Dissection of complex, fitness-related traits in multiple *Drosophila* mapping populations offers insight into the genetic control of stress resistance

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Abstract

We leverage two complementary *Drosophila melanogaster* mapping panels to genetically dissect starvation resistance, an important fitness trait. Using >1600 genotypes from the multiparental *Drosophila* Synthetic Population Resource (DSPR) we map numerous starvation stress QTL that collectively explain a substantial fraction of trait heritability. Mapped QTL effects allowed us to estimate DSPR founder phenotypes, predictions that were correlated with the actual phenotypes of these lines. We observe a modest phenotypic correlation between starvation resistance and triglyceride level, traits that have been linked in previous studies. However, overlap among QTL identified for each trait is low. Since we also show that DSPR strains with extreme starvation phenotypes differ in desiccation resistance and activity level, our data imply multiple physiological mechanisms contribute to starvation variability. We additionally exploited the *Drosophila* Genetic Reference Panel (DGRP) to identify sequence variants associated with starvation resistance. Consistent with prior work these sites rarely fall within QTL intervals mapped in the DSPR. We were offered a unique opportunity to directly compare association mapping results across labs since two other groups previously measured starvation resistance in the DGRP. We found strong phenotypic correlations among studies, but extremely low overlap in the sets of genomewide significant sites. Despite this, our analyses revealed that the most highly-associated variants from each study typically showed the same additive effect sign in independent studies, in contrast to otherwise equivalent sets of random variants. This consistency provides evidence for reproducible trait-associated sites in a widely-used mapping panel, and highlights the polygenic nature of starvation resistance.

Introduction

Periods of food scarcity and suboptimal nutrient resources present an important challenge for nearly all species (McCue 2010), and this form of environmental stress can limit the survival of individuals with poor nutritional status and reduced stress resistance (Harshman *et al.* 1999; Lee and Jang 2014). As a result, starvation stress resistance has direct implications for the fitness of individuals as they experience resource variability in natural populations. Starvation resistance is a classic quantitative, fitness-related trait that is associated with several other phenotypes that influence survival, lifespan, and reproduction (Service and Rose 1985; Da Lage *et al.* 1990; Rose *et al.* 1992; Toda and Kimura 1997; Karan and Parkash 1998; Hoffmann *et al.* 2005b; Sørensen *et al.* 2007; Lee and Jang 2014). In particular, increased starvation resistance is often negatively correlated with fecundity and positively correlated with longevity, energy storage, and other stress response traits (Service *et al.* 1985; Rose *et al.* 1992; Hoffmann and Parsons 1993; Chippindale *et al.* 1993; Harshman *et al.* 1999; Bochdanovits and de Jong 2003; Bubliy and Loeschcke 2005; Sørensen *et al.* 2007; Schwasinger-Schmidt *et al.* 2012; Lee and Jang 2014). Because of pervasive phenotypic and genetic correlations between starvation resistance and these other traits, the evolution of starvation resistance in natural populations is complex, and is driven by adaptation to heterogeneous environments, phenotypic plasticity, and extensive pleiotropy (Service and Rose 1985; Hoffmann and Parsons 1991, 1993; Chippindale *et al.* 1993; Toda and Kimura 1997; Karan and Parkash 1998; Harshman *et al.* 1999; Harbison *et al.* 2004; Pijpe *et al.* 2007; Rion and Kawecki 2007; Bauerfeind *et al.* 2014; Colinet *et al.* 2015; Everman and Morgan 2018).

Artificial selection for starvation resistance often results in a concomitant increase in desiccation resistance (Hoffmann and Parsons 1989a; Chippindale *et al.* 1996; Hoffmann and Harshman 1999; Harshman *et al.* 1999; Hoffmann *et al.* 2001), and selection specifically on desiccation resistance can also result in a corresponding rapid increase in starvation resistance (Hoffmann and Parsons 1989b). Nonetheless, surveys of natural populations in several *Drosophila* species have shown that starvation and desiccation resistance can also independently vary (Davidson 1990; Karan and Parkash 1998; Karan *et al.* 1998; Chippindale *et al.* 1998; Hoffmann and Harshman 1999; Gilchrist *et al.* 2008; Goenaga *et al.* 2013). Given these variable

94 patterns, Karan and Parkash (1998) and Da Lage *et al.* (1990) suggest that desiccation and
95 starvation resistance may not routinely be associated. Rather, both traits may be directly and
96 indirectly influenced by climate variability, and selection on other correlated traits such as
97 diapause or thermal tolerance in seasonally variable temperate environments (Hoffmann and
98 Parsons 1989b; Schmidt *et al.* 2005; Sørensen *et al.* 2007; Rion and Kawecki 2007; Goenaga *et al.*
99 2013; Rajpurohit *et al.* 2018).

100 Similar to desiccation, artificial selection for increased starvation resistance often results
101 in an increase in lipid levels in *D. melanogaster* (Chippindale *et al.* 1996, 1998; Djawdan *et al.*
102 1998; Harshman *et al.* 1999; Schwasinger-Schmidt *et al.* 2012; Goenaga *et al.* 2013; Hardy *et al.*
103 2018), suggesting that energy storage is one important mechanism that contributes to starvation
104 resistance. However, variation in the association between these traits has also been observed.
105 For example, while Chippindale *et al.* (1996) provided evidence of a strong positive correlation
106 between starvation resistance and lipid concentration following 60 generations of selection for
107 starvation resistance, Hoffmann *et al.* (2001) found that total lipid concentration and starvation
108 resistance in isofemale lines derived from natural populations were not correlated. Thus, the
109 association between starvation resistance and lipid level is likely dependent upon genetic
110 background and the evolutionary history of a population, resulting in across-population variation
111 in the strength and direction of the correlation between these traits.

112 Genetic dissection of starvation resistance can both lead to the identification of loci
113 impacting phenotypic variation and help understand how this trait is associated with desiccation
114 resistance and lipid level. Several studies have examined the genetic basis of starvation resistance
115 in *D. melanogaster* using a combination of selection experiments (Rose *et al.* 1992; Chippindale
116 *et al.* 1996; Harshman *et al.* 1999; Bochdanovits and de Jong 2003; Bubliy and Loeschcke 2005;
117 Schwasinger-Schmidt *et al.* 2012; Hardy *et al.* 2018; Michalak *et al.* 2018), gene expression studies
118 following exposure to starvation stress (Harbison *et al.* 2005; Sørensen *et al.* 2007), and genetic
119 mapping (Harbison *et al.* 2004; Mackay *et al.* 2012; Huang *et al.* 2014; Everman and Morgan
120 2018). These studies have provided extensive lists of candidate genes and variants, some of which
121 have been functionally validated (Lin *et al.* 1998; Clancy *et al.* 2001; Harbison *et al.* 2004, 2005;
122 Sørensen *et al.* 2007). However, up to this point few studies have undertaken an examination of

123 the genetic architecture of triglyceride or lipid content in the same genetically diverse panel used
124 to examine variation in starvation resistance. Doing so would allow a detailed comparison of
125 quantitative trait loci (QTL) that contribute to variation in each trait, provide insight into the
126 similarity of the genetic architectures of starvation resistance and correlated traits, and facilitate
127 a better understanding of their evolution, and the mechanisms underlying their variation.

128 In this study we use two powerful *D. melanogaster* mapping panels - the *Drosophila*
129 Synthetic Population Resource (DSPR) and the *Drosophila* Genetic Reference Panel (DGRP) - to
130 genetically dissect phenotypic variation, and to explore the phenotypic and genetic relationships
131 among traits, among mapping panels, and among laboratories. Our approach allows us to
132 accomplish three primary objectives. First, by measuring starvation resistance and triglyceride
133 level in the DSPR, we assess overlap in the loci that contribute to variation in each trait. Prior
134 work on these traits in flies suggests they would show similar genetic architectures with many
135 pleiotropic loci. However, despite a significant phenotypic correlation, we report limited overlap
136 among mapped loci contributing to variation in starvation resistance and triglyceride level,
137 suggesting that the genetic basis of these traits is largely independent in the DSPR. This highlights
138 the role that other physiological mechanisms, such as activity level and desiccation resistance
139 that also we explore here, may have in influencing starvation resistance.

140 Second, by measuring starvation resistance in both the DSPR and the DGRP under the
141 same environmental conditions, we address variation in the genetic architecture of this trait
142 between two distinct populations. In common with some previous studies using both panels to
143 dissect a trait (e.g. Najarro *et al.* 2015, 2017), we also find little overlap in the loci associated with
144 starvation resistance between mapping panels. This is likely the combined result of the
145 populations having unique genetic backgrounds (King and Long 2017), distinct evolutionary
146 histories, and differences in power to detect causative loci (Long *et al.* 2014).

147 Third, we leverage the ability to repeatedly measure trait variation on the same, stable
148 set of inbred genotypes to compare our DGRP starvation data to two previous starvation
149 resistance datasets collected by different laboratories (Mackay *et al.* 2012; Everman and Morgan
150 2018). We found that the sign of the additive effects of the most strongly-associated SNPs were
151 consistent across datasets. This suggests these SNPs contribute to variation in starvation

152 resistance in the DGRP, but have sufficiently small effects that they are regularly not identified
153 following genomewide multiple testing correction. This across-study comparison of the genetic
154 architecture of starvation resistance provides both technical insight into the use of genomewide
155 association (GWA) studies to understand the genetic basis of complex traits, and biological insight
156 into the phenotypic effects of loci that contribute to trait variation.

157 Materials and Methods

158

159 Mapping populations

160 Drosophila *Synthetic Population Resource*

161 The DSPR is a multiparental population that consists of two synthetic populations (pA and
162 pB) that were each established following an intercross of eight highly-inbred founder lines, with
163 one founder line shared between the two populations (King *et al.* 2012a). Flies were maintained
164 in pairs of subpopulations (pA1, pA2, pB1, pB2) at high population density for 50 generations
165 prior to the establishment of >1600 genotyped recombinant inbred lines (RILs) via 25 generations
166 of full-sib inbreeding (King *et al.* 2012a; b). Founder lines for the pA and pB panels have also been
167 sequenced at 50x coverage, enabling inference of the haplotype structure of each RIL via a hidden
168 Markov model (described in King *et al.* (2012a)).

169

170 Drosophila *Genetic Reference Panel*

171 The DGRP was established from mated females collected from a natural population in
172 Raleigh, North Carolina, with inbred lines derived from 20 generations of full-sib mating (Mackay
173 *et al.* 2012). Each of the 205 DGRP lines have been re-sequenced and genotyped allowing GWA
174 mapping to be carried out in the panel (Mackay *et al.* 2012; Huang *et al.* 2014).

175

176 Phenotyping assays and analysis

177 *Large-scale starvation resistance assay*

178 Strains from the DSPR and DGRP were duplicated from stocks, and flies were allowed to
179 lay eggs for up to 2 days. Vials were inspected twice daily, and laying adults were cleared when
180 necessary to maintain a relatively even egg density across experimental vials. While this visual

181 method of density control is less precise than counting eggs, experiments with 20 randomly-
182 selected DSPR RILs showed that the effect on starvation resistance of rearing flies via egg
183 counting or by visually-assessing egg number is extremely limited (variance explained = 0.9%; Fig
184 S1; see Table S1 for full breakdown of variance components).

185 In the following generation, experimental flies (2-4 days old) were sorted by sex over light
186 CO_2 anesthesia and placed in groups of same-sex individuals on new cornmeal-molasses-yeast
187 media for 1 day until the start of the starvation assay. The assay was initiated by placing flies on
188 1.5% agar media that additionally contained preservatives - a mix of propionic and phosphoric
189 acids, and benzoic acid (Tegosept, Genesee Scientific) dissolved in ethanol (see starvation media
190 recipe Text S1). Starvation media was made within 24 hours of the initiation of each block of the
191 assay and was not replaced throughout its duration. Vials were barcoded during the screen,
192 blinding experimenters to strain identification number, and assisting with efficient data collection
193 and analysis.

194 Death was assessed for each vial twice per day at approximately 0900 and 2100 hrs. The
195 first assessment of survival was made 24 hours after flies were transferred to starvation media.
196 Dead flies at this initial assessment point were not included in the analysis as their death may
197 have resulted from handling during the initial transfer to experimental vials rather than
198 starvation stress. Vials containing flies that had become entangled in the cotton vial plug at any
199 point during the assay were also excluded from the analysis. Flies were considered dead if they
200 were not moving or were unable to dislodge themselves from the starvation media. The
201 phenotype used for mapping was the mean time to death in hours per strain across the vial
202 replicates. Flies for this screen were reared and tested at approximately 23°C, 30-60% humidity,
203 with constant light.

204 We screened the DSPR (861 pA1/pA2 and 864 pB1/pB2 RILs) in a series of batches across
205 a seven-month period in 2010. Each batch included the majority of RILs that belonged to a
206 particular subpopulation. Starvation resistance was measured in 168 DGRP lines in a single batch
207 in 2012. In both mapping panels, survival was measured across 2 vial replicates per sex in ~85%
208 of strains, with ~90% of vials containing 10 flies (minimum flies per vial = 6). Finally, we measured

209 starvation resistance in the 15 DSPR founder lines, using 5 vial replicates per founder, in one
210 batch.

211 We assessed variation in starvation resistance due to subpopulation and sex in the DSPR
212 with a two-way ANOVA, including the interaction, and treated subpopulation (pA1, pA2, pB1,
213 pB2) and sex as fixed factors. Male and female-specific differences among the four
214 subpopulations were tested using Tukey's HSD post hoc comparisons with an experiment-wide
215 $\alpha = 0.05$. Differences in starvation resistance due to sex among the DGRP lines were analyzed
216 with a one-way ANOVA, treating sex as a fixed factor.

217

218 *Desiccation resistance assay*

219 To investigate the correlation between starvation and desiccation resistance, we
220 measured desiccation resistance in a subset of pA1/pA2 RILs that exhibited very low (17 RILs) or
221 very high (16 RILs) average female starvation resistance in the large-scale screen. Desiccation
222 resistance of female flies from all 33 strains was assessed in a single batch with two vial replicates
223 per RIL, where 92.9% of vials contained 10 flies (minimum flies per vial = 8). We placed
224 experimental flies, reared as described above, in empty vials plugged with cotton inside an
225 airtight desiccator (Cleatech, LLC). Relative humidity was reduced to < 5% throughout the
226 experiment by adding a large quantity of Drierite (calcium sulfate) to the chamber. Survival was
227 assessed every hour following initiation of the experiment, and mean desiccation resistance per
228 RIL was used in all analyses.

229

230 *Activity assay*

231 We employed the *Drosophila* Activity Monitoring System (DAM2, TriKinetics, Inc.) to
232 assess activity levels both prior to, and during starvation for a subset of DSPR RILs, selecting 16
233 (19) pA1/pA2 RILs that exhibited high (low) female starvation resistance in the large-scale screen.
234 Sixteen flies of each sex were tested per RIL. Flies for these assays were reared and tested at
235 25°C, 50% relative humidity, with a 12:12hr light:dark photoperiod. These environment
236 conditions are different from our large-scale screen, but in line with those used in previous
237 starvation resistance studies in *D. melanogaster* (Mackay *et al.* 2012; Everman and Morgan

238 2018). This change allowed us to examine the stability of DSPR starvation phenotypes across
239 assay environments.

240 One day prior to adding flies to monitor tubes, cornmeal-yeast-dextrose media was
241 poured into 100mm diameter petri dishes and allowed to set. Polycarbonate activity monitor
242 tubes (5mm diameter x 65mm length) were filled to approximately 10mm by pushing them into
243 the media, and the food plug in each tube was sealed with paraffin wax. A single fly was aspirated
244 into each tube, and the tubes were capped with small pieces of Droso-Plugs (Genesee Scientific).
245 Flies were allowed to acclimate to the tubes for 24 hours, and then we measured activity for the
246 next 24 hours under non-stressful conditions. Subsequently, each fly was tipped to a second
247 monitor tube containing starvation media (Text S1) and activity was continuously monitored until
248 each fly died.

249 To determine differences in activity under non-stressful conditions due to starvation
250 resistance rank (high versus low), we used a full three-way ANOVA model with interactions, and
251 treated starvation rank, sex, and light status (light versus dark) as fixed effects. The effect size of
252 the main effects and interactions were calculated using Cohen's F, which determines the effect
253 size as a ratio of the between-group and between-replicate standard deviations (R package:
254 sjstats) (Cohen 1988; Quinn and Keough 2002; Lüdecke 2018).

255

256 *Triglyceride level assay*

257 We duplicated 311 pA1/pA2 and 628 pB1/pB2 DSPR RILs from stocks to two replicate
258 vials, clearing parental flies when necessary to maintain relatively even egg density over test vials.
259 Eleven days following the start of egg laying we collected two sets of 10-12 females from each
260 parental vial, resulting in four collection vials from each RIL. Flies were aged for three additional
261 days before measuring triglyceride level.

262 Experimental females from each collection vial were anesthetized using CO_2, and groups
263 of 5 were arrayed into deep well plates (Axygen, P-DW-11-C) over ice, with each well pre-loaded
264 with a single glass bead. This resulted in 8 replicate samples of 5 females per RIL. Immediately
265 after finishing a plate, we added 400µl of cold homogenization buffer (10mM potassium
266 phosphate monobasic, 1mM EDTA, 0.05% Tween-20) to each well, homogenized for 45sec in a

267 Mini-BeadBeater-96 (Bio Spec Products, Inc.), and centrifuged for 4min at 2,500g. We then
268 moved 50µl of the supernatant to a standard PCR plate, incubated the plate in a thermocycler at
269 70°C for 5min, and then placed the plate on ice for 5min.

270 During the incubation steps, we added 30µl of homogenization buffer to 92 of the 96
271 wells of a flat-bottom, polystyrene assay plate (Greiner, 655101), and subsequently added 20µl
272 of the heat-deactivated fly homogenate to each. The four remaining wells of every assay plate
273 were dedicated to controls; one blank well contained 50µl of homogenization buffer only, and
274 three wells contained 5µl of Glycerol Standard Solution (SigmaAldrich, G7793, 2.5mg/ml) along
275 with 45µl of homogenization buffer.

276 The assay plate was then inserted into a BioTek Powerwave XS2 instrument pre-heated
277 to 37°C and read at 540nm (baseline absorbance scan). After the scan, and within 10min, we
278 added 100µl of Free Glycerol Reagent (SigmaAldrich, F6428) to each well. The plate was then re-
279 inserted into the instrument, incubated at 37°C for 5min, and read again at 540nm (free glycerol
280 absorbance scan). After this second scan, and again within 10min, we added 25µl of Triglyceride
281 Reagent (SigmaAldrich, T2449) to each well. The plate was again incubated at 37°C for 5min in
282 the machine and read for a third time at 540nm (triglyceride, or final absorbance scan).

283 For each sample, we obtained the final absorbance for each sample (FA_{sample}) and
284 calculated the initial absorbance (IA_{sample}) as the free glycerol measurement minus the baseline
285 measurement. We also generated the average final absorbance for the three standard wells
286 (FA_{std}) and the initial absorbance for the one blank well (IA_{blank}). We then estimated the true
287 serum triglyceride level as

288

289 $$\left[FA_{sample} - \left(IA_{sample} \times F\right)\right] / \left[FA_{std} - \left(IA_{blank} \times F\right)\right]$$

290

291 where F = 0.8. We then multiplied this value by the concentration of the glycerol standard
292 solution (2.5mg/ml) and used the average value across all 8 replicate samples as the RIL mean
293 triglyceride level for mapping and analysis. For precise details of the enzyme assay and
294 triglyceride calculation, see the SigmaAldrich Serum Triglyceride Determination kit product insert
295 (TR0100). Differences in triglyceride level due to DSPR subpopulation were investigated with a

296 one-way ANOVA followed by *post hoc* comparisons using Tukey's HSD (experiment-wide α =
297 0.05).
298

299 ## Correlations among traits

300 We assessed the relationship between the DSPR RIL mean starvation phenotypes from
301 the large-scale screen with those from the activity monitor experiment, the desiccation
302 resistance measures, and triglyceride level using general linear models. Subpopulation (pA1, pA2,
303 pB1, pB2) was included as a factor in the analysis examining starvation resistance and triglyceride
304 content.

305 Correlations among three DGRP starvation datasets - that from Mackay et al. (2012),
306 Everman and Morgan (2018), and the new screen we report here - were examined in a pairwise
307 manner using line means, accounting for multiple comparisons with a Bonferroni-adjusted alpha
308 level. Differences in the overall mean starvation resistance among the three datasets were
309 analyzed with a one-way ANOVA, treating study as a fixed factor.
310

311 ## Heritability

312 The genetic and phenotypic variances of starvation resistance and triglyceride content for
313 the pA and pB DSPR panels, and of starvation resistance for the DGRP panel, were estimated with
314 a linear mixed model using the lme and varcomp functions in R (R package: APE, Paradis *et al.*
315 2004; R package: nlme, Pinheiro *et al.* 2017). We calculated broad-sense heritability as the
316 proportion of the total variance of the strain-specific response explained by the estimated
317 genetic variance component (Lynch and Walsh 1998).
318

319 ## QTL mapping in the DSPR

320 Methods for QTL analysis, and the power and resolution of mapping using the DSPR panel
321 are discussed in detail in King *et al.* (2012a,b). Briefly, QTL mapping and peak analysis were
322 executed for starvation resistance and triglyceride data using the DSPRqtl R package
323 (github.com/egking/DSPRqtl; FlyRILs.org), regressing the mean trait response for each RIL on the
324 additive probabilities that each of the 8 founders contributed the haplotype of the RIL at each

325 mapped position. Significance thresholds were assigned following 1000 permutations of the data,
326 and positions of putative causative loci were estimated with 2-LOD support intervals, which
327 approximate 95% confidence intervals for QTL position in the DSPR (King *et al.* 2012a). Mean
328 starvation resistance varied between sexes in both the pA and pB panel ($F_{3,3440}$ = 18.317; p <
329 0.0001; Fig S2; Table S2), and subpopulation influenced female starvation resistance in the pA
330 panel (Tukey's HSD p < 0.0001; Fig S2). Therefore, QTL mapping was performed for males and
331 females of each panel separately, and subpopulation was included as a covariate in the analysis
332 of pA females. Mean female triglyceride level was similar between the pA1 and pA2
333 subpopulations (Tukey's HSD p = 0.75; Fig S3; Table S3) but varied between the pB1 and pB2
334 subpopulations (Tukey's HSD p < 0.0001; Fig S3; Table S3), so subpopulation was included as a
335 covariate in QTL analysis of the pB triglyceride data.
336

337 Analysis of DGRP starvation data

338 Variants associated with male and female starvation resistance in the DGRP were
339 identified using the DGRP2 web-based GWA mapping tool (http://dgrp2.gnets.ncsu.edu), which
340 takes into account variable *Wolbachia* infection status and large inversions that segregate among
341 the lines (Mackay *et al.* 2012; Huang *et al.* 2014). We performed GWA analysis on data collected
342 in this study and additionally reanalyzed starvation data from Mackay *et al.* (2012) and starvation
343 of young flies (5 – 7 days old) from Everman and Morgan (2018). We additionally assigned the
344 150 DGRP lines that are shared between the three datasets an across-study mean and performed
345 GWA analysis on this summary measure of starvation resistance.

