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Habitat and fishing control grazing potential on coral reefs

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Habitat and fishing control grazing potential on coral reefs

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Keywords

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Abstract

1. Herbivory is a key process on coral reefs which, through grazing of algae, can help sustain coral-dominated states on frequently-disturbed reefs and reverse macroalgal regime shifts on degraded ones.
2. Our understanding of herbivory on reefs is largely founded on feeding observations at small spatial scales, yet the biomass and structure of herbivore populations is more closely linked to processes which can be highly variable across large areas, such as benthic habitat turnover and fishing pressure. Though our understanding of spatiotemporal variation in grazer biomass is well developed, equivalent macroscale approaches to understanding bottom-up and top-down controls on herbivory are lacking.
3. Here, we integrate underwater survey data of fish abundances from four Indo-Pacific island regions with herbivore feeding observations to estimate grazing rates for two herbivore functions, cropping (which controls turf algae) and scraping (which promotes coral settlement by clearing benthic substrate), for 72 coral reefs. By including a range of reef states, from coral to algal dominance and heavily-fished to remote wilderness areas, we evaluate the influences of benthic habitat and fishing on the grazing rates of fish assemblages.
4. Cropping rates were primarily influenced by benthic condition, with cropping maximised on structurally complex reefs with high substratum availability and low macroalgal cover. Fishing was the primary driver of scraping function, with scraping rates depleted at most reefs relative to remote, unfished reefs, though scraping did increase with substratum availability and structural complexity.

5. Ultimately, benthic and fishing conditions influenced herbivore functioning through their effect on grazer biomass, which was tightly correlated to grazing rates. For a given level of biomass, we show that grazing rates are higher on reefs dominated by small-bodied fishes, suggesting that grazing pressure is greatest when grazer size structure is truncated.
6. Stressors which cause coral declines and clear substrate for turf algae will likely stimulate increases in cropping rates, in both fished and protected areas. In contrast, scraping functions are already impaired at inhabited reefs, particularly where structural complexity has collapsed, indicating that restoration of these key processes will require scraper biomass to be rebuilt towards wilderness levels.

Introduction

Herbivory is crucial to ecosystem function and community structure across terrestrial and aquatic ecosystems, playing a key role in cycling nutrients (Metcalfe et al. 2014), regulating species diversity and productivity (Royo et al. 2010, Rasher et al. 2013, Prieditis et al. 2017), and controlling habitat regime shifts (Zimov et al. 1995, Keesing and Young 2014, Verges et al. 2014). Herbivory processes are generally measured at local scales relevant to individual behaviours and population sizes, which restricts our understanding of how ecosystems function across larger spatial scales. Furthermore, anthropogenic pressures typically impact ecosystem processes, including herbivory, across much larger areas (Jackson 2008). Therefore, developing our understanding of both natural and anthropogenic drivers on herbivory at broad scales requires the integration of fine-scale herbivory observations with macroecological datasets. Such analyses are particularly relevant for coral reef ecosystems, which are facing multiple damaging human pressures and where herbivory is a key ecosystem function (Hughes et al. 2007, Cheal et al. 2010).

On tropical coral reefs, the removal of algae by herbivorous fishes is a critical process which clears space for coral settlement and growth (Bellwood et al. 2004). Herbivorous fishes can be categorized into browsers, which remove established macroalgae, and a diverse guild of grazers that feed on surfaces covered with algal turfs and associated microbial communities (Green & Bellwood 2009). Within the grazers, observations of feeding morphology and behaviour have identified two distinct grazing functions: cropping and scraping (Bellwood and Choat 1990, Polunin et al. 1995). Cropping species, primarily members of the Acanthuridae and Siganidae, remove the upper portions of the algae when feeding, which maintains algae in cropped states, promoting coral settlement and preventing transitions to fleshy macroalgae (Arnold et al. 2010). Scraping species in the tribe Scarinae gouge part of the underlying reef substratum together with microscopic epiphytes and epilithic and endolithic phototrophs when feeding (Choat and Clements 2018). In doing so, scrapers clear space for the settlement of benthic organisms, including corals (Bonaldo et al. 2014). Combined, cropping and scraping are considered essential functions which help sustain coral-dominated states (Bellwood et al. 2004, Hughes et al. 2007) and potentially reverse algal regime shifts (Graham et al. 2013).

Mature algae can proliferate in the absence of sufficient grazing pressure (Mumby et al. 2006, Burkepile and Hay 2008, Rasher et al. 2013), and correlative analyses of fished reef ecosystems have provided evidence of grazing biomass thresholds below which reefs become algae dominated (Graham et al. 2015, Robinson et al. 2018). Herbivorous fish populations are heavily exploited across much of the tropics (Edwards et al. 2014), which has compromised grazing functions on reefs which fail to maintain herbivore biomass thresholds (Bellwood et al.

2012, Graham et al. 2015, Robinson et al. 2018). However, fishing effects can be confounded by the influence of benthic productivity on herbivore populations (Russ et al. 2003, 2015), while species-specific habitat associations can also structure herbivore assemblages across a range of spatial scales (Hoey & Bellwood 2008, Doropoulos et al. 2013) and benthic compositions (Hoey & Bellwood 2011, Heenan et al. 2016). Such bottom-up influences on fish populations may be particularly strong when fish rely on habitat for both structure and food, such as algal-cropping fishes which are generally small and particularly dependent on the reef matrix for shelter (Wilson et al. 2008). Thus, herbivore assemblage structure is mediated by both habitat composition and fishing intensity but links between these drivers and grazing functions are not well resolved, particularly at macroecological scales.

Patterns in herbivore biomass are widely used to imply changes in herbivore functioning on coral reefs (e.g., Nash et al. 2016a, Robinson et al. 2018). However, biomass data overlooks size- and species-specific differences in feeding rates and functional roles. Therefore, measures of grazing impacts have been developed by integrating bite rate data with information on expected carbon intake for croppers (Marshell & Mumby 2015) or feeding behaviours for scrapers (Bellwood and Choat 1990, Bellwood et al. 2003). Furthermore, although allometric grazing ~ body size relationships (Lokrantz et al. 2008, Nash et al. 2013) indicate that the functional role provided by larger species is disproportionately greater (Bonaldo and Bellwood 2008), grazing potential may also depend on community size structure (Bellwood et al. 2012). Abundance decreases logarithmically with increasing body size, meaning that an assemblage of many small-bodied fish may be functionally equivalent to an assemblage of few large-bodied individuals (Munday and Jones 1998). Size-selective fishing which removes larger individuals (Robinson et al. 2017) and species (Taylor et al. 2014) is ubiquitous on many inhabited coral reefs and often leads to greater dominance of small-bodied fishes. However, contrasting evidence that loss of large fishes impairs bioerosion functions while compensatory increases in small fishes maintain grazing rates (Bellwood et al. 2012) suggests that links between size distributions and grazing functions are not fully resolved.

Here, we assess the drivers of herbivore functioning on coral reefs across four regions in the Indo-Pacific (Fig. S1). Our macroecological-scale analysis spans a benthic gradient from coral to macroalgal dominance and a fishing gradient from open-access fisheries to no-take fishing zones and remote wilderness areas. By integrating feeding observations with underwater visual census (UVC) data on fish abundance, we measured potential grazing rates at the scale of reef sites, which is highly relevant for understanding how benthic and fishing influences may alter ecosystem functioning (Nash et al. 2016a). We examine 1) how fishing pressure and benthic composition influences the grazing rates of two major feeding groups (croppers and scrapers), and 2) how grazing rates are controlled by both the biomass and size structure of grazing assemblages.

Materials and Methods

Survey methods

We surveyed 72 sites across Seychelles (n = 21), Maldives (11), the Chagos archipelago (25), and the Great Barrier Reef (GBR) (15) (Supplementary Methods). Grazing fish assemblages were surveyed using 8 replicate point counts of 7 m radius (Seychelles) or 4 replicate belt transects of 50 m length (Maldives, Chagos, GBR) conducted on hard-bottom reef slope habitat at 2-10 m depth. All sites were surveyed once, except for Seychelles where each site was surveyed in 2008, 2011, 2014 and 2017. Estimate of fish biomass using point counts and belt transects give comparable biomass estimates (Samoilys and Carlos 2000). Surveys were designed to minimise diver avoidance or attracting fish and were conducted by a single observer (NAJG). In point counts, large mobile species were censused before smaller territorial species. In belt transects, larger mobile fish were surveyed in a 5-m wide belt while simultaneously deploying the transect tape, and smaller site-attached damselfish species within a 2-m wide belt were recorded in the opposite direction. For both survey types, all diurnal, non-cryptic (>8 cm TL) reef-associated fish were counted and their TL estimated to the nearest centimetre. Length measurements were calibrated by estimating the length of sections of PVC pipe and comparing it to their known length prior to data collection each day. Fish lengths were then converted to body mass (grams) using published length-weight relationships (Froese and Pauly 2018) and standardised by survey area to give species-level biomass estimates that were comparable across datasets (kg ha^{-1}). The UVC dataset included 101 herbivore species (Table S1), with 11 species common to all four regions.

Herbivore species were further categorised as croppers or scrapers according to their morphology and feeding behaviour (Green and Bellwood 2009). While both groups feed primarily on the epilithial algal matrix (EAM) covered substrata, they differ in the amount of material/substratum that is removed during the feeding action. Croppers remove the upper portions of the algae and associated detritus and microbes leaving the basal portions of the algae intact on the substratum, while scrapers remove shallow pieces of the substratum together with the EAM, leaving distinct bite scars (Choat et al. 2002, Wilson et al. 2003, Hoey and Bellwood 2008).

Following fish surveys, benthic habitat composition was surveyed with eight 10-m line intercept transects (Seychelles), or four 50-m point intercept (benthos recorded every 50 cm) transects (Chagos, GBR, Maldives). We recorded the cover of hard corals, macroalgae and turf algae, as well as non-living substrate (rock, bare substrate, rubble and sand). The structural complexity of the reef was visually estimated on a six-point scale, ranging from 0 (no vertical relief) to 5 (complex habitat with caves and overhangs) (Polunin and Roberts 1993), which correlates strongly with a range of other methods for capturing the structural complexity of coral reefs (Wilson et al. 2007).

Herbivore feeding observations

Feeding observations of Indo-Pacific grazing fishes provided species-level estimates on bite rates of croppers and scrapers. Surveys were conducted in the Red Sea, Indonesia, and GBR. We analysed feeding observations for species observed in the UVC dataset (n = 39). Briefly, an individual fish of a target species was haphazardly selected and its body length (total length in cm) estimated. After a ~30 second acclimation period, each individual was followed for a minimum of 3 minutes during which the number of bites and the feeding substratum was recorded. We estimated the average feeding rate (bites per minute) for each observed fish. For scrapers, we also estimated the bite scar size using a separate dataset in which one diver followed individual fish and recorded the length and width of each bite scar, and estimated the total length of the fish.

Grazing rate estimates

We used feeding observations to convert UVC biomass estimates into the total grazing potential of croppers and scrapers. We defined grazing functions separately for each functional group whereby cropping function was measured as feeding intensity (bite rate data) and scraping function was measured as area grazed (bite rate and bite area data). We used a Bayesian hierarchical modelling framework that estimates species- and genera-level functional rates, which allowed us to estimate grazing rates for UVC species which were not observed in feeding surveys (n = 63). Cropper function was quantified in terms of potential feeding intensity, the total number of bites per minute, and derived from a predictive model which accounted for species- and genera-specific bite rates (Supplementary Methods, Table S2). We then used allometric relationships to convert bite rates into grams of carbon (g C) removed through EAM consumption (Marshell and Mumby 2015). For scrapers, we defined scraping function in terms of potential area of substrata cleared per minute. Feeding observations provided estimates of bite rates, which we modelled as a function of body size (*TL*, cm; $r = -0.43$) according to species- and genera-specific grazing rates (Supplementary Methods, Fig. S2, Table S2). We used bite area estimates to convert bite rates into area scraped per minute (m^2 $minute^{-1}$). Cropping and scraping rates were assigned to all observed species, corrected by fish biomass, then summed within surveys and averaged to give site-level estimates of potential grazing function (croppers = g C ha^{-1} min^{-1}, scrapers = m^2 ha^{-1} min^{-1}).

