the dispersal and history of a species of mycophageous Drosophila and found evidence of 420 changes in population distributions potentially due to the changing climate (SURVEY 2005) and population 421 structure possibly driven by segregating inversions and selfish elements. Because many species have 422 recently undergone range changes or expansions (EXCOFFIER et al. 2009; PORRETTA et al. 2012; WHITE et 423 al. 2013), we believe examining how this has affected genomic variation is important for population 424 modelling and even for future conservation efforts (EXCOFFIER et al. 2009; COE et al. 2012).

## 425 Methods

## 426 Fly collection, DNA isolation and sequencing

- 427 We collected wild *Drosophila* at the four mountainous locations across Arizona between the 22<sup>nd</sup> of August 428 and the 11th of September 2017: the Southwest research station in the Chiricahua mountains (CH, ~5,400 429 feet elevation, 31.871 latitude -109.237 longitude, 96 flies), in Prescott National Forest (PR, ~7,900 feet 430 elevation, 34.586 latitude -112.559 longitude, 96 flies), Madera Canyon in the Santa Rita mountains (SR, 431 ~4,900 feet elevation, 31.729 latitude -110.881 longitude, 96 flies) and Miller Peak in the Huachuca 432 mountains (HU, ~5,900 feet elevation, 31.632 latitude -110.340 longitude, 53 flies) (COE et al. 2012). Baits 433 consisted of store-bought white button mushrooms (Agaricus bisporus) placed in large piles about 30cm in 434 diameter, with at least 5 baits per location. We used a sweep net to collect flies over the baits in either the 435 early morning or late afternoon between one and three days after the bait was set. We sorted flies by sex 436 and species at the University of Arizona in Tucson, AZ and flash frozen at -80°C before shipping on dry 437 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). We also prepared the DNA of 40 *D. innubila* collected 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. We sequenced the libraries on four lanes of an Illumina HiSeq 4000 (150bp paired end) (Supplementary Table 1, Data to be deposited in the NCBI SRA).

#### 445 *Sample filtering, mapping and alignment*

446 We removed adapter sequences using Scythe (BUFFALO 2018), trimmed all data using cutadapt to remove 447 barcodes (MARTIN 2011) and removed low quality sequences using Sickle (parameters: -t sanger -q 20 -l 448 50) (JOSHI AND FASS 2011). We masked the *D. innubila* reference genome, using *D. innubila* TE sequences 449 generated previously and RepeatMasker (parameters: -s -gccalc -gff -lib customLibrary) (SMIT AND 450 HUBLEY 2013-2015; HILL et al. 2019). We then mapped the short reads to the masked D. innubila genome 451 using BWA MEM (LI AND DURBIN 2009), and sorted and indexed using SAMTools (LI et al. 2009). 452 Following mapping, we added read groups, marked and removed sequencing and optical duplicates, and 453 realigned around indels in each mapped BAM file using Picard and GATK 454 (HTTP://BROADINSTITUTE.GITHUB.IO/PICARD; MCKENNA et al. 2010; DEPRISTO et al. 2011). We then 455 removed individuals with low coverage of the D. innubila genome (less than 5x coverage for 80% of the 456 non-repetitive genome), and individuals we suspected of being misidentified as D. innubila following 457 collection due to anomalous mapping. This left us with 280 D. innubila wild flies (48 - 84 flies per 458 populations) from 2017 and 38 wild flies from 2001 with at least 5x coverage across at least 80% of the 459 euchromatic genome (Supplementary Table 1).