346 SNPs associated with starvation resistance were identified within each of the four
347 datasets following FDR correction for multiple comparisons (Benjamini and Hochberg 1995) in R
348 (p.adjust; R Core Team 2017). Since we found no significantly associated SNPs with an FDR
349 adjusted p-value < 0.05 for any starvation resistance dataset in either sex, we relaxed the
350 significance threshold to an FDR adjusted p-value < 0.2. As a significance threshold of $P < 10^{-5}$ is
351 commonly used in the DGRP (e.g. Mackay *et al.* 2012; Morozova *et al.* 2014; Huang *et al.* 2014;
352 Everman and Morgan 2018), we also present variants associated with starvation resistance in
353 each of the four datasets using this threshold.

354 There was minimal overlap in the identity of the above-threshold, starvation-associated
355 variants in each study. Thus, we sought to examine whether the sign of the additive effects of
356 these sets of variants was preserved across studies. Additive effects were calculated as one-half
357 the difference in starvation resistance between lines homozygous for the major allele and lines
358 homozygous for the minor allele (major allele frequency > 0.5), after accounting for *Wolbachia*
359 infection and TE insertions (Falconer and Mackay 1996; Huang *et al.* 2014). To determine the
360 proportion of SNPs that are expected by chance to have additive effects of the same sign across
361 studies, we obtained random samples of 50 SNPs from all of the DGRP SNP calls (~ 2 million SNPs)
362 and calculated the additive effects of the sampled SNPs across pairs of datasets for each sex. To
363 account for the possibility that the frequency spectrum of above-threshold ($P < 10^{-5}$), associated
364 SNPs is not represented by a set of randomly-selected variants, we stratified the random subsets
365 of 50 SNPs according to the distribution of allele frequencies of the top 50 SNPs associated with
366 starvation resistance for each sex in each study. Allele frequency bins used in this stratification
367 were 0.05 – 0.1, > 0.1 – 0.2, > 0.2 – 0.3, > 0.3 – 0.4, and > 0.4 – 0. 5. The exact stratification for
368 each sex and dataset is provided in Table S4. This process was repeated 1000 times for each
369 paired comparison of datasets (6 comparisons total) using an ordinary nonparametric
370 bootstrapping procedure with the R package boot (Davison and Hinkley 1997; Canty and Ripley
371 2017). For each iteration, we used a custom R function (see File S1) to calculate the proportion
372 of the 50 random stratified SNPs that had positive additive effects in both of the datasets being
373 compared.
374

375 ## Data availability

376 DGRP data from Mackay *et al.* (2012) are available online from
377 http://dgrp2.gnets.ncsu.edu, and DGRP data from Everman and Morgan (2018) are available
378 from Dryad (DOI: https://doi.org/10.5061/dryad.vq087). Data collected in this study is available
379 from Dryad (available upon acceptance), including all raw data for starvation resistance in the
380 DSPR and DGRP, raw desiccation resistance, triglyceride level, and activity data collected using
381 the DSPR, and all mapping results (see File S2). R code for bootstrapping analysis and additive
382 effect calculations in the DGRP is available in File S1.

383 ## Results and Discussion

384 ### Extensive phenotypic variation in starvation resistance in the DSPR and DGRP

385 Starvation resistance in both the DSPR and DGRP was highly variable among strains (Fig
386 1, Fig 2), and the broad sense heritability for starvation resistance was routinely high (Table 1).
387 Males were typically less starvation resistant than females (Fig S2, Fig S4, Table S2, Table S5),
388 although despite this male and female starvation resistance were significantly correlated in both
389 the DSPR (pA: R^2 = 53.0%; pB: R^2 = 57.0%; Fig S5) and DGRP (R^2 = 68.0%; Fig S6). Such sex-specific
390 differences in starvation resistance are likely influenced by a combination of higher glycogen and
391 triglyceride levels and larger body size, which are often observed for females relative to males
392 (Chippindale *et al.* 1996; Toda and Kimura 1997; Schwasinger-Schmidt *et al.* 2012; Goenaga *et al.*
393 2013).

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395
396
397 ***Table 1. Broad sense heritability for starvation resistance and triglyceride level.***

Panel	Trait	Sex	Heritability
DSPR pA	Starvation	Female	80.0%
DSPR pA	Starvation	Male	73.1%
DSPR pB	Starvation	Female	74.4%
DSPR pB	Starvation	Male	71.5%
DGRP	Starvation	Female	87.1%
DGRP	Starvation	Male	87.0%
DSPR pA	Triglycerides	Female	77.5%
DSPR pB	Triglycerides	Female	82.3%

398

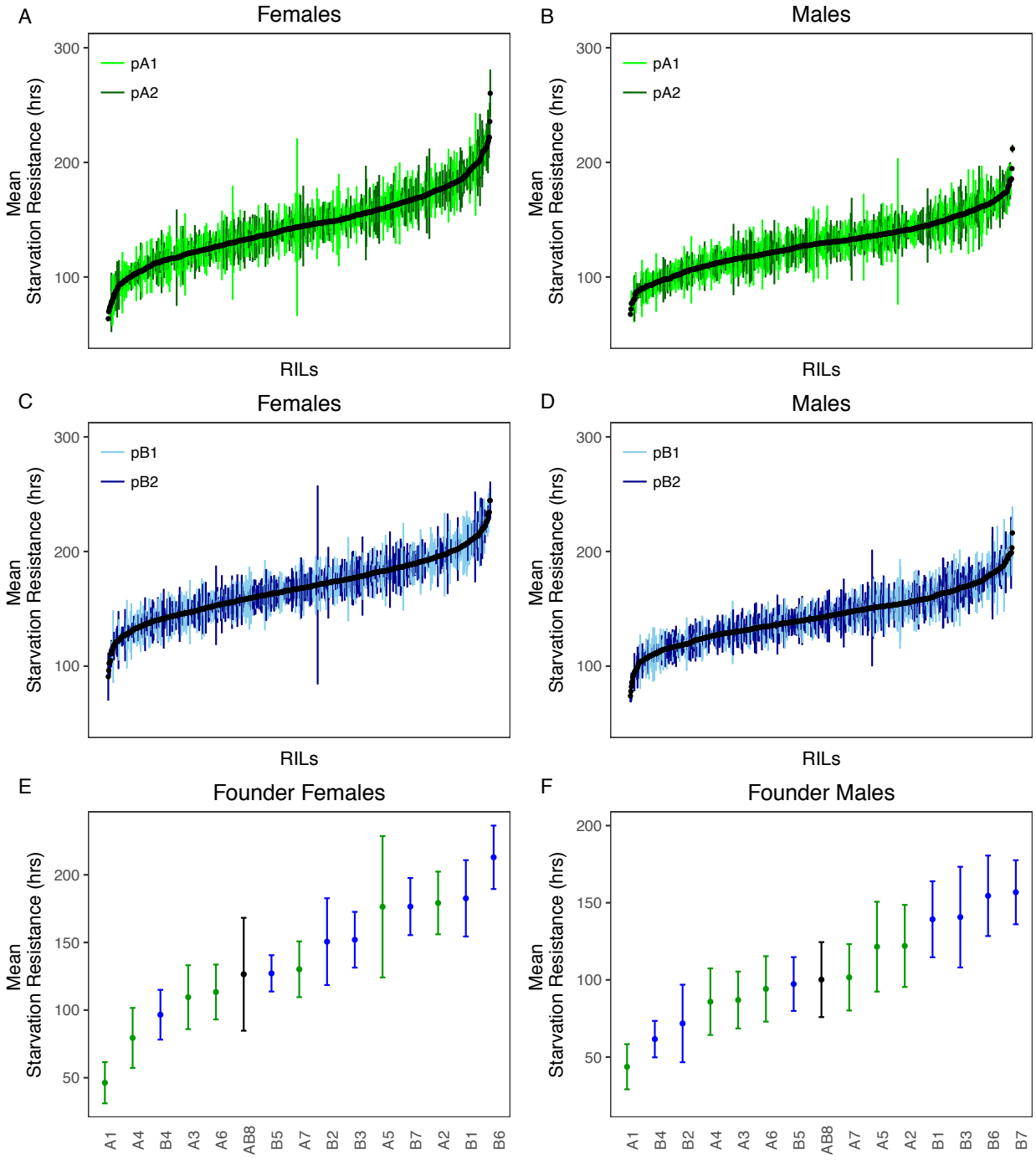
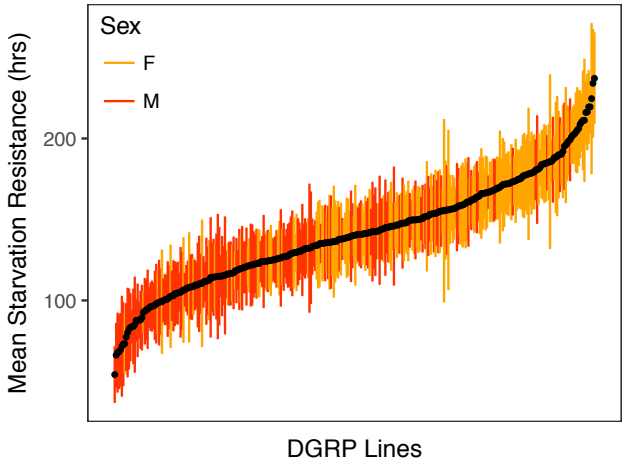


Figure 1. *Variation in mean (±SD) starvation resistance for each sex. A-D shows data for DSPR RIL panels pA and pB. E and F show data for the founder lines. In E and F, names of the founder lines are shown on the x-axis; the founder line AB8 is the founder shared by the two mapping panels.*



403

404 ***Figure 2.*** *Variation in mean DGRP (±SD) starvation resistance in females (orange) and males (red).*

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406
407

408 We screened the DSPR and DGRP for starvation resistance at 23°C and under constant
409 light conditions. Since starvation resistance is sensitive to the thermal environment (van
410 Herrewege and David 1997; Karan and Parkash 1998; Karan *et al.* 1998; Hoffmann *et al.* 2005a;
411 Bauerfeind *et al.* 2014) and may vary under different photoperiods (Sheeba *et al.* 2000; Xu *et al.*
412 2008; Seay and Thummel 2011), we sought to re-measure starvation resistance for a subset of
413 DSPR RILs at 25°C and with a 12 hour : 12 hour light/dark cycle, conditions that have been used
414 in other starvation studies (e.g. Mackay *et al.* 2012; Everman and Morgan 2018). Overall,
415 starvation resistance of the re-tested RILs was lower in both sexes compared to that measured
416 in the original large-scale starvation assay (effect of assay: $F_{1,136} = 31.60$, $p < 0.0001$; Fig S7A).
417 Despite this, starvation resistance in the subset of RILs was significantly correlated between the
418 two experiments (Females: $\beta = 0.43 \pm 0.04$, $t = 9.7$, $p < 0.0001$, $R^2 = 73.9\%$; Males: $\beta = 0.59 \pm 0.05$,
419 $t = 10.9$, $p < 0.0001$, $R^2 = 78.3\%$; Fig S7B).

420 We similarly compared starvation resistance phenotypes for the DGRP measured in the
421 current study with data generated by Mackay *et al.* (2012) and Everman and Morgan (2018). In
422 our study, the DGRP exhibited considerably higher resistance than in these previous works ($F_{2,532}$
423 $= 1457.5$, $p < 0.0001$; Fig S8). This discrepancy was not due to differences across studies in the
424 frequency with which flies were counted (every 4, 8, or 12 hours depending on the study, Fig S8).
425 To investigate whether the difference was due to the environmental conditions experienced by

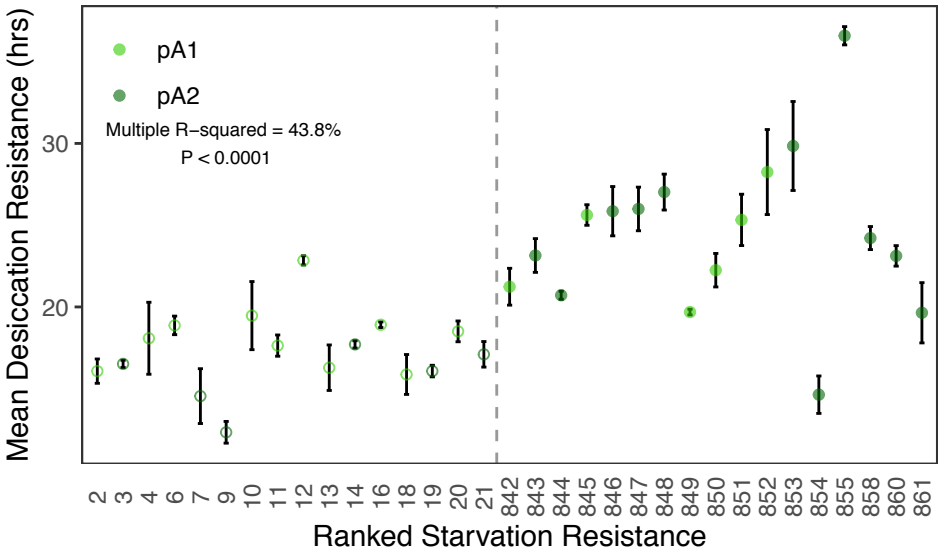
426 the experimental animals, we raised and tested 12 randomly-selected DGRP lines under the same
427 conditions as described for our initial screen (i.e., 23°C, 30-60% relative humidity, and constant
428 light) and under conditions that more closely mimic those described in Mackay *et al.* (2012) and
429 Everman and Morgan (2018) (i.e. 25°C, 50% relative humidity, and 12:12hr light:dark).
430 Furthermore, for both environments, we assayed starvation on agar media containing
431 preservatives (see Text S1), and on media lacking preservatives, as used by Everman and Morgan
432 (2018) and Mackay *et al.* (2012). The inclusion of preservatives in the assay media had the largest
433 effect on variation in starvation resistance among studies (Preservatives: $F_{1,327}$ = 1628.9, p <<
434 0.0001; variance explained = 81.2%; Fig S9), with rearing/testing environment explaining very
435 little of the variation (see Table S6 for the full breakdown of ANOVA variance components). We
436 speculate that the antibiotic properties of the preservatives extend lifespan under starvation
437 conditions by limiting growth of pathogenic microorganisms.
438 Even given the large across-study difference in mean starvation resistance in the DGRP,
439 we found moderately strong correlations in both sexes over datasets, ranging from 50.8% to
440 64.4% (Fig S10). The high correspondence among these three DGRP datasets, coupled with the
441 phenotypic correlation between the subset of DSPR strains assayed using two different
442 approaches (see above), suggests that fundamental aspects of the genetic control of starvation
443 resistance are generally consistent over experiments, even when environmental conditions such
444 as temperature are quite different. The differences we observe in starvation resistance between
445 studies may reflect ecologically-relevant phenotypic plasticity. The temporally variable thermal
446 environment is a particularly important source of selection for ectothermic organisms (Bell 2010;
447 Bergland *et al.* 2014). Plastic shifts in starvation resistance in response to temperature can have
448 important fitness benefits, including seasonal adaptation to fluctuating resource availability as
449 has been reported in the butterfly *Bicyclus anynana* (Pijpe *et al.* 2007) and following the induction
450 of diapause in *D. melanogaster* (Schmidt *et al.* 2005; Rion and Kawecki 2007). Collectively, these
451 previous studies and our data speak to the important influence of both phenotypic plasticity and
452 genotype on variation in starvation resistance in natural populations.
453

454 Starvation resistance is associated with desiccation resistance and low activity in the DSPR

455 Environmental stress can exert selection pressure on energy use and storage, and
456 environmental stressors that impact one type of stress resistance often impact a suit of other
457 stress-related traits (Hoffmann and Parsons 1989b). Several artificial selection studies for
458 starvation resistance have shown a correlated change in desiccation resistance, suggesting these
459 stress traits are related (Hoffmann and Parsons 1989a; b; Chippindale *et al.* 1996; Hoffmann and
460 Harshman 1999; Harshman *et al.* 1999; Hoffmann *et al.* 2001). For instance, a detailed study of
461 this correlated response by Hoffmann and Parsons (1989b) demonstrated a rapid phenotypic
462 response in both desiccation and starvation resistance following four generations of strong
463 selection for increased desiccation resistance, and in part this was attributed to selection acting
464 on a general stress response mechanism. Subsequent genomics studies have suggested that this
465 rapid phenotypic response is accompanied by rapid and extensive genomic change (Kang *et al.*
466 2016), and that extensive pleiotropy underlies desiccation resistance (Telonis-Scott *et al.* 2012,
467 2016; Kang *et al.* 2016; Griffin *et al.* 2017).

468 We investigated the association between starvation and desiccation resistance in the
469 DSPR by measuring female desiccation resistance in RILs chosen from the two tails of the
470 phenotypic distribution of female starvation resistance. We found that desiccation and starvation
471 resistance were significantly correlated (R^2 = 43.8, $F_{1,31}$ = 24.11, $p < 0.0001$; Fig 3). Since mean
472 desiccation resistance was considerably lower than mean starvation resistance for all lines tested
473 (compare Figs 1 and 3), flies experiencing desiccation conditions are unlikely to be dying from
474 starvation. In addition, since DSPR lines with very low starvation resistance do not also have low
475 larval viability (data from Marriage *et al.* 2014) or reduced adult lifespan (data from Highfill *et al.*
476 2016) it does not appear that DSPR lines with very low resistance to starvation and desiccation
477 are simply "sick" (Fig S11). The relationship between starvation and desiccation resistance in the
478 present study provides support for the genetic correlation and shared physiological mechanisms
479 that have been proposed to exist between these traits (Hoffmann and Parsons 1989a; b, 1993;
480 Harshman *et al.* 1999; Kennington *et al.* 2001). However, the correlation we observed is modest,
481 and does not rule out the possibility that the covariation observed between starvation and
482 desiccation resistance may be influenced by genetic variation in one or more other resistance-

483 associated traits. A more intensive sampling of the DSPR would be necessary to thoroughly
484 investigate the genetic correlation between starvation and desiccation resistance.



485

486 ***Figure 3.*** *Mean starvation and desiccation resistance are correlated in the DSPR ($F_{1,31}$ = 21.11, p < 0.0001).*
487 *Desiccation resistance is presented as RIL means (±SD). Open symbols indicate "low" starvation resistance*
488 *RILs; filled symbols indicate "high" starvation resistance RILs, and the dashed vertical line separates these*
489 *RIL classes.*

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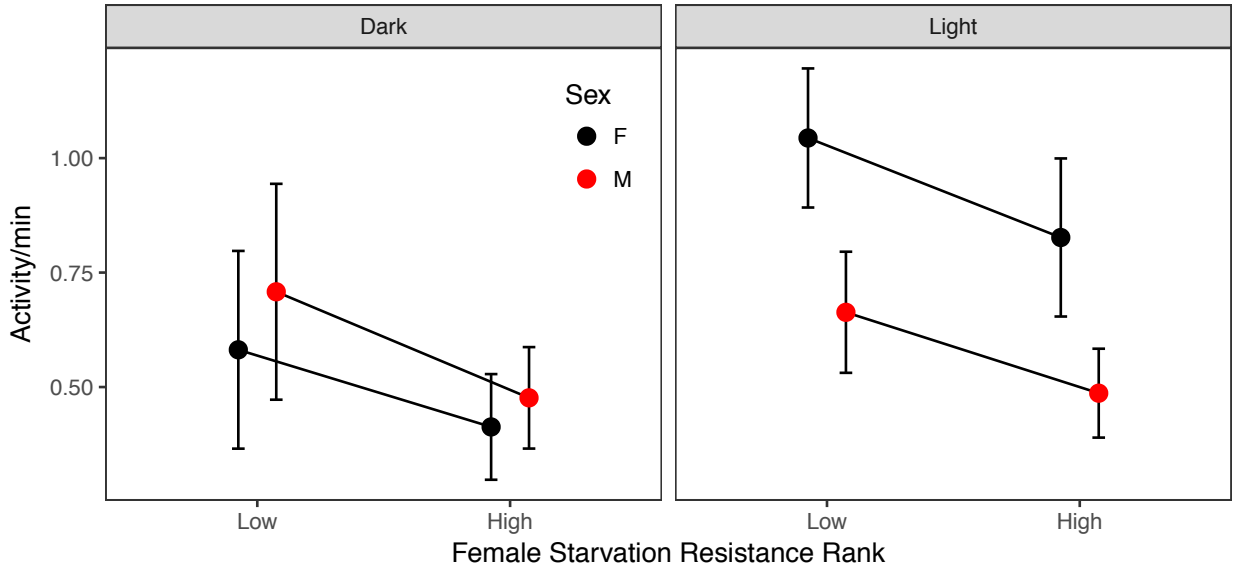
492

493 One physiological mechanism that may increase tolerance to environmental stressors is
494 a reduction in metabolic rate (Lighton and Bartholomew 1988; Hoffmann and Parsons 1989a; b,
495 1991; Chippindale *et al.* 1996; Djawdan *et al.* 1997; Marron *et al.* 2003; Rion and Kawecki 2007;
496 Schwasinger-Schmidt *et al.* 2012; Slocumb *et al.* 2015). Indeed, selection for both starvation and
497 desiccation resistance has been shown to lead to a correlated change in activity level, an indirect
498 proxy for metabolic rate (Hoffmann and Parsons 1989b, 1993; Schwasinger-Schmidt *et al.* 2012).
499 Here, we assessed activity of a subset of RILs exhibiting high and low starvation resistance to
500 understand how genetic variability in starvation resistance relates to activity levels under non-
501 stressful conditions. In the presence of nutritive media males and females differed in activity level
502 across the light and dark period ($F_{1,132}$ = 16.9, p < 0.0001; Fig 4; Table S7), with high starvation
503 resistance RILs exhibiting significantly lower activity levels than low starvation resistance RILs
504 ($F_{1,132}$ = 12.5, p < 0.001; Fig 4; Table S7). The effects of starvation resistance rank (high vs low),

505 sex, and the light status (light versus dark) on activity were similar in magnitude (Cohen's F: 0.21-
506 0.36; Table S7), suggesting that these factors contribute similarly to variation in waking activity
507 levels.



