1 Title: Genetic variation in heat tolerance of the coral *Platygyra daedalea* indicates potential for
2 adaptation to ocean warming.
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4 Running head: Genetics of corals' heat tolerance
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Abstract

Climate change induced increases in global ocean temperature represent the greatest
threat to the persistence of reef ecosystems and most coral populations are projected to
experience temperatures above their current bleaching thresholds annually by 2050. Adaptation
to higher temperatures is necessary if corals are to persist in a warming future. While many
aspects of heat stress have been well studied, few data are available for predicting the capacity
for adaptive cross-generational responses in corals. Consistent sets of heat tolerant genomic
markers that reliably predict thermal tolerance have yet to be identified. To address this
knowledge gap, we quantified the heritability and genetic variation associated with heat
tolerance in *Platygyra daedalea* from the Great Barrier Reef. We tracked the survival of
quantitative genetic crosses of larvae in a heat tolerance selection experiment. We also identified
allelic shifts in heat-selected survivors compared with paired, non-selected controls. The narrow-
sense heritability of survival under heat stress was 0.66 and a total of 1,069 single nucleotide
polymorphisms (SNPs) were associated with different survival probabilities. While 148 SNPs
were shared between several experimental crosses, no common SNPs were identified for all
crosses suggesting that specific combinations of many markers are responsible for heat tolerance.
However, we found two regions that overlap with previously identified loci associated with heat
tolerance in Persian Gulf populations of *P. daedalea* reinforcing the importance of these markers
for heat tolerance. These results illustrate the importance of high heritability and the complexity
of the genomic architecture underpinning host heat tolerance. These findings suggest that this *P.*
daedalea population has the genetic prerequisites for adaptation to increasing temperatures. This

study also provides knowledge for the development of high throughput genomic tools to screen
for variation within and across populations to enhance adaptation through assisted gene flow and
assisted migration.
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Introduction
Rising ocean temperatures induced by anthropogenic climate change have increased the
frequency and severity of coral bleaching events worldwide to the point that it is recognized as
the primary threat to coral reefs (Hutchings et al., 2019; Souter et al., 2020). Bleaching is the
process in which dinoflagellate symbionts (Family Symbiodiniaceae) are lost or expelled from
host corals (Lajeunesse et al., 2018). As symbiont photosynthesis provides the majority of host
nutritional requirements, mass bleaching caused by marine heat waves can result in rapid and
widespread coral mortality (Glynn, 1993). Global and regional bleaching events have resulted in
a substantial loss of coral cover worldwide over the last few decades (Hughes et al., 2017, 2018).
Over the last five years alone, 30% of reefs in the Great Barrier Reef (GBR) showed decreases in
coral cover (Hughes et al., 2019; Hughes et al., 2018; Long Term Reef Monitoring Program,
2019).
For coral reef ecosystems to persist in an ocean that will continue to warm for the
foreseeable future, corals will need to rapidly become more tolerant or resilient to higher
temperatures. Consequently, there is an urgent need to understand the adaptive capacity of corals
to this threat across ocean basins and populations to inform conventional management
approaches. This knowledge is also required for restoration and adaptation programs using
assisted gene flow and assisted migration to enhance the tolerance of local populations (van
Oppen et al., 2015) as adaptive capacity can help determine which populations should be used to
stock nursery and breeding programs. Working knowledge of the genomic regions that influence
heat tolerance, the extent to which coral host genetics contributes to heat tolerance, and its cross
generational persistence in a population will be essential to predict corals' responses to climate
change into the future.
Mechanisms of thermal and bleaching tolerance are diverse, and both hosts and
symbionts are known to contribute to the holobiont phenotype. For example, different symbiont
species can confer different thermal tolerances in the same host species (Fuller et al., 2020). In
addition, the presence of low background densities of heat tolerant symbiont species can aid in
recovery from heat stress (Bay et al., 2016; Berkelmans & van Oppen, 2006; Manzello et al.,
2019; Oliver & Palumbi, 2011; Silverstein et al., 2014). Many studies have also documented
genetic variation in coral host heat tolerance that supports the potential for adaptation (Bay et al.,
2017; Dixon et al., 2015; Howells et al., 2016; Kenkel et al., 2013; Matz et al., 2020; Meyer et
al., 2009; Woolsey et al., 2014). Differential gene expression has been associated with heat
tolerance in a number of coral species (e.g., Barshis et al., 2013; Kenkel et al., 2016; Ruiz-jones
& Palumbi, 2017; Traylor-Knowles et al., 2017). While some studies have identified genomic
markers in coding and non-coding genes that are associated with a heat tolerant phenotype, yet
no consensus set of 'candidate genes' has emerged (Bay & Palumbi, 2014; Dixon et al., 2015;
Dziedzic et al., 2019; Fuller et al., 2020; Jin et al., 2016) and the genetic architecture
underpinning heat tolerance remains largely unknown.
Coral heat tolerance is likely a complex polygenic trait controlled by many alleles of
small effect which complicates efforts to identify loci in part because of the statistical challenge
of distinguishing signal from noise (Fuller et al., 2020). Polygenicity is well documented in

diverse traits in agricultural species, model organisms, and humans (reviewed in Sella & Barton,
2019) but remains difficult to detect because most genotyping methods are biased towards
detecting loci of large effect (Wellenreuther & Hansson, 2016). Although methodological
challenges remain, genome-wide association studies have yielded insight into the genomic basis
of key performance traits in a number of species. For example, in plants, allele frequency
analysis has identified markers associated with traits such as yield, drought tolerance and heat
tolerance (Singh et al., 2017; Venuprasad et al., 2009; Vikram et al., 2012; Wang et al., 2019).
Consequently, great interest remains in identifying and developing genomic marker sets for
conservation, restoration and adaptation applications (Baums et al., 2019).
Even in the absence of predictive markers, knowledge of trait heritability can inform
breeding programs (Falconer & Mackay, 1996; Lynch & Walsh, 1998) and the development of
predictive models (Gienapp et al., 2013). If variation in a fitness-related phenotype is heritable
and not otherwise constrained, then selection can act to further propagate it within populations
(Charmantier & Garant, 2005). Narrow-sense heritability (h^2) describes the total phenotypic
variance in a trait that can be attributed to parental or additive genetics and is used to determine
the genetic contribution to traits such as heat tolerance (Falconer & Mackay, 1996; Lynch &
Walsh, 1998). To model the persistence of a trait in a population with the breeder's equation,
estimates of narrow-sense heritability are essential for calculating selection differentials
(Falconer & Mackay, 1996; Lynch & Walsh, 1998). Specifically, estimates of narrow-sense
heritability of heat tolerance in corals will allow us to calculate potential rates of adaptation
under various selection scenarios.
Current estimates of narrow-sense heritability suggest that the genetic variation to
support adaptation of increased heat tolerance is present in coral populations in many different
regions. For example, a heritability value of 0.89 was found for bleaching resistance in *Orbicella*
faveolata in the Caribbean (Dziedzic et al., 2019) and values between 0.48 and 0.74 for heat
tolerance in *Platygyra daedalea* in the Persian Gulf (Kirk et al., 2018). Heritability was similarly
high for two *Acropora* species from the GBR: (0.87) for heat tolerance in the larvae of the coral
A. millepora (Dixon et al., 2015) and (0.93) for survival during heat stress in the larvae of *A.*
spathulata (Quigley et al., 2020). However, heritability can vary substantially between species,
as well as between populations of conspecifics, and is influenced by the environment in which it
is measured (Falconer & Mackay, 1996; Lynch & Walsh, 1998). Therefore, a heritability
estimate from the Persian Gulf should not be used to estimate selection for heat tolerance or
potential for adaptation in a population on the GBR, but a comparison among regions can inform
understanding of the evolutionary potential of the species. To understand the heritability of heat
tolerance, selection for heat tolerance, and the potential for adaptation in GBR populations, more
heritability estimates of heat tolerance from populations on the GBR are needed (Bairos-Novak
et al., 2021).
Cross ocean basin comparisons of heritability and the genetic basis of heat tolerance are
important for assessing whether results from one region can be extrapolated to another.
Platygyra daedalea is a good candidate species for this type of study because it is widely
distributed in the Indo-Pacific and also occurs in the much warmer Persian Gulf (D'Angelo et al.,
2015). The Gulf regularly experiences temperatures of 33 - 34˚C in the summer and the
bleaching threshold for Gulf populations of *P. daedalea* is between 35 and 36˚C (Howells et al.,
2016; Riegl et al., 2011). Indeed, the bleaching threshold temperature for the coral assemblages
is 3 - 7˚C higher in the Gulf than in other regions (29 - 32˚C), including the GBR (Berkelmans,
2002; Coles & Riegl, 2013; Hoegh-Guldberg, 1999). In addition, heritability estimates and

