Worldwide reef surveys highlight different adaptive responses to heat between coral taxa

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Abstract

Coral reefs around the world are under threat due to widespread decline of hard corals caused by anomalous heat waves. Coral taxa of different morphologies are known to have different sensitivities to heat stress. However, little is known about how long-term effects of heat exposure differ between coral taxa, nor in particular how such effects might drive adaptive processes.

Here, we combined worldwide reef survey data with remotely sensed thermal anomalies to evaluate how local rates of taxa-specific coral cover were associated with heat stress. We found a negative association between coral cover and heat stress, where associations were weaker for corals with boulder-like morphology (massive corals), intermediate for corals with arborescent morphology (branching corals) and stronger for corals with encrusting or laminar morphology (plating corals). Additionally, we found that the negative association between branching coral cover and recent exposure to heat stress (measured as the year before a survey) was mitigated by the effect of long-term heat stress (measured since 1985), suggesting that adaptive processes occurred. In contrast, this "mitigating effect" of past heat-stress was not observed for either massive or plating corals.

We hypothesize that these different responses to recent and past heat stress mirror different
life history traits of coral taxa. For instance, branching corals are known to show higher
sensitivities to heat stress and faster growth rates, two characteristics that might boost
adaptive rates via evolutionary processes, when compared with other taxa.
The results of this work suggest that all coral taxa, regardless of their morphology, will be
"losers" of coral bleaching, and that conservation strategies should be adjusted to match taxa-
specific responses to heat stress.

Keywords

41 Coral reef – coral bleaching – global warming – heat stress – adaptation – remote sensing –
42 life history traits

Introduction

Over the past 30 years, coral reefs around the world have been suffering from a widespread
decline of hard corals (Ateweberhan, McClanahan, Graham, & Sheppard, 2011; Cramer et al.,
2020; De'ath, Fabricius, Sweatman, & Puotinen, 2012). This decline threatens the persistence
of the entire coral reef ecosystem, as hard corals constitute the physical architecture of these
habitats (Pratchett, Thompson, Hoey, Cowman, & Wilson, 2018). A leading cause of this
decline is coral bleaching, a stress response driven by heat stress that causes corals to
separate from the symbiotic algae living in their tissues (van Oppen & Lough, 2009). When
persisting over several weeks, a bleaching state will eventually lead to starvation of coral
(Diaz-Pulido & McCook, 2002). By now, extensive reef deterioration following anomalous
heat waves have been observed at a global extent, with reports of local loss of coral cover
(i.e. the percentage of the reef surface covered by live stony coral) up to 50% (Hughes,
Anderson, et al., 2018). By 2050, it is expected that coral bleaching conditions will become
persistent worldwide (Van Hooidonk, Maynard, & Planes, 2013).

The sensitivity to bleaching varies between coral species (Loya et al., 2001). Indeed, corals

with an arborescent morphology (hereafter referred to as branching corals) are considered

to be more sensitive to heat stress than corals with a laminar or an encrusting morphology

(hereafter referred to as plating corals) and corals with a boulder-like morphology (hereafter

referred to as massive corals; van Woesik, Sakai, Ganase, & Loya, 2011). Part of these

differences in thermal susceptibility mirror differences in morphological features that

modulate resistance to bleaching and thermal stress, such as the thickness of the tissues and

the size of the corallites (the skeletal cups containing individual polyps; Conti-Jerpe et al.,

2020; Mydlarz, McGinty, & Harvell, 2010; Robert van Woesik et al., 2012). In contrast, the

long-term effects of heat stress exposure are more difficult to assess based on coral

morphology alone (van Woesik et al., 2011). This is because differences in life history traits

between coral species (such as growth rates and fecundity) are likely to drive different rates

of recovery of coral cover after a heat stress event (Darling, Alvarez-Filip, Oliver, McClanahan,

& Côté, 2012; Osborne et al., 2017).

Substantial differences in heat tolerance can also be observed between corals from the same

species (Bay & Palumbi, 2014; Schoepf, Stat, Falter, & McCulloch, 2015). Such tolerance can

be the result of adaptive processes driven by previous exposure to thermal stress (Hughes et

al., 2019). These adaptive processes can occur at two levels: (1) at the physiological level

(acclimatization), when a colony temporarily adjusts its metabolism to cope with the stressful

condition; and (2) at the genetic level (evolutionary adaptation), when the repeated exposure

to stress selects the individuals of a population that carry the genetic traits conferring

increased tolerance (Palumbi, Barshis, Traylor-Knowles, & Bay, 2014). Understanding how

these adaptive processes contribute to increased heat tolerance in different coral species is

of paramount importance in order to organize effective conservation efforts (Matz, Treml, &

Haller, 2020; Schoepf et al., 2019).

