

1 **Tropical high-altitude insects show limited capacity to handle high temperatures**

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21 Abstract

22 Growing summer season, increased anthropogenic activities poses a continual challenge to
23 resident species Ectotherms like insects are especially vulnerable to rapid climatic changes.
24 High-altitude tropical insect populations have been rarely examined for their responses to high-
25 temperature. We exposed a tropical highland out-bred population of *Drosophila melanogaster*
26 from the Himalayas to growing summer conditions in outdoor mesocosm units. Population
27 response to thermal changes was tracked over ninety days at phenotypic and genotypic level.
28 Whole genomic resequencing data suggested a clear seasonal allelic shift. Interestingly, the
29 general heat responsive genes were missing in the summer due to monsoon allele shift; an
30 atypical response noted for high-altitude populations. Instead, candidates involved in kinases and
31 phosphorylation emerged as key players. Heat-knockdown time decreased over time indicating a
32 limited ability to handle increasing temperature. Merging data from both allelic shifts and heat-
33 knockdown time indicated a limited capacity for high-altitude insects in handling climate
34 warming.

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40 **Keywords:** *Drosophila*, Heat, Climate change, Rapid adaptation, High altitudes

41 1. Introduction

42 Natural habitats are heterogeneous and organisms come across diurnal and seasonal
43 temperature changes. Understanding how populations respond to environmental variations is a
44 fundamental question in ecological and evolutionary physiology. While it is known that
45 temperature influences the physiology and reproductive rates in a single generation, temperature
46 fluctuation across generations also influences thermo-sensitive life history traits (Klepsatel,
47 Girish, and Gálková 2020) In particular, *Drosophila* and indigenous species in certain native
48 conditions thrive in a short timeframe and may have compromised the ecosystem (Soto-Yéber et
49 al. 2018). To check this, several hypotheses exist in understanding the phenotypic plasticity that
50 these species show in changing environmental conditions.

51 Adaptive selection occurs at segregating allelic variants and requires time. There is an
52 alternative mechanism which comes without the cost of selection. Often a single genotype could
53 demonstrate different phenotypes in different environments and this property is known as
54 phenotypic plasticity (Bradshaw 1965). While it is a response to short-term environmental
55 change which could be over seasonal timescales (Bradford and Roff 1993; Brakefield and
56 Reitsma 1991), it has been well documented across insect taxa. For example, dung fly *Sepsis*
57 *cynipsea* shows body size variation throughout the summer (Blanckenhorn et al. 1999). A
58 seasonal change in color morphs' frequency has been observed in *Adalia bipunctata* beetle
59 (Brakefield 1985). Seasonal changes in color morphs of *Drosophila ananassae* have been found
60 associated with drought tolerance (Parkash et al. 2012). But the genetic basis of phenotypic
61 plasticity and its interplay with seasonal dynamics is poorly understood. Secondly, adaptive
62 tracking has not been given a required importance as it was considered that adaptive evolutionary

63 change is a very slow process and hard to expect any change in ecological time-scale (Stone,
64 Erickson, and Bergland 2020; Slobodkin 1980).

65 On the other hand, temporal variations are geographically widespread and are cyclic.
66 Many insect species reproduce multiple times during a growing season and pass through several
67 generations, which could shape the population ecology and trigger rapid adaptations. To
68 understand temporal variations, studies are required which undergo phenotypic plasticity at
69 various timescales (Siepielski, DiBattista, and Carlson 2009). The one basic assumption is that
70 traits which increase fitness in a given environment will be favored. A considerable work has
71 been done along these lines to understand the season dynamics (Rudman et al. 2022a; Dempster
72 1955) . Unfortunately, all of these studies come from temperate regions where climatic
73 conditions and environmental stress pressure is significantly lower when compared to tropical
74 hot conditions, which are on the threshold side. Furthermore, during fall, temperate regions show
75 more severe environmental conditions when compared to spring, whereas in the tropics, peak
76 summer season is a major determiner to population dynamics. For example, diapause is a
77 temperate phenomenon and never evolved in tropical populations (Schmidt and Conde 2006).
78 There are indeed changes in allele frequencies associated with thermal tolerance and desiccation
79 as some of the genes induce pleiotropic effects.

80 In the tropics, temperatures often cross permissible physiological limits for insects.
81 Studying thermo-sensitive genomic regions will shed light on molecular mechanisms behind
82 temperature sensation besides understanding insects' physiological responses to future climate
83 warming scenarios. The role of temperature responsive genes (TRGs) in ectothermic physiology
84 is not known much; our understandings are limited and have never been explored under natural
85 conditions and particularly not in high-altitude tropical mountain populations. The powerful

86 genetic tractability of the *Drosophila* model system has been instrumental in elucidating TRG
87 functions (Venkatachalam 2007). Noxious temperature sensing has evolved at both larval
88 (Daniel Tracey Jr, Wilson, and Laurent 2003) and adult (Xu et al. 2006) stages of Drosophilids..
89 Since *Drosophila* piggybacks between these thermal preferences, particularly in the early life
90 stages, state of relative conservation among species and unrelated temperature changes at species
91 origin was felt necessary to understand the TRGs oscillations.

92 In this work we collected a *Drosophila melanogaster* population from a high-altitude
93 locality in Western Himalayas, outcrossed it, released it into the mesocosm units in a tropical
94 orchard set-up with overlapping generations, and sub-sampled it at three different time points
95 (TPs) spanninga duration of ninety days. We aim to demonstrate that direct observation of
96 evolution in ecological time frames could resolve basic questions about adaptation. Therefore we
97 sampled the populations at different time-points (i.e. growing tropical summer season) for whole
98 genome re-sequencing and attempted to address fundamental questions around pace, temporal
99 dynamics, and underlying genomic architecture of populations under heat selection. We attempt
100 to discuss the underlying results of TRGs at different TPs and the raison d'etre behind phenotypic
101 plasticity in the insects for understanding their adaptive evolution. The dynamic responses in
102 these traits and variations at the plasticity level are complex and active areas of investigation to
103 understand how fast organisms are responding to these stresses (Maurya et al. 2021; Rudman et
104 al. 2022b).

