Linking social and ecological systems to sustain coral reef fisheries

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Running title: Coral reef social-ecological systems
Summary
The ecosystem goods and services provided by coral reefs are critical to the social and economic welfare of hundreds of millions of people, overwhelmingly in developing countries [1]. Widespread reef degradation is severely eroding these goods and services, but the socio-economic factors shaping the ways that societies use coral reefs are poorly understood [2]. We examine relationships between human population density, a multidimensional index of socio-economic development, reef complexity, and the condition of coral reef fish populations in five countries across the Indian Ocean. In fished sites, fish biomass was negatively related to human population density, but was best explained by reef complexity and a U-shaped relationship with socio-economic development. The biomass of reef fishes was 4-times lower at intermediate levels of economic development, compared to locations with both low and high development. In contrast, average biomass inside fisheries closures was 3-times higher than fished sites, and not associated with socio-economic development. Sustaining coral reef fisheries requires an integrated approach that uses tools such as protected areas to quickly build reef resources, while also building capacities and capital in societies over longer time frames to address the complex underlying causes of reef degradation.

Results and Discussion
Effectively confronting the coral reef crisis will require linking social and ecological systems to better understand and address the complex socio-economic drivers that influence how societies use and ultimately govern their use of coral reefs [2-3]. It is generally held that human use, driven primarily by population density, is a principal cause of coral reef degradation [4-7]. However, less is known about how other socio-economic factors such as economic development shape society’s impacts on coral reefs [8-9]. Sociological perspectives on human-environment interactions emphasize how socio-economic development can affect a societies’ impact on the environment, often in non-linear and sometimes positive ways [10-11]. To explore these linkages in coral reef fisheries, we collected data on a composite index of village-level infrastructure (as a proxy for local-scale socio-economic development), human population density, and structural complexity of reef habitat (rugosity) in 19 fished sites and 11 fisheries closures.

across five countries in the western Indian Ocean. We evaluated these drivers’ influence on the biomass of reef fishes, which is a variable sensitive to management and human impact [12].

Firstly, we examined whether the biomass of reef fishes targeted in the multi-species fishery could be explained independently by human population density, structural complexity and socio-economic development. In fished sites, human population numbers had a significant but weak negative relationship to the biomass of target reef fishes (n=19, $r^2=0.28$, $p=0.02$, Fig. 1a) and the benthic structural complexity had a moderate positive relationship (n=16, $r^2=0.54$, $p=0.001$, Fig. 1b), consistent with previous studies on reef fishes [4,7, 13-14]. Our novel finding is that the strongest relationship to fish biomass was the quadratic function of the socio-economic development index, which displayed a U-shaped relationship (n=19, $r^2=0.77$, $p<0.001$, Fig. 1c).

Secondly, we tested candidate models with all possible combinations of the three factors to determine the best combination of variables for explaining fish biomass in fished sites. Country was included as a random effect to account for non-independence of samples within countries [15]. A key and surprising finding from this study is that the best model included the quadratic socio-economic development index and reef structural complexity, but did not include human population density (likelihood ratio test of nested models with and without this term; ratio = 0.166, $p=0.684$). The quadratic term of the development index was highly significant in the selected model (likelihood ratio = 14.5, $p<0.001$). Thus, fish biomass is highest where community development is very low or high, but low where development is intermediate (Fig. 1c). Fish biomass at the bottom of the curve (Takaungu, Kenya) was $77 \pm 11.9$ kg/ha, approximately 1/4 of the biomass of the sites with the highest and lowest levels of development ($336 \pm SE 52$ kg/ha for Anse Volbert, Seychelles and $294 \pm SE 57.3$ kg/ha for Ambodilaitry, Madagascar, respectively) (Fig 1c).

These findings are consistent with the environmental Kuznets curve hypothesis, which predicts that increasing socio-economic development results in ecological degradation.
until a point when environmental conditions improve as societies become increasingly affluent and begin to demand environmental quality (creating a U shaped relationship between affluence and local environmental conditions) [10, 16-17]. The causal mechanisms behind a Kuznets curve relationship are generally classed in three broad categories: 1) a technique effect, whereby societies may change the technologies used to produce goods and services, which may have differing levels of impact on the environment; 2) a composition effect, whereby the composition of the economy could change to be less destructive to the local environment, for example switching from primary resource extraction to a service industry; and 3) a scale effect, whereby wealthier societies displace local impacts, for example, by drawing resources from other areas, often those poorer or less regulated [16, 18]. The parallel sociological perspective of ecological modernization, suggests that it is not economic development, per se, that leads changing environmental conditions, but rather the accompanying institutional changes, such as investments in scientific and natural-resource management organizations [19].