Over the last decade, the combination of field survey records with remote sensed data for

sea surface temperature has become one of the main tools used to measure coral decline

driven by heat stress. Multiple studies have shown associations between heat stress events

and coral cover decline both at local (Babcock et al., 2020; Head et al., 2019; Romero-Torres

et al., 2020; Selmoni, Lecellier, Vigliola, Berteaux-Lecellier, & Joost, 2020; Soto, Muller Karger,

Hallock, & Hu, 2011) and global scales (Selig, Casey, & Bruno, 2012). In other studies, the

metric of bleaching intensity was found to be positively associated with heat stress on the

Australian Great Barrier Reef (Hughes et al., 2017, 2019; Hughes, Kerry, et al., 2018), as well
as across larger spatial scales (for instance, the Indo-Pacific region; Hughes, Anderson, et al.,
2018; McClanahan, Darling, et al., 2020; McClanahan et al., 2019; McClanahan, Maina, et al.,
2020; Sully, Burkepile, Donovan, Hodgson, & van Woesik, 2019). Notably, some of these
studies investigated how past and recent thermal stress interact to drive coral cover
decline/bleaching intensity, suggesting the existence of coral adaptation but without being
able to distinguish acclimatization from evolutionary adaptation (Head et al., 2019; Hughes
et al., 2019; McClanahan, Maina, et al., 2020; Selmoni, Lecellier, et al., 2020; Sully et al., 2019).
Very few of these works focus on the differences in heat stress responses between coral
species (Hughes, Kerry, et al., 2018; McClanahan, Darling, et al., 2020), and thus very little
information was collected to assess differences in heat stress adaptive potential between
coral species.

In this study, we combine global coral cover data obtained from the Catlin Seaview Survey
project (González-Rivero et al., 2016, 2014) with time series of sea surface temperature
anomalies computed from remote sensing data by the Coral Reef Watch (Skirving et al., 2020).
The goal here is to investigate worldwide how coral decline and adaptation may be driven by
heat stress in coral taxa with different types of morphologies. We first evaluate how the
degree of heat stress experienced by reefs over the past 20 years is associated with local
differences in coral cover. These differences concern (1) overall coral cover rates, (2) a
diversity index derived from taxa-specific coral cover rates and (3) coral cover rates of three
groups of coral taxa based on morphology: branching, plating and massive. Next, we assess
how such rates of coral cover and diversity driven by recent heat stress (measured across the
year before the survey) are modulated (i.e. accentuated or mitigated) either by short-term
heat stress (measured across the two previous years) or long-term heat stress exposure
(measured across more than 20 previous years). We hypothesize taxa-specific responses to
heat stress mirror the differences in heat tolerance and the potential for evolutionary
adaptation determined by distinct life history traits.

Material and methods

Coral cover data

The coral cover data were obtained from the Catlin Seaview Survey (CSS) project (González-Rivero et al., 2016, 2014). When we accessed the data in January 2021, there were a total of 860 surveys performed along 579 transects (380 transects visited one time, 140 two times, 36 three times, and 23 four times) across twelve study areas around the world between 2012 and 2018 (Fig. 1a). The CSS project applied the same standardized framework to each survey when recording and analyzing field data. First, field surveys were performed using an underwater 360° camera that took a picture every three seconds along a 1.6-2 km transect, maintaining a constant depth of 10 meters. Next, field pictures were processed using automated image recognition based on machine learning algorithms. Such algorithms were trained using labels that were manually annotated by coral taxonomists (Beijbom, Edmunds, Kline, Mitchell, & Kriegman, 2012).

For each reef survey, the CSS data provided (1) the overall measure of hard coral cover rates and (2) the coral cover rates for 78 groups of corals labelled using visual characteristics (for instance "*Acropora* corymbose/tabular/plate", "*Porites*" or "*Pocilloporidae*"). Since coral species differ between regions, only part of these 78 labels were observed in each of the different study areas. We therefore manually created broader groups of corals based on morphology to facilitate comparability between study areas. This was done in three steps. First, we retrieved the CSS metadata showing the list of species for each of the 78 coral labels. Second, we consulted the Coral Trait Database (Madin et al., 2016) to determine morphology (according to Veron, 2000) of the species in each label group. Third, we established three broader groups of corals based on morphology: branching (including coral morphologies "branching", "digitate", "corymbose" and "tabular"), plating ("encrusting" and "laminar") and massive ("massive" and "sub-massive"). For each of these broad groups of coral morphologies (hereafter referred to as coral taxa), we computed the average hard coral cover for each survey using a custom script (see "Data and code availability" section) in the R environment (v. 3.5.1; R Core Team, 2021).

Diversity index

The CSS coral cover data of the 78 coral labels were used to compute the Simpson's diversity

index (Simpson, 1949) of each survey. For each study area, we first excluded the coral labels

that were not observed in any of the surveys (*i.e.,* average coral cover equal to 0). Next, the

coral cover of the retained labels were used to compute the Simpson diversity index for each

survey using the "diversity" function of the VEGAN R package (v. 2.5; Oksanen et al., 2019).

Note that VEGAN returns the diversity index as 1-(Simpson's diversity), so that higher values

(i.e. towards 1) correspond to higher diversity.