Explanatory covariates

First, to account for fishing effects ranging from the remote and protected Chagos archipelago to heavily-exploited reefs in Seychelles, we estimated fishable biomass as a proxy for exploitation pressure. This proxy, based on total fish community biomass, is highly sensitive to exploitation pressure and predicted by human population size, access to markets, and fisheries management (Cinner et al. 2016), and has been used to represent large-scale fishing gradients in numerous studies (e.g. McClanahan et al. 2011, Graham et al. 2017). Reefs were also assigned a categorical fishing pressure covariate to distinguish between protected (i.e. no-take areas), exploited, and remote reefs (Supplementary Methods).

Second, benthic surveys provided site-level estimates of benthic composition. We estimated structural complexity and the site-level cover for four major habitat-forming groups (live hard coral, macroalgae, available substrate, and rubble) by averaging across replicates at each site. Available substrate was the total cover of rock, bare substrate, and turf algae, and represents the area of substrate available for EAM growth.

Third, we estimated the biomass of each functional group (kg ha^{-1}) and a large fish indicator (LFI) as a measure of size structure (Robinson et al. 2017). We use the LFI to measure the relative abundance of large-bodied fish, which are considered key contributors to grazing functions because of their high per-capita consumption rates (Lokrantz et al. 2008) and long foraging movements (Nash et al. 2013). We defined large fish separately for each group as the length at the 75% quantile of the size distribution in the full dataset, such that the LFI was the relative abundance of fish greater than 15 cm for croppers and 30 cm for scrapers. Biomass and the LFI were estimated for each replicate and then averaged for each reef.

Statistical modelling

We modelled variation in herbivore functioning according to 1) gradients in benthic habitat composition and fishing pressure and 2) grazing rates estimated from grazer biomass and assemblage size structure. To place modelled effect sizes on a common scale, we scaled and centered all continuous covariates to a mean of zero and standard deviation of one and converted the categorical fishing status covariate into two dummy variables (fished - protected, fished - remote) (Schielzeth 2010). We used multimodel inference to assess parameter effect sizes. For each function, we fitted a global linear mixed effects model with five benthic fixed effects (hard coral, macroalgae, available substrate, rubble, structural complexity) and three fishing fixed effects (fishable biomass, remote reef, protected reef), for gamma-distributed errors (ϵ). Potential covariance among reefs in the same dataset and year was modelled using nested random intercept terms where, for each observation *i* at each reef *j* in dataset *k*:

$$grazing_{ijk} = \beta_0 + \beta_1 hardcoral_{ijk} + \beta_2 substrate_{ijk} + \beta_3 rubble_{ijk} + \beta_4 macroalgae_{ijk} + \beta_5 complexity_{ijk} + \beta_6 fishablebiomass_{ijk} + \beta_7 fished.protected_{ijk} + \beta_8 fished.remote_{ijk} + reef_j + dataset_k + \epsilon_{ijk} \quad \text{Eq. 1}$$

From the global model, we fitted all possible subset models (Bartoń 2013) and assessed their support using Akaike's Information Criterion corrected for small sample sizes (AICc), where the top-ranked model had the lowest AICc score (Burnham and Anderson 2003). Initial modelling indicated support for multiple competing models (i.e. ΔAICc < 2), so we visualised relative covariate effect sizes by extracting standardised t-values for all models within 7 AICc units of the top-ranked model and, for each model, rescaling t-values so that 1 is the strongest predictor in a given model, and weighing that value by the models' AICc weight (Cade 2015). These scaled t-values represent the relative effect size of each covariate between 0 (unimportant) and 1 (important). Next we generated model predictions to visualise the effect of each covariate with scaled t-value > 0.4, excluding remaining fixed effects and random effects and correcting predictions by each models' AICc weight, with prediction uncertainty represented by the AICc-weighted sample variance (Robinson et al. 2017). Our multi-model approach accounts for uncertainty in the 'best' fitted model when AICc scores indicate several models are equally valid (Burnham and Anderson 2003). We avoid potential biases in model-averaged coefficient sizes by presenting effect sizes as standardised t-values, which are more informative measures of covariate importance than sums of AICc weights (Cade 2015).

Benthic and fishing influences on assemblage-level grazing rates will be underpinned by differences in the number and size of grazing fishes (Hoey & Bellwood 2008). Indeed, as grazing estimates were derived from feeding data combined with UVC biomass data we expected grazer biomass to correlate strongly with grazing rates. Although size-selective overfishing is expected to have disproportionate impacts on grazing function (because grazing rates increase with body size; Lokrantz et al. 2008), depletion of large-bodied fish may be offset by increased abundances of smaller individuals (Bellwood et al. 2012). Thus, we examined how grazing functions vary with assemblage size structure by modelling the effects of grazer biomass and the proportion of large-bodied fishes (LFI; number of individuals > 15 cm for croppers or 30 cm for scrapers) on grazing rates. For each function, we fitted a generalized linear mixed effects model with interaction between biomass and LFI, for each observation *i* at each reef *j* in dataset *k*, and Gamma-distributed errors:

$$grazing_{ijk} = A + B.biomass_{ijk} * C.LFI_{ijk} + reef_j + dataset_k + \epsilon_{ijk} \qquad \text{Eq. 2}$$

We weighed model support for each covariate and the interaction between biomass and the LFI with AICc (Burnham and Anderson 2003), selecting the top-ranked model for interpretation and visualization. We visualized the continuous interaction by estimating grazing rates across the range of observed grazer biomass at two LFI values: dominance by small fishes was represented by an assemblage with LFI = 0.25 (i.e. 25% of individuals were large-bodied), and dominance by large fishes was represented by an assemblage with LFI = 0.75 (i.e. 75% of individuals were large-bodied).

All data were analysed in R (R Core Team 2018), using packages *lme4* (linear mixed effect models, Bates et al. 2015), *MuMIn* (multimodel inference, Bartoń 2013), and *rethinking* (Bayesian models, McElreath 2017).

Results

For cropping fishes, 9 species were assigned individual bite rates (representing 32.9% of biomass for this group), and remaining species were assigned genera-specific (54.4%) or an average cropper bite rate (12.6%). Assemblage-level cropping rates ranged from 0.04 to 5.52 g C ha^{-1} min^{-1}, with cropping highest on GBR and Chagos reefs (Fig. S3A). Irrespective of region, cropping was maximised in complex habitats with high substrate availability and low macroalgal cover (Fig. 2A-C), while hard coral or rubble cover were weak influences (Fig. 1). Cropping rates were weakly affected by fisheries management status, and were similar across remote, protected and fished reefs (Fig. 1).

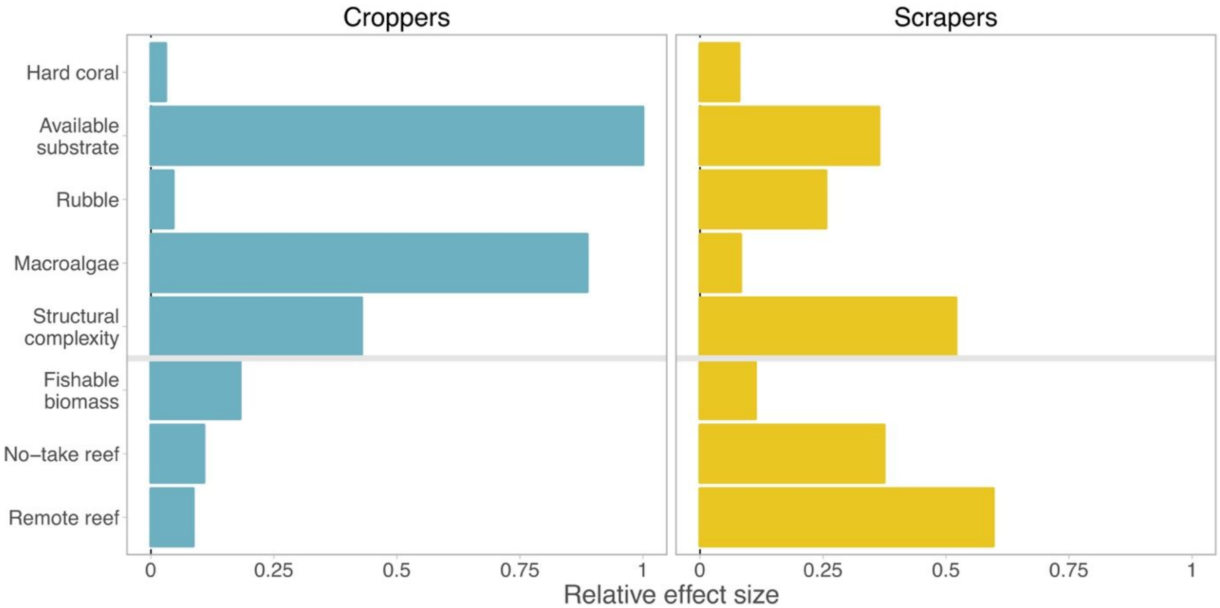

Figure 1. Relative effect of benthic composition and fishing pressure on modelled grazing rates for croppers (left) and scrapers (right). Bars are relative effect size ratios of each covariate for top-ranking model sets (models ≤ 7 AICc units of top-ranked model), scaled to indicate very weak (0) or very important (1). See Table S3 for covariate effect sizes across the top-ranking model sets.

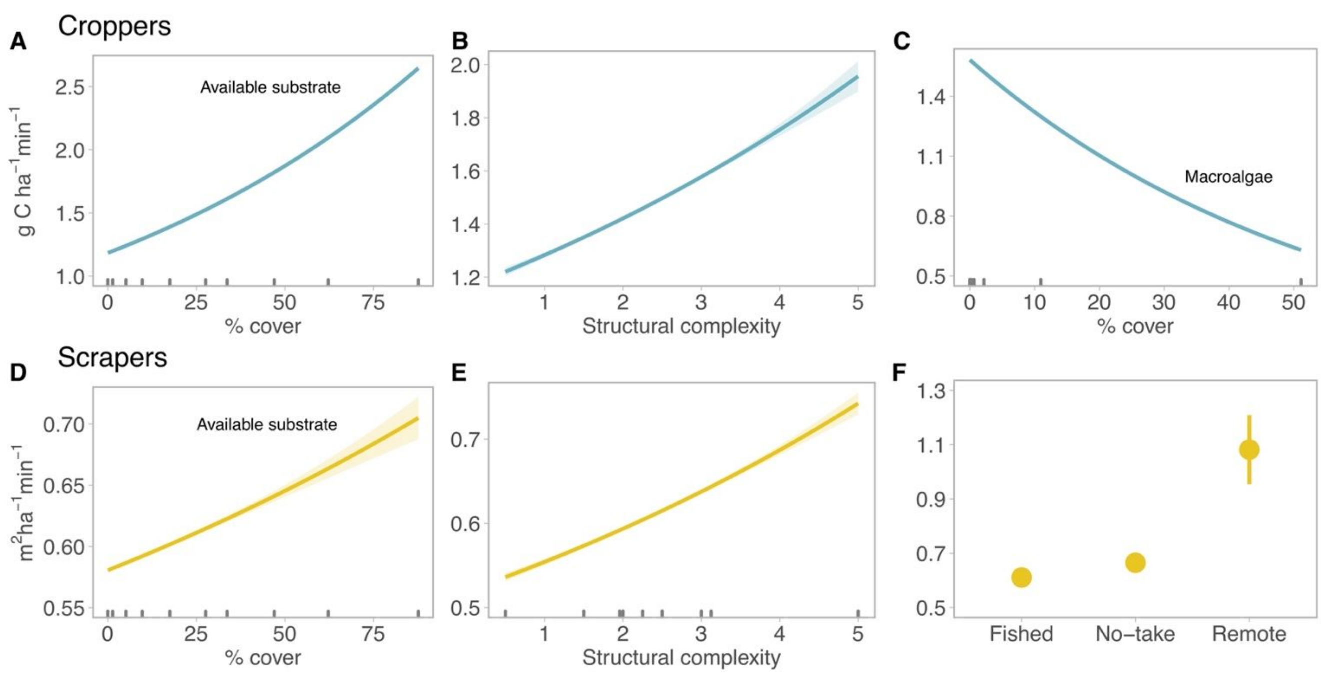

Figure 2. Predicted effects of benthic and fishing drivers on potential cropping (A-C) and scraping (D-F) rates. Benthic effects are available substrate (A, D) and structural complexity

(B, E) for both grazing groups, and macroalgae (C) for croppers. Fishing effects are management status for scrapers (F). Lines and points are grazing rates as predicted by top model sets (≤ 7 AICc units from top-ranking model) holding other covariates to their means, with each model prediction weighted by its AICc weight and error represented as sample variance. All visualized covariates had relative effect size ratios > 0.4 (Fig. 1). Decile rugs indicate the spread of observed data.

