

28 **Introduction**

29 Rising seawater temperatures are bleaching corals and decimating reefs world-wide
30 (Hughes et al., 2018; Smale et al., 2019). How reef ecosystems will respond to warmer oceans
31 will be a function of genetic variation and how it is segregated among populations and
32 communities. A critical priority is to understand differences in physiological tolerance among
33 populations to identify the functional diversity within and among species available to cope with
34 warmer oceans (Edmunds et al., 2014; van Woesik, Sakai, Ganase, & Loya, 2011). To understand
35 how reef species will fare under warmer oceans and design successful management strategies, it
36 is critical to accurately delineate species (or independently evolving lineages) as it will allow us
37 to 1) study the physiological significance of genetically distinct but morphologically similar
38 groups within species complexes, 2) properly estimate the relative abundance of each cryptic
39 species, and 3) accurately quantify population sizes, genetic connectivity among populations, and
40 establish whether many cryptic lineages drive ecosystem-level changes with narrow distributions
41 or a few species with independent demographics.

42 The potential for cryptic species, morphologically indistinguishable yet genetically
43 distinct groups (Bickford et al., 2007; Knowlton, 1993), remains an overlooked aspect of coral
44 reef biology, even within common reef dwellers (Prada et al. 2008; Prada & Hellberg 2013; Prada
45 et al., 2014; Rosser, 2015; Warner, van Oppen, & Willis, 2015). Such hidden taxonomic diversity
46 frequently harbors distinct physiological variation and allows cryptic lineages to occupy different
47 habitats and respond differently to climatic variations (Struck et al., 2018). Failure to recognize
48 the different susceptibilities of cryptic species to climate change stressors can underestimate
49 threats to local populations, and ultimately to entire ecosystems (Fišer, Robinson, & Malard,

50 2018). Therefore, uncovering cryptic diversity stands as a major research priority to account for
51 ecosystem dynamics and forecasting future states (Bálint et al., 2011). The advent of genome-
52 wide data has facilitated the detection of such diversity, a task thus far not always achievable with
53 traditional molecular markers alone (Leaché & Oaks, 2017).

54
55 A key Caribbean reef builder severely affected by climate change is the mountainous star
56 coral, *Orbicella faveolata*. Currently placed under threatened status (NOAA & NMFS, 2014),
57 this massive coral has served as a model species for numerous physiological studies across the
58 Caribbean (Colombo-Pallotta, Rodríguez-Román, & Iglesias-Prieto, 2010; DeSalvo et al., 2010)
59 as well as for understanding the genetic basis of adaptation to warmer temperatures and bleaching
60 (Manzello et al., 2018; Wright, Correa, Quigley, & Davies, 2019). A recent study (Dziedzic,
61 Elder, Tavalire, & Meyer, 2019) found variation in the physiological response to thermal stress
62 among individuals of *O. faveolata* across reefs in Bocas del Toro, Panama, which the authors
63 interpreted as indicative of genetic variation within species to adapt to rising ocean temperatures.
64 Here, we reanalyzed their data and find that their study is more consistent with the presence of
65 multiple cryptic lineages (only one of which is heat-tolerant) rather than as a result of diversity
66 within a single species.

67

68 **Materials and Methods**

69 To test whether genetic variation in *O. faveolata* from Bocas del Toro represents multiple
70 cryptic lineages (rather than a single species), we carried out six analyses using the original data
71 from (Dziedzic et al., 2019): 1) a Discriminant Analysis of Principal Components (DAPC), 2) a

72 Bayesian clustering method, 3) a neighbor joining (NJ) distance tree based on the UPGMA
73 algorithm, 4) a species tree under the Bayesian multispecies coalescent framework of SNAPP, 5)
74 a phylogenomic tree using the coalescent method SVDquartets, and 6) a Bayes factor
75 contingency table to associate lineage assignment and bleaching tolerance scores.

76
77 Multi-locus raw sequence data (SRA: BioProject PRJNA413258) from 39 samples was
78 downloaded from the National Center for Biotechnology and Information (NCBI) server. We
79 mapped the data to a reference genome (Prada et al., 2016) and scored 383,160 Single Nucleotide
80 Polymorphisms (SNP) following the *dDocent* pipeline (Puritz, Hollenbeck, & Gold, 2014). We
81 removed all indels, retaining only biallelic SNPs that were genotyped in at least 70% of
82 individuals, had a minor allele frequency of 0.05, and a minimum coverage of 5x. To reduce
83 linkage disequilibrium, we kept only SNPs at least 1,000 bp apart. Remaining loci were screened
84 for statistical outliers potentially under strong selection using BayeScan (Foll, 2012). Our final
85 dataset consisted of 3,560 high-confidence SNPs. All filtering steps were done using VCFtools
86 (Danecek et al., 2011).