Figure 1. The study regions. a) Distribution of the twelve study regions. b) For each study region, the map shows the average Degree Heating Week (DHW) values for the 1985-2020 period. The blue squares indicate areas were coral cover surveys from the Catlin Seaview Surveys project were performed, with the numbers corresponding to the number of surveys per area. The reef surfaces were derived from the "Global Distribution of Coral Reefs" dataset of the UNEP (UNEP-WCMC, WorldFish-Center, WRI, & TNC, 2021)

 Heat stress data

The heat stress data were retrieved from the Coral Reef Watch database, as part of the 5km-

resolution sea surface temperature products (Skirving et al., 2020). The variable we used to

describe heat stress was the Degree Heating Week (DHW, Fig. 1b), as it has been shown to be

directly correlated with coral bleaching occurrence and severity (Kayanne, 2017; G. Liu et al.,

2014; McClanahan et al., 2019). For a given day, DHW is calculated as the sum of the

temperature hotspots (that is, temperatures that exceeded by 1°C the maximal monthly

average) from the preceding 12-weeks period. We used the CSS metadata to retrieve the

coordinates of every survey location (at the mid-point of the transects), and extracted the

daily DHW from 1985 to 2020 for each coordinate using the RASTER R package (v. 3.0; Hijmans,

2016). We discarded 82 surveys for which DHW values could not be extracted due to

coordinates located on pixels adjacent to the coastline. In total, we extracted daily DHW

values for 778 surveys.

For each reef survey, we computed five different metrics of average DHW (Fig. 2). The first

metric represents recent heat stress ($\overline{DHW}_{1yr}$) and is calculated as the average DHW during

the year that preceded the survey date (Fig. 2a). Similarly, we computed an estimation of past

heat stress as the average DHW across the 10 years ($\overline{DHW}_{10yrs}$; Fig. 2b) and the 20 years

($\overline{DHW}_{20yrs}$; Fig. 2c) that preceded the survey. Additionally, we computed two metrics of past

heat stress which excluded the first year before the sampling date. The first of these metrics

covers the previous two years ($\overline{DHW}_{1-3yrs}$; Fig. 2d), while the second covers all of the

previous years ($\overline{DHW}_{>1yr}$; Fig. 2e). These two variables were developed as estimators of past

heat stress that do not overlap with recent heat stress ($\overline{DHW}_{1yr}$), so that they could be used

in bi-variate model construction (see the "Statistical analysis" section).

Figure 2. Degree heating week (DHW) time series averages. The figure displays the differences in the calculation of the five DHW averages used in this study. The green arrows indicate a hypothetical date for the coral cover survey, and the red boxes show the periods of interest for the calculation of average DHW with the different methods.

Statistical analysis

Previous studies have employed repeated measures of coral cover over time to determine

the trends of coral growth or decline, and then used statistical analyses to match these trends

with sea surface temperature data (Babcock et al., 2020; Head et al., 2019; Selig et al., 2012;

Soto et al., 2011). This approach is not transposable to the CSS dataset, as most of the

transects were surveyed only once. For this reason, we employed a different method to

investigate how local differences in coral cover rates were associated with historical patterns

of heat stress measured at the survey sites.

In practice, we investigated the association between coral cover rates and heat stress using

generalized linear mixed-models (GLMMs), where the random factors controlled for spatial

autocorrelation between survey sites (Dormann et al., 2007). We employed four levels of

random factors representing spatial autocorrelation at different spatial scales and

progressively nested into each other (*i.e.* the factors covering smaller spatial scales were

nested into those covering larger spatial scales). The first two random factors were:

1) the oceanic regions as defined by the CSS metadata (five levels: Australia, Southeast Asia,

Pacific, Atlantic and Indian Ocean);

2) the study areas (twelve levels, Fig. 1a);

We designed additional random factors to control for spatial autocorrelation at a local scale

(within each study area). This was done by applying the following clustering approach in the

R environment. First, we computed the Euclidean distances between surveys sites, and

applied a hierarchical clustering using the Ward distance method. The result was a tree of

distances between survey sites that was used to compute the two additional random factors

(3 and 4):

3) survey sites located up to 100 km from each other;

4) survey sites located up to 7 km from each other.

The two threshold values (100 km and 7 km) used in the clustering procedure were chosen as

they corresponded to the average value and the minimal value, respectively, of the mean

Euclidean distances measured between survey sites within each study area.

GLMMs were built using the GLMMTMB R package (v. 1.0; Brooks et al., 2017). We investigated

five response variables recorded at each survey site: overall coral cover (Fig. 3a), Simpson's

Diversity (Fig. 3b), coral cover of branching corals (Fig. 3c), coral cover of plating corals

(Fig. 3d) and coral cover of massive corals (Fig. 3e). As the values of these variables are rates

between 0 and 1, we used the beta regression method to model their response (Ferrari &

Cribari-Neto, 2004). The five explanatory variables were the DHW averages (Fig. 2) described

in the "Heat stress data" section: $\overline{DHW}_{1yr}$ (Fig. 3f), $\overline{DHW}_{10yrs}$ (Fig. 3g), $\overline{DHW}_{20yrs}$ (Fig. 3h),

$\overline{DHW}_{1-3yrs}$ (Fig. 3i), and $\overline{DHW}_{>1yr}$ (Fig. 3l).

We ran the analysis for two types of GLMMs: univariate and bi-variate (Table 1). In univariate

GLMMs, we investigated the association of each of the five response variables with either

$\overline{DHW}_{1yr}$, $\overline{DHW}_{10yrs}$ or $\overline{DHW}_{20yrs}$. The goal was to evaluate how local patterns of coral cover

and diversity can be explained by recent and past heat stress.

