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# Trawl fishing impacts on the status of seabed fauna in diverse regions of the globe

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## Abstract

Bottom trawl fishing is a controversial activity. It yields about a quarter of the world's wild seafood, but also has impacts on the marine environment. Recent advances have quantified and improved understanding of large-scale impacts of trawling on the seabed. However, such information needs to be coupled with distributions of benthic invertebrates (benthos) to assess whether these populations are being sustained under current trawling regimes. This study collated data from 13 diverse regions of the globe spanning four continents. Within each region, we combined trawl intensity

distributions and predicted abundance distributions of benthos groups with impact and recovery parameters for taxonomic classes in a risk assessment model to estimate benthos status. The exposure of 220 predicted benthos-group distributions to trawling intensity (as swept area ratio) ranged between 0% and 210% (mean = 37%) of abundance. However, benthos status, an indicator of the depleted abundance under chronic trawling pressure as a proportion of untrawled state, ranged between 0.86 and 1 (mean = 0.99), with 78% of benthos groups > 0.95. Mean benthos status was lowest in regions of Europe and Africa, and for taxonomic classes Bivalvia and Gastropoda. Our results demonstrate that while spatial overlap studies can help infer general patterns of potential risk, actual risks cannot be evaluated without using an assessment model that incorporates trawl impact and recovery metrics. These quantitative outputs are essential for sustainability assessments, and together with reference points and thresholds, can help managers ensure use of the marine environment is sustainable under the ecosystem approach to management.

#### KEYWORDS

benthic invertebrates, ecosystem-based fisheries management, risk assessment, species distribution modelling, sustainable fisheries, trawling

## 1 | INTRODUCTION

Bottom trawling (such as beam, otter trawls and dredge; hereafter “trawling”) is important for global food security, providing about 20 million tonnes of global catch (Amoroso et al. 2018). However, the ecological impacts of trawling on the marine environment have been a concern across the globe (Jennings & Kaiser, 1998; Puig et al., 2012; Pusceddu et al., 2014; Thrush & Dayton, 2002). Overall, there is limited large-scale quantitative evidence of the risks trawling pose to the environment and to benthic organisms that encounter physical contact with trawl gear (Mazor et al., 2017; Pitcher et al., 2017).

Ecosystem-based management (EBM) is an approach that is being adopted around the globe for managing fisheries (Astles et al., 2006; Pikitch et al., 2004). This management approach considers the suite of interactions within a given ecosystem rather than addressing issues in isolation (Holsman et al., 2017). Risk assessment is an essential component of EBM and provides critical information for prioritizing management interventions (Holsman et al., 2017; Stelzenmüller et al., 2015). In the absence of a quantitative approach, there has typically been a reliance on qualitative risk assessments of seabed trawl impacts, using expert opinion and stakeholder knowledge, or rank scoring approaches to guide management decisions (Astles et al., 2006; Fletcher, 2005; Lorance et al., 2011). However, transparent evidence-based quantitative assessments are possible with access to technologies that provide information on fishing activity (e.g. Vessel Monitoring Systems (VMS) and satellite Automatic Identification Systems (AIS) for fishery effort information) and advances in statistical modelling methods (Pitcher et al., 2017).

Recent efforts have synthesized our current understanding of trawling extent and impacts around the world (Hiddink et al., 2017; Amoroso et al., 2018; Sciberras et al., 2018). For

example, regional trawl footprint data were collated by Amoroso et al., (2018), providing a broad-scale spatial coverage of current trawl effort. The study found that 14.5% of the total studied area (7.7 million km<sup>2</sup>) was trawled, but varied considerably among 24 regions of the world. Systematic review methodologies and meta-analyses have been used to compile depletion and recovery information of trawl fishing disturbances on seabed invertebrates (Hiddink et al., 2017; Sciberras et al., 2018), highlighting those species groups that are more sensitive to trawl impacts (e.g. long-lived biota; Hiddink et al., 2019). Given these advances, they now need to be applied to knowledge of spatial distributions of seabed fauna to assess the impact and sustainability of benthos in trawled regions.

Understanding the sensitivity of benthic invertebrates (benthos) to trawling disturbance is of fundamental ecological importance because they perform essential ecosystem processes such as reworking sediments, forming habitat structures and oxygenating the seafloor (Solan et al., 2004). Furthermore, their status is commonly used as an indicator for measuring ecosystem health or disturbance (Hiddink, Jennings, & Kaiser, 2006; Przeslawski, Ah Yong, Byrne, Wörheide, & Hutchings, 2008). Despite their importance, knowledge of benthos distributions across broad spatial scales (>1,000 km<sup>2</sup>) is limited (Reiss et al., 2015); most likely attributable to high costs of surveys, limits in taxonomic expertise, and lengthy sample processing time (Fisher, Knowlton, Brainard, & Caley, 2011). New methods have been proposed to predict and expand knowledge of spatial distributions of benthos at regional scales of 1,000's of km<sup>2</sup> (e.g. Baltic Sea: Gogina and Zettler (2010); North Sea: Reiss, Cunze, König, Neumann, and Kröncke (2011); Australian waters; Mazor et al. (2017)); these methods can be coupled with known distributions of trawl intensity to compute

**TABLE 1** Study regions and characteristics of areas where benthos groups are predicted. Note that more sites may have been surveyed but were left out due to missing environmental data. See Table S1, and Figures S1–S13 for further information on each survey

