

1 Large effective population size masks population genetic structure in  
2 *Hirondellea* amphipods within the deepest marine ecosystem, the  
3 Mariana Trench.

4

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11 Running title: Genetic structure in hadal amphipods

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## 14 Abstract

15 The examination of genetic structure in the deep-ocean hadal zone has focussed on  
16 divergence between the tectonic trenches to understand how environment and geography may  
17 drive species divergence and promote endemism. There has been little attempt to examine  
18 localised genetic structure within trenches, partly because of logistical challenges associated  
19 with sampling at an appropriate scale, and the large effective population sizes of species that  
20 can be sampled adequately may mask underlying genetic structure. Here we examine genetic  
21 structure in the superabundant amphipod *Hirondellea gigas* in the Mariana Trench at depths  
22 of 8126-10545 m. RADSeq was used to identify 3,182 loci containing 43,408 SNPs across  
23 individuals after stringent pruning of loci to prevent paralogous multicopy genomic regions  
24 being erroneously merged. PCA of SNP genotypes resolved no genetic structure between  
25 sampling locations, consistent with a signature of panmixia. However, DAPC identified  
26 divergence between all sites driven by 301 outlier SNPs in 169 loci and significantly  
27 associated with latitude and depth. Functional annotation of loci identified differences  
28 between singleton loci used in analysis and paralogous loci pruned from the dataset and also  
29 between outlier and non-outlier loci, all consistent with hypotheses explaining the role of  
30 transposable elements driving genome dynamics. This study challenges the traditional  
31 perspective that highly abundant amphipods within a trench form a single panmictic  
32 population. We discuss findings in relation to eco-evolutionary and ontogenetic processes  
33 operating in the deep sea, and highlight key challenges associated with population genetic  
34 analysis in non-model systems with inherent large effective population sizes and genomes.

35

36 Keywords: Deep-sea, amphipod, hadal, Mariana, RADSeq, genetic structure

## 37 Introduction

38 The deep-ocean hadal zone remains one of the most poorly understood and least explored  
39 ecosystems on Earth. It extends below 6000 m to full ocean depth at ca. 11,000 m in the  
40 Mariana trench, and whilst it accounts for the bottom 45% of the oceans bathymetric range it  
41 represents only 2% of the total seafloor area (Jamieson, 2015). Most of the hadal zone is made  
42 up of the deep-ocean trenches that form at subduction zones between converging tectonic  
43 plates, and most of these trenches are situated around the Pacific rim where they form a  
44 discontinuous chain of discrete topographic seafloor features separated by intervening abyssal  
45 plains (Stewart & Jamieson, 2018).

46

47 An enduring question has been to what extent the individual trenches are demographically and  
48 evolutionarily independent, each shaping the patterns of local intra- and inter-specific genetic  
49 structure through the combined effects of extreme environmental conditions and their relative  
50 geographic isolation (Ritchie et al., 2017b; Weston & Jamieson, 2022). The traditional  
51 narrative has been that the trenches promote a high level of species endemism for a specialised  
52 hadal fauna that is distinct from shallower deep-sea fauna on the abyssal plains by virtue of  
53 adaptation to extreme hydrostatic pressures and reduced food availability (Beliaev, 1989;  
54 Wolff, 1959, 1970). Many taxa show a pattern of phylogeographic structure that is consistent  
55 with this view, such as the Lysianassoid amphipod genus *Hirondellea* where several species  
56 can have restricted geographic ranges associated with individual trenches (France, 1993;  
57 Lowry & Stoddart, 2010) while others do not (Weston & Jamieson, 2022). Even in those taxa  
58 with a cosmopolitan distribution across ocean basins, such as the amphipods of the *Eurythenes*  
59 *gryllus* complex, there are monophyletic morphospecies found within individual trenches that  
60 indicate localised evolutionary divergence and potentially incipient speciation (Eustace et al.,  
61 2016; Havermans, 2016; Havermans & Smetacek, 2018; Havermans et al., 2013; Ritchie et al.,

62 2015; Weston et al., 2021). A similar picture is seen in the genus *Paralicella*, where high levels  
63 of genetic divergence exist between neighbouring Kermadec and New Hebrides trenches based  
64 upon microsatellite DNA polymorphisms. However an apparent lack of structure for a different  
65 *Paralicella* species across the Pacific Ocean questions a total lack of gene flow among  
66 individual trenches (Ritchie et al., 2017b).

67

68 Most research on how trenches influence the spatial distribution of genetic diversity has  
69 focussed on a broad geographic scale between trenches (Weston & Jamieson, 2022). However,  
70 there is also evidence for localised population and ontogenetic structure within trenches  
71 (Blankenship et al., 2006; Eustace et al., 2016; Lacey et al., 2018; Lacey et al., 2016; Weston  
72 et al., 2021). Multiple different deep-sea amphipod species across different trenches show the  
73 same patterns of ontogenetic stratification, with juveniles found in in the shallower depths  
74 within their individual species bathymetric distribution and only adults present in the deeper  
75 regions. It has been considered that at shallower depths the juveniles will experience reduced  
76 predation and intra-specific competition, gain a nutritional benefit, and the reduced hydrostatic  
77 pressures will confer a physiological and metabolic advantage during growth (Blankenship et  
78 al., 2006; Hessler et al., 1978; Lacey et al., 2018). How such ontogenetic structure affects  
79 population genetic structure through the effects of differential selection and cohort formation  
80 across depth distributions remains unknown.

81

82 A major impediment to collating the body of data and evidence necessary to generalise about  
83 how the hadal trenches shape patterns of intra-specific genetic structure is the sheer logistical  
84 challenge of sampling the deep sea at an appropriate geographic scale and for an adequate  
85 number of individuals (Jamieson et al., 2010). Indeed, many of the high profile studies on deep-  
86 sea genetic structure work within a phylogeographic rather than a population genetic

87 framework (Baco et al., 2016; Havermans et al., 2013; Taylor & Roterman, 2017); focus on  
88 topographic features such as hydrothermal vents with a proclivity for marked genetic structure  
89 driven by natural selection (Vrijenhoek, 2010; Xiao et al., 2020); or confirm hypotheses of  
90 wide ranging dispersal in cosmopolitan species from relatively sparse sampling (France &  
91 Kocher, 1996). Advances in autonomous underwater vehicle technology have increased the  
92 capability for sampling the deep sea in a more systematic and less opportunistic manner. This  
93 can provide much better insight of how stochastic and deterministic microevolutionary forces  
94 shape the spatial distribution of genetic diversity over different geographic scales (Jamieson,  
95 2015).