508

509 ***Figure 4.*** *Activity level on regular media for males (red) and females (black) from a subset of high and low*
510 *female starvation resistance RILs. Panels indicate the light and dark periods of a 24-hour monitoring*
511 *period. Activity while awake was influenced by both a sex-by-light interaction ($F_{1,132}$ = 16.9, P < 0.001) and*
512 *by starvation resistance class (i.e. high or low; $F_{1,132}$ = 12.5, P < 0.001).*

513
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515
516 The differences in activity between high and low starvation resistance lines on regular,
517 nutritive media (Fig 4) were preserved under starvation stress conditions, with the high
518 starvation resistant lines being less active than the low starvation resistant lines throughout the
519 starvation process (Fig S13; Table S8). This pattern aligns with that from previous studies. For
520 instance, Schwasinger-Schmidt *et al.* (2012) found that activity of flies with high starvation
521 resistance was reduced following 15 generations of selection for starvation resistance in both
522 males and females. Slocumb *et al.* (2015) also found that waking activity was reduced in lines
523 selected to have high starvation resistance. Although previous associations between increased
524 starvation tolerance and lower activity levels, metabolic rate, and changes in behavior have been
525 observed (Murphey and Hall 1969; Hoffmann and Parsons 1989a; Blows and Hoffman 1993;

526 Djawdan *et al.* 1997; Karan *et al.* 1998; Schwasinger-Schmidt *et al.* 2012; Masek *et al.* 2014), our
527 findings present a novel addition to our understanding of how increased starvation resistance
528 may occur. Behavioral components of energy conservation are likely to play a role in how
529 individuals compensate for stressful conditions (van Dijk *et al.* 2002; McCue 2010; Masek *et al.*
530 2014) and represent an additional facet of the complex nature of phenotypic variability in
531 starvation resistance.
532

533 Starvation resistance and triglyceride level are correlated in the DSPR

534 Periods of starvation have been shown to significantly reduce triglyceride levels in both
535 males and females (Schwasinger-Schmidt *et al.* 2012), and others have suggested that fat stores
536 and starvation resistance may be genetically correlated (Service *et al.* 1985; Rose *et al.* 1992;
537 Chippindale *et al.* 1996; Harshman *et al.* 1999; Schwasinger-Schmidt *et al.* 2012; Slocumb *et al.*
538 2015). To investigate the relationship between these traits in the DSPR, we measured mean
539 female triglyceride level in a subset of the pA and pB DSPR RILs and found substantial phenotypic
540 and genetic variation among RILs (Table 1; Fig 5). Mean starvation resistance and triglyceride
541 level were positively correlated in both DSPR panels, although the correlation in the pA and pB
542 panels was significantly different (Fig 6). Overall, variation in mean starvation resistance
543 explained 23.7% of the variation observed in mean triglyceride level among the DSPR RILs across
544 the two mapping panels (R^2 = 23.7%, $F_{3,929}$ = 95.96, $p < 0.0001$; Fig 6), suggesting that a proportion
545 of variation in female starvation resistance can be explained by variation in triglyceride level in
546 the DSPR.

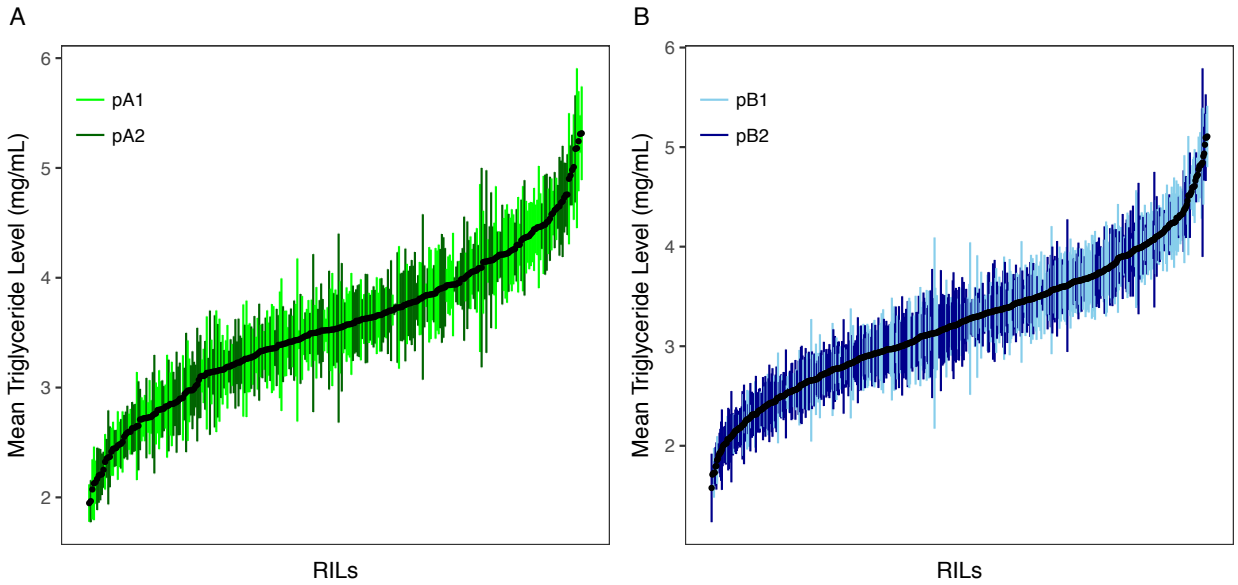


Figure 5. *Variation in mean DSPR triglyceride level (±SD) for females in the pA panel (A) and pB panel (B).*

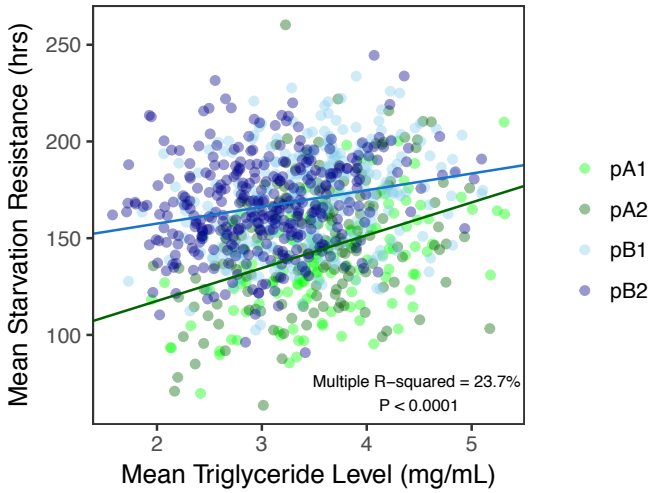


Figure 6. *Mean starvation resistance and triglyceride level are positively correlated in females ($F_{3,929}$ = 95.96, p < 0.0001). The strength of the correlation varied between the two mapping panels (interaction: $F_{1,929}$ = 9.32, p < 0. 01). Points are colored to indicate subpopulation for each mapping panel, although subpopulation was not included in the regression analysis.*

The correlation between triglyceride level (or total lipid level, depending on the study) and starvation resistance has been measured in numerous natural and artificially selected *Drosophila* populations, and a positive relationship is often described (Chippindale *et al.* 1996; Djawdan *et al.* 1997; van Herrewege and David 1997; Harshman *et al.* 1999; Schwasinger-Schmidt

561 *et al.* 2012; Goenaga *et al.* 2013; Slocumb *et al.* 2015; Hardy *et al.* 2018). For example, in
562 isofemale lines derived from populations distributed across approximately 14.4 degrees of
563 latitude, Goenaga *et al.* (2015) found that 12% of the variation in female starvation resistance
564 was accounted for by lipid content. Similarly, Chippindale *et al.* (1996) found a very strong
565 positive relationship between total lipids and starvation resistance following extended selection
566 for increased starvation resistance and suggested that lipid levels may directly determine
567 starvation resistance. However, a strong correspondence between lipid content and starvation
568 resistance is not always observed in strains derived from natural populations (Robinson *et al.*
569 2000; Hoffmann *et al.* 2001; Jumbo-Lucioni *et al.* 2010). For example, Jumbo-Lucioni *et al.* (2010)
570 found no correlation between triacylglycerol levels and starvation resistance measured in inbred
571 lines derived from a natural population. Hoffman *et al.* (2001) suggested that variation in the
572 strength of the correlation between triglycerides and starvation resistance may be due to the
573 evolutionary history of the study population. Evolutionary tradeoffs between increased lipid
574 storage and other aspects of fitness may also influence the correlation between starvation
575 resistance and lipid levels (Huang *et al.* 2014; Hardy *et al.* 2015, 2018). Furthermore, artificial
576 selection may increase starvation resistance via mechanisms that preferentially modify lipid
577 accumulation or metabolism, rather than by impacting energy level or energy-saving behavioral
578 strategies (Hoffmann and Parsons 1989a; Blows and Hoffman 1993; Hoffmann *et al.* 2001;
579 Marron *et al.* 2003; Rion and Kawecki 2007; Masek *et al.* 2014; Slocumb *et al.* 2015). The
580 relationship observed between triglyceride levels and starvation resistance in our study supports
581 the hypothesis that triglyceride levels and starvation resistance are likely physiologically related.
582 Equally, it is evident from our data that triglyceride level likely influences starvation resistance to
583 a lesser degree than proposed by Chippindale *et al.* (1996) and Hoffmann and Harshman (1999),
584 and that starvation resistance and triglyceride level have the potential to evolve independently
585 under natural selection.
586

587 ## Starvation resistance QTL allow prediction of DSPR founder phenotypes

588 We identified 8 QTL for starvation resistance in the pA panel and 7 QTL in the pB panel,
589 several of which overlapped between sexes (Fig S14). Both sets of QTL explained a substantial

590 amount of variation in starvation resistance, with individual peaks accounting for 3.7-13.2% of
591 the variation (Table 2). The total variance explained by QTL in the pA (pB) population was 26.1%
592 (32.8%) in females and 17.5% (37.9%) in males, assuming QTL are independent and additive
593 (Table 2). None of the QTL identified in the pA and pB mapping panels overlapped, and since
594 power to detect 5% QTL is expected to be high in our study (King *et al.* 2012a) and all DSPR
595 phenotyping was completed within seven months using the same design and environmental
596 conditions, this likely reflects genetic variation among the different sets of founders used to
597 establish the two sets of lines.
598
599
600 ***Table 2. Summary of QTL identified for starvation resistance and triglyceride level in the DSPR.***

Starvation Resistance QTL: pA DSPR panel								
QTL	**Sex**	**Peak LOD**	**Chr**	**Physical Interval (Mb)**	**Genetic Interval (cM)**	**Variance Explained**	**No. Genes**	**No. Potential Starvation Candidate Genes[a]**
SA1	M	7.41	2L	0.34-0.64	0.36-1.02	3.90	52	3
SA2	F	10.85	2R	4.90-5.53	59.39-60.64	6.01	122	10
SA2	M	10.59	2R	4.87-5.58	59.32-60.74	5.52	126	10
SA3	F	8.05	2R	12.85-13.43	80.75-82.81	4.23	110	9
SA4	F	7.03	2R	14.18-14.72	84.14-86.68	3.70	105	7
SA5	F	10.13	3L	3.40-3.92	6.89-8.88	5.29	52	4
SA6	F	13.21	3R	0.79-2.37	47.10-47.52	6.85	244	11
SA7	M	7.66	3R	15.55-15.86	66.52-67.36	4.03	52	4
SA8	M	7.73	3R	21.07-21.48	86.77-88.20	4.06	80	3
Starvation Resistance QTL: pB DSPR panel								
QTL	**Sex**	**Peak LOD**	**Chr**	**Physical Interval (Mb)**	**Genetic Interval (cM)**	**Variance Explained**	**No. Genes**	**No. Potential Starvation Candidate Genes[a]**
SB1	F	7.68	2L	7.02-7.30	23.69-25.20	4.05	33	1
SB2	M	8.35	2L	10.10-10.50	39.24-40.93	4.39	110	13
SB3	F	7.13	2L	12.14-12.65	45.94-46.96	3.77	44	4
SB3	M	7.05	2L	12.13-12.63	45.92-46.93	3.72	45	4
SB4	F	12.28	2R	1.10-2.70	54.93-55.73	6.41	148	10
SB4	M	19.40	2R	2.01-2.69	55.21-55.72	9.91	81	8
SB5	F	18.98	3L	18.94-19.42	45.48-45.75	9.73	54	5
SB5	M	26.40	3L	19.00-19.43	45.51-45.76	13.24	46	3
SB6	F	9.07	3R	8.86-9.20	52.50-53.29	4.77	48	7

SB6	M	12.77	3R	8.57-9.21	52.33-53.30	6.64	90	9
SB7	F	7.72	3R	25.65-26.15	99.12-100.11	4.08	111	8

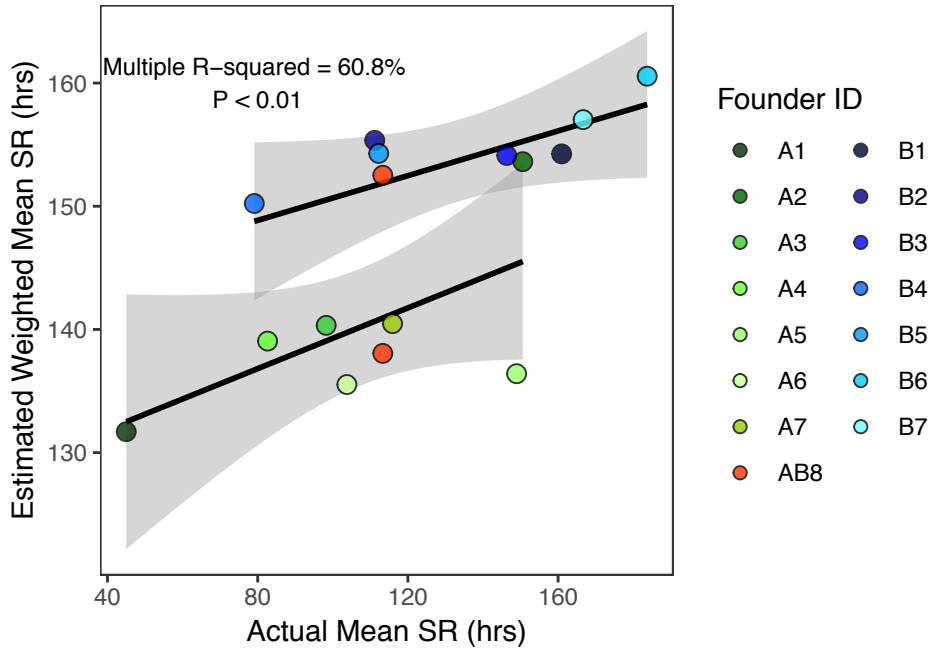
Triglyceride Level QTL: pB DSPR panel

QTL	Sex	Peak LOD	Chr	Physical Interval (Mb)	Genetic Interval (cM)	Variance Explained	No. Genes	No. Potential Starvation Candidate Genes[a]
TB1	F	8.73	3L	3.45-4.09	7.08-9.52	6.22	84	7
TB2	F	7.90	3R	7.18-8.01	50.50-51.54	5.64	118	5
TB3	F	8.45	3R	8.31-8.71	51.96-52.54	6.03	54	8
TB4	F	7.71	X	17.34-17.43	57.17-57.44	5.51	4	0

[a]Genes and functions associated with starvation resistance are listed in Table S9. The number of genes associated with starvation resistance (SR) was determined by cross-referencing genes within each QTL interval with previously reported candidates and biological functions reported by FlyBase (Lin et al. 1998; Clancy et al. 2001; FlyBase Curators et al. 2004; Harbison et al. 2004; Sørensen et al. 2007; Nuzhdin et al. 2007).

601
602
603
604 Because we can estimate the effect associated with each founder haplotype at each
605 mapped QTL in the DSPR, it follows that a combination of the estimates across all QTL can be
606 used to predict the actual phenotypes of the original founder strains. We measured starvation
607 resistance in the 15 DSPR founder lines (Fig 1E, F) to test this prediction. It is likely that the
608 strength of the correlation between the estimated and actual trait response is influenced by the
609 number and effect size of each QTL mapped for the trait. To account for differences in the degree
610 to which starvation resistance is influenced by QTL of varying effect sizes, we calculated the
611 predicted mean trait for each founder line weighted by the variance explained by each QTL.
612 As anticipated, using a general linear model the weighted mean predicted starvation
613 resistance of the founders based on QTL effects was significantly correlated with the sex-
614 averaged mean starvation resistance measured for the founder lines (R^2 = 60.8%, $F_{3,13}$ = 6.12, $p <$
615 0.01; Fig 7). The slope of this relationship is relatively small ($\beta = 0.13 \pm 0.07$), suggesting that
616 while a large component of variation in starvation resistance is clearly genetic (supported by
617 heritability estimates for each panel, Table 1), substantial variation in the phenotype is
618 unaccounted for by additive genetic effects at mapped QTL. This unaccounted-for genetic
619 variation in starvation resistance is likely due to many QTL with very small effects beyond our
620 power to detect them (King *et al.* 2012a) and/or epistatic interactions among QTL (Evans *et al.*;

621 Mackay 2014). Epistasis may be especially important when comparing actual founder strain
622 phenotypes with those inferred via QTL effects due to the many generations of recombination
623 employed while establishing the DSPR from the inbred founders. However, the strength of the
624 correlation between predicted and actual responses does suggest that QTL identified from the
625 DSPR mapping panels identify causative loci that influence the level of starvation resistance
626 among the progenitors of the RILs.



627

628 ***Figure 7.*** *Estimated starvation resistance weighted by the variance explained by each QTL and actual*
629 *starvation resistance measured for the 15 founder lines of the DSPR were significantly correlated ($\beta = 0.13$*
630 *± 0.07; $F_{3,13} = 6.21$, $p < 0.01$). AB8 identifies the founder line shared between the pA and pB mapping panels*
631 *estimated independently in each QTL analysis. Grey shading indicates the 95% CI of the regression.*

632
633
634

635 Limited overlap between the genetic architecture of starvation resistance and triglyceride 636 level in the DSPR

637 To further understand the relationship between starvation resistance and triglyceride
638 level in the DSPR we compared the genetic architectures of these two traits. We mapped four
639 distinct QTL for triglyceride level in the pB population, each of which accounted for 5.5-6.2% of
640 the variation in this trait (Table 2; Fig S15), in total explaining 23.4% of the variation in pB. No QTL

641 for triglyceride level were detected in the pA panel, likely due to the reduced number of pA RILs
642 assessed (pA N = 311; pB N = 628). However, even with this reduced power, the QTL map for pA
643 suggests that the genetic architecture for triglyceride level is different between the two mapping
644 panels, as there is no evidence of near-significant peaks in pA within QTL intervals statistically
645 identified in the pB panel (Fig S15).

646 Given the phenotypic correlation between triglyceride level and starvation resistance in
647 the DSPR (Fig 6) and similar correlations previously reported in other studies (Chippindale *et al.*
648 1998; Hoffmann and Harshman 1999), one might predict overlap of QTL associated with these
649 traits. However, we see only limited evidence for this. Triglyceride QTL TB1 (mapped in the pB
650 panel) and starvation QTL SA5 (mapped in the pA panel) do physically overlap, but given the
651 complete lack of evidence for QTL for the same trait co-localizing in both the pA and pB DSPR
652 mapping panels, it is unlikely the variant(s) underlying these QTL are the same. To investigate the
653 relationship between the two QTL that do overlap within the same panel (SB6 and TB3), we
654 assessed the influence of haplotype structure at the overlapping QTL on the positive phenotypic
655 correlation between triglyceride level and starvation resistance (Fig 6). In this analysis, we first
656 identified the founder haplotype for each RIL at the positions of the overlapping QTL peaks, and
657 calculated the average phenotype of each of the founder haplotypes. We then assessed the
658 correlation between haplotype-specific mean triglyceride level and starvation resistance with a
659 general linear model. After accounting for the haplotype structure at the overlapping peaks, we
660 found that mean starvation resistance and triglyceride level were significantly correlated ($F_{1,7}$ =
661 7.72, $p < 0.05$, R^2 = 52.4%; Fig S16), suggesting some pleiotropic variants may be responsible for
662 this pair of overlapping starvation resistance and triglyceride level QTL.