We used socio-economic survey data from these communities to further examine how a combination of the technique, composition, and scale effects, and also aspects of local socio-cultural institutions may play a role in our observation of a Kuznets relationship for coral reef fishes in the western Indian Ocean (Table 2). Sites with low levels of development are characterized by high levels of dependence on fishing as a primary occupation, minimal engagement in salaried employment, and few boats with engines (Table 2, Fig 2). Although these low development sites tend to have weak national governments [20], the presence of customary socio-cultural institutions, such as taboos, may act to restrict fishing effort (although this later indicator was only suggestive at \( p=0.054 \), Table 2). Together, these factors suggest that in low development sites, technological constraints and social institutions may limit people’s exploitation of marine resources. Reduced dependence on marine resources, variable access to boats but increasing access to engines and other technologies, high use of spear guns, and a lack of customary management institutions characterize communities with intermediate levels of development (Table 2, Fig. 2). Factors such as reduced dependence on marine resources and increased technological efficiency can break down customary socio-cultural
institutions that may be critical in managing marine resources [21]. For example, in Kenya, which has some sites with the poorest fishery condition, customary institutions were once widespread, but have largely broken down in recent years [22], with destructive fishing techniques now practiced in some of these locations [6]. Sites with high socio-economic development are generally characterized by effective national government [20], low dependence on fishing, reduced use of potentially damaging gear such as gill nets and higher use of more benign gear such as reef handlines, high levels of engagement in salaried employment, and high levels of access to boats with engines that allow for fishing further afield (Table 2, Fig. 2).

The role of fisheries closures

Fisheries closures can help to sustain reef fisheries by increasing fish biomass within their boundaries, protecting corals and other habitat for reef fishes from damage caused by uses such as destructive fishing practices, and providing ‘spillover’ of adult fishes close to reserve boundaries (generally <500m) [23]. Fisheries closures exist along the full socio-economic development gradient of our study sites and, on average, have approximately 3-times the fish biomass of fished sites, with the difference between the lowest biomass in fished sites and the highest in a closure (~1200 kg/ha) being ~16-fold (both of which were in Kenya) (Fig. 3). Variation in the biomass of fishes within closures can be partially attributed to differences in park compliance, buffer zones, closure size, and age [12, 24-25]. Importantly, there is no clear relationship between biomass in closures and the gradient of development, suggesting that effective marine parks are not just a measure of community affluence [3]. This context suggests that while community development can result in modest variation of fish resources, improvements in fish biomass may be derived from local governance such as well-enforced fisheries closures at most stages of socio-economic development. The poor relationship between development and fish biomass in closures (Fig. 3) suggest that other factors such as social capital, organization, and governance are important elements of successful closures [3, 6].
Although fish biomass was considerably higher inside most fisheries closures, closures alone are unlikely to sustain coral reef fisheries throughout the region. This is in part because they cover too small an area to maintain system-wide resilience, with the current spatial extent of closures in the region ranging from 0.5-15% of the total reef area per country [6]. Following large-scale disturbances such as the 1998 coral bleaching event, the small and dispersed fisheries closures in the western Indian Ocean were not able to prevent declines in key components of reef ecosystems (e.g. reef structural complexity and small-bodied herbivores) or promote faster recovery than fished areas [26]. Vastly expanding the area covered by fisheries closures may promote system-wide resilience to some disturbances, for example by improving ecosystem connectivity and enhancing the biomass of key herbivorous fish groups [27]. However, significant closed area expansion is likely to be met with considerable resistance from stakeholders and in many cases is socially and politically unrealistic. There is clearly a need to develop management strategies that foster resilience throughout the entire seascape, not just inside protected areas [2, 28].

An Integrated approach necessary to sustain coral reef fisheries
Sustaining coral reef fisheries will require moving towards an integrated social-ecological systems approach that better understands and incorporates the socio-economic factors that shape the ways that societies interact with reefs [29]. By linking social science and ecology at a regional scale, this study provides a novel contribution to our understanding of how societies’ socio-economic conditions can influence reef fisheries. In regions such as East Africa, where persistent poverty is often coupled with resource degradation [3, 21, 30-31], improving human welfare and institutional capacities will be an essential component of sustaining broader coral reef seascapes. Escaping these so called “poverty traps” [30-31] will require governments and donors involved in the management of reefs to make meaningful investments in programs that improve governance, build social and physical infrastructure, address burgeoning population growth rates, and provide alternatives to heavy reliance on reef-based livelihoods [3, 32]. From the findings of this research, we suggest prioritization should be given to: 1) assisting low development sites to navigate the transition to improved welfare without dwelling in the intermediate
development stage where resources are likely to be most degraded; and 2) improving environmental conditions and welfare in intermediate development sites in ways that do not use the extraction of reef resources as a major basis of development.