96

97 Here we exploit a sampling campaign to the Mariana Trench that yielded multiple samples of  
98 the amphipod *Hirondellea gigas* to provide the first reports of within-species population  
99 genetic structure over a small spatial geographic scale within a single hadal trench. *H. gigas*  
100 (Birstein & Vinogradov, 1955) is the dominant scavenging amphipod species found deeper  
101 than 6000 m in the hadal trenches of the NW Pacific Ocean (Eustace et al., 2013; France, 1993).  
102 It is readily caught in large numbers using baited traps deployed on deep-sea autonomous  
103 lander vehicles, which simulates the natural occurrence of surface-derived carrion falls  
104 (Blankenship & Levin, 2007). Multiple individuals were sampled along a transect that  
105 encompasses both the overriding and subducting tectonic plates and the trench axis of the  
106 Mariana Trench. We utilise double digest restriction site-associated DNA sequencing (ddRAD)  
107 (Peterson et al., 2012) to genotype individuals at multiple thousand loci across the *Hirondellea*  
108 genome. The propensity for sampling *H. gigas* from within the NE Pacific Ocean hadal zones  
109 makes it an obvious target species to examine within-species population genetic structure in  
110 trenches. However, this also presents two major challenges.

111

112 The first challenge is that the ease of attracting *H. gigas* to baited traps suggests an extremely  
113 large population size. Indeed, the number of individuals attracted to bait are typically hundreds  
114 to thousands within hours (Gallo et al., 2015; Hessler et al., 1978). Such hyperabundance can  
115 influence the detection and interpretation of population genetic divergence (Waples &  
116 Gaggiotti, 2006). Population structure in neutral markers is primarily driven by the effects of  
117 genetic drift, with effects inversely proportional to population size (Hedrick et al., 1976;  
118 Waples, 1998). As such there will be very slow accumulation of genetic divergence between  
119 large subpopulations, even in the absence of gene flow. Such so-called “panmictic inertia”  
120 causes a disconnect between genetic differentiation as estimated from neutral polymorphisms  
121 and the actual demographic connectivity between locations. The use of genome-wide markers  
122 generated using RADSeq in analysis means that focus can shift away from just examining  
123 purely neutral markers influenced by demographic processes, and instead allow examination  
124 of patterns of population structure at outlier loci that are highly differentiated relative to neutral  
125 expectations reflecting the influence of selection. The spatial distribution of such adaptive  
126 genetic variation can be used to infer local adaptation and be particularly useful for identifying  
127 the footprint of genetic structure in marine systems where high levels of gene flow and large  
128 population sizes are common (Deagle et al., 2015).

129

130 The second challenge is that deep-sea amphipods have extremely large genome sizes (Ritchie  
131 et al., 2017a). These can range up to ca. 34 Gbp in *Alicellea gigantea*, with *Hirondellea dubia*  
132 having a ca. 5 Gbp genome with a large component of repetitive and paralogous DNA (Ritchie  
133 et al., 2017a). This makes it especially difficult to ensure that the relatively short DNA  
134 sequences generated using approaches such as RADseq are assembled into true orthologous  
135 loci from which single nucleotide polymorphism (SNP) genotypes and allele frequencies can  
136 be drawn to infer patterns of population genetic structure. The potential for confounding

137 paralogous gene sequence variants has been highlighted in species that have undergone  
138 historical genome duplication events, such as the salmonids (Davidson et al., 2010; Waples et  
139 al., 2016) and also in species with large genomes (Deagle et al., 2015) especially where the  
140 focal organism lacks a reference genome and so RADseq relies on *de novo* locus assembly  
141 (Nadukkalam Ravindran et al., 2018). In the case of deep-sea amphipods, there is no evidence  
142 of ancestral genome duplication, but some indications that large genomes are a consequence of  
143 transposable element proliferation which will duplicate genomic regions as a by-product of  
144 transposition (Brown & Thatje, 2014; Ritchie et al., 2017a). Both genome duplication and the  
145 actions of repetitive elements increase the potential that sequence contigs from polymorphic  
146 multicopy genomic regions are erroneously collapsed into a single locus, artificially inflating  
147 heterozygosity and biasing allele frequency. Several bioinformatic pipelines are available for  
148 RADseq analysis to enrich for true analogues within and between samples by identifying  
149 erroneous paralogues based upon optimising sequence similarity metrics, thus preventing over-  
150 splitting of true alleles into separate loci or conversely under-splitting and incorrectly merging  
151 paralogous sequences into a single locus (Harvey et al., 2015; McKinney et al., 2017;  
152 Nadukkalam Ravindran et al., 2018; O’Leary et al., 2018; Willis et al., 2017). Here we  
153 combine several of these approaches at high stringency to identify true independent loci from  
154 which proper biological inference can be made. This study therefore presents both new insight  
155 into the patterns of population genetic structure within hadal amphipods from the Mariana  
156 Trench, but also provides a cautionary tale for population genomics studies on non-model  
157 organisms with inherently large genomes.

158

## 159 [Materials & Methods](#)

160

161 **Sample collection**

162 Amphipods were collected using the full ocean depth rated autonomous baited camera and trap  
163 lander; Hadal-Lander C (Linley et al., 2016) during a research cruise on the RV Falkor  
164 (Schmidt Ocean Institute cruise FK141109). The lander was equipped with a conductivity,  
165 temperature and depth (CTD) sensor (SBE-19 plus V2; SeaBird Electronics Inc, USA) to  
166 record depth every 10 s throughout deployment, and an array of three 30 cm x 6 cm funnel  
167 traps baited with 200 g of locally sourced fish. Samples were taken from five sites across the  
168 Mariana Trench from north to south, encompassing both the overriding and subducting tectonic  
169 plates and the trench axis (Figure 1).

170

171 Mature male samples were identified according to Barnard & Ingram (1990) and Barnard &  
172 Karaman, (1991). DNA was extracted according to Ritchie et al., (2017b). All samples were  
173 confirmed as *H. gigas* from mitochondrial COI DNA barcodes obtained using LCO1490 and  
174 HCO12198 PCR primers (Folmer et al., 1994) according to Ritchie et al., (2015).

175

176 **ddRAD sequencing, locus assembly and paralog filtering**

177 Individuals were genotyped using double-digest RADSeq (Peterson et al., 2012) with the two  
178 6-bp restriction enzymes *Pst*I and *Eco*RI. Samples were multiplexed across four libraries and  
179 sequenced across two lanes on an Illumina HiSeq 2000/2500 instrument generating 125 bp  
180 paired-end reads. Read pairs with evidence of at least 4 bp of 3' read-through into the *Eco*RI  
181 restriction site plus Illumina adapter (forward reads) or the *Pst*I restriction site plus 8bp barcode  
182 and Illumina adapter (reverse reads) were removed using CUTADAPT 2.3 (Martin, 2011). The  
183 remaining reads were demultiplexed and quality-filtered (-c -r -q options) using the  
184 *process\_radtags* utility of STACKS 2.58 (Rochette et al., 2019). Barcodes were rescued



185 allowing one nucleotide mismatch and read pairs containing residual Illumina adapters with at  
186 most two mismatches were discarded.