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106 **2. Materials and Methods**

107 **2.1 *Drosophila* collection and maintenance**

108 We used a highland population of *D. melanogaster* which was collected from Western
109 Himalayas (Rohru, Himachal Pradesh, India; 31.2046⁰N, 77.7524⁰E, Altitude 1554 meters). The
110 flies were collected using banana baits in the year 2018. Each collected female was placed in a
111 separate tube at the site of collection. Almost 270 such lines were made in the field using females
112 captured during the field trip. Upon 10 to 15hr arrival to the lab, vials were placed at room
113 temperature (25⁰C). The successful lines were examined based on the taxonomic keys. Overall,
114 40 lines of *D. melanogaster* were successful and were maintained in the laboratory, on standard
115 ‘agar-jaggery-yeast-maize’ media at 25⁰C, with a 12/ 12light/dark cycle. All the 40 lines were
116 used in this work.

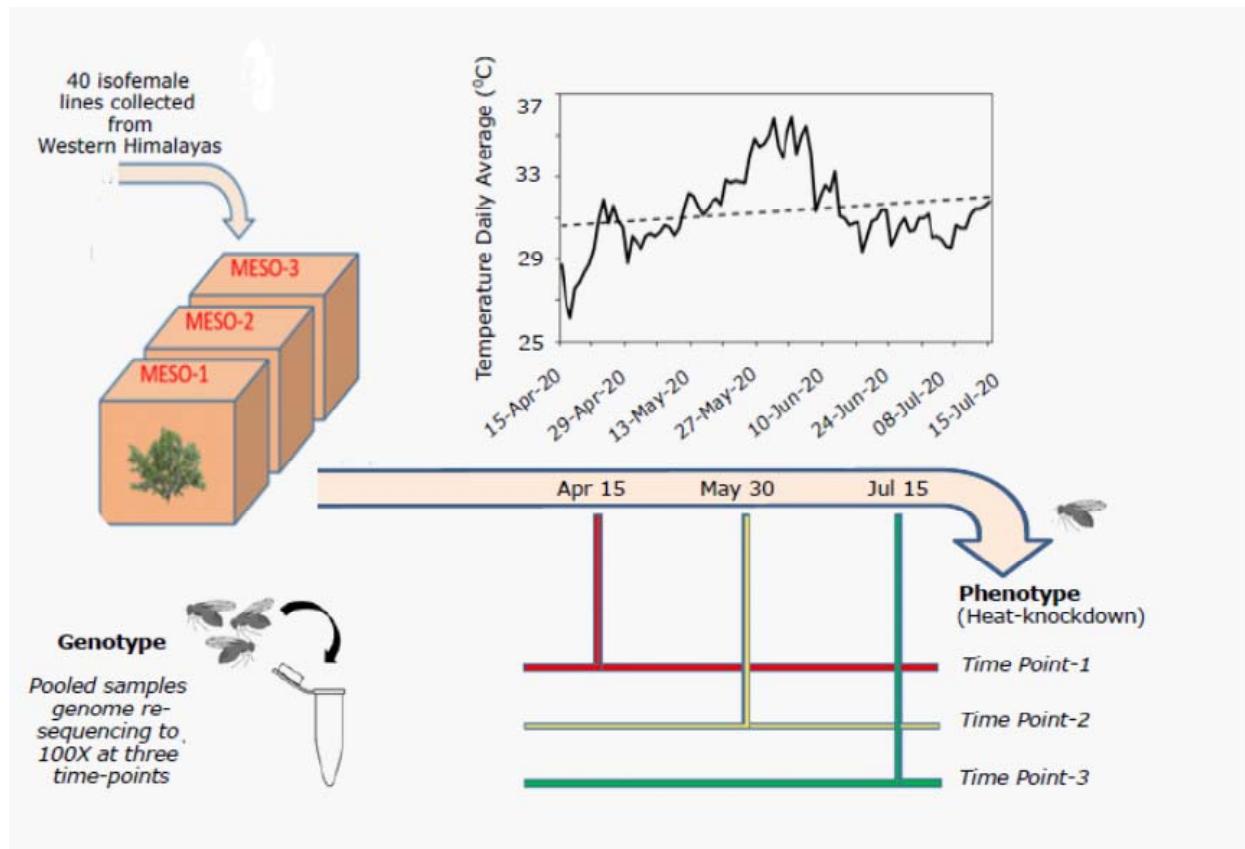
117 **2.2 *Out-crossing***

118 Forty isofemale lines of *D. melanogaster* were outcrossed (founding numbers 1000
119 adults: 500 males + 500 females) in a 0.61m X 0.61m X 0.61m dimension cages made up of
120 plexiglas acrylic sheet. For this, 10 males and 10 females each from 40 isofemale lines were
121 used. Each generation of eggs was collected in twenty half-pint size culture bottles. As soon as
122 the pupae became darker and ready to enclose, bottles were moved to the plexiglass cage and
123 plugs were removed. This allowed random mating in the founding base population. Once all the
124 flies had enclosed in the cage, eggs were collected and transferred to fresh food bottles for the
125 next generation. These lines were outcrossed for 5 generations keeping populations’ size
126 between 5-6 thousand individuals every generation.

127 **2.3 Mesocosm Cages**

128 To track adaptive changes in real time, and whether or not changes in temperature
129 (increasing temperature over growing tropical summer) lead to changes in allelic frequencies, we
130 tracked populations (3-fold replication) in outdoor mesocosm cages for over the period of 90
131 days (Fig. 1). For this, we used three outdoor cages (Ahmedabad University Experimental
132 Evolution Station). Each mesocosm was a 1.5x1.5x1.5 meters in dimension (custom designed at
133 Ahmedabad University Fabrication Shop) outdoor insect rearing enclosure surrounding a mature
134 (dwarf) sapota tree. Three cages were used for this experiment (MESO-1, MESO-2, and MESO-
135 3). Each cage was founded with 1000 males and 1000 females collected from the 4th generation
136 of laboratory cage (i.e. outcrossed population). Every morning at 09.00 am fresh food (50 ml of
137 standard ‘agar-jaggery-yeast-maize’) in a half-pint bottle was placed in each enclosure for the
138 entire duration of the experiment (15th April 2020 to 15th July 2020). Flies were allowed to
139 oviposit on the fresh food for 24 hours. Each morning, bottles with eggs laid in the last 24 hours,
140 were sealed with cotton plugs and larvae were allowed to develop (inside the same cages); upon
141 eclosion, bottles were opened and adults were released into the cages. Thus populations were
142 cultured under a natural regime of overlapping generations. Temperature and RH in the cages
143 were recorded using HOBO U23 Pro v2 data loggers (Onset Computer Corp., Bourne, MA,
144 USA).