87
88 Initially we examined genome-wide variation through a DAPC using the R package
89 ADEGENET (Jombart, 2008; Jombart, Devillard, & Balloux, 2010). We estimated the number
90 of clusters (from 1 to 8) using *find.clusters* in ADEGENET and selected the optimal number of
91 groups along using a Bayesian Information Criterion (BIC) approach. To avoid overfitting, we
92 used the *optim.a.score* function to determine the number of PC axes to be retained. We then used
93 a Bayesian clustering method as implemented in STRUCTURE 2.3.4 (Pritchard, Stephens, &

94 Donnelly, 2000) to infer the number of genetic clusters (from 1 to 8) and potential admixture.
95 Structure was performed on unlinked SNP datasets and run in parallel using StrAutoParallel v 1.0
96 (Chhatre & Emerson, 2017) using an admixture model with correlated allele frequencies. Burnin
97 was set to 250,000 followed by 500,000 MCMC generations. We evaluated the optimal K in
98 STRUCTURE HARVESTER (Earl, 2012) following Evanno, Regnaut, and Goudet (2005).
99 Structure results were plotted in Structure Plot (Ramasamy, Ramasamy, Bindroo, & Naik, 2014).

100
101 To further test for the presence of multiple cryptic lineages, we used a NJ tree based on
102 the UPGMA algorithm using the pairwise genetic distance matrix of genotypes with the R
103 package APE (Paradis & Schliep, 2019), with 1,000 bootstrap replicates to assess branch support.
104 We then built a species tree under the Bayesian multispecies coalescent framework of SNAPP
105 v1.3 (Bryant, Bouckaert, Felsenstein, Rosenberg, & Roy Choudhury, 2012) implemented in
106 BEAST2 v2.5 (Bouckaert et al., 2018), with a path sampling of 24 steps (MCMC length =
107 1,000,000, pre-burnin = 1,000). We excluded individuals with missing data and a dataset of
108 unlinked biallelic SNPs, no outgroup, and the genetic-clustering results used for cluster
109 assignments (Kornilios et al., 2019). Because SNAPP is computationally intensive (our analysis
110 took one month), each group included only 3 individuals, for a total of 12 individuals. Marginal
111 likelihood estimates were obtained for each different model run. Each species delimitation model
112 was ranked by their marginal likelihood estimate following Leaché, Fujita, Minin, and Bouckaert
113 (2014). Log files were combined using Log Combiner v 1.1 and input into Tracer v1.6 (Rambaut,
114 Drummond, Xie, Baele, & Suchard, 2018). Convergence and ESS > 200 were assessed using
115 Tracer after a 10% burnin. A maximum clade credibility tree was generated with Tree Annotator

116 v 2.3 (R. Bouckaert et al., 2018). Both the consensus tree and all tree topologies were drawn in
117 DensiTree v2.2 (R. R. Bouckaert, 2010). Additionally, phylogenomic relationships among
118 individuals were inferred under the coalescent method SVDquartets 1.0 (Chifman & Kubatko,
119 2014) implemented in PAUP* 4.0b10 (Swofford, 2003). All possible quartets were evaluated
120 with prior assignment, using all individuals, and non-parametric bootstrapping with 1,000
121 replicates for branch support. iTOL was used for tree visualization and edition (Letunic & Bork,
122 2016).

123

124 Lastly, to test the null hypothesis of no difference in tolerance or susceptibility across
125 cryptic lineages, we performed a contingency table Bayes factor test (Morey & Rouder, 2015)
126 using the R package BayesFactor under an independent multinomial distribution, and estimated
127 the difference in probability of colonies being tolerant or susceptible given their cryptic lineage.

128

129 **Results**

130 Our genetic clustering analyses based on bi-allelic and unlinked SNPs suggest the
131 presence of three (STRUCTURE) or four (DAPC) genetic clusters (Figs. 1, 2, S1 and S2),
132 henceforth referred to as PAN_1, PAN_2, PAN_3, and PAN_4. The NJ tree with all individual
133 colonies recovered the same three divergent clades identified with Structure (PAN_1, PAN_3 and
134 Pan_4), plus an extra well-supported clade within PAN_1 that corresponded to the individuals
135 grouped as PAN_2 (Fig. 2).