Continent	Region	Survey Area km <sup>2</sup>	Trawl SAR exposure % of survey area (km <sup>2</sup> )	Depth Range	Benthic Surveys	No. of Survey Sites*	Survey Years	Gear Types for Benthic Invertebrate Survey
North America	Bering Sea	632,677	9.00% (56,912)	12 – 1809	6	1,333	2008, 2009, 2010	Otter trawl shelf, and otter trawl slope
	Aleutian Islands	104,340	2.19% (2,285)	47 – 1,185	3	366	2010	Otter trawl
	Gulf of Alaska	348,490	3.24% (11,292)	0 – 1,130	3	817	2009	Otter trawl
	West Coast	152,480	9.51% (14,497)	30 – 1,349	3	1887	2008, 2009, 2010	Otter trawl
Europe	North Sea	571,694	78.92% (451,183)	13 – 244	1	267 (epifauna) 1,187 (infauna)	1999/2000 – 2002	Beam trawl and grab
	Kattegat/ Western Baltic Sea	99,465	69.10% (68,729)	0 – 94	1	706	2000 – 2013	Grab
Australia/Oceania	Gulf of Carpentaria	381,919	4.07% (15,530)	10 – 102	2	104	1990	Dredge and grab
	Great Barrier Reef	179,944	10.35% (18,633)	5 – 103	6	1940	2003 – 2005	Prawn trawl and sled
	South East	165,783	13.64% (22,612)	7 – 1,015	4	408	1 survey = 1993 – 19963 surveys = 1979 – 1983	Sled and grab
	Western Australia	529,665	0.9% (4,714)	50 – 1,311	3	238	2005	Beam Trawl, sled and grab
Africa	Chatham/ Challenger New Zealand	443,421	3.68% (16,310)	60 – 2000	3	142 (DTIS) 146	2007	Deep towed imaging system (DTIS), epibenthic seamount sled and beam trawl
	Benguela/Agulhas South Africa	219,831	41.66% (91,575)	29 – 889	1	223	2011	Otter trawl
	Namibia	171,927	112.42% (193,275)	90 – 812	1	222	2008, 2009, 2010	Gisund super two-panel bottom trawl

benthos status (relative to an untrawled state—calculated from impact rates, recovery rates and exposure to trawling) and help inform the extent to which trawling is sustainable in different areas of the seabed (Mazor et al., 2017). Combined, the information can be used assist managers in the choice of best practices to minimize impacts and ensure sustainability in the local context (McConnaughey et al., 2020).

Here, we quantify the status of benthos in 13 case-study regions from four continents (Australia, Europe, Africa and North America). Each region was chosen based on the availability of trawl intensity data and benthos survey data. To assess the status of benthos under current trawling practices, we modelled their current-day abundance distributions (based on recent survey samplings), and spatially combined these with maps of trawling intensity (Amoroso et al., 2018) and published recovery and depletion estimates derived from global meta-analyses (Hiddink et al., 2017; Sciberras et al., 2018; Hiddink et al., 2020), using a quantitative risk assessment method (Pitcher et al., 2017). Our findings aim to advance understanding of the current impacts and risks (to benthos) of trawling on the seafloor for regions across the globe.

## 2 | METHOD

### 2.1 | Study regions

Thirteen large-scale study regions across the globe were selected for analysis based on data availability (Table 1; Table S1). The geographical extent of each region was bounded by the latitude, longitude and depth range of the sites for which benthos data from systematic surveys were available to avoid excessive extrapolation of benthos predictions. For maps of study regions, see Figures S1–S13.

### 2.2 | Trawl intensity

Trawl intensity data were acquired from Amoroso et al., (2018). These data were calculated using VMS or fishing log-book data, to produce a swept area ratio (SAR: the annual cumulative area swept by trawl gear within a given grid cell of seabed, divided by the area of that grid cell) of trawling within a grid cell (either 1 km<sup>2</sup>, 0.01° or 1 × 1 min grids of longitude and latitude), over a 3- to 5-year period (typically 2008–2010). To ensure trawling activity is representative, we only included regions where >70% of trawling activity was accounted for (Amoroso et al., 2018). To enable comparisons across regions where <100% of trawling activity was reported, we scaled-up trawling effort (F by 100/coverage%) for each region and by gear type to represent total trawl intensity (i.e. 100% trawl activity for each region), and recalculated regional SARs and footprints. This scaling and recalculation assumes that collated data are representative of the spatial distribution of the total.

## 2.3 | Benthos distributions

### 2.3.1 | Benthos data

Benthos data from seabed surveys were sought for regions where trawl intensity data were available from Amoroso et al., (2018). Ultimately, data were collated from 13 of 24 regions. Benthos abundances in surveys were recorded as counts or weight and were standardized by sampled area. We included surveys of both infauna and epifauna where possible and attempted to match survey years to the trawl data. Survey sampling gear varied among regions, but sampling was predominantly conducted using an otter trawl, benthic sled and/or grab (Table 1).

Eight taxonomic classes of benthos were examined: Anthozoa (i.e. sea anemones and corals), Ascidiacea (sea squirts), Asteroidea (seastars), Bivalvia (bivalved shelled molluscs), Gastropoda (sea snails and slugs (alt: coiled, conical or shell-less molluscs), Malacostraca (crabs and shrimps), Ophiuroidea (brittle stars) and Polychaeta (segmented worms). These classes were the subject of meta-analyses in which depletion and recovery information have recently been estimated (Hiddink et al., 2017; Sciberras et al., 2018; Hiddink et al., 2020; Figure 1). Following Mazor et al. (2017), we further divided taxonomic classes into benthos groups, that is, groups of species/taxa within a class that have similar spatial distributions and relationships with environmental variables. The clustering approach uses Multivariate Regression Trees (MRT) to group sites based on the sampled abundances of taxa and their relation with environmental variables, and assigns taxa to these site groups using the Dufrêne and Legendre (1997) indicator-species metric (DLI) (Mazor et al., 2017). Benthos groups were used because of inconsistencies in the level of reported taxonomic hierarchy among surveys and therefore serve as the lowest resolution of benthic data considered for this study.