For each of the five response variables, two bi-variate GLMMs were computed. The first

investigated the association of the response variable with (1) $\overline{DHW}_{1yr}$ and (2) the interaction

between $\overline{DHW}_{1yr}$ and $\overline{DHW}_{1-3yrs}$, while the second model employed (1) $\overline{DHW}_{1yr}$ and (2)

the interaction between $\overline{DHW}_{1yr}$ and $\overline{DHW}_{>1yr}$ as explanatory variables. The goal of these

bi-variate models was to evaluate how past heat stress (measured on the short-term for the

first model, and on the long-term for the second) modulates (i.e. accentuates or mitigates)

the effect of recent heat stress on coral cover and diversity.

Table 1. List of the models computed. The table displays the formulae of the univariate and bi-variate models computed for every response variable (overall coral cover: $CC_{overall}$, Simpson's diversity: SD, branching coral cover: $CC_{branching}$, plating coral cover: $CC_{plating}$, massive coral cover: $CC_{massive}$). The explanatory variables used in univariate models are the averages of degree heating weak ($\overline{DHW}$) measured during the year ($\overline{DHW}_{1yr}$), the ten years ($\overline{DHW}_{10yrs}$) and the twenty years ($\overline{DHW}_{20yrs}$) before the survey. In bi-variate models, the explanatory variables are (1) $\overline{DHW}_{1yr}$ and (2) the interaction between $\overline{DHW}_{1yr}$ and $\overline{DHW}$ measured during either the two previous years ($\overline{DHW}_{1-3yrs}$) or all the previous years ($\overline{DHW}_{>1yr}$). α indicates the intercept, β the effects of the explanatory variables.

Response variable	Univariate models	Bi-variate models
Overall coral cover	1) $CC_{overall} \sim \alpha + \beta(\overline{DHW}_{1yr})$ 2) $CC_{overall} \sim \alpha + \beta(\overline{DHW}_{10yrs})$ 3) $CC_{overall} \sim \alpha + \beta(\overline{DHW}_{20yrs})$	4) $CC_{overall} \sim \alpha + \beta_1(\overline{DHW}_{1yr}) + \beta_2(\overline{DHW}_{1yr}:\overline{DHW}_{1-3yrs})$ 5) $CC_{overall} \sim \alpha + \beta_1(\overline{DHW}_{1yr}) + \beta_2(\overline{DHW}_{1yr}:\overline{DHW}_{>1yr})$
Simpson's diversity	6) $SD \sim \alpha + \beta(\overline{DHW}_{1yr})$ 7) $SD \sim \alpha + \beta(\overline{DHW}_{10yrs})$ 8) $SD \sim \alpha + \beta(\overline{DHW}_{20yrs})$	9) $SD \sim \alpha + \beta_1(\overline{DHW}_{1yr}) + \beta_2(\overline{DHW}_{1yr}:\overline{DHW}_{1-3yrs})$ 10) $SD \sim \alpha + \beta_1(\overline{DHW}_{1yr}) + \beta_2(\overline{DHW}_{1yr}:\overline{DHW}_{>1yr})$
Branching coral cover	11) $CC_{branching} \sim \alpha + \beta(\overline{DHW}_{1yr})$ 12) $CC_{branching} \sim \alpha + \beta(\overline{DHW}_{10yrs})$ 13) $CC_{branching} \sim \alpha + \beta(\overline{DHW}_{20yrs})$	14) $CC_{branching} \sim \alpha + \beta_1(\overline{DHW}_{1yr}) + \beta_2(\overline{DHW}_{1yr}:\overline{DHW}_{1-3yrs})$ 15) $CC_{branching} \sim \alpha + \beta_1(\overline{DHW}_{1yr}) + \beta_2(\overline{DHW}_{1yr}:\overline{DHW}_{>1yr})$
Plating coral cover	16) $CC_{plating} \sim \alpha + \beta(\overline{DHW}_{1yr})$ 17) $CC_{plating} \sim \alpha + \beta(\overline{DHW}_{10yrs})$ 18) $CC_{plating} \sim \alpha + \beta(\overline{DHW}_{20yrs})$	19) $CC_{plating} \sim \alpha + \beta_1(\overline{DHW}_{1yr}) + \beta_2(\overline{DHW}_{1yr}:\overline{DHW}_{1-3yrs})$ 20) $CC_{plating} \sim \alpha + \beta_1(\overline{DHW}_{1yr}) + \beta_2(\overline{DHW}_{1yr}:\overline{DHW}_{>1yr})$
Massive coral cover	21) $CC_{massive} \sim \alpha + \beta(\overline{DHW}_{1yr})$ 22) $CC_{massive} \sim \alpha + \beta(\overline{DHW}_{10yrs})$ 23) $CC_{massive} \sim \alpha + \beta(\overline{DHW}_{20yrs})$	24) $CC_{massive} \sim \alpha + \beta_1(\overline{DHW}_{1yr}) + \beta_2(\overline{DHW}_{1yr}:\overline{DHW}_{1-3yrs})$ 25) $CC_{massive} \sim \alpha + \beta_1(\overline{DHW}_{1yr}) + \beta_2(\overline{DHW}_{1yr}:\overline{DHW}_{>1yr})$

For each GLMM, we reported the estimate (β, with its standard deviation) and the p-value

from the Wald test for the fixed effects (Brooks et al., 2017). The visualization of the

regression lines was performed using the VISREG package (v. 2.7; Breheny & Burchett, 2017)

and custom R functions.