136

137 The maximum clade credibility tree from SNAPP support four clades (Fig. 3A) according
138 to the marginal likelihood estimate (Table S1). The SVDquartets tree with individuals as terminal
139 branches was similar to the NJ tree (Fig. 2), favoring the relationships between PAN_1 with
140 PAN_3, and PAN_2 with PAN_4, with the latter being highly supported by bootstrap iteration
141 (Fig. 3B).

142
143 Finally, to see whether cryptic variation correlates with variation in bleaching response,
144 we used Bayes factors. We found that bleaching susceptibility, as reported in the original study
145 (Fig 4), and genotypic group are 42.7 times (Fig S3.A and Table S2) more likely to be associated
146 than not associated given the observed data. For instance, the difference in probability of PAN_2
147 being more tolerant to bleaching than PAN_3 is greater than 0.8 (Fig S3.B). Moreover, algal
148 symbiont genera partitioned differentially among the lineages (Fig S4). PAN_1 seems highly
149 promiscuous associating with species from four genera: *Symbiodinium*, *Brevolium*,
150 *Cladocopium*, and *Durusdinium*, while the other lineages are mostly restricted to *Durusdinium*
151 species.

152 153 **Discussion**

154 Our reanalysis of Dziedzic et al.'s data suggests that *O. faveolata* from Bocas del Toro
155 Reefs is not a single cosmopolitan species but rather is composed of multiple cryptic lineages
156 with independent evolutionary trajectories. Congruence among species delimitation methods
157 suggests that four cryptic lineages coexist under the nominal species *O. faveolata* in this
158 Caribbean region. Furthermore, thermal tolerance occurs in three of these lineages (PAN_1,
159 PAN_2, and PAN_4), with one of them (PAN_2) being exclusively composed of heat-tolerant

160 colonies, while PAN_3 composed only of heat-sensitive ones. Remarkably, symbiont affinity in
161 these four lineages exhibits dominance of the genus *Durussdinium* in three of them (PAN_2,
162 PAN_3, and PAN_4). In contrast, only one lineage (PAN_1) presents a promiscuous association
163 pattern of symbiosis and associates with species from four genera.

164

165 *Cryptic lineages and thermal tolerance*

166 The prevalence of heat-tolerant colonies within the PAN_1 and PAN_2 lineages implies that,
167 under recurrent or stronger bleaching events, the survival of less tolerant lineages will be severely
168 compromised. Given the dramatic pace of the Caribbean reef decline, without a thorough
169 appreciation of the cryptic species composing this pivotal ecosystem, silent diversity losses will
170 take place as we grapple to capture and protect the true existing biological diversity (Richards,
171 Berry, & van Oppen, 2016). Consequently, a deeper understanding of how cryptic coral lineages
172 vary physiologically is critical to forecasting reef ecosystem composition and resilience.

173 Ecosystem resilience to disturbances hinges not only on individual species trajectories but on the
174 standing community composition under regional conditions and their temporal variability
175 (Edmunds et al., 2014). Therefore, the disappearance of cryptic evolutionary lineages decreases
176 evolutionary potential by disrupting current diversification processes that will impact future
177 biodiversity (Bálint et al., 2011).

178 As we increase our ability to delineate species using genomic data, we need to
179 accommodate for the ensuing ecological implications this broader recognition of diversity will
180 bring to our understating of species interactions, symbiotic relationships, physiological
181 thresholds, and ecosystem integrity. Coral reefs' future largely lies in their physiological response

182 to heat, which depends directly on their symbiotic algal counterpart. Therefore, a potentially
183 more restricted specificity among coral-algal symbioses, as it may be occurring in cryptic coral
184 species, challenges our notions of the thermal tolerance corals can withstand under increasing
185 warming sea surface temperatures (Thornhill, Lewis, Wham, & LaJeunesse, 2014). Although this
186 poses an even bleaker scenario for coral reefs, it opens up an opportunity for more integrative
187 studies that take into account cryptic diversity to define whether coral thermal tolerance is more
188 or less widely distributed among the most common reef-builder species. Importantly, this would
189 better inform conservation and management decisions because those species we thought of being
190 cosmopolitan might be local or environmentally confined. Coral endangered species, such as *O.*
191 *faveolata*, would be in even at greater peril as their population sizes must have been
192 overestimated.