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146

147 **Figure 1. Outdoor mesocosm experimental set-up.** Forty *D. melanogaster* isofemale lines
148 were pooled to generate an outbred population. Lines were outcrossed for 5 generations in a
149 24x24x24 inch dimension cages under laboratory conditions (25⁰C). From this outcrossed
150 population (maintaining the size close to 5000-6000 individuals every generation) we released
151 500 males and 500 females in each outdoor mesocosm cages (MESO-1, MESO-2, and MESO-3).
152 For the next 90 days food was changed every day and the previous bottle with eggs was capped
153 and placed in the cage itself. Upon eclosion adults were released in the existing cage and bottles
154 were discarded. Cages were sampled at three time-points (TP) for genomic and phenotypic
155 analysis: at the beginning (TP1), in the middle (TP2) and at the end (TP3).

156 **2.4 Adaptive tracking of populations**

157 Samples for genomic and phenotypic measurements were collected at three time-points
158 for adaptive tracking. Samples for TP1 were collected from the founding adults while the TP2
159 and TP3 samples were collected on the 45th and 90th day, respectively. Eggs were collected in
160 30 bottles at lower density (10 bottles for each replicate cage). These bottles were allowed to
161 develop under a constant temperature (25°C) in the laboratory and phenotypic data was collected
162 using F3 individuals.

163 **2.5 DNA sample preparation for genomic analysis**

164 Pooled genome resequencing was performed with 100 *D. melanogaster* females in
165 triplicate from each outdoor cage, a total of nine samples were collected for whole genome
166 sequencing analysis. The DNA extraction procedures involved homogenizing the flies (n = 100)
167 in 200 µL of lysis buffer (100 mM Tris-Cl, 100 mM EDTA, 100 mM NaCL, 0.5% SDS). The
168 homogenate was incubated at 65°C for 30 minutes. Following this, proteins were precipitated by
169 adding 800 µL of 2 parts 5M potassium acetate, 5 parts of 6M lithium chloride solution and 15
170 minutes ice incubation. The supernatant was collected after centrifugation of the homogenate at
171 12K RPM for 15 minutes at room temperature. DNA was precipitated by adding 800 µL of
172 isopropanol and centrifugation at 12K RPM for 15 minutes. The pellet was collected and washed
173 with 70% ethanol. After removing the ethanol, the pellet was allowed to dry at room temperature
174 and was re-suspended in 100 µL of TE buffer.

175 **2.6 Library construction and quality control**

176 1 µg of DNA was used as input material for the DNA sample preparations. Sequencing
177 libraries were generated using NEBNext® DNA Library Prep Kit following manufacturer's
178 recommendations and indices were added to each sample. The genomic DNA is randomly

179 fragmented to a size of 350bp by shearing, then DNA fragments were end polished, A-tailed, and
180 ligated with the NEBNext adapter for Illumina sequencing, and further PCR enriched by P5 and
181 indexed P7 oligos. The PCR products were purified (AMPure XP system) and resulting libraries
182 were analyzed for size distribution by Agilent 2100 Bioanalyzer and quantified using real-time
183 PCR.

184 **2.7 *Next Generation Sequencing***

185 The whole genome sequencing (WGS) of natural populations was done to check the role
186 of mutations in functional candidates. Library preparation and fragmentation was done using
187 DNA-350 bp by default, while the sequencing was performed on a HiSeq PE150. The samples
188 were mixed for library construction using a PCR-free library preparation guide. The downstream
189 analysis and annotation was done with raw reads run through an in-house benchmarked pipeline
190 (Meena et al. 2018), tweaked for a whole exome pipeline (<https://github.com/prashbio/WES>).

191 After the allele frequencies were mapped, we read all mutation positions in MESO-1,
192 MESO-2 and MESO-3 for all the three time point scales (T1|T2|T3) and created a data structure
193 which contained three types of positions: single, double and triple mutations. For example,
194 single types of mutations are those that are positioned only once across the 3 different TPs,
195 double- any two TP scales while triple – across all three of them. This was done for MESO-1,
196 MESO-2 and MESO-3 (experimental scale) and the resulting files were checked for interpreting
197 the possible SNPs. However, we asked whether mutations that are identical are present at least
198 in 2 time points at the same position and finally a composite matrix containing mutational
199 positions of MESO-1, MESO-2, and MESO-3 was done.

200 **2.8. *Heat-knockdown assay***

201 We tested F3 females (ca. 3 days old) for their thermal tolerance to an increasing
202 temperature using the dynamic heat ramp assay. In this assay, temperatures ecologically relevant
203 to the species are the base-point after which they are steadily increased till the fly is knocked
204 down (Lutterschmidt and Hutchison 1997). Individual flies were collected in glass tubes of
205 0.05m³ volume. The tubes were immersed in a water bath set at room temperature (ca. 25°C).
206 Temperature was increased at the rate of 1°C/ minute with the upper thermal limit being 43°C.
207 We measured three technical replicates with 24 flies each from three cages across three time
208 points. Nested ANOVA (experimental cages within time-points) was used to assess the variation
209 in knock-down time. Post-hoc analysis was performed using the *glht* () function from the
210 “multcomp” package (Hothorn, Bretz, and Westfall 2015). For genomic analysis, the flies were
211 aspirated from the cages itself.