### 2.3.2 | Environmental predictors for modelling benthos

Thirty-four environmental variables previously reported to be associated with distributions of a range of benthic invertebrates (Mazor et al., 2017) were used to model the distributions of benthos in each region (Table 2). All variables were available at a global extent at various spatial scales and were processed into consistent grids to match the resolution of the trawl intensity data provided for each region. Environmental layers (e.g. data from the NASA Ocean Biology Processing Group) were processed using R (R Core Team 2018; package “ncdf4”; Pierce, 2017, and package “raster” Hijmans 2019) to convert netCDF files into rasters. Annual averages for environmental variables were calculated from the monthly means of all available years. Seasonal range composites were calculated from the range of January to December monthly means, averaged across all years. All environmental variables (using raster format)



**FIGURE 1** Box plots by region (Table S1 for more details) of: a) the percentage of benthos-group abundance exposed to trawling (SAR exposure), b) depletion values  $d$ , c) recovery parameters  $R$ , d) the relative status of benthos groups using mean values and lower confidence interval for recovery. The black lines represent the median value

were transformed into the relevant projection and coordinate system (to match the gridded trawl intensity data) with resampling by cubic convolution to the desired cell size (either  $1 \text{ km}^2$ ,  $0.01^\circ$  or  $1 \times 1 \text{ min}$  grids of longitude and latitude). Rasters were then clipped to the boundaries of each study region. Other environmental layers required three-dimensional interpolation to extract properties at the seafloor using a bathymetry layer (e.g. CSIRO Atlas of Regional Seas; Ridgway, Dunn, & Wilkin, 2002). Predictors that did not vary among surveyed sites ( $SD = 0$ ) or contained missing data for considerable parts of a region were excluded from individual analysis. Where predictors were largely complete ( $>90\%$  of grid), `na.spline` (package “`zoo`”; Zeileis, 2019) was used to interpolate missing predictor data.

### 2.3.3 | Predicting benthos distributions

Benthos-group abundance distributions were predicted for each region using R package “`randomForest`” (Liaw & Wiener, 2002). For each region, we applied one of three methods to obtain a site-by-taxon matrix following Mazor et al. (2017): i) a single-gear approach—benthos were sampled by one device; abundance data were arranged into a conventional site-by-taxon matrix, ii) multiple gear approach—benthos were sampled by two different devices that sampled an overlapping composition of benthos at the same sites; a multiplicative scaling factor was estimated for each taxon sampled by different gears (note gear that targeted and predominantly sampled epifauna (e.g. trawls) and infauna (e.g. grabs) were not combined), and iii) disparate data sets

**TABLE 2** Thirty-four environmental variables used to predict benthos abundance distributions

Variable	Values	Source	Years	Scale
Temperature at seafloor (°C)	Annual Average Seasonal Range	CSIRO Atlas Of Regional Seas (CARS 2009)	up to 2009	1/2°
Salinity at seafloor (psu)	Annual Average Seasonal Range	CSIRO Atlas Of Regional Seas (CARS 2009)	up to 2009	1/2°
Oxygen at seafloor (ml/l)	Annual Average Seasonal Range	CSIRO Atlas Of Regional Seas (CARS 2009)	up to 2009	1/2°
Silicate at seafloor (µmol/l)	Annual Average Seasonal Range	CSIRO Atlas Of Regional Seas (CARS 2009)	up to 2009	1/2°
Phosphate at seafloor (µmol/l)	Annual Average Seasonal Range	CSIRO Atlas Of Regional Seas (CARS 2009)	up to 2009	1/2°
Nitrate at seafloor (µmol/l)	Annual Average Seasonal Range	CSIRO Atlas Of Regional Seas (CARS 2009)	up to 2009	1/2°
Depth 1 arc-minute	Mean	ETOPO Amante, C. and B.W. Eakins (2009)	1940 to 2008	1 arc-minute
Chlorophyll <i>a</i> concentration (mg/m <sup>3</sup> )	Annual Average Seasonal Range	NASA Ocean Biology Processing Group (OBPG) Aqua-Modis Level 3 Browser, Standard Mapped Image (SMI), Chlorophyll calculated with OC3 algorithm.	2002 – 2016	0.041° (4 km)
Attenuation coefficient (K490)	Annual Average Seasonal Range	NASA Ocean Biology Processing Group (OBPG) Aqua-Modis Level 3 Browser, Standard Mapped Image (SMI), Diffuse attenuation coefficient at 490 nm, KD2 algorithm.	2002 – 2016	0.041° (4 km)
Particulate Organic Carbon mg/m <sup>3</sup> (POC)	Annual Average Seasonal Range	NASA Ocean Biology Processing Group (OBPG) Aqua-Modis Level 3 Browser, Standard Mapped Image (SMI), Particulate Organic Carbon, D. Stramski, 2007 (443/555 version)	2002 – 2016	0.041° (4 km)
Photosynthetically Active Radiation (PAR)	Annual Average Seasonal Range	NASA Ocean Biology Processing Group (OBPG) Aqua-Modis Level 3 Browser, Standard Mapped Image (SMI), Photosynthetically Available Radiation, R. Frouin	2002 – 2016	0.041° (4 km)
Sea Surface Temperature Night-time (SST_Night)	Annual Average Seasonal Range	NASA Ocean Biology Processing Group (OBPG) Aqua-Modis Level 3 Browser, Standard Mapped Image (SMI), SST 11 µ night-time.	2002 – 2016	0.041° (4 km)
Sea Surface Temperature Daytime (SST_Day)	Annual Average Seasonal Range	NASA Ocean Biology Processing Group (OBPG) Aqua-Modis Level 3 Browser, Standard Mapped Image (SMI), SST 11 µ daytime.	2002 – 2016	0.041° (4 km)
Net Primary Production (NPP)	Annual Average Seasonal Range	Ocean Productivity – Oregon State University Behrenfeld MJ, Falkowski PG (1997) Photosynthetic rates derived from satellite-based Chlorophyll concentration. <i>Limnology and Oceanography</i> 42:1–20.	2002 – 2016	1/6°
Benthic Irradiance (BIR)	Annual Average Seasonal Range	*Calculated in R BIR = PAR × exp(-K490 × depth)	2002 – 2016	0.041° (4 km)
Export Particulate Organic Carbon flux (EPOC)	Annual Average Seasonal Range	Calculated in R using the exponential decay model Pace et al. 1987 EPOC = 3.523 × NPP × depth <sup>-0.734</sup> .	2002 – 2016	0.041° (4 km)
Gravel	Mean	Sediment from dbSEABED	up to 2015	0.01° where present
Sand	Mean	Sediment from dbSEABED	up to 2015	0.01° where present
Mud	Mean	Sediment from dbSEABED	up to 2015	0.01° where present