193
194 *The importance of delineating cryptic lineages to identify the genomic architecture to bleaching*

195 In *O. faveolata*, microsatellite loci suggest population connectivity at large spatial (>
196 1,000 km) scales over its entire distribution (Severance & Karl, 2006). A more recent study,
197 however, found population structure among close reefs to be at odds with the predominant ocean
198 circulation patterns in the area (Rippe et al., 2017). This discrepancy might not only reflect
199 inherently difficulties associated with the interpretation of these markers (Fukami, 2008) but may
200 also suggest the presence of unrecognized diversity, as we have uncovered here. It remains to be
201 seen whether the four lineages that we delineated here occur across the Caribbean and the Gulf of
202 Mexico, or if they occur under specific habitats at smaller geographical scales (Prada et al. 2008).
203 Similarly, other scleractinian corals exhibit fine genetic structuring along with restricted

204 geographic areas that are more consistent with the existence of cryptic lineages than within-
205 species genetic diversity (McFadden et al., 2017; Ohki, Kowalski, Kitanobo, & Morita, 2015;
206 Richards et al., 2016; Warner et al., 2015). Therefore, accurate delimitation of species boundaries
207 is crucial to quantify species diversity and elucidate biologically meaningful patterns of gene
208 flow and dispersal among populations (Prada & Hellberg, 2013; Wham & LaJeunesse, 2016).

209

210 Likewise, genome-wide association studies (GWAS) when there is underlying cryptic
211 diversity (stratification), fail to account for the non-independent distribution of the genetic
212 variation, which is constrained by the particular evolutionary history of each lineage (Sul,
213 Martin, & Eskin, 2018). In this case, genetic variants found to be associated with thermal
214 tolerance might have been unwarily conflated with cryptic lineages. We did not pursue a GWAS
215 reanalysis because the small number of individuals composing each cryptic lineage would have
216 further underpowered the inferences made from an already small sample size.

217

218 *Cryptic lineages and symbiotic relationships*

219 Highly specialized symbiotic relationships between genetically distinct hosts and their
220 counterparts can underlie tight and long-lasting patterns of coevolution, reflecting unique
221 prospects to withstand environmental fluctuations under particular environments, such as those
222 projected under a climate change scenario (Voolstra et al., 2011). For hard corals, such specificity
223 to engage in particular physiological and ecological interaction with their endosymbiotic algae
224 translates directly into specific environmental thresholds given their dependence on symbiosis for
225 energy acquisition (Ziegler, Eguíluz, Duarte, & Voolstra, 2018). Taxonomic revision of

226 symbiotic microalgae (Symbidinaceae) found in corals and other invertebrates has revealed that
227 what it was once assumed to be a single panmictic species, it encompasses multiple genera and
228 species with a more restricted suit of hosts (LaJeunesse et al., 2018). Whether what we
229 considered as coral's potential for harboring diverse endosymbiotic communities might have
230 reflected not only the taxonomic uncertainty on symbionts but also the existence of cryptic
231 diversity among host coral species. Under this view and in light of our reanalysis, the original
232 conclusion that *O. faveolata* could naturally harbor genetic variation to adapt to rising ocean
233 temperatures (Dziedzic et al., 2019) could be a case in point of this richer diversity both on the
234 coral host as well as its symbionts.

235

236 *Conclusions*

237 Accurately delineating species, even cryptic ones, is key for reef conservation and
238 management because it allows us to accurately estimate the relative abundance and population
239 sizes of these biological units. This, in turn, helps us understand patterns of gene flow to
240 determine genetic connectivity among populations. Moreover, accurate detection and description
241 of cryptic species enhance our ability to establish if ecosystem-level changes are driven by
242 various cryptic lineages narrowly distributed, or instead by few species with overlapping or
243 distinct demographics.

244

245 Ultimately, whether coral populations can survive to the rapid increase in water
246 temperature while maintaining ecosystem resilience, is still under heavy scrutiny. Given the
247 current degradation experienced by coral reefs worldwide, the projected increases in

248 environmental variability, and regardless of the predominant mechanism that corals may have to
249 cope with future changes, identifying evolutionary units of biological diversity is critical to
250 consolidate conservation efforts. Failure to do so in either host or symbiont leads to an
251 overestimation of the ecological and physiological ranges of individual species, undermining our
252 view of how reefs will respond to rapidly changing conditions.