663 The limited overlap in the QTL intervals associated with starvation and triglyceride level
664 suggests that the genetic bases of this pair of traits are largely independent, or at least not tightly
665 linked at QTL with moderate to large effects. In natural populations, increased starvation
666 resistance may evolve as a result of selection on diverse traits including metabolic rate, activity
667 level, lifespan, development rate, thermal tolerance, and fecundity (Service *et al.* 1985;
668 Hoffmann and Parsons 1989b, 1993; Rose *et al.* 1992; Chippindale *et al.* 1993; Djawdan *et al.*
669 1997; Harshman *et al.* 1999; Bochdanovits and de Jong 2003; Marron *et al.* 2003; Bubliy and

670 Loeschcke 2005; Rion and Kawecki 2007; Schwasinger-Schmidt *et al.* 2012), and triglyceride levels
671 may be influenced by genetic variation in each of these traits. Our evidence of minimal overlap
672 between the genetic architectures of starvation resistance and triglyceride levels, coupled with
673 a phenotypic correlation between these traits, may be indicative of a series of complex
674 correlations between traits that influence stress tolerance, energy metabolism, and life history
675 in the DSPR.
676

677 Candidate genes underlying fitness trait variation

678 Across all QTL identified for starvation resistance and triglyceride level in this study,
679 several genes within mapped QTL intervals have functions related to these and other correlated
680 traits (Table 2, Table S9). Of particular interest are the 30 genes that fall within our QTL intervals
681 that were identified in previous starvation resistance studies (Clancy *et al.* 2001; Harbison *et al.*
682 2005; Sørensen *et al.* 2007) (Table S9). Gene ontology analyses performed for each trait and panel
683 revealed enrichment of genes within pA starvation QTL related to glutathione metabolic process
684 (6.91-fold enrichment, FDR corrected P = 0.000146; Table S10), as well as several categories that
685 were enriched for genes implicated by mapped triglyceride QTL (Table S10). This enrichment
686 could assist with the resolution of the functional genes within QTL regions. However, it should be
687 noted that the sets of genes implicated by QTL mapping in the DSPR (3-244 genes per interval in
688 this study) are extremely unlikely to all contribute to trait variation, and their presence within
689 QTL intervals cannot alone be taken as evidence for causality.

690 Upon examination of the genes within the overlapping starvation resistance and
691 triglyceride interval (TB3 and SB6), we found several genes that have either been predicted or
692 experimentally demonstrated to be associated with traits related to starvation resistance and
693 triglycerides or metabolism (Table S9 and references therein). Genes that fall within the intervals
694 of the overlapping peaks include those that influence adult lifespan (e.g. *ry*, *Men*, *Gnmt*
695 (Simonsen *et al.* 2006; Paik *et al.* 2012; Obata and Miura 2015)), lipid metabolic processes
696 (including *Lip3*, *CG11598*, *CG11608*, *CG18530* (FlyBase Curators *et al.* 2004)), insulin signaling
697 (e.g. *poly* (Bolukbasi *et al.* 2012)), response to starvation (e.g. *mthl12*, *Gnmt* (UniProt Curators
698 2002; Obata *et al.* 2014)), larval feeding behavior (e.g. *Hug* (Melcher and Pankratz 2005)),

699 circadian rhythm and sleep (e.g. *timeout*, *Men* (Harbison *et al.* 2004; Benna *et al.* 2010)), and
700 triglyceride homeostasis (*Gnmt* (Obata *et al.* 2014)) (Table S9). These genes are promising
701 candidates for future studies seeking to examine the functional genetic relationship between
702 these two traits.
703

704 Different mapping approaches reveal unique genetic architectures for starvation 705 resistance

706 Dissection of a quantitative genetic trait using different approaches can allow greater
707 resolution of the genetic architecture, and provide insight into how alleles unique to different
708 mapping panels contribute to phenotypic variation. To gain this additional understanding we
709 assessed the genetic architecture of starvation resistance in the DGRP using GWA mapping of
710 four starvation resistance datasets: new data collected in this study, data from Mackay *et al.*
711 (2012), data from Everman and Morgan (2018), and a consensus, across-study starvation
712 resistance measure calculated as the mean response across the three starvation datasets (150
713 lines were measured across the three studies). Using an FDR threshold of 20%, between 0 and 12
714 SNPs were associated with starvation resistance in each dataset and sex (Table 3; Table S11).
715 Aside from 3 SNPs that overlap between the across-study mean dataset and the Mackay *et al.*
716 (2012) dataset, none of these above-threshold SNPs were the same (Table 3; Table S11). Using
717 the more lenient significance threshold of $P < 10^{-5}$ (see Table 3 for the equivalent FDR values),
718 between 17 and 48 SNPs were associated with starvation resistance for each dataset and sex
719 (Table 3). However, overlap in associated SNPs among datasets was still minimal (Fig S17). The
720 SNPs identified using the more lenient significance threshold include all those identified at the
721 FDR threshold of 20%, so all subsequent analyses are performed on the larger set of associated
722 SNPs, and we acknowledge that these sets may include larger fractions of false-positive
723 associations.
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726 ***Table 3. Summary of DGRP GWA results and lines measured in this study, Mackay* et al. *(2012) and***
727 ***Everman and Morgan (2018).***

Study[a]	Significance Threshold	Equivalent FDR Female	Equivalent FDR Male	No. Strains	No. SNPs[b]: Female	No. SNPs[b]: Male
This Study	FDR 0.2	-	-	168	2	3
	$P < 10^{-5}$	0.78	0.79	168	23	19
Mackay et al. (2012)	*FDR 0.2*	*-*	*-*	*203*	*11*	*0*
	$P < 10^{-5}$	*0.49*	*0.82*	*203*	*39**	*17**
Everman and Morgan (2018)	*FDR 0.2*	*-*	*-*	*164*	*0*	*0*
	$P < 10^{-5}$	*0.67*	*0.65*	*164*	*25**	*22**
Across-Study Mean Response	FDR 0.2	-	-	150	12	0
	$P < 10^{-5}$	0.43	0.80	150	48	22

[a]Text in italics indicates results from the reanalysis of starvation data previously presented by Mackay et al. (2012) and Huang et al. (2014) and Everman and Morgan (2018).
[b]Asterisks indicate results that match those reported in the original studies.

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731 Across all four datasets reporting starvation resistance in the DGRP, 12 SNPs (associated
732 with seven genes) identified using the $P < 10^{-5}$ significance threshold fell within QTL intervals
733 identified for starvation resistance in the DSPR (Fig S18; Table S11). In females, one SNP
734 associated with starvation resistance from Mackay *et al.* (2012) is within QTL SA4 (gene
735 *CG30118*), and one SNP associated with starvation resistance in this study is present in QTL SA3
736 (gene *mbl*). One SNP associated with starvation resistance from Everman and Morgan (2018) is
737 within QTL SB7 (gene *hdc*), and one SNP from the average of starvation resistance across the
738 DGRP datasets is within QTL SB5 (gene *Gbs-76A*). In males, one SNP associated with starvation
739 resistance from Mackay *et al.* (2012) is within QTL SB3 and was not associated with a gene. Four
740 SNPs (one from this study and three from the average of starvation resistance) fell within QTL
741 SB5 and were all near the gene *fz2*; one additional SNP associated with the average of starvation
742 resistance was also within SB5 and was associated with the gene *pip*. One SNP associated with
743 starvation resistance from Everman and Morgan (2018) was within QTL SB6 (gene *beat-Vc*). Only
744 one of these seven genes has been previously associated with starvation resistance (*CG30118*;
745 (Sørensen *et al.* 2007)), and none have reported functions specifically related to starvation
746 resistance or general stress response (FlyBase Curators *et al.* 2004). Furthermore, none of the
747 overlapping genes survived an FDR threshold of 0.2, increasing the possibility that these genes

748 may be false positives. Therefore, with the possible exception of *CG30118*, these genes may not
749 be promising candidates, despite their overlap among studies.

750 Compared to genes implicated by QTL identified in the DSPR, which include several that
751 have been previously associated with starvation or related phenotypes (e.g., lifespan or lipid
752 content), DGRP GWAS hits implicate fewer *a priori* strong candidate genes. Additionally, we did
753 not observe any GO enrichment following analyses of SNPs associated with the four starvation
754 resistance datasets, although we acknowledge that the limited number of implicated genes likely
755 compromised the power of these analyses. Of the total 127 unique genes associated with
756 starvation resistance in the DGRP across studies and sexes, only two have been previously
757 identified as associated with starvation resistance in other mapping populations (*CG30118*, *scaf6*;
758 Table S11; (see Table S2 in Sørensen *et al.* 2007)). More generally, five had previously been
759 associated with the determination of adult lifespan (e.g. *cnc* and *Egfr*; Table S11; (Sykiotis and
760 Bohmann 2008; Kamakura 2011)), and 5 have been previously associated with lipid metabolism
761 or metabolic processes (e.g. *GlcAT-P* and *Ugt86Dj*; Table S11; (FlyBase Curators *et al.* 2004;
762 Gaudet *et al.* 2010)). Given the relative lack of power of a GWA study using less than 200
763 genotypes (Long *et al.* 2014), and our use of a permissive genomewide threshold, it could be that
764 many of the GWAS associations are incorrect, explaining why associations do not typically tag
765 known candidates. Equally, it could be the case that a series of novel pathways are involved in
766 natural variation for starvation resistance, and that traditional candidates - often identified via
767 mutagenesis screens rather than through examination of segregating allelic variation - typically
768 do not harbor the functional natural variants detectable in a GWAS (Mackay *et al.* 2009).

769 The general lack of correspondence among the loci associated with starvation resistance
770 in each mapping panel does not invalidate either approach as a strategy to uncover functional
771 variation. It is likely that many genes contribute to variation in this trait with effects that are
772 either fairly small, or that have effects only in a specific genetic background (i.e., exhibit genetic
773 epistasis), and we would not expect to routinely identify such loci. In addition, comparison of the
774 genetic architecture of quantitative genetic traits across multiple panels is complicated by a
775 number of additional factors. The genetic structure of the mapping panel (e.g., whether it is a
776 multiparental panel like the DSPR or a population-based association study panel like the DGRP)

777 influences the analytical strategy, and the power and resolution of mapping (Long et al. 2014).
778 The complement of alleles present in the panel, and the frequency with which they segregate,
779 will also affect the ability to identify the same locus across mapping panels (e.g., King and Long
780 2017). This point is especially true for the comparisons made here, since the DSPR represents a
781 global sampling of genetic variation represented by the 15 founder strains, whereas the variation
782 present in the DGRP is a direct reflection of the genetic variability in a single population at a single
783 point in time. Therefore, a lack of overlap in the identified QTL for a complex, highly-polygenic
784 trait between the DSPR and the DGRP is perhaps not unexpected.
785

786 Repeatability in the SNPs associated with starvation resistance across DGRP studies

787 The public availability of starvation resistance data for the DGRP from multiple studies
788 provides a novel opportunity to investigate the reliability and repeatability of associations
789 identified for a classic quantitative trait in the same mapping panel across independent
790 phenotypic screens (Lithgow *et al.* 2017). Despite the moderately high phenotypic correlation
791 between starvation resistance measured in the three studies (Fig S8), only a single variant was
792 implicated in more than one of the three studies (Fig S17).

793 The lack of overlap in SNPs associated with starvation resistance could be due to
794 differences in the rearing/testing environments of the three studies (discussed above), where
795 genotype-by-environment effects - often pervasive for complex traits (Gurganus *et al.* 1998) -
796 could lead to different sites impacting variation in different studies. However, it is potentially
797 more likely that the sets of associated SNPs have real, but extremely small effects on starvation
798 resistance variation, and power in a GWA panel of modest size is too low to consistently detect
799 them (Boyle *et al.* 2017). If true, one would predict that - in contrast to random SNPs - the "top
800 SNPs" identified within each starvation resistance dataset would have additive effects of the
801 same sign across all studies. In essence, significantly associated SNPs with positive effects on
802 starvation resistance from data collected in this study would be expected to have positive
803 additive effects on starvation resistance measured in Mackay *et al.* (2012) and Everman and
804 Morgan (2018) more often than expected by random chance.

805 To test this prediction, we first collected the sign of the additive effects of SNPs that
806 survived the significance threshold ($P < 10^{-5}$) in each dataset for both sexes (Table S11), and
807 determined whether these top SNPs had additive effects of the same sign in every other dataset.
808 We then established a null distribution of SNP additive effect signs across pairs of datasets. This
809 was accomplished by taking samples of 50 SNPs segregating in the DGRP and extracting the sign
810 of the additive effect of each SNP in the pair of datasets, regardless of the association statistic
811 for that SNP. The proportion of the 50 SNPs that had a positive additive effect on the trait was
812 recorded for each of 1000 iterations and used to build an expected distribution of SNP effect sign
813 sharing for each pair of datasets. We note that to compensate for any allele frequency bias in the
814 variants that are actually most associated with phenotype, we ensured that the frequencies of
815 the randomly-sampled SNPs were stratified to match the top 50 SNPs associated with each sex
816 and dataset (see Materials and Methods; Table S3).
817 Finally, we compared the proportion of top SNPs for which the sign stayed the same
818 across studies to our expected distribution for each pair of datasets (Fig 8). We found that for the
819 random samples of SNPs, the probability that the additive effects were positive in both datasets
820 compared was greater than 50% for most comparisons (distribution means ranging from 45.9-
821 72.1%; Fig 8). This implies that a random SNP is slightly more likely to have the same sign effect
822 across data sets, which may be explained in part by the phenotypic correlation between the
823 datasets (Fig S10). Even given this skew, far more top SNPs than expected by chance had additive
824 effects of the same sign in each of the other starvation resistance studies (Fig 8; Table S3; Fig
825 S19). This may suggest that there is phenotypic signal even in SNPs that have very small effects,
826 and that are not clearly associated with starvation resistance in the GWAS (see Yang *et al.* 2010).
827 Generally, the consistency of the additive effects of SNPs associated with starvation resistance in
828 the DGRP calculated across datasets implies that starvation resistance is a highly polygenic trait,
829 with a large number of QTL with very small effects that influence variation in this trait (Mackay
830 2004; Boyle *et al.* 2017).
831

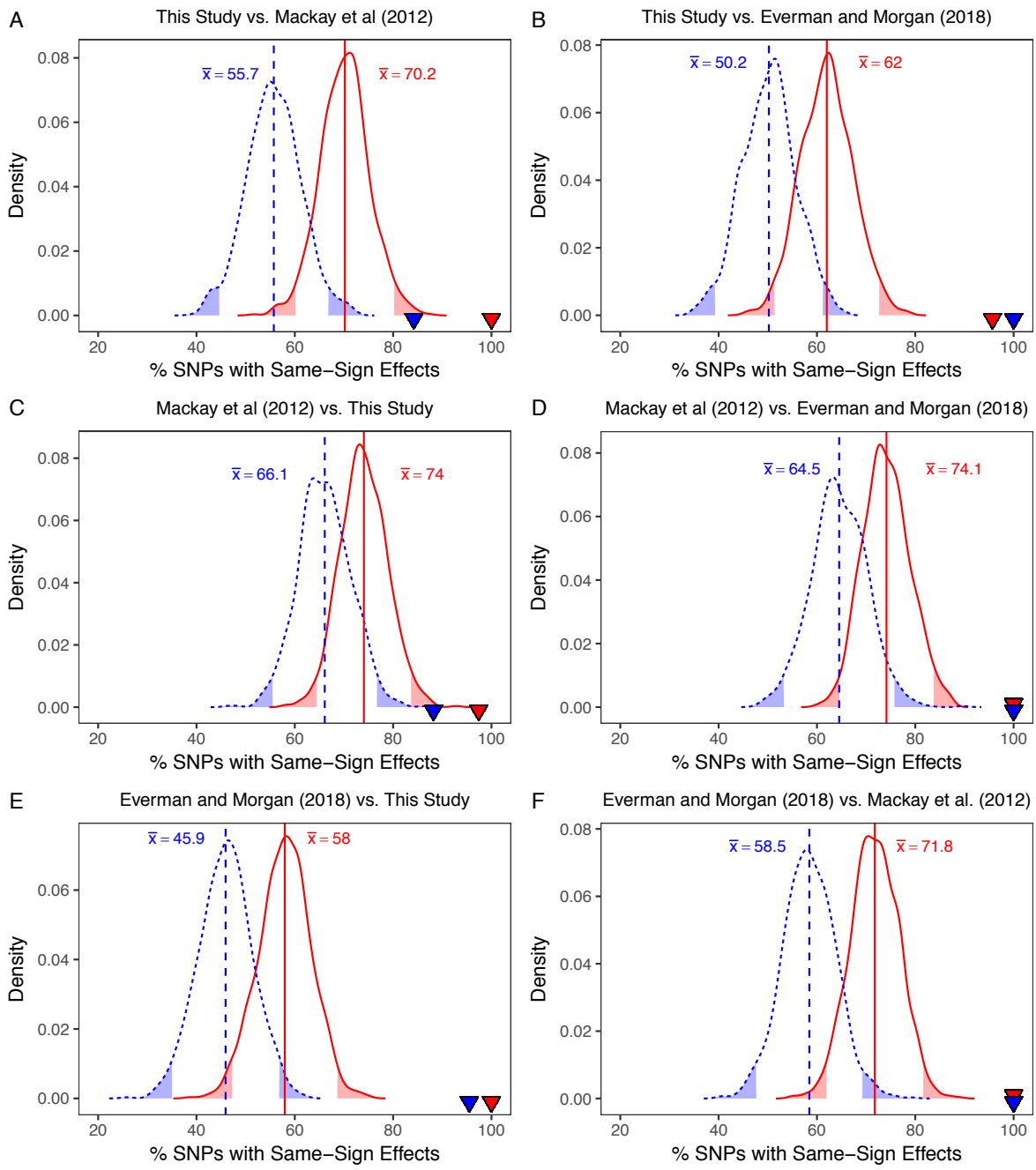

Figure 8. *Distribution of the proportion of SNPs that are expected to have additive effects of the same sign in pairwise comparisons of data from this study, Mackay* et al. *(2012) and Everman and Morgan (2018). Data shown were obtained for random samples of 50 SNPs with 1000 iterations. In each plot, red indicates female data, blue indicates male data, and corresponding vertical lines and text annotation indicate the mean of the sex-specific samples. The shaded tails represent the upper and lower 95% confidence intervals of each distribution. The triangles in each plot represent the sex-specific observed proportion of top SNPs from each GWA analysis ($P < 10^{-5}$) that had additive effects that were the same sign across studies (Fig S19). For example, in A, 100% of the SNPs associated with female starvation resistance in this study had additive effects of the same sign when calculated for starvation resistance reported in Mackay* et al. *(2012). The dataset comparison is indicated above each plot. In every case, top SNPs from one study were more*

843 *likely to have the same additive effect sign in a second study than a random set of frequency-matched*
844 *SNPs.*

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848 ## Conclusions

849 In this study, we have described the complex quantitative genetics of starvation
850 resistance in two large *D. melanogaster* mapping panels that have been thoroughly genomically
851 characterized. The DSPR and DGRP panels have the advantage of increased, stable genetic
852 diversity and provide a unique comparison to many previous quantitative genetic studies of
853 starvation resistance that may be limited by genetic diversity or mapping power. Correlations
854 between starvation resistance and the additional traits described in this study offer insight into
855 the genetic control of related stress response traits and provide support for the hypothesis that
856 the genetic architecture of stress traits varies by population and is dependent upon sex,
857 environment, and the evolutionary history of the populations studied. The relationships between
858 the traits analyzed in this study also offer insight into the broader responses of organisms to
859 starvation stress, given the high conservation of mechanisms related to starvation resistance in
860 diverse species (Partridge *et al.* 2005; Rion and Kawecki 2007). Here, we have demonstrated that
861 traits related to survival under starvation conditions, energy storage, activity levels, and survival
862 under desiccation stress are phenotypically correlated in the DSPR, consistent with previous
863 artificial selection studies as well as some natural populations. However, we also clearly
864 demonstrate that starvation resistance and triglyceride level are largely genetically independent
865 traits, indicating that evolutionary constraint between these two traits is unlikely. We
866 additionally describe the highly polygenic nature of starvation resistance using the DGRP,
867 leveraging previously published phenotypes on the same lines to compare the genetic
868 architecture of the trait across three studies. Our work shows that despite a lack of overlap across
869 studies in the identity of the variants associated with phenotype at a nominal genomewide
870 threshold, the sign of the additive effects of such top SNPs are conserved across studies
871 conducted by different labs. In turn this suggests that these variants do contribute to the
872 phenotype but have sufficiently small effects that they are not routinely captured following a

severe, genomewide correction for multiple tests. From this, we gain a much more detailed understanding of the genetic control of trait variation in a genetically diverse panel and provide insight into the utility of across-study and across-panel comparisons.

