

1 **Increased socially-mediated plasticity in gene**  
2 **expression accompanies rapid adaptive evolution**

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22

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25

#### 26 **Statement of authorship**

27 Conceived and designed the experiments: NWB, Sonia P, MGR, MZ. Performed the  
28 experiments: Sonia P, NR. Analyzed the data: NWB, XL, Sonia P, Steve P, YF. Contributed  
29 reagents/materials/analysis/tools: NWB, NR. Wrote the paper: NWB, Sonia P, MGR, MZ.

30

#### 31 **Data accessibility statement**

32 Sequence data has been deposited in the NCBI BioProject database (accession number:  
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34

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## 44 **Abstract**

45 Recent theory predicts that increased phenotypic plasticity can facilitate adaptation as traits  
46 respond to selection. When genetic adaptation alters the social environment, socially-  
47 mediated plasticity could cause co-evolutionary feedback dynamics that increase adaptive  
48 potential. We tested this by asking whether neural gene expression in a recently arisen,  
49 adaptive morph of the field cricket *Teleogryllus oceanicus* is more responsive to the social  
50 environment than the ancestral morph. Silent males (flatwings) rapidly spread in a Hawaiian  
51 population subject to acoustically-orienting parasitoids, changing the population's acoustic  
52 environment. Experimental altering crickets' acoustic environments during rearing revealed  
53 broad, plastic changes in gene expression. However, flatwing genotypes showed increased  
54 socially-mediated plasticity, while normal-wing genotypes exhibited negligible expression  
55 plasticity. Increased plasticity in flatwing crickets suggests a coevolutionary process coupling  
56 socially flexible gene expression with the abrupt spread of flatwing. Our results support  
57 predictions that phenotypic plasticity should rapidly evolve to be more pronounced during  
58 early phases of adaptation.

## 59 **Introduction**

60 Adaptive mutations are likely to cause correlated phenotypic effects that extend beyond  
61 traits directly targeted by selection (Raymond *et al.* 2001). The fate of a new mutation during  
62 establishment and spread will therefore depend on the balance of costs and benefits of  
63 those associated effects, and phenotypic plasticity has been proposed as a mechanism that  
64 can mitigate the costs. Despite more than a century of debate focusing on how plasticity  
65 impacts rates of evolutionary change, the challenge of empirically testing the link between  
66 plasticity and the establishment of new mutations has defied resolution (Baldwin 1896;  
67 West-Eberhard 2005; Ghalambor *et al.* 2007; Scoville and Pfrender 2010; Stoks *et al.* 2015).  
68 An influential model of this process predicts that increased plasticity associated with traits  
69 directly affected by abrupt (“extraordinary”) changes in selection should evolve over tens of  
70 generations, followed by a much longer period during which adaptive, previously plastic,  
71 phenotypes become genetically assimilated (Lande 2009). Increased plasticity can also  
72 increase the likelihood of adaptive evolutionary responses, even if some of the plasticity is  
73 initially counter-selected (Ghalambor *et al.* 2007; 2015).

74         An overlooked and unresolved question about the relationship between plasticity  
75 and rapid adaptive evolution concerns the extended phenotypic consequences of new  
76 mutations. Genomic invasion of mutations of large effect can indirectly cause major social  
77 changes that provoke plastic phenotypic responses, generating coevolutionary feedback  
78 (Bailey 2012). For example, adaptive mutations that affect social behaviour will alter the  
79 social environment as they spread, potentially altering the expression of other traits such as  
80 aggression or mating behaviour that are sensitive to the social environment (Schradin 2013).  
81 Pre-existing plasticity may enable persistence of new mutations with otherwise negative  
82 effects, but provided there is sufficient genetic variation for that plasticity, it could also

83 coevolve with adaptive mutations if they alter the environment that cues plastic responses  
84 (West-Eberhard 2005; Lande 2009). This scenario requires only a new genotype under  
85 selection that creates environmental feedback, plus genetic variation for plasticity, and it  
86 makes testable predictions about how plasticity modulates the rate of evolution.

87         We tested these predictions by capitalizing on the recent and rapid spread of a male-  
88 silencing wing morph in the Pacific field cricket (*Teleogryllus oceanicus*). Silence protects  
89 males in Hawaii from attack by an acoustically-orienting parasitoid fly, *Ormia ochracea*, and  
90 the phenotype, flatwing, segregates as a Mendelian trait on the X chromosome (Zuk *et al.*  
91 2006; Tinghitella 2008; Pascoal *et al.* 2014). Males who carry flatwing mutation(s) develop  
92 wings that are incapable of normal sound production. These flatwing males appeared in  
93 2003 and spread to near-fixation over approximately 20 generations, so dynamics of this  
94 system reflect the early stages of rapid adaptive evolution (Zuk *et al.* 2006). Flatwing males  
95 are protected from parasitoid attack, but they face difficulty in mate attraction because in  
96 this species, male calling song is the only known long-range mating signal and females  
97 cannot sing. Male song thus constitutes a dominant component of the social environment,  
98 and plasticity mediated by the acoustic environment appears to be advantageous in *T.*  
99 *oceanicus* populations that contain a large proportion of flatwing males. Females reared in  
100 environments lacking song are more responsive, which may enable them to compensate for  
101 the lack of signalling males by responding more quickly and with less discrimination to the  
102 few calling males that remain in the population (Bailey and Zuk 2008). Males reared in  
103 silence invest less in reproductive tissues but are more likely to adopt alternative  
104 reproductive tactics that increase the likelihood of encountering females (Bailey *et al.* 2010),  
105 present decreased immunity (Bailey *et al.* 2011) and show increased locomotion (Balenger  
106 and Zuk 2015).

107           Here, we asked whether enhanced socially-mediated plasticity is associated with the  
108 rapidly-evolving flatwing genotype, as theoretical arguments and models predict (West-  
109 Eberhard 2005; Lande 2009). We quantified transcriptome plasticity to the social  
110 environment in crickets that did or did not carry alleles for flatwing, and tested whether the  
111 genotypes respond to the social environment differently. We specifically evaluated the  
112 effects of prior social experience during development and maturation, rather than an  
113 instantaneous or short-term response as might be activated during mate choice and  
114 phonotaxis (Immonen and Ritchie 2012). We focused on longer-term effects of the acoustic  
115 environment because such exposure mimics variation that crickets would experience while  
116 developing in wild populations dominated by singing normal-wing males or silent flatwing  
117 males.

118           We examined socially-mediated gene expression using tissue derived from cricket  
119 heads, which comprised central and peripheral nervous tissues plus associated sensory  
120 structures contained within the head capsule, assayed during a relevant developmental  
121 interval of adulthood. In crickets, head capsule tissue contains the central brain structures,  
122 which themselves contain approximately 100 times more cells than any one of the ganglia  
123 distributed along the ventral nerve cord (Schildberger *et al.* 1989). We examined gene  
124 expression in tissue contained within head capsules (hereafter referred to as 'neural' or  
125 'brain' tissue for convenience) because we were interested in genes and transcripts that  
126 might influence behavioural responses to the acoustic environment. Such responses need  
127 not rely exclusively on gene expression in the brain, but the tissue-specificity of our  
128 approach allowed us to exclude expression differences that might be associated with  
129 downstream effects of the obvious morphological variation between morphs (Zuk *et al.*  
130 2006; Pascoal *et al.* 2014).

131 Examining neural expression allowed us to bypass difficulties that can arise from  
132 selecting and measuring plasticity of traits at the level of organismal phenotype. A growing  
133 literature focuses on how genomic approaches to the study of phenotypic plasticity can  
134 illuminate causal expression differences underlying plastic responses (Aubin-Horth and Renn  
135 2009), or differential expression arising as a downstream consequence of earlier plastic  
136 changes (Aubin-Horth et al. 2005; Nyman et al. 2017). Others have characterized gene  
137 expression differences underlying environmentally-induced polyphenisms, as in morphs of  
138 the locust *Locusta migratoria* (Wang et al. 2014) or alternative male phenotypes in the bulb  
139 mite *Rhizoglyphus robini* (Stuglik et al. 2014). The present study had a different aim: our  
140 tests were focused on the prediction that rapid adaptation is facilitated by associated  
141 increases in phenotypic plasticity, and we focused on plasticity's relationship with a  
142 genetically-determined polymorphism evolving under selection. Thus, we tested whether  
143 flatwing and normal-wing genotypes show different neural transcriptome responses to the  
144 social environment in *T. oceanicus*, which would provide evidence that transcriptome  
145 plasticity to the social environment is coevolving with the segregating trait, flatwing, which  
146 directly alters that social environment. Our findings support the theoretical prediction that  
147 increased phenotypic plasticity characterizes early stages of rapid adaptation.

148

## 149 **Material and methods**

### 150 **Crickets and acoustic environment manipulation**

151 We used 3 replicate lines each of Kauai pure-breeding flatwing and normal-wing *T. oceanicus*  
152 to test whether neural gene expression in mutant and normal-wing crickets responds  
153 differently to changes in the acoustic environment. The lines were generated through a  
154 series of crosses to ensure homozygosity at the locus or loci causing the flatwing genotype

155 (the phenotype segregates as an X-linked, single locus trait), but the lines were not isogenic  
156 (Zuk *et al.* 2006; Pascoal *et al.* 2014; Pascoal *et al.* 2016a). Stock crickets were reared in 16 L  
157 plastic containers under common garden conditions in a temperature-controlled chamber at  
158 25 °C with a 12:12h light:dark cycle. They were provided with moistened cotton and  
159 cardboard egg cartons for shelter and fed Burgess Supa Rabbit Exel Junior and Dwarf rabbit  
160 pellets *ad libitum*. When sex differences became apparent, males and females were isolated  
161 in 118 mL plastic cups and thereafter reared individually and maintained twice weekly as for  
162 the stock crickets. Isolated crickets were randomly assigned to one of four temperature-  
163 controlled incubators under two treatments. We adapted previously-described methods  
164 (Kasumovic *et al.* 2011; Thomas *et al.* 2011; Bailey and Zuk 2012; Bailey and Macleod 2014;  
165 Pascoal *et al.* 2016b) to manipulate crickets' perceptions of their acoustic environment. Two  
166 incubators were kept in silence ("no song" treatment mimicking a population with few or no  
167 normal-wing males) and two incubators playing back two different average Kauai male  
168 calling songs simultaneously ("song" treatment) mimicked a population with a high density  
169 of singing males. Average calling song parameters were determined from laboratory  
170 recordings made at  $25 \pm 2$  °C of  $n = 24$  normal-wing males from a Kauai stock population,  
171 and the two average Kauai songs were artificially constructed by excising pulses representing  
172 the correct length and carrier frequency from recordings, and manually arranging them into  
173 the required pattern of pulse intervals (Table S1). Since *T. oceanicus* are mainly active at  
174 night, we played back song only during the dark phase of the crickets' light:dark cycle. All  
175 conditions other than the presence or absence of song were kept uniform in the two  
176 treatments. Just after adult eclosion, the left wing scrapers were removed from all crickets  
177 to prevent singing which would interfere with the silent treatment (flatwing males and



178 females cannot sing but were also clipped to control for confounding effects due to cutting).

179 One week later, cricket tissues were dissected and stored in RNALater at -20 °C.

180

### 181 **RNA extractions, library preparation and sequencing**

182 RNA extraction, library preparation and sequencing were performed as described in Pascoal

183 *et al.* (2016a). Briefly, we extracted total RNA from cricket heads (n=48; 3 biological

184 replicates for each sex, morph, social treatment and incubator, Table S2) using TRIzol plus

185 RNA purification kits (Life Technologies) and PureLink DNase treatment (Invitrogen),

186 followed by Qubit (Invitrogen) and Bioanalyser (Agilent) quantification and quality control.

187 We depleted total RNA with RiboZero following the manufacturer's protocol. Purified RNA

188 was checked for depletion and then libraries were constructed using the ScriptSeq protocol

189 (Epicentre). After fragmentation and conversion to cDNA, samples were purified with

190 Ampure XP beads, barcoded, PCR amplified for 14 cycles, and multiplexed. We checked

191 quantity and quality of final pools and performed qPCR using Illumina Library Quantification

192 Kits (Kapa) on a Roche Light Cycler LC480II. Denatured DNA was loaded at 9 pM with 1%

193 fragmented phage PhiX DNA spiked-in, then sequenced on an Illumina HiSeq 2000 (2×100 bp

194 paired end reads).

195

### 196 **RNA-seq data analysis**

197 Data analysis was conducted following the same pipeline as described in Pascoal *et al.*

198 (2016a). Briefly, CASAVA version 1.8.2 (Illumina), Cutadapt version 1.2.1 (Martin 2011) and

199 Sickle version 1.200 with a minimum window quality score of 20 were used for initial

200 processing and quality control of the data (Table S3). We used Trinity (Grabherr *et al.* 2011)

201 to create a combined transcriptome assembly using *in silico* normalisation of trimmed read  
202 data and a k-mer size of 25bp (Table S4). In common with other transcriptome assemblies,  
203 we recovered a large number of contigs and unitigs (Grabherr et al. 2011) (Table S4). These  
204 may relate to different isoforms or different exons deriving from the same gene, and  
205 differential expression of these transcripts between genes may therefore reflect differences  
206 in either transcription or splicing of genes, both of which may be biologically important.  
207 Quantification of transcript abundances was done with RSEM (Li and Dewey 2011): reads  
208 were mapped to the *de novo* transcriptome assembly using BOWTIE 2 (Langmead and  
209 Salzberg 2012), and expected raw read counts for downstream differential expression (DE)  
210 analysis were generated using the mapping BAM files. Prior to DE analysis, we applied a  
211 minimum expression level filter by only retaining transcripts that had non-zero counts in at  
212 least 6 samples, which is the number of samples in a group and thus the minimum number  
213 of non-zero samples likely to be biologically informative. It is possible to implement  
214 additional filtering by removing transcripts for which expression levels are lower than 1  
215 count per million (cpm) in a specified number of groups; however, this must be balanced  
216 against the anti-conservative effect of increasing the false discovery rate when the number  
217 of DE transcripts recovered is reduced. We therefore present results based on data filtered  
218 as above, but performed additional filtering for the analysis presented in Figure 1 and  
219 verified that it does not qualitatively change the main patterns recovered (Fig. S6).