253

References

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428 **Data Accessibility**

- 429
- 430 • Sequence data archived at NCBI's Sequence Read Archive (SRA) BioProject
 - 431 PRJNA413258.
 - 432
 - 433 • Scripts used for analysis will be made publicly available at <https://github.com/matiasgoco>

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436 **Author Contributions**

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438 M. G-C. and C.P. conceived the study. M.G-C compiled and reanalyzed data. M. G-C. and C.P.

439 wrote the paper. Both authors read and approved the paper.

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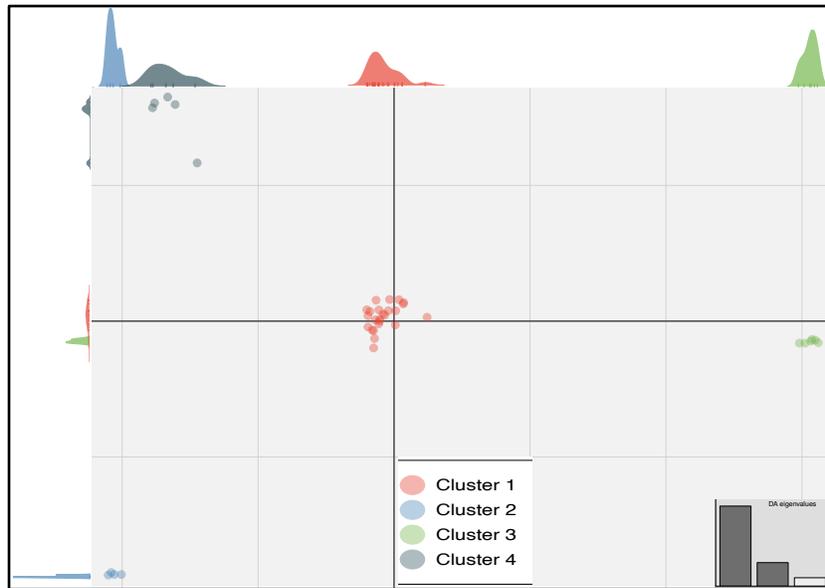
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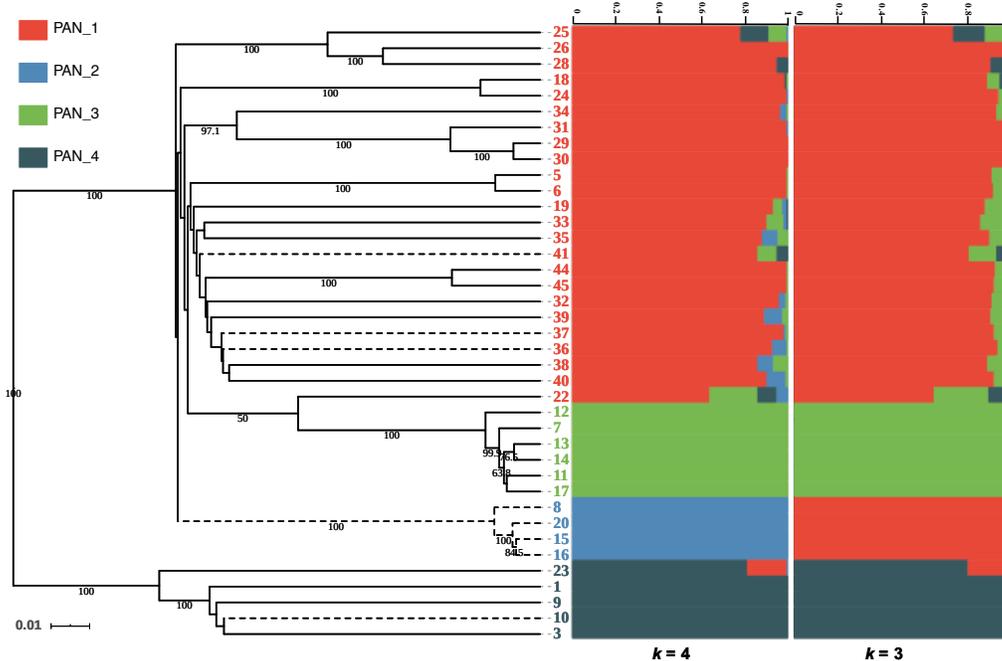
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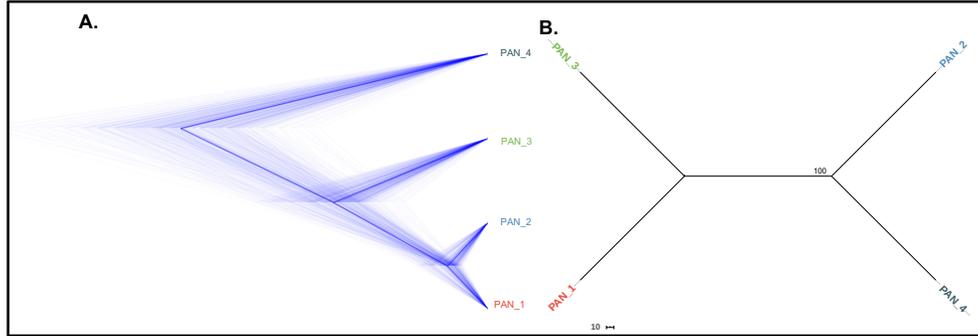