Efforts to improve human welfare in a reef governance context will likely be ineffective and sometimes even counter-productive unless they are coupled with effective policies and governance, for two key reasons. Firstly, relying on the assumption that resource conditions will improve with socio-economic development does not account for potentially irreversible change in coral reef ecosystems [33]. Irreversible change may occur as a result of the heavy degradation at the bottom of the curve and prevent a rebound of fishery resources as development increases [16]. Policy tools such as closures will be critical in helping sustain fisheries and preventing these local ecological phase-shifts, particularly for sites with transitioning economies. Along with closures, there is a need to identify successful aspects of fisheries management from sites that sit along the low or high development sites and determine whether and how such measures might be applicable to other areas, particularly intermediate societies. Such policies may involve fostering or restoring traditional values and institutions [21], instituting property rights [34], switching to fishing practices that exploit different and more sustainable resources, or implementing restrictions on gear types that cause habitat damage [6].

Secondly, aspects of economic growth can contribute to larger-scale degradation of reef ecosystems. As societies become more affluent, they are able to extract resources from further a field [16, 35] and they contribute increasingly to larger-scale and more complex problems confronting reefs, such as coastal modification (e.g. dredging and land reclamation), land-based pollution (e.g. incorporating pesticides and fertilizers in agriculture), and high carbon emissions [10-11]. To minimize the potential negative effects of economic growth on reef systems, socio-economic development needs to be coupled with effective legislation, institutional strengthening, and regional agreements. For example, in Kenya, recent Beach Management Unit legislation provides a form of property rights to coastal fishers, which essentially restricts their ability to fish in distant fishing grounds and simultaneously provides incentives for stewardship of local
resources. At a national level, this type of legislation may help to prevent more distant ecosystems from becoming degraded when there are improvements in local welfare. At a larger scale, multilateral agreements may be required that discourage wealthier countries from consuming the nearshore fishery resources of the poor. Furthermore, governments and donor agencies should make sustainability a cornerstone of development programs, so projects that aim to improve human welfare as part of reef management, do not inadvertently result in increasing contributions to larger-scale threats to coral reefs [11].

These economic and policy approaches for sustaining coral reefs and associated fisheries operate on different, but complimentary, spatial and temporal scales. Policy approaches such as closures can operate on relatively fast temporal scales, with initial responses in fish populations detectable within 3-5 years [36], but their effects are highly localized. Protected areas may provide a lifeline to threatened fisheries regardless of societal trajectory, but there is also a need to govern the entire seascape, particularly with increased occurrence of global threats, such as coral bleaching, which can undermine reef systems both inside and outside protected areas [27-28]. Conversely, socio-economic development that reduces reliance on reef resources may take decades or generations, but is likely to influence how resources are used throughout a society’s entire fishing grounds, which are often much larger than protected areas in the region [37]. Sustaining coral reef fisheries will require using policy responses such as closures to build resources locally while simultaneously addressing key socio-economic drivers of decline to confront both local and larger-scale drivers of reef degradation.

**Experimental Procedures**

**Socio-economic Field Studies**

**Study sites**

We studied 19 coastal communities and adjacent coral reef sites in the western Indian Ocean spanning five countries: Kenya, Tanzania, Seychelles, Mauritius, and Madagascar. Study sites were selected to provide a gradient of economic development and human population density both within and between countries. At each site we investigated the following socio-economic indicators: community-level infrastructure (as a measure of
economic development); human population density; the proportion of the community involved in fishing (and that ranked it as their primary livelihood strategy); the proportion engaged in salaried employment; the proportion of fishers that use gillnets, reef handlines, spearguns, traps, small seine nets, and pelagic gear; the proportion of fishers that own boats and engines, and the presence of customary socio-cultural institutions such as taboos that may restrict fishing.

**Population density**

Population density data was collected using the Socioeconomic Data and Applications Center (SEDAC) grided population of the world database (available Online http://sedac.ciesin.org/gpw/global.jsp). Geographic coordinates of field sites were overlaid on the grided population database. When a field site was near the border of two grids, those grids were averaged to give a mean population density. Grid cells were 4.66 km$^2$.