187

188 The sequence data were assembled using the *de-novo* STACKS pipeline with additional  
189 filtering at various points to enrich for unique non-paralogous loci (hereafter, singletons)  
190 (Supplemental Figure S1) (Nadukkalam Ravindran et al., 2018). Initial loci were assembled  
191 from the forward reads within each individual in *ustacks*, implementing a strict haplotyping  
192 approach to identifying paralogs, where no more than two haplotypes may be observed in an  
193 individual and allele merging must be optimised to maximise biallelic loci and minimise  
194 monoallelic loci per individual (Ilut et al., 2014; Willis et al., 2017). Alleles were built from  
195 primary stacks of at least three identical reads ( $m=3$ ) and merged into loci via ungapped  
196 alignments allowing up to seven mismatches ( $M=7$ , equivalent to 94% sequence identity) and  
197 at most two primary stacks per locus. The deleveraging algorithm was disabled to ensure that  
198 overmerged multi-allelic loci originating from diverged paralogous genomic regions were  
199 blacklisted. We selected  $M=7$  based on a parameter sweep from  $M=0$  to  $M=15$ , where  $M=7$   
200 minimised the average number of monoallelic loci per individual (though biallelic loci had  
201 already peaked at  $M=4$ ; Supplemental Document S1), yielding sets of high-quality mono- and  
202 biallelic loci that are relatively distinct in sequence space within each individual. Initial SNP  
203 and haplotype calls were made by aligning secondary stacks to loci with at most nine  
204 mismatches ( $N=9$ ), calling SNPs using the bounded model with an upper error bound of 0.05,  
205 and retaining haplotypes from primary stacks only.

206

207 Loci from all individuals were then merged into a catalogue using *cstacks* in ungapped  
208 alignment mode. To prevent paralogs that were diverged between individuals from being  
209 merged into the same catalogue locus we required a 100% match ( $n=0$ ) between loci from

210 different individuals. This strict approach ensures that paralog variants are recorded as separate  
211 loci but does not preclude *bona fide* heterozygous catalogue loci to be formed from individuals  
212 with biallelic loci. Correspondence between catalogue loci and sample alleles was obtained  
213 using *sstacks* in ungapped alignment mode, and catalogue loci that linked to alleles from  
214 multiple loci within an individual were blacklisted since these originate from within-individual  
215 paralogs that were missed by the initial locus assembly. The catalogue locus sequences were  
216 then clustered by sequence similarity to identify diverged between-individual paralogs  
217 (Nadukkalam Ravindran et al., 2018). Sequence clustering was carried out at 40% identity in  
218 ungapped end-to-end alignments using VSEARCH 2.15.1 (Rognes et al., 2016), and all loci  
219 that clustered with other loci were blacklisted as putative paralogs. A parameter sweep  
220 indicated that the numbers of clusters reached a minimum at 40% (Supplemental Document  
221 S1), suggesting that this degree of identity produces the most conservative set of singleton loci  
222 in this highly diverse dataset.

223

224 Genotyping was completed by incorporating the reverse reads using *tsv2bam* and calling final  
225 genotypes from information across all individuals using *gstacks* with the Maruki-low model  
226 ( $\alpha = 0.05$ ) (Maruki & Lynch, 2017). The final assembled paired-end catalogue loci  
227 consensus sequences were examined for microbial and human contamination by querying  
228 against all bacterial, viral, fungal and protozoan genomes available in the NCBI REFSEQ  
229 database using BLASTN 2.9.0 (Camacho et al., 2009) with an e-value cut-off of 0.001, and  
230 classifying against the PlusPF-16GB database using KRAKEN2 2.0.8 (Wood et al., 2019).  
231 Loci identified with at least one of these methods were blacklisted.

232

233 A final paralogue check was done based on heterozygosity and allele read ratios in the final  
234 genotypes of the paired-end catalogue contigs (McKinney et al., 2017). SNPs with minimum

235 allele frequency of 0.01, maximum heterozygosity of 0.6 and found in loci common among all  
236 five populations were extracted using *STACKS populations*, and paralogous loci were  
237 identified using HDplot (<https://github.com/gjmckinney/HDplot>), enforcing a maximum  
238 deviation from even read ratio (z-score) of +/- 2 to classify a locus as a singleton. The final  
239 blacklist of paralogs consisted of the intersection set between loci blacklisted by the  
240 VSEARCH and HDplot methods (between-individuals methods), plus those blacklisted by  
241 haplotyping (within-individuals method). Final genotypes at singleton loci that were common  
242 among all five populations and identified in at least 50% of individuals per population were  
243 extracted using *populations*, requiring a minimum minor-allele frequency of 0.01 and a  
244 maximum observed heterozygosity of 0.6.

245

#### 246 Inference of genetic structure

247 Genetic structure among individuals and sampling sites was examined using principal  
248 components analysis (PCA) and discriminant analysis of principal components (DAPC) in  
249 *adegenet* 2.1.3 (Jombart, 2008). Missing data were replaced with the dataset-wide mean allele  
250 frequency at the corresponding SNP. PCA eigenvalues were used to determine the number of  
251 principal components retained for DAPC such that 80% of the variance was preserved. Genetic  
252 structure among sampling sites was further examined using pairwise  $F_{ST}$  (Nei & Chesser, 1983)  
253 and population-specific  $F_{ST}$  (local  $F_{ST}$ ) (Weir & Goudet, 2017), both calculated using  
254 *FinePop2* (Kitada et al., 2017, 2021). Relationships between individual-based scores on each  
255 of the first three linear discriminant functions (DAPC) and latitude or sampling depth were  
256 identified using Spearman's rank correlation. The pairwise  $F_{ST}$  matrix was summarised using  
257 PCA and relationships between the first two principal components and latitude or depth were  
258 identified likewise. Isolation by latitude or depth based on the pairwise  $F_{ST}$  matrix was tested  
259 using Mantel tests with 9999 permutations (*vegan* package) (Oksanen et al., 2013). Population-

260 specific  $F_{ST}$  was regressed onto latitude or depth variables using generalised least-squares  
261 models, accounting for autocorrelation in the response (Kitada et al., 2021).

262

263 SNPs driving the observed divergence between sites were identified by hierarchically  
264 clustering DAPC variable contributions to the linear discriminant functions into “outlier” and  
265 “non-outlier” SNPs using Ward’s method (*snpzip* function in *adegenet*).  $F_{ST}$ -based outliers  
266 were identified using OutFLANK (Lotterhos & Whitlock, 2015) and outliers associated with  
267 individual-based structure were identified using multivariate Mahalanobis distances based on  
268 the first six principal components using PCADAPT (Luu et al., 2017). The union of the three  
269 sets was retained as a final set of outliers.

270

271 Finally, genetic structure among individuals based on outliers and non-outliers was inferred  
272 using two Bayesian methods. SNP-based inference was carried out by thinning SNPs by  
273 linkage disequilibrium using PLINK 1.9 (Chang et al., 2015) with an  $r^2$  threshold of 0.5 and  
274 then using fastSTRUCTURE with the logistic prior model and inferring between two ( $K=2$ )  
275 and five ( $K=5$ ) genetic clusters (Raj et al., 2014). Haplotype-based inference was carried out  
276 using RADpainter and fineRADstructure (Malinsky et al., 2018). The co-ancestry matrix  
277 inferred by RADpainter was inspected using PCA and used to assign individuals to an optimal  
278 number of genetic clusters using fineRADstructure with  $10^6$  iterations burn-in,  $10^6$  iterations  
279 MCMC and  $10^4$  thinning interval. The clustering dendrogram was obtained from the best  
280 posterior state identified from at most  $10^5$  search attempts.