In addition, we evaluated the goodness-of-fit of each of the GLMMs by measuring the

difference between the model's Akaike Information Criterion (AIC; Bozdogan, 1987) and the

AIC of a null model. For univariate models, the corresponding null model was a GLMM using

a constant value as explanatory variable. For bi-variate models, the null model was the

univariate GLMMs using $\overline{DHW}_{1yr}$ as explanatory variable. This approach allowed us to

evaluate how accounting for the effect of past heat stress on recent heat stress (i.e. the

interaction term of bi-variate models) improves the goodness-of-fit, compared to a model

including recent heat stress only ($\overline{DHW}_{1yr}$). According to the rules of thumb for model

selection (Burnham & Anderson, 2004), a model with $dAIC < -2$ (where

$dAIC = AIC_{null\ model} - AIC_{model}$) has an improved goodness-of-fit, compared to the null

model.

Figure 3. Distribution of coral cover, diversity and thermal stress across the study areas.

Results

Univariate models

We first built univariate models to evaluate how coral cover and diversity from field surveys were associated with average heat stress ($\overline{DHW}$) calculated over different time periods (1 year, 10 years and 20 years) before the survey dates.

Overall coral cover and Simpson's diversity

All univariate association models between coral cover rates and heat stress showed a higher goodness-of-fit than the null model ($dAIC(\overline{DHW}_{1yr}) = -77.9$, $dAIC(\overline{DHW}_{10yrs}) = -45.1$, $dAIC(\overline{DHW}_{20yrs}) = -32.9$; Fig. 4a). All of these models showed significant negative associations ($p < 0.00001$), where a stronger effect was observed for DHW averages computed over longer periods ($\beta(\overline{DHW}_{1yr}) = -0.03 \pm 0.01$, $\beta(\overline{DHW}_{10yrs}) = -0.12 \pm 0.02$, $\beta(\overline{DHW}_{20yrs}) = -0.23 \pm 0.04$).

The same negative associations were observed between heat stress and Simpson's Diversity (Fig. 4b), though only the association with $\overline{DHW}_{1yr}$ was significant ($p = 0.01$, $\beta = -0.01 \pm 0.003$) and with a higher goodness-of-fit than the null model ($dAIC = -4.13$).

Taxa-specific coral cover

We then focused on taxa-specific rates of coral cover, and observed negative associations with heat stress for all the taxa (Fig. 5a-c). However, the magnitude of such negative effects differed between taxa. Overall, the largest magnitude of effects was observed for plating corals ($\beta(\overline{DHW}_{1yr}) = -0.01 \pm 0.001$, $\beta(\overline{DHW}_{10yrs}) = -0.04 \pm 0.01$, $\beta(\overline{DHW}_{20yrs}) = -0.09 \pm 0.1$; Fig. 5b), followed by branching corals $\beta(\overline{DHW}_{1yr}) = -0.01 \pm 0.002$, $\beta(\overline{DHW}_{10yrs}) = -0.03 \pm 0.01$, $\beta(\overline{DHW}_{20yrs}) = -0.05 \pm 0.1$; Fig. 5a), and finally by massive corals $\beta(\overline{DHW}_{1yr}) = -0.01 \pm 0.001$, $\beta(\overline{DHW}_{10yrs}) = -0.02 \pm 0.01$, $\beta(\overline{DHW}_{20yrs}) = -0.03 \pm 0.14$; Fig. 5c). All of the taxa-specific models were significant ($p < 0.02$), with a higher goodness-of-fit than the null models ($dAIC < -2$), with the exception of the association between $\overline{DHW}_{20yrs}$ and branching coral cover ($dAIC = -1.62$).

Bi-variate models

We employed bi-variate models to evaluate how the negative associations of both coral cover

and diversity with recent heat stress ($\overline{DHW}_{1yr}$) were modulated by heat stress averages

calculated either over the short-term (the previous 2 years, $\overline{DHW}_{1-3yrs}$) or the long-term (all

of the previous years since 1985, $\overline{DHW}_{>1yr}$; Fig. 4a-b, Fig. 5a-c).

Overall coral cover and Simpson's diversity

For coral cover (Fig. 4c, Fig. 4e) and Simpson's diversity (Fig. 4d, Fig. 4f), none of the bi-variate

models showed a higher goodness-of-fit ($dAIC > -2$) when compared with the null model

(i.e. the univariate model with $\overline{DHW}_{1yr}$).

Taxa-specific coral cover

As we focused on taxa-specific coral cover rates, we observed that the types of associations

differed between taxa. In branching corals, the bi-variate model with $\overline{DHW}_{1-3yrs}$ resulted in

a higher goodness-of-fit than the null model ($dAIC = -4.5$) and the effect of the interaction

term ($\overline{DHW}_{1yr}:\overline{DHW}_{1-3yrs}$) was significant (p $= 0.01$) and of positive sign (Fig. 5d). Indeed,

the negative effect of $\overline{DHW}_{1yr}$ on branching coral cover appeared to decrease in magnitude

with higher levels of $\overline{DHW}_{1-3yrs}$.