Feeding data were more highly resolved for scraping herbivores, with all fishes assigned size-specific bite areas, and either species- (27 of 35 species, 80.9% of UVC) or genera-specific bite rates (19.1%). Scraping rates were greatest on GBR reefs (> 1 m^2 min^{-1} ha^{-1}) and lowest on Maldives reefs (< 0.3 m^2 min^{-1} ha^{-1}) (Figure S4B). Scraping rates increased with available substrate (Fig. 2D) and structural complexity (Fig. 2E), but in contrast to croppers, were relatively invariant with macroalgal cover (Fig. 1). Remote reefs had the greatest scraping rates, which were considerably lower on fished and protected reefs (Figs. 1, 2D). After accounting for these coarse protection effects, scraping was only weakly associated with total fishable biomass (Fig. 1).

Herbivore biomass is often used as a proxy for the magnitude of their function, but the relationship between biomass and function is rarely tested. Here, cropping rates were strongly and positively correlated with cropper biomass (R^2 = 0.99, Fig. 3A), indicating that the drivers of biomass variation would match tightly to the modelled drivers of cropper function. Similarly, scraping rates increased with scraper biomass but with greater levels of unexplained variation (R^2 = 0.81) which occurred across the biomass gradient (Fig. 3B). Size structure (LFI, the proportion of large-bodied individuals in each assemblage) modified function ~ biomass relationships, with potential grazing function increasing as assemblages became dominated by smaller-bodied individuals (Fig. 3, Table 1). Size structure effects were moderately stronger for scrapers (parameter coefficient = -0.317 ± 0.03 standard error) than croppers (-0.087 ± 0.001).

For example, at average grazer biomass levels (croppers = 65 kg ha^{-1}, scrapers = 370 kg ha^{-1}), grazing rates were 15% (croppers) and 21% (scrapers) greater in small-bodied assemblages (LFI = 25%) than in large-bodied assemblages (LFI = 75%).

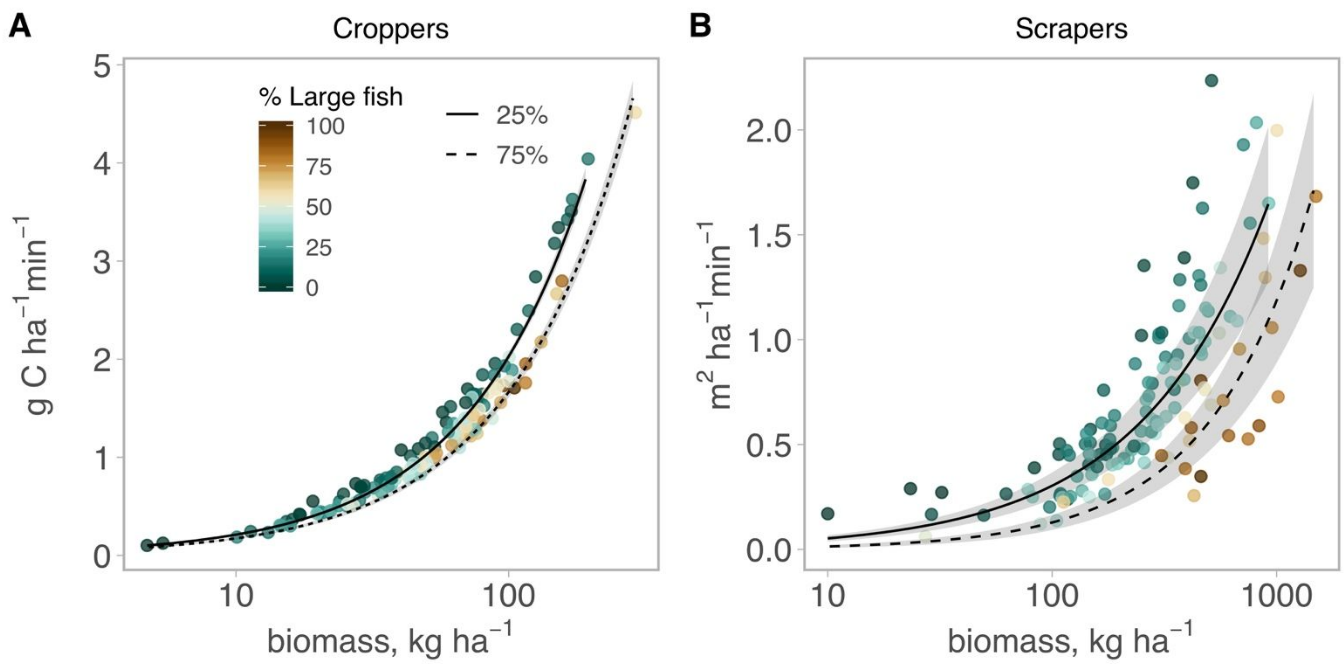

Figure 3. Association between grazing function, grazer biomass, and assemblage size structure. Reef-level estimates of cropper algal consumption (A) and scraper area grazed (B) plotted against UVC biomass (log_{10} scale), coloured by the LFI. Lines are model fits of grazing ~ biomass relationships for small-bodied assemblages (solid line: 25% of individuals are large-bodied fish) and large-bodied assemblages (dashed line: 75% of individuals are large-bodied fish), shaded with two standard errors. Large fishes are defined as ≥ 15 cm for croppers and ≥ 30 cm for scrapers.

Table 1. AIC selection for grazing function ~ grazer biomass + LFI models. Parameter coefficients, AICc and AICc weights are shown for all competing models, ranked by AICc and with the top-ranked model in bold.

Intercept	Biomass	LFI	LFI*biomass	AICc	ΔAICc	AICc weight
Croppers						
0.024	**0.728**	**-0.087**	**-**	**-296.935**	**0**	**0.748**
0.025	0.727	-0.086	-0.002	-294.759	2.176	0.252
0.077	0.681	-	-	-208.064	88.871	0
0.414	-	0.183	-	226.190	523.125	0
0.362	-	-	-	4.000	239.595	0
Scrapers						
-0.581	**0.693**	**-0.317**	**0.084**	**-117.791**	**0**	**1**
-0.542	0.654	-0.306	-	-100.337	17.454	0
-0.526	0.522	-	-	-45.345	72.446	0
-0.445	-	-	-	97.598	215.389	0
-0.446	-	0.074	-	98.559	216.350	0

Discussion

Evaluating herbivory through a macroecology lens provides insights into the functioning of a broad range of coral reefs, including coral, rubble and algal benthic states in both remote and exploited ecosystems. We found that herbivore assemblage grazing rates varied substantially across the Indo-Pacific, and in accordance with top-down (i.e. fishing pressure) and bottom-up (i.e. benthic habitat) drivers which were specific to each functional group. Cropping rates were primarily controlled by bottom-up influences, with function maximised in complex habitats that feature high substrate availability and low macroalgae cover. Conversely, for parrotfishes, scraping rates were maximised on remote reefs in the Chagos archipelago which is isolated from fishing pressures, and increased with available substrate and structural complexity. Benthic and fishing influences were underpinned by the strong dependence of grazing rates on fish biomass, although we also demonstrate that reefs dominated by small-bodied fishes exert moderately greater grazing rates.

Cropping rates were primarily mediated by benthic habitat type, in particular structural complexity, macroalgae cover, and substrate availability. Our results emphasize the strong dependence of small-bodied reef fishes on benthic composition (Munday and Jones 1998, Wilson et al. 2010), and demonstrate that potential cropping function is relatively unaffected by top-down fishing effects, likely because cropping assemblages are mostly comprised of small-bodied fishes which are not targeted in many reef-associated fisheries (Hicks & McClanahan 2012). Strong relationships between benthic composition and the grazing function of small-bodied reef fish likely reflects the importance of resource availability, which has been shown to have stronger control on cropping surgeonfishes than fishing pressure (Russ et al. 2018). For example, the decrease in cropping rates with increasing macroalgae may be due to feeding avoidance in macroalgal-dominated areas (Hoey & Bellwood 2011), as well as lower accessibility of turf algae under macroalgal canopies (Roff et al. 2015). In contrast, reefs with high EAM (i.e. substrate availability) support expansive and easily accessible turf mats which are targeted by large grazer populations (Williams & Polunin 2001), which in turn limit the development of larger macroalgae. Strong benthic effects imply that cropper functioning will respond more strongly to habitat disturbances, such as coral bleaching, severe storms or enrichment of algal communities, than to fishing. Indeed, disturbances which increase substrate availability for turf algal growth, such as coral mortality from heat stress, typically stimulate an increase in grazer abundance (Wilson et al. 2006, Gilmour et al. 2013, Russ et al. 2018). However, since structural complexity was also shown to be a strong driver, any positive rebound of cropping function may be negated if disturbances also erode structural complexity (Graham et al. 2006).

Scraping was strongly influenced by fishing pressure at inhabited reefs, with exploitation suppressing scraping rates far below those supported at remote, unfished reefs. This effect was stronger than influences of benthic cover and small-scale fishing protection, suggesting that bottom-up control of scraping assemblages on reefs is a relatively weak influence on their function, and that small-scale fishing protection does not conserve wilderness levels of scraping function. Movement of fish across reserve boundaries, particularly larger-bodied parrotfish which have larger home ranges (Green et al. 2014), and poor compliance with fishing regulations (Bergseth et al. 2018) likely limited the effectiveness of these small MPAs, many of which are adjacent to fishing grounds. Indeed, local extirpation of one parrotfish species (*Bolbometopon muricatum*) across the Indo-Pacific has also diminished bioerosion and coral predation functions (Bellwood et al. 2012). Scraping rates also increased moderately with structural complexity, further underlining the importance of coral reef structure in supporting herbivory (Nash et al. 2016a). As with croppers, the positive effect of available substrate on scraping rates is consistent with evidence that many scraping species respond positively to disturbances that clear substrate area (e.g. coral declines, Wilson et al. 2006), with increases in scraping function likely to promote coral recovery (Gilmour et al. 2013).

By modelling observed grazing rates and omitting benthic and fishing covariates, we demonstrated how grazing rates can vary simply as a function of biomass and size structure. Because grazing rates were positively correlated with grazer biomass and grazing calculations were derived from body mass estimates, this suggests that benthic and fishing drivers are proximate drivers of grazing function through their effect on biomass. However, for a given level of biomass, assemblages dominated by small-bodied fishes had a higher grazing potential than those dominated by large-bodied fishes. These findings are consistent with evidence that grazing functions on exploited reefs may be maintained by high densities of small-bodied parrotfish (Bellwood et al. 2012). Smaller fish have higher mass-specific metabolic rates (Gillooly et al. 2001) and thus may feed more intensively per unit of fish biomass than large fish. Therefore, this may explain why the LFI relationship was strongest for scraping rates which were modelled using size-specific feeding data. In contrast, large-bodied fishes comprised a greater fraction of assemblage biomass on high-biomass reefs (e.g. > 500 kg ha^{-1}, Fig. 3), suggesting that reefs where grazing functions are maintained by few large individuals may be particularly vulnerable to fishing effects.