preliminary data on genomic associations exist for a population of *P. daedalea* in the Persian
Gulf (Kirk et al., 2018). This provides a unique opportunity to compare heritability and the
genomic basis of heat tolerance between populations from these relatively warmer and cooler
environments, which is important for understanding how increased thermal tolerance has evolved
and for determining whether the genetic mechanisms underpinning heat tolerance in this species
are conserved across ocean basins.
To understand the potential for a GBR population of *P. daedalea* to persist in a warming
environment through selection and genetic adaptation we estimated narrow-sense heritability and
identified genomic regions associated with heat tolerance. We also compared the genomic basis
of heat tolerance in this species on the GBR to those of the Persian Gulf *P. daedalea* (Kirk et al.,
2018). Estimating heritability of heat tolerance allowed us to quantify the host genetic
contribution to this trait. We identified genomic markers associated with heat tolerant
phenotypes by conducting a heat selection experiment and then assessing changes in allele
frequency in the survivors. Sequencing our heat-selected pools of bi-parental crosses allowed us
to identify markers associated with heat tolerance and examine how widespread these markers
were across families in both genic and intergenic genomic regions. This study addresses
fundamental questions regarding the adaptive potential of corals to heat stress and provides
evidence of their ability to survive and persist in a changing environment.
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159 **Materials and Methods**
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Experimental crosses and larval culture
Six colony fragments of *Platygyra daedalea* were collected from Esk Reef, an inshore
reef in the central region of the Great Barrier Reef (18˚ 46'21.95" S, 146˚ 31' 19.33" E) under
Australian Institute of Marine Science (AIMS) permit G12/35236.1. The corals were maintained
in the National Sea Simulator at AIMS at Esk Reef ambient temperature (27˚C) in flow-through
sea water aquaria with 25% reduced ambient sunlight. In November 2016 (the 5th night after the
full moon, at 18:30) the six parental colonies were isolated in ~60L bins filled with 0.04-µm
filtered sea water (FSW) (Weeriyanun et al., 2021). Colonies spawned between 18:30-19:00
GMT+10. Gamete bundles were collected and washed with FSW. Sperm and eggs were then
separated using 60 µm mesh sieves and eggs were thrice rinsed to eliminate residual sperm. We
combined gametes from pairs of colonies according to a partial diallel design (Figure 1) to
produce 10 families (hereafter termed crosses). A full diallel cross was not possible because
colonies D and E did not produce enough eggs for the purposes of this experiment. Each cross
was maintained at a density of 0.5 to 1.0 planula larvae per ml in separate 12L culture vessels
with flow-through FSW at 27˚C and gentle aeration.
176
Heat stress experiments
To quantify the variation in coral heat tolerance, we measured larval survival during
controlled heat stress exposure following (Dixon et al., 2015). Six replicates of 20 larvae from
each cross were randomly allocated to each of two temperature treatments: ambient reef
temperatures from the site of collection (27˚C, Supplemental Figure 1) or an acute heat stress
(35.5˚C). Larval replicates were contained in individual mesh bottomed 20ml wells (to allow for
water circulation) within 60L flow through FSW tanks (N = 6 control and N = 6 heat per cross).
A ramping rate of 1˚C per hour was used to reach 35.5˚C and water temperatures were

maintained thereafter by computer-controlled heaters. Mortality was monitored by manually
counting with a hand counter the number of surviving larvae in each well every 12 hours for the
next ten days. Larval survival data can be found at
https://github.com/hollandelder/Platygyradaedalea_HeritabiltyandMarkersofThermalTolerance.
189
Statistical analysis of heat tolerance
Survival under heat stress was modeled using a Kaplan Meier Survival Analysis and a
Cox Proportional Hazards (CPH) model that compared larval survival between crosses at 35.5˚C
and 27˚C (Cox, 2007; Kaplan and Meier, 2017) using the Survival package in R (Therneau,
2015). Our model fit individual survival, allowing slopes to vary for temperature, cross, and the
interaction of temperature and cross. We also calculated the uncensored survival probabilities as
in Kirk et al., 2018. The code for the Kaplan Meier survival plots and CPH model can be
referenced at
https://github.com/hollandelder/Platygyradaedalea_HeritabiltyandMarkersofThermalTolerance.
CPH in the Survival package in R does not account for random effects, so we also used a
generalized linear mixed model (GLMM) to determine if position contributed to variation in
survival. Please see the supplemental text for complete GLMM analysis details.
We used survival fractions at the experimental endpoint and Kaplan Meier survival
probabilities to estimate narrow-sense heritability of the variation in heat tolerance (Falconer &
Mackay, 1996; Wilson et al., 2010) using a restricted maximum likelihood (REML) mixed-
effects model implemented in the software package WOMBAT (Meyer, 2007) with the fixed
effects of temperature, sire, and dam.
207
Heat stress selection experiment to investigate the genomic basis for heat stress tolerance
We conducted a selection experiment to identify changes in allele frequency in response
to acute heat exposure. Four replicates of 500 larvae per cross were allocated into 1L flow-
through culture containers for a total of 2000 larvae in each of two temperature treatments (27˚C
and 35.5˚C). Treatment conditions and ramping profiles were identical to those described for the
survival analysis. We sampled 100 larvae from each replicate for a total of 400 larvae per
treatment after 53 h, the time at which more than 50% mortality was evident in most crosses, and
larvae were preserved in 100% ethanol for subsequent genetic analysis.
216
Genomic library preparation
Genomic DNA from each replicate pool was extracted from preserved samples using the
Omega Bio-tek E.Z.N.A. Tissue Kit (Omega Bio-tek, Norcross, GA) and genotyped using
2bRAD, a sequencing-based approach for SNP genotyping (Wang et al., 2012) that has been
previously applied in corals https://github.com/z0on/2bRAD_denovo (Dixon et al., 2015;
Howells et al., 2016). Libraries were sequenced on four lanes of Illumina HiSeq2500 (50bp SE)
at Ramaciotti Centre for Genomics at the University of New South Wales (Sydney, Australia).
224
Sequence processing
Raw reads were quality filtered to remove any that had ten or more bases with a Q-score
< 30. This is a stringent filter, but 95% of reads were retained for most samples at this threshold
(Supplemental Table 1). Reads exhibiting a 12bp match to the Illumina sequencing adaptors
were also removed. Remaining high quality reads were mapped against a 2bRAD reference
previously developed from larvae of Persian Gulf *P. daedalea* (SRA accession: SRP066627,

Howells et al., 2016) to facilitate subsequent cross-region comparisons using the gmapper
command of the SHRiMP software package (Rumble et al., 2009) with flags --qv-offset 33 -Q --
strata -o 3 -N 1 as in (Kirk et al., 2018). On average, 91% of reads mapped to the reference
(Supplemental Table 1). After alignment to the reference, we further filtered reads with
ambiguous (i.e. those mapping equally well to two or more different regions in the reference) or
weak matches (those that did not span at least 33 of a 36 base sequence and matched fewer than
30 bases within that span of a sequence). SNPs were called using the CallGenotype.pl script
found at https://github.com/Eli-Meyer/2bRAD_utilities as in (Dziedzic et al., 2019; Kirk et al.,
2018) using observed nucleotide frequencies and a minor allele frequency of 0.25 to be
considered heterozygous. Monomorphic loci were removed as well as those occurring in fewer
than 25% of samples within a cross. From here on, we refer to mapped, quality filtered, and
genotyped reads as tags. Samples with too many missing tags (<10,000 remaining) following
read filtering were also removed. We required that each tag be genotyped in at least seven out of
eight replicates for each cross to compare alleles at that locus across replicates in each treatment.
Tags with more than two SNPs were also removed because they are likely to come from
repetitive regions (Treangen & Salzberg, 2012). Finally, we down-sampled to one SNP per tag to
control for non-independence and linkage but prioritized minimizing the number of missing
SNPs. That is to say, the SNP found in most samples for a given tag was the one kept. If a
sample did not have a SNP in the same base pair position in the tag compared to other copies of
the tag, the SNP in the alternative position was kept. All 2bRAD genotyping and sequence
filtering was conducted using scripts available at (https://github.com/Eli-Meyer/2brad_utilities).
252
Allele frequency analysis
We then compared allele frequencies between control larvae and the survivors of heat
stress in replicate samples from each treatment within each cross. Since these samples
represented pools of large numbers of larvae (100) rather than single diploid individuals, we
followed the same general approach previously described as Pool-Seq (Kofler et al., 2017). We
applied a stringent coverage threshold of 80x within cross to minimize sampling error that is
likely to distort estimates of allele frequencies at lower coverage considering the number of
individuals in each sample (Zhu et al., 2012). To find the percentage of loci not genotyped in all
crosses following filtering, we extracted and compared the number of tags to all SNPs genotyped
in each cross using scripts available at
https://github.com/hollandelder/Platygyradaedalea_HeritabiltyandMarkersofThermalTolerance.
We interpreted SNP frequencies at a locus as allele frequencies in the pool. To test for
differences in allele frequencies between control larvae and those surviving heat stress, we used
logistic regression to analyze the number of observations of each allele in each sample with
generalized linear models in R (glm with argument "cross=binomial(logit)") where treatment
was a fixed effect (R Core Team, 2017). We used a Benjamini-Hochberg false discovery rate
(FDR) correction to adjust p-values (Waite & Campbell, 2006). An allele was determined to be
significant at an FDR-corrected p-value ≤ 0.05. We also calculated the proportion of alleles that
increased or decreased in frequency as a result of heat stress treatment for each cross.
272
Comparison of genotyped SNPs between crosses
All scripts and code for recreating the analyses described below can be found at
https://github.com/hollandelder/Platygyradaedalea_HeritabiltyandMarkersofThermalTolerance.
Pearson correlation tests as implemented in the R package ggpubr (Kassambara, 2020) tested for