281

## 282 **Functional annotation of loci**

283 The assembled ddRAD loci consensus sequences were functionally annotated to identify  
284 functional enrichment among subsets of loci. Loci containing repetitive elements were

285 identified using REPEATMASKER 4.0.9 with the arthropoda database. Loci overlapping  
286 coding sequences were identified via alignment against the *Hirondellea gigas* transcriptome  
287 assembly (NCBI TSA accession GEZX01) using BLASTN 2.9.0 (Camacho et al., 2009) with  
288 an e-value cut-off of 0.001. Protein and GeneOntology (GO) annotations (biological process  
289 ontology) for each locus were obtained via alignment against all arthropod proteins available  
290 on UniProt (taxonomy ID 6656) using BLASTX 2.9.0 with an e-value cut-off of 0.001.  
291 Enrichment of GO terms among outlier loci against a background of non-outlier loci were  
292 carried out using hypergeometric tests in CLUSTERPROFILER 3.14.2 (Yu et al., 2012) with  
293 an FDR-corrected significance threshold of 0.05.

294

## 295 Results

### 296 ddRAD data assembly

297 An average of 3,306,212 read pairs per sample passed quality filtering and were assembled to  
298 19,905-34,486 loci per individual with a mean coverage of 12-35x per locus (Supplemental  
299 document S1). Sequencing depth and numbers of loci were similar among sampling sites  
300 (Kruskal-Wallis test:  $P > 0.05$ ; Supplemental document S1), though one library produced  
301 significantly fewer loci than other libraries ( $P < 0.05$ ; Supplemental document S1). The locus  
302 catalogue comprised 457,130 loci, of which 11,407 (2%) were removed as microbial/human  
303 contaminants and 3,182 non-paralogous loci containing 43,408 SNPs were present in all five  
304 sampling sites and at least 50% of individuals at each site.

305

306 Inspection of locus-sharing patterns among individuals revealed systematic data missingness  
307 associated with ddRAD library, which was partially confounded by sampling site. PCA on  
308 Jaccard distances and DAPC between libraries separated individuals from sites C or D in one  
309 particular library from individuals from the same site in other libraries (Supplemental Figure

310 2). This same separation was also apparent in SNP genotypes, even in loci shared across all  
311 individuals. The library in question had fewer loci overall (Supplemental document S1) and  
312 more overlap in read pairs (45% versus 13-25% pairs merged using FLASH 1.2.11, (Magoč &  
313 Salzberg, 2011), suggesting a difference in size-selection regime. To account for this batch  
314 effect, loci or SNPs that were among the top 10% features driving the DAPC separation in  
315 locus sharing or SNP genotypes between individuals within sites C or D were removed (Deagle  
316 et al., 2015; O’Leary et al., 2018). This removed 70% of SNPs with heterozygote deficiency in  
317 sites C and D (Supplemental Figure 3) and left 824 loci containing 5,147 SNPs with minimal  
318 residual discrimination between libraries and good discrimination between sites C and D  
319 (Supplemental Figure 2). Notwithstanding, all further analyses were also confirmed with a  
320 reduced dataset, comprising only individuals from sites C, D and E from the outlier library.

321

## 322 Genetic structure

323 PCA on SNP genotypes indicated substantial genetic differentiation among individuals but not  
324 sampling sites, whereby the two axes accounted for 4.98% and 3.92% of the variance (Figure  
325 2a). In contrast, DAPC on individuals grouped by sampling site revealed three major genetic  
326 groups on the first two discriminant functions, comprising sites A+B, C+D and E (Figure 2b).  
327 The third discriminant function separated sites A from B and C from D (Figure 2b). The genetic  
328 structure described by the first two discriminant functions was significantly associated with  
329 latitude (axis 1;  $\rho = - 0.870$ ;  $P < 0.001$ ) and trench depth (axis 2;  $\rho = - 0.757$ ;  $P < 0.001$ )  
330 respectively (Figure 3). These associations were stable across a broad range of PCA variance  
331 preserved in the DAPC (Supplemental Table 1). Based on DAPC variable loadings, this genetic  
332 structure was driven by 301 outlier SNPs. No significant  $F_{ST}$ -outliers were identified, and PCA-  
333 based Mahalanobis distances yielded 23 outliers, of which 6 (26%) were not identified by  
334 DAPC, yielding a combined set of 307 outlier SNPs and 4,840 non-outlier SNPs.

335

336 The pairwise  $F_{ST}$  matrix based on the outlier SNPs supported the same three genetic clusters  
337 as identified via DAPC (Figure 4a) and the first two principal components tended to be  
338 associated with latitude and depth respectively ( $\rho = -0.9$ ;  $P = 0.08$ ). Mantel tests supported a  
339 weak relationship with latitude ( $r = 0.66$ ;  $P = 0.06$ ) but not with depth ( $r = -0.07$ ;  $P = 0.55$ ).  
340 Pairwise  $F_{ST}$  based on non-outlier SNPs described a much weaker genetic structure, primarily  
341 driven by divergence of site B, but showing no associations with latitude or depth (Figure 4a).  
342 Population-specific (local)  $F_{ST}$  corroborated the contrast in population structure between  
343 outlier and non-outlier loci (Figure 4b), whereby it was weakly associated with depth in outliers  
344 ( $\beta = 3.87e-05$ ;  $SE = 2.13e-05$ ;  $Z = 1.815$ ;  $P = 0.070$ ) and strongly associated with latitude in  
345 non-outliers ( $\beta = -0.418$ ;  $SE = 0.130$ ;  $Z = -3.216$ ;  $P = 0.001$ ).

346

347 Individual-based Bayesian inference of genetic structure failed to resolve any structure using  
348 non-outlier SNPs and yielded genetic clusters inconsistent with geographic or bathymetric  
349 structure using outlier SNPs (Supplemental Figure 4). Similarly, the haplotype-based  
350 coancestry matrix resolved minimal structure using non-outlier loci and failed to identify  
351 structure congruent with sampling site, geography or bathymetry using outlier loci  
352 (Supplemental Figure 5). Analysis of a reduced dataset comprising only samples from a single  
353 library covering sites C, D and E corroborated genetic structure between these three sites  
354 (Supplemental Figure 6), and local  $F_{ST}$  corroborated genetic similarity between the deep sites  
355 C and D compared to the shallow site E (Supplemental Figure 7).

356

### 357 [Functional annotation of loci](#)

358 Out of the total raw 457,130 loci, 33,174 (7%) contained repetitive sequence elements, 218,874  
359 (48%) matched a *Hirondellea gigas* transcript and 65,921 (14%) matched an arthropod UniProt

360 protein. The various filters employed to enrich for singleton loci identified 25,402 loci as  
361 paralogs common between at least 50% of individuals from each sampling site - these  
362 contained significantly more repetitive elements (5.9% vs. 3.6%;  $\chi^2_1 = 6.821$ ;  $P = 0.009$ ),  
363 significantly fewer transcripts (53.0% vs. 57.7%;  $\chi^2_1 = 6.765$ ;  $P = 0.009$ ) and similar numbers  
364 of proteins (17.0% vs. 18.8%;  $\chi^2_1 = 1.650$ ;  $P = 0.199$ ) compared to the singleton loci. Among  
365 these singletons, proportions of annotated loci did not differ significantly between 170 outlier  
366 and 654 non-outlier loci, ranging from 2.4%-4% for repetitive elements, 52.3%-59.2% for  
367 transcripts and 18.2%-21.2% for proteins (Figure 5).