## 460 Nucleotide polymorphisms across the population samples

461 For the 318 sequenced samples with reasonable coverage, we called SNPs using GATK (MCKENNA et al. 462 2010; DEPRISTO et al. 2011) which generated a multiple strain VCF file. We then used BCFtools 463 (NARASIMHAN et al. 2016) to remove sites with a GATK quality score (a composite PHRED score for 464 multiple samples per site) lower than 950 and sites absent (e.g. sites of low quality, or with 0 coverage) 465 from over 5% of individuals. This filtering left us with 4,522,699 SNPs and small indels across the 168Mbp 466 genome of *D. innubila*. We then removed SNPs found as a singleton in a single population (as possible 467 errors), leaving us with 3.240,198 SNPs. We used the annotation of D. innubila and SNPeff (CINGOLANI 468 et al. 2012) to identify SNPs as synonymous, non-synonymous, non-coding or another annotation. 469 Simultaneous to the *D. innubila* population samples, we also mapped genomic information from outgroup 470 species D. falleni (SRA: SRR8651761) and D. phalerata (SRA: SRR8651760) to the D. innubila genome 471 and called divergence using the GATK variation calling pipeline to identify derived polymorphisms and 472 fixed differences in D. innubila.

#### 473 *Population genetic summary statistics and structure*

- 474 Using the generated total VCF file with SNPeff annotations, we created a second VCF containing only
- 475 synonymous polymorphism using BCFtools (NARASIMHAN *et al.* 2016). We calculated pairwise diversity
- $\label{eq:476} \mbox{ per base, Watterson's theta, Tajima's D (TAJIMA 1989) and F_{ST} (WEIR AND COCKERHAM 1984) (versus all$
- 477 other populations) across the genome for each gene in each population using VCFtools (DANECEK *et al.*
- 478 2011) and the VCF containing all variants. Using ANGSD to parse the synonymous polymorphism VCF

479 (KORNELIUSSEN et al. 2014), we generated synonymous unfolded site frequency spectra for the D. innubila

480 autosomes for each population, using the D. falleni and D. phalerata genomes as outgroups to the D.

481 *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 (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 (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.

## 490 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 97.5<sup>th</sup> percentile and 2.5<sup>th</sup> 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 97.5<sup>th</sup> 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 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 97.5<sup>th</sup> percentile for composite likelihood ratio per chromosome.
- 504 Using the total VCF with outgroup information, we next calculated Dxy per SNP for all pairwise 505 population comparisons (NEI AND MILLER 1990), as well as within population pairwise diversity and dS 506 from the outgroups, using a custom python script. We then found the average Dxy and dS per gene and 507 looked for gene enrichments in the upper 97.5<sup>th</sup> percentile, versus all other genes.

## 508 Inversions

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

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

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

512 VCF quality score lower than 200. We also called inversions using Pindel (YE *et al.* 2009) in these same 513 samples and again removed low quality inversion calls. We next manually filtered samples and merged 514 inversions with breakpoints within 1000bp at both ends and significantly overlapping in the 515 presence/absence of these inversions across strains (using a  $\chi^2$  test, *p*-value < 0.05). We also filtered and 516 removed large inversions which were only found with one of the two tools. Using the remaining filtered 517 and merged inversions we estimated the frequency of each inversion within the total population.

## 518 Signatures of recurrent selection

519 We filtered the total VCF with annotations by SNPeff and retained only non-synonymous (replacement) or 520 synonymous (silent) SNPs. We then compared these polymorphisms to the differences identified to D. 521 falleni and D. phalerata to polarize changes to specific branches. Specifically, we sought to determine sites 522 which are polymorphic in our *D. innubila* populations or are substitutions which fixed along the *D. innubila* 523 branch of the phylogeny. We used the counts of fixed and polymorphic silent and replacement sites per 524 gene to estimate McDonald-Kreitman-based statistics, specifically direction of selection (DoS) 525 (MCDONALD AND KREITMAN 1991; SMITH AND EYRE-WALKER 2002; STOLETZKI AND EYRE-WALKER 526 2011). We also used these values in SnIPRE (EILERTSON et al. 2012), which reframes McDonald-Kreitman 527 based statistics as a linear model, taking into account the total number of non-synonymous and synonymous 528 mutations occurring in user defined categories to predict the expected number of these substitutions and 529 calculate a selection effect relative to the observed and expected number of mutations (EILERTSON et al. 530 2012). We calculated the SnIPRE selection effect for each gene using the total number of mutations on the 531 chromosome of the focal gene. Using FlyBase gene ontologies (GRAMATES et al. 2017), we sorted each 532 gene into a category of immune gene or classed it as a background gene, allowing a gene to be classed in 533 multiple immune categories. We fit a GLM to identify functional categories with excessively high estimates 534 of adaptation, considering multiple covariates:

535 536

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

+ 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).