220           Read numbers mapping to each transcript were modelled with negative binomial  
221 error distributions using edgeR (Robinson *et al.* 2010). We implemented generalized linear  
222 models (GLMs) containing each of the three factors of interest (sex, morph and acoustic  
223 treatment) plus all two-way and three-way interactions. Normalisation factors were  
224 calculated to correct for differences in library size among samples, which might otherwise

225 cause bias in differential gene expression analysis. The “TMM” (Trimmed Mean M-values)  
226 method in edgeR (Robinson *et al.* 2010) was applied, with default parameters. Common,  
227 trended and tag-wise dispersion parameters were estimated. Tagwise dispersion was used  
228 for fold change estimating and significance testing. The estimated log<sub>2</sub> fold change for the  
229 each of the models and contrasts were tested in edgeR using a likelihood-ratios (LR) test  
230 (Wilks 1938). P-values associated with logFC (log<sub>2</sub> fold change) were adjusted for multiple  
231 testing such that genes with a false discovery rate (FDR) adjusted P-value < 5% were defined  
232 as significantly differentially expressed (Benjamini and Hochberg 1995).

233         Pairwise comparisons of major interest (i.e. normal-wing male song vs. normal-wing  
234 male no song; flatwing male song vs. flatwing male no song; normal-wing female song vs.  
235 normal-wing female no song; flatwing female song vs. flatwing female no song; all females  
236 vs. flatwing males and all females vs. normal-wing males) were also tested. To visualise  
237 whether and how overall patterns of gene expression separated samples by sex, genotype  
238 and acoustic treatment, a multidimensional scaling (MDS) plot was drawn using the plotMDS  
239 function in edgeR applied to all transcripts. We used Trinotate ([trinotate.sourceforge.net/](http://trinotate.sourceforge.net/))  
240 to annotate the transcriptome and DE sequences and Blast2GO (<http://www.blast2go.com>)  
241 (Conesa *et al.* 2005) to create gene ontology outputs.

242

### 243 **Nanostring validation**

244 To validate the RNA-seq data, we used Nanostring technology with a subset of 32 target  
245 probes that allowed us to analyse the same 48 samples used for the RNA-seq experiment.  
246 Nanostring technology directly obtains sample read count numbers without the need for  
247 cDNA synthesis and intermediate PCRs. Each selected probe represents an individual  
248 transcript or a group of transcript isoforms with the same gene expression pattern. For the

249 list of probes to test (nCounter CodeSet) we included: i) gene annotations of interest, ii)  
250 transcripts that were simultaneously DE in different contrasts (referred as overlap probes),  
251 iii) up- and down-regulated transcripts for each of the individual contrasts and iv) transcripts  
252 that were not DE in RNA-seq. 100 ng of total RNA, as quantified by Qubit assay, was used for  
253 each hybridization assay in a volume of 5  $\mu$ l. Hybridisation buffer, reporter CodeSet and  
254 Capture probe set was added to each sample and incubated overnight (16-18H) at 65°C,  
255 according to manufacturer's instructions. Samples were handled in groups of 12. After  
256 hybridization, the samples were washed and loaded onto an nCounter cartridge. Each  
257 prepared cartridge was loaded into the counter with the associated CodeSet definition file  
258 allowing count generation for each transcript, including the negative and positive controls.

259         Data analysis was performed using the NanoString software nSolver Analysis  
260 Software Version 2.5.34. Background subtraction was done using all internal Nanostring  
261 negative controls, normalization was obtained using the internal Nanostring positive  
262 controls and 3 reference transcripts that were not DE in the RNA-seq experiment, and fold  
263 change ratios were estimated using data partitioning with NormalMaleSong treatment as  
264 baseline. Different normalization (just using the internal positive controls) and fold change  
265 methods (pairwise) were also tested but did not differ from the previous results. We chose  
266 to use the portioned method for fold change analysis because the same baseline was used in  
267 the RNA-seq global GLM analysis (dataset upon which the CodeSet selection was based). A  
268 direct fold change comparison for the different contrasts (sex, morph and acoustic  
269 treatment) between Nanostring and RNA-seq datasets was performed. Regression and  
270 paired t-test sample analyses were performed in SPSS Statistics 22.

271

## 272 **Results**

### 273 **Neural gene expression**

274 We assembled and characterised *de novo* transcriptomes for *T. oceanicus* (Tables S3-S5),  
275 generating a combined assembly to facilitate differential expression (DE) analysis. *T.*  
276 *oceanicus* lacks an annotated reference genome and is distantly-related to commonly  
277 employed model insects such as *Drosophila melanogaster*, so we performed expression  
278 analyses *de novo* at the level of isoforms. We recovered a characteristically large number of  
279 contigs and unitigs as a result, and we collectively refer to these as ‘transcripts’ for  
280 convenience. Our comparisons did not depend on the presence of annotation information,  
281 so we utilised the entire set of annotated and unannotated transcripts and followed this  
282 with homology-based identity and functional categorization where possible. Nanostring  
283 analysis performed on the same 48 samples used for RNA-seq yielded consistent results (see  
284 Figs. S1 and S2).

285 In a model that combined data from all treatments, sex differences accounted for the  
286 largest number of differentially-expressed neural transcripts (Fig. 1). Gene expression also  
287 differed between flatwing and normal-wing genotypes, and between acoustic treatments  
288 (Fig. 1). Gene Ontology (GO) terms associated with the latter group of socially-mediated  
289 plastic transcripts included sensory perception of sound, smell, touch; locomotion; and  
290 spermatogenesis, which correspond with known behavioural, physiological and  
291 morphological responses to the acoustic environment in this species, in particular the  
292 tendency of males to strategically allocate sperm resources depending on the perceived  
293 presence of rival males (Bailey *et al.* 2010; Gray and Simmons 2013).

294

295 **Flatwing and normal-wing neural transcriptomes respond differently to the acoustic**

296 **environment**

297 There were considerable differences in neural gene expression between flatwing and  
298 normal-wing genotypes, and annotations of interest included *rhomboid*, *hedgehog*, and  
299 *wingless*. Crucially, the morph genotypes showed different neural gene expression responses  
300 to the acoustic treatments. Interaction terms in the global model of gene expression  
301 illustrated the latter point: 7,927 transcripts showed different responses across acoustic  
302 treatments in males versus females (sex\*acoustic treatment interaction), and 6,982  
303 transcripts showed different responses across acoustic treatments in flatwing versus normal-  
304 wing crickets (morph\*acoustic treatment interaction) (Fig. 1).

305         The large number of transcripts that showed different patterns of socially-mediated  
306 transcriptome plasticity in flatwing versus normal-wing genotypes (Fig. 1) supported the  
307 prediction that socially-mediated transcriptome plasticity is coevolving with the genetic  
308 mutation(s) that cause flatwing. Given our interest in the differential sensitivity of flatwing  
309 and normal-wing crickets to the social environment, we followed up our global analysis of  
310 transcriptome variation with individual pairwise contrasts testing differential expression  
311 between “song” and “no song” treatments in each of the four classes of cricket: normal-wing  
312 and flatwing males and females. This analysis was designed to investigate whether and how  
313 sexes and morphs differ in socially-mediated plastic gene expression, and it confirmed our  
314 main result: flatwing and normal-wing genotypes show strikingly different patterns of  
315 transcriptome plasticity (Fig. 2). Very few transcripts were differentially expressed between  
316 acoustic environments in normal-wing crickets, whereas flatwing crickets showed  
317 considerable transcriptomic responses to the social environment (Fig. 2, see also Fig. S3).

318         Thus the dominant pattern underlying transcripts recovered from the  
319 morph\*acoustic interaction term in the main GLM is differential expression in flatwings

320 across social environments, but little to negligible socially-mediated plasticity in normal-wing  
321 crickets. Gene expression also responded differently to the social environment in male  
322 versus female neural tissue: there was no overlap of DE transcripts between the sexes. The  
323 lack of overlap is in agreement with the finding above that a significant number of  
324 transcripts show sexually dimorphic responses to the acoustic environment. While flatwing  
325 genotypes showed greater plasticity than normal-wing genotypes ( $\chi^2 = 767.30$ ,  $df = 1$ ,  $p <$   
326  $0.001$ ), flatwing males showed greater transcriptome sensitivity to the acoustic environment  
327 than flatwing females ( $\chi^2 = 206.32$ ,  $df = 1$ ,  $p < 0.001$ ). The pattern of sex differences was  
328 reversed in normal-wing crickets, although this is based on a very small number of DE  
329 transcripts recovered in the normal-wing comparison ( $n = 15$  in normal-wing females, versus  
330 zero in normal-wing males) ( $\chi^2 = 15.00$ ,  $df = 1$ ,  $p = 0.001$ ).

331 In pairwise comparisons, only 15 transcripts showed socially-mediated plasticity in  
332 normal-wing females. Nevertheless, GO analysis recovered annotations including response  
333 to stimulus and locomotion among these, again consistent with prior findings about  
334 flexibility in female mate choice and searching behaviours. Flatwing males showed 610  
335 differentially expressed transcripts between acoustic treatments and 30% ( $n=179$ ) had  
336 annotations including GO terms such as localization, response to stimulus, signalling,  
337 reproduction, reproductive process, and locomotion. Female flatwings had 201 DE  
338 transcripts but only 6% ( $n=12$ ) had associated annotations; this may reflect male-biased  
339 availability in public datasets.

340 A final set of analyses tested how morph genotype, acoustic treatment effects and  
341 their interaction impacted the transcriptomes of each sex separately. These broadly  
342 supported our previous findings, and indicated that although both sexes show expression  
343 variation depending on whether they carry flatwing vs. normal-wing alleles, the bulk of

344 plastic expression variation between morph genotypes appears to be driven by males. We  
345 interrogated patterns of socially-mediated plasticity between the morphs in greater detail by  
346 performing a clustering analysis of the 5,547 transcripts recovered in the morph\*acoustic  
347 interaction term in the males-only analysis (Fig. 3). This analysis was only done for males  
348 owing to a paucity of differentially-expressed transcripts in females (see Table S6 and Fig.  
349 S4). The analysis produced 11 clusters describing differences in the way that gene expression  
350 was governed by the social environment in normal-wing versus flatwing males. Overall,  
351 expression differences appeared to be more extreme between social environments in  
352 flatwing males, although some transcripts showed reversed patterns of socially-mediated  
353 plasticity. For example, cluster 1 transcripts were downregulated in the “song” treatment  
354 compared to the “no song” treatment in flatwing males, whereas they were upregulated in  
355 the “song” treatment in normal-wing males. A similar reversal occurred in the opposite  
356 direction in cluster 3. Such patterns exemplify crossing reaction norms. In contrast,  
357 transcripts in cluster 7 and 11 appear to be downregulated in the “song” environment in  
358 flatwing males, but with little to no differential expression in normal-wing males. An  
359 assessment of functional annotations associated with transcripts in each cluster revealed  
360 several suggestive patterns related to behavioural phenotypes. For example, both clusters 7  
361 and 11 contained transcripts with GO terms describing locomotor behaviour, and sensory  
362 perception of sound was annotated in clusters 7, 9, 10, and 11. Additional behavioural  
363 annotations included flight from cluster 6, inter-male aggression from cluster 7, and male  
364 courtship from cluster 11.

365         Nearly half (45%) of the 5,547 transcripts implicated in the male morph\*acoustic  
366 interaction had an associated annotation. Metabolic and cellular processes were highly  
367 represented, and biologically relevant recovered GO terms include response to stimulus,



368 developmental process, reproduction, locomotion, reproductive process, behaviour,  
369 immune system process and growth (Fig. S5). These enriched GO terms are suggestive of  
370 differences in the mechanisms by which flatwing and normal-wing genotypes respond to  
371 acoustic cues in their rearing environment. Previous experiments have provided evidence  
372 that each of these processes are affected by exposure to the acoustic environment during  
373 development and rearing, providing corroboration for gene expression data, and potential  
374 candidates for future study of the functional genomics of socially-mediated plasticity.

375

### 376 **Transcriptome feminisation and sex differences in plasticity**

377 The nearly 7,000 transcripts identified as significant in the overall sex\*morph interaction  
378 (Fig. 1) suggested that brain transcriptomes showed different levels of sex-biased expression  
379 in the two morphs. A comparison of differential expression between flatwing males versus  
380 all females, and between normal-wing males versus all females, revealed that there were  
381 fewer sex differences in flatwing male brain transcriptomes compared to normal-wing male  
382 brain transcriptomes (Fig. 4a) ( $\chi^2 = 2011.79$ ,  $df = 1$ ,  $p < 0.001$ ). Flatwing males thus had more  
383 female-like patterns of neural gene expression. We used multidimensional scaling (MDS) to  
384 plot similarities among samples in expression measured across all transcripts (Fig. 4b). The  
385 first and second dimensions separated the sexes and morph genotypes, respectively. As with  
386 the previous analysis, flatwing male brain transcriptomes appeared more female-like than  
387 those of normal-wing males, but this feminisation was most prominent in flatwing males  
388 that had been reared in silence (Fig. 4b). Thus, flatwing males not only showed the greatest  
389 degree of transcriptome plasticity in response to acoustic signals in their environment, but  
390 their exposure to song appeared to mitigate female-like patterns of gene expression in the  
391 brain. Despite the fact that expression of the flatwing phenotype is sex-limited, female

392 carriers of the flatwing mutation(s) also showed altered neural gene expression compared to  
393 normal-wing females. On average, expression patterns differed the most between normal-  
394 wing males and flatwing females, although neural expression differences between  
395 genotypes were more pronounced in males than in females (Fig. 4b).

396         The pattern of transcriptome feminisation in flatwing males is consistent with the  
397 well-documented female-like venation patterns on their forewings (Zuk *et al.* 2006), and it is  
398 notable that both *doublesex* and *fruitless* were identified as differentially expressed between  
399 the sexes. However, female-like expression patterns of flatwing brains are not consistent  
400 with the idea that the causative mutation(s) underlying flatwing exert effects that are strictly  
401 compartmentalised to wing venation. Instead, flatwing and normal genotypes appear to  
402 constitutively differ in the expression of brain transcripts, suggesting widespread genomic  
403 effects associated with the mutation(s) arising either through pleiotropy, linkage  
404 disequilibrium, or coevolution (Pascoal *et al.* 2016a).