approach—benthos were sampled by multiple surveys disparate in one or more of spatial extent, time, taxonomic resolution and identification, sampling device and abundance metrics; in this case, random forest models predict taxa to unsampled sites combined with a scaling approach that normalizes taxa data to represent the proportion of abundance it contributes within its data sets.

Model performance was measured by the  $R^2$  of overall fit of predicted against observed values and by the cross-validated out-of-bag (OOB)  $R^2$  values (estimated internally using bootstrapped samples that leave out about one-third of the data; Breiman, 2001). Predictor importance was extracted from the models as per Mazor et al., (2017) by obtaining the random forest predictor importance measure (%IncMSE). Predictor importance across models was calculated by scaling importance by its proportionate contribution to model performance (OOB  $R^2$ ) for each benthos group. These proportions were then averaged across all models, per region and per taxonomic class to estimate overall predictor importance. Models with poor prediction performance (cross-validated OOB  $R^2 < 5\%$ ) were excluded from the status assessment.

## 2.4 | Trawl SAR exposure of predicted benthos distributions

We quantified trawl SAR exposure (i.e. proportion of benthos abundance currently distributed in areas that are trawled) as a percentage, by spatially overlaying benthos-group distributions and trawl intensity (SAR). Specifically, we summed the product of the predicted benthos-group abundance in trawled grid cells multiplied by the trawl SAR of each cell and then divided by total group abundance in all cells, as per Mazor et al., (2017). We note that SAR exposure  $> 100\%$  may occur for benthos abundance in cells with SAR  $> 1$  which are repeatedly exposed and thus the repeated exposure can be greater than the total abundance in all cells.

## 2.5 | Benthos status assessment model

Here we applied a quantitative risk assessment method derived from the logistic population-growth equation (Pitcher et al., 2017) to estimate “relative benthos status” (RBS):

$$RBS = 1 - F \frac{d}{r},$$

where  $F$  is the trawling SAR,  $d$  is trawl depletion rate per trawl pass, and  $r$  is population growth/recovery rate. Depletion rate parameters, specific to taxonomic classes, were obtained from Sciberras et al. (2018, for trawl gears only), and recovery rates were derived from Hiddink et al., (2020), respectively (Table S2; see Supporting Information methods for details of derivation). Depletion rates also differ by trawl gear types and by habitats, and recovery rates also vary with habitat types. To account for this, taxonomic class-level average depletion and recovery rates were scaled according to gear types and habitat types (see Supporting Information methods). Absolute status, expressed as a proportion, was estimated from the

product of RBS multiplied by the predicted abundance distribution (grid-cell abundances) and divided by the total benthos-group predicted abundance. A status of 1 indicates a state where the benthos population is not depleted by trawling and 0 being entire depletion. We characterized the uncertainty range in the status estimate by using the mean values for depletion and recovery, and by using the lower 95% confidence interval (CI) for recovery. We used the lower 95% CI as it was considered more consistent with the concept of a precautionary approach. It was sufficient to use just the CI for recovery without uncertainty in depletion because the uncertainties in these parameters are inversely related. Benthos status was also calculated to consider only trawled areas (grid cells with  $F > 0$ ) of our study regions to examine how status may change by spatial extent and specifically within trawled-only areas.

To investigate the relationship between trawl SAR exposure and benthos status, we plotted the trawl SAR exposure, benthos status and sensitivity ( $d/R$ ) of each benthos group. Sensitivity  $d$  (trawl depletion rate per trawl pass) and  $R$  (population growth/recovery rate) was calculated as described in Supporting Information methods.

## 3 | RESULTS

### 3.1 | Benthos distributions

A total of 220 benthos-group distributions were modelled from our 13 study regions and 8 taxonomic classes (Table 3; Table S3). Average explanatory model performance across all benthos-group models, measured by the  $R^2$  of the overall fitted against observed values, was 0.75 (median = 0.82), and the cross-validated  $R^2$  of predicted against OOB values was 0.37 (median = 0.34). Model performance varied greatly by region (Figure S14), but not by taxonomic class (Figure S15). The most important predictors across all models were the seasonal range of photosynthetically active radiation (PAR), the average temperature at the seafloor ( $^{\circ}\text{C}$ ), the average salinity at the seafloor (psu) and oxygen at the seafloor (ml/l) (Figures S16; S17). The pattern of predictor importance was highly variable across regions (Figure S16); however, some regions are particularly influenced by sediments, such as the Gulf of Carpentaria and the Great Barrier Reef. Predictor importance was less variable among taxonomic classes (Figure S17). Different benthos groups had different orders of predictor importance, but appeared more consistent across taxonomic classes compared to regions.