540 To confirm these results, we also used AsymptoticMK (HALLER AND MESSER 2017) to calculate 541 asymptotic  $\alpha$  for each gene category. We generated the non-synonymous and synonymous site frequency 542 spectrum for each gene category, which we then used in AsymptoticMK to calculate asymptotic  $\alpha$  and a 543 95% confidence interval. We then used a permutation test to assess if functional categories of interest 544 showed a significant difference in asymptotic  $\alpha$  from the rest of categories.

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## 556 Supplementary Methods

We used dnaPipeTE (GOUBERT *et al.* 2015) to quantify the extent that repetitive element content differed across the populations. Our approach assumed a genome size of 168Mbp, with the number of randomly sampled reads equal to 1-fold coverage of the genome, resampling each strain 2 times to get the average estimate of each strains TE content. Following TE identification, we grouped sequences by known superfamilies and compared the proportion of the genome composed of each superfamily across strains in the populations. We also used a reciprocal blast (e-value < 0.0000001) (ALTSCHUL *et al.* 1990) to identify TE families present in each strain.

We confirmed TE families shared between the previous RepeatModeler (SMIT AND HUBLEY 2008) annotation of the *D. innubila* reference genome and the dnaPipeTE annotation using blast (e-value < 10e-08) (ALTSCHUL *et al.* 1990). After confirming that they did not differ in content, we called TE insertions in each strain across the genome using PopoolationTE2 (KOFLER *et al.* 2016), then merged the output and calculated the frequency of insertions, grouping by TE order, population and if the insertion was exonic, intronic, non-coding or flanking a gene (500bp up or downstream of start or end). When considering individual TE families in *D. innubila*, we used the RepBase TE names and identifications (BAO *et al.* 2015).

# 571 Supplementary Results

We characterized the repetitive content across our samples using dnaPipeTE (GOUBERT *et al.* 2015) and called TE insertions per line using PopoolationTE2 (KOFLER *et al.* 2016). The reference *D. innubila* genome contains 154 different TE families along with varying satellites and simples repeats, with resequenced individuals varying from 4.4% to 38.4% of reads matching repetitive sequences. Strains varied from 1913 to 7479 TE insertions per strain in the non-repetitive portion of the genome. Like nuclear polymorphism, we find little population structure by shared TE insertions, though strains do seem to disperse primarily by the number of insertions (Supplementary Figure 11B).

Similar to *D. melanogaster* (CHARLESWORTH AND LANGLEY 1989; CHARLESWORTH *et al.* 1997;
PETROV *et al.* 2011; KOFLER *et al.* 2012; KOFLER *et al.* 2015), *D. innubila* harbors a significant excess of
low frequency TE insertions compared to the SFS of synonymous variants (Supplementary Figure 11A,
GLM Count ~ Frequency \* SNP or TE, t-value = -16.401, p-value = 1.889e-60), with no difference in the
insertion frequency spectra between populations (GLM Count ~ Frequency \* TE order \* Population, tvalue = -0.341, p-value = 0.733). This implies in every population, TE insertions are on average mildly
deleterious and removed via purifying selection.