correlations between the number of significant markers and the number of markers genotyped in
each cross, as well as the number of significant markers and the mean survival following ten
days of heat stress. We calculated the genetic distance between sets of parents by conducting
pairwise comparisons of the high coverage loci using a custom Perl script which tested for
numbers of shared markers in each parent set. The number of shared markers determined genetic
distance between parents.
We then examined the presence and absence of SNPs between all pairwise combinations
of crosses. To determine if the difference in significant SNPs identified between crosses was the
result of filtering and genotyping steps, we calculated the percentage of tags missing in all
crosses. Tidyverse and dplyr R software packages were used to extract the number of tags before
and after all the filtering steps, and at the 80x coverage threshold and to compare those to all tags
genotyped in each cross (Wickham et al., 2019, 2020). To calculate the percentage of tags shared
between each pair of crosses we compared the total significant tags genotyped in one cross to the
sum of the significant tags genotyped in each focal pair. We also found whether significant tags
were found to be insignificant or not genotyped in each cross. For tags with overlap in four or
more crosses, we determined whether the allele in the cross was increasing or decreasing in
frequency.
294
Functional enrichment analysis
To investigate the functional implications of candidate tolerance loci, we mapped the
2bRAD reference (SRA accession: SRP066627, Howells et al., 2016) to the *P. daedalea*
transcriptome (SRA accession PRJNA403854 Kirk et al., 2018) using Bowtie2 (Langmead et al.,
2009) requiring a minimum MAPQ score of 23. Gene names and ontology terms were assigned
to mapped 2bRAD tags based on the existing transcriptome annotations (Kirk et al., 2018). For
tags that were identified in the transcriptome, we determined if the allele in the tag was
increasing or decreasing in frequency. A rank-based gene ontology (GO) enrichment analysis
was conducted to identify significantly enriched terms among annotated tags containing
significant SNPs as identified in the allele frequency analysis using the GO_MWU suite of
scripts (Wright et al., 2015) , https://github.com/z0on/GO_MWU). The p-value from the allele
frequency analysis was negative log-transformed and used as the continuous variable to identify
GO terms enriched among the most significant SNPs using a Mann Whitney U test followed by
Benjamini-Hochberg false discovery rate correction.
309
Ocean Basin Comparison
As reads from our experiment were mapped against the Persian Gulf *P. daedalea*
reference (SRA accession: SRP066627, Howells et al., 2016), we were able to compare tags
containing significant SNPs identified in this study to the tags containing significant SNPs
identified by Kirk and colleagues in their larval heat stress study (Kirk et al., 2018). This was
done using dplyr and tidyverse software packages in R to extract all matching significant tags in
this study and match them to the subset of tags containing significant SNPs from Kirk and
colleagues' study (Wickham et al., 2019, 2020).
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319 **Results**
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Survival in heat stress experiments

A Cox proportional hazards (CPH) model revealed significant effects of heat on larval
survival (p-value < 0.0001) (Supplemental Figure. 2). Individual factors in the model,
temperature and cross, as well as the interaction were significant (p-value < 0.0001). Endpoint
survival probabilities ranged more among crosses at 35.5˚C (0.405 to 0.750) than at 27˚C (0.861
to 0.975) (Figure. 2). Cross DB had the greatest survival probability at 35.5˚C (0.75+/- 0.04),
while cross CA had the lowest survival probability (0.41 +/- 0.04) (Figure. 2). Coral E provided
sperm in both EC and EF which both rank among the top four crosses, with (0.68+/- 0.04) and
(0.64+/- 0.04) probability of survival, respectively. Coral C provided eggs in crosses DC and BC
and provided sperm in cross CA, and this coral is associated with the bottom three survival
probabilities 0.48 (+/- 0.04), 0.44 (+/- 0.05), and 0.40 (+/- 0.04).
CPH does not allow for testing of random effects. Since position of the net wells were
randomly assigned, position could contribute to variation in thermal tolerance and thus we
applied a generalized linear mixed model (GLMM) to test for position effects. We found that
position did not contribute significantly to variation. Please see supplemental GLMM text for
more details.
337
338   *Quantitative genetic analysis of variation in heat tolerance*
339   Variation in survival during heat stress was highly heritable, with additive genetic
340   variation explaining 66% of variation in survival (i.e. $h^2 = 0.66$; Table 1). We also estimated h^2
341   using the survival probabilities from the Kaplan Meier analysis and recovered a similar estimate
342   ($h^2 = 0.64$). Both metrics reflect the total mortality over the entire time course (Table 1).
343
Genomic analysis of markers under selection during heat stress
Comparative analysis of larvae surviving an acute heat stress exposure versus paired
control larval pools for each cross revealed that experimental treatment altered allele frequencies
at hundreds of loci. Logistic regression of multilocus SNP genotypes (Figure. 3) uncovered tens-
to-hundreds of alleles in each cross whose frequencies were significantly altered by heat stress
(Table 2). Overall, there was a general pattern of increase in major allele frequency in most
crosses except for crosses BC, DC, and FA (Figure. 4). This also meant that minor alleles were
decreasing in frequency in most crosses. The number of alleles significantly associated with
survival under heat stress differed widely among crosses; however, the number of significant
alleles was not correlated with the number of tags genotyped in each sample ($P = 0.48$, $R^2 = -$
0.26), genetic distance between the parents ($P = 0.46$, $R^2 = -0.27$), or the heat tolerance of each
cross ($P = 0.48$, $R^2 = -0.27$). The variation was also not associated with a particular parent. For
example, parents C and D each produced crosses with few SNPs under selection (BC and DB),
but the combination of these parents in cross DC had the most, with 336 SNPs associated with
survival under heat stress (Table 2).
To evaluate the generalizability of genetic associations we searched for overlap in
significant tags among crosses. A proportion of tags were consistently associated with heat stress
across multiple crosses. In total, 148 significant tags were found in more than one cross (Table 3
and Supplemental Table 2). A pairwise comparison of the percent of shared significant tags
between a focal pair of crosses, revealed that 1-6% of significant tags were found to be
significant in another cross (Figure. 5, Table 3, and Supplemental Table 2). The same two tags
were found in seven crosses, a single tag overlapped in six crosses, and the same four tags
overlapped in five crosses (Table 3). The major allele in six different tags that were found to be
significantly changing in frequency in four or more crosses was found to increase in frequency

(Table 3). Five tags showed mixed change in frequency for the major allele, in some crosses it
increased and in others it decreased (Table 3). Only one tag showed consistent decrease in
frequency of the major allele across crosses in which it was identified (Table 3). However, most
tags associated with heat tolerance in one cross were not associated with heat tolerance in
unrelated crosses. For example, cross DC shared no parents with crosses EF or FA and 82 - 92%
of the tags identified in each cross were not associated with heat stress in the other two.
To determine whether this lack of overlap was a result of the necessarily stringent
filtering required for the analysis of pool-seq data (Kofler et al., 2011; Zhu et al., 2012), we
investigated overlap in tags among crosses at earlier stages of our analysis. After initial quality
filtering, we found only 30 – 63% of tags were genotyped in another cross (median of 52%,
Supplemental Table 3). This range did not appreciably change after the 80X coverage threshold
was applied, with 21 - 66% of tags identified in other crosses (median of 49%, Supplemental
Table 4), indicating that the coverage threshold was not driving this pattern. However, statistical
thresholds did have a moderate impact, as the percentage of significant tags in one focal cross
not identified in other pairwise comparisons ranged from 11 - 61% (median of 42%,
Supplemental Table 5).
We identified whether each significant tag was either not genotyped or was genotyped
and found to be insignificant. We found that a majority of significant tags (64%) were genotyped
in at least five out of ten crosses (Supplemental Table 7). Of the tags that were found to be
significant in at least four or more crosses, they were found to be insignificant in at least one and
in most cases more than one other cross (Table 4, Supplemental Table 6).
389
Functional significance of enriched tags
Seven GO terms were significantly enriched among the nine annotated tags that changed
in allele frequency in response to heat stress in any cross (Figure 6). Tags in genes regulating cell
projection, receptor signaling, regulation of apoptosis, regulation of hydrolase, and negative
regulation of metabolism were over-represented among the most significantly differentially
represented tags under heat stress conditions (Table 5). For the subset of significant tags shared
among crosses, 9 markers could be unambiguously assigned to a position in the transcriptome all
of which were annotated (Table 5). Furthermore, two of these tags which were identified in three
crosses were annotated as genes previously implicated in heat tolerance Monocarboxylate
transporter 10 (K1PRR1) and Isocitrate dehydrogenase subunit 1 (E9C6G6) (Table 5) (Drury et
al., 2021; Kenkel et al., 2013; Pathmanathan et al., 2021; Polato et al., 2010). In two tags that
mapped to transcripts and overlapped in other crosses, we found consistent decrease in the major
allele for the Monocarboxylate transporter and consistent increase of the major allele for the
isocitrate dehydrogenase subunit (Table 5).
404
Ocean basin comparison
Finally, we compared significant SNPs arising from any cross with previously identified
SNPs enriched in response to heat stress in Persian Gulf *P. daedalea* (Kirk et al., 2018). We
found two tags significantly associated with both survival and heat tolerance in our dataset
overlapped with sites identified in (Kirk et al., 2018) (Table 6). The specific SNPs identified in
each tag differed among studies (Table 5) and the tags did not match any genes in the reference
transcriptome, so the potential functional significance of this shared variation remains unknown.
412
413   **Discussion**