To integrate UVC data across the Indo-Pacific we generalized across cropper species which are known to perform distinct feeding roles. For example, croppers have well-documented differences in morphology, diet (e.g. detritivores or turf), and feeding behaviours (Choat et al. 2002, Wilson et al. 2003, Tebbett et al. 2017), though large-scale studies such as ours typically aggregate all cropping species into a single functional group (e.g. Heenan et al. 2016). We defined cropping function using species- or genera-specific bite rates, with a high proportion of individuals assigned average grazing rates. As such, current practices for estimating cropping function at assemblage scales are largely reflective of biomass levels rather than species-specific differences in feeding rate. For scraping functions, which are more consistent among species (Bellwood and Choat 1990, Bonaldo et al. 2014) and more finely resolved with species- and size-specific bite rates, our results suggest that grazing rates can partially decouple from grazing biomass. Such patterns support recent findings that grazing metrics which include species-specific feeding behaviours are better predictors of benthic change than grazing biomass (Steneck et al. 2018). For both functions, our approach of modelling genera- and species-specific bite rates from observations collected in several regions enabled us to leverage observational data in a hierarchical framework which predicts grazing rates of new, related species, given uncertainties in species and genera (and body size for scrapers). For example, we were able to assign bite rates to species observed in UVC but not observed in feeding surveys, with estimates that were informed by the feeding behaviour of closely related congeners. Such models could be further improved with additional feeding data on other herbivore species in different regions, and could even be developed to account for temperature effects on grazing rates (Bruno et al. 2015) and examine how herbivory might respond to ocean warming.

Random intercepts in the predictive models indicated that regional differences in grazing rates were unexplained by benthic and fishing covariates, which is likely due to unmeasured processes that control feeding rates and herbivore biomass. For example, herbivore biomass variation (and thus grazing function) has been linked to differences in benthic (Russ et al. 2003) and oceanic productivity (Heenan et al. 2016). Similarly, behavioural observations indicate that grazing intensity is constrained by wave exposure (Bejarano et al. 2017) and sedimentation (Goatley & Bellwood 2012), while scraping rates can be higher in no-take fishing areas (Nash et al. 2016b) which may have led us to underestimate grazing function on protected reefs. More broadly, our space-for-time approach precludes detection of non-linear changes in grazing rates that may arise when herbivore assemblages reorganize in response to acute disturbances (Han et al. 2016). Temporal analyses which link habitat suitability, primary productivity, and herbivory would greatly develop our understanding of how grazing functions influence long-term changes in reef state.

By integrating feeding rates with UVC data across a gradient of grazing biomass, we generated reef-level estimates of potential grazing pressure at four Indo-Pacific coral reefs. Our study demonstrates how benthic habitat and fishing pressure influence the functional potential of herbivore assemblages, at relevant scales for understanding ecosystem-level responses to disturbances such as bleaching (Nash et al. 2016a). Cropping pressure is likely to increase in response to stressors which clear substrate space for turf growth, though responses to physical disturbances will vary across species according to their life history characteristics (e.g. recruitment rates, Russ et al. 2018). Intact reef structure will be critical for maintenance of scraping functions, though reefs in close proximity to human populations are unlikely to return to wilderness levels of grazing pressure, even with protection from fishing (MacNeil et al. 2015). For a given level of biomass, dominance by smaller-bodied fishes will enhance grazing, though we stress that biomass was by far the most important predictor of grazing functions and recovery or protection of fish biomass will help ensure herbivory processes are functionally intact on degraded coral reefs (Williams et al. 2016).

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Authors' contributions

JR conceived the study. AGL, AH, KN, SW and NG designed field surveys and collected ecological data. JR, JMI, JD, JH, SH analysed data and wrote the first draft of the manuscript. All authors contributed to interpretation of results and provided editorial comments.

Data accessibility

Data and R scripts are provided at github.com/jpwrobinson/grazing-grads.

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Habitat and fishing control grazing potential on coral reefs

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Supplementary Methods

Region details

In Seychelles, 21 reefs were surveyed in 2008, 2014, and 2017 on two inhabited islands (Mahe, Praslin). Surveys were conducted on the reef slope at 9-12 m depth, and stratified to include carbonate fringing reefs, granitic rocky reefs with coral growth, and patch reef habitats on a sand, rubble, or rock base (Fig. S1B). Surveys were repeated for either 8 (2008) or 16 (2011, 2014, 2017) replicates at each reef, which were located at least 15 m away from each other. To ensure that survey effort was comparable among Seychelles reefs, we only considered surveys from the first 8 replicates (per site per survey year). Overall, the surveys covered up to 0.5 km of reef front and 2,500 m^2 of reef habitat, including 672 point counts over 4 surveyed years. Reefs were categorised by their exploitation status, with 9 sites in small protected areas and 12 sites supporting artisanal fisheries.

In the Chagos archipelago, 25 reefs were surveyed on four uninhabited atolls in 2010 (Fig. S1B). Surveys were stratified to include sheltered (9) and exposed (9) habitats, and four replicate transects were conducted at each site, resulting in 100 total transects. All reefs were categorised as remote.

In Maldives, 11 reefs were surveyed on one atoll (Huvadhoo) in 2013 (Fig. S1B). Surveys were conducted on the reef slope for 4 replicates per reef, resulting in 44 total transects. All reefs were categorised as fished.

In Australia, five reefs were surveyed on the central Great Barrier Reef in 2010 and 2011 (Wheeler, Davies, Rib, Trunk, John Brewer) (Graham et al. 2014) (Fig. S1B). Reefs were stratified to include 3 wave exposed and 3 wave sheltered locations (6 per reef), which were further divided into reef slope (7-9 m depth), reef crest (2-3 m depth), and reef flat (100 m distance from crest). Each location and habitat type was surveyed with four replicate transects. We used data for surveys conducted on the reef slope, which produced a dataset of 24 transects per reef and 120 transects in total. Davies, Rib, Trunk and John Brewer were categorised as fished, and Wheeler was categorised as protected (no-take zone).

Benthic categories

To understand the range of benthic habitat types across the dataset, we categorised reefs according to their benthic regime, using a correlation-based PCA and K-means clustering (Jouffray et al. 2015). The optimal number of clusters was found using an elbow method with k = 2-15 range, and then applied to the K-means clustering. For reefs in Seychelles which were surveyed in multiple years, we estimated regimes at each site by averaging cover values over time. Across all reefs, we detected three benthic regimes characterised by 1) hard coral dominance, 2) high availability of bare substrate, and 3) rubble reefs (Fig. S1B). Coral dominance was the most common regime, detected at 53 reefs across all four regions, whereas bare substrate regimes were only present in Seychelles (14) and Chagos (7). Rubble reefs were only present in Seychelles (6 reefs).

Bite rate models

Cropper function was quantified in terms of potential feeding intensity, the total number of bites per minute, and derived from a predictive model which accounted for species- and genera-specific bite rates (Eqs. 1,2). In our cropper feeding data, bite rates were only weakly correlated with TL (Pearson's r = -0.18), and so we assumed bite rates were unrelated to body size.

$$biterate = Gamma(\mu, \theta) \qquad \text{Eq. 1}$$

$$log(\mu) = \beta_0 + species_i + genus_j + dataset_k \qquad \text{Eq. 2}$$

From this model, we generated species- and genera-level posterior predictions of grazing rates and assigned to each individual cropping fish observed in UVCs. Fish belonging to genera which were not present in the feeding observation dataset were assigned average feeding rates irrespective of species and genera (i.e. the model intercept, `\beta_0`). Following Van Rooij et al. (1998), daily carbon intake was linked to body mass (M, grams) as

$$gC = 0.0342.M^{0.816} \qquad \text{Eq.3}$$

which we then divided by the predicted number of bites per day to produce an estimate of grams carbon consumed per minute by each individual cropping fish. Our approach accounts for bite size increasing with body size, meaning that larger fish will have greater carbon intakes (Marshell & Mumby 2015). We summed estimates within UVC replicates (i.e. point count or transect) and averaged across replicates to give site-level estimates of potential cropping function.

Scraping function was quantified terms of potential area of substrata cleared per minute. Feeding observations provided estimates of bite rates, which we modelled as a function of body size (TL, cm; r = -0.43) according to species- and genera-specific grazing rates, for gamma distributed errors (Eqs. 4, 5).

$$biterate = Gamma(\mu, \theta) \qquad \text{Eq. 4}$$

$$log(\mu) = \beta_0 + \beta_1 * TL + species_i + genus_j + dataset_k \qquad \text{Eq. 5}$$

To account for potential differences in scraping action among species and across body sizes, we used a second underwater feeding observation dataset of scraper bite areas. Bite scar area (cm^2) was modelled as a function of body size (TL, cm; r = 0.83), for Gamma distributed errors (Eqs. 6,7).

$$scarsize = Gamma(\mu, \theta) \qquad \text{Eq. 6}$$

$$log(\mu) = \beta_0 + \beta_1 * TL \qquad \text{Eq. 7}$$

By including size (TL) as an explanatory covariate, models accounted for scar area increasing with body size (Fig. S2A) and bite rates decreasing with body size (Fig. S2B). For each observed scraper in the UVC dataset, we generated posterior predictions for bite rate and scar size according to its species identity and body size. Species which were not observed in feeding observations were assigned genera-level bite rates. These predictions were converted to area scraped per minute (bite rate x scar size = area scraped) (m^2 ha^{-1} min^{-1}), summed within surveys and averaged to give site-level estimates of potential scraping function.

All models fitted to feeding data were fitted with weakly informative priors (Table S2) using Markov Chain Monte Carlo sampling implemented in Stan. We sampled three chains of 3,000 iterations (warmup = 1,500) each for model checks, and one long chain of 5,000 iterations (warmup =1,500) for generating grazing predictions. Model convergence was assessed by inspecting posterior predictions, Gelman-Rubin diagnostic (Rhat), and the number of effective samples (Table S2).

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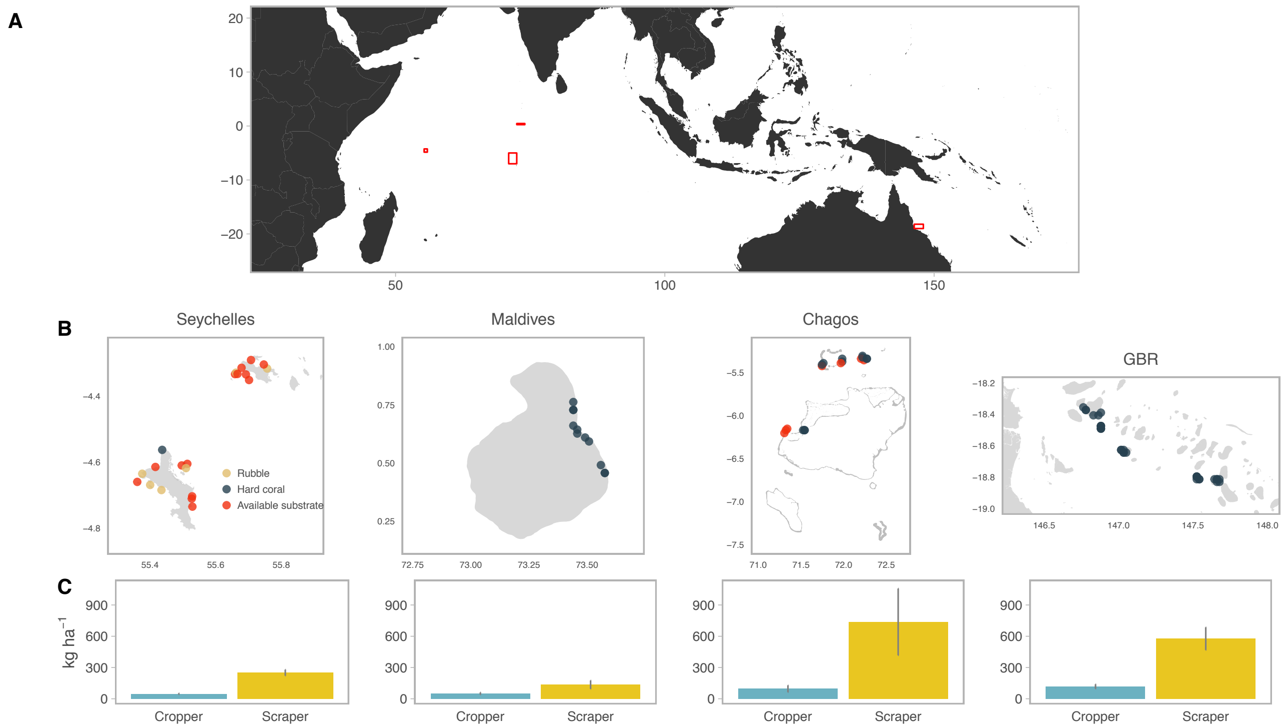

Figure S1 | Map of study sites with benthic habitat regimes (B) and herbivore biomass levels (C). Survey sites are coloured by regimes identified in k-cluster analysis (coral = blue, substrate = red, rubble = yellow), and bar plots show mean grazing biomass (± 2 standard errors) for croppers and scrapers.