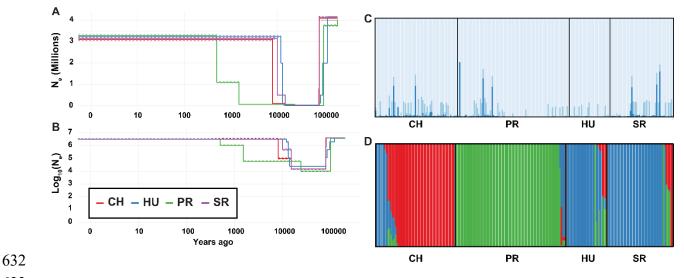
586 Using dnaPipeTE (GOUBERT *et al.* 2015), we find a significantly higher density of RC & TIR 587 elements compared to other repeat orders (Supplementary Figure 11C, t-value = 3.555 p-value = 3.745e-588 04), consistent with the reference genome (HILL *et al.* 2019). The density of repetitive content is also higher

589 genome wide in the CH and PR populations compared to HU and SR (Supplementary Figure 11C & D, t-590 value = 2.856, p-value = 4.291e-03). This is in keeping with a more recent bottleneck for these species 591 reducing effective population size and efficacy of selection, resulting in bursts of repeat activity with 592 relaxed selection for removal of insertions. These changes are primarily driven by an expansion of simple 593 repeats in the CH population (Supplementary Figure 11D, GLM t-value = 3.978, p-value = 7.31e-05) and 594 an expansion of TIR elements in the PR population (Supplementary Figure 11D, GLM t-value = 3.914, p-595 value = 9.52e-05). Specifically, we see expansions of the satellite CASAT HD (GLM t-value = 5.554, p-596 value = 8.832e-08) and the simple repeat sequences CAACAA, CTC and GTGT in the CH population when 597 compared to all other populations (GLM t-value = 9.204, p-value = 2.555e-17). In the PR population we 598 find significantly higher abundances of a TE families closely related to *Tetris Dvir* (GLM t-value = 13.641, 599 p-value = 2.889e-32), *Helitron-2N1 DVir* (GLM t-value = 12.381, p-value = 2.789e-28) and *Chapaev3-*600 I PM (GLM t-value = 11.472, p-value = 1.662e-24) compared to other populations. We do not find any 601 evidence that particular TE orders are more abundant on any one chromosome in D. innubila (GLM t-value 602 = 1.854, p-value = 0.633), though do find TEs are at significantly higher insertion densities in the inverted 603 regions of Muller element A than at the regions of the genome (Wilcoxon Rank Sum Test W= 19763, p-604 value = 0.01488). This suggests the lack of recombination in the inverted region is allowing the 605 accumulation of repetitive content on Muller element A.

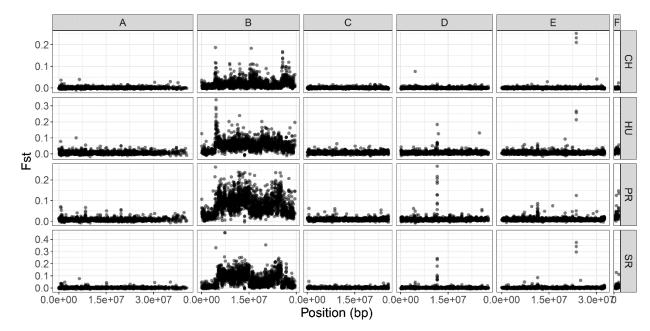
606 TE insertions are usually assumed to be at least mildly deleterious (CHARLESWORTH AND LANGLEY 607 1989; PETROV et al. 2011). In D. innubila, TE density is lower in regions flanking genes or within genes 608 compared to non-coding regions (GLM t-value = -6.538, p-value = 6.23e-11), consistent with the 609 deleterious assumption. However, the frequency of TE insertions was significantly higher in exonic regions 610 compared to introns and UTRs (Supplementary Figure 11A, GLM t-value = 4.040, p-value = 5.34e-05), 611 across all populations, which we may have observed as these are wild caught flies and so may have more 612 recessive deleterious insertions segregating in the population than are seen in inbred samples. Overall the 613 repetitive content in Drosophila innubila appears to be mildly deleterious, with TE insertions shared 614 between locations by migration. Despite this there are some major differences in the repeat content of each 615 population, possibly due to the stochastic effect of population bottlenecks.