### 3.2 | Trawl SAR exposure

Across all regions, the mean percentage of the predicted abundance of benthos groups exposed to trawling was 36.63% (median = 8.90%), with a range between 0% and 209.90% (Figure 1). The European regions, Kattegat/Western Baltic Sea and North Sea had the highest overlap of trawl activity with distributions of benthos, with an average exposure of 142.53% and 134.48%, respectively.



**TABLE 3** Number of derived benthos groups (method following Mazor et al., 2017) across region and per taxonomic class

Region	Benthos Groups	Anthozoa	Ascidacea	Asteroidea	Bivalvia	Gastropoda	Malacostraca	Ophiuroidea	Polychaeta
Aleutian Islands	10	1	2	2	1		2	2	
Bering Sea	23	4	2	4	1	3	5	2	2
Gulf of Alaska	17	3	2	3	1	2	4	2	
West Coast USA	17	3		4		3	4	3	
Kattegat/ Western Baltic Sea	7				2	2		1	2
North Sea	40	2	2	5	6	6	9	5	5
Benguela/ Agulhas South Africa	18	2	1	4		2	4		
Namibia	3						3	3	2
Chatham/ Challenger New Zealand	22	3		4	2	3	3	3	4
Great Barrier Reef	16	2	1	2	3	2	3	3	
Gulf of Carpentaria	16	1	3	1	3	1	3	2	2
South East Australia	13				1	1	4	3	4
Western Australia	18	2		1	2	2	4	2	5
Total Number	220	23	13	30	22	27	48	31	26

The regions with moderate overlap were the African regions, Namibia (107.70%) and Southern Benguela and Agulhas ecoregions of South Africa (37.57%). Regions with the least overlap of trawling with benthos groups were Western Australia (1.13%), Gulf of Alaska (2.32%) and Aleutian Islands (2.41%).

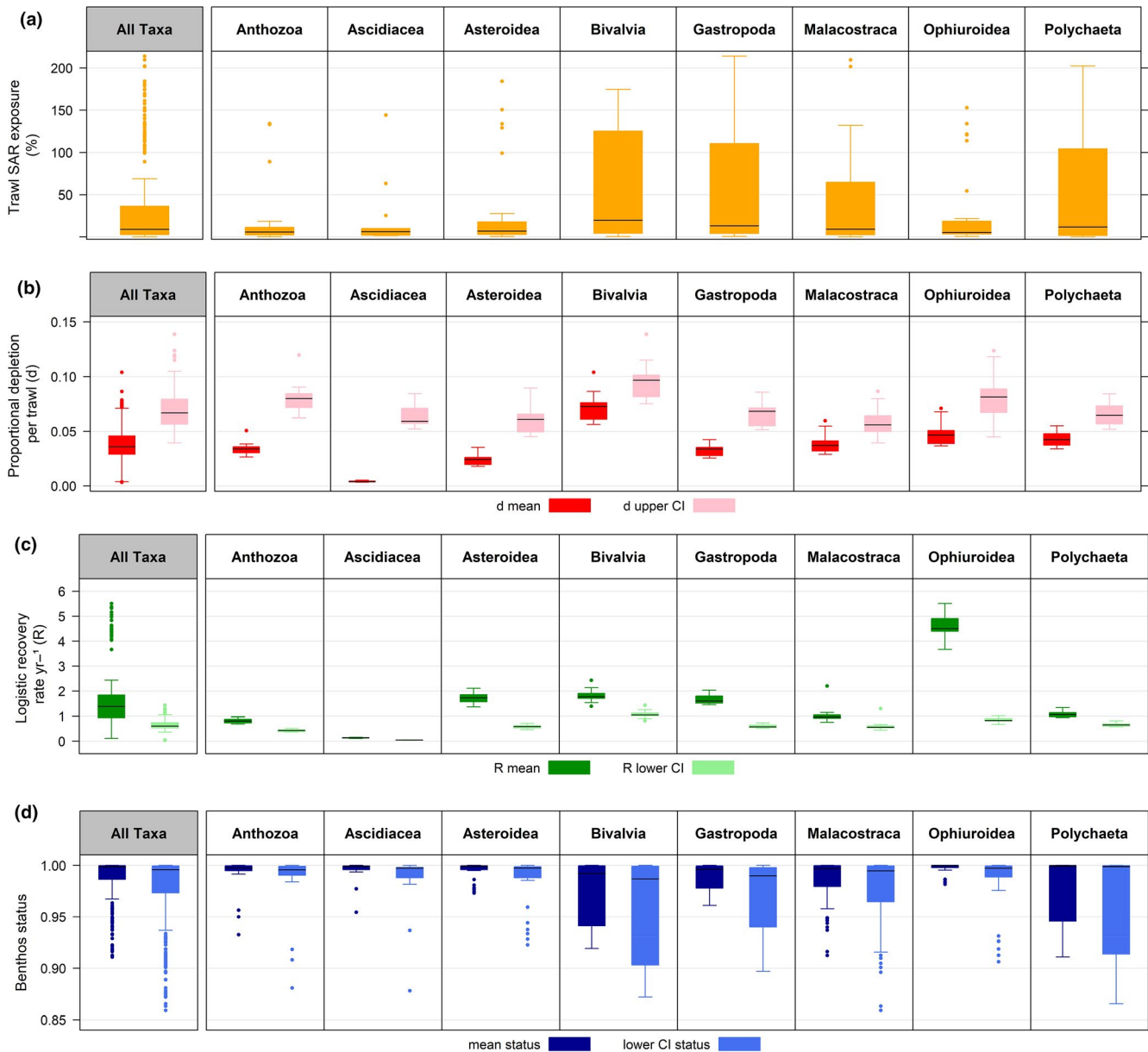
Among taxonomic classes, the range of trawl exposures (Figure 2a) was less than that among regions (Figure 1a). Taxonomic classes that had the highest mean percentage of their distributions overlapping with trawling across all regions were Bivalvia (55.70%), Gastropoda (53.58%) and Polychaeta (46.44%) (Figure 2). The classes with the least trawl exposure were Anthozoa (20.52%) and Ascidacea (21.31%).