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Supplemental Figures

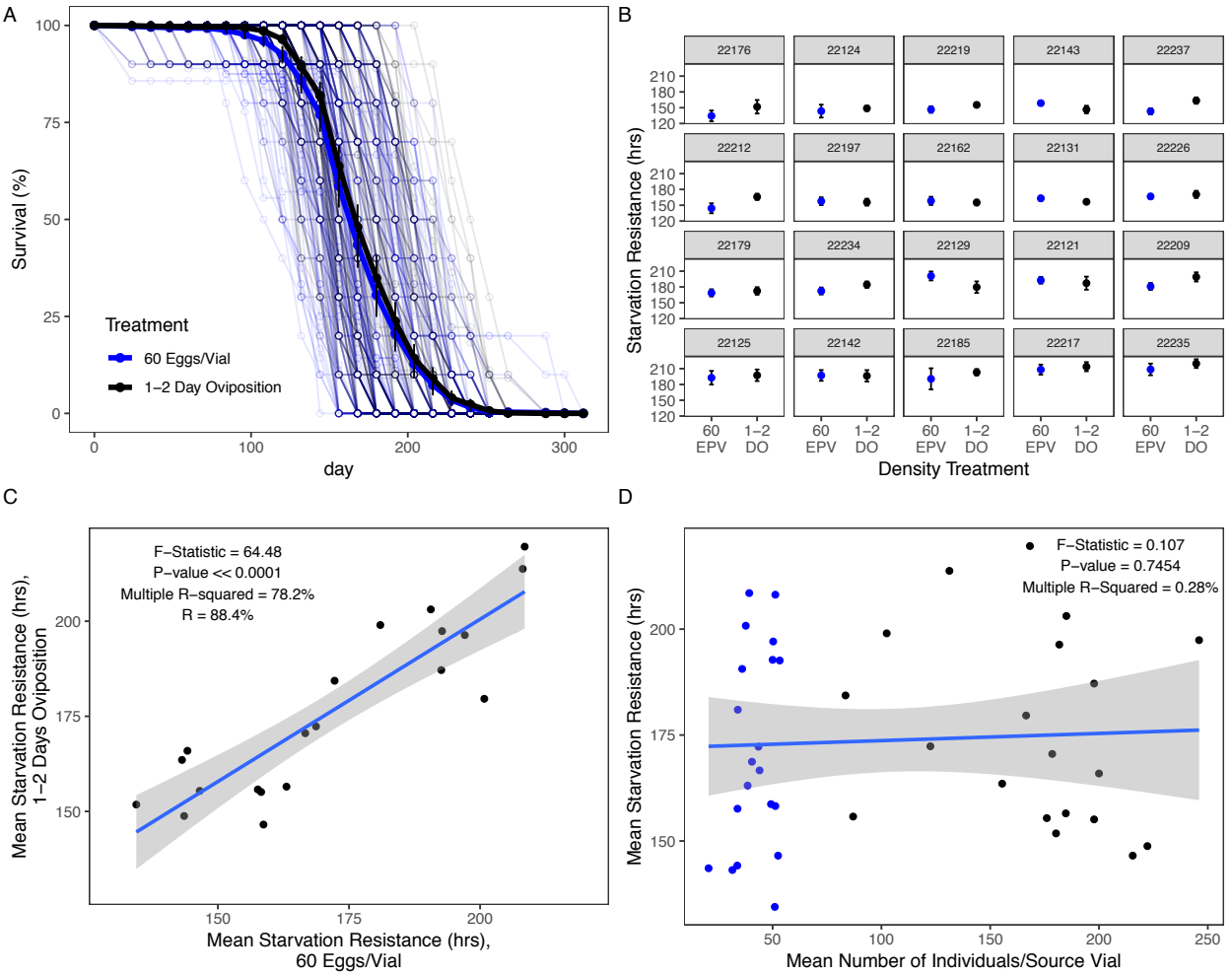

Figure S1. *Egg density in vials used to generate experimental individuals has a limited influence on starvation resistance. This was tested in 20 randomly-selected DSPR RILs by rearing experimental individuals according to two treatments: with 60 eggs per vial (60 EPV) or with 1-2 days of oviposition (1-2 DO). The total number of adults emerging from each source vial was also counted. Starvation resistance of the experimental individuals from each density treatment was measured as described for the large-scale starvation screen. A. Percent survival per vial at each 12-hr assessment point was very similar throughout the course of this experiment regardless of rearing density. Bold lines and points indicate the overall mean (±95% CI) survival for each treatment group at each 12-hr assessment point. B. Mean (±95% CI) starvation resistance for each of the 20 randomly-selected DSPR RILs was rarely influenced by the density treatment. Overall, density treatment had a minor effect on the average lifespan of each DSPR RIL ($F_{1,19}$ = 18.15, P < 0.0001, % Variance Explained = 0.90%; see Table S1 for full breakdown of variance components). C. Mean starvation resistance by DSPR RIL was strongly correlated between the two density treatments. D. The mean number of individuals per source vial of experimental flies did not explain a significant amount of variation in starvation resistance. In A, B, and D, black corresponds to the 1-2 day oviposition treatment; blue corresponds to the 60 eggs per vial treatment. In C and D, grey shading represents the 95% CI of the regression.*

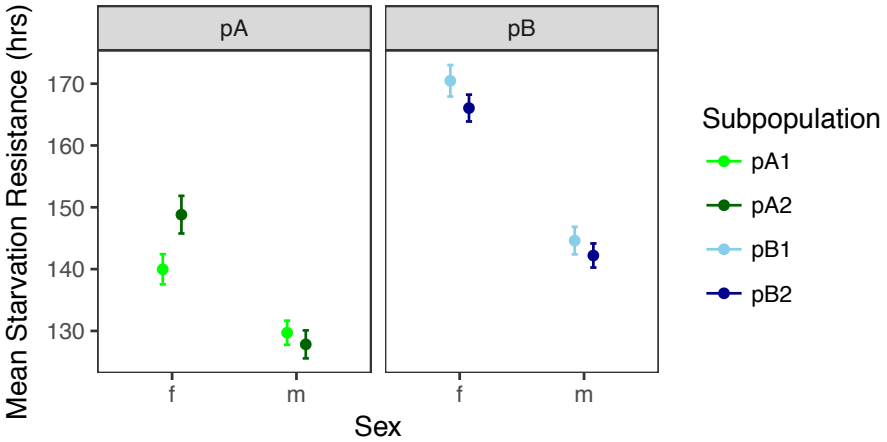

Figure S2. *Mean (± 95% CI) starvation resistance per subpopulation and sex. Sex and subpopulation interact to influence starvation resistance ($F_{3,3440}$ = 18.317; $p < 0.0001$), though only females from the pA1 and pA2 subpopulations were significantly different within a panel (Tukey's HSD adj. $p < 0.0001$).*

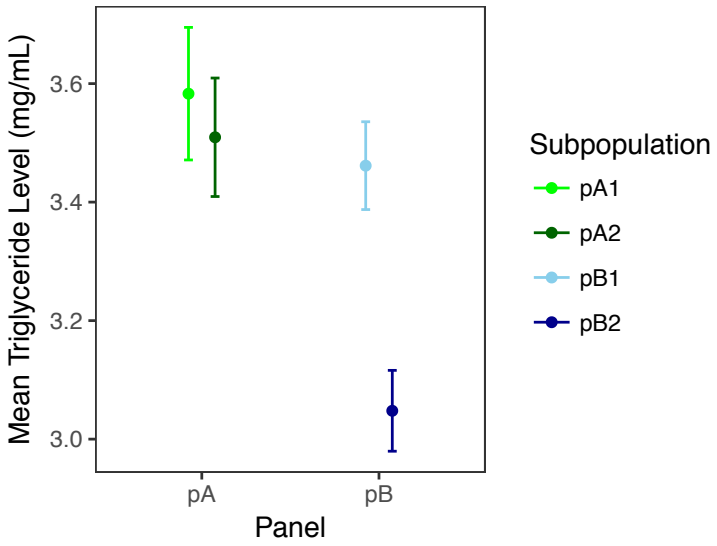

Figure S3. *Mean female triglyceride level per subpopulation (± 95% CI). Subpopulation influenced triglyceride level ($F_{3,935}$ = 37.099; $p < 0.0001$), though this was driven by differences between the pB subpopulations (Tukey's HSD adj. $p < 0.0001$)*

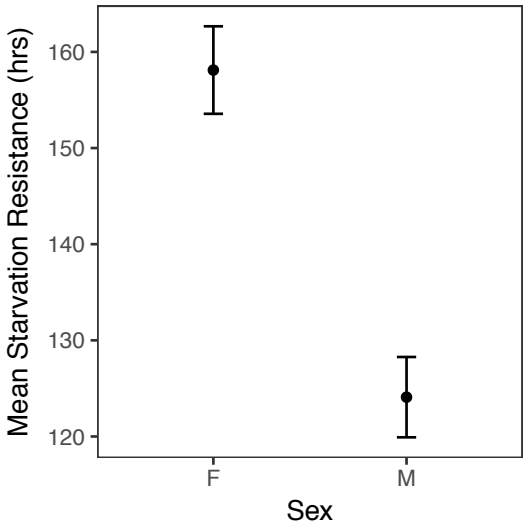

Figure S4. *Mean starvation resistance (±95% CI) for males and females in the DGRP ($F_{1,334}$ = 118.21, p < 0.0001).*

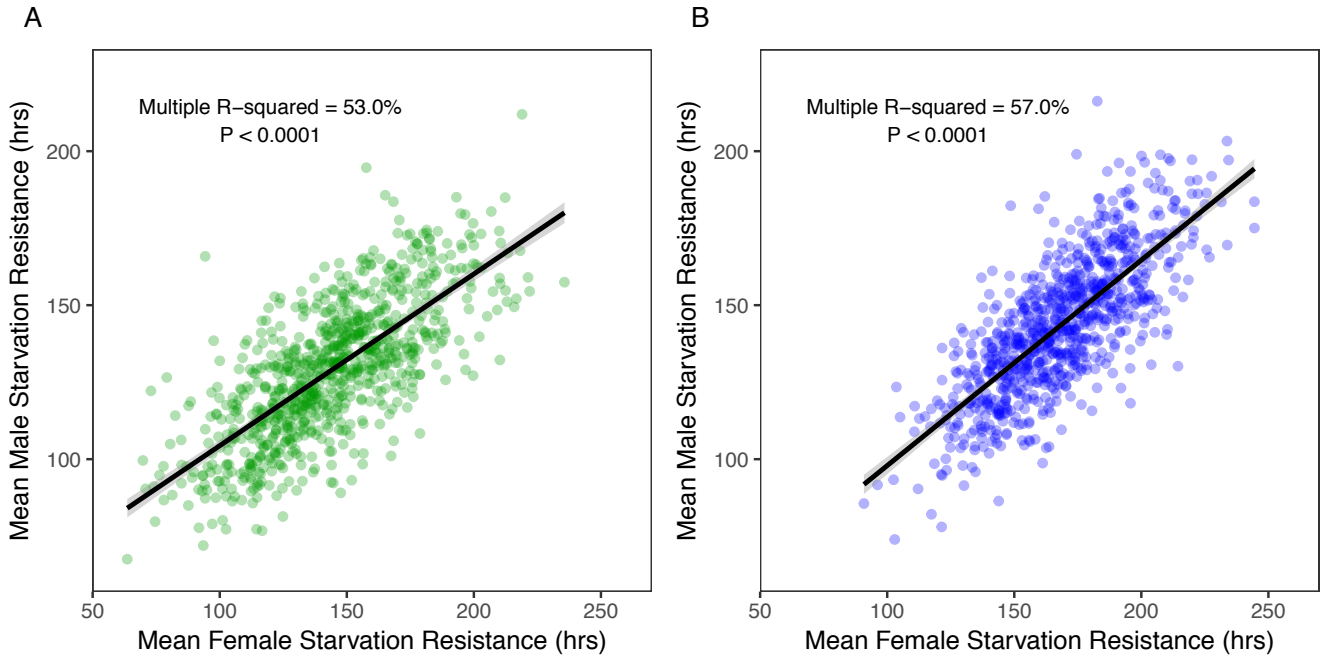

Figure S5. *Correlation between sex-specific mean starvation resistance for the DSPR pA mapping panel (A) and pB mapping panel (B). Sex-specific responses were significantly correlated for both panels (pA: $F_{1,859}$ = 968.5, P < 0.0001; pB: $F_{1,860}$ = 1138, p < 0.0001). Grey shading around the regression line in both plots indicates the 95% confidence interval.*

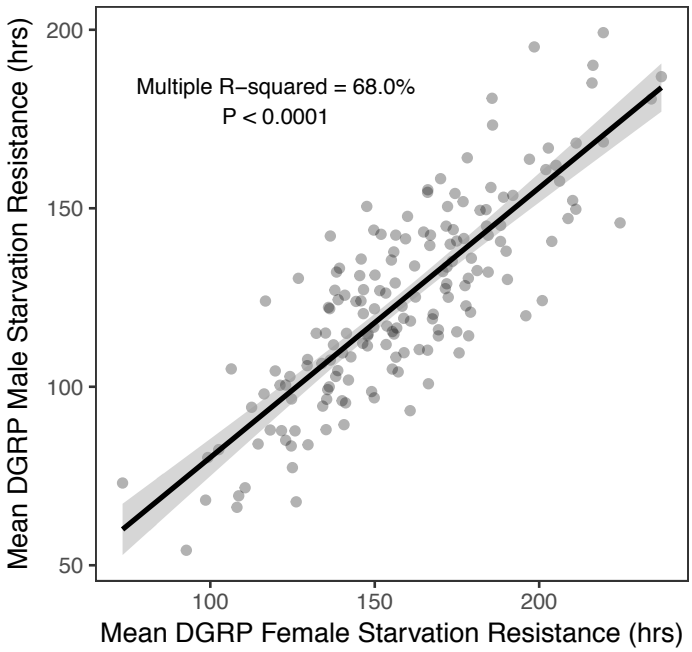

Figure S6. *Male and female mean starvation resistance was significantly correlated in the DGRP ($F_{1,334}$ = 118.21, P < 0.0001). Grey shading around the regression line indicates the 95% confidence interval.*

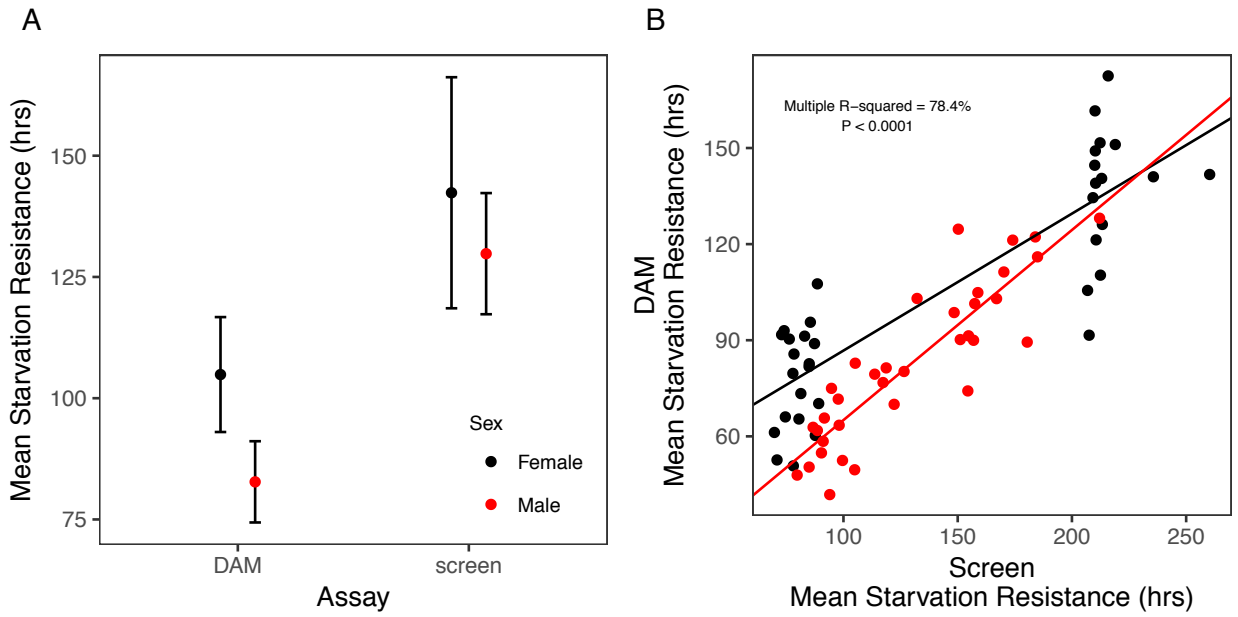

Figure S7. *A. Starvation resistance was significantly higher in the large-scale starvation screen of all DSPR RILs compared to the DAM (Drosophila Activity Monitor) starvation assay for the selected subset of RILs (Assay: $F_{1,136} = 31.60$, $p < 0.0001$). Mean starvation resistance across RIL means is presented ($\pm$95% CI). B. Mean starvation resistance measured in the large-scale starvation resistance screen (x-axis) was correlated with mean starvation resistance measured in the DAM assay (y-axis) in the DSPR (Females: $\beta = 0.43 \pm 0.04$, $t = 9.7$, $p < 0.0001$, $R^2 = 73.9\%$; Males: $\beta = 0.59 \pm 0.05$, $t = 10.9$, $p < 0.0001$, $R^2 = 78.3\%$). The multiple R^2 value in the plot includes the interaction between starvation resistance measured under different assay conditions with sex.*

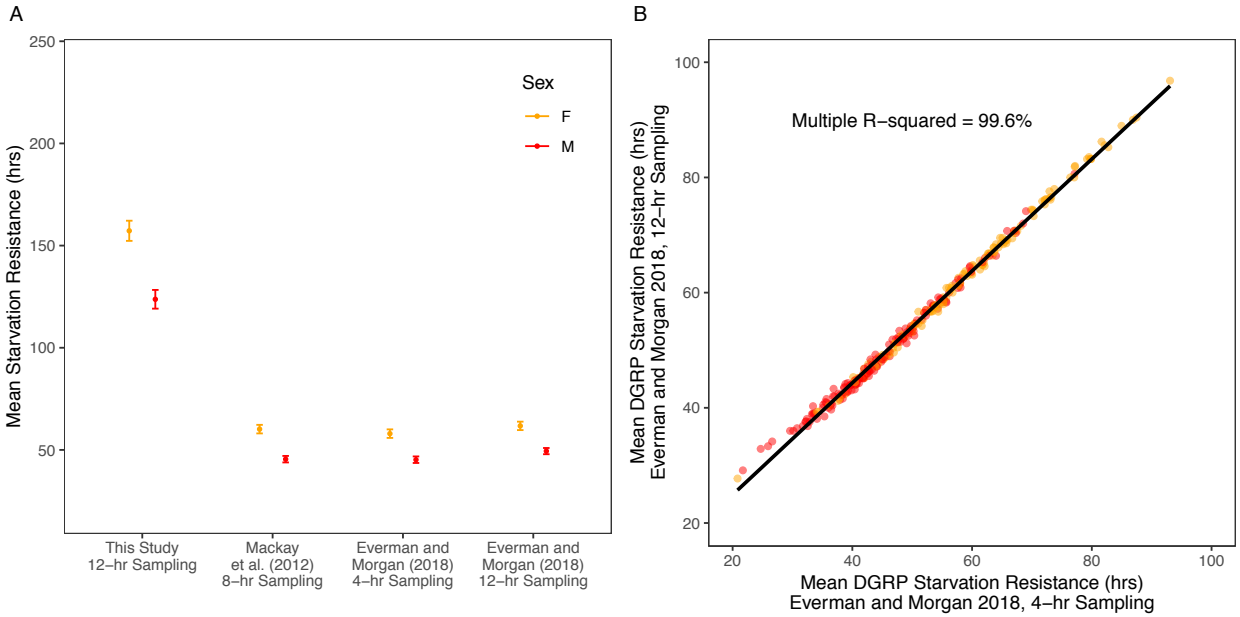

Figure S8. *A. Mean starvation resistance (± 95% CI) was significantly higher in this study compared to Mackay* et al. *(2012) and Everman and Morgan (2018) ($F_{2.532}$ = 1457.5, P < 0.0001). The increased mean and variation in starvation resistance observed in this study was not driven by differences in the frequency at which survival was assessed, since a re-analysis of data from Everman and Morgan (2018) with a longer interval between fly counting events matching the interval from the present study, revealed essentially no difference in the phenotypes assayed. B. Mean starvation resistance by line and sex measured according to the 4-hr sampling interval was highly correlated to our re-analysis of the Everman and Morgan (2018) data using a 12-hr sampling interval.*

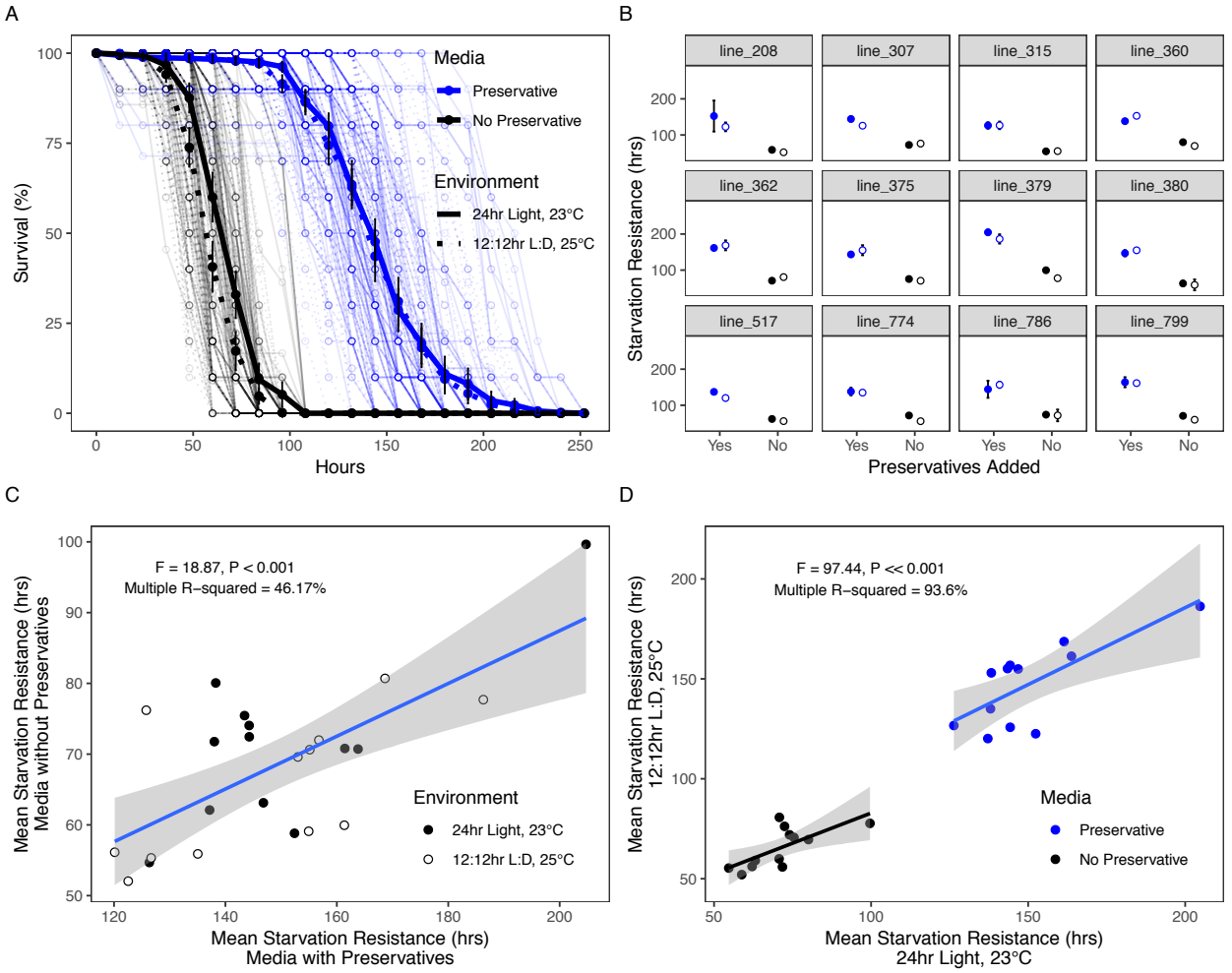

Figure S9. *Flies maintained on starvation media with preservatives lived much longer than flies on starvation media without preservatives, regardless of environmental conditions (24-hr light, 23°C as used in the large-scale starvation screen vs. 12:12hr L:D, 25°C as used in Mackay* et al. *(2012) and Everman and Morgan (2018)). This was tested in 12 randomly-selected DGRP lines. A. Percent survival per vial was different between the two media treatments, and differed slightly due to environment, but only when media did not contain preservatives. Black lines and points indicate media with no preservatives; blue lines and points indicate media with preservatives; solid lines indicate the 24hr Light, 23°C environment; dashed lines indicate the 12:12hr L:D, 25°C treatment. The bold points and lines for each treatment indicate the overall mean (±95% CI) survival of each treatment group at each 12-hr assessment point. B. Mean (±95% CI) starvation resistance for each of the 12 randomly selected DGRP lines was rarely influenced by the environment treatment (closed symbols = 24hr Light, 23°C; open symbols = 12:12hr L:D, 25°C), but media preservatives consistently resulted in higher survival for each DGRP line. C. Mean starvation resistance by DGRP line was significantly correlated between the two media treatments. D. DGRP line, preservatives in the media, and environmental conditions together explained nearly all phenotypic variation in starvation resistance. The full reporting of variance components is presented in Table S6. In C and D, grey shading represents the 95% CI of the regression.*