This may have occurred due to a founder effect following the population bottleneck, where a majority of CH founders by chance had a higher proportion of particular satellites or simple repeats (CHARLESWORTH *et al.* 2003), but this is unlikely given the gene flow between populations. Alternatively, the bottleneck could have fixed segregating recessive variation which limits the regulation of repetitive content in the genome, leading to its expansion. However, if this was the case and satellite expansion is even mildly deleterious, we would expect migratory rescue of repeat regulation machinery. A third

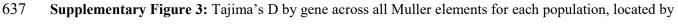
- 622 possibility is that satellite expansion is associated with local evolutionary dynamics either involved in
- 623 adaptation or genetic conflict (GARRIDO-RAMOS 2017; LOWER *et al.* 2018).
- 624 Supplementary Figure 1: A. Population size history of *Drosophila innubila* backwards in time for each
- 625 population. **B.** Population size history on the Log10 scale of *Drosophila innubila* backwards in time for
- 626 each population. C. Results of Structure software (FALUSH et al. 2003) for estimating population structure
- 627 between locations for 100,000 sampled synonymous polymorphisms from all autosomes, with a K=3
- 628 (estimated optimal K value). Note that this plot summarizes all autosomes (excluding Muller B) and the X
- 629 chromosome due to very little structure between locations for all chromosomes. **D.** Results of Structure
- 630 software (FALUSH et al. 2003) for estimating population structure between locations for 16
- 631 polymorphisms on the autosomes, with a K=3 (estimated optimal K value).



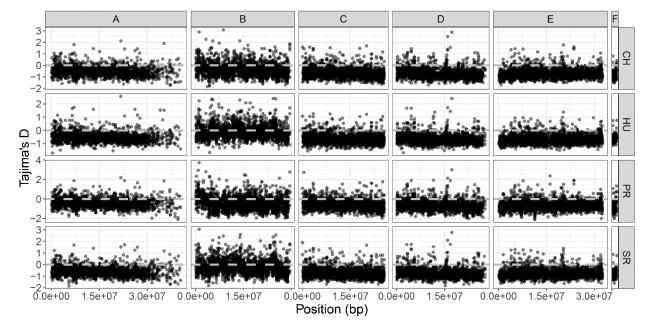
634 Supplementary Figure 2: Fst by gene across all Muller elements for each population, located by loci (in



bp) on the Muller element.



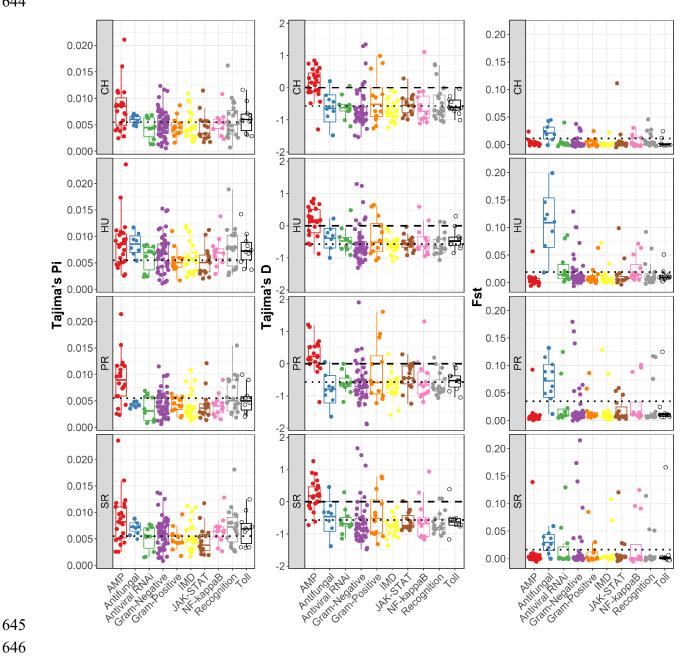
638 loci (in bp) on the Muller element. The grey dashed line shows a Tajima's D of 0.