**Community-level development**

To measure community-level development, we recorded the presence of 16 community-scale infrastructure items [38] in each community by interviewing community leaders and triangulating results with direct observation. We ran Factor Analysis on the presence or absence of infrastructure items to reduce these 16 items into a scale of socio-economic development. This resulted in one factor that explained 51% of the variance [3]. The marginal variance explained by the subsequent factor was low (11%), so only the first factor was extracted. Factor loadings for the specific items were: hard top road = 0.893, phone service= 0.865, restaurant = 0.865, electric service = .0842, piped water = 0.831, public transportation = 0.802, fuel station = 0.758, food market = 0.735, doctor =0.734, hotel = 0.695, septic tanks = 0.665, secondary school =0.662, hospital = 0.506, primary school = 0.498, medical clinic = 0.457, sewage treatment = 0.384. We used the subsequent factor scores for each community as a measure of community-level socio-economic development. Because the Kuznets curve predicts a U-shaped relationship between affluence and environmental conditions, this economic development index was included in regression models as a second-order polynomial.
Resource use, dependence, and governance indicators
To investigate potential causal mechanisms related to the observed environmental Kuznets curve, we conducted more detailed socio-economic assessments in each site. We conducted 1412 household surveys in the 19 fished sites. Sampling of households within villages was based on a systematic design, where a fraction of every $i^{th}$ household (e.g. 2\textsuperscript{nd}, 3\textsuperscript{rd}, 4\textsuperscript{th}) was determined by dividing the total village population by the sample size [39]. There were 23-143 surveys conducted per site, depending on the population of the village. We examined dependence on fishing and salaried employment (e.g. teaching, government work, etc.) by asking respondents to list the jobs people in the household engaged in for food or money. We then asked respondents to rank these activities in order of importance. Fishers were asked about the type of boat and gear they used to determine the following indicators: proportion of fishers with boats, proportion of fishers with boats that have engines, and the type of gear used by fishers. In sites with few fishermen, additional systematic surveys were conducted from the population of fishers [3]. We also examined the presence of sociocultural institutions such as taboos that may help manage marine resources using data in [40].

Ecological Field Studies
Study sites
We collected ecological data from a total of 30 locations: 19 fished sites and 11 fisheries closures. Field sites were selected to be as similar as possible in terms of reef structure, depth, and a dominance of a hard bottom substratum [27]. All sites were located on shallow reef lagoons and slopes on fringing reefs (<7 m depth). When sampling protected areas, sites were located in the centre of the closures.

Reef fish biomass
Biomass of fishes (kg/ha) was selected as an indicator of the condition of reef fish assemblages and treated as the response variable in regressions. Fish biomass is a sensitive indicator of fishing pressure in these multi-species fisheries, which is the dominant local human impact on fish communities in the region [41]. Biomass was based
on fishes > 10cm in length, from diurnally active, non-cryptic families that were extensively surveyed across all sites. Data on fish biomass was collected using underwater visual census by two experienced observers (T.R. McClanahan and N.A.J. Graham) whose detection ability is very similar [42]. All diurnally active, non-cryptic, reef-associated fishes were identified to family or species level and counted, and their size was estimated to 5 or 10 cm intervals at each site. In Kenya, Tanzania, Mauritius, and Madagascar, three to five 100 m × 5 m belt transects were used to count and estimate the numbers and size of fishes [43]. In Seychelles, sixteen 7-m radius point counts were completed at each of 3 sites within each closure [14]. In both methods double counting was avoided by observers disregarding individuals that left the survey boundary and re-entered. Both methods covered a similar area of reef per site (~2000m²) and data were standardised to kg/ha. There may be small amounts of variation associated with different survey techniques and habitats, however methods papers have found little difference between strip transects and point counts in estimating fish abundance [44], and all sites were in shallow fringing reef habitats. Wet weight (biomass) was estimated from the individual fish length data using length-weight relationships for species or families [45].

**Reef complexity**

We also examined habitat rugosity and a nominal term for country to account for two potentially confounding factors. Rugosity, or the topographic complexity of the reef substratum, has been associated with the biomass of reef-associated fish [46-47]. At each site, 5-16 replicate measures of rugosity were calculated by measuring the linear distance covered by 10-m lengths of chain or weighted rope fitted to the contour of the reef surface [13]. Rugosity was, however, only available for 16 of the 19 field sites.