474
 475 **Figure 1.** Discriminant analysis of principal components (DAPC) (Optimal $k = 4$ and number of PCA =
 476 6). Note the full segregation of the data, suggestive of non-randomly reproducing groups.
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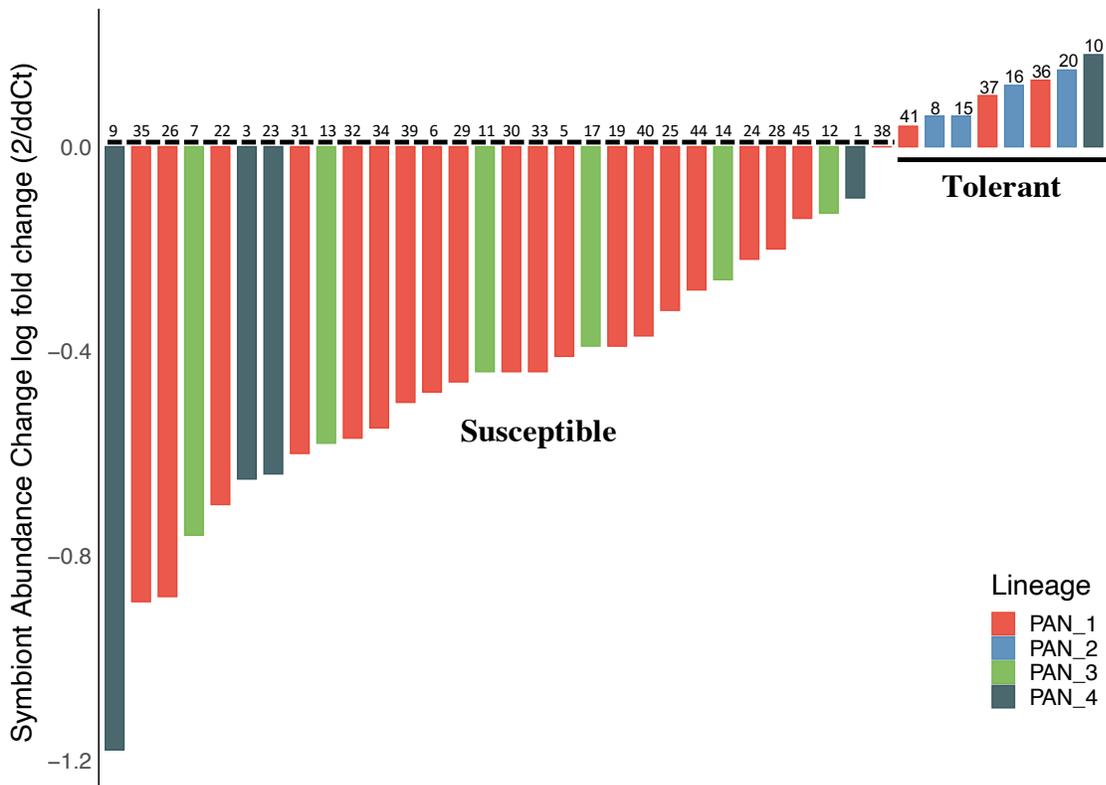
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 479 **Figure 2.** Neighbor Joining (NJ) tree using the pairwise genetic distance matrix (node numbers are
 480 bootstrap support values), depicting the groups inferred from the genetic clustering methods (dashed
 481 branches represent heat-tolerant samples). Hierarchical Bayesian population clustering with
 482 STRUCTURE depicting $k = 3$ (optimal) and $k = 4$.