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Adaptive potential of Platygyra daedalea
Our analysis demonstrates substantial variation in heat tolerance is evident among larval
crosses. Moreover, 66% of this variation was attributable to additive genetic effects, indicating
significant adaptive potential. Coral parent E was generally associated with crosses that had high
survival probabilities and Coral parent C was associated with crosses that had lower survival
probabilities (Figure 2). Our results add to only four published narrow-sense heritability
estimates for heat tolerance in corals, all of which find this trait is highly heritable ranging from
0.58 – 0.87 (Bairos-Novak et al., 2021; Dixon et al., 2015; Dziedzic et al., 2019; Kirk et al.,
2018; Quigley et al., 2020). Heritability estimates from different populations can yield valuable
information about selection in a given geographic region as estimates are dependent on the
environment and the standing genetic variation in the population for which the estimate is made
(Falconer & Mackay, 1996). Heritability estimates will be low if all parents express the same
phenotype and genetic variation contributing to a trait is low (Falconer & Mackay, 1996).
However, they will be high if there is phenotypic variation and genetic variation contributing to
the trait is also high (Falconer and Mackay, 1996). A novel comparison is the contrast of the
heritability estimate for the same phenotype, larval survival in acute heat stress (0.48) found by
Kirk and colleagues for *P. daedalea* in the Persian Gulf (Kirk et al., 2018). Persian Gulf *P.*
daedalea have survived in a very shallow and warm sea for thousands of years (Purkis & Riegl,
2012), which may mean that a degree of selection for heat tolerance has already taken place
resulting in less genetic diversity in the extant population (Falconer & Mackay, 1996). The lower
heritability value is evidence to support this and the comparatively higher estimate of heritability
for this trait in the central GBR *P. daedalea* population suggests that until very recently, heat
stress may not have been the predominant selective pressure on this population, consistent with
historical bleaching datasets for this region (Hughes et al., 2018).
Responses to selection depend on genetic variation in traits (Falconer & Mackay, 1996;
Lynch & Walsh, 1998). Consequently, our study contributes to understanding the quantitative
genetic framework necessary to model heat tolerance under various selection regimes. In its
simplest form, this can be expressed in the breeder's equation, which describes the response to
selection as the product of the selection differential (S) and heritability of the trait under
selection. Selection differentials have not been widely estimated in corals (Kenkel et al., 2015;
Weeriyanun et al., 2021), but considering the widespread reductions in coral populations
following coral bleaching (Bellwood et al., 2004; Bruno & Selig, 2007; De'ath et al., 2012;
Hoegh-Guldberg et al., 2007) they are likely high for sensitive species. For truncation selection,
where all individuals with trait values above a threshold contribute offspring to the next
generation, selection differentials are equal to the difference in mean trait values between the
selected population and the original population (Falconer & Mackay, 1996; Lynch & Walsh,
1998) expressed as standard deviations of the population mean. If the selection differential
equals one standard deviation, all parents with traits one standard deviation above the population
mean contribute to phenotypic variation in the next generation. Based on the distribution of
cumulative survival probabilities in heat stress from our experiment (0.57 ± 0.18) and the
heritability of variation in heat tolerance we estimated from these data (0.66), the Breeder's
Equation predicts a response to selection of 0.12. We would expect that if larvae of the next
generation after selection were subjected to the same heat stress treatment, the average mortality
would be 0.48 (0.57 - 0.12), a 20% reduction in mortality under acute heat stress in a single
generation. If instead of the high h^2 estimated in our study (0.66), we modeled the responses to

selection using a minimum h^2 value of 0.1, we would predict only a 3% reduction in mortality in
the following generation. While the estimates above are underpinned by many assumptions, the
comparison highlights how empirical estimations of narrow-sense heritability can be used to
predict adaptive responses. The relevance of these estimates for population responses depends on
the extent to which larval survival during heat stress is directly under selection in natural
populations, or alternatively the effectiveness of larval heat tolerance as a proxy for heat
tolerance in adult colonies. Considering the lack of work in multi-generational comparisons, this
would be a fruitful direction of future study.
468
Allele frequency analysis supports multilocus adaptation
Physiological and fitness traits can be determined by single or few loci of large effect, but
more commonly they are underpinned by many loci of small effect (Sella & Barton, 2019).
Surveying a larger portion of the population and limiting our detection of allele frequency to
within the offspring of two known parents has helped to detect tens-to-hundreds of alleles
supporting survival during heat stress. Comparing between crosses allowed us to investigate the
generality of these patterns among different genetic backgrounds. This comparison revealed a
number of markers with consistent changes in allele frequencies in multiple crosses (Table 3 and
Supplemental Table 2). The tags that show allele frequency change in a consistent direction are
important targets for future studies of markers of thermal tolerance. Those that show inconsistent
direction of frequency change could be further evidence of the polygenicity of this trait. One
allele might be important for thermal tolerance response, but if the right combination of alleles
are not found at other sites in the genome thermal tolerance will not be as high.
However, many of the associations detected in each cross were not shared in comparisons
of multiple crosses. Our investigation into the lack of overlap among markers revealed that 49%
of markers in a given cross on average were not genotyped in other crosses even before the
application of the 80X within cross coverage filter (Supplementary tables 3-5), suggesting that
this loss is not a result of bioinformatic filtering applied but instead may result from aspects of
the sequencing methodology and the biology of the animals involved. Null alleles are one
common explanation for missing markers during sequencing (e.g. Crooks et al., 2013).
Alternatively, this bias could also result from the pooled sequencing approach. There were 100
individuals in each of eight replicate pools. Although amplification of libraries was kept to a
minimum (no more than 17 cycles of PCR) a certain amount of PCR bias is unavoidable and in a
given cross certain regions may have been amplified more frequently (Dohm et al., 2008). A
small number of additional sites were lost through the filtering process, but stringent thresholds
are required to reduce false positive allele calls given the small size of our RAD tags (Kofler et
al., 2011). In this study, a pooled sequencing approach was chosen to maximize the resources
available to survey as much of the population as possible. However, this is a clear tradeoff of
pooled sequencing with this number of individuals per pool and sequencing relatively short
fragments of the genome. Sequencing individual larvae may help alleviate this issue (e.g. Dixon
et al. 2015).
Statistical thresholds also had an impact and the observation that certain SNPs were
found to be significant in some crosses but not others could also result from the underlying
distribution of those alleles in our parental population. In isolated populations the loci
contributing to a trait are predicted to be fewer and of larger effect size (Savolainen et al., 2013).
However, in populations experiencing immigration, the locus effect size and the number of loci
that contribute to a trait become more variable (Savolainen et al., 2013). Corals can disperse in

their larval phase before metamorphosis into a sedentary juvenile, so it is entirely possible that
individuals originating from disparate reefs were among those chosen for our parental population
(Davies et al., 2015). The observation that many markers associated with heat stress were
specific to individual crosses suggests that many alleles of small effect contribute to variation in
heat tolerance and that many different combinations of these small effect alleles can affect heat
tolerance as is the growing consensus for thermal tolerance in coral (Bay & Palumbi, 2014;
Fuller et al., 2020; Kirk et al., 2018; Thomas & Palumbi, 2017). The fact that many of our
significant tags were found to be insignificant in other crosses also supports this (Table 4,
Supplemental Table 7). These differences may also result in part from epistatic interactions
where the effects of variation at these loci depend on the interaction of two or more genes and
composition of alleles at these genes which is dependent on genetic background (Day et al.,
2008). The large number of alleles changing in frequency in cross DC could result from an
unfortunate combination of parental alleles at many loci producing larvae with multiple
deleterious alleles. This could be the result of additive genetic effects, but it could also be an
indicator of epistasis, which is harder to detect but found frequently in polygenic traits in model
organisms (Soyk et al., 2020). Nevertheless, repeated observation of some of these associations
across different genetic backgrounds provides strong evidence that these markers are linked to
genetic factors contributing to variation in heat tolerance.
Future studies can use the alleles that increased in frequency and the genomic regions
identified in this and other studies to develop biomarkers to identify resilient coral populations
for assisted gene flow or use in selective breeding programs. Starting with genetic markers
associated with habitat variation in previous studies (Capper et al., 2015), we can test these
associations through a combination of laboratory experiments and extensive population surveys
of allele frequencies (Jin et al., 2016). These studies revealed that the frequencies of alleles
associated with enhanced tolerance for environmental stress range from 0.48 in some populations
to 0.92 in others (Jin et al., 2016). Further studies are needed to extend these surveys across
species, markers, and geographic regions to develop an ecosystem-scale perspective on the
potential for coral adaptive responses to ocean warming.
534
SNPs associated with heat tolerance occur in genes related to extracellular matrix and
metabolism
Among the annotated SNPs exhibiting significant changes in allele frequency in response
to experimental selection, a number of functional enrichments were identified in genes
associated with the organization of the cellular matrix and cellular components, the regulation of
apoptosis, regulation of hydrolase, and metabolism. In addition, we found a subset of tags
mapped to genes previously implicated in heat tolerance, two of which were observed in multiple
crosses (Table 5). These genes are prime candidates for further functional studies aimed at
delineating the potential mechanistic role of these polymorphisms (Cleves et al., n.d., 2018,
2020).
Four markers were associated with genes that play roles in the extracellular matrix, with
enrichments in cell projection organization, surface receptor signaling, and cell component
organization (Table 5, Figure. 6). GPI Mannosyltransferase 2 functions in the assembly of
glycosylphosphatidylinositol (GPI) anchors that attach proteins to cell membranes (Ji et al.,
2005). In corals this protein is in the pathway that changes the mannose content of the
extracellular matrix (Lee et al., 2016). The abundance of Gammaproteobacteria is reduced with
the change in mannose in coral mucus (Lee et al., 2016) Metalloendopeptidase, a metal-