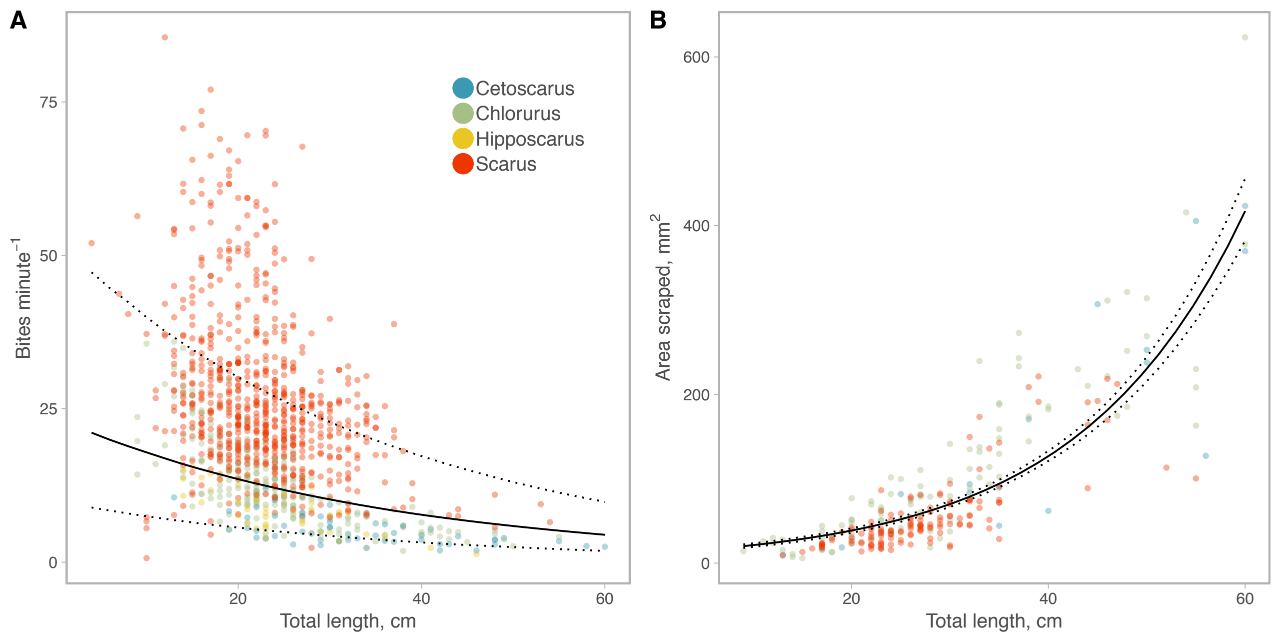

Figure S2 | Size effects on scraper bite rates (A) and bite area (B). Lines indicate median posterior predictions with 95% certainty intervals, excluding species and genera effects, across the range of observed body sizes (total length, cm). Points are observed bite rates or bite areas coloured by genera.

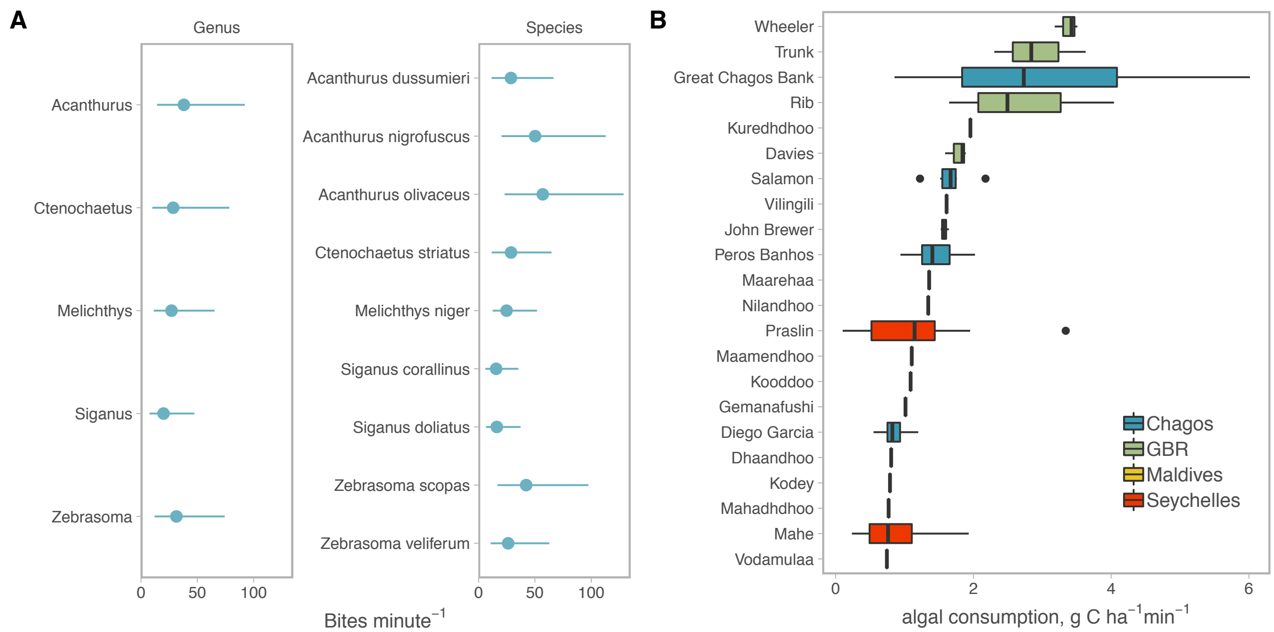

Figure S3 | Cropper bite rate predictions (A) and observed cropper function in UVC (B) Predicted bite rates are median posterior predictions with 95% certainty intervals (A), and boxplots are site-level observed cropping function for each reef, coloured by UVC region.

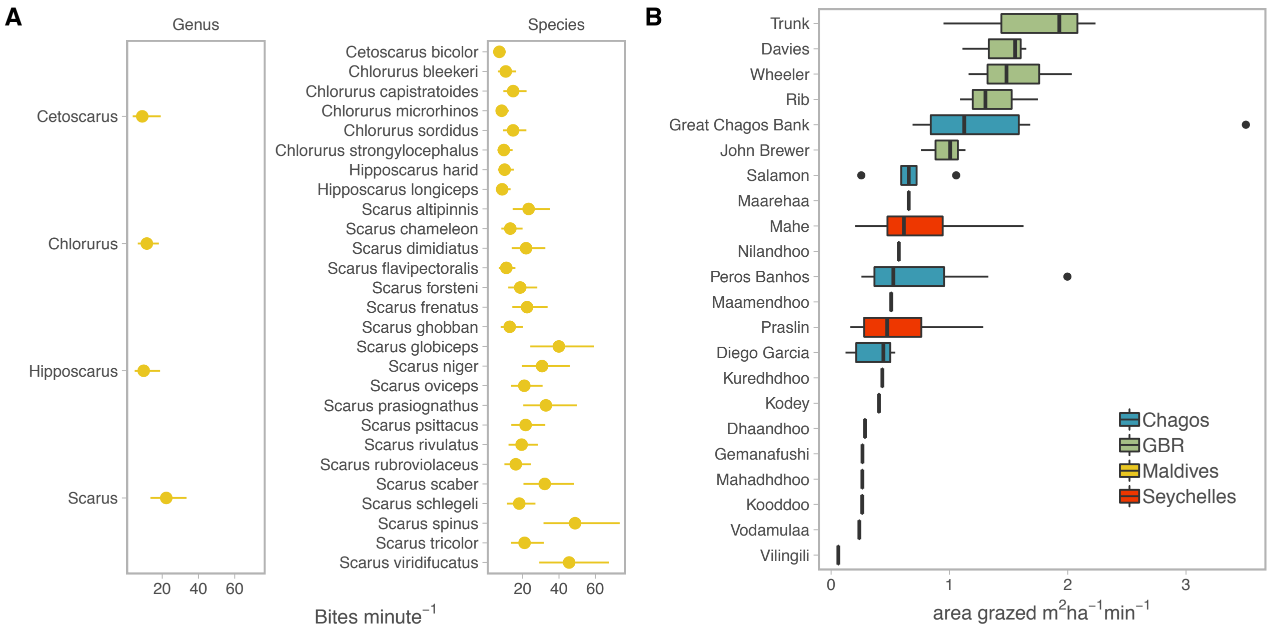

Figure S4 | Scraper bite rate predictions (A) and observed scraping function in UVC (B) Predicted bite rates are median posterior predictions with 95% certainty intervals (A), and boxplots are site-level observed scraping function for each reef, coloured by UVC region.

Croppers	*Acanthurus auranticavus, Acanthurus blochii,* ***Acanthurus dussumieri****, Acanthurus leucocheilus, Acanthurus leucosternon, Acanthurus lineatus, Acanthurus nigricans, Acanthurus nigricauda,* ***Acanthurus nigrofuscus****, Acanthurus nigroris,* ***Acanthurus olivaceus****, Acanthurus tennenti, Acanthurus tennentii, Acanthurus triostegus, Acanthurus tristis, Centropyge bicolor, Centropyge bispinosa, Centropyge vrolikii, Chrysiptera biocellata, Ctenochaetus binotatus,* ***Ctenochaetus striatus****, Ctenochaetus truncatus, Dischistodus melanotus, Dischistodus perspicillatus, Dischistodus prosopotaenia, Dischistodus pseudochrysopoecilus,* ***Melichthys niger****, Plectroglyphidodon lacrymatus, Plectroglyphidodon leucozonus, Plectroglyphidodon phoenixensis, Pomacentrus amboinensis, Pomacentrus bankanensis, Pomacentrus indicus, Pomacentrus nagasakiensis, Pomacentrus trilineatus, Pomacentrus wardi,* ***Siganus corallinus****,* ***Siganus doliatus****, Siganus puelloides, Siganus puellus, Siganus punctatus, Siganus spinus, Siganus stellatus, Siganus vulpinus, Stegastes apicalis, Stegastes fasciolatus, Stegastes lividus, Stegastes nigricans, Zebrasoma desjardinii,* ***Zebrasoma scopas****,* ***Zebrasoma veliferum***
Scrapers	***Cetoscarus bicolor****, Chlorurus atrilunula,* ***Chlorurus bleekeri****,* ***Chlorurus capistratoides****, Chlorurus enneacanthus,* ***Chlorurus microrhinos****,* ***Chlorurus sordidus****, Chlorurus stronglycephalus,* ***Hipposcarus harid****,* ***Hipposcarus longiceps****,* ***Scarus altipinnis****, Scarus capistratoides, Scarus caudofasciatus,* ***Scarus chameleon****,* ***Scarus dimidiatus****, Scarus falcipinnis,* ***Scarus flavipectoralis****,* ***Scarus forsteni****,* ***Scarus frenatus****,* ***Scarus ghobban****,* ***Scarus globiceps****,* ***Scarus niger****,* ***Scarus oviceps****, Scarus prasiognathos,* ***Scarus psittacus****,* ***Scarus rivulatus****,* ***Scarus rubroviolaceus****,* ***Scarus scaber****,* ***Scarus schlegeli****,* ***Scarus spinus****,* ***Scarus tricolor****,* ***Scarus viridifucatus***

Table S1 | Nominal cropping and scraping herbivores surveyed in UVC. Species with feeding observations are indicated in bold.