### 3.3 | Benthos status

Across all benthos groups in all regions, the average status was 0.9878 (mean) and 0.9759 (lower CI) (Figure 1,2). However, for individual benthos groups, status ranged from 0.9110 to 1 (mean), and 0.8592 to 1 (lower CI). The North Sea region had the lowest average status of 0.9538 (mean) and 0.9097 (lower CI), followed by the Kattegat/Western Baltic Sea (0.9554 mean; 0.9189 lower CI) (Figure 1d,3). These regions also had the largest range of status (max-min). The majority of regions (8 of 13) had an average

status > 0.99 (both mean and lower CI values; Figure 3), whereas for taxonomic classes, only half of the benthos groups had an average status > 0.98 (both mean and lower CI values; Figure 2d). The class Bivalvia had the lowest average status (0.9738 mean; 0.9587 lower CI), followed by Malacostraca (0.9841 mean; 0.9742 lower CI) and Gastropoda (0.9895 mean; 0.9718 lower CI). Similarly to regions, taxonomic classes with the lowest average status also had the largest range of values. Benthos status when calculated for only trawled areas (grid cells with SAR > 0) of our study regions (Figure S18; Table S3) were slightly lower (range from 0.8754 to 0.9999, and lower CIs from 0.8020 to 0.9999; average status 0.9807 and 0.9610 (lower CI)) compared to benthos status for our entire study regions (Figure 1) (means ranging from 0.9110 to 1, and lower CIs from 0.8592 to 1).

We found that higher trawl SAR exposure was related to a lower benthos-group status ("lower" in relation to our results—where status 0.98 was the lower confidence interval) (Figure 4). Benthos status also depended on the sensitivity ( $d/R$ ) of the benthos group to trawling impacts and their ability to recover. Sensitivity ranged from 0.0076 to 0.0697, and higher sensitivity to trawling (dark points on Figure 4) was related to a lower benthos status. However, this relationship did vary and some groups in Europe with higher sensitivity have greater exposure to beam trawls and dredges; the spatial footprints of these gear types are



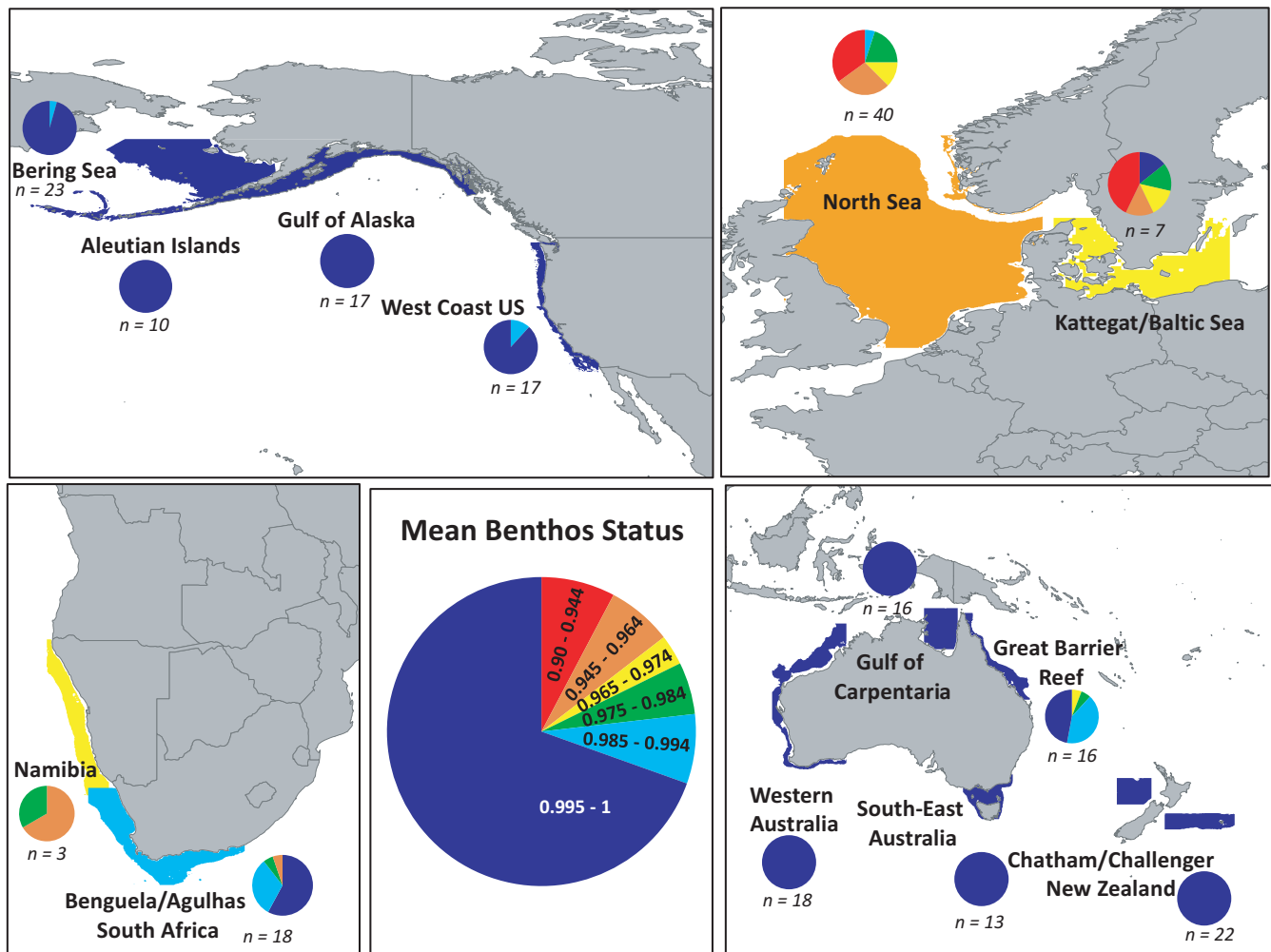
**FIGURE 2** Box plots by taxonomic class (Table 3 for more details) of a) the percentage of benthos-group abundance exposed to trawling (SAR exposure) b) depletion values  $d$ , c) recovery parameters  $R$ , d) the relative benthos status using mean values and lower confidence interval for recovery. The black lines represent the median value