dependent protease, and Zinc finger MYM-type protein both restructure and organize the
extracellular matrix (Lodish et al., 2020). The Rab9 effector with kelch motifs controls docking
of endosomes in the process of exocytosis (Díaz et al., 1997). While metalloendopeptidase and
this zinc finger protein were not found in other coral studies, they are found in heat stress in
other invertebrates (Juárez et al., 2021; Prieto et al., 2019; Zhang et al., 2015). On the other hand,
Rab proteins have been identified as important for signaling the start of phagocytosis
mechanisms and cell destruction in anemones during oxidative stress and bleaching (Chen et al.,
2005; Downs et al., 2009). In this case it is likely the Rab 9 effector is involved in signaling the
start of the cell death during heat stress. SNPs in these genes indicate that control of cell
reorganization and structure is important during heat stress.
Negative regulation of macromolecule metabolic process was also identified as a
significant GO category (Figure 6). Three coral crosses had a SNP in the Krebs cycle enzyme
isocitrate dehydrogenase (IDH) and a monocarboxylate transporter both of which are associated
with this annotation. These proteins have been identified in gene expression studies as important
to the coral thermal stress response (Drury et al., 2021; Kenkel et al., 2013; Pathmanathan et al.,
2021; Polato et al., 2010). IDH participates in reducing reactive oxygen species and is one of the
44 proteins identified as the minimal stress proteome (Kultz, 2005). The monocarboxylate
transporter moves lactate, pyruvate, and ketone bodies across plasma membranes, thus it is
associated with the regulation of metabolism (Halestrap, 2012). This transporter has been
identified as important in recent coral heat stress studies and is upregulated under heat stress
(Drury et al., 2021; Pathmanathan et al., 2021). The SNPs in IDH and the monocarboxylate
transporter associated with heat tolerance in this study further strengthens their proposed role in
the thermal stress response in corals.
Additional genes annotated with GO categories like metabolism, regulation of hydrolase,
and regulation of apoptosis, may also be important for handling reactive oxygen species. While
the 3oxoacyl (acyl-carrier-protein) reductase has a larger role in fatty acid biosynthesis, it also
catalyzes a reduction reaction (Chan & Vogel, 2010). Similarly, IDH is responsible for
regulation of metabolism, but the reaction it catalyzes is also a reduction. These genes have been
identified as important for controlling reactive oxygen species levels resulting from increased
metabolism during heat stress (Dixon et al., 2015) and our results add further evidence of the
possible importance of these SNPs to corals' heat stress response.
The consistent change in the frequency of the major alleles in the monocarboxylate
transporter and isocitrate dehydrogenase across multiple crosses also suggests that these alleles
are important for heat stress survival. They are prime targets for future studies assessing markers
of thermal tolerance.
587
Ocean Basin Comparison
Comparisons of populations of *P. daedalea* in more typical reef temperature conditions
(this study) and unusually warm conditions (Kirk et al., 2018) will be important to understand
the genomic basis of heat tolerance. Two markers from the present study were significantly
associated with heat tolerance in the Persian Gulf population (Table 6). Neither of these markers
matched genes in the transcriptome and without a genomic resource it is not possible to identify
their location in the genome or proximal genes. However, their repeated identification strongly
supports an important role in coral thermal tolerance. Further work could include screening for
differential abundance of these alleles among natural thermal gradients and potential validation
for gene function through knock-down from CRISPR-Cas9 induced mutagenesis (e.g., Cleves et

al., 2018). The lack of shared markers between these cross-ocean basin populations may be
further evidence that many alleles of small effect contribute to heat tolerance and that they vary
in their distribution across populations.
601
Conclusions
Our findings demonstrate substantial genetic variation and high heritability of heat
tolerance (i.e., > 60%) in the coral *Platygyra daedalea* from the GBR supporting the potential for
adaptation. Our analysis identified hundreds of markers associated with heat tolerance, including
markers that were repeatedly associated with heat tolerance across different genetic backgrounds.
This supports the conclusion that heat tolerance is a complex polygenic trait determined by many
alleles of small effect that are likely to be species and population specific. Despite this
complexity, we nevertheless found markers shared in multiple crosses that may be important for
heat tolerance in *P. daedalea.* Two markers were also identified in Persian Gulf populations, and
a number occurred in genes previously implicated in coral thermal tolerance. Our findings build
on the growing body of evidence that natural populations of corals harbor substantial genetic
variation in heat tolerance that can support adaptive responses to selection and identify genetic
markers for heat tolerance in a globally distributed coral species. These heat tolerance markers
are essential to any type of reseeding or assisted gene flow and assisted migration programs,
which may be necessary to maintaining reefs in the future as we work to reduce CO_2 emissions
and mitigate the effects of climate change.
618

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Code and Data Availability
Sequence Access: NCBI, Accession: PRJNA542930
https://github.com/hollandelder/Platygyradaedalea
_HeritabiltyandMarkersofThermalTolerance
https://github.com/Eli-Meyer/sequence_processing
https://github.com/z0on/GO_MWU)
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636 **Authors' contributions**
637 E.M., H.E., and L.B. conceived the experiments. H.E., V.M., J.M., and L.B. conducted
638 experiments, L.B. and A.B. provided advice and logistical support through the project, H.E.
639 conducted genetic and statistical analysis and wrote the manuscript. L.B. and E.M. helped with
640 genetic and statistical analysis and L.B., V.W. and A. B. helped write the manuscript.
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References

Bairos-Novak, K. R., Hoogenboom, M. O., van Oppen, M. J. H., & Connolly, S. R. (2021). Coral adaptation to climate change: Meta-analysis reveals high heritability across multiple traits. In *Global Change Biology* (Vol. 27, Issue 22, pp. 5694–5710). John Wiley and Sons Inc. https://doi.org/10.1111/gcb.15829

Barshis, D. J., Ladner, J. T., Oliver, T. a, Seneca, F. O., Traylor-Knowles, N., & Palumbi, S. R. (2013). Genomic basis for coral resilience to climate change. *Proceedings of the National Academy of Sciences of the United States of America*, *110*(4), 1387–1392. https://doi.org/10.1073/pnas.1210224110

Baums, I. B., Baker, A. C., Davies, S. W., Grottoli, A. G., Kenkel, C. D., Kitchen, S. A., Kuffner, I. B., LaJeunesse, T. C., Matz, M. v., Miller, M. W., Parkinson, J. E., & Shantz, A. A. (2019). Considerations for maximizing the adaptive potential of restored coral populations in the western Atlantic. *Ecological Applications*, *29*(8), 1–23. https://doi.org/10.1002/eap.1978

Bay, L. K., Doyle, J., Logan, M., & Berkelmans, R. (2016). Recovery from bleaching is mediated by threshold densities of background thermo-tolerant symbiont types in a reef-building coral. *Royal Society Open Science*, *3*(160322).

Bay, R. a, & Palumbi, S. R. (2014). Report Multilocus Adaptation Associated with Heat Resistance in Reef-Building Corals. *Current Biology*, 1–5. https://doi.org/10.1016/j.cub.2014.10.044

Bay, R., Rose, N., Logan, C., & Palumbi, S. (2017). Genomic models predict successful coral adaptation if future ocean warming rates are reduced. *Science Advances*, 1–10. https://doi.org/10.1126/sciadv.1701413

Bellwood, D. R., Hughes, T. P., Folke, C., & Nystro, M. (2004). Confronting the coral reef crisis. *Nature*, *429*(June), 827–833.

Berkelmans, R. (2002). Time-integrated thermal bleaching thresholds of reefs and their variation on the Great Barrier Reef. *Marine Ecology Progress Series*, *229*, 73–82.

Berkelmans, R., & van Oppen, M. J. H. (2006). The role of zooxanthellae in the thermal tolerance of corals: a "nugget of hope" for coral reefs in an era of climate change. *Proceedings. Biological Sciences / The Royal Society*, *273*(1599), 2305–2312. https://doi.org/10.1098/rspb.2006.3567

Bruno, J. F., & Selig, E. R. (2007). Regional Decline of Coral Cover in the Indo-Pacific : Timing , Extent , and Subregional Comparisons. *PloS One*, *8*. https://doi.org/10.1371/journal.pone.0000711

Capper, R. L., Jin, Y. K., Lundgren, P. B., Peplow, L. M., Matz, M. v, & Oppen, M. J. H. van. (2015). Quantitative high resolution melting : two methods to determine SNP allele frequencies from pooled samples. *BMC Genetics*, *16*(62), 1–13. https://doi.org/10.1186/s12863-015-0222-z

Chan, D. I., & Vogel, H. J. (2010). Current understanding of fatty acid biosynthesis and the acyl carrier protein. In *Biochemical Journal* (Vol. 430, Issue 1, pp. 1–19). https://doi.org/10.1042/BJ20100462

Charmantier, A., & Garant, D. (2005). Environmental quality and evolutionary potential : lessons from wild populations. *Proceedings of the Royal Society*, *272*, 1415–1425. https://doi.org/10.1098/rspb.2005.3117

Chen, M. C., Hong, M. C., Huang, Y. sen, Liu, M. C., Cheng, Y. M., & Fang, L. S. (2005). ApRab11, a cnidarian homologue of the recycling regulatory protein Rab11, is involved in