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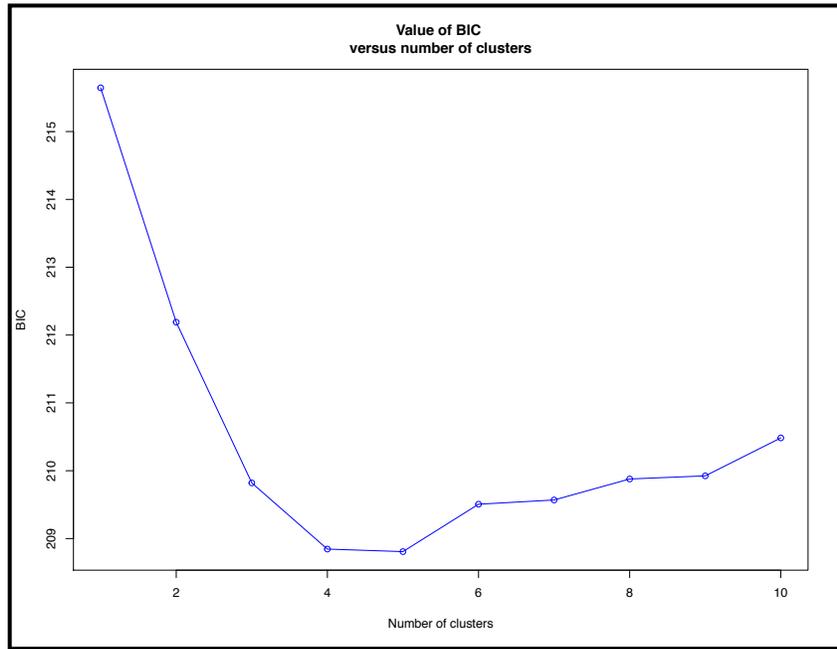
Figure 3. A) Trees inferred from the Bayesian coalescence analysis of SNAPP (the congruent tree is in thick blue), B) SVDquartets tree from analysis in PAUP*.



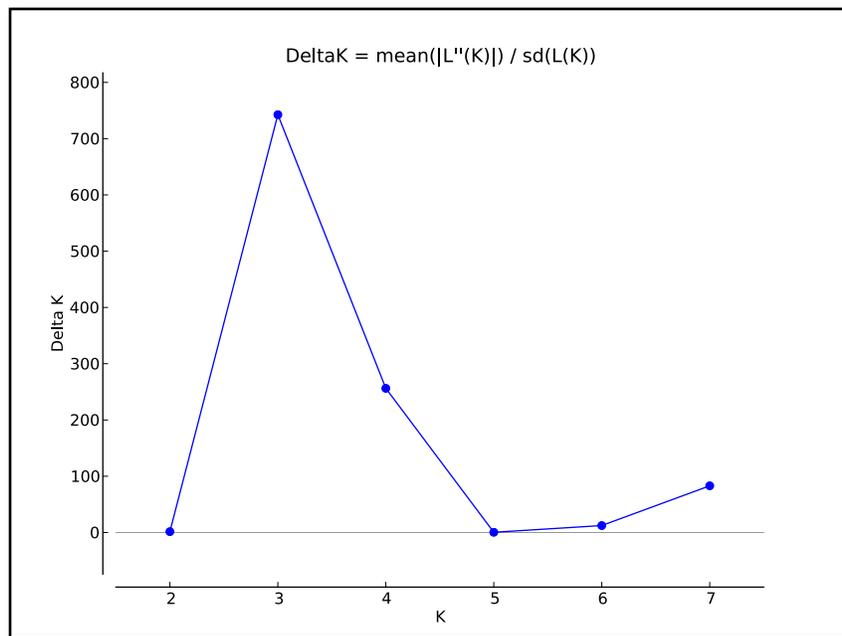
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Figure 4. Log-fold change (2^{-ddCt}) algal symbiont abundance per colony using qPCR after four weeks of experimental conditions (Modified from Dziedzic et al., 2019). Colors denote the cryptic lineages uncovered in this study.

493 **Supplementary material**
494



495 **Figure S1.** Value of Bayes information criterion versus number of clusters found in DAPC.
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498 **Figure S2.** Most probable number of clusters according to the value of Delta K in Structure.
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No. Groups	Marginal Likelihood Estimate	Rank
2	7292164.4623	3
3	11303237.4275	2
4	14163377.8363	1