405

## 406 **Discussion**

407 There is much debate and controversy concerning the role of phenotypic plasticity in  
408 evolutionary change, and both adaptive and non-adaptive plasticity have been proposed to  
409 increase the likelihood of adaptive evolution (West-Eberhard 2005; Ghalambor *et al.* 2015).

410 Plasticity can create opportunities for divergent selection to act, accelerate responses to  
411 selection, pre-adapt populations to respond to novel selective pressures, increase the  
412 likelihood of diversification, or deflect the effects of selection (West-Eberhard 1989; West-  
413 Eberhard 2003; DeWitt and Scheiner 2004; West-Eberhard 2005; Wund 2012; Zuk *et al.*  
414 2014). These predictions have received mixed empirical support. Comparative work has

415 linked diversity with patterns of ancestral plasticity in spadefoot toad species (Gomez-  
416 Mestre and Buchholz 2006), and patterns of plasticity have been found to recapitulate  
417 macroevolutionary patterns of trait divergence in *Polypterus*, the ray-finned fishes (Standen  
418 *et al.* 2014). Despite the intense interest and focus this topic has received, however,  
419 plasticity is often treated as a static property, rather than an evolvable quantity. For  
420 example, the idea that pre-existing phenotypic plasticity acts as a pre-adaptation is  
421 appealing, and has received support in the cricket system we used here (Bailey *et al.* 2008;  
422 Tinghitella *et al.* 2009; Zuk *et al.* 2014), yet we still do not understand how plasticity interacts  
423 with traits under selection throughout the ongoing process of adaptive evolution. Our  
424 findings in *Teleogryllus oceanicus* reveal a genetic association between a rapidly evolving  
425 genotype and plasticity in neural gene expression supporting the view that plasticity itself is  
426 subject to evolutionary forces, and, in particular, can increase during the early stages of  
427 adaptive evolution in line with theoretical predictions (West-Eberhard 2005; Garland and  
428 Kelly 2006; Lande 2009). Box 1 and Fig. 5 provide a graphical description and explanation of  
429 this process.

430         Prior work has revealed acoustically-mediated plasticity in a broad spectrum of traits  
431 related to mating and reproduction in *T. oceanicus* from the island of Kauai, where alleles  
432 causing the erasure of sound-producing structures on male wings have rapidly spread,  
433 almost always in a manner that would be expected to increase fitness in a silent  
434 environment dominated by silent flatwing males (Zuk *et al.* 2006; Pascoal *et al.* 2014; Zuk *et*  
435 *al.* 2014). The constitutive difference in acoustically-mediated plastic gene expression in *T.*  
436 *oceanicus* crickets carrying flatwing versus normal-wing genotypes is consistent with the  
437 rapid evolution of increased plasticity in neural gene expression in flatwing genotypes –  
438 increased plasticity to the acoustic environment accompanied the rapid spread of flatwing.

439 In contrast, we recovered very few socially-mediated plastic transcripts in crickets carrying  
440 normal-wing genotypes; in individual comparisons for normal-wing males, there were none.  
441 Flatwings of either sex, however, showed hundreds of transcripts DE between social  
442 environments. While it is possible that a single, or very few, transcripts could control  
443 responses to the social environment at the phenotypic level in female crickets carrying  
444 normal-wing genotypes, for example if some genes within regulatory networks exert greater  
445 control over such plasticity than others, they nevertheless exhibited a different pattern of  
446 neural transcriptome plasticity than females carrying the recently-derived flatwing  
447 genotype. Both the order of magnitude difference in the number of socially-cued DE  
448 transcripts between morph genotypes in pairwise comparisons and the existence of nearly a  
449 dozen distinct expression clusters in the morph\*acoustic environment interaction for males,  
450 indicated that numerous genetic modules are implicated in responses to acoustic social cues.

451 It is unclear whether the socially-mediated plasticity in gene expression we have  
452 documented is causally linked to adaptive phenotypic responses. For example, enhanced  
453 adaptive plasticity is expected following episodes of rapid adaptation to extreme  
454 environmental pressures (Lande 2009), although this may be accompanied by the release of  
455 cryptic genetic variation for both adaptive and non-adaptive plasticity (Fischer et al. 2016). In  
456 situations where non-adaptive plastic responses to environmental change enhance  
457 responses to directional selection by exposing cryptic variation, those plastic responses that  
458 persist in newly-adapted populations may be of lower magnitude, but are likely to lie along  
459 adaptive phenotypic trajectories (Ghalambor *et al.* 2015, though see Crispo et al. 2010). We  
460 note that exposure to song in the acoustic environment of *T. oceanicus* appeared not to  
461 change neural transcriptomes in the same direction as morph-associated changes, but  
462 instead predominately shifted transcriptome profiles along a sex-biased gene expression axis

463 (x-axis on MDS plot in Fig. 4b) in a male-biased direction.

464 Evidence from other systems suggests that stress responses may represent a  
465 frequent underlying mechanism for acoustically-induced expression changes. Acoustically-  
466 mediated plasticity has been suggested to facilitate adaptive responses to the presence of  
467 signalling rivals in other cricket species (*T. commodus*; Kasumovic et al. 2011) and to  
468 anthropogenic noise pollution in birds (the nightingale *Luscinia megarhynchos*; Brumm  
469 2004). In *Drosophila melanogaster*, courtship song signals activate stress-related gene  
470 expression pathways (Immonen and Ritchie 2012), and in the zebrafish *Danio rerio*, gene  
471 expression changes in the inner ear have been linked to recovery from trauma caused by  
472 over-exposure to extremely loud (179 dB) stimuli (Schuck et al. 2011). A future objective in *T.*  
473 *oceanicus* is therefore to determine whether enhanced brain transcriptome plasticity  
474 associated with flatwing genotypes is causally linked to adaptive phenotypic responses,  
475 either as a mechanistic driver of those responses or as a consequence of them (Mateus *et al.*  
476 2014; Aubin-Horth et al. 2005).

477 We would not have expected a difference in plastic responses of flatwing and  
478 normal-wing genotypes if the average genotype in the population had been subjected to  
479 similar selection favouring the rapid evolution of socially-mediated plasticity. It appears that  
480 the initial spread of flatwing was facilitated by pre-existing plasticity, followed by further  
481 differential selection on plasticity in flatwing versus normal-wing genotypes. It is important  
482 to note that pre-existing genotypic variation in plasticity is necessary for plasticity to  
483 subsequently evolve: the existence of reaction norm variation prior to dramatic  
484 environmental change favouring increased plasticity is a key assumption of the Lande (2009)  
485 model. There is evidence for such reaction norm variation in *T. oceanicus* (Bailey and Zuk  
486 2012), and it seems likely that the different morphs experience distinct selective pressures

487 because of the differences in both parasitoid attack rates and mating tactics employed by  
488 either type of male (Zuk *et al.* 2006). Because of the short timeframe in which the evolution  
489 and spread of flatwing has taken place, the difference in plasticity between flatwing versus  
490 normal-wing genotypes strongly suggests a pleiotropic effect of flatwing allele(s) or loci  
491 maintained in linkage disequilibrium. Rapid evolution of *de novo* physical linkage is an  
492 unlikely scenario. Two intriguing possibilities are that both morphs may demonstrate  
493 plasticity at the level of observable reproductive or physiological phenotypes, yet be subject  
494 to different environmental triggers or neurogenomic mechanisms of socially-mediated  
495 plasticity, or that selection has favoured canalized responses to the social environment in  
496 normal-wing genotypes, with correspondingly different consequences for plastic changes in  
497 the brain transcriptome (Cardoso *et al.* 2015).

498         The constitutive differences in how flatwing and normal-wing transcriptomes  
499 respond to cues in the social environment support key theoretical predictions about the  
500 coevolution of plasticity with novel adaptations. Lande (2009) and others (West-Eberhard  
501 2005; Garland and Kelly 2006) predict a rapid evolutionary increase in plasticity at the onset  
502 of dramatic environmental changes. In Hawaiian *T. oceanicus*, the acoustic environment  
503 underwent an abrupt and profound change because of the rapid spread of silent males: in  
504 the span of several dozen generations, the population on Kauai shifted from one in which  
505 long-range acoustic signals were the dominant mode of social communication, to a  
506 population effectively depauperate in song (Zuk *et al.* 2006). Feedback between the rapid  
507 change from a song-rich to a silent environment, and plasticity in response to the acoustic  
508 environment, appears to have created a situation favourable for the rapid coevolution of  
509 socially-cued plasticity and alleles that cause the silent flatwing phenotype. Over time,  
510 genetic assimilation is predicted to more permanently link these traits, but it is likely to

511 occur on the order of hundreds to thousands of generations, not dozens (Box 1) (Lande  
512 2009). Similar feedback effects are pervasive in evolving systems (Crespi 2004), and the  
513 relationship between flatwing and transcriptome plasticity in *T. oceanicus* demonstrates  
514 how the general impact of phenotypic plasticity on evolutionary change in other systems is  
515 likely to be inextricably linked to its own coevolution with traits under selection.

516

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525

## 526 **References**

- 527 Aubin-Horth N, Renn SCP. 2009. Genomic reaction norms: using integrative biology to understand  
528 molecular mechanisms of phenotypic plasticity. *Mol Ecol.* 18: 3763-3780.
- 529 Aubin-Horth N, Landry CR, Letcher BH, Hofmann HA. 2005. Alternative life histories shape brain gene  
530 expression profiles in males of the same population. *Proc Roy Soc Lond B.* 272: 1655–1662.
- 531 Baldwin JM. A new factor in evolution. 1896. *Am Nat.* 30: 441-451, 536-553.
- 532 Bailey NW. 2012. Evolutionary models of extended phenotypes. *Trends Ecol Evol.* 27: 561-569.

533 Bailey NW, Macleod E. 2014. Socially flexible female choice and premating isolation in field crickets  
534 (*Teleogryllus* spp.) J Evol Biol. 27:170-180

535 Bailey NW, Zuk M. 2012. Socially flexible female choice differs among populations of the Pacific field  
536 cricket: geographic variation in the interaction coefficient  $\psi$  ( $\Psi$ ). Proc Roy Soc Lond B. 279: 3589-  
537 3596.

538 Bailey NW, Gray B, Zuk M. 2011. Exposure to sexual signals during rearing increases immune defence  
539 in adult field crickets. Biol Lett. 7: 217-220.

540 Bailey NW, Gray B, Zuk M. 2010. Acoustic experience shapes alternative mating tactics and  
541 reproductive investment in male field crickets. Curr Biol. 20: 845-849.

542 Bailey NW, McNabb JR, Zuk M. 2008. Preexisting behavior facilitated the loss of a sexual signal in the  
543 field cricket *Teleogryllus oceanicus*. Behav Ecol. 19: 202-207.

544 Bailey NW, Zuk M. 2008. Acoustic experience shapes female mate choice in field crickets. Proc Roy  
545 Soc Lond B. 275: 2645-2650.

546 Balenger SL, Zuk M. 2015. Roaming Romeos: male crickets evolving in silence show increased  
547 locomotor behaviours. Anim Behav. 101: 213-219.

548 Benjamini Y, Hochberg Y. 1995. Controlling the false discovery rate: a practical and powerful  
549 approach to multiple testing. J Roy Stat Soc B. 57: 289-300.

550 Brumm H. 2004. The impact of environmental noise on song amplitude in a territorial bird. J Anim  
551 Ecol. 73:434-440.

552 Cade W. 1975. Acoustically orienting parasitoids: fly phonotaxis to cricket song. Science. 190: 1312-  
553 1313.

554 Cardoso SD, Teles MC, Oliveira RF (2015) 2015. Neurogenomic mechanisms of social plasticity. J Exp  
555 Biol. 218: 140-149.



556 Crespi BJ. 2004. Vicious circles: positive feedback in major evolutionary and ecological transitions.  
557 Trends Ecol Evol. 19: 627-633.

558 Crispo E, DiBattista JD, Correa C, Thibert-Plante X, McKellar AE, Schwartz AK, Berner D, De León LF,  
559 Hendry AP. 2010. The evolution of phenotypic plasticity in response to anthropogenic disturbance.  
560 Evol Ecol Res. 12:47-66.

561 Conesa A, Götz S, García-Gómez JM, Terol J, Talón M, Robles M. 2005. Blast2GO: a universal tool for  
562 annotation, visualization and analysis in functional genomics research. Bioinform. 21: 3674–3676.

563 DeWitt TJ, Scheiner SM. 2004. Phenotypic plasticity: functional and conceptual approaches. Oxford  
564 University Press, Oxford.

565 Fischer EK, Ghalambour CK, Hoke KL. 2016. Can a network approach resolve how adaptive vs  
566 nonadaptive plasticity impacts evolutionary trajectories? Int Comp Biol 56:877-888.

567 Garland Jr T, Kelly SA. 2006. Phenotypic plasticity and experimental evolution. J Exp Biol. 209: 2344-  
568 2361.

569 Ghalambor CK, Hoke KL, Ruell EW, Fischer EK, Reznick DN, Hughes KA. 2015. Non-adaptive plasticity  
570 potentiates rapid evolution of gene expression in nature. Nature. 525: 372-375.

571 Ghalambor CK, McKay JK, Carroll S, Reznick DN. 2007. Adaptive versus non-adaptive phenotypic  
572 plasticity and the potential for contemporary adaptation in new environments. Funct Ecol. 21:  
573 394-407.

574 Gibson G, Dworkin I. 2004. Uncovering cryptic genetic variation. Nat Rev Genet. 5: 681-690.

575 Gomez-Mestre I, Buchholz DR. 2006. Developmental plasticity mirrors differences among taxa in  
576 spadefoot toads linking plasticity and diversity. Proc Natl Acad Sci USA. 103: 19021-19026.

577 Grabherr MG, Haas BJ, Yassour M, Levin JZ, Thompson DA, Amit I, et al. 2011. Full-length  
578 transcriptome assembly from RNA-Seq data without a reference genome. Nature Biotech. 29: 644-  
579 652.