212

213 **3. Results**

214 ***3.1. Allele frequencies***

215 The MESO-1, 2, & 3 replicates were run through the whole genome sequencing (WGS)
216 pipeline where we inferred causal SNPs across these sub-population (Table 1). The minor allele
217 frequencies were tabulated from the depth with number of reads showing variation (DP4) with
218 1) forward ref alleles; 2) reverse ref; 3) forward alt; 4) reverse alt alleles, used in variant calling
219 using the formula, variant allele frequency(VAF) = (forward alt + reverse alt alleles) / (forward
220 ref alleles + reverse ref + forward alt+ reverse alt alleles). After setting up the MAF<=0.15
221 assuming that 15% of the SNPs would have this MAF, we further retained heterozygous SNPs.
222 All 9 raw reads yielded 9,963,857 SNPs per sample. The common most SNPs between the
223 MESO-1, 2, & 3 cages and the outliers were tabulated by mapping the chromosomal positions to

224 flybase reference genome (flybase.org last accessed date: September 22, 2021) (Table 1/Fig. 2).
225 By removing those specific minor allele frequencies (MAFs) above 0.15, we could assume that
226 there is a clear distribution of contamination. We therefore sought to assess the possibility of
227 observed TRG with MAF <0.15 which gave us an average of 240 SNPs per sample
228 (Supplementary Table 1). This raises the possibility that allele frequency could be generated by
229 *D. melanogaster* alleles consistently at various time point scales. We further assessed the
230 frequency of SNPs that are fixed for alternate alleles and cross checked this with the previously
231 identified SNPs associated with seasonal changes (Temperature) and tabulated all SNPs for time
232 points (Table 1).

233

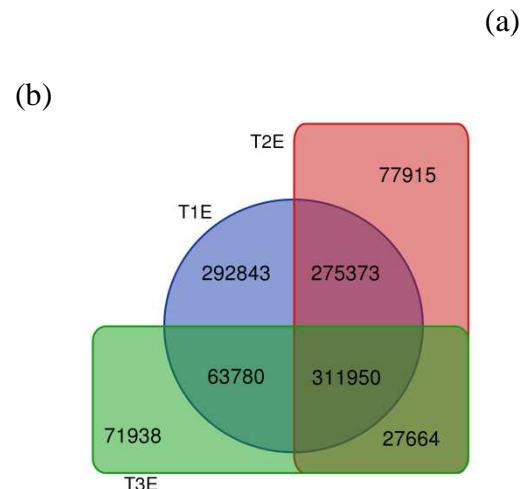
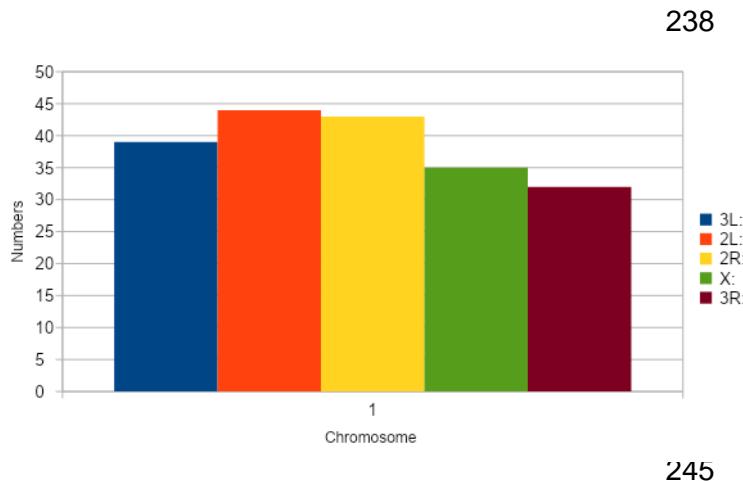
234 **Table 1: Average number of SNPs after filtering**

Total identified SNPs/	89,674,715/9,963,857
Average SNPs per TP sample	
>=10 DP <=400	5,015,244
Total number of SNPs with MAF <0.15/	3,573,280/397,031
Average MAF <0.15/sample	
Total used in analysis/	5,015,244/ 240
Average bona fide SNPs per TRG	

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248 **Figure 2.** (a) Flybase SNPs mapped to our experimental candidates across all chromosomes (b)
249 The SNPs mapped across the three time point experimental pools. The numbers indicate the
250 individual time point SNPs apart from the most common SNPs at the intersection.

251

252

253 **3.2. Positional mapping with Flybase SNPs**

254 The downstream annotation was processed with chromosomal alias obtaining 1870 total
255 number of scaffolds with an average length of 17,195,935 per chromosome. The chromosomal
256 distribution of SNPs is largely at the interface of 2L and X (Data not shown; Supplementary
257 Table 1, Fig. 3). We also mapped the SNPs to that of reported 193 SNPs from flybase to check if
258 TRG mutations exist. We found Single TP SNPs: (n=noPOS: 4,666), Double TP SNPs: POS:
259 (n=2,536) and no Triple TP SNPs (n=POS: 0).The final list of protein coding genes were mapped
260 across various chromosomes (Supplementary Table 1). As the current analysis represented

261 identifying common mutations, we further looked into mapping the existing known SNPs (for
262 heat stress).

263

264 ***3.3. Population comparisons***

265 The time point samples and replicates mapped to the temperature regime were treated as
266 T1_MESO-1, T2_MESO-1 and T3_MESO-1 which we call them as populations. As we obtain
267 the data, the coverage across these three sub-populations were filtered for common mutations or
268 SNPs that have been retained from T1_MESO-1,2&3 through T3_MESO-1,2&3 or the unique
269 set of SNPs through them (Supplementary Table 1). While removing the SNPs with total counts
270 with the same allelic proportion, we deem that they have not undergone any changes through the
271 three experiments. On the contrary, those SNPs that have an overlap with changes in SNPs
272 across the same positions were filtered and we therefore limited the downstream analysis to such
273 overlapping SNPs from three sub-populations or experiments. As we combined all the reads, we
274 have checked the significant differences between the experiments for overlapping SNPs using
275 the chi-square test as a standard practice. The *p*-value heuristics resulting from the analysis based
276 on three population comparisons were in agreement. Another approach was to check the SNP
277 mean density which could be done using an existing method (proposed by Burke et al 2010
278 (Burke et al. 2010); Winbush & Singh 2021 (Winbush and Singh 2021)) which measures the
279 quantile score but our study is not a measure of nucleotide diversity from a threshold value
280 instead the number of SNPs emerging from a whole genome sequencing study.