The same effect on branching coral cover was observed for the interaction term that included

$\overline{DHW}_{>1yr}$ (Fig. 5g). In this case, we observed a higher goodness-of-fit of the model ($dAIC =$

-28.6) and a larger magnitude of the effect of the interaction term (p < 0.00001). For

instance, the effect of $\overline{DHW}_{1yr}$ on branching coral cover was $\beta = -0.024 \pm 0.003$ for

$\overline{DHW}_{>1yr}$ between 0 and 0.1, decreased to $\beta = -0.016 \pm 0.002$ for $\overline{DHW}_{>1yr}$ between

0.2 and 0.3, and was close to zero ($\beta = +0.001 \pm 0$) for $\overline{DHW}_{>1yr}$ above 0.4.

For the plating coral taxon, we found that the bi-variate model with $\overline{DHW}_{>1yr}$ had a higher

goodness-of-fit than the null model ($dAIC = -8.8$, Fig. 5h). Remarkably, here the effect of

the interaction term was negative (p $= 0.001$), resulting in the negative association between

$\overline{DHW}_{1yr}$ and plating coral cover to increase its magnitude for higher levels of $\overline{DHW}_{>1yr}$.

In contrast, neither bi-variate models for massive coral cover resulted in increased goodness-

of-fit ($dAIC > -2$) when compared to the null model.

All the coefficients and statistics describing both the univariate and bi-variate association

models are displayed in the Supplementary Table 1.

Figure 4. Association between coral cover and diversity with heat stress. The plots display the association of two variables measured from reef surveys (a, c and e: overall coral cover; b, d, f: Simpson's index of coral taxa diversity) with different variables describing past heat stress. In a and b, the plots display the association of coral cover and diversity with the average degree heating week ($\overline{DHW}$) calculated during the year ($\overline{DHW}_{1yr}$, magenta), the 10 years ($\overline{DHW}_{10yrs}$, blue) and the 20 years ($\overline{DHW}_{20yrs}$, dark blue) that preceded the survey dates. In c and d, the plots show the association of coral cover and diversity with $\overline{DHW}_{1yr}$ across different ranges of average DHW during the two previous years ($\overline{DHW}_{1-3yrs}$). In e and f, the plots show the association of coral cover and diversity with $\overline{DHW}_{1yr}$ across different ranges of average DHW during all the previous years (i.e. since 1985; $\overline{DHW}_{>1yrs}$). For each association model, the plots display the goodness-of-fit compared to a null model ($dAIC$) and the p-values (p) of the fixed effects.

Figure 5. Taxa-specific association between coral cover and heat stress. The plots display the association of coral cover rates from reef surveys for three coral taxa (a, d, g: branching corals; b, e, h: plating corals; c, f, i: massive corals) with different variables describing past heat stress. In a, b and c, the plots display the association of coral cover with the average degree heating week ($\overline{DHW}$) calculated during the year ($\overline{DHW}_{1yr}$, magenta), the 10 years ($\overline{DHW}_{10yrs}$, blue) and the 20 years ($\overline{DHW}_{20yrs}$, dark blue) that preceded the survey dates. In d, e and f, the plots show the association of coral cover with $\overline{DHW}_{1yr}$ across different ranges of average DHW during the two previous years ($\overline{DHW}_{1-3yrs}$). In g, h and i, the plots show the association of coral cover with $\overline{DHW}_{1yr}$ across different ranges of average DHW during all the previous years (i.e. since 1985; $\overline{DHW}_{>1yrs}$). For each association model, the plots display the goodness-of-fit compared to a null model ($dAIC$) along with the p-values (p) of the fixed effects.

Discussion

The global decline of coral cover and diversity driven by heat stress

We observed that local rates of coral cover were negatively associated with past heat stress exposure (Fig. 4a). We hypothesize that such negative association describes a causal link, as these observations are consistent with previous studies in the field (Babcock et al., 2020; Head et al., 2019; Hughes, Kerry, et al., 2018; Selig et al., 2012; Soto et al., 2011).

Recent exposure (*i.e.* within one year) to intense heat stress is known to drive a significant decline of coral cover. This was previously shown during the mass coral bleaching event in 2016, which caused a 50% relative decline of coral cover (*i.e.* the ratio of coral cover after and before the event) in the north of the Australian Great Barrier Reef (Hughes, Kerry, et al., 2018) and a 29% decline in the Chagos Archipelago (Indian Ocean; Head et al., 2019). Average DHW values for these two regions in 2016 were ~2.3 and ~3.5°C-week, respectively, and represent heat waves of extreme magnitude. In the present study, recent heat stress was generally below 1°C-week (Fig. 3f), yet even this lower heat stress was significantly associated with coral decline. For instance, one year at an average DHW of 1°C-week was estimated to result in an absolute coral cover decline of ~3% (Fig. 4a).

The recurrence of heat stress events across larger temporal frames (10-20 years) can drive even more severe declines in coral cover. For example, a previous study surveying reefs from north-western Australia for over 20 years indicated that recurrent heat stress was associated with a relative decline of coral cover by 38-92% (Babcock et al., 2020). Our results corroborate this finding, and we estimate that a 20-year period with an average DHW of 0.5°C-week corresponds to an absolute coral cover decline of ~11% (Fig. 4a). To put this number in context, an average 0.5°C-week corresponds to the long-term heat observed for most of the reefs in the southern Caribbean and in some reefs in the northern Great Barrier Reef since 1985 (Fig. 1).