narrower than those of otter trawls and thus contribute less to cell SAR but lead to higher depletion rates ( $d$ ). Other factors that prevent a strict relationship with sensitivity are that distributions of benthos groups and of trawling (and different gear types) are complex and differ with sediment distributions.

## 4 | DISCUSSION

This study presents a large-scale assessment of the status of seabed invertebrate communities and provides insight into the sustainability of bottom trawling in regions across the globe. Unlike other large-scale assessments that have examined trawl footprints (Amoroso

et al., 2018), or status of sedimentary habitats in relation to trawling (Pitcher et al., in review), this work incorporates sampling data from surveys of benthos enabling a more direct quantification of trawl impacts on different types of benthos. Our results indicate that benthos groups may have up to 210% of their distribution exposed to trawl activity (as SAR intensity), yet the lowest benthos status at a regional scale was 0.86, decreasing to 0.80 within trawled footprint areas (Figure S18). In 11 of our 13 case-study regions, all benthos groups had a status  $> 0.95$ , and only a quarter (22%) of benthos groups had a status  $< 0.95$  (i.e. reduced by 0.05–0.14 owing to trawling activity). Overall benthos status was relatively high (mean status = 0.99; lower confidence interval = 0.98; mean status in trawled areas = 0.98; lower confidence interval in trawled areas = 0.96).

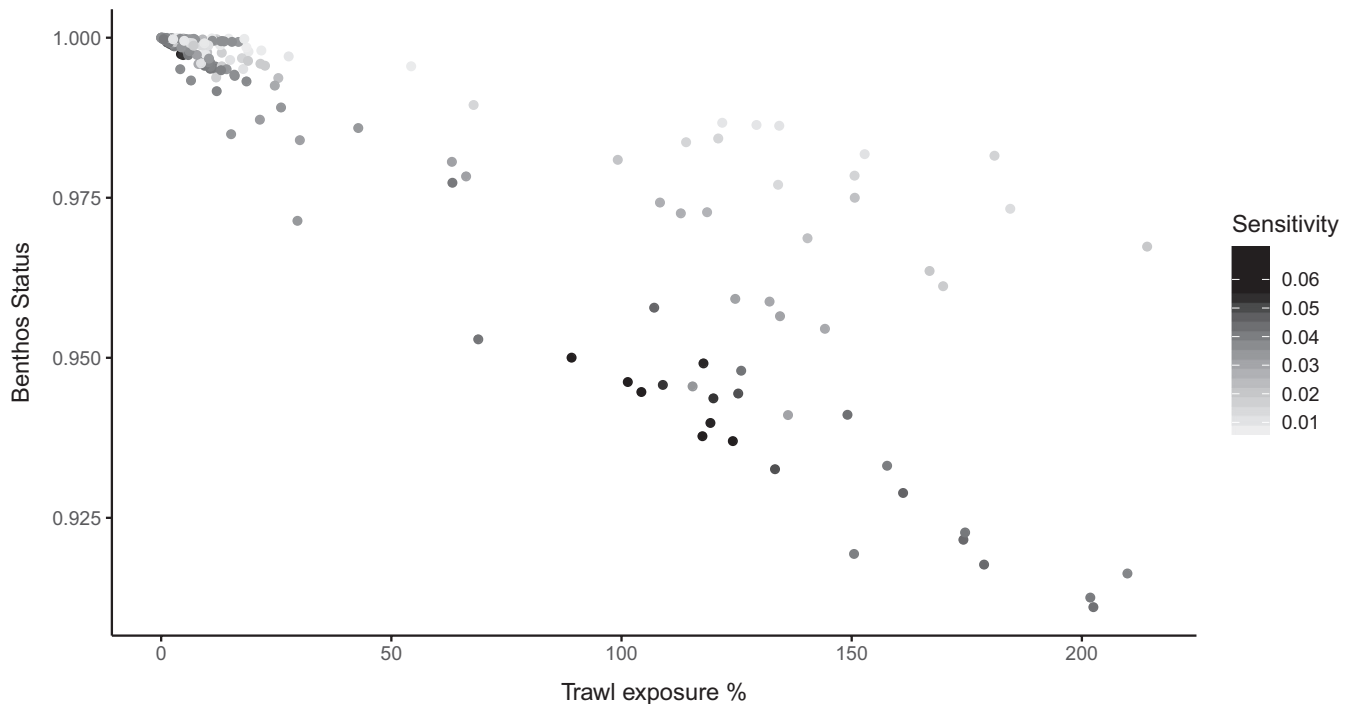


**FIGURE 3** Map of mean benthos-group status across 13 case-study regions (for study region maps, see Figure S1-S13). For each region,  $n$  is the total number of benthos groups assessed, and pie charts represent the proportion of benthos groups with a particular benthos status—coloured according to the overall mean benthos status pie chart. Figure appears in colour in the online version only

Hence, regional-scale impacts of trawling on the seabed communities assessed in this study seemed less than might be expected from results of previous studies (Hiddink et al., 2017; Amoroso et al., 2018; Sciberras et al., 2018).