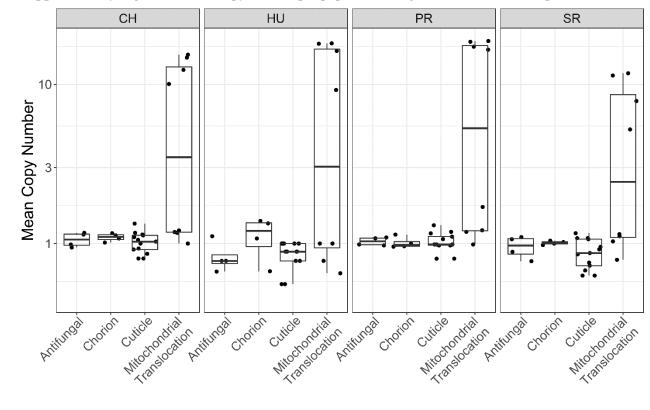
639 640

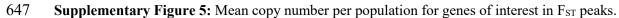
641 Supplementary Figure 4: Population genetic statistics (Pairwise diversity, Tajima's D and Fst) for genes 642 in each immune category, for each population. Each plot has a dotted line to show the genomic background 643 statistics for each population. The Tajima's D plot contains a dashed line to show 0.

644

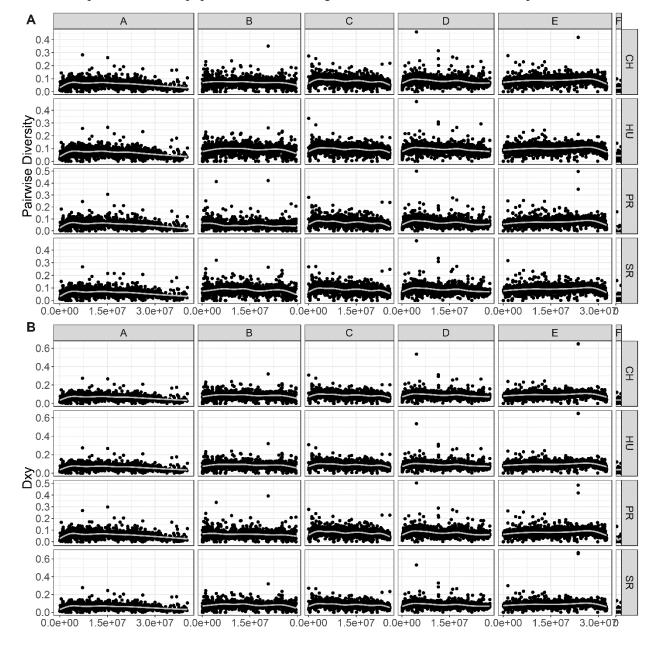


646

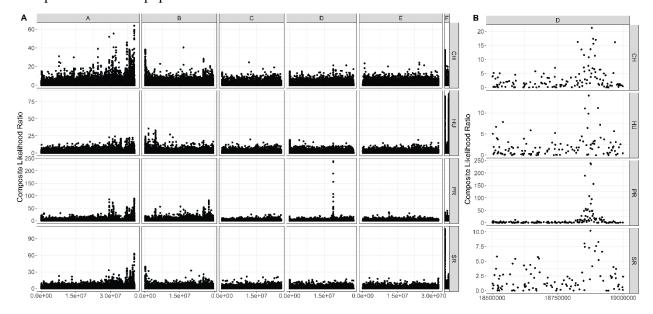


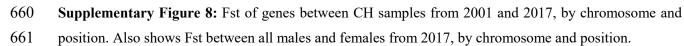


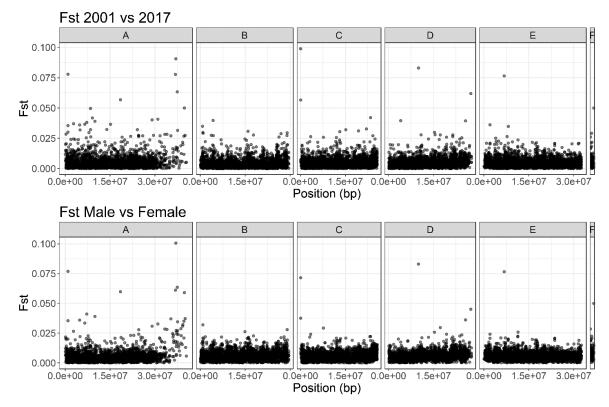
- **Supplementary Figure 6: A.** Within population pairwise diversity per gene across the *D. innubila* genome.
- **B.** D<sub>XY</sub> per gene for each population. Instead of showing all pairwise comparisons, we show one randomly
- 652 chosen comparison for each population, due to no significant differences between comparisons.



655 Supplementary Figure 7: Composite likelihood score for a selective sweep in 1kbp windows of the 656 genome estimated using Sweepfinder2. Separated by chromosome and population. A. Genome wide 657 composite likelihood score. B. Focus on 18.5-19Mbp of Muller element D, to show strongest selective 658 sweep in each the PR population.