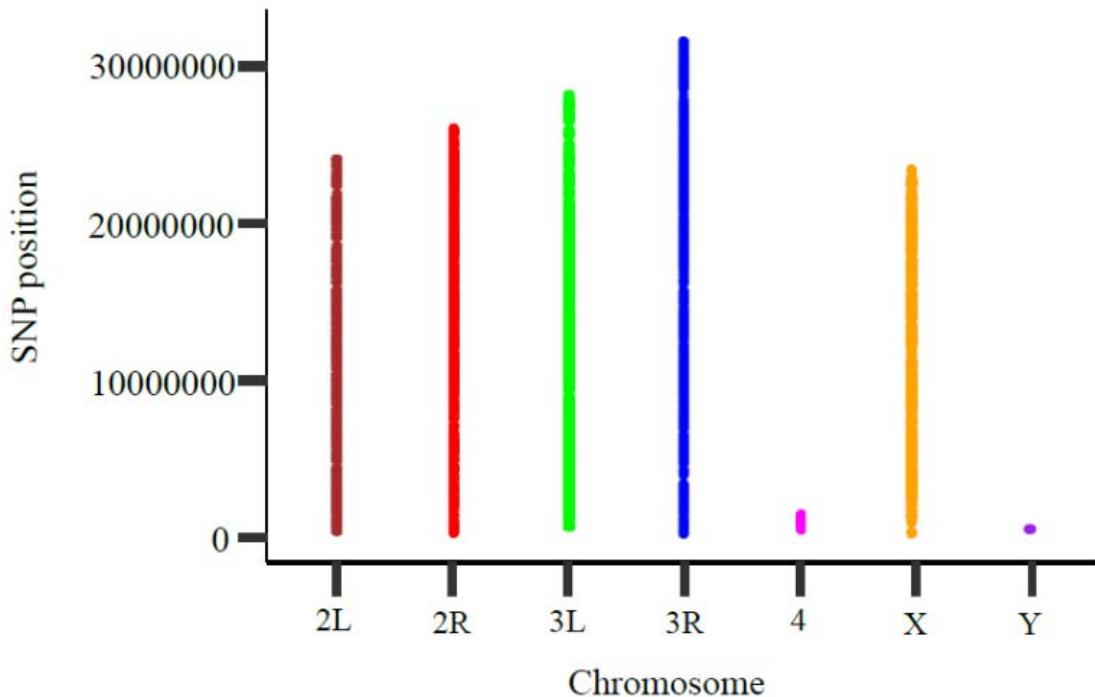
281 While we have found SNPs that have been deleted from T1 to T3 (Supplementary File 2),
282 the MAFs in earlier cases were set to as big as MAF<0.89 which is 89%. Our stringent cutoff
283 (Winbush& Singh, 2021) indicates that highly sensitive SNPs emerge from our analysis and

284 when these were tried to map them to flySNP base, none or not many of them are mapped
285 indicating that these are novel (data not shown). Nevertheless, we lacked sequencing data on
286 similar scales where whole genome sequencing (WGS) was done so that we could have
287 identified convergent or divergent regions contributing between the three sub-populations. For
288 each comparison, we observed that 77,915 (24.97%), 71,938 (23.06%) and 27,664 (8.8%) SNPs
289 in the respective three time points existed indicating that the temperature affects the sensitivity of
290 the flies. We then mapped these SNPs to summarize the genes that have escaped the mutations in
291 T3_MESO-1, 2 &3. When we map them to the table depicting counts for genes associated with
292 these SNPs for T3_MESO-1, 2 & 3, we observe that a large number of diseased phenotypes are
293 associated with these conditions. However, despite consistent read depth of coverage and SNPs,
294 we observe a partial overlap suggesting that the accorded convergence can be attributed to
295 slightly reduced heterozygosity which may invite a bias. However, this was reconfirmed from
296 our validation set of founder experiments.

297 As we identified the genes with diseased phenotypes we were interested to see whether or
298 not the candidate loci of these genes have an influence on the pathways. To check this, our
299 panther gene ontology annotation derived enriched genes unique to temperature. In summary,
300 our analysis identified a large number of candidate genes and SNPs associated with heat
301 handling wiring machinery. Rather, our approaches indicate the variation across different
302 subpopulations, as we argue that several diseased phenotypes have emerged from our analysis.
303 Taken together, temperature sensitivity plays a very important role and serves as a determinant to
304 mesocosm experimental study. Although genetic drift augurs well for taking temperature as a
305 context, functional validation for observational variation is presumptive for maintaining such

306 selection pressure. It may be argued that the function of the genes stemming from allelic
307 variation has allowed us to determine candidate TRGs.