	Parameter	Prior	Mean	Lower 89%	Upper 89%	Effective samples	$\hat{R}$
Cropping bite rate	X	$N(3.43, 10)$	3.346	2.655	4.080	357	1.00
	θ	$Exp(2)$	4.937	4.546	5.239	1500	1.00
	species	$N(0, \sigma_s)$	0.414	0.172	0.622	486	1.00
	genus	$N(0, \sigma_G)$	0.453	0.004	0.839	188	1.03
	region	$N(0, \sigma_d)$	0.372	0.004	0.753	356	1.00
	$\sigma_s, \sigma_G, \sigma_d$	$Cauchy(0, 1)$					
Scraping bite rate	A	$N(3.10, 10)$	3.161	2.491	3.794	718	1.00
	B	$N(0, 5)$	-0.028	-0.031	-0.025	3500	1.00
	θ	$Exp(1)$	1.624	1.512	1.733	2708	1.00
	species	$N(0, \sigma_s)$	0.408	0.302	0.501	1872	1.00
	genus	$N(0, \sigma_G)$	0.650	0.184	1.085	830	1.00
	region	$N(0, \sigma_d)$	0.282	0.049	0.532	737	1.00
	$\sigma_s, \sigma_G, \sigma_d$	$Cauchy(0, 1)$					
Scraping bite area	A	$N(4.45, 5)$	2.459	2.354	2.568	1182	1.00
	B	$N(0, 2)$	0.060	0.057	0.062	1052	1.00

Table S2 | Bayesian priors and model convergence indicators for feeding rate models (Eqs 1,2, 4-7). Priors are weakly informative, except for intercept priors which were set at the mean bite rate or bite area (on a log scale). Parameter symbols are defined in Eqs. 4 – 7, and θ is the scale parameter for the Gamma distribution. $N(0, 10)$ is a normal distribution with mean = 0 and standard deviation = 10, $Cauchy(0, 1)$ is a Cauchy distribution with location = 0 and scale = 1. Estimates for random effect variances not shown.

Cropping

			Standardized effect si
Intercept	**Fished.Protected.dummy**	**Fished.Unfished.dummy**	**complexity**
0.410	-	-	0.087
0.431	-	-	0.092
0.396	0.125	-	0.093
0.417	0.132	-	0.099
0.328	-	0.526	0.091
0.432	-	-	0.087
0.345	-	0.336	0.087
0.310	0.134	0.534	0.097
0.432	-	-	0.094
0.451	-	-	0.093
0.404	-	-	-
0.328	0.127	0.353	0.093
0.454	-	-	0.100
0.415	0.116	-	0.093
0.419	0.125	-	0.101
0.439	0.132	-	0.106
0.341	-	0.577	0.091
0.431	0.125	-	0.099
0.351	-	0.530	0.099
0.361	-	0.385	0.087
0.428	-	-	-
0.455	-	-	0.095
0.367	-	0.342	0.095
0.392	0.096	-	-
0.333	0.134	0.538	0.105
0.322	0.125	0.573	0.097
0.324	-	0.530	-
0.343	-	0.320	-
0.426	-	-	-
0.474	-	-	0.100
0.341	0.117	0.390	0.093
0.350	0.127	0.359	0.101
0.399	-	-	-
0.416	0.102	-	-
0.437	0.116	-	0.100
0.365	-	0.582	0.099
0.454	0.125	-	0.106
0.309	0.106	0.539	-
0.447	-	-	-
0.382	-	0.391	0.095
0.328	0.099	0.334	-

0.423	-	-	-
0.412	0.087	-	-
0.358	-	0.368	-
0.338	-	0.582	-
0.345	0.125	0.577	0.105
0.386	0.096	-	-
0.321	-	0.530	-
0.339	-	0.319	-
0.363	0.117	0.397	0.101
0.420	-	-	-
0.432	0.095	-	-
0.409	0.102	-	-
-0.586	0.172	0.786	0.090
-0.560	-	0.784	0.084
-0.496	-	0.903	-
-0.506	-	0.886	0.069
-0.467	0.172	-	0.092
-0.526	0.154	0.874	0.075
-0.562	0.179	0.768	0.087
-0.518	0.138	0.896	-
-0.443	-	-	0.086
-0.562	-	0.792	-
-0.367	-	-	0.071
-0.600	0.178	0.908	0.094
-0.394	0.157	-	0.077
-0.446	0.180	-	0.089
-0.585	0.156	0.794	-
-0.537	-	0.769	0.079
-0.360	-	-	-
-0.555	0.169	0.783	0.100
-0.590	0.172	0.783	0.088
-0.572	-	0.892	0.087
-0.584	0.165	0.813	0.088
-0.527	-	0.781	0.094
-0.383	0.140	-	-
-0.422	-	-	0.081
-0.559	-	0.825	0.083
-0.558	0.160	0.774	-
-0.535	-	0.774	-
-0.445	-	-	-
-0.511	-	0.985	0.071
-0.539	0.160	0.985	0.078
-0.568	-	0.784	0.082

-0.467	0.155	-	-
-0.435	0.169	-	0.102
-0.461	0.176	-	0.095
-0.576	0.185	0.894	0.091
-0.503	-	0.985	-
-0.474	-	0.889	0.078
-0.469	0.173	-	0.090
-0.499	-	0.901	-
-0.504	-	0.884	0.067
-0.408	-	-	0.096
-0.502	0.151	0.874	0.083
-0.489	-	0.906	-
-0.462	0.167	-	0.091
-0.528	0.154	0.869	0.074
-0.445	-	-	0.084
-0.528	0.142	0.987	-
-0.438	-	-	0.088
-0.436	-	-	0.085
-0.566	0.178	0.765	0.085
-0.443	0.160	-	-
-0.421	-	-	-
-0.539	0.176	0.767	0.094
-0.561	-	0.836	-
-0.520	0.138	0.891	-
-0.564	-	0.789	-
-0.549	-	0.878	0.083
-0.596	0.160	0.890	-
-0.572	-	0.877	-
-0.569	0.175	0.905	0.103
-0.439	0.184	-	0.092
-0.587	0.156	0.791	-
-0.338	-	-	0.080
-0.602	0.178	0.908	0.092
-0.447	0.180	-	0.087
-0.391	0.162	-	0.080
-0.514	0.137	0.896	-
-0.583	0.149	0.829	-
-0.364	-	-	0.073
-0.370	-	-	0.070
-0.369	0.154	-	0.085
-0.421	0.178	-	0.096
-0.397	0.158	-	0.076
-0.513	-	0.767	0.088

	-0.543	-	0.767	0.078
	-0.550	0.169	0.780	0.099
	-0.363	-	-	-
	-0.553	-	0.791	-
	-0.598	0.171	0.929	0.092
	-0.539	-	0.888	0.098
	-0.357	-	-	-
	-0.579	0.155	0.794	-
	-0.523	-	0.826	0.095
	-0.523	-	0.777	0.094
	-0.550	0.161	0.815	0.099
	-0.352	-	-	-
	-0.571	-	0.925	0.086
	-0.574	-	0.892	0.085
	-0.448	-	-	-
	-0.570	0.164	0.876	-
	-0.424	-	-	0.080
	-0.416	-	-	0.084
	-0.395	-	-	0.090
	-0.470	0.156	-	-
	-0.431	0.169	-	0.101
	-0.387	0.141	-	-
	-0.545	-	0.862	-
	-0.537	-	0.771	-
	-0.560	0.160	0.771	-
	-0.438	-	-	-
	-0.589	0.166	0.809	0.087
	-0.380	0.142	-	-
	-0.430	0.173	-	0.104
	-0.566	-	0.821	0.081
	-0.465	0.176	-	0.093
	-0.484	-	0.985	0.080
	-0.442	-	-	-
	-0.405	-	-	0.096
	-0.578	0.185	0.894	0.089
	-0.463	0.157	-	-
	-0.461	0.150	-	-
Scraping	-0.515	0.157	0.985	0.086
	-0.380	0.140	-	-
	-0.437	-	-	-
	-0.513	-	0.983	0.070
	-0.506	-	0.983	-
	-0.542	0.161	0.982	0.077

-0.472	-	0.883	0.078
-0.553	0.183	0.892	0.098
-0.427	0.162	-	0.101
-0.397	-	-	0.097
-0.532	-	0.773	-
-0.558	0.160	0.774	-
-0.458	0.171	-	0.094
-0.464	0.154	-	-
-0.403	-	-	0.099
-0.465	0.168	-	0.089
-0.446	0.160	-	-
-0.496	-	0.985	-
-0.424	-	-	-
-0.440	-	-	0.086
-0.486	-	0.901	-
-0.500	0.151	0.869	0.083
-0.439	-	-	0.083
-0.433	-	-	0.087
-0.535	0.175	0.763	0.094
-0.563	-	0.831	-
-0.599	0.161	0.892	-
-0.574	-	0.878	-
-0.531	0.143	0.985	-
-0.570	-	0.914	-
-0.437	0.162	-	-
-0.565	0.175	0.905	0.103
-0.416	-	-	-
-0.441	0.184	-	0.090
-0.551	-	0.879	0.081
-0.524	-	0.876	0.091
-0.594	0.154	0.919	-
-0.337	-	-	0.080
-0.418	-	-	-
-0.419	0.176	-	0.096
-0.585	0.150	0.823	-
-0.444	0.160	-	-
-0.549	-	0.836	-
-0.525	0.142	0.987	-
-0.548	-	0.787	-
-0.508	-	0.763	0.088
-0.417	0.183	-	0.100
-0.335	-	-	0.082
-0.564	0.167	0.930	0.103

-0.535	-	0.888	0.097
-0.563	-	0.875	-
-0.365	0.158	-	0.087
-0.394	0.162	-	0.079
-0.368	0.154	-	0.085
-0.574	0.155	0.789	-
-0.512	0.137	0.891	-
-0.367	-	-	0.072
-0.591	0.159	0.889	-
-0.534	-	0.927	0.098
-0.600	0.172	0.927	0.091
-0.519	-	0.821	0.094
-0.574	0.148	0.830	-
-0.551	0.161	0.813	0.099
-0.360	-	-	-
-0.392	-	-	0.090
-0.572	0.165	0.877	-
-0.351	-	-	-
-0.441	-	-	-
-0.547	-	0.864	-
-0.426	0.172	-	0.104
-0.418	-	-	0.082
-0.573	-	0.923	0.084
-0.390	-	-	0.092
-0.444	-	-	-
-0.350	-	-	-
-0.466	0.158	-	-
-0.432	-	-	-
-0.464	0.151	-	-
-0.384	0.144	-	-
-0.436	-	-	-
-0.459	0.154	-	-
-0.399	-	-	0.098
-0.394	-	-	0.096
-0.424	0.162	-	0.100
-0.425	-	-	-
-0.482	-	0.982	0.081
-0.570	0.164	0.876	-
-0.528	-	0.770	-
-0.550	0.181	0.892	0.098
-0.542	-	0.862	-
-0.379	0.140	-	-
-0.423	0.166	-	0.104