**Analyses**

We used multiple linear regression to compare the ability of human population density (natural log transformed), level of development (based on a quadratic function of the factor scores of community-level infrastructure) and rugosity of habitat at fish count sites to explain reef fish biomass. Variables were fitted as fixed effects in a mixed model using the nlme library in R. To account for non-independence within countries, we added
country as a random term, significantly improving the model (likelihood ratio test on models fitted with REML adjusted for testing at the margin; ratio = 9.30, p = 0.001) [15]. The interclass correlation, indicating the relationship between points within the same country, was 0.998 [15].

All possible regression model combinations of fixed variables were compared for their fit to the data using low-sample-corrected Akaike’s information criterion (AICc) and Bayesian information criteria (BIC values) based on maximum likelihood estimation [15, 48] (Table 1). The significances of individual terms were tested by likelihood ratio tests [15]. Selected models were assessed for heteroscedacity and normality of residuals by visual assessment of plots and by addition of the varIdent variance structure to the random part, but this did not improve the model fit (likelihood ratio = 7.63, p= 0.1057).

To investigate whether there were differences in the assessed socio-economic conditions in different parts of the U-shape curve, we used natural groupings of the data to divide communities into three groups. This resulted in groupings of the four sites with the highest development, the five sites with the lowest development, and ten sights with moderate development. We then used ANOVA to test for significant differences in socio-economic conditions in these groups (Table 2). We used GLS model with the varIdent function in R to overcome violations of homogeneity in four indicators: percent of households engaged in fishing, percent of households that rank fishing as a primary occupation, the percent of households engaged in salaried employment, and the percent of fishers with boats that have engines.

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References


Table legends

Table 1. Comparison of candidate models with three fixed effects for reef-fish biomass: a quadratic function of our socio-economic development index, habitat rugosity index and natural log of human population density. All models include a random effect of country. Model 5, including the development index and habitat complexity, has the lowest BIC and AICc score, confirming it as the best fit. df= degrees of freedom; n= sample size; AICc= Akaike information criterion corrected for small sample sizes; BIC= Bayesian information criterion; ΔAICc and ΔBIC= difference from the criterion scores of the most favoured mode; AICc weight= Likelihood weight based on the AICc values of all tested models [45].

Table 2. The average percent of low, medium, and high development communities involved in select occupational and fishing activities (range in parentheses).
Figure legends

Figure 1. Fit of reef fish biomass as a function of a) human population density ($r^2=0.28$), b) habitat rugosity index ($r^2=0.54$), and c) community-level socio-economic development index ($r^2=0.77$). Solid lines show curves fitted from linear (a & b) and quadratic (c) regressions. Data distinguished by country where MD = Madagascar, SZ = Seychelles, KY = Kenya, MS = Mauritius, TZ = Tanzania.

Figure 2. Fishing practices common in different stages of socio-economic development: a) a fisher from a low development site in a small wooden canoe; b) a fisher from a moderate development site using a spear gun, and c) fishers in a high development site hand line fishing from a motorized boat (source: Seychelles Fishing Authority).

Figure 3. The biomass of reef fish in protected sites (filled symbols) and fished sites (open symbols) along a gradient of economic development. The solid line is the best-fit curve fitted with the quadratic regression of fished site biomass and development. The fish biomass from protected sites was not included in the regression analysis.
Table 1.

<table>
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<th>Model</th>
<th>Fixed model terms</th>
<th>df</th>
<th>n</th>
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<th>BIC</th>
<th>ΔAICc</th>
<th>ΔBIC</th>
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Table 2

<table>
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<tr>
<td>Average % of households that listed fishing as their primary occupation</td>
<td>48 (28-75) %</td>
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<tr>
<td>Average % of households that engaged in regular salaried employment (manufacturing, teaching, etc)</td>
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<tr>
<td>Average % of fishers using gill net</td>
<td>20 (9-36)</td>
</tr>
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<td>Average % of fishers using hand lines</td>
<td>21 (13-35)</td>
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<td>Average % of fishers using seine net</td>
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<td>Average % of fishers using pelagic nets and lines</td>
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<td>Average % of fishers with boats</td>
<td>90 (84-100) %</td>
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<td>Average % of boats with engines</td>
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<tr>
<td><strong>Presence of socio-cultural governance institutions</strong></td>
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\( ^a \) GLS model with varIdent function fitted to overcome violation of homogeneity
\( ^b \) Number of communities in group with customary sociocultural institutions that may help to govern marine resource use
\( ^c \) Chi\(^2\) statistic (p-value estimated by Monte Carlo simulation)
Figure 1.

Figure 2

Figure 2