580 Gray B, Simmons LW. 2013. Acoustic cues alter perceived sperm competition risk in the field cricket  
581 *Teleogryllus oceanicus*. Behav Ecol. 24: 982-986.

582 Immonen E, Ritchie MG. 2012. The genomic response to courtship song stimulation in female  
583 *Drosophila melanogaster*. Proc R Soc Lond B. 279:1359-1365.

584 Kasumovic MM, Hall MD, Try H, Brooks RC. 2011. The importance of listening: juvenile allocation  
585 shifts in response to acoustic cues of the social environment. J Evol Biol. 24:1325-1334.

586 Lande R. 2009. Adaptation to an extraordinary environment by evolution of phenotypic plasticity and  
587 genetic assimilation. *J Evol Biol*. 22: 1435-1445.

588 Langmead B, Salzberg SL. 2012. Fast gapped-read alignment with Bowtie 2. Nature Method. 9: 357-  
589 359.

590 Li B, Dewey CN. 2011. RSEM: accurate transcript quantification from RNA-Seq data with or without a  
591 reference genome. BMC Bioinform. 12: 323.

592 Logue DM, Abiola IO, Rains D, Bailey NW, Zuk M, Cade WH. 2010. Does signalling mitigate the costs  
593 of agonistic interactions? A test in a cricket that has lost its song. Proc Roy Soc Lond B. 277: 2571-  
594 2572.

595 Martin M. 2011. Cutadapt removes adapter sequences from high-throughput sequencing reads.  
596 EMBnet.journal. 17: 10-12.

597 Mateus ARA, Marques-Pita M, Oostra V, Lafuente E, Brakefield PM, Zwaan BJ, Beldade P. 2014.  
598 Adaptive developmental plasticity: Compartmentalized responses to environmental cues and to  
599 corresponding internal signals provide phenotypic flexibility. BMC Biol. 12: 97.

600 Nyman C, Fischer S, Aubin-Horth N, Taborsky B. 2017. Effect of the early social environment on  
601 behavioural and genomic responses to a social challenge in a cooperatively breeding vertebrate.  
602 Mol Ecol. 26: 3186-3203.

603 Pascoal S, Liu X, Ly T, Fang Y, Rockliffe N, Paterson S, et al. 2016a. Rapid evolution and gene  
604 expression: a rapidly-evolving Mendelian trait that silences field crickets has widespread effects on  
605 mRNA and protein expression. *J Evol Biol.* 29: 1234-1246.

606 Pascoal S, Mendrok M, Mitchell C, Wilson AJ, Hunt J, Bailey NW. 2016b. Sexual selection and  
607 population divergence I. The influence of socially flexible cuticular hydrocarbon expression in male  
608 field crickets (*Teleogryllus oceanicus*). *Evolution.* 70: 82-97.

609 Pascoal S, Cezard T, Eik-Nes A, Gharbi K, Majewska J, Payne E, et al. 2014. Rapid convergent evolution  
610 in wild crickets. *Curr Biol.* 24: 1369-1374.

611 Raymond M, Berticat C, Weill M, Pasteur N, Chevillon C. 2001. Insecticide resistance in the mosquito  
612 *Culex pipiens*: what have we learned about adaptation? *Genetica.* 112-113: 287-296.

613 Robinson M, McCarthy D, Smyth G. 2010. edgeR: a Bioconductor package for differential expression  
614 analysis of digital gene expression data. *Bioinform.* 26: 139-140.

615 Rotenberry JT, Swanger E, Zuk M. 2015. Alternative reproductive tactics arising from a continuous  
616 behavioral trait: Callers versus satellites in field crickets. *Am Nat.* 185: 469-490.

617 Rotenberry JT, Zuk M, Simmons LW, Hayes C. 1996. Phonotactic parasitoids and cricket song  
618 structure: an evaluation of alternative hypotheses. *Evol Ecol.* 10: 233-243.

619 Schildberger K, Huber F, Wohlers DW. 1989. Central auditory pathway: neuronal correlates of  
620 phonotactic behavior. In: *Cricket Behavior and Neurobiology* (Eds. Huber F, Moore TE, Loher W.)  
621 Cornell University Press, Ithaca, NY.

622 Schradin C. 2013. Intraspecific variation in social organization by genetic variation, developmental  
623 plasticity, social flexibility or entirely extrinsic factors. *Phil Trans Roy Soc Lond B.* 368: 20120346.

624 Schuck JB, Sun H, Penberthy WT, Cooper NGF, Li X, Smith ME. 2011. Transcriptome analysis of the  
625 zebrafish inner ear points to growth hormone mediated regeneration following acoustic trauma.  
626 *BMC Neurosci.* 12:88.

627 Scoville AG, Pfrender ME. 2010. Phenotypic plasticity facilitates recurrent rapid adaptation to  
628 introduced predators. *Proc Natl Acad Sci USA*. 107: 4260-4263.

629 Standen EM, Du TY, Larsson HC. 2014. Developmental plasticity and the origin of the tetrapods.  
630 *Nature*. 513: 54-58.

631 Stoks R, Govaert L, Pauwels K, Jansen B, De Meester L. 2015. Resurrecting complexity: the interplay  
632 of plasticity and rapid evolution in the multiple trait response to strong changes in predation  
633 pressure in the water flea *Daphnia magna*. *Ecol Lett*. 19: 180-190.

634 Stuglik MT, Babik W, Prokop Z, Radwan J. 2014. Alternative reproductive tactics and sex-biased gene  
635 expression: the study of the bulb mite transcriptome. *Ecol Evol*. 4: 623-632.

636 Thomas ML, Gray B, Simmons LW. 2011. Male crickets alter the relative expression of cuticular  
637 hydrocarbons when exposed to different acoustic environments. *Anim Behav*. 82:49-53.

638 Tinghitella RM, Wang JM, Zuk M. 2009. Preexisting behavior renders a mutation adaptive: flexibility  
639 in male phonotaxis behavior and the loss of singing ability in the cricket *Teleogryllus oceanicus*.  
640 *Behav Ecol*. 20: 722-728.

641 Tinghitella RM. 2008. Rapid evolutionary change in a sexual signal: genetic control of the mutation  
642 'flatwing' that renders male field crickets (*Teleogryllus oceanicus*) mute. *Heredity*. 100: 261-267.

643 Wang X, Fang X, Yang P, Jiang X, Jiang F, Zhao D, et al. 2014. The locust genome provides insight into  
644 swarm formation and long-distance flight. *Nat Comm*. 5: 2957.

645 West-Eberhard MJ. 2005. Developmental plasticity and the origin of species differences. *Proc Natl*  
646 *Acad Sci USA*. 102: 6543-6549.

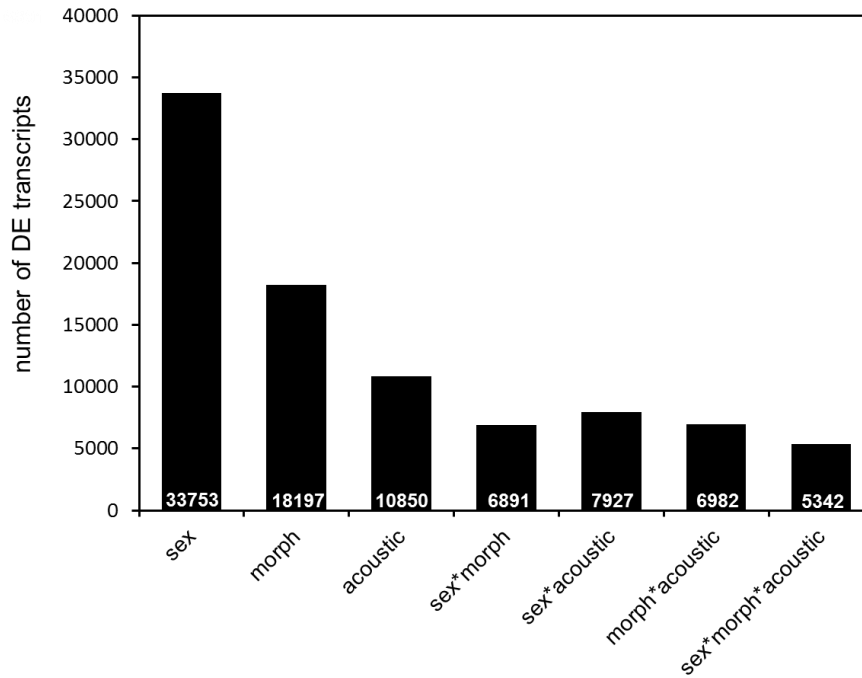
647 West-Eberhard MJ. 1989. Phenotypic plasticity and the origins of diversity. *Annu Rev Ecol Syst*. 20:  
648 249-278.

649 West-Eberhard MJ. 2003. *Developmental plasticity and evolution*. Oxford University Press, Oxford.

- 650 Wilks S. 1938. The large-sample distribution of the likelihood ratio for testing composite hypotheses.  
651 Annal Mathemat Stat. 9: 60-62.
- 652 Wund MA. 2012. Assessing the impacts of phenotypic plasticity on evolution. Integ Comp Biol. 52: 5-  
653 15.
- 654 Zuk M, Bastiaans E, Langkilde T, Swanger E. 2014. The role of behaviour in the establishment of novel  
655 traits. Anim Behav. 92: 333-344.
- 656 Zuk M, Rotenberry JT, Tinghitella RM. 2006. Silent night: adaptive disappearance of a sexual signal in  
657 a parasitized population of field crickets. Biol Lett. 2: 521-524.
- 658 Zuk M, Rotenberry JT, Simmons LW. 1998. Calling songs of field crickets (*Teleogryllus oceanicus*) with  
659 and without phonotactic parasitoid infection. Evolution. 52: 166-171.
- 660 Zuk M, Simmons LW, Cupp, L. 1993. Calling characteristics of parasitized and unparasitized  
661 populations of the field cricket *Teleogryllus oceanicus*. Behav Ecol Sociobiol. 33: 339-343.

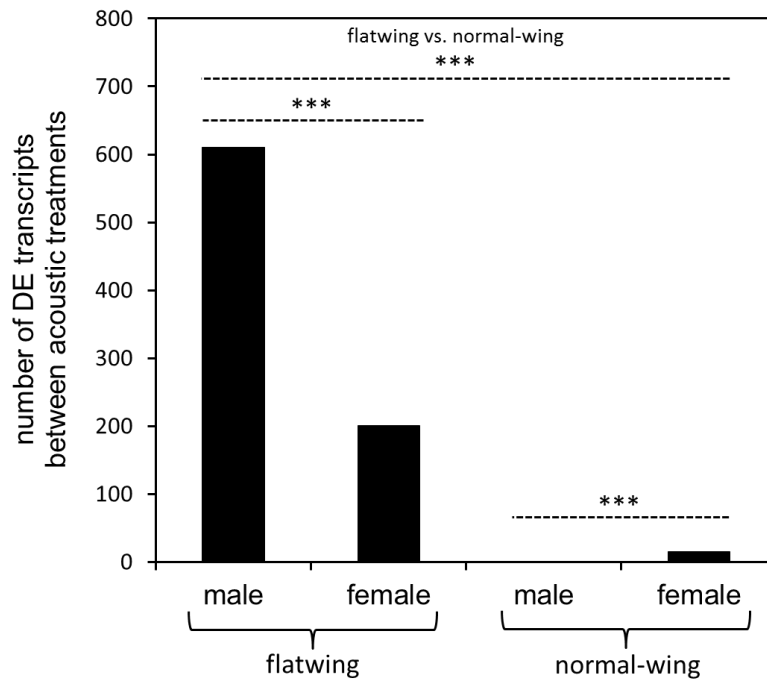
662 **Figures**

663



664

665 **Figure 1 Differential transcript expression in cricket neural tissue.** Expression differences  
666 were inferred using generalized linear models (GLMs). The bars show numbers (given in  
667 white text) of transcripts that were DE between sexes, between wing morphs, and between  
668 acoustic treatments. Interaction terms indicate transcripts whose differential expression was  
669 not heterogeneous, i.e. not in the same direction or magnitude in different groups.



670

671 **Figure 2 Socially-mediated plasticity in gene expression is constitutively different between**

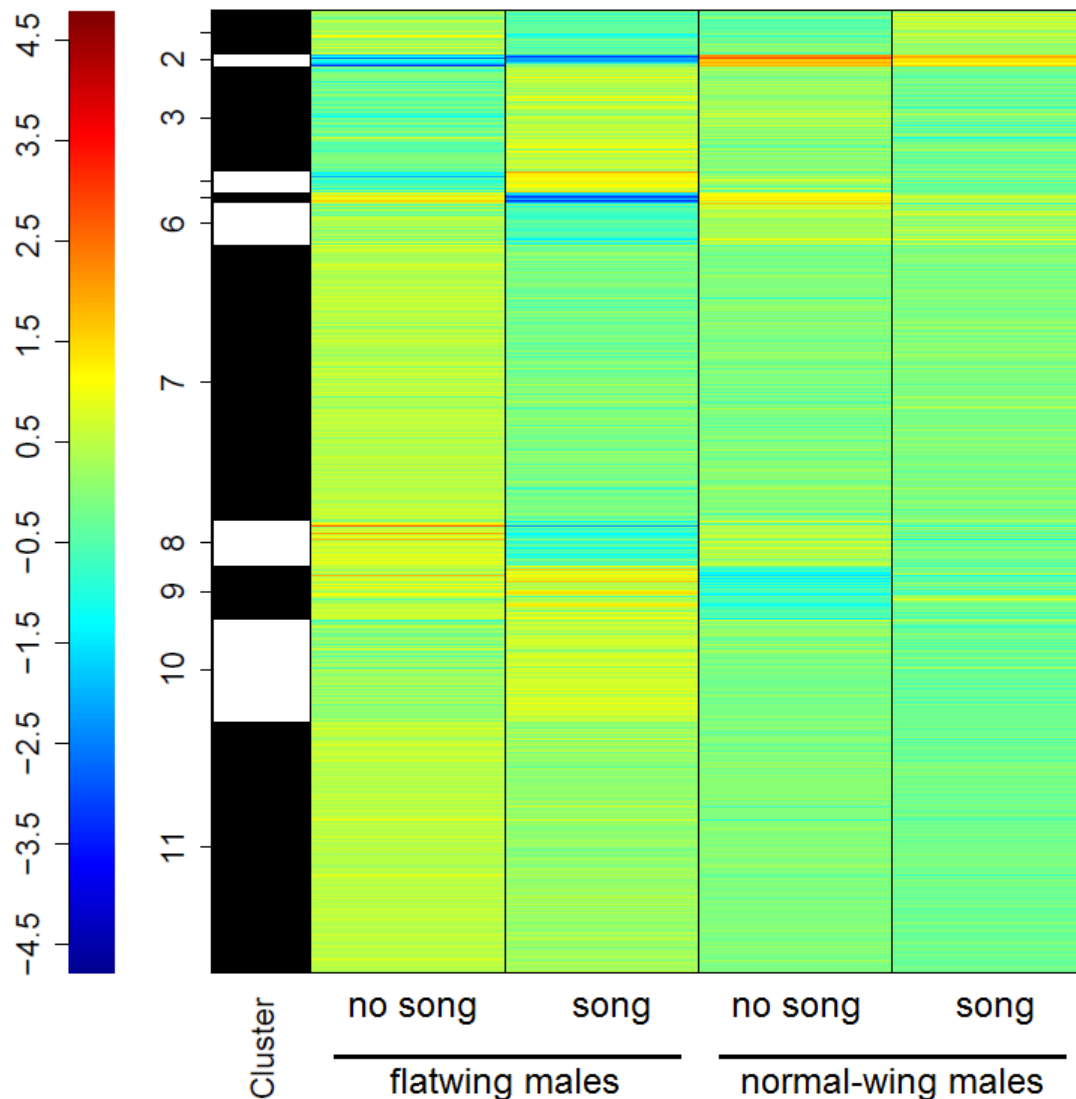
672 **morph genotypes.** The number of differentially-expressed transcripts in the brains of adult  
673 crickets that had been reared in song vs. silence is indicated for each morph and sex.