-0.554	0.159	0.770	-
-0.458	0.152	-	-
-0.513	0.157	0.982	0.086
-0.461	0.172	-	0.092
-0.394	-	-	0.099
-0.433	-	-	-
-0.378	0.142	-	-
-0.453	0.148	-	-
-0.458	0.156	-	-
-0.440	0.163	-	-
-0.495	-	0.982	-
-0.418	-	-	-
-0.435	-	-	0.086
-0.572	-	0.912	-
-0.520	-	0.876	0.091
-0.414	-	-	-
-0.441	0.159	-	-
-0.597	0.155	0.917	-
-0.544	-	0.830	-
-0.413	0.180	-	0.098

ze

fish.biom	hard.coral	macroalgae	rubble	substrate	df	logLik
0.101	-	-0.174	-	0.212	8	-95.276
-	-	-0.176	-	0.213	7	-96.660
0.097	-	-0.173	-	0.215	9	-94.472
-	-	-0.174	-	0.216	8	-95.773
-	-	-0.173	-	0.211	8	-95.954
0.104	-	-0.172	0.040	0.239	9	-94.984
0.087	-	-0.173	-	0.211	9	-94.998
-	-	-0.171	-	0.214	9	-95.034
0.101	-0.036	-0.179	-	0.210	9	-95.121
-	-	-0.174	0.035	0.236	8	-96.438
0.112	-	-0.176	-	0.220	7	-97.621
0.082	-	-0.171	-	0.214	10	-94.165
-	-0.037	-0.181	-	0.210	8	-96.499
0.100	-	-0.171	0.031	0.235	10	-94.300
0.097	-0.035	-0.178	-	0.213	10	-94.316
-	-0.036	-0.179	-	0.214	9	-95.612
-	-	-0.170	0.042	0.239	9	-95.640
-	-	-0.173	0.026	0.233	9	-95.656
-	-0.039	-0.179	-	0.209	9	-95.768
0.089	-	-0.169	0.045	0.240	10	-94.643
-	-	-0.179	-	0.222	6	-99.279
0.104	-0.036	-0.177	0.041	0.237	10	-94.825
0.086	-0.037	-0.178	-	0.208	10	-94.829
0.110	-	-0.175	-	0.221	8	-97.163
-	-0.039	-0.177	-	0.212	10	-94.842
-	-	-0.169	0.033	0.235	10	-94.850
-	-	-0.175	-	0.221	7	-98.467
0.096	-	-0.174	-	0.219	8	-97.342
0.115	-	-0.174	0.039	0.246	8	-97.356
-	-0.037	-0.179	0.036	0.234	9	-96.274
0.084	-	-0.168	0.036	0.236	11	-93.944
0.082	-0.037	-0.176	-	0.211	11	-93.991
0.112	0.008	-0.175	-	0.221	8	-97.612
-	-	-0.178	-	0.224	7	-98.770
0.100	-0.036	-0.176	0.032	0.233	11	-94.142
-	-0.040	-0.176	0.043	0.237	10	-95.446
-	-0.036	-0.178	0.026	0.231	10	-95.493
-	-	-0.173	-	0.222	8	-97.907
-	-	-0.177	0.034	0.245	7	-99.083
0.088	-0.037	-0.175	0.045	0.238	11	-94.468
0.093	-	-0.173	-	0.220	9	-96.853

-	0.009	-0.177	-	0.223	7	-99.268
0.113	-	-0.173	0.033	0.243	9	-96.981
0.098	-	-0.171	0.044	0.248	9	-97.009
-	-	-0.172	0.043	0.249	8	-98.165
-	-0.040	-0.174	0.033	0.233	11	-94.652
0.110	0.010	-0.173	-	0.222	9	-97.150
-	0.005	-0.174	-	0.221	8	-98.464
0.096	0.007	-0.173	-	0.219	9	-97.336
0.083	-0.038	-0.173	0.036	0.234	12	-93.766
0.115	0.009	-0.172	0.039	0.246	9	-97.346
-	-	-0.176	0.027	0.241	8	-98.646
-	0.011	-0.176	-	0.224	8	-98.753
-	-	-	-	-	7	-40.220
-	-	-	-	-	6	-41.541
-	-	-	0.123	0.134	7	-40.453
-	-	-	0.112	0.121	8	-39.345
-	-	-	-	-	6	-41.635
-	-	-	0.099	0.115	9	-38.276
-	-	-	-	0.052	8	-39.580
-	-	-	0.114	0.130	8	-39.603
-	-	-	-	-	5	-42.952
-	-	-	-	-	5	-43.122
-	-	-	0.105	0.118	7	-40.912
-0.076	-	-	-	-	8	-39.780
-	-	-	0.093	0.114	8	-39.799
-	-	-	-	0.054	7	-40.947
-	-	-	-	-	6	-42.063
-	-	-	-	0.048	7	-41.005
-	-	-	0.117	0.132	6	-42.125
-	-0.048	-	-	-	8	-39.964
-	-	-0.028	-	-	8	-40.044
-0.068	-	-	-	-	7	-41.201
-	-	-	0.022	-	8	-40.113
-	-0.051	-	-	-	7	-41.247
-	-	-	0.109	0.129	7	-41.250
-	-	-	-	0.050	6	-42.386
-	-	-	0.032	-	7	-41.307
-	-	-	-	0.056	7	-41.329
-	-	-	-	0.055	6	-42.449
-	-	-	-	-	4	-44.641
-0.062	-	-	0.108	0.119	9	-39.064
-0.072	-	-	0.095	0.114	10	-37.897
-	-	-0.028	-	-	7	-41.369

-	-	-	-	-	5	-43.605
-	-0.048	-	-	-	7	-41.392
-0.056	-	-	-	-	7	-41.395
-0.079	-	-	-	0.053	9	-39.115
-0.051	-	-	0.121	0.134	8	-40.269
-	-0.043	-	0.113	0.117	9	-39.133
-	-	-0.031	-	-	7	-41.424
-	-	-0.023	0.120	0.133	8	-40.334
-	-	-0.020	0.109	0.119	9	-39.249
-	-0.053	-	-	-	6	-42.660
-	-0.038	-	0.100	0.112	10	-38.109
-	-0.012	-	0.124	0.134	8	-40.435
-	-	-	0.016	-	7	-41.574
-	-	-0.021	0.097	0.114	10	-38.173
-	-	-0.030	-	-	6	-42.748
-0.059	-	-	0.112	0.130	9	-39.357
-0.049	-	-	-	-	6	-42.777
-	-	-	0.027	-	6	-42.787
-	-	-0.028	-	0.052	9	-39.407
-	-	-	-	0.058	6	-42.812
-	-	-	-	0.056	5	-43.927
-	-0.037	-	-	0.048	9	-39.429
-	-	-	0.035	-	6	-42.857
-	-	-0.024	0.111	0.128	9	-39.473
-	-	-0.033	-	-	6	-42.892
-0.070	-	-	-	0.049	8	-40.655
-0.061	-	-	-	-	7	-41.798
-0.054	-	-	-	-	6	-42.918
-0.076	-0.047	-	-	-	9	-39.532
-0.059	-	-	-	0.054	8	-40.686
-	-	-0.033	-	-	7	-41.821
-	-0.044	-	0.106	0.114	8	-40.706
-0.078	-	-0.030	-	-	9	-39.580
-	-	-0.030	-	0.054	8	-40.738
-0.053	-	-	0.091	0.113	9	-39.591
-	-0.006	-	0.115	0.130	9	-39.599
-	-	-	0.028	-	7	-41.892
-0.045	-	-	0.103	0.117	8	-40.768
-	-	-0.023	0.102	0.116	8	-40.789
-	-0.038	-	0.094	0.111	9	-39.646
-	-0.036	-	-	0.050	8	-40.810
-	-	-0.024	0.090	0.112	9	-39.665
-	-0.041	-	-	0.044	8	-40.817

-	-	-0.028	-	0.048	8	-40.832
-	-0.057	-0.036	-	-	9	-39.686
-	-	-0.026	0.114	0.130	7	-41.976
-	-0.015	-	-	-	6	-43.095
-0.074	-	-	0.019	-	9	-39.699
-0.068	-0.051	-	-	-	8	-40.911
-0.033	-	-	0.116	0.132	7	-42.047
-	-0.010	-	-	-	7	-42.051
-	-0.057	-	0.036	-	8	-40.950
-	-0.062	-0.036	-	-	8	-40.965
-	-0.051	-	0.026	-	9	-39.818
-	-0.011	-	0.118	0.132	7	-42.111
-0.065	-	-	0.030	-	8	-40.995
-0.070	-	-0.030	-	-	8	-41.008
-	-	-0.036	-	-	5	-44.374
-0.065	-	-	-	0.057	8	-41.032
-	-	-0.031	-	0.050	7	-42.180
-0.051	-	-	-	0.050	7	-42.202
-	-0.042	-	-	0.045	7	-42.203
-	-	-0.036	-	-	6	-43.321
-	-0.058	-0.038	-	-	8	-41.079
-	-	-0.027	0.106	0.126	8	-41.087
-0.057	-	-	-	0.055	7	-42.224
-	-	-0.032	-	0.054	7	-42.224
-	-	-0.033	-	0.056	8	-41.098
-	-	-	0.029	-	5	-44.453
-	-	-0.027	0.020	-	9	-39.955
-0.040	-	-	0.108	0.129	8	-41.138
-0.056	-0.047	-	-	-	8	-41.157
-	-	-0.026	0.030	-	8	-41.162
-0.059	-	-0.032	-	-	8	-41.165
-0.063	-0.044	-	0.110	0.116	10	-38.852
-0.034	-	-	-	-	5	-44.560
-	-0.063	-0.038	-	-	7	-42.347
-0.081	-	-0.029	-	0.053	10	-38.915
-0.040	-	-	-	-	6	-43.491
-	-	-	0.022	-	6	-43.494
-0.072	-0.038	-	0.097	0.111	11	-37.733
-	-0.004	-	0.109	0.129	8	-41.249
-	-0.015	-	-	-	5	-44.617
-0.064	-	-0.022	0.106	0.118	10	-38.953
-0.053	-	-0.024	0.118	0.132	9	-40.135
-0.073	-	-0.023	0.093	0.113	11	-37.774

-	-0.051	-0.027	0.110	0.115	10	-38.970
-0.079	-0.036	-	-	0.049	10	-38.971
-	-0.051	-	0.020	-	8	-41.301
-	-0.057	-	0.031	-	7	-42.442
-	-0.006	-	-	0.054	7	-42.445
-	20214594370	-	-	0.056	8	-41.329
-0.055	-	-	0.014	-	8	-41.347
-	-0.008	-	-	-	6	-43.597
-0.049	-0.052	-	-	-	7	-42.488
-	-	-0.029	0.014	-	8	-41.377
-	-	-0.035	-	0.058	7	-42.540
-0.051	-0.011	-	0.122	0.134	9	-40.254
-	-	-0.035	-	0.056	6	-43.666
-0.051	-	-0.031	-	-	7	-42.557
-	-0.019	-0.026	0.121	0.131	9	-40.291
-	-0.046	-0.027	0.097	0.110	11	-37.942
-	-	-0.028	0.025	-	7	-42.608
-0.047	-	-	0.025	-	7	-42.627
-	-0.046	-0.034	-	0.047	10	-39.175
-	-	-0.030	0.032	-	7	-42.659
-0.063	-	-0.035	-	-	8	-41.529
-0.057	-	-0.034	-	-	7	-42.665
-0.061	-	-0.025	0.109	0.128	10	-39.208
-0.051	-	-	0.033	-	7	-42.676
-0.044	-	-	-	0.059	7	-42.676
-0.078	-0.057	-0.037	-	-	10	-39.226
-0.037	-	-	-	0.057	6	-43.831
-0.062	-	-0.032	-	0.054	9	-40.457
-0.072	-	-0.030	-	0.049	9	-40.458
-0.070	-0.041	-	-	0.044	9	-40.470
-0.058	-	-	0.026	-	8	-41.650
-	-0.052	-0.029	0.103	0.112	9	-40.514
-	-0.006	-	-	0.056	6	-43.924
-	-0.046	-0.036	-	0.049	9	-40.522
-	-	-0.032	0.026	-	8	-41.676
-	0.002	-	-	0.059	7	-42.811
-	-0.019	-	0.036	-	7	-42.812
-0.059	-0.005	-	0.113	0.130	10	-39.354
-	-0.026	-0.037	-	-	7	-42.817
-	-0.051	-0.035	-	0.043	9	-40.553
-0.060	-0.035	-	-	0.051	9	-40.558
-0.045	-0.044	-	0.105	0.114	9	-40.561
-0.074	-0.050	-	0.023	-	10	-39.417