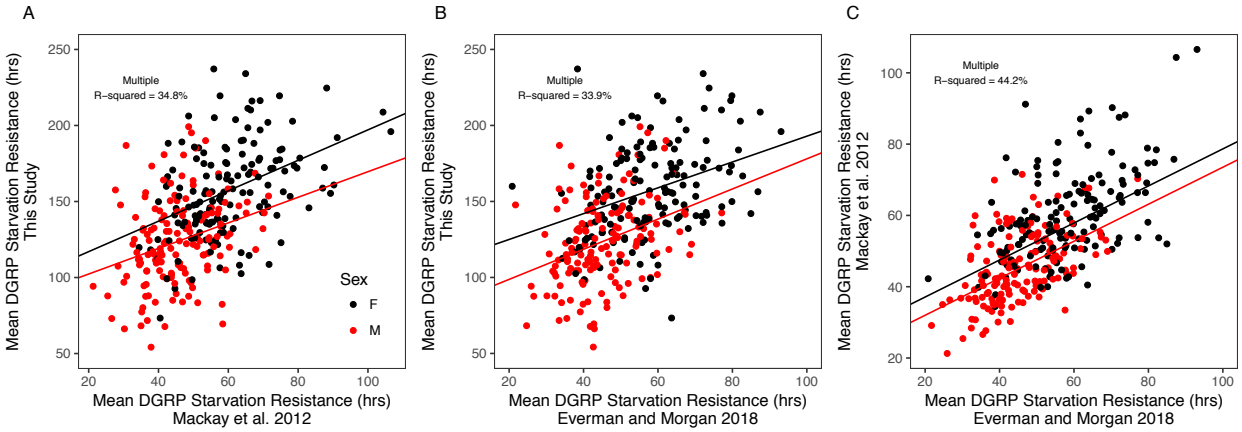

Figure S10. *Correlation between sex-specific mean starvation resistance in the DGRP panel for 150 lines that overlap between this study, Mackay* et al. *2012, and Everman and Morgan 2018. Red points indicate males and black points indicate females. All comparisons showed that the three independent measures of starvation resistance were significantly correlated (A: $F_{3,296} = 52.69$, $p < 0.0001$; B: $F_{3,296} = 50.66$, $p < 0.0001$; C: $F_{3.296} = 78.18$, $p < 0.0001$).*

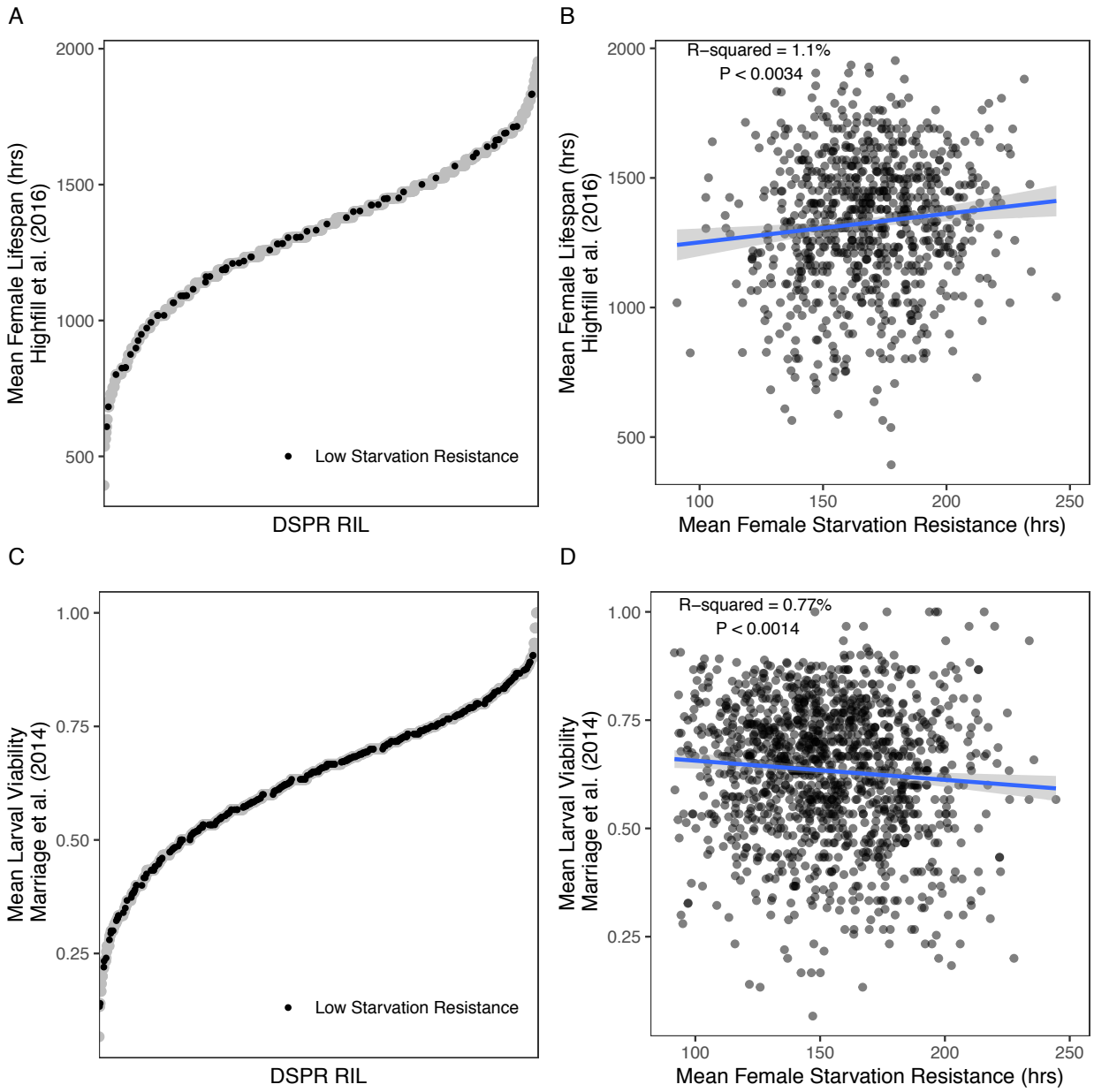

Figure S11. *DSPR RILs with low starvation resistance ranged from low to high for other measures of fitness. A. Distribution of mean female lifespan in DSPR pB RILs (Highfill* et al. *(2016). RILs with low female starvation resistance (in the bottom 25% of the distribution) are shown in solid black symbols, while other RILs are shown in gray. B. The correlation between mean female starvation resistance and mean female lifespan is minimal, suggesting there is no association between low starvation resistance and reduced lifespan. C. Distribution of mean larval viability measured as the proportion of 1st instar larvae reared under control conditions that emerged as adults (Marriage* et al. *(2014). RILs with low female starvation resistance (in the bottom 25% of the distribution) are shown in solid black symbols, while other RILs are shown in gray. D. The weak correlation between mean female starvation resistance and mean larval viability again suggests that there is no association between low starvation resistance and low larval viability. We also fail to find strong associations between starvation resistance and other measures of fitness in the DGRP (Fig S12).*

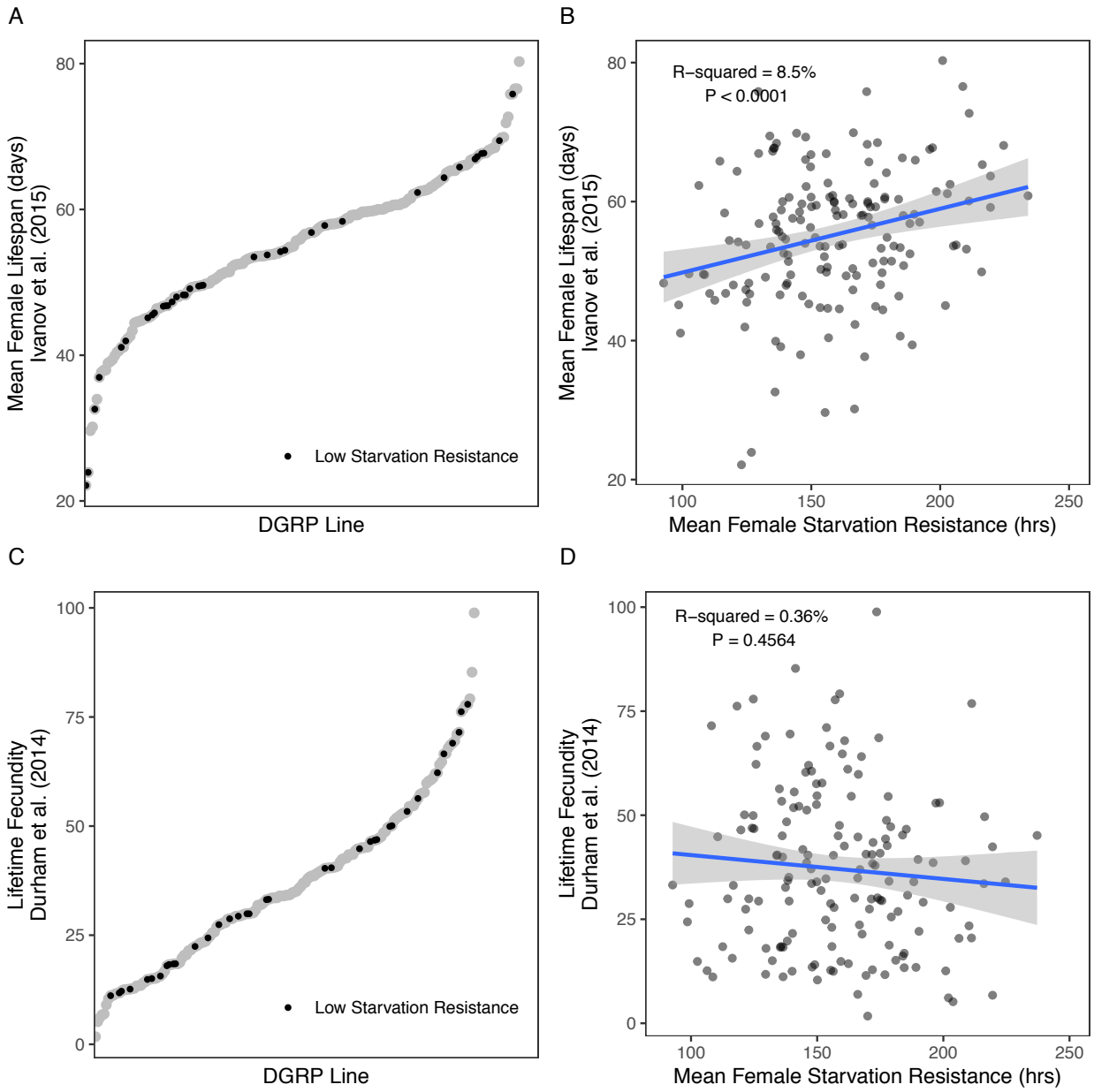

Figure S12. *DGRP Lines with low starvation resistance ranged from low to high for other measures of fitness. A. Distribution of mean female lifespan in the DGRP (Ivanov* et al. *(2015). Lines with low female starvation resistance (in the bottom 25% of the distribution) are shown in solid black symbols, while other lines are shown in gray. B. The correlation between mean female starvation resistance and mean female lifespan is minimal, suggesting there is little association between low starvation resistance and reduced lifespan. C. Distribution of lifetime fecundity (Durham* et al. *(2014). Lines with low female starvation resistance (in the bottom 25% of the distribution) are shown in solid black symbols, while other lines are shown in gray. D. The weak correlation between mean female starvation resistance and fecundity suggests that there is no association between low starvation resistance and this measure of fitness.*

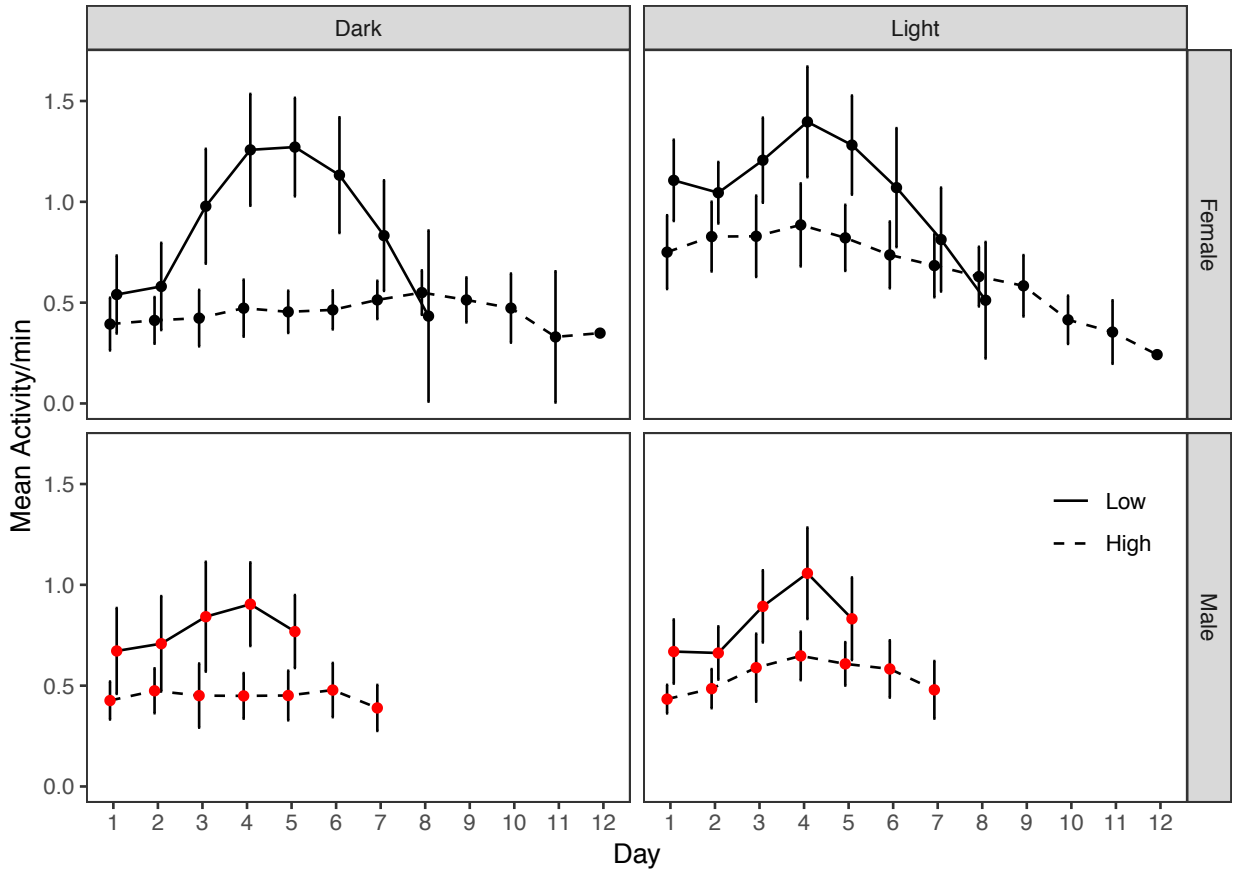

Figure S13. *Mean activity levels for females (top panel, black points) and males (bottom panel, red points) (±95% CI) across the 12-hr daily light or dark periods during starvation in the DAM (Drosophila Activity Monitor) assay until death. Low starvation resistance RILs (solid line) tended to be more active during starvation compared to high starvation resistance RILs (dashed line) under both light and dark conditions. Light status (dark versus light) influenced the overall activity level of females but did not influence male activity. Data were analyzed with a repeated measures ANOVA; results are presented in Table S8. Similar to the pre-starvation period (Fig 4), waking activity levels of individuals during the DAM starvation experiment were primarily driven by starvation resistance rank in both sexes (females: $F_{1,8} = 14.87$, $p < 0.01$; males: $F_{1,6} = 13.87$, $p < 0.01$; Table S8). Light status did not influence activity between days for either sex (females: $F_{1,8} = 0.84$, $p = 0.39$; males: $F_{1,6} = 0.23$, $p = 0.65$; Table S8). However, light status did significantly influence activity in females within each day (females: $F_{1,8} = 43.09$, $p < 0.0001$; Table S8), indicating that female activity in both high and low starvation resistance RILs was consistently higher during times when lights were on. Male activity of high and low starvation resistance RILs tended to remain constant under different light conditions. The larger influence of light status on females is consistent with patterns observed in activity during the pre-starvation period (Fig 4).*

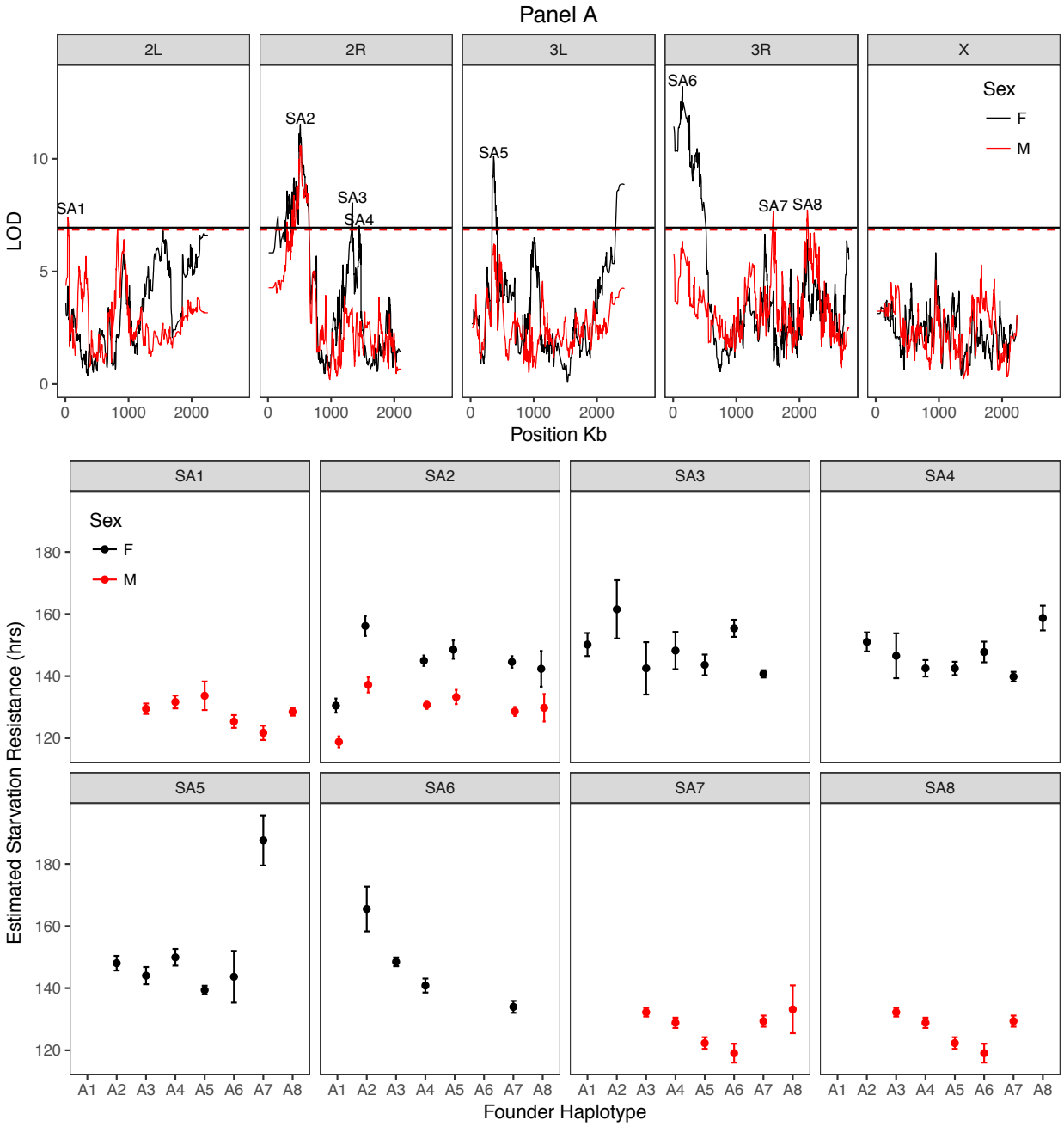

Figure S14. *Starvation resistance QTL and estimated allele effects at each QTL. Data are presented as RIL means (±SE) for estimated starvation resistance when the founder haplotype was present in more than 5 RILs. As has been seen in a number of studies using the DSPR and other multiparental populations (King* et al. *2012b; Giraud* et al. *2014; Najarro* et al. *2015), the estimated phenotypic effects of each founder haplotype suggest that multiple alleles may be present at our starvation QTL, since the effects do not fall into two clear "high" and "low" classes.*