308



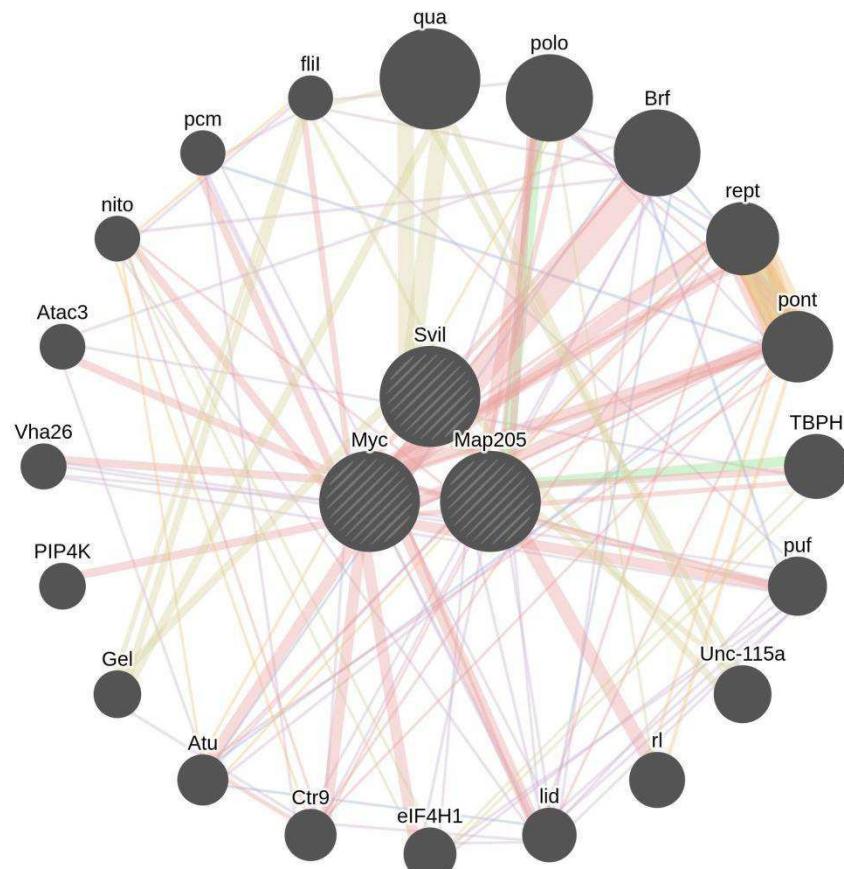
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310 **Figure 3.** Variable polymorphisms shown to have a large deviation from time point scale 1 to
311 time point 3 taking into account read depth and the number of sampled chromosomes. The X-
312 axis shows the chromosomes with the Y-axis showing the single nucleotide polymorphism
313 (SNP) positions.

314

315 To check genomic signatures affected by seasonal fluctuations, we pooled sequencing
316 data containing the SNPs by estimating the allele frequencies (AF). From each population and
317 time point, the three replicate cages were merged and the files were segregated to average read
318 depth of 20–200× coverage (Supplementary Table 1). We observed that these estimated AFs

319 have been shown to be accurate for understanding the magnitude of genetic variation through the
320 time point scales. From Fig. 1, we reason that variable polymorphisms had a large deviation from
321 time point scale 1 to time point 3 taking into account read depth and the number of sampled
322 chromosomes. Of the 3,119,540 common SNPs tested, we identified substantial numbers falling
323 between the statistical values, meaning they are called integrated thermal responsive SNPs.
324 However, the AFs corresponding to selection coefficients per generation were not done with this
325 statistical power as they are variable. Our rationale was to assess evolutionary features
326 underlying rapid adaptive states in response to selection pressure. As the data is variable with
327 the random disturbance and is different across elements of the vector, we asked whether the
328 SNPs are enriched among functional genetic elements. To find out whether or not the SNPs are
329 genic, we screened total SNPs present on specific genes common to these time points. A putative
330 interaction map indicates that the genes are largely spread across with the key pathways
331 attributing to these factors are MAP kinases and other important signaling pathways (Fig. 4). In
332 general, we find four important genes and a few un-characterized genes which could be
333 attributed to pleiotropy and linkage disequilibrium largely due to TRGs/seasonal variation.



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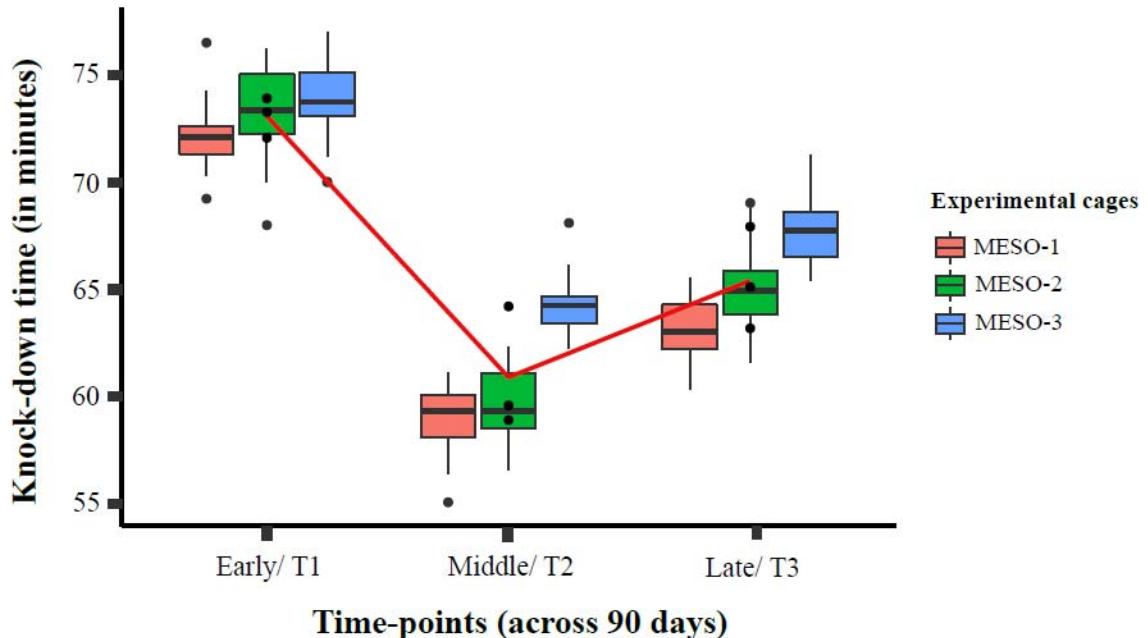
335 **Figure 4.** A putative interaction map of Svil, Myc and MAP205 with interacting partners
336 connected through edges (lines). The pink edges indicate that they are known to be physically
337 interacting with each other while other edge colors denote pathways, domain association, co-
338 localization, neighborhood and genetic interactions if any.

339

340 **3.4. Heat tolerance**

341 Knock-down times of flies significantly differed among all three time-points ($df = 2$, $F =$
342 966.30, $P < 0.0001$) (Supplementary Table 2). Heat knockdown time was significantly lower for
343 flies from T2 (middle time-point) (median = 60.38 minutes, interquartile range (IQR) = 4.55, $N =$
344 72) in comparison to early (T1) (median = 73.22 minutes, IQR = 3.07, $N = 72$) (estimate = -

345 13.151, $t = -27.115$, $P < 0.0001$) and late (T3) (median = 65.39 minutes, IQR = 3.71, N = 72)
346 (estimate = 4.275, $t = 8.814$, $P < 0.0001$) time-points. Variation in heat knockdown response is
347 partially explained by variation in experimental cages too (df = 6, $F = 42.12$, $P < 0.0001$) (Fig. 5).