368

369 A total of 66, 116 and 345 GeneOntology (GO) biological process annotations were assigned  
370 to the outlier, non-outlier and paralog loci, respectively. None of these GO terms were  
371 significantly enriched ( $FDR > 0.2$ ) against background locus sets comprising all 3182 singleton  
372 loci or (for paralogs) the whole raw locus catalogue. However, the set of GO terms assigned to  
373 paralog loci was markedly differed from those in singletons with only 54 out of 345 terms  
374 (15.7%) being shared. The two most frequently observed GO terms among outliers and non-  
375 outliers are related to DNA transposon activity (*DNA integration* [GO:0015074], *transposition*,  
376 *DNA-mediated* [GO:0006313]), followed by terms related to epigenetics and protein  
377 ubiquitination (Figure 5, Supplemental Document S2). The transposon terms represent several  
378 reverse transcriptase/integrase enzymes and retrotransposon polyproteins (Tf2, TC1, Gag-pol,  
379 TY3B-I/G). The most frequently observed GO term among paralogs was *viral genome*  
380 *replication* [GO:0019079], followed by various fundamental cellular processes (Supplemental  
381 Document S2). The GO terms unique to outlier loci comprised processes involved in protein  
382 modification, photoperception (*sina* and *SLAH1* genes) and neural function among various  
383 fundamental RNA processing terms (Supplemental Document S2).

384



## 385 Discussion

386 This study has resolved fine-scale population genetic structure in *Hirondellea gigas* deep-sea  
387 amphipods in the Mariana Trench, highlighting a signature of isolation by geography and  
388 bathymetry across a small set of strongly diverged loci against a highly diverse genomic  
389 background impacted by large population size, large genome size and extensive locus paralogy  
390 caused by transposable elements. Our analyses thus provide new insight into eco-evolutionary  
391 processes operating in the deep ocean hadal zone, emphasising how both stochastic and  
392 deterministic microevolutionary forces are shaping the spatial distribution of genetic variation  
393 and so could contribute across range of spatial scales to population and phylogeographic  
394 structure and potentially the speciation process.

395 The large genome sizes characteristic of many deep-sea amphipods, which range from c. 4-34  
396 Gbp (Ritchie et al., 2017a), cause considerable technical challenges when attempting to resolve  
397 population genetic structure using approaches such as *de novo* RADseq. Proliferation of  
398 transposable elements increases levels of paralogy across the genome and thus increases the  
399 risk that sequence reads from paralogous multicopy regions are incorrectly merged into a single  
400 locus, leading to a bias in estimates of allele frequency and any downstream population genetic  
401 analysis inference. This problem is not unique to deep-sea amphipods but applies to all  
402 organisms with large and dynamic genomes, particularly those having undergone genome  
403 duplication or subject to extensive TE activity (Davidson et al., 2010; Deagle et al., 2015;  
404 Waples et al., 2016; ).

405 We made several modifications to the standard ddRAD STACKS pipeline to ensure that the  
406 final dataset yielded robust inference of genetic structure. First, we enriched for non-paralogous  
407 loci within individuals during *ustacks* assembly by increasing the number of allowed  
408 mismatches between alleles, disabling the deleveraging algorithm and allowing at most two

409 alleles per locus. These settings are different to standard recommendations (Paris et al., 2017),  
410 but are effective in blacklisting biologically implausible loci comprising multiple alleles  
411 originating from diverged paralogs. We empirically identified the ideal threshold in our dataset  
412 to strike a balance between minimising the numbers of monoallelic loci and maximising the  
413 numbers of biallelic loci (Ilut et al., 2014; Willis et al., 2017). Second, we observed high levels  
414 of between-individual loci similarity (Supplemental Document S1), leading to a decrease in  
415 catalogue loci by >40% when allowing seven mismatches in *cstacks* as suggested by default  
416 recommendations (Paris et al., 2017). An effective way of avoiding overmerging of paralogs  
417 diverged between individuals into confounded catalogue loci is to undermerge loci in *cstacks*  
418 and blacklist catalogue loci that cluster at relatively low sequence similarity (Nadukkalam  
419 Ravindran et al., 2018). In our dataset, even a similarity threshold of 90% reduced the number  
420 of unique loci across all individuals by more than 50%, and the number of locus clusters did  
421 not stabilise until 40% similarity, consistent with extensive locus paralogy (Supplemental  
422 Document S1). Finally, we removed loci with uneven per-allele read depths, which may  
423 originate from identical paralogous genomic regions that our previous measures failed to  
424 identify (McKinney et al., 2017). Our strict modified RAD-Seq pipeline may be overly  
425 conservative, possibly falsely blacklisting diverged singleton loci (Nadukkalam Ravindran et  
426 al., 2018), but is clearly essential when dealing with data with extensive paralogy.

427 Although our focus was not on identifying and characterising TE content, for which more  
428 specialised pipelines exist (Chak & Rubenstein, 2019), comparing the functional annotations  
429 of the singleton versus paralogous locus sets allowed us to gain some insight into the role of  
430 TEs in affecting eco-evolutionary processes operating in *H. gigas* in the deep sea. The  
431 paralogous loci contained significantly more repetitive elements and returned fewer hits with  
432 the *H. gigas* transcriptome compared to the singleton loci. The singleton and paralogous loci  
433 also had near non-overlapping gene ontology assignments, with the singleton loci being

434 associated with GO terms that included regulation of integration, transposition and methylation  
435 that would also be involved in policing the activity of transposable elements. Combined, these  
436 observations reinforce the significance of transposable elements in driving genome size in  
437 deep-sea amphipods (Naville et al., 2019; Ritchie et al., 2017a), and are consistent with  
438 hypotheses implicating TE as drivers of evolutionary innovation necessary to allow the  
439 colonisation of the deep ocean from shallower bathyal depths (Brown & Thatje, 2014). There  
440 is growing recognition and focus around how TE proliferation can generate genomic novelty  
441 through a variety of mechanisms that can lead to rapid adaptation across a broad taxonomic  
442 range and under varying ecological contexts (Bourque et al., 2018; Ricci et al., 2018; Wells &  
443 Feschotte, 2020). For deep-sea amphipods what is now required is the mapping of different TE  
444 families across host genomes to examine location relative to genes that could influence the  
445 adaptive potential to extreme hydrostatic pressure, especially for those situations where there  
446 is clear gene family expansion or signatures of positive selection (Kobayashi et al., 2018, 2019;  
447 Lan et al., 2017).