674 Differential expression was separately assessed for each of the four types of crickets using  
675 pairwise comparisons between the “song” and “no song” acoustic treatment groups.

676 Asterisks highlight significant differences in the proportion of differentially expressed

677 transcripts for the comparisons indicated (Chi-square tests using a total of  $n = 1,545,564$

678 observations for all groups. All  $p < 0.001$  after Bonferroni correction at  $\alpha = 0.0003$



679

680 **Figure 3 Comparison of socially-mediated gene expression in flatwing vs. normal-wing**

681 **males.** Transcripts whose expression was significant in the morph\*acoustic interaction of the

682 male-specific expression analysis are depicted. The significance of the interaction term

683 indicated that the two morph genotypes regulate expression of that transcript differently in

684 response to the acoustic environment. Transcripts are grouped into 11 clusters describing

685 similar patterns of socially-mediated plasticity. The color gradient represents the difference

686 in log<sub>2</sub> fold change compared to the across-treatment average, with larger values (red)

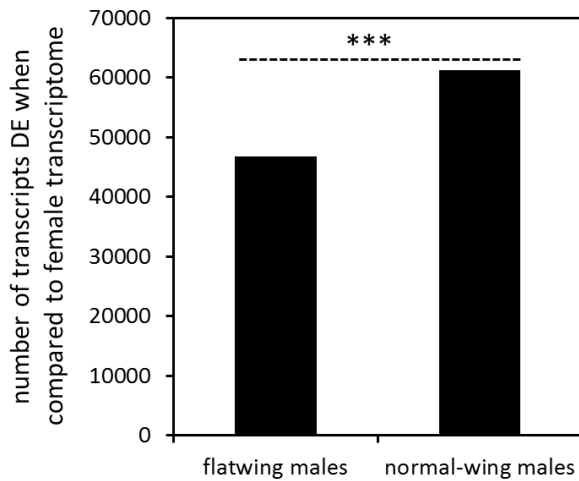
687 indicating up-regulation, and smaller values (blue) indicating down regulation. For each

688 gene, data from all samples are zero-centred to facilitate visual interpretation.

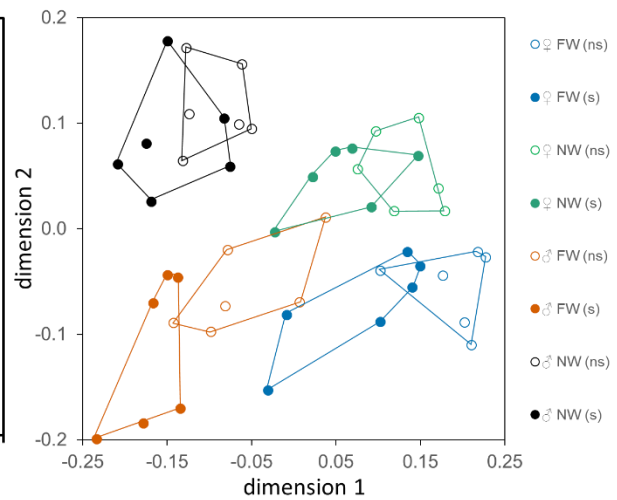
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690 **A**



**B**

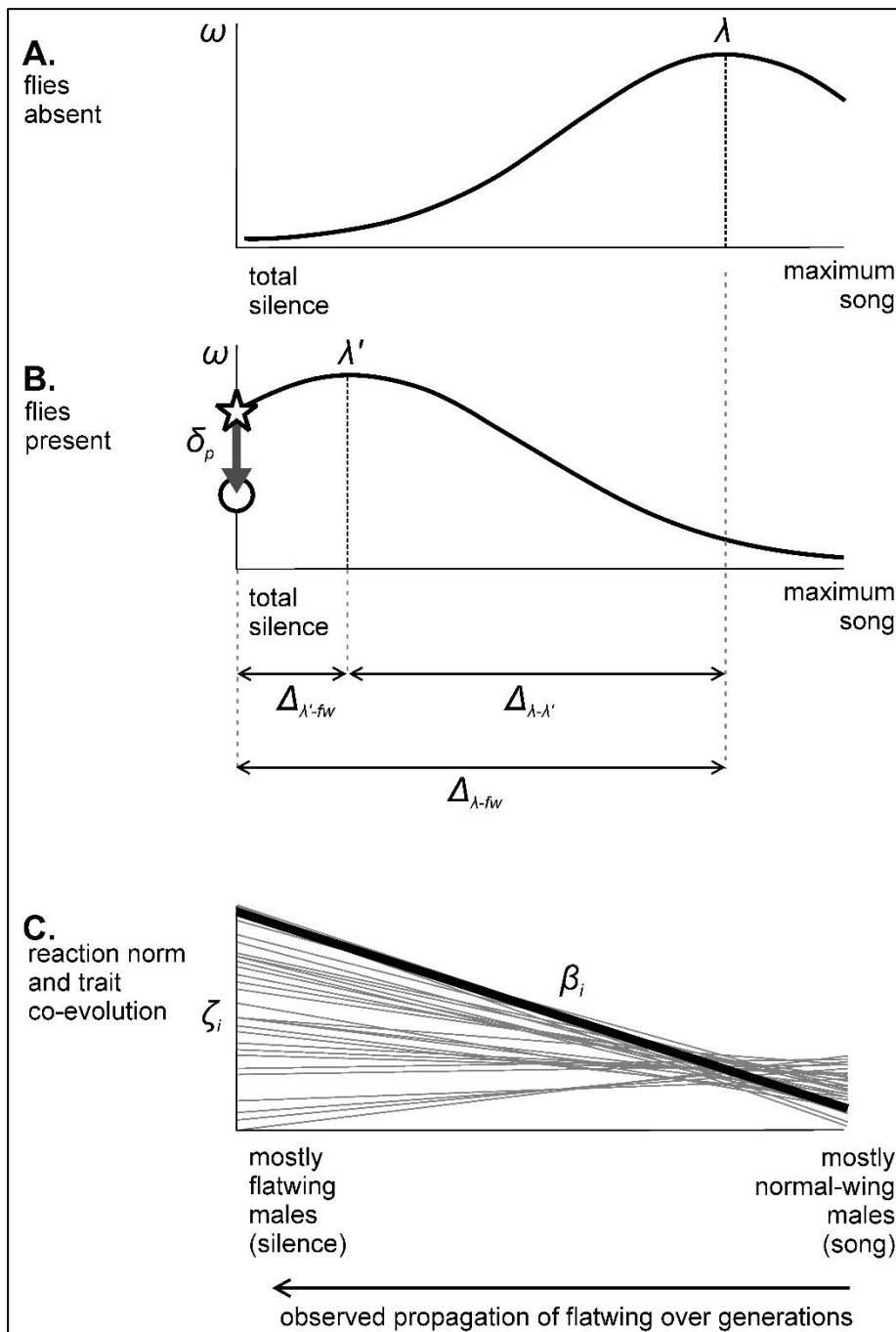


691

692

693 **Figure 4 Neural transcriptomes are feminised in flatwing males. (a)** Number of transcripts  
694 differentially expressed between flatwing males versus all females and between normal-  
695 wing males versus all females. Greater similarity between flatwing males and females than  
696 between normal-wing males and females indicates transcriptional feminisation of flatwing  
697 male neural tissue; asterisks indicate a significant difference ( $\chi^2 = 2011.79$ ,  $df = 1$ ,  $p < 0.001$ ).

698 **(b)** Multidimensional scaling (MDS) plot showing overall patterns of neural gene expression  
699 in each of the 48 samples, for all mapped transcripts. Open symbols represent crickets  
700 reared in silence and solid symbols represent those reared with song. Polygons have been  
701 drawn to enclose all the replicates of each type of cricket. The factors “sex”, “morph”, and  
702 “acoustic treatment” explain 8%, 4%, and 3% of the total variation (Bray distance) in  
703 transcriptome profiles, respectively.



704

705 **Figure 5 Schematic illustration depicting coevolution between phenotypic plasticity and a**  
 706 **novel adaptive phenotype, as described in Box 1.** Panels (a)-(c) illustrate a scenario of rapid  
 707 evolution of male silence in *Teleogryllus oceanicus*, and a hypothetical role for plasticity  
 708 based on Lande 2009.

709

710 **Box 1.** Rapid coevolution of socially-mediated plasticity and a trait under selection. The  
711 evolutionary loss of male song in *Teleogryllus oceanicus* is used as an example (Figure 5).

712  
713 **[A] Hypothetical Gaussian fitness function for male singing tendency in an ancestral**  
714 **environment.** The y-axis represents relative male fitness ( $\omega$ ), which depends on how much  
715 males sing (x-axis). Song is advantageous owing to its role in mate attraction, courtship and  
716 aggression, but energetic and mechanical constraints reduce male fitness beyond an optimal  
717 level of song production,  $\lambda$  (Fig. 5A).

718  
719 **[B] Shift of the optimal male singing tendency when acoustically-orienting parasitoids are**  
720 **present.** The y-axis still represents relative male fitness ( $\omega$ ) and the x-axis how much males  
721 sing. Song still functions in mate acquisition and thus carries a sexually selected benefit.  
722 However, optimal levels of male song production are now lower ( $\lambda'$ ) because of  
723 countervailing natural selection exerted by fatal parasitoids that use it to locate hosts. The  
724 shift in optimum male phenotype along the x-axis is indicated by  $\Delta_{\lambda-\lambda'}$ , and can be  
725 conceptualised as selection on quantitative variation underlying the tendency to sing, by  
726 forcing a shift in the distribution of singers vs. non-singers in the population or alternatively  
727 through a change in average behaviour across males. Early field studies found support for  
728 the latter (Cade 1975; Zuk *et al.* 1993; Rotenberry *et al.* 1996; Zuk *et al.* 1998). Despite the  
729 benefits of song reduction, complete cessation of singing still carries costs, for example  
730 because of the need to acquire mates via other means (Bailey *et al.* 2010, Rotenberry *et al.*  
731 2015) and poorer performance in agonistic encounters (Logue *et al.* 2010).

732 The star indicates the phenotype of obligately silent flatwing males. The invasion of  
733 flatwing allele(s) into the population marks the emergence of a new, discrete phenotype  
734 favoured because it places males closer to the optimal phenotype when flies are present. If  
735 there were no flies, the flatwing male phenotype would carry a severe cost owing to its  
736 distance from the population optimum,  $\Delta_{\lambda-fw}$ , yet when flies are present it clearly confers an  
737 advantage despite having “overshot” the optimal phenotype,  $\Delta_{\lambda'-fw}$ . Flatwing is also known  
738 to cause a range negative pleiotropic effects in males that express it: they cannot advertise  
739 for or court females, and they experience dysfunction in agonistic encounters (Zuk *et al.*  
740 2006; Bailey *et al.* 2008; Logue *et al.* 2010). Flatwing males also have reduced investment in  
741 reproductive tissues (Bailey *et al.* 2010) and partially-feminised cuticular hydrocarbon  
742 profiles (unpublished data). The fitness decrement due to negative pleiotropy in flatwing  
743 males,  $\delta_p$ , is indicated by the solid grey arrow, which shows how the potential maximum  
744 fitness benefits of flatwing (star) exceed the realised fitness benefits (circle). Plasticity to the  
745 changed signalling environment caused by the spread of silent flatwing males is known to  
746 enable males to mitigate consequences of obligate silence, reducing the fitness decrement  
747  $\delta_p$  associated with flatwing (Fig. 5B).