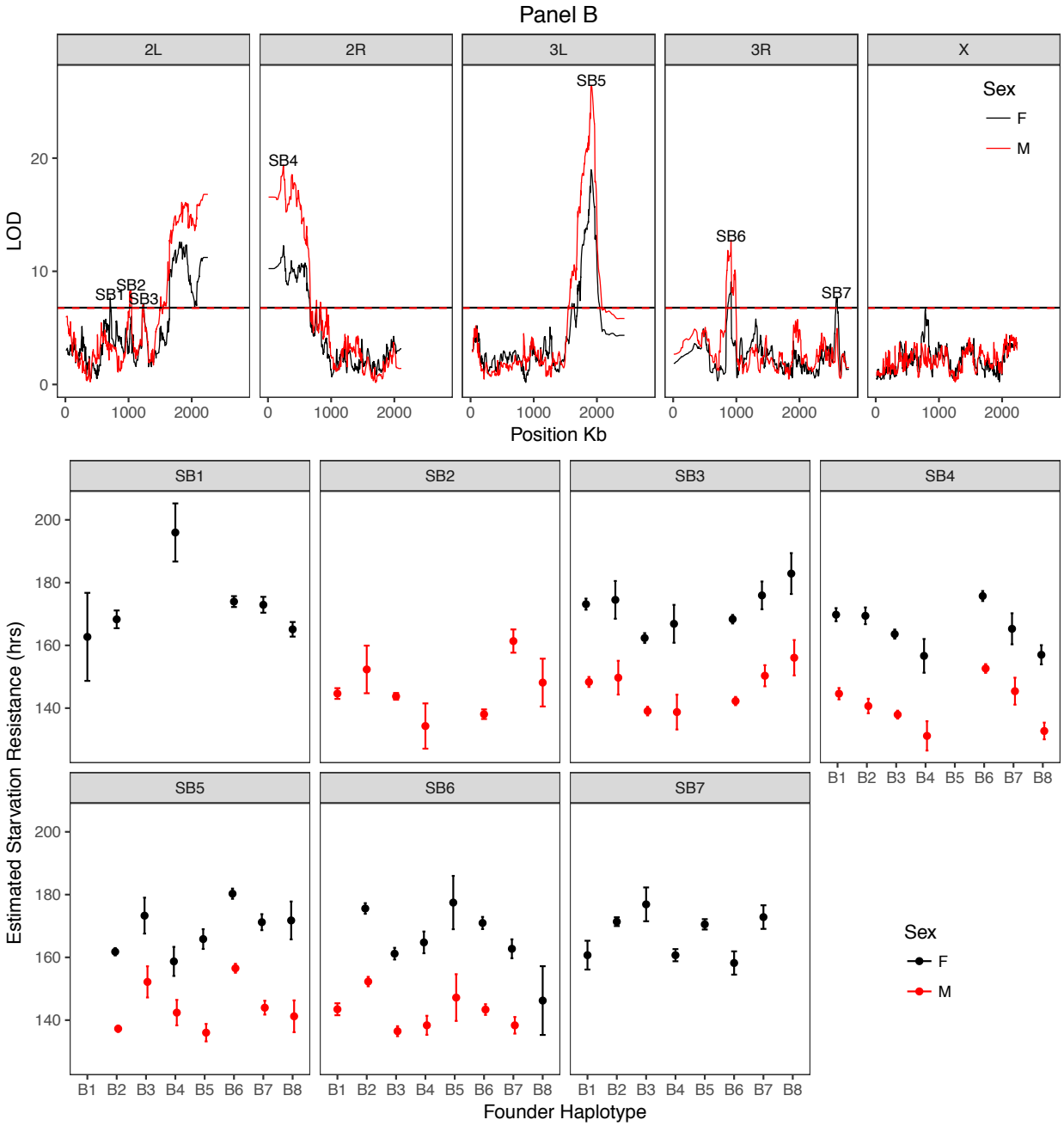

Figure S14 continued.

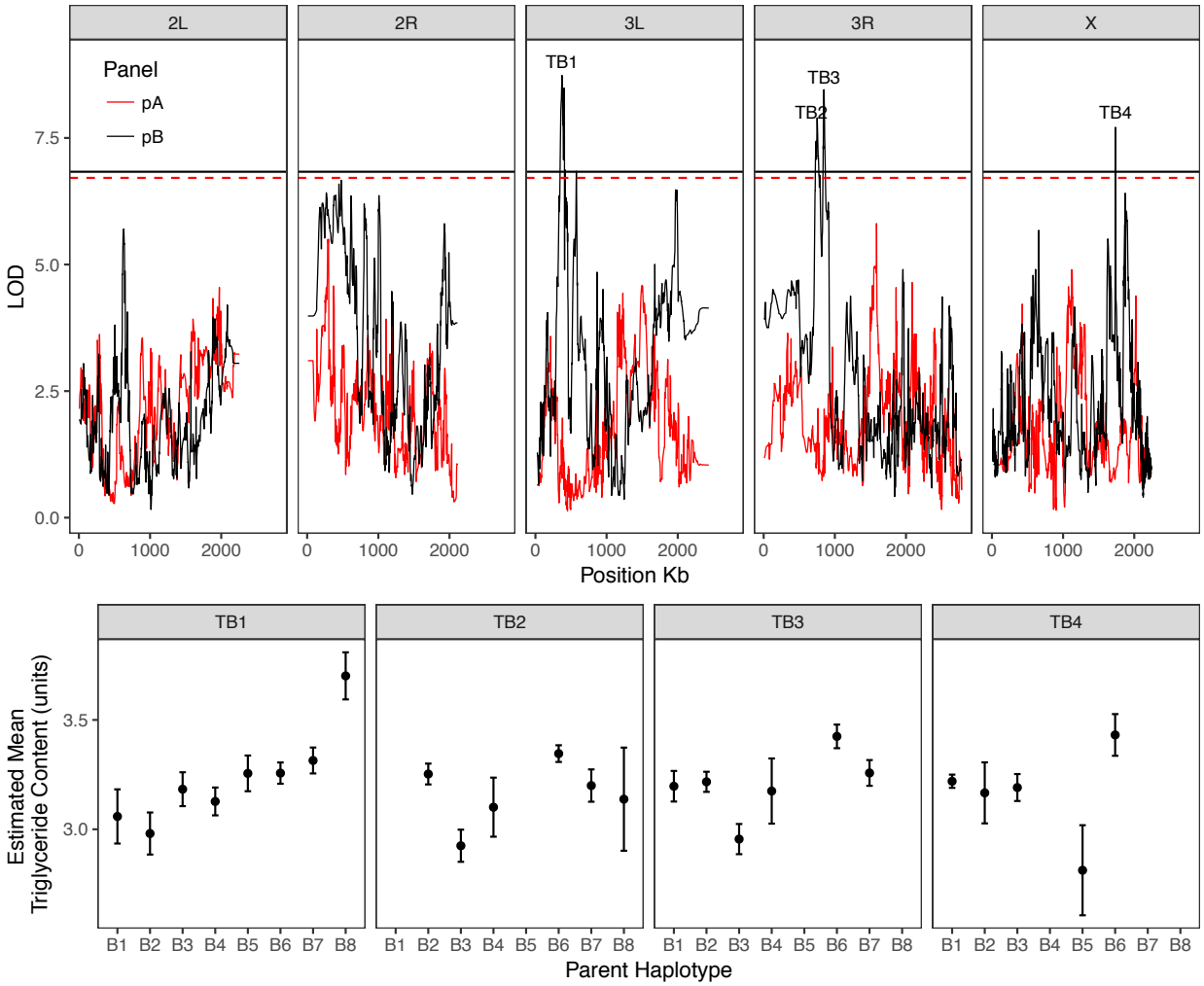

Figure S15. *Triglyceride level QTL and estimated allele effects of founders at each QTL. Data are presented as RIL means (±SE) for estimated triglyceride level when the founder haplotype was present in more than 5 RILs. Similar to starvation resistance, the estimated phenotypic effects of each founder haplotype suggest that multiple alleles may be present at our triglyceride QTL.*

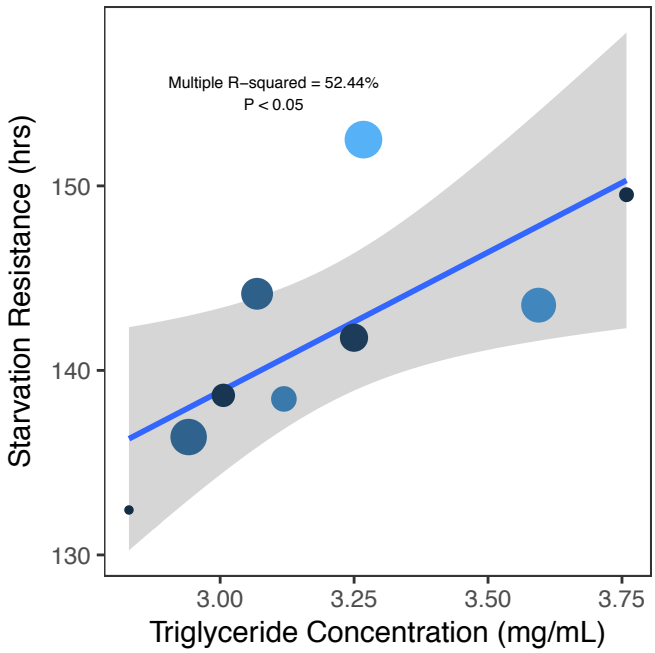

Figure S16. *Triglyceride level and starvation resistance were correlated after accounting for variation due to founder haplotype at the overlapping peaks TB3 and SB6 ($F_{1,7}$ = 7.72, P < 0.05). Data presented are averages for each founder haplotype in the pB panel, including "NA" for RILs that could not be assigned with confidence to a known haplotype. Point size relates the number of RILs per haplotype for the starvation resistance peak (smallest = 1 RIL; largest = 193 RILs); point color relates the number of RILs per haplotype for the triglyceride level peak (black = 1 RIL; lightest blue = 181 RILs). Grey shading around the regression line indicates the 95% confidence interval.*

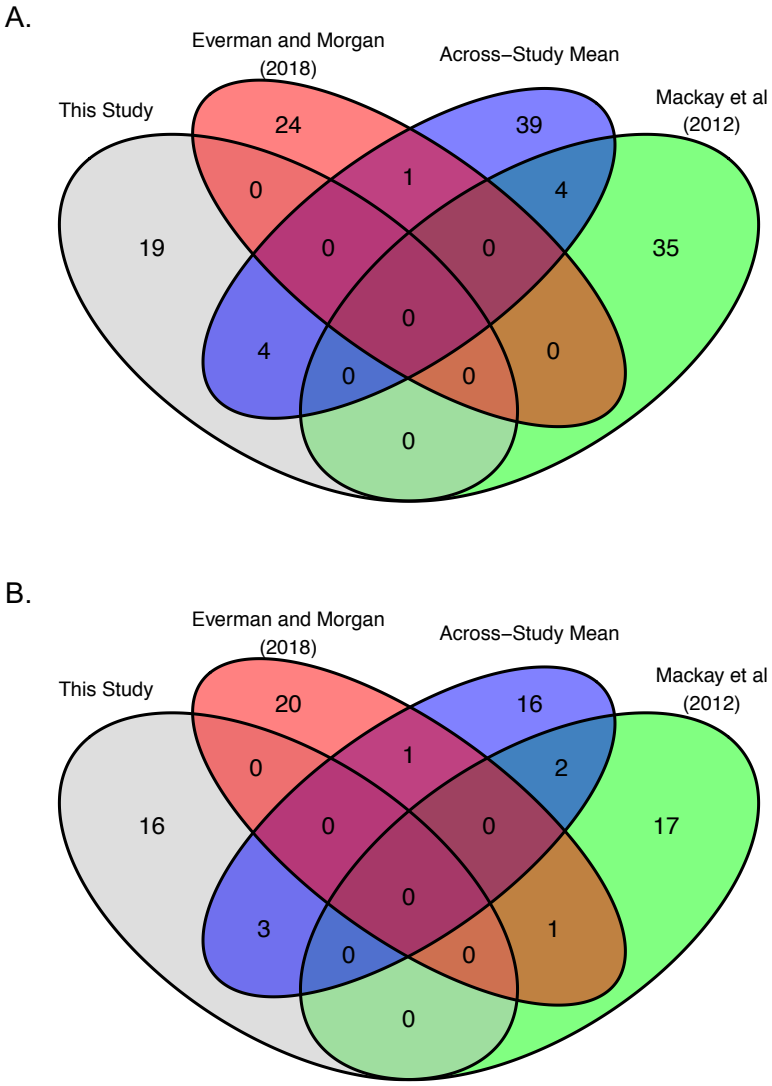

Figure S17. *Overlap in SNPs associated with starvation resistance for each DGRP dataset using the $P < 10^{-5}$ significance threshold. Overlap between data sets was minimal. Plot A presents results for females; plot B presents results for males.*

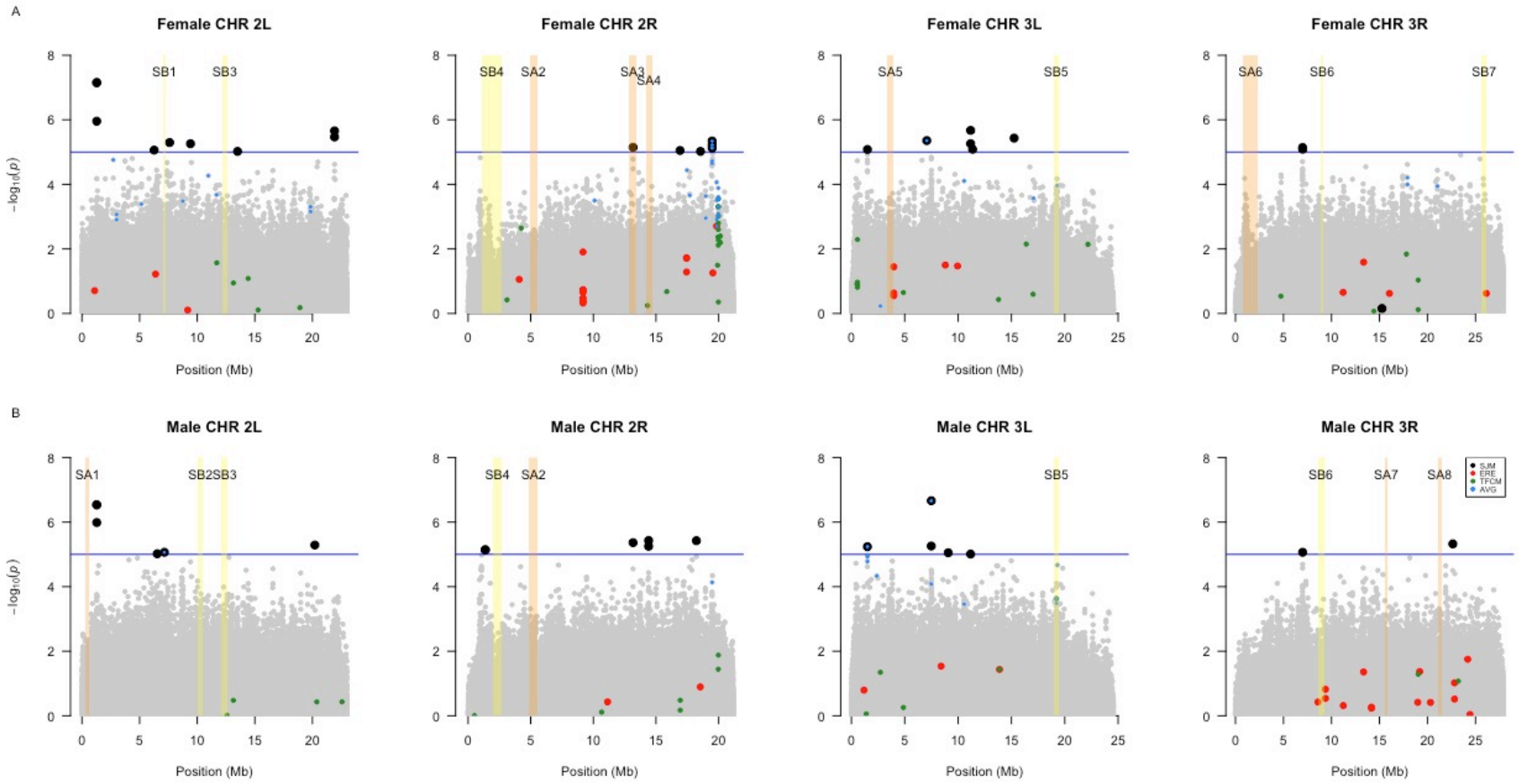
A
Female CHR 2L
Female CHR 2R
Female CHR 3L
Female CHR 3R
SB1
SB3
SB4
SA2
SA3
SA4
SA5
SB5
SA6
SB6
SB7
B
Male CHR 2L
Male CHR 2R
Male CHR 3L
Male CHR 3R
SA1
SB2SB3
SB4
SA2
SB5
SB6
SA7
SA8
SJM
ERE
TFCM
AVG
$-\log_{10}(p)$
Position (Mb)

Figure S18. *Manhattan plots of mean starvation resistance in the DGRP with SNPs that were associated with starvation resistance in previous studies and intervals of sex-specific QTL identified for starvation resistance in the DSPR highlighted. Plots are broken up by chromosome arm in A for females and in B for males. In all plots, points highlighted in black indicate SNPs that are associated with starvation resistance in the DGRP from data obtained in this study; red points indicate SNPs associated with starvation resistance in Everman and Morgan 2018; green points indicate SNPs associated with starvation resistance in Mackay et al. 2012; blue points indicate SNPs associated with starvation resistance averaged across the three datasets. A genomewide significance threshold of* $P < 10^{-5}$ *is shown with the blue line. Yellow shaded boxes and labels correspond to QTL intervals around peaks mapped in the pB DSPR panel; orange shaded boxes correspond to QTL intervals around peaks mapped in the pA DSPR panel.*

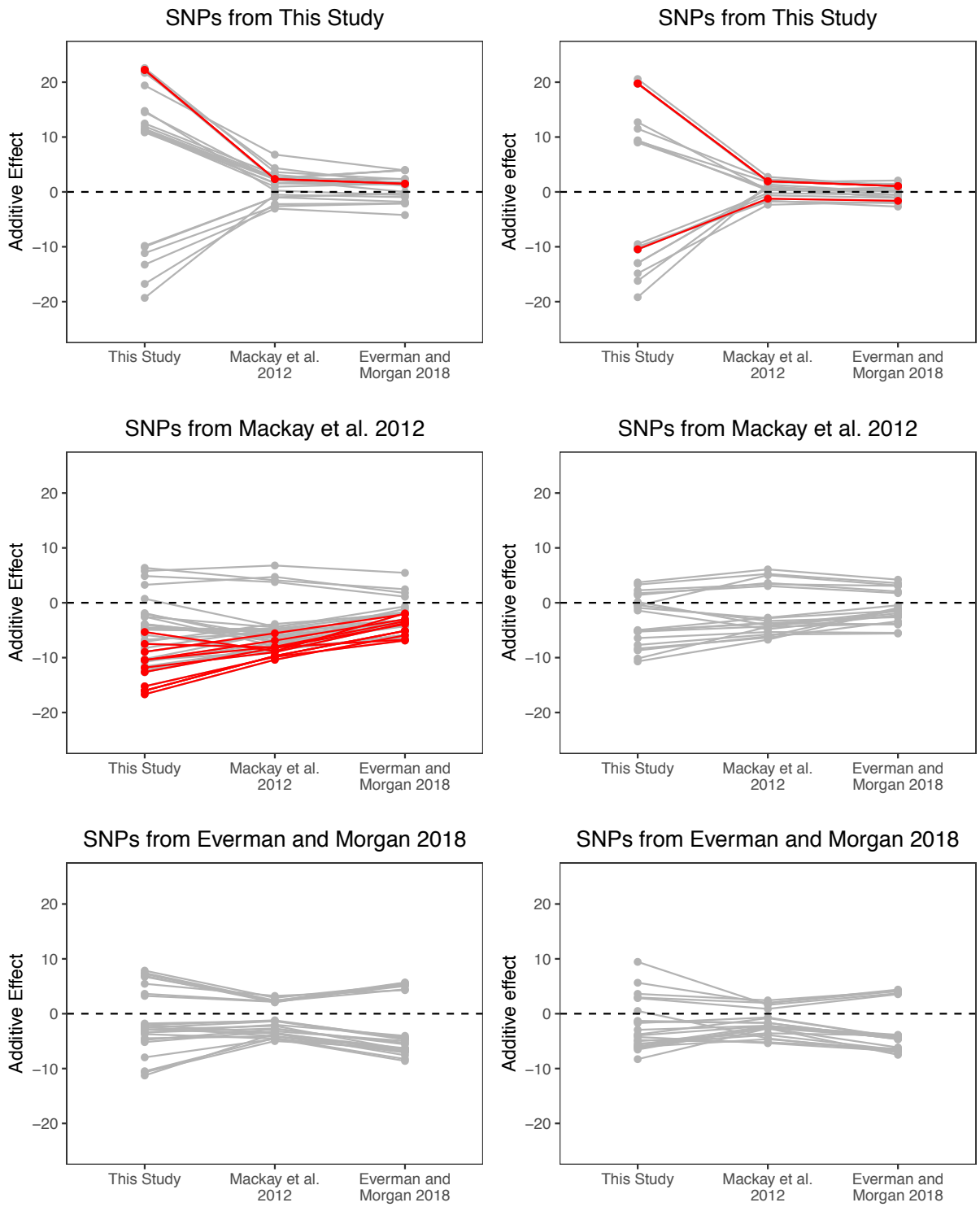

Figure S18. *Additive effects of SNPs associated with starvation resistance (at $P < 10^{-5}$) in each study, along with their additive effects estimated in the other two studies. Female data is presented in the left column of plots; male data is presented in the right column of plots. SNPs that passed the FDR threshold of 0.2 are highlighted in red. Generally, SNPs had similar effects (of the same +/- sign) on starvation resistance in all three experiments.*

Supplemental Tables

Table S1. Analysis of variance of the effect of DSPR rearing density on starvation resistance.