448 A further inherent challenge for identifying patterns of population genetic structure in *H. gigas*  
449 is that the high local abundance of these amphipods could show, at most, only subtle levels of  
450 genetic divergence (Hedrick et al., 1976). Given that the effect of genetic drift on allele  
451 frequencies is inversely proportional to population size, even in the complete absence of gene  
452 flow among locations there will only be very slow accumulation of genetic differences at  
453 neutral loci that makes inferring dispersal and population connectivity problematic (Waples &  
454 Gaggiotti, 2006). This genomic inertia can cause a disconnect between patterns of demographic  
455 and population genetic structure (Lowe & Allendorf, 2010) and can be common occurrence in  
456 marine systems where species can have large effective population sizes and fewer physical  
457 barriers to gene flow (Baco et al., 2016; Deagle et al., 2015; Etter et al., 1999). Indeed, our  
458 initial data exploration using PCA suggested no genetic structure between sampling locations

459 within the Mariana Trench, consistent with a signature of panmixia. However, a parallel DAPC  
460 analysis, that acts to maximise variance between sampling locations whilst minimising  
461 variance within them, identified clear genetic structure across all sampling sites. This  
462 differentiation was driven by a relatively small number of SNPs and loci, which display higher  
463 levels of divergence among populations than the neutral genomic average and thus can be  
464 considered to reflect the effects of adaptive divergence. Markers under divergent selection will  
465 change in frequency more readily than neutral polymorphisms even in large populations,  
466 though the capacity to identify genomic regions under selection is difficult with low levels of  
467 linkage disequilibrium (Pritchard & Przeworski, 2001).

468 An intriguing finding was that the multivariate structure described by the first two discriminant  
469 functions was significantly correlated with both latitude and trench depth. Relationships with  
470 latitude may reflect a signature of isolation by distance, though given the structure was driven  
471 primarily by a limited set of outlier loci this would also suggest some gradual signature of  
472 divergence based upon some environmental gradient. These outlier loci represented GO terms  
473 associated with DNA integration, transposition and methylation, in line with the broader suite  
474 of singleton loci, again reflecting the potential important of transposable element activity and  
475 policing in amphipod adaptation and evolution (Brown & Thatje, 2014; Ritchie et al., 2017a).  
476 Critically, the association with trench depth challenges any oversimplistic perspective about  
477 patterns of dispersal, gene flow and genetic structure in the deep sea and the hadal zone. The  
478 traditional paradigm was centred around the depth-differentiation hypothesis (Rex & Etter,  
479 2010) which postulated that there will be decreases in both environmental heterogeneity and  
480 barriers to gene flow with increasing depth and hence less structure. Whilst data from some  
481 deep-sea species do support the depth differentiation hypothesis (Coward et al., 2014; Quattrini  
482 et al., 2015; Ritchie et al., 2013) it is apparent that as more studies examine a broader taxonomic  
483 spread and range of habitats, the more exceptions there are to the general rule. Disjunct

484 topographical features such as seamounts, fracture zones and ocean ridges do disrupt gene flow,  
485 and hydrothermal vents offer localised selection pressures that drive adaptive divergence (Baco  
486 et al., 2016). The hadal trenches, as a disconnected network of ultra-deep island-like habitats,  
487 are also recognised as drivers of genetic structure in the deep sea. This is evidenced both by  
488 the levels of endemism found in the hadal zone, and also from some population genetic studies  
489 that have compared across trenches in cosmopolitan amphipod species (Chan et al., 2020;  
490 Havermans et al., 2013; Ritchie et al., 2017b; Weston et al., 2022). Patterns of genetic structure  
491 have been framed in the context of understanding how the intervening abyssal plains operate  
492 as barriers to gene flow and how differences in selection between abyssal and hadal zones may  
493 drive divergence. The current study provides whole new insight into the microgeographic  
494 scales over which population genetic structure can accrue, rejecting any concept that highly  
495 abundant amphipods within a trench form a single panmictic population. Moreover, it  
496 highlights that environmental differences operating within a trench and over microgeographic  
497 scales do influence the spatial distribution of genetic variation.

498

499 Within-trench population structure has been hinted at before from patterns of ontogenetic  
500 variation in amphipods, with a consistent trend across species and trenches of juveniles residing  
501 in the shallower depths of the bathymetric range (Blankenship et al., 2006; Eustace et al., 2016;  
502 Lacey et al., 2018; Weston et al., 2021). Spatially this would mean that juveniles at shallower  
503 sites move in a downslope direction as they grow and mature, which could explain the patterns  
504 of isolation-by-depth observed in the current data given the environmental gradients associated  
505 with trench depth and the effects of genetic drift operating in a similar way to a founder effect  
506 as juveniles moved to deeper reaches of the trench. This would be especially pronounced if  
507 juveniles formed periodic cohorts. Chronobiological triggers that could entrain reproduction  
508 do occur even at full ocean depth (Taira et al., 2004), and temporal and seasonal shifts in

509 surfaced derived food supply with associated responses of deep-sea benthic fauna are well-  
510 documented (Durden et al., 2020; Gooday & Lamshead, 1989; Kalogeropoulou et al., 2010).  
511 It is also known that *Hirondellea* amphipods do not move long distances through the water  
512 column, with video footage from the autonomous landers showing they so not move more than  
513 a metre or so from the surface of the trench, and amphipod traps that are suspended above the  
514 trench floor rarely catching any individuals (Jamieson et al., 2009). This, coupled with a life  
515 cycle in *Hirondellea* that involves fertilisation and juvenile development in a marsupium,  
516 would predict that longer distance dispersal is limited. Moreover, the trench topology could  
517 provide multiple localised barriers to dispersal, certainly from the trench axis onto the  
518 shallower aspects of the plates, and in the case of the Mariana Trench, there are five physically  
519 partitioned areas at hadal depths within this one trench (Jamieson & Stewart, 2021).

520

521 Overall, this study exemplifies why deep-sea population genetics needs to embrace a seascape  
522 genomics approach (Liggins et al., 2020; Nielsen et al., 2020; Riginos et al., 2016) to  
523 understand better how the deep-sea environment shapes genomic diversity both in terms of  
524 how geographical features influence the connectivity of populations and how environmental  
525 gradients affect demographic structure and selective regimes. The drivers of genetic structure  
526 encompass both stochastic and deterministic microevolutionary forces spanning a range of  
527 spatial and temporal scales often underpinned by subtle if not enigmatic environmental effects.  
528 This requires both the capacity to examine genome-wide structure in non-model systems with  
529 potentially low levels of underlying genomic resources, plus also sampling at an appropriate  
530 scale that can encompass both micro- and macrogeographic patterns. The latter is being  
531 achieved through advances in deep-ocean lander and submersible technology, and we have  
532 illustrated how the former is achievable with scrutiny of the data generated and recourse to  
533 potential sources of bias and error such as large genome size and effective population size.

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541

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#### 839 [Data Accessibility Statement](#)

840 Raw DNA sequence reads are deposited in NCBI SRA bioproject PRJNA834786. Sample  
841 genotypes and analysis scripts are available on GitHub repository  
842 [https://github.com/wenzelm/hgigas\\_ddRAD](https://github.com/wenzelm/hgigas_ddRAD).