The diversity of coral taxa was found to be negatively associated with heat stress exposure, where such association was particularly noticeable for recent heat stress (Fig. 4b). We hypothesize that this decline in diversity is due to an asymmetrical sensitivity to heat stress between different coral taxa, as suggested by previous studies (Hughes, Kerry, et al., 2018;

Loya et al., 2001; McClanahan, Darling, et al., 2020). Indeed, all taxa-specific coral cover was

found to be negatively associated with heat stress, but the magnitude of these associations

varied between taxa (Fig. 5a-c). For instance, after one year at an average DHW of 1°C-week

we estimated a coral cover decline of 0.7% for massive corals, 0.9% for branching corals and

1.3% for plating corals. Similar differences were observed for long-term heat stress, where a

20-year period at an average DHW of 0.5°C-week was estimated to drive a decline of coral

cover by 1.7% for massive corals, 2.3% for branching corals and 4.7% for plating corals.

Life history traits mirror heat responses

Coral cover of massive corals showed the lowest rates of decline in association with heat

stress (Fig. 5c). This was not surprising, as the majority of massive corals are considered to

constitute "stress-tolerant" taxa (Darling et al., 2012; Hughes, Kerry, et al., 2018; Loya et al.,

2001). Several life history traits are suggested to drive heat tolerance in massive corals, such

as (1) thicker tissues that provide protection against heat stress (Mydlarz et al., 2010); (2)

larger corallites, which are indicators of a trophic strategy that can compensate for the

absence of the symbiotic input during a bleaching event (Conti-Jerpe et al., 2020); and (3)

slower growth rates and longer generation times, promoting energy storage (Darling et al.,

2012; van Woesik et al., 2012). Of note, the negative association between massive coral cover

and recent heat stress did not appear to be significantly modulated by the levels of past heat

stress (Fig. 5f, Fig. 5i). This might suggest that adaptive processes to heat stress rarely

occurred in massive corals over the past decades.

In contrast, branching corals displayed higher levels of coral cover decline in association with

recent heat stress (Fig. 5a). This is consistent with previous studies (Hughes, Kerry, et al.,

2018; Loya et al., 2001). Indeed, branching corals have thinner tissues, which underpins a

higher sensitivity to heat exposure, as well as smaller corallites, which are indicators of a

Symbiodinium-dependent feeding strategy (Conti-Jerpe et al., 2020; Darling et al., 2012;

Mydlarz et al., 2010). We also found that the decline of branching coral cover appeared to be

mitigated by past thermal stress (Fig. 5d, Fig. 5g), and that such mitigation was stronger for

long-term heat stress (measured for more than 20 years before surveys), than for short-term

heat stress (measured during the two years before surveys). Of note, previous works

estimated that the time frame used for computing short-term heat stress (*i.e.* two years)
corresponds to the time required for acclimatization processes that boost heat tolerance in
corals (Palumbi et al., 2014). Our results therefore suggest that branching corals might be
adapting to heat stress, and that such adaptation might occur predominantly through
evolutionary processes, rather than through acclimatization. To support these hypotheses,
recent genotype-environment association studies detected signals of heat adaptation in the
branching species of *Acropora millepora* in the south-western Pacific (Fuller et al., 2020;
Selmoni et al., 2021) and *Acropora digitifera* in Japan (Selmoni, Rochat, Lecellier, Berteaux-
Lecellier, & Joost, 2020).

The rates of evolutionary adaptation to a rapidly changing environment are expected to be
enhanced by (1) the intensity of natural selection and (2) the shortness of generation time
(Catullo, Llewelyn, Phillips, & Moritz, 2019; Kawecki & Ebert, 2004; Quigley, Bay, & van Oppen,
2019). These two elements support the view of branching corals adapting faster than massive
corals through evolutionary processes. Indeed, the selection imposed by heat stress is likely
to be more intense for branching corals, as their morphological characteristics provide
reduced protection against this kind of stress, compared to massive corals. Furthermore, the
majority of branching corals are classified as having "competitive" life history traits, defined
by higher growth rates and larger colony sizes than "stress-tolerant" massive corals (Darling
et al., 2012). As the fecundity of a coral colony is known to strongly correlate with colony size
(Álvarez-Noriega et al., 2016), the fast growth of branching corals is likely to result in shorter
reproductive cycles, leading to a faster evolutionary adaptation to environmental conditions
(Catullo et al., 2019). Of note, coral bleaching experiments in north-western Australia
reported increased heat resistance for branching corals living in thermally variable
environments, whereas massive corals from the same reefs showed lower increases in heat
resistance (Schoepf et al., 2015). This work corroborates the hypothesis of a differential
potential for heat adaptation between branching and massive corals.

Remarkably, plating corals were the taxon that showed the highest rates of coral cover
decline in association with heat stress (Fig. 5b). Plating corals are predominantly found in taxa
with a "generalist" or with a "weedy" life history trait (Darling et al., 2012). Life history traits
of these taxa reflect intermediate traits when compared with "stress-tolerant" massive corals
and "competitive" branching corals. On the one hand, corallite size is smaller than that of
"stress-tolerant" corals, which might indicate a higher sensitivity to heat stress (Conti-Jerpe

et al., 2020; Darling et al., 2012). On the other hand, growth rates are slower than
"competitive" taxa, which might cause longer generation times (Álvarez-Noriega et al., 2016;
Darling et al., 2012). We therefore hypothesize that higher rates of coral cover decline
associated with heat stress in plating corals might be due to the fact that (1) plating corals are
more sensitive to heat stress than massive corals and (2) plating coral rates of evolutionary
adaptation are slower than for branching corals. In this regard, it is interesting to observe that
the negative association between plating coral cover and recent heat stress was accentuated
by the effect of long-term heat stress (Fig. 5h). This observation might indicate that the slow
generation time of plating corals hampers coral cover recovery between heat waves.