-0.070	-0.062	-0.038	-	-	9	-40.604
-0.054	-0.014	-	-	-	7	-42.895
-0.053	-0.037	-	0.092	0.110	10	-39.441
-0.055	-	-0.025	0.088	0.111	10	-39.441
-	-0.045	-0.029	0.091	0.108	10	-39.449
-	-0.020	-0.037	-	-	8	-41.773
-	-0.013	-0.026	0.112	0.127	10	-39.453
-0.047	-	-0.024	0.100	0.116	9	-40.634
-0.060	-0.008	-	-	-	8	-41.790
-0.065	-0.057	-	0.034	-	9	-40.645
-0.076	-	-0.028	0.017	-	10	-39.516
-	-0.066	-0.034	0.034	-	9	-40.699
-	-0.014	-	0.029	-	8	-41.870
-	-0.061	-0.034	0.024	-	10	-39.560
-0.036	-	-0.026	0.113	0.130	8	-41.889
-	-0.052	-0.037	-	0.044	8	-41.907
-0.067	-	-0.034	-	0.057	9	-40.773
-	-0.018	-0.028	0.115	0.129	8	-41.939
-	-	-0.033	0.027	-	6	-44.216
-0.060	-	-0.034	-	0.055	8	-41.975
-0.059	-0.057	-0.039	-	-	9	-40.823
-0.053	-	-0.032	-	0.050	8	-41.979
-0.067	-	-0.027	0.028	-	9	-40.830
-0.051	-0.042	-	-	0.045	8	-42.022
-0.037	-	-0.036	-	-	6	-44.278
-0.033	-0.011	-	0.117	0.132	8	-42.035
-0.043	-	-0.037	-	-	7	-43.189
-	-0.025	-0.039	-	-	6	-44.305
-	-	-0.035	0.020	-	7	-43.232
-0.042	-	-0.028	0.104	0.126	9	-40.962
-0.032	-	-	0.028	-	6	-44.382
-	-0.019	-0.039	-	-	7	-43.281
-0.051	-0.062	-0.039	-	-	8	-42.156
-	-0.067	-0.036	0.029	-	8	-42.157
-	-0.060	-0.037	0.018	-	9	-41.005
-	-0.019	-	0.030	-	6	-44.414
-0.065	-0.051	-0.029	0.107	0.114	11	-38.669
-0.065	0.001	-	-	0.058	9	-41.032
-	-0.016	-0.035	-	0.053	8	-42.193
-0.081	-0.046	-0.036	-	0.048	11	-38.688
-0.057	-0.005	-	-	0.055	8	-42.221
-	-0.011	-0.028	0.106	0.126	9	-41.073
-0.054	-0.050	-	0.018	-	9	-41.083

-	-0.011	-0.034	-	0.055	9	-41.085
-0.038	-	-	0.021	-	7	-43.391
-0.074	-0.046	-0.029	0.093	0.109	12	-37.543
-0.057	-	-0.031	0.012	-	9	-41.131
-0.047	-0.057	-	0.030	-	8	-42.286
-0.034	-0.014	-	-	-	6	-44.538
-0.040	-0.003	-	0.108	0.129	9	-41.137
-	-0.011	-	0.023	-	7	-43.479
-0.040	-0.007	-	-	-	7	-43.485
-0.047	-	-0.036	-	0.059	8	-42.384
-0.053	-0.019	-0.027	0.119	0.131	10	-40.095
-0.040	-	-0.036	-	0.056	7	-43.554
-0.049	-	-0.030	0.023	-	8	-42.433
-0.053	-	-0.032	0.031	-	8	-42.457
-0.072	-0.051	-0.037	-	0.043	10	-40.179
-	-0.016	-0.037	-	0.054	7	-43.640
-	-0.009	-0.037	-	0.057	8	-42.532
-0.061	-	-0.033	0.023	-	9	-41.408
-	-0.029	-0.035	0.034	-	8	-42.563
-0.062	-0.045	-0.037	-	0.050	10	-40.245

Model diagnostics

AICc	ΔAIC	AIC-weight
207.732	0.000	0.094
208.231	0.499	0.073
208.433	0.701	0.066
208.726	0.994	0.057
209.088	1.357	0.048
209.456	1.724	0.040
209.484	1.752	0.039
209.556	1.825	0.038
209.729	1.998	0.035
210.057	2.325	0.029
210.152	2.420	0.028
210.163	2.432	0.028
210.179	2.448	0.028
210.433	2.702	0.024
210.466	2.735	0.024
210.711	2.980	0.021
210.767	3.035	0.021
210.799	3.068	0.020
211.024	3.292	0.018
211.120	3.388	0.017
211.236	3.504	0.016
211.484	3.753	0.014
211.492	3.760	0.014
211.507	3.775	0.014
211.516	3.785	0.014
211.532	3.801	0.014
211.845	4.114	0.012
211.864	4.133	0.012
211.892	4.160	0.012
212.035	4.303	0.011
212.106	4.375	0.011
212.200	4.469	0.010
212.405	4.674	0.009
212.450	4.718	0.009
212.503	4.771	0.009
212.725	4.994	0.008
212.820	5.089	0.007
212.995	5.263	0.007
213.076	5.345	0.007
213.155	5.423	0.006
213.193	5.462	0.006

213.447	5.716	0.005
213.450	5.719	0.005
213.505	5.774	0.005
213.511	5.779	0.005
213.523	5.792	0.005
213.788	6.057	0.005
214.107	6.376	0.004
214.160	6.429	0.004
214.177	6.446	0.004
214.180	6.448	0.004
214.472	6.741	0.003
214.687	6.956	0.003
95.343	0.000	0.027
95.754	0.412	0.022
95.810	0.467	0.021
95.860	0.517	0.021
95.942	0.599	0.020
96.028	0.686	0.019
96.331	0.988	0.016
96.377	1.034	0.016
96.380	1.037	0.016
96.720	1.378	0.013
96.727	1.384	0.013
96.730	1.388	0.013
96.769	1.426	0.013
96.796	1.454	0.013
96.798	1.455	0.013
96.914	1.571	0.012
96.921	1.579	0.012
97.098	1.756	0.011
97.260	1.917	0.010
97.306	1.963	0.010
97.396	2.053	0.010
97.397	2.054	0.010
97.404	2.062	0.010
97.445	2.102	0.009
97.517	2.174	0.009
97.561	2.218	0.009
97.571	2.228	0.009
97.598	2.255	0.009
97.603	2.260	0.009
97.612	2.269	0.009
97.642	2.300	0.008

97.685	2.343	0.008
97.686	2.344	0.008
97.694	2.351	0.008
97.705	2.363	0.008
97.709	2.367	0.008
97.741	2.399	0.008
97.752	2.409	0.008
97.839	2.496	0.008
97.974	2.632	0.007
97.992	2.650	0.007
98.036	2.693	0.007
98.042	2.699	0.007
98.051	2.708	0.007
98.165	2.822	0.006
98.169	2.826	0.006
98.190	2.847	0.006
98.225	2.883	0.006
98.245	2.903	0.006
98.289	2.946	0.006
98.296	2.953	0.006
98.331	2.988	0.006
98.334	2.991	0.006
98.386	3.043	0.006
98.422	3.080	0.006
98.455	3.113	0.006
98.480	3.138	0.006
98.499	3.156	0.005
98.508	3.165	0.005
98.539	3.196	0.005
98.544	3.201	0.005
98.545	3.203	0.005
98.582	3.239	0.005
98.635	3.293	0.005
98.647	3.304	0.005
98.657	3.315	0.005
98.673	3.330	0.005
98.687	3.344	0.005
98.707	3.364	0.005
98.749	3.407	0.005
98.768	3.426	0.005
98.790	3.448	0.005
98.805	3.462	0.005
98.806	3.463	0.005

98.834	3.491	0.005
98.846	3.504	0.005
98.856	3.513	0.005
98.862	3.520	0.005
98.873	3.531	0.005
98.993	3.650	0.004
98.998	3.655	0.004
99.005	3.663	0.004
99.071	3.729	0.004
99.100	3.757	0.004
99.112	3.770	0.004
99.125	3.782	0.004
99.161	3.818	0.004
99.187	3.844	0.004
99.225	3.882	0.004
99.236	3.893	0.004
99.264	3.921	0.004
99.308	3.965	0.004
99.310	3.967	0.004
99.313	3.971	0.004
99.329	3.986	0.004
99.345	4.002	0.004
99.351	4.008	0.004
99.352	4.009	0.004
99.366	4.023	0.004
99.381	4.039	0.004
99.385	4.042	0.004
99.448	4.105	0.003
99.485	4.143	0.003
99.495	4.152	0.003
99.500	4.157	0.003
99.522	4.179	0.003
99.596	4.254	0.003
99.597	4.254	0.003
99.649	4.306	0.003
99.654	4.312	0.003
99.659	4.316	0.003
99.665	4.323	0.003
99.668	4.326	0.003
99.711	4.368	0.003
99.725	4.382	0.003
99.746	4.403	0.003
99.748	4.405	0.003

99.757	4.415	0.003
99.760	4.417	0.003
99.773	4.430	0.003
99.787	4.444	0.003
99.792	4.450	0.003
99.828	4.485	0.003
99.865	4.522	0.003
99.866	4.523	0.003
99.879	4.536	0.003
99.926	4.583	0.003
99.983	4.640	0.003
99.984	4.641	0.003
100.005	4.662	0.003
100.017	4.674	0.003
100.058	4.716	0.003
100.084	4.741	0.002
100.120	4.777	0.002
100.157	4.815	0.002
100.168	4.825	0.002
100.222	4.880	0.002
100.229	4.886	0.002
100.233	4.890	0.002
100.235	4.892	0.002
100.254	4.912	0.002
100.256	4.913	0.002
100.269	4.927	0.002
100.335	4.992	0.002
100.389	5.046	0.002
100.391	5.048	0.002
100.415	5.073	0.002
100.471	5.128	0.002
100.503	5.161	0.002
100.520	5.177	0.002
100.520	5.178	0.002
100.523	5.180	0.002
100.526	5.183	0.002
100.526	5.184	0.002
100.527	5.184	0.002
100.538	5.195	0.002
100.581	5.238	0.002
100.591	5.248	0.002
100.598	5.256	0.002
100.651	5.309	0.002

100.683	5.341	0.002
100.694	5.351	0.002
100.700	5.357	0.002
100.700	5.358	0.002
100.716	5.373	0.002
100.718	5.375	0.002
100.723	5.381	0.002
100.743	5.401	0.002
100.751	5.408	0.002
100.766	5.423	0.002
100.850	5.508	0.002
100.874	5.531	0.002
100.910	5.567	0.002
100.938	5.595	0.002
100.949	5.606	0.002
100.985	5.643	0.002
101.021	5.678	0.002
101.049	5.706	0.002
101.103	5.761	0.001
101.121	5.778	0.001
101.122	5.779	0.001
101.128	5.785	0.001
101.136	5.794	0.001
101.214	5.871	0.001
101.229	5.886	0.001
101.240	5.897	0.001
101.280	5.938	0.001
101.283	5.940	0.001
101.368	6.025	0.001
101.399	6.056	0.001
101.436	6.093	0.001
101.466	6.123	0.001
101.483	6.140	0.001
101.485	6.143	0.001
101.486	6.143	0.001
101.499	6.157	0.001
101.538	6.195	0.001
101.540	6.197	0.001
101.557	6.215	0.001
101.576	6.233	0.001
101.613	6.270	0.001
101.621	6.279	0.001
101.641	6.298	0.001

101.645	6.302	0.001
101.685	6.342	0.001
101.707	6.365	0.001
101.737	6.394	0.001
101.744	6.401	0.001
101.749	6.406	0.001
101.750	6.408	0.001
101.861	6.518	0.001
101.874	6.531	0.001
101.938	6.596	0.001
102.008	6.665	0.001
102.012	6.669	0.001
102.037	6.694	0.001
102.085	6.743	0.001
102.176	6.834	0.001
102.182	6.840	0.001
102.234	6.891	0.001
102.292	6.949	0.001
102.296	6.954	0.001
102.308	6.965	0.001