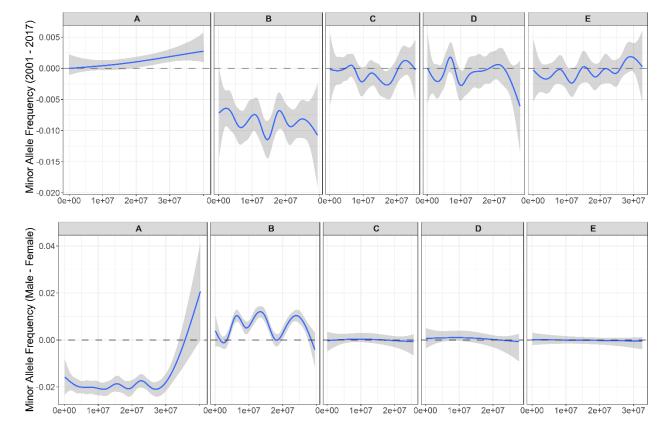






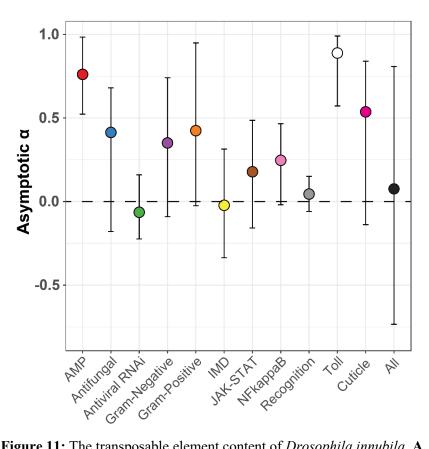
662

- 663 Supplementary Figure 9: Minor allele frequency difference curve across the genome (averaged over 2000
- 664 SNPs, sliding 1000 SNPs). Shows average difference in the minor allele frequencies (based on total 2017
- sample). Comparisons between 2001 Chiricahua and 2017 Chiricahua, and between all 2017 males and
- 666 2017 female samples.





**Supplementary Figure 10:** Asymptotic  $\alpha$  for immune categories, cuticle development proteins and all proteins, with 95% confidence intervals for categories. Categories marked with a \* are significantly higher than the background following a permutation test (<0.05). 0 is marked with a dashed line. In the 'All' category, the median and 95% confidence interval is calculated across all functional categories, while in the specific functional categories the 95% confidence intervals for Asymptotic  $\alpha$  are calculated using AsymptoticMK (MESSER AND PETROV 2012; HALLER AND MESSER 2017).