Limitations

In this work, we assumed that local rates of coral cover were driven by the effect of degree
heating weeks as measured via remote sensed data. However, coral cover is a general
indicator of the state of health of the reef and can be affected by other environmental
stressors, many of which are not necessarily related to heat stress (De'ath et al., 2012). Two
complementary approaches could be used to mitigate this limitation. First, the association
analysis could include a wider range of environmental variables that describe other potential
stressors for corals survival (such as water pH, turbidity or human pollution; Bessell-Browne
et al., 2017; P. J. Liu, Meng, Liu, Wang, & Leu, 2012; Mollica et al., 2018). Such analyses could
also include additional metrics describing heat stress, in order to determine how the
magnitude versus the frequency of thermal anomalies contribute to coral cover decline
(McClanahan et al., 2019; Sully et al., 2019). Second, the analysis of coral cover data can be
complemented with bleaching records, as these represent an incontrovertible signal of heat
stress in corals. While this has already been done in previous studies (Hughes, Kerry, et al.,
2018; McClanahan, Darling, et al., 2020), one main limitation is that worldwide bleaching
records are scattered across space and time (Donner, Rickbeil, & Heron, 2017). This issue
could be tackled with future advances in bleaching detection via remote sensing, which could
provide a standardized view of bleaching events at a worldwide scale (Hedley et al., 2016).
The three broad categories of coral taxa that we used in this work (branching, plating and
massive) are a simplification of the morphological characteristics of corals. Indeed, corals

display a continuum of morphologies, with substantial variation observed even between

corals from the same taxa (Darling et al., 2012; Pratchett et al., 2015). We could therefore

replace these categorical classifiers of coral morphs with a set of continuous variables

describing colony shape (such as size, branch extension, radial extension, vertical extension,

horizontal extension and corallite size). This would allow future research to investigate how

the association between coral cover and heat stress is expressed along a continuum of

morphological characteristics, so that the resulting predictions on heat tolerance could be

projected to any coral species.

Perspectives

Coral taxa with different morphological traits and distinct life history traits are all suffering

significant declines in coral cover due to heat stress. If confirmed by future studies, this would

imply that there are no coral bleaching winners (Loya et al., 2001; van Woesik et al., 2011),

but only losers of different types.

For massive corals, the decline appears to be less severe, but the presumptive low rates of

evolutionary adaptation raise an alarm: how much additional heat can massive corals tolerate

without resorting to evolutionary processes? For branching corals, however, evolutionary

adaptation might promote coral cover recovery. However, such recoveries will likely depend

on how widespread the distribution of adapted colonies are presently, and at what rate the

adaptive traits would spread across reef systems in the years to come. For plating corals, the

situation appears more concerning, as neither heat sensitivity nor adaptive processes seem

to have the potential for opposing coral cover decline.

The hypotheses formulated in this work should now be tested using field experiments.

Genotype-environment association studies have already demonstrated their power for

characterizing the potential of evolutionary adaptation in coral populations at local scales (for

instance in the central Great Barrier Reef, Fuller et al., 2020; or in New Caledonia, Selmoni et

al., 2021). This approach should now be extended to wider spatial scales (for instance the

entire south-western Pacific) and consider multiple species with different life history traits, in

order to obtain a more comprehensive view of coral adaptation across reef systems and coral

taxa. In addition, such analyses should be systematically complemented with physiological

assays testing heat tolerance of corals in aquarium experiments (see e.g. Savary et al., 2021),

so that the limits of thermal resistance can be estimated for colonies from different taxa, as

well as for adapted/non-adapted colonies from the same taxon.

Recent studies on coral evolutionary adaptive potential suggest that reef conservation efforts

should be oriented toward reefs exposed to recurrent stress, as these might be the sources

for adapted corals in the years to come (Matz et al., 2020; Selmoni, Lecellier, et al., 2020). As

coral taxa might have different potentials for adaptation, the effectiveness of such

conservation strategies will also depend on the community structure of the reefs. Verification

of the hypothesis raised in this study would thus have major implications for coral reef

conservation strategies.

Conclusions

We analyzed the association between local rates of coral cover and heat stress at a global

scale and found that both recent and past heat stresses are associated with a decline of coral

cover and of coral taxonomic diversity. The magnitude of the negative effect of heat stress on

coral cover differed between coral taxa, with lower magnitude in massive, intermediate

magnitude in branching and higher magnitude in plating corals. Furthermore, we found that

the negative association between branching coral cover and recent heat stress appeared to

be mitigated by long-term past heat stress, which might suggest the existence of adaptive

processes of evolutionary origin. We hypothesize that the observed differences in the effects

of heat stress responses mirror distinct life history traits of coral taxa. If confirmed by future

studies, such taxa-specific differences would have major implications for coral reef

conservation strategies.

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