the establishment and maintenance of the Aiptasia-Symbiodinium endosymbiosis.
Biochemical and Biophysical Research Communications, *338*(3), 1607–1616.
https://doi.org/10.1016/j.bbrc.2005.10.133
Cleves, P. A., Shumaker, A., Lee, J. M., Putnam, H. M., & Bhattacharya, D. (2020). Unknown to
Known: Advancing Knowledge of Coral Gene Function. In *Trends in Genetics* (Vol. 36,
Issue 2, pp. 93–104). Elsevier Ltd. https://doi.org/10.1016/j.tig.2019.11.001
Cleves, P. A., Strader, M. E., Bay, L. K., Pringle, J. R., & Matz, M. v. (2018). CRISPR/Cas9-
mediated genome editing in a reef building coral. *PNAS*, *115*(20), 5235–5240.
https://doi.org/10.1073/pnas.1722151115
Cleves, P. A., Tinoco, A. I., Bradford, J., Perrin, D., Bay, L. K., & Pringle, J. R. (n.d.). *Reduced*
thermal tolerance in a coral carrying CRISPR-induced mutations in the gene for a heat-
shock transcription factor. https://doi.org/10.1073/pnas.1920779117/-/DCSupplemental
Coles, S. L., & Riegl, B. M. (2013). Thermal tolerances of reef corals in the Gulf : A review of
the potential for increasing coral survival and adaptation to climate change through assisted
translocation. *Marine Pollution Bulletin*, *72*(2), 323–332.
https://doi.org/10.1016/j.marpolbul.2012.09.006
Core Team, R. (2017). *R: A Language and Environment for Statistical Computing*.
https://www.r-project.org/
Cox, D. R. (2007). Regression Models and Life-Tables. *Journal of the Royal StatisticalSociety.*
Series B (Methodological), *34*(2), 187–220.
D'Angelo, C., Hume, B. C. C., Burt, J., Smith, E. G., Achterberg, E. P., & Wiedenmann, J.
(2015). Local adaptation constrains the distribution potential of heat-tolerant Symbiodinium
from the Persian/Arabian Gulf. *ISME Journal*, *9*(12), 2551–2560.
https://doi.org/10.1038/ismej.2015.80
Davies, S. W., Treml, E. A., Kenkel, C. D., & Matz, M. v. (2015). Exploring the role of
Micronesian islands in the maintenance of coral genetic diversity in the Pacific Ocean.
Molecular Ecology, *24*(1), 70–82. https://doi.org/10.1111/mec.13005
Day, T., Nagel, L., van Oppen, M. J. H., & Caley, M. J. (2008). Factors affecting the evolution
of bleaching resistance in corals. *American Naturalist*, *171*(2).
https://doi.org/10.1086/524956
De'ath, G., Fabricius, K. E., Sweatman, H., & Puotinen, M. (2012). The 27 – year decline of
coral cover on the Great Barrier Reef and its causes. *PNAS*, *109*(44), 17995–17999.
https://doi.org/10.1073/pnas.1208909109
Díaz, E., Schimmöller, F., & Pfeffer, S. R. (1997). A Novel Rab9 Effector Required for
Endosome-to-TGN Transport. In *The Journal of Cell Biology* (Vol. 138, Issue 2).
Dixon, G., Davies, S., Aglyamova, G., Meyer, E., Bay, L. K., & Matz, M. v. (2015). Genomic
Determinants of coral heat tolerance across latitudes. *Science*, *348*(6242), 1460–1462.
Dohm, J. C., Lottaz, C., Borodina, T., & Himmelbauer, H. (2008). Substantial biases in ultra-
short read data sets from high-throughput DNA sequencing. *Nucleic Acids Research*,
36(16). https://doi.org/10.1093/nar/gkn425
Downs, C. A., Kramarsky-Winter, E., Martinez, J., Kushmaro, A., Woodley, C. M., Loya, Y., &
Ostrander, G. K. (2009). Symbiophagy as a cellular mechanism for coral bleaching.
Autophagy, *5*(2), 211–216. https://doi.org/10.4161/auto.5.2.7405
Drury, C., Bean, N., Harris, C., Hancock, J., Hucekba, J., Martin, C., Roach, T., Quinn, R., &
Gates, R. D. (2021). Intrapopulation adaptive varience supports selective breeding in a reef
building coral. *BioArXIV*, 1–19.

Dziedzic, K. E., Elder, H., & Meyer, E. (2019). Heritable variation in bleaching responses and its
functional genomic basis in reef-building corals (Orbicella faveolata). *Molecular Ecology*,
1–16.
Falconer, D. S., & Mackay, T. F. C. (1996). *Introduction to Quantitative Genetics*.
Fuller, Z. L., Mocellin, V. J. L., Morris, L. A., Cantin, N., Shepherd, J., Sarre, L., Peng, J., Liao,
Y., Pickrell, J., Andolfatto, P., Matz, M., Bay, L. K., & Przeworski, M. (2020). Population
genetics of the coroal Acropora millepora: Toward genomic prediction of bleaching.
Science, *369*(6501). https://doi.org/10.1126/science.aba4674
Gienapp, P., Lof, M., Reed, T. E., McNamara, J., Verhulst, S., & Visser, M. E. (2013).
Predicting demographically sustainable rates of adaptation: Can great tit breeding time keep
pace with climate change? *Philosophical Transactions of the Royal Society B: Biological
Sciences*, *368*(1610). https://doi.org/10.1098/rstb.2012.0289
Glynn, P. W. (1993). Coral reef bleaching: ecological perspectives. *Coral Reefs*, *12*, 1–17.
https://doi.org/10.1007/BF00303779
Halestrap, A. P. (2012). The monocarboxylate transporter family-Structure and functional
characterization. In *IUBMB Life* (Vol. 64, Issue 1, pp. 1–9). https://doi.org/10.1002/iub.573
Hoegh-Guldberg, O. (1999). Climate change, coral bleaching and the future of the world's coral
reefs. *Marine and Freshwater Research*, *50*(8), 839–866. https://doi.org/10.1071/MF99078
Hoegh-Guldberg, O., Harvell, C. D., Sale, P. F., Edwards, A. J., Caldeira, K., Knowlton, N.,
Eakin, C. M., Muthiga, N., Bradbury, R. H., Dubi, A., Hatziolos, M. E., Mumby, P. J.,
Hooten, A., Steneck, R. S., Greenfield, P., Gomez, E., & Iglesias-Prieto, R. (2007). Coral
Reefs Under Rapid Climate Change and Ocean Acidification. *Science*, *318*(5857), 1737–
1742.
Howells, E. J., Abrego, D., Meyer, E., Kirk, N., & Burt, J. A. (2016). Host adaptation and
unexpected symbiont partners enable reef-building corals to tolerate extreme temperatures.
Global Change Biology, 1–13. https://doi.org/10.1111/gcb.13250
Hughes, T., Kerry, J., Baird, A., Connolly, S., Chase, T., Dietzel, A., Hill, T., Hoey, A.,
Hoogenboom, M., Jacobson, M., Kerswell, A., Madin, J., Mieog, A., Paley, A., Pratchett,
M., Torda, G., & Woods, R. (2019). Global warming impairs stock - recruitment dynamics
of corals. *Nature*, *568*, 387–390. https://doi.org/10.1038/s41586-019-1081-y
Hughes, T. P., Kerry, J. T., Álvarez-noriega, M., Álvarez-romero, J. G., Anderson, K. D., Baird,
A. H., Babcock, R. C., Beger, M., Bellwood, D. R., Berkelmans, R., Bridge, T. C., Butler, I.
R., Byrne, M., Cantin, N. E., Comeau, S., Connolly, S. R., Cumming, G. S., Dalton, S. J.,
Diaz-Pulido, G., … Wilson, S. K. (2017). Global warming and recurrent mass bleaching of
corals. *Nature*, *543*, 373–377. https://doi.org/10.1038/nature21707
Hughes, T. P., Kerry, J. T., Baird, A. H., Connolly, S. R., Dietzel, A., Eakin, C. M., Heron, S. F.,
Hoey, A. S., Hoogenboom, M. O., Liu, G., Mcwilliam, M. J., Pears, R. J., Pratchett, M. S.,
Skirving, W. J., Stella, J. S., & Torda, G. (2018). Global warming transforms coral reef
assemblages. *Nature*, *556*, 492–496.
Hutchings, P., Kingsford, M., & Hoegh-Guldberg, O. (2019). *The Great Barrier Reef: Biology,
Environment, and Management*. Csiro Publishing.
Ji, Y. K., Hong, Y., Ashida, H., Shishioh, N., Murakami, Y., Morita, Y. S., Maeda, Y., &
Kinoshita, T. (2005). PIG-V involved in transferring the second mannose in
glycosylphosphatidylinositol. *Journal of Biological Chemistry*, *280*(10), 9489–9497.
https://doi.org/10.1074/jbc.M413867200

Jin, Y. K., Lundgren, P., Lutz, A., Raina, J., Howells, E. J., Paley, A. S., Willis, B. L., & Oppen,
M. J. H. van. (2016). *Genetic markers for antioxidant capacity in a reef-building coral*.
May.
Juárez, O. E., Escobedo-Fregoso, C., Arredondo-Espinoza, R., & Ibarra, A. M. (2021).
Development of SNP markers for identification of thermo-resistant families of the Pacific
oyster Crassostrea gigas based on RNA-seq. *Aquaculture*, *539*.
https://doi.org/10.1016/j.aquaculture.2021.736618
Kaplan, E. L., & Meier, P. (2017). Nonparametric Estimation from Incomplete Observations.
Journal of the American Statistical Association, *53*(282), 457–481.
Kassambara, A. (2020). *ggpubr R package: Create Publication Ready Plots*.
https://rpkgs.datanovia.com/ggpubr/
Kenkel, C. D., Matz, M. v, & Tx, A. (2016). *Enhanced gene expression plasticity as a*
mechanism of adaptation to a variable environment in a reef-building coral. *3*, 1–21.
Kenkel, C. D., Meyer, E., & Matz, M. v. (2013). Gene expression under chronic heat stress in
populations of the mustard hill coral (Porites astreoides) from different thermal
environments. *Molecular Ecology*, *22*(16), 4322–4334. https://doi.org/10.1111/mec.12390
Kenkel, C. D., Setta, S. P., & Matz, M. v. (2015). Heritable differences in fitness-related traits
among populations of the mustard hill coral, Porites astreoides. *Heredity*, *115*(6), 509–516.
https://doi.org/10.1038/hdy.2015.52
Kirk, N., Howells, E., Abrego, D., Burt, J., & Meyer, E. (2018). Genomic and transcriptomic
signals of thermal tolerance in heat-tolerant corals (Platygyra daedalea) of the
Arabian/Persian Gulf. *Molecular Ecology*, *27*, 5180–5194.
Kofler, R., Orozco-terWengel, P., de Maio, N., Pandey, R. V., Nolte, V., Futschik, A., Kosiol,
C., & Schlötterer, C. (2011). Popoolation: A toolbox for population genetic analysis of next
generation sequencing data from pooled individuals. *PLoS ONE*, *6*(1).
https://doi.org/10.1371/journal.pone.0015925
Kofler, R., Pandey, R. V., Schlötterer, C., Populationsgenetik, I., Vienna, V., & Wien, A.-.
(2017). PoPoolation2 : identifying differentiation between populations using sequencing of
pooled DNA samples (Pool-Seq). *Bioinformatics*, *27*(24), 3435–3436.
https://doi.org/10.1093/bioinformatics/btr589
Kultz, D. (2005). Molecular And Evolutionary Basis of the Cellular Stress Response. *Annual*
Review Physiology, *67*, 225–257.
https://doi.org/10.1146/annurev.physiol.67.040403.103635
Lajeunesse, T. C., Parkinson, J. E., Gabrielson, P. W., Jeong, H. J., Reimer, J. D., Voolstra, C.
R., Santos, S. R., Lajeunesse, T. C., Parkinson, J. E., Gabrielson, P. W., Jeong, H. J., &
Reimer, J. D. (2018). Systematic Revision of Symbiodiniaceae Highlights the Antiquity and
Diversity of Coral Endosymbionts Article Systematic Revision of Symbiodiniaceae
Highlights the Antiquity and Diversity of Coral Endosymbionts. *Current Biology*, *28*(16),
2570-2580.e6. https://doi.org/10.1016/j.cub.2018.07.008
Langmead, B., Trapnell, C., Pop, M., & Salzberg, S. L. (2009). Ultrafast and memory-efficient
alignment of short DNA sequences to the human genome. *Genome Biology*, *10*(R25).
Lee, S. T. M., Davy, S. K., Tang, S. L., & Kench, P. S. (2016). Mucus sugar content shapes the
bacterial community structure in thermally stressed Acropora muricata. *Frontiers in*
Microbiology, *7*(MAR). https://doi.org/10.3389/fmicb.2016.00371
Lodish, H., Berk, A., Kaiser, C. A., Krieger, M., Bretscher, A., Ploegh, H., Martin, K. C., Yaffe,
M. B., & Amon, A. (2020). *Molecular Cell Biology* (9th ed.). Freeman.