674

675 **Supplementary Figure 11:** The transposable element content of *Drosophila innubila*. A. Insertion

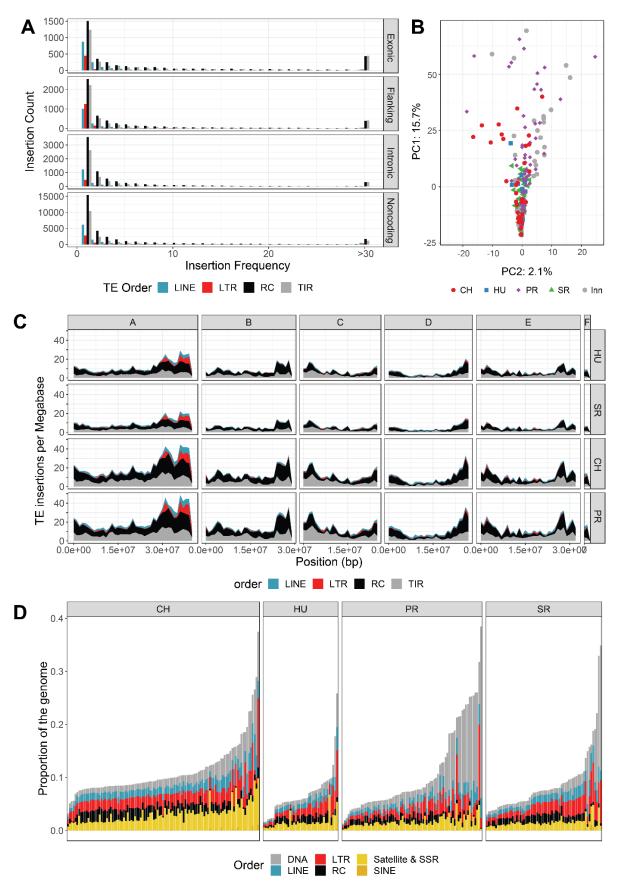
676 frequency spectra for TEs in *D. innubila* separated by TE order and the location of insertion (e.g. coding

677 region, non-coding, intronic). **B.** Principle component analysis (showing PCs 1 & 2) of TE insertions

678 across *D. innubila* strains shows little population structure. Strains are labelled by their population (both

679 shape and color). C. Mean TE insertion density per 1Mb window (sliding 1Mb) for each population of *D*.

- 680 *innubila*, identified using PopoolationTE2. TE insertions are colored by their order. **D.** Proportion of the
- 681 genome made up of repetitive content for each strain, as found with dnaPipeTE. Strains are ordered by
- total TE content from most to least, with bars colored by TE order.



- 684 Supplementary Table 1: Summary of *Drosophila innubila* fly's DNA collected and sequenced for this
- study, including summary of coverage for X chromosome, autosomes. Also contains SRA accessions for
- 686 each strain.
- 687 **Supplementary Table 2:** GLM for population genetic statistics in immune gene categories relative to the
- 688 background for each population.
- 689 Supplementary Table 3: Summary of gene ontology enrichments for F<sub>ST</sub> in each population, separated
- 690 by processes, components and functions.
- 691 Supplementary Table 4: Summary of GLM for elevated McDonald-Krietman statistics GO categories in
- 692 D. innubila.
- 693 Supplementary Table 5: Summary of gene ontology enrichments for D<sub>XY</sub> in each population.
- 694 Supplementary Data 1: VCF file for SNPs in *D. innubila*, used in estimation of population genetic
- 695 statistics and in GWAS.
- 696 Supplementary Data 2: Population genetic statistics calculated for each gene in *D. innubila* using
- 697 VCFtools for each population.
- 698 Supplementary Data 3: McDonald-Kreitman statistics calculated for each gene in D. innubila using
- 699 SnIPRE for each population.
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