348

349 **Figure 5:** Box-plots representing variation in knock-down times in response to heat ramping
350 assays, across three time-points (early, middle and late) from three replicate experimental
351 cages. Flies from the middle time-point (45th day) display the significantly lowest tolerance
352 to increase in temperatures compared to the early (day 0) and late (90th day) time-points.

353

354 **4. Discussion**

355 An organisms' response to changing environmental conditions determines its fitness.
356 Adaptations to rapid environmental fluctuations are known (Franks and Hoffmann 2012).

357 Responses to rapid environmental change could be micro-evolutionary or mediated through
358 phenotypic plasticity (Sgrò, Terblanche, and Hoffmann 2016). Trait plasticity in response to
359 climate change has been documented across diverse taxa, e.g. in birds (Charmantier et al. 2008),
360 mammals (Nussey, Wilson, and Brommer 2007), and insects (Ragland and Kingsolver 2008).
361 Still, climate-mediated-response studies display a temperate geographical bias (Feeley, Stroud,
362 and Perez 2017). For example, a “global” study (Seebacher, White, and Franklin 2015) found
363 physiological rates (Q_{10}) of key metabolic enzymes to be on the rising trend in the past few years
364 (calculated from 1990-2010). The authors however admit biased conclusions owing to a greater
365 representation of sampled studies from North America and Europe in contrast to tropical regions.
366 Again, while temperature is the central focus of most of these climate-mediated-response studies,
367 the confounding effects of another strong seasonal predictor – photoperiod -- are often
368 overlooked (Bradshaw and Holzapfel 2008). Diapause, an adaptation to survive freezing winter
369 temperatures but sensed through shorter photoperiods (Manenti, Sten, and Loeschcke 2021;
370 Danks 1987) is common for temperate rather than tropical species (Denlinger 1986).

371

372 Contrarily, tropics exhibit lesser variation in photoperiod compared to temperate regions.
373 Climate in the tropics is generally described as warm and humid, although higher altitudes
374 experience cooler temperatures. Besides, tropical landscapes span diverse ecosystems ranging
375 from desert, scrub woodlands to moist deciduous rainforests. Seasons across these diverse
376 landscapes are largely under the influence of both temperature and precipitation changes in the
377 tropics. Temperature fluctuations throughout the year are generally less pronounced in the
378 tropics in comparison to temperate regions (Janzen 1967). However, the warming and cooling
379 phases of El Nino and La Nina, respectively could affect intra-annual climate variability in the

380 tropics (Malhi and Wright 2004). Further, since tropical species have evolutionarily faced low
381 temperature fluctuations (Sheldon et al. 2018), a small degree of temperature rise could pose
382 challenges in terms of survival and fitness (Deutsch et al. 2008). Thus, climate mediated
383 temperature responses may strikingly differ in the lower latitudinal tropics versus temperate
384 regions at higher latitudes (Sheldon 2019).

385 Global warming is predicted to induce habitat shifts and lower altitude/ tropical
386 populations are less likely to be buffered to thermal extremes unless they shift poleward
387 (Kellermann et al. 2012) (but see (Overgaard, Kearney, and Hoffmann 2014)). The question of
388 how tropical organisms with narrowly adapted thermal ranges will cope with temperature
389 fluctuations imposed by rapid climatic change needs inputs from multiple perspectives. Our
390 study reports the genetic variation linked to TRGs in a tropical *Drosophila* species and thus
391 paves way for further research into thermal acclimation in tropical species.

392

393 While temperature may be the proximate cause influencing thermal acclimation, the
394 ultimate modulators are molecules sensing temperature changes. While adaptive response studies
395 test phenotypic outcomes (e.g. through thermal performance curves), studies correlating this
396 ability to adapt to the underlying genetic variation are still far and few. In recent years adaptive
397 responses to thermal changes have been reflected through the gene expression profiles too
398 (Kelly, 2019). Thus, selecting candidate genes responding putatively to thermal stress could
399 provide crucial cues in understanding thermal acclimation (Sørensen and Loeschke 2007). We
400 therefore aimed at targeting TRGs as candidates to check allelic variation across the growing
401 season under tropical conditions.

402 We used whole genome sequencing and filtered common SNPs for TRGs across the three
403 time-points (T1 to T3). Results demonstrate considerable allelic variation across the three time-
404 points. Temperature would increase from T1 to T2 and then decrease from T2 to T3. We
405 therefore surmise that allelic variation could execute through physiological plasticity in response
406 to temperature extremes in this tropical species. Higher allelic variation was observed in the first
407 two time-points in comparison to the third (T3), where temperature could plateau due to the
408 onset of monsoon. Further studies could determine if the observed allelic variation corroborates
409 positively with survival and fitness in the peak hot season. In temperate orchards, considerable
410 allelic variation with respect to seasons (winter versus fall) has been noted (Bergland et al. 2014)
411 maintained through balancing selection. However, in this case, considerable allelic variation for
412 temperature responsive genes was seen within a span of 90 days (corresponding to peak summer
413 season). It therefore remains to be analyzed if selection is indeed strong, relaxed or balancing
414 through seasonal oscillations like temperate counterparts.

415 Plasticity may incur costs (Krebs and Loeschke 1994). These costs along with other
416 constraints (e.g. biochemical) may limit thermal acclimation (Seebacher, White, and Franklin
417 2015) and should be further investigated. While the mechanistic basis of temperature regulating
418 TRGs is still being explored (Singh et al. 2019), our study which reveals novel SNP's in
419 *Drosophila* TRGs opens up new avenues for testing thermal acclimation hypotheses. The
420 pleiotropic effect of temperature sensation on other functions can be seen through the association
421 of different pathways integrated through the protein interaction map. Nonetheless, genes
422 associated are novel and not typical of those associated with thermal acclimation, e.g. heat shock
423 proteins (Sørensen and Loeschke 2007).