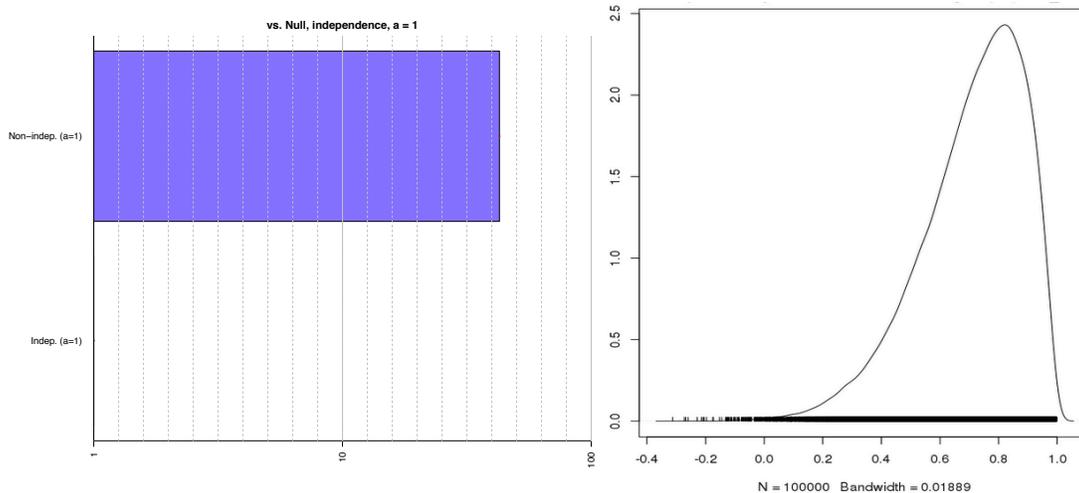
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Table S1. Path sampling results for the four clusters delimitation models.

Bleaching	Genotypic group			
	PAN_1	PAN_2	PAN_3	PAN_4
Tolerant	3	4	0	1
Susceptible	20	0	6	4

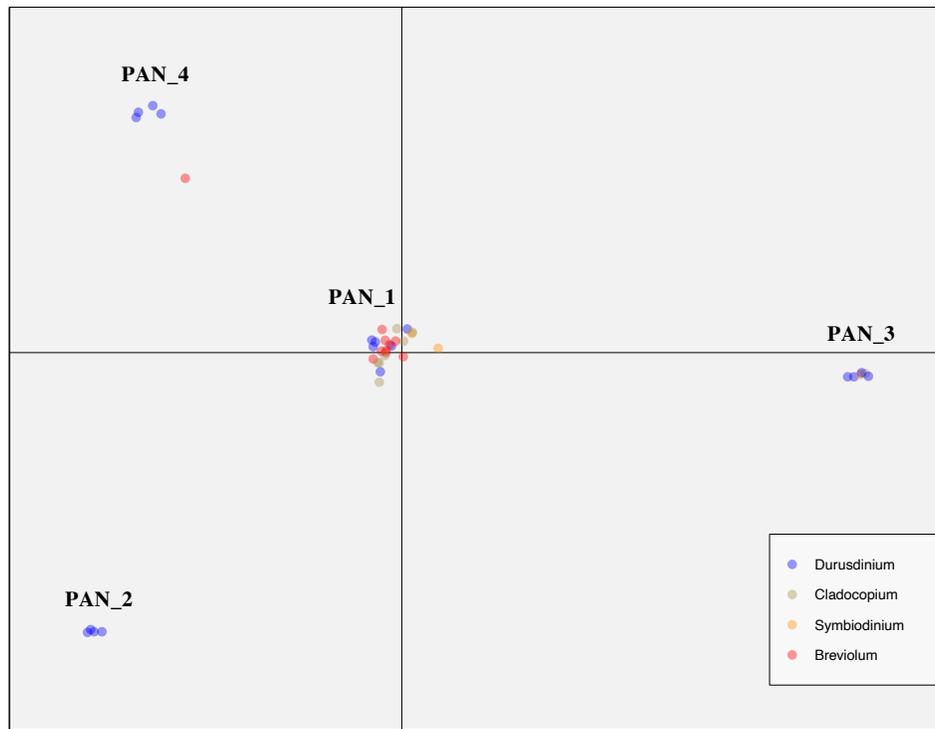
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Table S2. Frequencies for tolerant and susceptible colonies to bleaching according to their genotype group



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Figure S3. A) Bayes factor plot of contrasting hypothesis of non-association (H0) vs association (H1) between genotypic group and bleaching tolerance. **B)** Increase in probability of bleaching tolerance of genotypic group PAN_2 when compared to PAN_3.



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Figure S4. Discriminant analysis of principal components (DAPC) as in Fig.1 but colored by symbiont genera, previously known as clades.