843

#### 844 [Benefit-Sharing Statement](#)

845 Benefits Generated: A research collaboration was developed with scientists from the  
846 countries providing genetic samples, the results of research have been shared with the  
847 provider communities and the broader scientific community. More broadly, our group is  
848 committed to international scientific partnerships, as well as institutional capacity building.

#### 849 [Author Contributions](#)

850 SP and AJJ designed the research and collected samples. SP performed labwork prior to  
851 RADSeq. MW analysed the data. All authors wrote and corrected the manuscript.

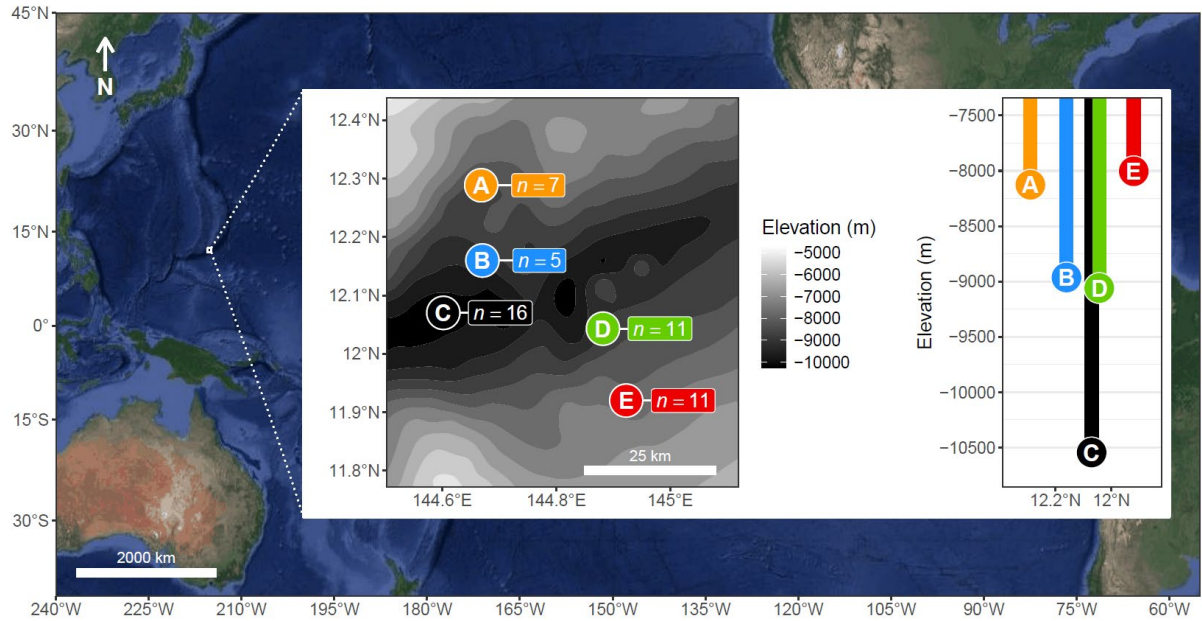
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855 Table and Figure legends

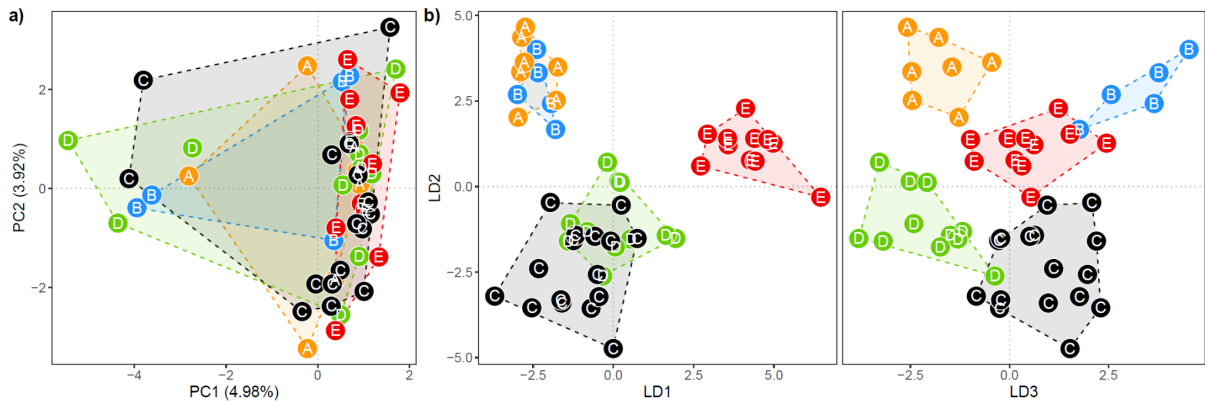
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858 **Figure 1:** Sampling locations of *Hironidellea gigas* across the Mariana Trench in the Pacific  
859 Ocean. Inset left: bathymetric contour map indicating sample coordinates, depth and sample  
860 size ( $n$ ). Inset right: sample depths organised by latitude (north-south transect). Maps obtained  
861 via *ggmap* (Kahle & Wickham, 2013) and *elevatR* (Hollister et al., 2021) R packages.  
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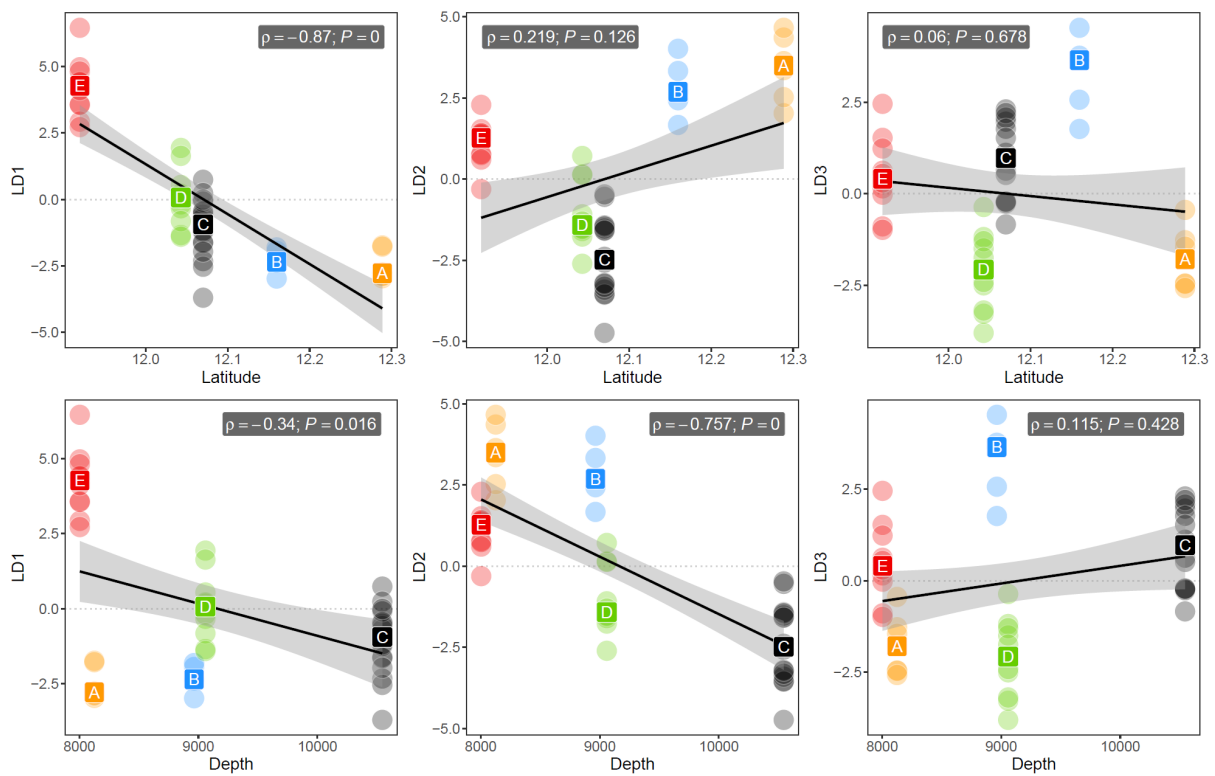
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865 **Figure 2:** Genetic structure among individuals from five sampling sites (A-E). a) Scatter plot  
866 of the first two eigenvectors (PC1 and PC2) following principal components analysis of  
867 genotypes. b) Scatter plots of the first three discriminant functions (LD2 vs LD1 and LD2 vs  
868 LD3) from discriminant analysis of principal components (DAPC) among genotypes.  
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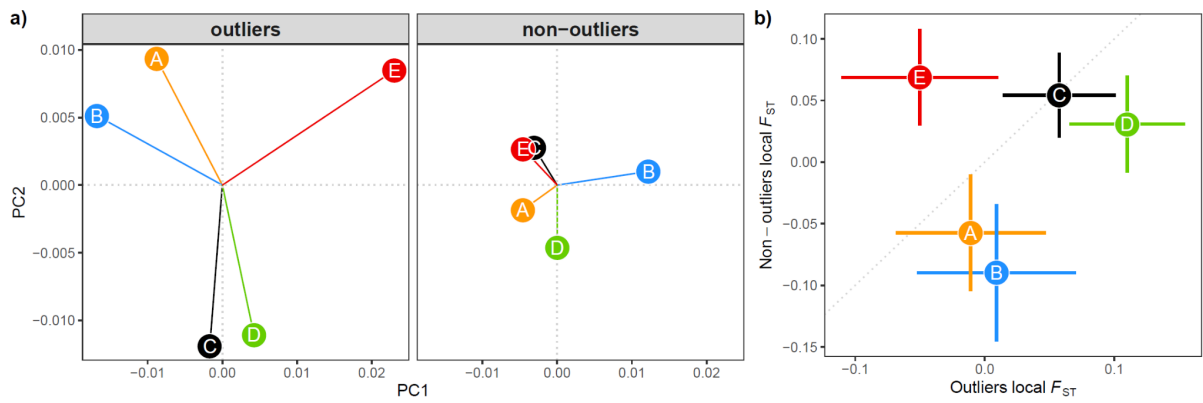