424 We found two independent lines of evidence: First, a large number of kinases and
425 phosphorylases are associated with stress, possibly linked to heat/temperature as evident from
426 the protein interaction map (Fig. 4). Second, the large number of key regulators and transcription
427 factors that are associated with cell proliferation and growth. Underlying these, *Myc* emerged as
428 a key gene with the interactants possibly associated with genetic arrangements, and most
429 importantly reducing the mutational load or burden. This is also in agreement with the fact that
430 the *Myc* plays a key role in insufficiency of haplotypes reducing the mutation load which further
431 is the key factor for extended lifespan or pro-aging (Greer et al. 2013; Morrison, Murakami, and
432 Cleghon 2000). On the other hand, phosphatidylinositol 5-phosphate 4-kinase (PIP4K) is known
433 to regulate the growth during fly development (Gupta et al. 2013). While its activity is
434 implicated in cellular responses, regulating the growth factors, the protein interaction networks
435 clearly reveal that there are transient mutations underpinning the changes in heat/temperature
436 stress as these changes could be because of the activity of downstream kinases. We envisage that
437 the pathways associated with PIP4K switch signals modulating the strength of a signal. Taken
438 together, the kinases and phosphatases contribute to the outcome or ability to acclimate of the
439 flies in these conditions. As they adapt, essential regulatory processes and changes mediating
440 phosphorylation and kinases are known. Further, our genomic sequencing has emerged as an
441 approach to understand such “adaptive loci” influencing these signaling pathways.

442 Mesocosm experiments as ours closely resemble natural conditions are hitherto reported
443 only in temperate populations (e.g. (Rudman et al. 2022b)) but not for tropical species. Ours is
444 therefore the first study to track genetic variation at more realistic and natural time-scales in a
445 tropical fruit fly species. We considered a time frame where summer temperature steadily
446 increases, peaks and then wanes. Our results from heat ramp assay corroborate this, in that

447 increase in temperatures of the external environment (cages) also altered the thermal sensitivity.
448 Thus, the peak in the temperature during T2 (Figure 1) also corresponded with flies from T2
449 being the most sensitive to thermal stress and hence the faster knock-down time in T2 compared
450 to T1 and T3 (Fig. 5). However, further studies could ascertain if seasonal variation in alleles
451 related to TRGs are connected/ pleiotropic with fitness-related traits. Thus, in the temperate
452 regions, diapause associated genes are known to be up and down regulated corresponding to
453 winter and fall seasonal variation, respectively (Zhao et al. 2015) with outcomes in fecundity
454 (ovariole size (Schmidt et al. 2005)). In tropical scenarios, where adaptations like diapause are
455 less known, one could expect adaptivity to thermal extremes to evolve through alternate gene
456 regulatory networks. Since *Drosophila melanogaster* originated from Africa (David and Capy
457 1988) and then dispersed to other continents, reaction norms for temperature tolerance could be
458 expected to be broadly similar across the tropics than across temperate regions. Surprisingly, our
459 study populations from higher altitudes (Rohru, western Himalayas) exhibited limited tolerance
460 to higher temperatures, thus re-affirming the speculation that tropical species are indeed sensitive
461 to higher temperature thresholds. It remains to be tested if differential response to temperature
462 extremes does exist across altitudinal populations. Given clinal variation for desiccation
463 resistance of Drosophilids from the Indian subcontinent (Rajpurohit et al., 2013), exploring
464 thermal tolerance and associated genomic variation is an interesting research avenue.

465 Seasonal changes in the tropics are also linked to rainfall. Compared to temperate
466 regions, tropics exhibit striking changes in humidity from ca. 50% (summer) to ca. 90% (rains)
467 across seasons. Few studies report the impact of temperature-humidity interactions on life history
468 traits in the tropics. Thus, high (not moderate) temperatures in conjunction with low relative
469 humidity have been shown to lower fecundity in fruit flies (Maurya et al. 2021). Higher

470 temperatures trigger up-regulation of genes related to desiccation tolerance (Rajpurohit et al.
471 2013). Flies under high desiccation stress displayed a lag in mating initiation time (Arya et al.,
472 2021). It is possible that the lowered fecundity of females at high temperatures could be due to
473 effect of temperature individually in the male sex (low sperm count) (David et al. 2005).
474 Temperature-humidity interactions could thus affect both mating behaviour and gametogenesis,
475 crucial for reproductive fitness. Sexually dimorphic role of TRGs through temperature is
476 therefore an interesting avenue for future work.

477

478 **Conclusions**

479 Tropical species are likely to be affected by global increase in temperature since the
480 thermal range to which they could respond is narrow compared to their temperate counterparts.
481 Our study hints that the narrow temperature tolerance of tropical species could be linked with
482 alternate molecular pathways for thermal sensitivity. We found a considerable number of newer
483 candidate alleles whose frequencies are significantly shifting as temperature increases over the
484 growing season. Our study highlights that genomic signatures could be corroborated with field
485 based studies to understand organismal responses to changing environmental conditions. Our
486 findings of allelic variation have implications not only in tracking fitness outcomes across
487 seasons but also in pest management. Future work could combine molecular and field based
488 approaches for a comprehensive understanding of climate mediated changes.

489

490 **Competing interests:** None to declare.

491

492 **Authors' contribution**

493 SR: Conceived the idea, Experiment design, Experimentation, Managing resources, and Writing
494 the manuscript.

495 PSS: Raw data processing, Reads collection, and Writing code

496 HM: Analyzing data, Writing the manuscript

497 HA: Data collection

498 RA: Figures, Reads collection

499 PS: Raw data processing, Reads collection, Writing code, Writing the manuscript

500 VL: Conceived the idea, Managing Resources, Writing manuscript

501

502 **Data Availability**

503 All supplementary files are deposited in the Dryad repository

504 <https://doi.org/10.5061/dryad.7pvmcvdw> .

505 Raw sequence reads of the genome will be made available through NCBI upon manuscript

506 acceptance or upon reviewer request

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