Long Term Reef Monitoring Program. (2019). *Annual Report on Reef Condition: Mixed bill of*
health for the Great Barrier Reef. https://www.aims.gov.au/reef-monitoring/gbr-condition-
summary-2018-2019

Lynch, M., & Walsh, B. (1998). *Genetics and Analysis of Quantitative Traits*.

Manzello, D. P., Jankulak, M., Matz, M. v, Enochs, I. C., Valentino, L., Carlton, R. D.,
Kolodziej, G., Serrano, X., & Towle, E. K. (2019). Role of host genetics and heat - tolerant
algal symbionts in sustaining populations of the endangered coral Orbicella faveolata in the
Florida Keys with ocean warming. *Global Change Biology*, *25*, 1016–1031.
https://doi.org/10.1111/gcb.14545

Matz, M. v., Treml, E. A., & Haller, B. C. (2020). Estimating the potential for coral adaptation to
global warming across the Indo-West Pacific. *Global Change Biology*, *26*(6), 3473–3481.
https://doi.org/10.1111/gcb.15060

Meyer, E., Davies, S., Wang, S., Willis, B., Abrego, D., Juenger, T., & Matz, M. (2009). Genetic
variation in responses to a settlement cue and elevated temperature in the reef-building coral
Acropora millepora. *Marine Ecology Progress Series*, *392*, 81–92.
https://doi.org/10.3354/meps08208

Meyer, K. (2007). WOMBAT — A tool for mixed model analyses in quantitative genetics by
restricted maximum likelihood (REML). *Journal of Zhejiang University Science B*, *8*(11),
815–821. https://doi.org/10.1631/jzus.2007.B0815

Oliver, T. A., & Palumbi, S. R. (2011). Many corals host thermally resistant symbionts in high-
temperature habitat. *Coral Reefs*, *30*, 241–250. https://doi.org/10.1007/s00338-010-0696-0

Pathmanathan, J. S., Williams, A., Stephens, T. G., Su, X., Eric, N., Conetta, D., Putnam, H. M.,
& Bhattacharya, D. (2021). Multi-omic characterization of the thermal stress phenome in
the stony coral Montipora capitata. *BioArXIV*, 1–25.

Polato, N. R., Voolstra, C. R., Schnetzer, J., DeSalvo, M. K., Randall, C. J., Szmant, A. M.,
Medina, M., & Baums, I. B. (2010). Location-specific responses to thermal stress in larvae
of the reef-building coral Montastraea faveolata. *PloS One*, *5*(6), e11221.
https://doi.org/10.1371/journal.pone.0011221

Prieto, D., Markaide, P., Urrutxurtu, I., Navarro, E., Artigaud, S., Fleury, E., Ibarrola, I., &
Urrutia, M. B. (2019). Gill transcriptomic analysis in fast- and slow-growing individuals of
Mytilus galloprovincialis. *Aquaculture*, *511*.
https://doi.org/10.1016/j.aquaculture.2019.734242

Purkis, S. J., & Riegl, B. M. (2012). Geomorphology and Reef Building in the SE Gulf. In *Coral*
Reefs of the Gulf: Adaptation to Climatic Extremes in the World's Hottest Sea (pp. 33–50).
https://doi.org/10.1007/978-94-007-3008-3

Quigley, K. M., Bay, L. K., & van Oppen, M. J. H. (2020). Genome-wide SNP analysis reveals
an increase in adaptive genetic variation through selective breeding of coral. *Molecular*
Ecology, *May*. https://doi.org/10.1111/mec.15482

Riegl, B., Purkis, S., Al-cibahy, A., Abdel-moati, M., & Hoegh-Guldberg, O. (2011). Present
Limits to Heat-Adaptability in Corals and Population-Level Responses to Climate
Extremes. *PloS One*, *6*(9). https://doi.org/10.1371/journal.pone.0024802

Ruiz-jones, L. J., & Palumbi, S. R. (2017). Tidal heat pulses on a reef trigger a fine-tuned
transcriptional response in corals to maintain homeostasis. *Science Advances*, *3*(e1601298),
1–11.

Rumble, S. M., Lacroute, P., Dalca, A. v., Fiume, M., Sidow, A., & Brudno, M. (2009). SHRiMP: Accurate mapping of short color-space reads. *PLoS Computational Biology*, *5*(5), 1–11. https://doi.org/10.1371/journal.pcbi.1000386

Savolainen, O., Lascoux, M., & Merilä, J. (2013). Ecological genomics of local adaptation. *Nature Reviews Genetics*, *14*(November), 807–820. https://doi.org/10.1038/nrg3522

Sella, G., & Barton, N. H. (2019). Thinking about the Evolution of Complex Traits in the Era of Genome-Wide Association Studies. *Annual Review of Genomics and Human Genetics*, *20*, 461–493. https://doi.org/10.1146/annurev-genom-083115-022316

Silverstein, R. N., Cunning, R., & Baker, A. C. (2014). Change in algal symbiont communities after bleaching, not prior heat exposure, increases heat tolerance of reef corals. *Global Change Biology*, 236–249. https://doi.org/10.1111/gcb.12706

Singh, D., Singh, C. K., Sewak, R., Tomar, S., & Pal, M. (2017). Genetics and Molecular Mapping of Heat Tolerance for Seedling Survival and Pod Set in Lentil. *Crop Science*, *57*, 3059–3067. https://doi.org/10.2135/cropsci2017.05.0284

Souter, D., Planes, S., Wicquart, J., Logan, M., Obura, D., & Staub, F. (2020). *Status of Coral Reefs of the World: 2020 Chapter 2. Status of Coral Reefs of the World*.

Soyk, S., Benoit, M., & Lippman, Z. B. (2020). *New Horizons for Dissecting Epistasis in Crop Quantitative Trait Variation*. https://doi.org/10.1146/annurev-genet-050720

Therneau, T. (2015). *A Package for Survival Analysis in S. verson 2.38* (p. https://CRAN.R-project.org/package=survival).

Thomas, L., & Palumbi, S. R. (2017). The genomics of recovery from coral bleaching. *Proceedings of the Royal Society B: Biological Sciences*, *284*(1865). https://doi.org/10.1098/rspb.2017.1790

Traylor-knowles, N., Rose, N. H., Sheets, E. A., & Palumbi, S. R. (2017). Early Transcriptional Responses during Heat Stress in the Coral Acropora hyacinthus. *Biological Bulletin*, *232*, 91–100.