748  
749 **[C] Evolution of phenotypic plasticity during “extraordinary” environmental change caused**  
750 **by proliferation of silent flatwing males.** Here, the y-axis represents a generic trait  $\zeta_i$  that  
751 mitigates negative pleiotropic effects of flatwing by responding to the acoustic social  
752 environment—for example, the tendency of males to adopt satellite mating tactics. The x-  
753 axis now represents the proportion of flatwing males present in the population, which  
754 determines the amount of song present within the environment. Here, we consider the shift  
755 towards a silent social environment an “extraordinary” environmental change, cf. Lande  
756 (2009). An optimal reaction norm with slope  $\beta_i$  is indicated by the thick line, and selection

757 will favour individuals expressing phenotypes close to this line. If there is genetic variation  
758 for plasticity, for example as a result of past environmental stochasticity caused by  
759 demographic fluctuations or environmental signal interference (indicated by “silence” and  
760 “song” in parentheses on the x-axis), then reaction norms for individual genotypes are  
761 predicted to be distributed as indicated by the light grey lines, with little genetic variance  
762 available to selection under ordinary environmental circumstances that characterise  
763 populations rich in singing, normal-wing males, but with increasing exposure of cryptic  
764 genetic variation as the social environment shifts due to the proliferation of flatwing males  
765 (Gibson and Dworkin 2004). As the environment changes (following the lower arrow from  
766 right to left along the x-axis), phenotypes that mitigate negative effects of flatwing (i.e.  
767 reducing  $\delta_p$ ) will be positively selected, favouring reaction norms with increasingly large  
768 slopes  $\beta$ . Short-term reaction norm evolution over a timescale of tens to hundreds of  
769 generations is expected to be rapid, whereas a longer period of genetic assimilation is  
770 predicted to occur subsequently over many thousands of generations (Lande 2009). The  
771 evolution of flatwing crickets in Hawaii is very recent as they appear to have arisen  
772 approximately 15 years ago, thus the rapid spread of flatwings represents the earliest phase  
773 of this process (Zuk *et al.* 2006) (Fig. 5C). Figure based on Lande (2009) (Fig. 1).

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782 **Supplementary Material**

783 *Supplemental Tables*

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785 **Table S1.** Average calling song parameters (10 songs per individual) recorded from 24 normal-wing males from a laboratory Kauai stock  
786 population. Grand means used to construct song playbacks are indicated in bold on the bottom row. *T. oceanicus* song consists of a trill-like  
787 series of higher-amplitude pulses (commonly referred to as the “long chirp”) followed by a series of lower-amplitude pulses typically clustered  
788 into pairs, the “short chirp” (Bailey and Macleod 2013). LC: long chirp; SC: short chirp; IPI: inter-pulse interval; ICI: inter-chirp interval.

ID	# long chirps	# short chirps	total duration	carrier freq.	LC total duration	LC pulse duration	LC pulse interval	LC - SC interval	SC chirp duration	SC pulse duration	SC IPI	SC ICI	SC total duration	inter-song interval
<b>W1</b>	4.9	6.5	1.207	5030.8	0.276	0.034	0.024	0.062	0.068	0.027	0.012	0.069	0.867	0.087
<b>W2</b>	6.6	7.2	1.563	4739.6	0.389	0.039	0.028	0.079	0.078	0.030	0.015	0.085	1.093	0.078
<b>W3</b>	4.1	4.1	0.737	5206.1	0.212	0.030	0.028	0.047	0.066	0.031	0.009	0.064	0.478	0.061
<b>W4</b>	6.0	4.1	0.936	4809.6	0.344	0.031	0.032	0.065	0.070	0.029	0.013	0.078	0.525	0.087
<b>W5</b>	5.6	6.0	1.210	4646.6	0.311	0.034	0.030	0.066	0.060	0.029	0.011	0.082	0.833	0.098
<b>W6</b>	5.9	6.8	1.289	5052.3	0.317	0.032	0.029	0.065	0.058	0.022	0.014	0.081	0.944	0.091
<b>W7</b>	4.7	4.4	0.925	4811.6	0.277	0.037	0.026	0.057	0.070	0.023	0.014	0.063	0.589	0.064
<b>W11</b>	5.5	6.0	1.281	4751.4	0.325	0.040	0.021	0.072	0.079	0.033	0.017	0.081	0.886	0.125
<b>W12</b>	5.9	5.6	1.170	4976.4	0.347	0.038	0.022	0.063	0.067	0.029	0.010	0.070	0.761	0.085
<b>W13</b>	5.8	7.0	1.462	4856.7	0.344	0.034	0.026	0.068	0.064	0.027	0.010	0.079	1.050	0.095
<b>W14</b>	6.5	8.8	1.908	5051.0	0.400	0.038	0.024	0.056	0.069	0.035	0.010	0.088	1.450	0.098
<b>W15</b>	6.4	8.0	1.500	4741.7	0.357	0.029	0.030	0.068	0.064	0.025	0.009	0.068	1.077	0.086
<b>W16</b>	5.2	4.1	0.899	5363.5	0.326	0.035	0.029	0.061	0.073	0.027	0.014	0.065	0.511	0.093
<b>W17</b>	5.2	5.0	0.988	4758.4	0.316	0.039	0.022	0.048	0.071	0.031	0.009	0.057	0.620	0.062
<b>W18</b>	5.7	3.6	0.888	4821.9	0.346	0.037	0.029	0.062	0.081	0.033	0.013	0.064	0.479	0.069
<b>W19</b>	5.1	5.7	1.234	4838.4	0.327	0.038	0.032	0.071	0.079	0.037	0.012	0.078	0.835	0.093
<b>W20</b>	7.3	8.5	1.883	4655.8	0.448	0.033	0.036	0.080	0.076	0.028	0.013	0.080	1.356	0.068
<b>W21</b>	7.9	6.4	1.418	5576.2	0.467	0.036	0.021	0.055	0.072	0.032	0.007	0.071	0.896	0.086
<b>W22</b>	6.2	6.8	1.312	4965.7	0.385	0.042	0.022	0.056	0.076	0.037	0.008	0.064	0.876	0.098
<b>W23</b>	5.8	6.1	1.328	5051.9	0.340	0.073	0.023	0.065	0.071	0.030	0.009	0.092	0.922	0.120
<b>W24</b>	7.6	5.2	1.157	5008.9	0.428	0.037	0.020	0.055	0.072	0.032	0.009	0.067	0.673	0.100
<b>W26</b>	5.3	5.9	1.208	4810.9	0.303	0.039	0.015	0.055	0.068	0.030	0.007	0.074	0.850	0.095
<b>W27</b>	5.1	8.6	1.583	4659.1	0.285	0.034	0.026	0.067	0.072	0.029	0.011	0.072	1.231	0.090
<b>W28</b>	5.8	6.8	1.464	4740.6	0.337	0.035	0.022	0.076	0.077	0.034	0.012	0.081	1.052	0.102
<b>grand mean</b>	<b>5.8</b>	<b>6.1</b>	<b>1.273</b>	<b>4913.5</b>	<b>0.342</b>	<b>0.037</b>	<b>0.026</b>	<b>0.063</b>	<b>0.071</b>	<b>0.03</b>	<b>0.011</b>	<b>0.074</b>	<b>0.869</b>	<b>0.089</b>

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790 **Table S2.** Experimental design summary.

<b>sex</b>	<b>morph genotype</b>	<b>acoustic treatment</b>	<b>incubator</b>	<b>biological replicates</b>
<b>males</b>	normal-wing	song	A	3
			B	3
		no song	C	3
			D	3
	flatwing	song	A	3
			B	3
		no song	C	3
			D	3
<b>females</b>	normal-wing	song	A	3
			B	3
		no song	C	3
			D	3
	flatwing	song	A	3
			B	3
		no song	C	3
			D	3

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792 **Table S3.** Summary of raw and trimmed sequence data. The proportion of trimmed reads that were unpaired after trimming was generally low  
793 (<1%), indicating that the data were of good quality.

Sample	Index	Pool	Untrimmed reads	Trimmed reads	R1/R2 pairs	R0 reads (%) <sup>1</sup>
Sample_65_Normal21_4	ATCACGAT-TCTTTCCC	m10	51213110	50388088	24948575	490938 (0.97)
Sample_66_Normal6_2	CGATGTAT-TCTTTCCC	m10	45735160	45209613	22,385,500	438,613 (0.97)
Sample_67_Normal8_3	TTAGGCAT-TCTTTCCC	m10	53038412	52423136	25,947,709	527,718 (1.01)
Sample_68_Normal6_20	TGACCAAT-TCTTTCCC	m10	62684826	61589560	30,521,397	546,766 (0.89)
Sample_69_Normal8_12	ACAGTGAT-TCTTTCCC	m10	51146242	50240263	24,869,694	500,875 (1.00)
Sample_70_Normal21_11	GCCAATAT-TCTTTCCC	m10	47476394	43246580	21,409,818	426,944 (0.99)
Sample_71_Normal6_14	CAGATCAT-TCTTTCCC	m10	55389280	53792333	26,621,039	550,255 (1.02)
Sample_72_Normal8_16	ACTTGAAT-TCTTTCCC	m10	63735242	62889776	31,101,990	685,796 (1.09)
Sample_73_Normal21_13	GATCAGAT-TCTTTCCC	m10	39795704	39292576	19,426,286	440,004 (1.12)
Sample_74_Normal6_8	TAGCTTAT-TCTTTCCC	m10	39213464	38706053	19,178,113	349,827 (0.90)
Sample_75_Normal8_5	GGCTACAT-TCTTTCCC	m10	40659330	40193367	19,878,742	435,883 (1.08)
Sample_76_Normal21_19	CTTGTAAT-TCTTTCCC	m10	52701328	51845955	25,671,123	503,709 (0.97)
Sample_77_FW11_3	AGTCAAAT-TCTTTCCC	m11	43404602	43103150	21,409,018	285,114 (0.66)
Sample_78_FW17_1	AGTTCCAT-TCTTTCCC	m11	33319756	33046763	16,392,856	261,051 (0.79)
Sample_79_FW23_1	ATGTCAAT-TCTTTCCC	m11	38747894	38381221	19,030,316	320,589 (0.84)
Sample_80_FW11_5	CCGTCCAT-TCTTTCCC	m11	43424404	43095693	21,415,093	265,507 (0.62)
Sample_81_FW17_5	GTAGAGAT-TCTTTCCC	m11	35454676	35145354	17,429,104	287,146 (0.82)
Sample_82_FW23_6	GTGAAAAT-TCTTTCCC	m11	35627960	35253705	17,471,064	311,577 (0.88)
Sample_83_FW11_9	GTGGCCAT-TCTTTCCC	m11	48753966	48021554	23,852,141	317,272 (0.66)
Sample_84_FW17_18	CGTACGAT-TCTTTCCC	m11	46751338	46357206	23,015,914	325,378 (0.70)
Sample_85_FW23_9	GAGTGGAT-TCTTTCCC	m11	39954704	39522528	19,618,474	285,580 (0.72)
Sample_86_FW11_15	CACCGGAT-TCTTTCCC	m11	43454276	43070149	21,375,323	319,503 (0.74)
Sample_87_FW17_16	CACGATAT-TCTTTCCC	m11	50661862	50097667	24,870,404	356,859 (0.71)
Sample_88_FW23_14	TCCCGAAT-TCTTTCCC	m11	47736978	47339870	23,497,415	345,040 (0.73)
Sample_89_Normal6_21	ATCACGAT-TCTTTCCC	m12	48,182,384	47,798,219	23,711,773	374,673 (0.78)
Sample_90_Normal8_1	CGATGTAT-TCTTTCCC	m12	53,957,708	53,488,753	26,530,364	428,025 (0.80)



<b>Sample_91_Normal21_17</b>	TTAGGCAT-TCTTTCCC	m12	45,211,642	44,737,241	22,165,232	406,777 (0.91)
<b>Sample_92_Normal6_18</b>	TGACCAAT-TCTTTCCC	m12	51,896,536	51,385,601	25,508,876	367,849 (0.72)
<b>Sample_93_Normal8_5</b>	ACAGTGAT-TCTTTCCC	m12	51,634,454	50,965,943	25,297,584	370,775 (0.73)
<b>Sample_94_Normal21_18</b>	GCCAATAT-TCTTTCCC	m12	42,417,674	41,797,660	20,735,523	326,614 (0.78)
<b>Sample_95_Normal6_12</b>	CAGATCAT-TCTTTCCC	m12	50,339,926	49,856,365	24,735,866	384,633 (0.77)
<b>Sample_96_Normal8_11</b>	ACTTGAAT-TCTTTCCC	m12	44,997,936	44,361,627	22,008,624	344,379 (0.78)
<b>Sample_97_Normal21_10</b>	GATCAGAT-TCTTTCCC	m12	38,004,450	37,638,634	18,675,819	286,996 (0.76)
<b>Sample_98_Normal6_14</b>	TAGCTTAT-TCTTTCCC	m12	37,369,320	36,638,383	18,167,764	302,855 (0.83)
<b>Sample_99_Normal8_14</b>	GGCTACAT-TCTTTCCC	m12	28,662,486	28,110,963	13,930,418	250,127 (0.89)
<b>Sample_100_Normal21_13</b>	CTTGTAAT-TCTTTCCC	m12	34,753,666	34,101,236	16,909,101	283,034 (0.83)
<b>Sample_101_FW11_2</b>	AGTCAAAT-TCTTTCCC	m13	41,281,256	40,713,985	20,234,480	245,025 (0.60)
<b>Sample_102_FW17_1</b>	AGTTCCAT-TCTTTCCC	m13	43,094,538	42,818,603	21,289,617	239,369 (0.56)
<b>Sample_103_FW23_1</b>	ATGTCAAT-TCTTTCCC	m13	39,692,492	39,412,319	19,581,222	249,875 (0.63)
<b>Sample_104_FW11_6</b>	CCGTCCAT-TCTTTCCC	m13	42,186,768	41,753,255	20,748,837	255,581 (0.61)
<b>Sample_105_FW17_8</b>	GTAGAGAT-TCTTTCCC	m13	41,076,258	40,742,216	20,252,790	236,636 (0.58)
<b>Sample_106_FW23_7</b>	GTGAAAAT-TCTTTCCC	m13	42,621,620	42,178,361	20,970,276	237,809 (0.56)
<b>Sample_107_FW11_11</b>	GTGGCCAT-TCTTTCCC	m13	38,435,250	38,101,992	18,909,722	282,548 (0.74)
<b>Sample_108_FW17_17</b>	CGTACGAT-TCTTTCCC	m13	32,553,664	32,319,194	16,046,684	225,826 (0.70)
<b>Sample_109_FW23_11</b>	GAGTGGAT-TCTTTCCC	m13	40,169,036	39,704,569	19,729,851	244,867 (0.62)
<b>Sample_110_FW11_13</b>	CACCGGAT-TCTTTCCC	m13	41,294,536	40,916,496	20,314,203	288,090 (0.70)
<b>Sample_111_FW17_16</b>	CACGATAT-TCTTTCCC	m13	43,717,190	43,338,556	21,519,773	299,010 (0.69)
<b>Sample_112_FW23_13</b>	TCCCGAAT-TCTTTCCC	m13	41,002,624	40,034,832	19,876,536	281,760 (0.70)

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<sup>1</sup> % of all trimmed reads

796 **Table S4.** Assembly metrics for *de novo* transcriptome constructed using data from 48 sequencing libraries.