Source	**df**	**SS**	**MS**	**F**	**P**	**% Var. Exp.**
Density	1	1758.00	1758.20	18.15	< 0.0001	0.90
DSPR RIL	19	148747	7828.80	80.83	<< 0.0001	80.2
Density x DSPR RIL	19	8905	468.70	4.84	< 0.0001	4.80
Residual	269	26054	96.90			

Table S2. Analysis of variance of mean starvation resistance in the DSPR.

Source	df	SS	MS	F	P
Subpopulation[a]	3	330782	110261	183.702	< 0.0001
Sex	1	342197	342197	570.121	< 0.0001
Subpopulation x Sex	3	32982	10994	18.317	< 0.0001
Residual	3440	2064747	600		

[a] We note that since most RILs from a given subpopulation were tested in the same batch, batch effects may contribute to some of the subpopulation-to-subpopulation differences we report.

Table S3. Analysis of variance of mean triglyceride level in the DSPR.

Source	**df**	**SS**	**MS**	**F**	**P**
Subpopulation	3	46.81	15.6020	37.099	< 0.0001
Residual	935	393.22	0.4206		

Table S4. Stratification of the top 50 SNPs associated with starvation resistance in the DGRP across five frequency bins.

Study	Sex	Allele Frequency Bin	No. SNPs
This Study	F	0.05 - 0.1	14
This Study	F	> 0.1 - 0.2	9
This Study	F	> 0.2 - 0.3	13
This Study	F	> 0.3 - 0.4	6
This Study	F	> 0.4 - 0.5	8
This Study	M	0.05 - 0.1	11
This Study	M	> 0.1 - 0.2	12
This Study	M	> 0.2 - 0.3	13
This Study	M	> 0.3 - 0.4	3
This Study	M	> 0.4 - 0.5	11
Mackay *et al.* 2012	F	0.05 - 0.1	28
Mackay *et al.* 2012	F	> 0.1 - 0.2	11
Mackay *et al.* 2012	F	> 0.2 - 0.3	5
Mackay *et al.* 2012	F	> 0.3 - 0.4	2
Mackay *et al.* 2012	F	> 0.4 - 0.5	4
Mackay *et al.* 2012	M	0.05 - 0.1	21
Mackay *et al.* 2012	M	> 0.1 - 0.2	13
Mackay *et al.* 2012	M	> 0.2 - 0.3	7
Mackay *et al.* 2012	M	> 0.3 - 0.4	2
Mackay *et al.* 2012	M	> 0.4 - 0.5	7
Everman and Morgan 2018	F	0.05 - 0.1	11
Everman and Morgan 2018	F	> 0.1 - 0.2	13
Everman and Morgan 2018	F	> 0.2 - 0.3	10
Everman and Morgan 2018	F	> 0.3 - 0.4	8
Everman and Morgan 2018	F	> 0.4 - 0.5	8
Everman and Morgan 2018	M	0.05 - 0.1	18
Everman and Morgan 2018	M	> 0.1 - 0.2	13
Everman and Morgan 2018	M	> 0.2 - 0.3	12
Everman and Morgan 2018	M	> 0.3 - 0.4	4
Everman and Morgan 2018	M	> 0.4 - 0.5	3

Table S5. Analysis of variance of mean starvation resistance in the DGRP.

Source	**df**	**SS**	**MS**	**F**	**P**
Sex	1	97213	97213	118.21	< 0.0001
Residual	334	274680	822		

Table S6. Analysis of variance of the effect of preservatives and environment on starvation resistance in the DGRP.

Source	df	SS	MS	F value	P	% Var. Exp.
Environment	1	2400	2400	16.71	< 0.0001	0.30
Preservatives	1	604184	604184	4205.70	< 0.0001	81.17
DGRP Line	11	63653	5787	40.28	< 0.0001	8.55
Environment x Preservatives	1	155	155	1.08	0.30	0.02
Environment x DGRP Line	11	7289	663	4.61	< 0.0001	0.98
Preservatives x DGRP Line	11	14718	1338	9.31	< 0.0001	1.98
Environment x Preservatives x DGRP Line	11	4961	451	3.14	< 0.001	0.67
Residual	327	46976	144			

Table S7. Analysis of variance of activity during the 24-hour period prior to the DAM (Drosophila Activity Monitor) starvation assay.

Source	df	SS	MS	F value	P	Effect Size
Starvation Resistance Rank (High vs. Low)	1	1.37	1.37	12.48	< 0.001	0.31
Sex	1	0.61	0.61	5.57	< 0.05	0.21
Lights On/Off	1	1.55	1.55	14.11	< 0.001	0.33
Starvation Resistance Rank x Sex	1	0.00	0.00	0.01	0.92	0.01
Starvation Resistance Rank x Lights On/Off	1	0.00	0.00	0.00	0.98	0.00
Sex x Lights On/Off	1	1.85	1.85	16.87	< 0.0001	0.36
Starvation Resistance Rank x Sex x Lights On/Off	1	0.02	0.02	0.21	0.64	0.04
Residuals	132	14.49	0.11	0.00		

Table S8. Repeated measures analysis of variance across days for activity during the DAM (Drosophila Activity Monitor) starvation assay for males and females.

Female Activity During Starvation
Between Days

Source	df	SS	MS	F value	P
Starvation Resistance Rank (High vs. Low)	1	11.15	11.15	14.19	< 0.01
Lights On/Off	1	0.66	0.66	0.84	0.39
Starvation Resistance Rank (High vs. Low) x Lights On/Off	1	0.48	0.48	0.61	0.46
Residuals	8	6.29	0.79		

Within Day

Source	df	SS	MS	F value	P
Starvation Resistance Rank (High vs. Low)	1	19.06	19.06	115.58	< 0.0001
Lights On/Off	1	7.10	7.11	43.09	< 0.0001
Starvation Resistance Rank (High vs. Low) x Lights On/Off	1	0.16	0.16	0.99	0.32
Residuals	575	94.81	0.17		

Male Activity During Starvation
Between Days

Source	df	SS	MS	F value	P
Starvation Resistance Rank (High vs. Low)	1	5.47	5.47	13.79	< 0.01
Lights On/Off	1	0.09	0.09	0.23	0.65
Starvation Resistance Rank (High vs. Low) x Lights On/Off	1	0.67	0.67	1.69	0.24
Residuals	6	2.38	0.40		

Within Day

Source	df	SS	MS	F value	P
Starvation Resistance Rank (High vs. Low)	1	5.03	5.04	46.70	< 0.0001
Lights On/Off	1	0.32	0.32	2.93	0.09
Starvation Resistance Rank (High vs. Low) x Lights On/Off	1	0.12	0.12	1.14	0.29
Residuals	471	50.78	0.11		

Table S9. Data from genes mapped to the region under QTL intervals for starvation resistance in the pA and pB DSPR mapping panels and triglyceride level in the pB DSPR mapping panel based on Flybase release version FB2018_1*. Highlighted genes indicate those previously identified in QTL mapping studies of starvation resistance.*

Table S10. Gene ontology analysis of genes that are included within QTL intervals for starvation resistance and triglyceride level.

Starvation Resistance GO Analysis: pA DSPR Panel

Category	Fold Enrichment	FDR
glutathione metabolic process (GO:0006749)	6.91	1.46E-04
cellular modified amino acid metabolic process (GO:0006575)	3.85	1.79E-02

Triglyceride Level GO Analysis: pB DSPR Panel

Category	Fold Enrichment	FDR
heat shock-mediated polytene chromosome puffing (GO:0035080)	42.66	4.25E-04
polytene chromosome puffing (GO:0035079)	38.39	3.35E-04
sensory perception of sweet taste (GO:0050916)	27.42	7.71E-04
detection of chemical stimulus involved in sensory perception of taste (GO:0050912)	19.47	7.16E-04
chaperone cofactor-dependent protein refolding (GO:0051085)	16	5.94E-03
'de novo' posttranslational protein folding (GO:0051084)	16	5.19E-03
protein refolding (GO:0042026)	15.36	5.65E-03
'de novo' protein folding (GO:0006458)	13.24	9.68E-03
cellular response to unfolded protein (GO:0034620)	12.8	4.87E-03
response to unfolded protein (GO:0006986)	12.44	4.78E-03
cellular response to topologically incorrect protein (GO:0035967)	9.95	1.06E-02
response to topologically incorrect protein (GO:0035966)	9.74	1.00E-02
cellular response to heat (GO:0034605)	8.93	5.11E-02
chaperone-mediated protein folding (GO:0061077)	8.93	4.79E-02
response to hypoxia (GO:0001666)	7.22	4.89E-02
sensory perception of taste (GO:0050909)	6.65	2.84E-02

Gene lists used in each analysis included all genes unique to male and female analyses for each trait. GO analysis was performed using the PANTHER Overrepresentation Analysis (Released 2018-05-21) with Fisher's Exact test with FDR multiple test correction. We found no GO enrichment among the genes implicated by QTL mapped for starvation resistance in the pB panel or genes associated with SNPs implicated in the DGRP.

Table S11. Data from GWA, generated from the DGRP Freeze 2.0 pipeline, based on Flybase release version FB2018_1. *All SNPs shown passed the $P < 10^{-5}$ significance threshold; highlighted SNPs passed the FDR threshold of 0.2.*

Supplemental Text

Text S1. *Starvation media recipe.*

1000 ml	De-ionized water
15 g	Agar
12 ml	Propionic/Phosphoric acid mix*
2 g	Tegosept (Genesee Scientific, Cat: 20-258) dissolved in 20 ml 95% EtOH

* Acid mix:

330 ml	De-ionized water
259 ml	Propionic acid
31 ml	o-Phosphoric acid (85%)

Bring agar in water to a boil, reduce heat and simmer for 20 minutes. Remove from heat, stir in acid mix and tegosept. Cool slightly before pouring vials.

Supplemental Files

File S1. All code associated with the bootstrapping analysis of SNPs associated with starvation resistance measured in the DGRP in this study, Mackay *et al.* (2012), and Everman and Morgan (2018).

File S2. *Description of each dataset associated with this study.*

README for Datafiles (in RawData.zip) Accompanying Everman et al. 2019

DSPR_Data/:

DSPR_1.txt

Tab-separated txt file of raw DSPR Starvation and Desiccation Resistance reported in hours for each fly per experimental vial.
Data are presented in Figure 1, Figure 3, Figure 6, Figure S2, Figure S5, Figure S7, Figure S11.
Column headers:

Mapping.Panel = DSPR, Drosophila Synthetic Population Resource
Trait = StarvationResistance or DesiccationResistance
RIL.ID = Recombinant Inbred Line ID from the DSPR
VialID = Unique number corresponding to each experimental vial
Sex = (m) Male or (f) Female
FlyID = Unique identifier for individual flies in each vial
LifespanHrs = Lifespan in hours for each individual fly

DSPR_2.txt

Tab-separated txt file of raw Founder Starvation Resistance reported in hours for each fly per experimental vial.
Data are presented in Figure 1.
Column headers:

Trait = StarvationResistance
DSPR.founder = DSPR founder line ID
Sex = (m) Male or (f) Female
RepVial = Vial replicate number
FlyID = Unique identifier for individual flies in each vial
LifespanHrs = Lifespan in hours for each individual fly

DSPR_3.txt

Tab-separated txt file of raw female DSPR Triglyceride Levels reported per well.
Data are presented in Figure 5, Figure 6, Figure S3.
Column headers:

Mapping.Panel = DSPR, Drosophila Synthetic Population Resource
Trait = TriglycerideLevel
RIL.ID = Recombinant Inbred Line ID from the DSPR
NumericPlateID = Unique number corresponding to each plate
WellID = ID corresponding to each well of the plate
SampleID = Unique identifier for samples
TrueSerumTriglyConc = Triglyceride level per sample based on five females per sample

DSPR_4.txt

Tab-separated txt file of raw DSPR Starvation Resistance reported in hours using Drosophila Activity Monitors (DAM).
Data are presented in Figure S7.
Column headers:

Mapping.Panel = DSPR, Drosophila Synthetic Population Resource
Trait = StarvationResistance
RIL.ID = Recombinant Inbred Line ID from the DSPR
StarvationClass = Categorical variable (HighStarvClass or LowStarvClass) based on data from DSPR_1.txt
Sex = (M) Male or (F) Female
MonitorID = Unique identifier for DAM
MonitorTubeID = Unique identifier for each tube in each DAM
LifespanHrs = Lifespan in hours for each fly

DSPR_5.txt

Tab-separated txt file of DSPR Activity reported under non-stressful conditions using Drosophila Activity Monitors (DAM).
Data are presented in Figure 4.
Column headers:

Mapping.Panel = DSPR, Drosophila Synthetic Population Resource
RIL.ID = Recombinant Inbred Line ID from the DSPR
Trait = Activity
Sex = (m) Male or (f) Female
StarvationClass = Categorical variable (HighStarvClass or LowStarvClass) based on data from DSPR_1.txt

N = Number of flies tested
ActLight.Mean = Mean activity levels under light conditions
ActLight.SD = Standard deviation for activity levels under light conditions
ActDark.Mean = Mean activity levels under dark conditions
ActDark.SD = Standard deviation for activity levels under dark conditions

DSPR_6.txt

Tab-separated txt file of DSPR Activity reported under starvation conditions using Drosophila Activity Monitors (DAM).
Data are presented in Figure S13.
Column headers:

Mapping.Panel = DSPR, Drosophila Synthetic Population Resource
RIL.ID = Recombinant Inbred Line ID from the DSPR
LightStatus = (L) light or (D) dark
SamplingPeriod = Day of experiment
FemaleMeanActivity = Mean activity of females
MaleMeanActivity = Mean activity of males

DSPR_7.txt

Tab-separated txt file of DSPR Average Starvation Resistance reported in hours under different rearing density treatments.
Data are presented in Figure S1.
Column headers:

Mapping.Panel = DSPR, Drosophila Synthetic Population Resource
Trait = StarvationResistance
RIL.ID = Recombinant Inbred Line ID from the DSPR
Treatment = Density (60 eggs placed into rearing vials) or Population (females laid eggs for 1-2 days with visual assessment of density)
TotalAdults = Total number of flies that emerged from each vial
VialID = Unique identifier for each experimental vial
N = Number of flies per experimental vial
LifespanHrs = Average lifespan in hours for each experimental vial

DSPR_8.txt

Tab-separated txt file of DSPR % survival on starvation media under different rearing density treatments.
Data are presented in Figure S1.
Column headers:

Mapping.Panel = DSPR, Drosophila Synthetic Population Resource
Trait = StarvationResistance
RIL.ID = Recombinant Inbred Line ID from the DSPR
Treatment = Density (60 eggs placed into rearing vials) or Population (females laid eggs for 1-2 days with visual assessment of density)
TotalAdults = Total number of flies that emerged from each vial
VialID = Unique identifier for each experimental vial
ScreenpointHrs = Hour intervals at which flies in each vial were counted
Survival% = Percent of flies in each vial that were alive at each screenpoint

DSPR_9.txt

Tab-separated txt file of DSPR LOD score from QTL mapping analysis.
Data are presented in Figure S14 and Figure S15.
Column headers:

DSPR.Panel = pA or pB mapping panel from the DSPR
Trait = StarvationResistance or TriglycerideLevels
sex = (m) Male or (f) Female
chr = Chromosome (X, 2L, 2R, 3L, 3R)
Ppos = Position on chromosome based on assembly 5.0
Gpos = Genetic position
LOD = LOD score from QTL mapping analysis

DSPR_10.txt

Tab-separated txt file of DSPR observed and estimated starvation resistance.
Data are presented in Figure 7.
Column headers:

DSPR.founder = pA or pB mapping panel from the DSPR
ObservedStarvationResistance = Mean observed starvation resistance of each founder line
EstimatedStarvationResistance = Weighted estimated mean starvation resistance of each founder line determined from QTL analysis

DSPR_11.txt

Tab-separated txt file of DSPR starvation resistance and triglyceride level after accounting for variation due to haplotype at overlapping peaks.
Data are presented in Figure S16.
Column headers:

FounderHaplotype = Predicted haplotype at overlapping QTL
StarvationResistance = Mean starvation resistance
TriglycerideLevel = Mean triglyceride level
N_Starv = Number of RILs with the corresponding predicted founder haplotype for Starvation QTL
N_Tri = Number of RILs with the corresponding predicted founder haplotype for Triglyceride QTL

##
DGRP_Data/:

DGRP_1.txt

Tab-separated txt file of raw DGRP Starvation Resistance reported in hours for each fly per experimental vial. Average response per sex and line was used in GWA.
Data are presented in Figure 2, Figure S4, Figure S5, Figure S8, Figure S12, used to generate Figure 17, Figure S18, Figure S19.
Column headers:

Mapping.Panel = DGRP, Drosophila Genetic Reference Panel
Trait = StarvationResistance
RAL.ID = Line ID based on RAL identifier
Bloomington.ID = Bloomington stock ID
Sex = (M) Male or (F) Female
VialID = Unique identifier for each experimental vial
FlyID = Unique identifier for individual flies in each vial
LifespanHrs = Lifespan in hours for each individual fly

DGRP_2.txt

Tab-separated txt file of DGRP Average Starvation Resistance reported in hours in different environments and on different starvation media types.
Data are presented in Figure S9.

Column headers:

Mapping.Panel = DGRP, Drosophila Genetic Reference Panel
Trait = StarvationResistance
RAL.ID = Line ID based on RAL identifier
Environment = Flies were maintained at 25°C with a 12:12hr L:D cycle (25°C_12hr) or 23°C with constant light (23°C_24hr)
Media = Preservatives or NoPreservatives in the starvation media
VialID = Unique identifier for each experimental vial
N = Number of flies per experimental vial
LifespanHrs = Average lifespan in hours for each experimental vial

DGRP_3.txt

Tab-separated txt file of DGRP % survival in different environments and on different starvation media types.
Data are presented in Figure S9.
Column headers:

Mapping.Panel = DGRP, Drosophila Genetic Reference Panel
Trait = StarvationResistance
RAL.ID = Line ID based on RAL identifier
Media = Preservatives or NoPreservatives in the starvation media
Environment = Flies were maintained at 25°C with a 12:12hr L:D cycle (25°C_12hr) or 23°C with constant light (23°C_24hr)
VialID = Unique identifier for each experimental vial
ScreenpointHrs = Hour intervals at which flies in each vial were counted
Survival% = Percent of flies in each vial that were alive at each screenpoint

DGRP_4.txt

Tab-separated txt file of DGRP Starvation Resistance for lines shared between this study, Mackay et al. 2012, and Everman and Morgan 2018.
Data are presented in Figure S8, Figure S10.
Column headers:

RAL.ID = Line ID based on RAL identifier
Sex = (M) Male or (F) Female
LifespanHrs_EvermanetAl.2019 = Mean starvation resistance reported in this study
LifespanHrs_MackayEtAl.2012 = Mean starvation resistance reported in Mackay et al. 2012
LifespanHrs_Everman&Morgan.2018 = Mean starvation resistance reported in Everman and Morgan 2018

DGRP_5.txt

Tab-separated txt file of bootstrap results of the sign of SNPs across DGRP data collected in this study, Mackay et al. 2012 and Everman and Morgan 2018.
These data are compiled from original bootstrap files; code presented in File S1 is formatted to read each file by DataID.
Data are presented in Figure 8.
Column headers:

%SNPs_SameSign = Percent of SNPs that have the same sign between studies following boostrap analysis of random samples of SNPs.
Density = Density calculated from original bootstrap files
Sex = Female or Male
DataID = Dataset ID for plotting in File S1
Comparison = Direction of comparison of SNPs (ThisStudy_vs_MackayEtAl2012, ThisStudy_vs_Everman&Morgan2018, Everman&Morgan2018_vs_ThisStudy, Everman&Morgan2018_vs_MackayEtAl2012, MackayEtAl2012_vs_ThisStudy, or MackayEtAl2012_vs_Everman&Morgan2018)

DGRP_6.txt

Tab-separated txt file of adjusted mean starvation resistance DGRP data collected in this study, Mackay et al. 2012 and Everman and Morgan 2018.
These data are compiled from original GWA-generated files; code presented in File S1 is formatted to read each file by Study.
Data are used in bootstrap analysis.
Column headers:

RAL.ID = Line ID based on RAL identifier
AdjustedMeanStarvationResistance = Adjusted mean phenotype from GWA of each study
Study = Study in which the original starvation resistance data was collected (ThisStudy, MackayEtAl2012, or Everman&Morgan2018)
Sex = (f) female or (m) male

DGRP_7.txt

Comma-separated txt file of SNP frequencies for bootstrap analysis, used in File S1, generated from dgrp.t.txt in code file.

Data are used in bootstrap analysis.
Column headers:

rAF = Reference allele frequency
aAf = Alternate allele frequqncy
SNP = SNP ID

DGRP_8.txt

Large space-separated matrix of SNP calls for bootstrap analysis, used in File S1, called as dgrp.t.txt in code file. Data originally available from dgrp2.gnets.ncsu.edu.
Data are used in bootstrap analysis.
Column headers:

id = Line ID based on RAL identifier, formatted "line_XXX"
*Remaining Columns: SNP id

Additional Files:
/GWAS_AVG_Starvation/

Contains output files for the average starvation resistance calculated for overlapping DGRP lines between this study, Mackay et al. 2012 and Everman and Morgan 2018.

/GWAS_SJM_Starvation/

Contains output files for the average starvation resistance from this study.