Treangen, T. J., & Salzberg, S. L. (2012). Repetitive DNA and next-generation sequencing: Computational challenges and solutions. *Nature Reviews Genetics*, *13*(1), 36–46. https://doi.org/10.1038/nrg3117

van Oppen, M. J. H., Oliver, J. K., Putnam, H. M., & Gates, R. D. (2015). Building coral reef resilience through assisted evolution. *Proceedings of the National Academy of Sciences*, *112*(8), 2307–2313. https://doi.org/10.1073/pnas.1422301112

Venuprasad, R., Dalid, C. O., del Valle, M., Zhoa, D., Espiritu, M., Sta Cruz, M. T., Amante, M., Kumar, A., & Atlin, G. N. (2009). Identification and characterization of large-effect quantitative trait loci for grain yield under lowland drought stress in rice using bulk-segregant analysis. *Theoretical Applied Genetics*, *120*, 177–190. https://doi.org/10.1007/s00122-009-1168-1

Vikram, P., Swamy, B. P. M., Dixit, S., Ahmed, H., Cruz, M. T. S., Singh, A. K., Ye, G., & Kumar, A. (2012). Field Crops Research Bulk segregant analysis : " An effective approach for mapping consistent-effect drought grain yield QTLs in rice ." *Field Crops Research*, *134*, 185–192. https://doi.org/10.1016/j.fcr.2012.05.012

Waite, T. A., & Campbell, L. G. (2006). Controlling the false discovery rate and increasing statistical power in ecological studies. *Ecoscience*, *13*(4), 439–442. https://doi.org/10.2980/1195-6860(2006)13[439:CTFDRA]2.0.CO;2

Wang, M., Liu, W., Jiang, B., & Peng, Q. (2019). Genetic Analysis and Related Gene Primary Mapping of Heat Stress Tolerance in Cucumber Using Bulked Segregant Analysis. *Horticulture Science*, *54*(3), 423–428. https://doi.org/10.21273/HORTSCI13734-18

Wang, S., Meyer, E., McKay, J., & Matz, M. (2012). 2b-RAD: a simple and flexible method for genome-wide genotyping. *Nature Methods*, *9*(August), 808–810. http://www.nature.com/nmeth/journal/v9/n8/abs/nmeth.2023.html

Weeriyanun, P., Collins, R. B., Macadam, A., Kiff, H., Randle, J. L., & Quigley, K. M. (2021). Predicting selection-response gradients of heat tolerance in a wide-ranging reef-building coral. *BioRxiv*. https://doi.org/10.1101/2021.10.06.463349

Wellenreuther, M., & Hansson, B. (2016). Detecting Polygenic Evolution: Problems, Pitfalls, and Promises. In *Trends in Genetics* (Vol. 32, Issue 3, pp. 155–164). Elsevier Ltd. https://doi.org/10.1016/j.tig.2015.12.004

Wickham, H., Averick, M., Byan, J., Chang, W., McGowan, L. D., Francois, R., Grolemund, G., Hayes, A., Henry, L., Hester, J., Kuhn, M., Pedersen, T. L., Miller, E., Bache, S. M., Muller, K., Ooms, J., Robinson, D., Seidel, D. P., Spinu, V., … Yutani, H. (2019). Welcome to the Tidyverse. *The Journal of Open Source Software*, *4*(43), 1–6.

Wickham, H., Francois, R., Henry, L., & Muller, K. (2020). *dplyr: A Grammar of Data Manipulation* (R package version 0.8.4). https://cran.r-project.org/package=dplyr

Wilson, A. J., Réale, D., Clements, M. N., Morrissey, M. M., Postma, E., Walling, C. A., Kruuk, L. E. B., & Nussey, D. H. (2010). An ecologist's guide to the animal model. *Journal of Animal Ecology*, *79*(1), 13–26. https://doi.org/10.1111/j.1365-2656.2009.01639.x

Woolsey, E. S., Keith, S. a., Byrne, M., Schmidt-Roach, S., & Baird, a. H. (2014). Latitudinal variation in thermal tolerance thresholds of early life stages of corals. *Coral Reefs*, *34*, 471–478. https://doi.org/10.1007/s00338-014-1253-z

Wright, R. M., Aglyamova, G. v., Meyer, E., & Matz, M. v. (2015). Gene expression associated with white syndromes in a reef building coral, Acropora hyacinthus. *BMC Genomics*, *16*(1), 1–12. https://doi.org/10.1186/s12864-015-1540-2

Zhang, Y., Sun, J., Mu, H., Li, J., Zhang, Y., Xu, F., Xiang, Z., Qian, P. Y., Qiu, J. W., & Yu, Z. (2015). Proteomic basis of stress responses in the gills of the pacific oyster Crassostrea gigas. *Journal of Proteome Research*, *14*(1), 304–317. https://doi.org/10.1021/pr500940s

Zhu, Y., Bergland, A. O., González, J., & Petrov, D. A. (2012). Empirical validation of pooled whole genome population re-sequencing in Drosophila melanogaster. *PLoS ONE*, *7*(7), 1–7. https://doi.org/10.1371/journal.pone.0041901

Tables

Table 1. Estimates of narrow-sense heritability of thermal tolerance obtained from different survival metrics. Censored and uncensored KM estimates refer to cumulative probability of survival estimated with or without censoring individuals surviving at the experimental endpoint.

Data	*Survival*	h^2
Survival at 240hr	59%	0.66
KM Estimate	59%	0.64

950
951

Table 2. The number of loci at which allele frequencies were
significantly altered in survivors of thermal stress in each cross.

Cross	Significant markers	Markers genotyped	Parental distance	Survival (10 d)
AB	76	15,217	0.247	0.637
AF	34	19,399	0.245	0.551
BA	95	11,616	0.247	0.742
BC	69	22,820	0.257	0.442
CA	75	23,173	0.246	0.406
DB	44	23,999	0.256	0.758
DC	336	17,548	0.246	0.483
EC	77	16,734	0.262	0.683
EF	124	13,904	0.257	0.654
FA	139	16,691	0.245	0.639

Table 3. Tags identified as changing in frequency after heat stress in four or more crosses. The color of the cross indicates whether the allele in the tag was increasing in frequency (blue) or decreasing in (red).

Marker	SNP Location	Marker Found in Crosses	Number Crosses
denovoLocus34739	1, 21	AB, AF, BC, CA, DC, EF, FA	7
denovoLocus44191	16, 17	AB, AF, BA, CA, DB, EF, FA	7
denovoLocus9359	1	AB, BC, CA, DC, EF, FA	6
denovoLocus121909	11	AB, BA, CA, EC, FA	5
denovoLocus1291	4, 6	AB, BA, BC, DB, EF	5
denovoLocus22543	1, 25	BA, CA, EC, EF, FA	5
denovoLocus93322	3, 25	AB, AF, DC, EC, EF	5
denovoLocus13062	26, 36	CA, EC, EF, FA	4
denovoLocus18723	1	BA, DB, DC, EC	4
denovoLocus29413	1	BC, DC, EF, FA	4
denovoLocus36225	1, 4	AB, AF, EC, EF	4
denovoLocus51123	4	AB, AF, DC, EF	4

995 **Table 4**. The number of crosses a tag was identified as either significant, insignificant,
996 or not genotyped for tags that were found to be significantly changing in frequency in
997 four or more crosses.

Tag	Number Crosses Significant	Number Crosses Insignificant	Number Crosses Not Genotyped
denovoLocus34739	7	1	2
denovoLocus44191	7	0	3
denovoLocus9359	6	3	1
denovoLocus121909	5	4	1
denovoLocus22543	5	3	2
denovoLocus93322	5	3	2
denovoLocus1291	5	1	4
denovoLocus18723	4	5	1
denovoLocus13062	4	4	2
denovoLocus36225	4	4	2
denovoLocus29413	4	3	3
denovoLocus51123	4	1	5

Table 5. Tags mapped to transcripts that changed in frequency after heat stress. Listed in order of the number of crosses in which the tag was found. Color of the cross indicates the direction in the change of frequency: red represents a decrease and blue represents an increase.

Crosses	Number of Tags Mapped to Transcript	Transcripts	Descriptions (Annotation)
AB, BA, EF	1	c39807	Monocarboxylate transporter 10 (K1PRR1)
AF, BA, CA	1	c15457	Isocitrate dehydrogenase subunit 1 (E9C6G6)
AB	1	c36043	Metalloendopeptidase (A7RVF7)
AB	1	c9426	60S_ribosomal_protein_L6_(Fragment)
BC	1	c10517	3-oxoacyl-[acyl-carrier-protein] reductase (K1QNK4)
DC	1	c7510	Putative dna polymerase type b (A0A069DZN0)
DC	1	c11010	Zinc finger MYM-type protein 3 (K1QD06)
FA	1	c16893	Rab9_effector_protein_with_kelch_motifs
FA	1	c12147	GPI mannosyltransferase 2 (K1PUQ8)

1021 Table 6. Tags shared between the Persian Gulf and GBR populations that are significant to
1022 thermal tolerance.

Cross	Tag	SNP Base Pair GBR	SNP Base Pair Persian Gulf	Major Allele GBR	Major Allele Persian Gulf
CA	denovoLocus12624	35	28	T	GT
DC	denovoLocus60661	16	21	T	C

Figures

Figure 1. Crossing design and heat stress experiments. (A) Six *Platygyra daedalea* colonies were sampled from Esk Reef on the GBR. (B) Crossing design used to produce ten crosses among those parental colonies. (C) Representative micrograph (10X) of planula larvae used for these experiments (53 hours post-fertilization).

Figure 2. Heat tolerance differs among crosses. Kaplan Meier plots of cumulative probability of survival for each family in control 27.5°C (blue) and elevated 35.5°C (red) temperature. Panels are arranged in order of heat tolerance, based on survival probability at 240 hr. Shaded regions represent 95% confidence intervals.

1042
1043 Figure 3. Heat stress alters allele frequencies in surviving larvae. Panels show correlations
1044 between the average allele frequencies in survivors of replicate heat stress treatments relative to
1045 the corresponding controls. Markers with significantly altered allele frequencies (logistic
1046 regression, FDR < 0.05) are highlighted in red.
1047

1048
1049 Figure 4. The proportion of major alleles that changed in frequency after heat stress. Crosses are
1050 indicated on the x-axis and the proportions of major alleles that increased (blue) and decreased
1051 (red) are depicted on the y-axis.
1052

1053
1054 Figure 5. Heatmap of the percent of significant tags shared between pairs of crosses. The percent
1055 of shared significant tags for each cross was calculated from the number of significant tags
1056 shared between the pair and the total number of potential tags that could have been shared. These
1057 comparisons demonstrate independent associations between heat stress and genotype at these
1058 loci across different genetic backgrounds.
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1061   Figure 6. Gene Ontology Tree. Gene ontology categories enriched in significant markers. The
1062   size of the font indicates the significance of the category per the inset key. Markers were mapped
1063   to the *Platygyra daedalea* transcriptome and the matching transcripts were assigned the log
1064   adjusted p-value from the allele frequency logistic regression.
1065