<b>Assembled sequences</b>	<b>Size range</b>	<b>Mean length</b>	<b>N50<sup>1</sup></b>	<b>Number of sequences &lt;500 bp</b>	<b>Number of sequences 500 bp – 5 kb</b>	<b>Number of sequences 5 kb – 10 kb</b>	<b>Number of sequences 10 kb – 50 kb</b>	Trinity “genes” with a single “isoform”	Trinity “genes” with multiple “isoforms” (%) <sup>2</sup>
1,545,564	201-20,090	543	712	1,071,927	471,717	1,873	47	1,378,017	167,547 (13.34%)

797 <sup>1</sup> N50 = 50% of bases occur in sequences of this size or larger

798 <sup>2</sup> Calculated as the number of “Components and subcomponents” associated with >1 “seq” (% of total “Components and subcomponents”; n=1,256,357)

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**Table S5.** Summary of trimmed sequence data mapping against *de novo* transcriptome assembly.

Sample	R1+R2 reads (trimmed)	R1+R2 reads mapped to the assembly	% R1+R2 reads mapped to the assembly
Sample_65_Normal21_4	49,897,150	38,105,620	76.37%
Sample_66_Normal6_2	44,771,000	34,452,978	76.95%
Sample_67_Normal8_3	51,895,418	39,197,006	75.53%
Sample_68_Normal6_20	61,042,794	47,335,750	77.55%
Sample_69_Normal8_12	49,739,388	37,080,698	74.55%
Sample_70_Normal21_11	42,819,636	32,965,310	76.99%
Sample_71_Normal6_14	53,242,078	40,366,564	75.82%
Sample_72_Normal8_16	62,203,980	47,127,042	75.76%
Sample_73_Normal21_13	38,852,572	28,302,580	72.85%
Sample_74_Normal6_8	38,356,226	29,574,182	77.10%
Sample_75_Normal8_5	39,757,484	30,169,664	75.88%
Sample_76_Normal21_19	51,342,246	38,445,788	74.88%
Sample_77_FW11_3	42,818,036	32,830,286	76.67%
Sample_78_FW17_1	32,785,712	24,668,246	75.24%
Sample_79_FW23_1	38,060,632	28,749,404	75.54%
Sample_80_FW11_5	42,830,186	33,162,330	77.43%
Sample_81_FW17_5	34,858,208	25,838,696	74.13%
Sample_82_FW23_6	34,942,128	26,527,616	75.92%
Sample_83_FW11_9	47,704,282	37,237,598	78.06%
Sample_84_FW17_18	46,031,828	35,899,416	77.99%
Sample_85_FW23_9	39,236,948	30,365,624	77.39%
Sample_86_FW11_15	42,750,646	33,422,012	78.18%
Sample_87_FW17_16	49,740,808	38,280,754	76.96%
Sample_88_FW23_14	46,994,830	36,001,654	76.61%
Sample_89_Normal6_21	47,423,546	37,130,662	78.30%
Sample_90_Normal8_1	53,060,728	41,761,678	78.71%
Sample_91_Normal21_17	44,330,464	34,546,178	77.93%
Sample_92_Normal6_18	51,017,752	39,657,376	77.73%
Sample_93_Normal8_5	50,595,168	39,503,388	78.08%
Sample_94_Normal21_18	41,471,046	32,403,190	78.13%
Sample_95_Normal6_12	49,471,732	38,614,454	78.05%
Sample_96_Normal8_11	44,017,248	34,804,912	79.07%
Sample_97_Normal21_10	37,351,638	29,378,420	78.65%
Sample_98_Normal6_14	36,335,528	28,426,270	78.23%
Sample_99_Normal8_14	27,860,836	21,636,444	77.66%
Sample_100_Normal21_13	33,818,202	26,369,794	77.98%
Sample_101_FW11_2	40,468,960	32,288,476	79.79%
Sample_102_FW17_1	42,579,234	33,197,186	77.97%
Sample_103_FW23_1	39,162,444	30,570,240	78.06%
Sample_104_FW11_6	41,497,674	32,508,230	78.34%
Sample_105_FW17_8	40,505,580	31,198,968	77.02%
Sample_106_FW23_7	41,940,552	32,857,162	78.34%

<b>Sample_107_FW11_11</b>	37,819,444	29,694,136	78.52%
<b>Sample_108_FW17_17</b>	32,093,368	24,838,774	77.40%
<b>Sample_109_FW23_11</b>	39,459,702	31,028,792	78.63%
<b>Sample_110_FW11_13</b>	40,628,406	32,040,942	78.86%
<b>Sample_111_FW17_16</b>	43,039,546	34,247,024	79.57%
<b>Sample_112_FW23_13</b>	39,753,072	31,194,008	78.47%

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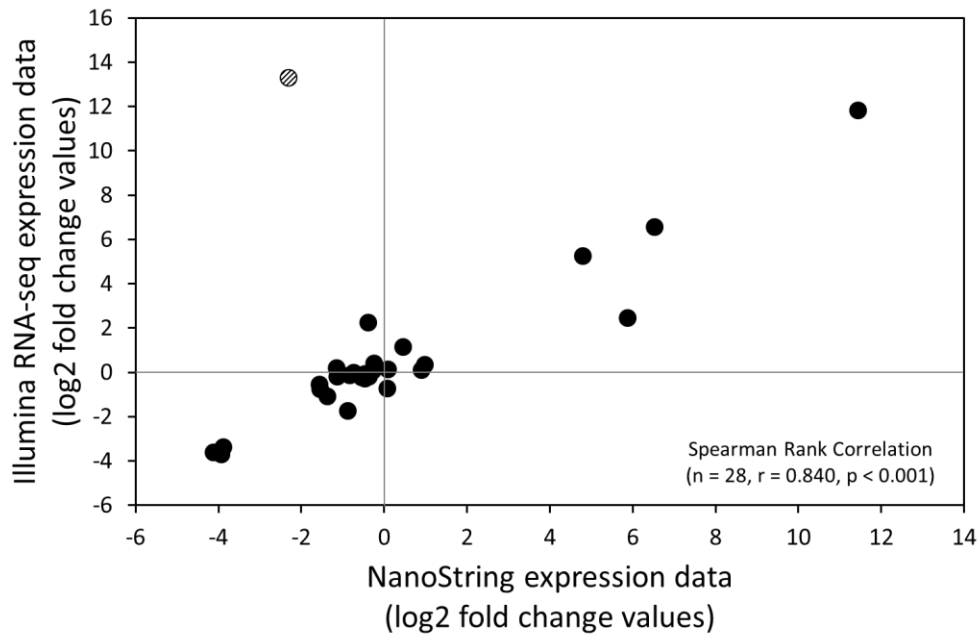
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804 **Table S6.** Generalized linear models performed separately for each sex: summary of DE  
805 transcripts by contrast.

<b>GLM by Sex</b>	<b>Males</b>			<b>Females</b>		
	<b>DE</b>	<b>DE_Up</b>	<b>DE_Down</b>	<b>DE</b>	<b>DE_Up</b>	<b>DE_Down</b>
<b>morph (ref. FW)</b>	16586	4737	11849	463	278	185
<b>acoustic treatment (ref. No Song)</b>	6225	4817	1438	3	1	2
<b>morph x acoustic treatment</b>	5547	4099	1488	2	0	2

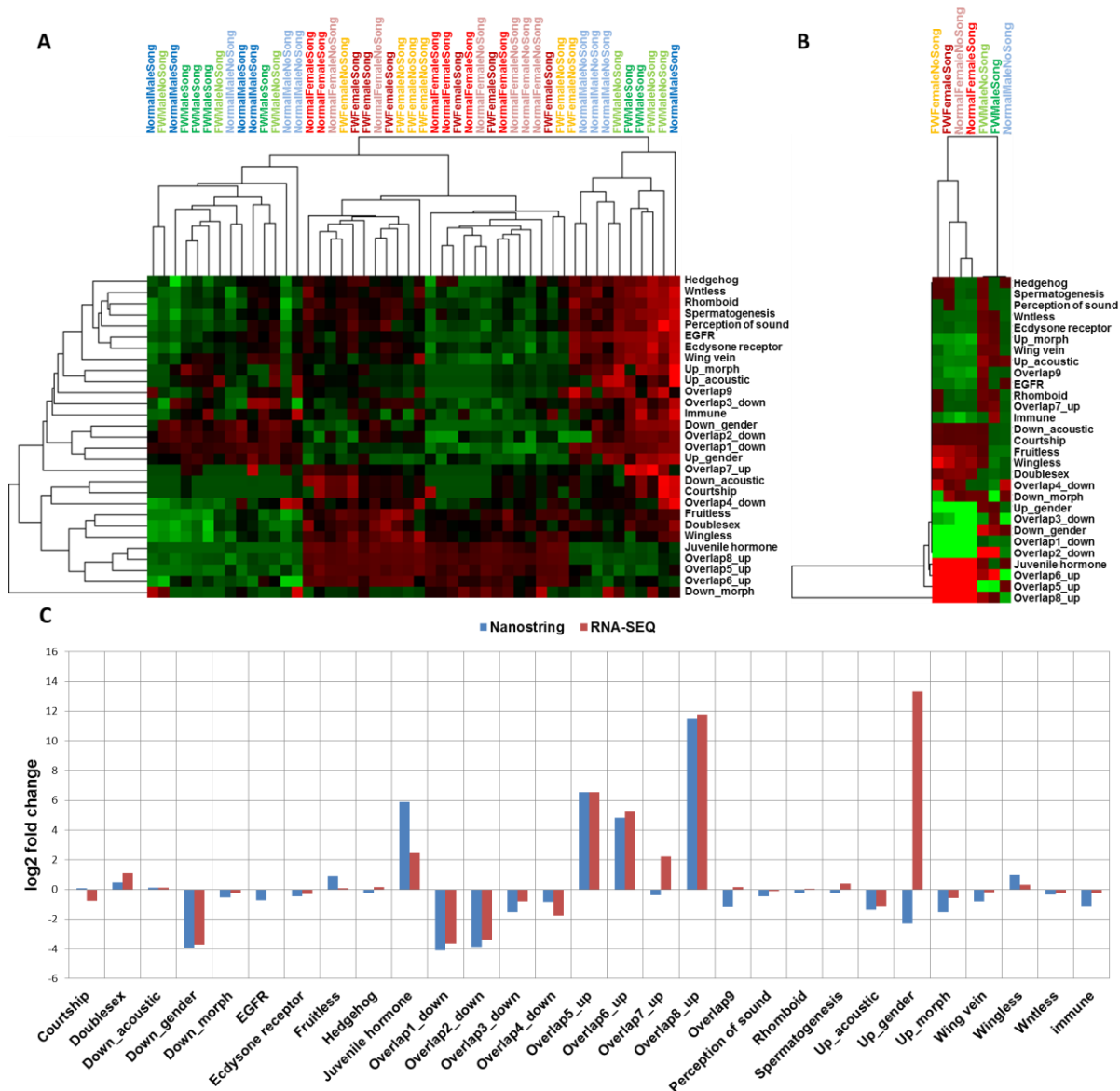
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807 *Supplemental Figures and Results*



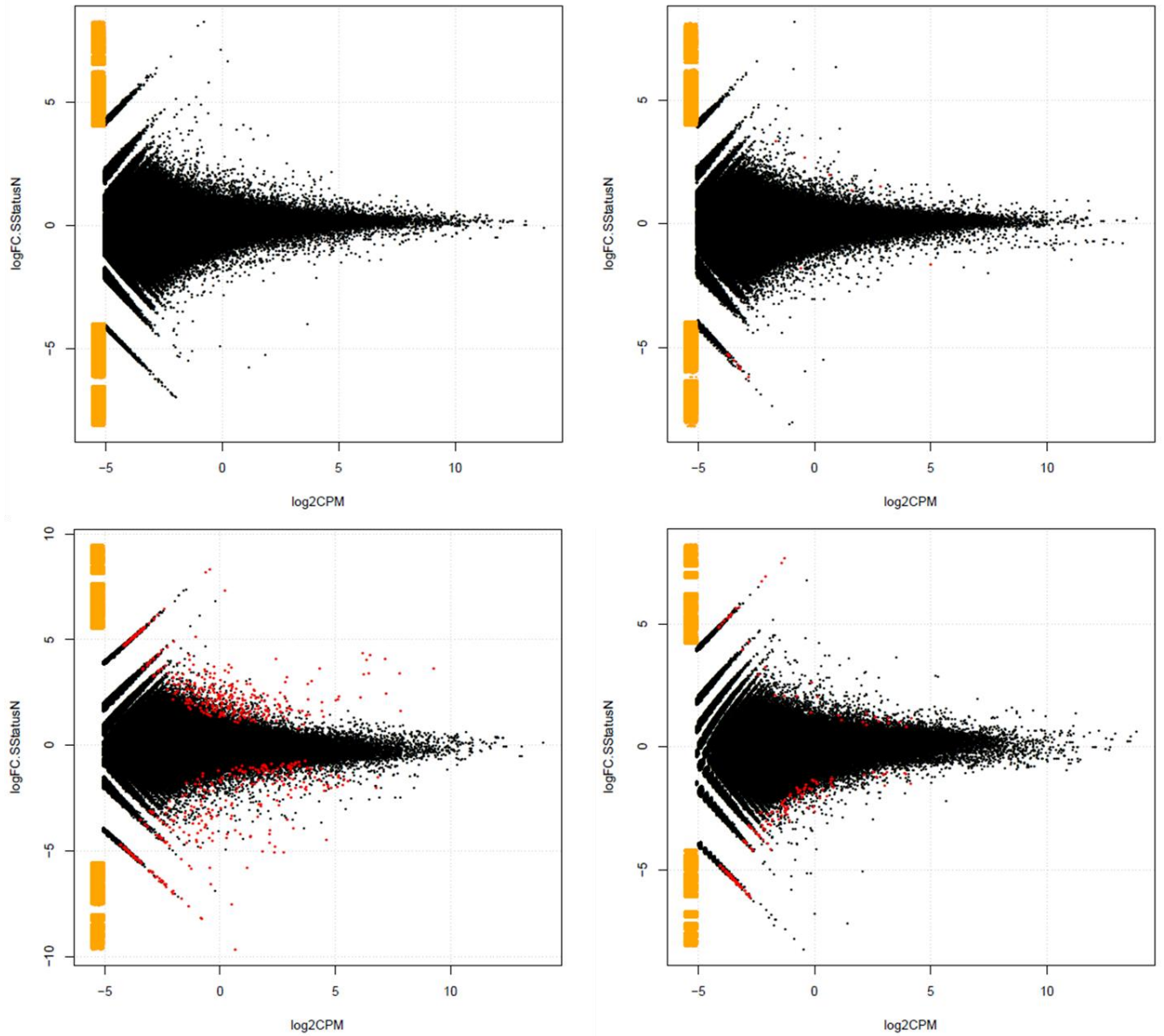
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810 **Figure S1. Comparison of RNA-seq and NanoString expression analyses.** Log2 fold change  
811 values comparing expression between sexes are indicated for each technique. The three  
812 control transcripts that were not DE in the RNA-seq experiment are excluded from the plot,  
813 because they did not have associated log2 fold change values. The hatched datapoint  
814 represents a failed NanoString probe (see Fig. S2 for details). Results of a Spearman Rank  
815 correlation using the remaining data are shown inset. Paired samples t-tests were also  
816 performed for expression data from each technique by the three main contrasts, which  
817 similarly found no evidence for an overall difference in expression patterns quantified using  
818 the two techniques (Sex: n = 28, t = -1.150, p = 0.260 excluding one probe that  
819 malfunctioned; n = 29, t = -1.333, p = 0.193 if it is included. Morph: n = 29, t = 0.258, p =  
820 0.798. Acoustic treatment: n = 29, t = -0.184, p = 0.855).