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872 **Figure 3.** Relationships between genetic structure, latitude and trench depth. Individual scores  
873 on the first three linear discriminant functions (LD1, LD2 and LD3) from DAPC are plotted  
874 against latitude (top panels) or depth (bottom panels) and are overlaid with a linear regression  
875 model. Spearman's rho and P value are indicated in panel corners.

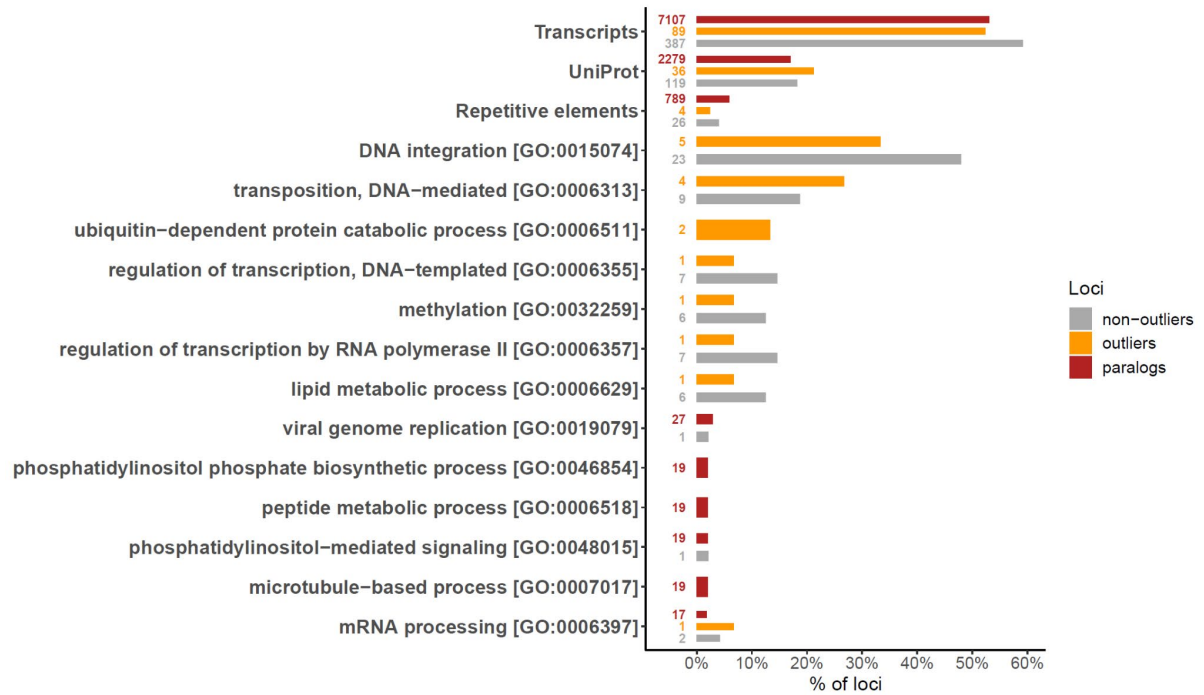
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879 **Figure 4.** Relationships between  $F_{ST}$ -based genetic structure in outlier versus non-outlier SNPs  
880 among sampling sites. a) Scatter plots of the first two principal components (PC1 and PC2) of  
881 pairwise  $F_{ST}$ . b) Scatter plot of population-specific (local)  $F_{ST} \pm 2 \cdot SE$  with a hypothetical 1:1  
882 relationship indicated by a dotted grey line.  
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885 **Figure 5.** Functional annotation of ddRAD loci identified as paralogs, divergence outliers or  
 886 singleton non-outliers. Numbers and proportions are given for loci annotated as transcripts  
 887 (*Hirondellea gigas* transcriptome), UniProt proteins (arthropoda), repetitive elements  
 888 (RepeatMasker) and the six most frequently observed GeneOntology (*biological process*)  
 889 terms in each locus category. Percentages for GO terms are ratios based on total loci with GO  
 890 annotations.

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