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**Figure S2. Nanostring validation data based on 32 probes.** (A) Normalized Nanostring data and (B) fold change ratio agglomerative cluster heatmaps. Background subtraction was done using all the negative controls, normalization was obtained using the positive controls and 3 reference transcripts that were not DE in the RNA-seq experiment and fold change ratios were estimated using data partitioning with normal-wing male song treatment as baseline. (C) Fold change comparison between sexes for Nanostring (blue) and RNA-seq (red) datasets for the tested probes. Note that these exclude the three negative controls, as they lacked associated log<sub>2</sub>FC values. The data for the “Up\_gender” probe corresponds to the outlier in Figure S1. The overall low number of read counts for this probe across samples indicated that it likely represented a failed Nanostring probe. Such failures may arise when the folding structure of a transcript prevents access to the probe during annealing.



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835 **Figure S3. MA plots showing significantly, differentially-expressed transcripts between**  
 836 **song versus no-song treatments (in red), separated by cricket type.** Log<sub>2</sub> fold change  
 837 (log<sub>2</sub>FC) is plotted against log<sub>2</sub> counts per million (log<sub>2</sub>CPM). Normal-wing males (top left),  
 838 normal-wing females (top right), flatwing males (bottom left), and flatwing females (bottom  
 839 right).

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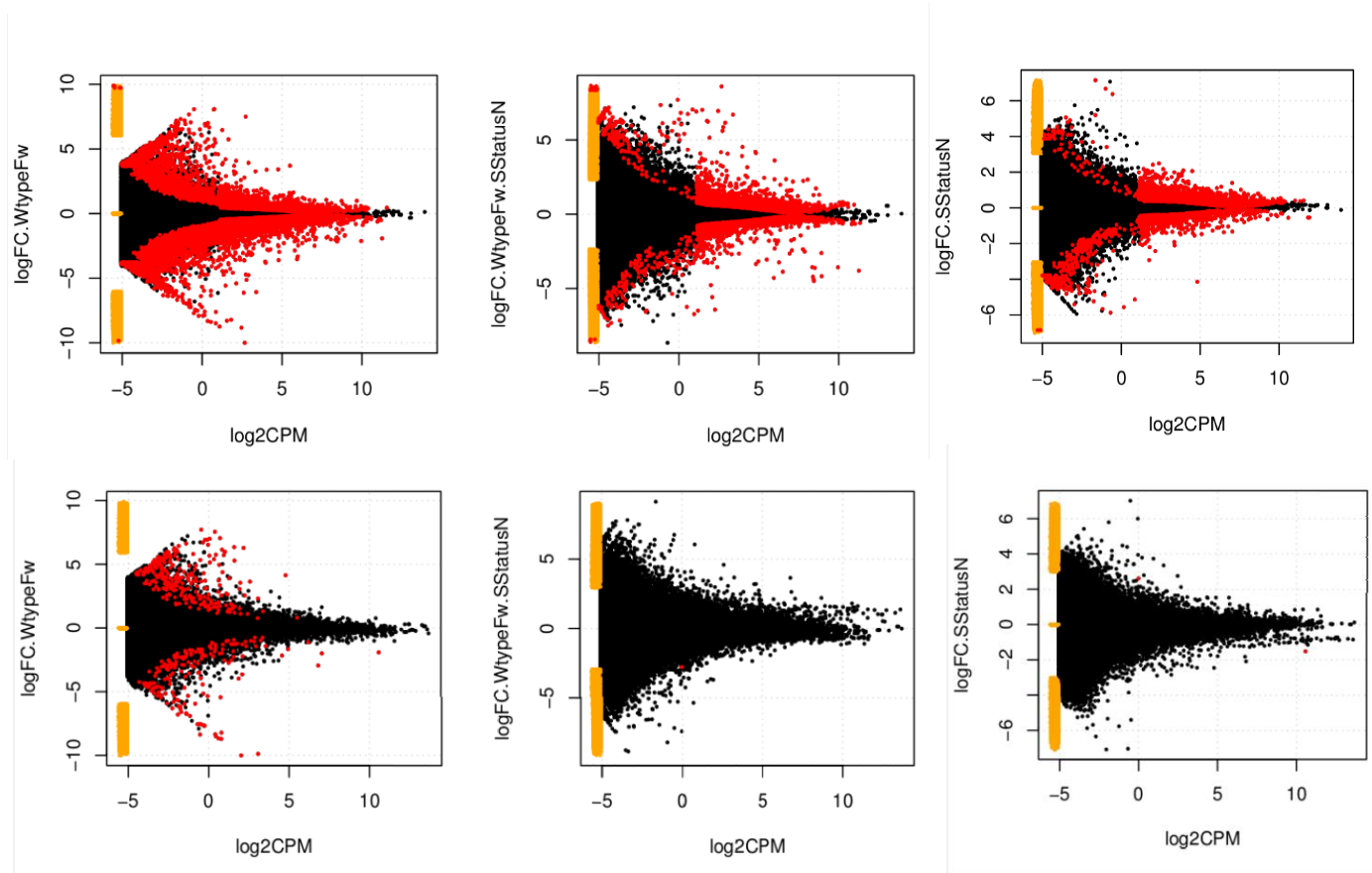
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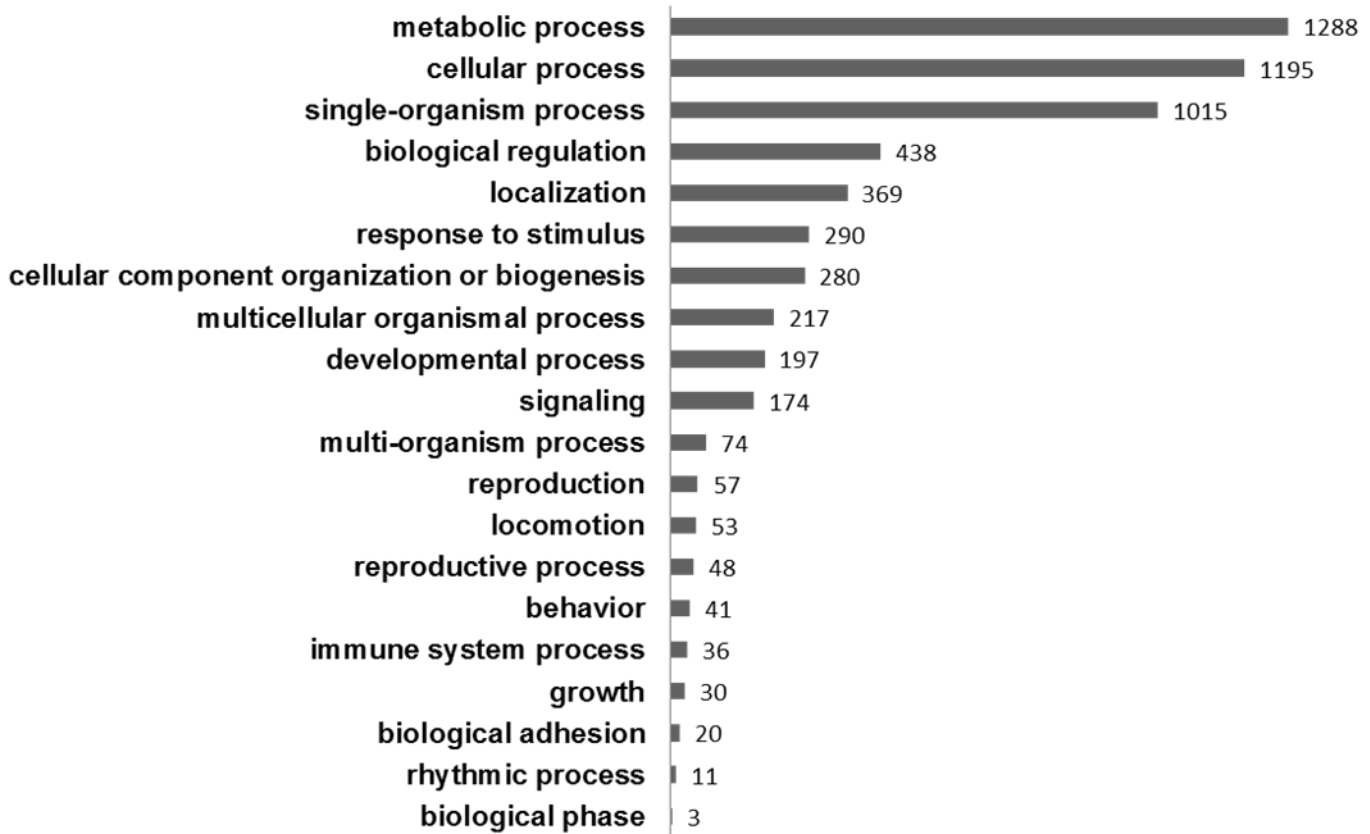


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848 **Figure S4. MA plots showing transcripts that were significantly, differentially expressed in generalized linear models (GLMs) undertaken**  
 849 **separately for each sex.** The top row shows males, and the bottom row shows females. The contrast by morph genotype is shown in the left  
 850 column, the interaction between morph genotype and acoustic treatment is in the middle column, and the acoustic treatment contrast is on  
 851 the right. Low expression genes ( $\log_2\text{CPM} < -5$ ) are colored in orange, and significant DE genes are colored in red. FC: fold change, CPM: counts  
 852 per million, Wtype: wing type (morph), SStatus: song status (acoustic treatment).



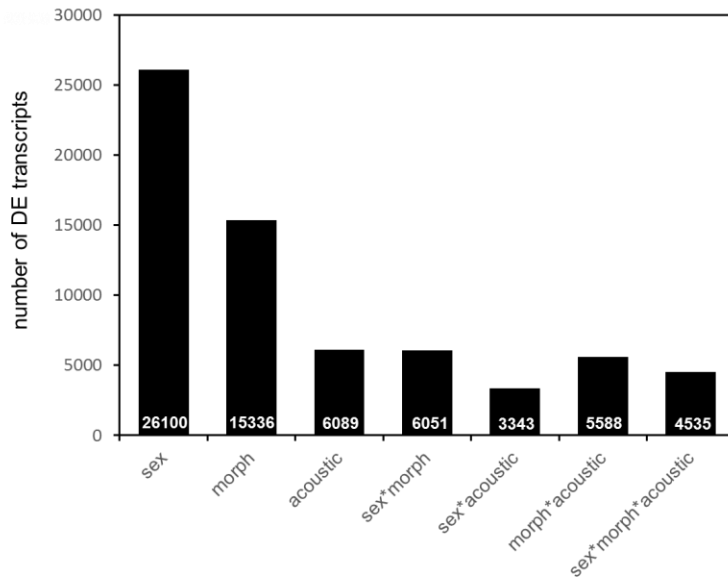
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856 **Figure S5. Gene ontology (GO) terms for the 5,547 transcripts that were significant in the**  
857 **morph\*acoustic treatment interactions in male-only generalized linear models.** Terms for  
858 Biological Process, Level 2 are shown, and thus describe functional annotation information  
859 for transcripts that showed different magnitudes or directions of plastic change in response  
860 to the acoustic environment in normal-wing versus flatwing males.



861  
 862 **Figure S6. Differential transcript expression in the global GLM presented in Figure 1 of the**  
 863 **main text, when a more stringent filter for lowly-expressed transcripts is applied.** In this  
 864 analysis, transcripts were removed if they were not expressed at a level of at least 1 count  
 865 per million in a minimum of 3 samples. Fewer DE transcripts are recovered overall (n =  
 866 36,561 with more stringent filtering, versus n = 46,511), but the general pattern of DE genes  
 867 represented across the GLM contrasts remains similar. The morph\*acoustic interaction is of  
 868 particular importance, because it indicates transcripts for which flatwing versus normal-wing  
 869 genotypes of crickets respond differently to the social environment (i.e. divergent reaction  
 870 norms). More stringent filtering did not decrease the relative fraction of transcripts that  
 871 were significantly DE in this interaction category (15.28% of all DE transcripts with zero-  
 872 count plus 1cpm filter, versus 15.01% of all DE transcripts with zero-count filter only).