European regions (the North Sea and Skagerrak/Kattegat) have trawl footprints covering > 50% of their continental shelf (Amoroso et al., 2018) and had the lowest average benthos status between 0.95–0.96 (Figure 3). Regions of Africa with trawl footprints of ~ 10%–30% of their continental shelves (Amoroso et al., 2018) displayed an average benthos status between 0.97–0.99 (Figure 3). Regions, such as North America and Australia, with lower trawl footprints (<10%) displayed higher benthos status (i.e. >0.99). Although average benthos status per region relates to the overall trawl SAR exposure, there are differences for particular benthos groups due to their sensitivity to trawling (Figures 1, 4). For example, average benthos status for the North Sea region was 0.95, but one *Bivalvia* group had a lower status of 0.92 due to higher trawl exposure (174.64%) and sensitivity (0.04) (Figure 5a).

Spatial overlays of human activities on habitats or species distribution maps are often used to infer threats and risks (Evans et al., 2011; Trebilco et al., 2011) and can be informative for prioritizing areas where there is greater potential risk of impact, and for indicating where more information is needed (Ban, Alidina, & Ardron, 2010). However, our results show that while there is a general trend that greater overlaps of benthos distributions with trawling result in lower benthos status (Figure 4), the rates of impact and the recovery rates (sensitivity; Table S3) of organisms are also important (Pitcher, 2014). Simple spatial overlap analyses that do not consider these dynamics are problematic for determining specific management actions (Tulloch et al., 2015). For example, Benguela/Agulhas South Africa's Asteroidean group has considerably higher trawl exposure (129.32%) than the Great Barrier Reef Malacostraca group (15.19%), yet their status is relatively similar (0.9864 and 0.9849, respectively; Figure 5). This similarity is due to the higher recovery ( $R = 1.81$ ) and thus lower sensitivity (0.01) to trawl impacts for Benguela/Agulhas South Africa's Asteroidea in comparison to the higher sensitivity (0.03) for Malacostraca in the Great Barrier



**FIGURE 4** Relationship between benthos status (mean values) and trawl SAR exposure (Table S3). Each point represents a predicted benthos group ( $n = 220$ ), and sensitivity ( $d/R$ ), where  $d$  (trawl depletion rate per trawl pass) and  $R$  (population growth/recovery rate) is calculated as described in Supporting Information methods

Reef. Thus, when quantifying risks, the dynamics of biological processes (e.g. the depletion and recovery component in our assessment model) need to be incorporated, as presented in this study, to avoid misdirecting management actions and to ensure effective outcomes.

Comparisons across regions and taxa are complex when different quantities and sources of data are used. For instance, our study indicates that the taxonomic class Bivalvia has a slightly lower benthos status than other classes. However, this may be related to the higher number of bivalve groups located in heavily trawled regions of Europe. Likewise, for Namibia, our results are based only on three Malacostraca groups, as these were the only taxa for which data were available for the region. It is likely that the average benthos status calculated for this region is not representative of other benthos taxa. Species distribution model performance also ranged widely among regions, with poorer performance in some regions such as the Aleutian Islands and Kattegat/Western Baltic Sea (Figure S14). Differences in performance are possibly related to the range of taxa or environmental variables in each region, where model performance has been found to be higher for taxa with narrower environmental gradients compared to those with larger areas of occupancy (Grenouillet, Buisson, Casajus, & Lek, 2011). Other caveats of this study include the spatial scale of benthic surveys, where some countries sampled the same or similar spatial extents to that of their trawl fishery grounds, while others have used a broader regional approach (Figures S1 – S13). This may lead to indications of greater relative trawl exposure and lower status in the former and the opposite in the latter, simply due to study extent. To address this issue, we also provided benthos status for trawled-only areas (only for grid cells

with  $SAR > 0$ ) and found comparable results with only a slight decrease of benthos status within trawled-only areas in comparison to our full study area extents (Figure S18). Lower benthos status may also occur if this study attempted to predict relative to a pristine pre-trawled baseline as many regions have had long histories of trawling which is likely to have modified benthic community composition and distribution. It is important to note that we have only considered eight common taxonomic classes and have not included biogenic habitats or most types of colonial organisms (e.g. bryozoans, porifera and hydrozoans). These organisms are expected to be more sensitive to trawling (Althaus et al., 2009; Collie, Hall, Kaiser, & Poiner, 2000) and, depending on how they are distributed in relation to where trawling occurs, would likely have a lower benthos status than the classes of biota assessed in this study. For example, Anthozoa and Ascidiacea had lower trawl exposure as such species are commonly found on hard substrata that are less exposed to trawling (Lambert, Jennings, Kaiser, Hinz, & Hiddink, 2011; Pitcher et al., 2016). Benthos data in this study were predominantly sampled in unconsolidated habitat types that are conducive to survey by trawl gears; thus, our outcomes will not reflect benthos in hard ground habitats which may be more sensitive (Lambert et al., 2011). Nevertheless, some limitations are inherent when conducting broad-scale, multiregional studies, that are dependent on existing available data.

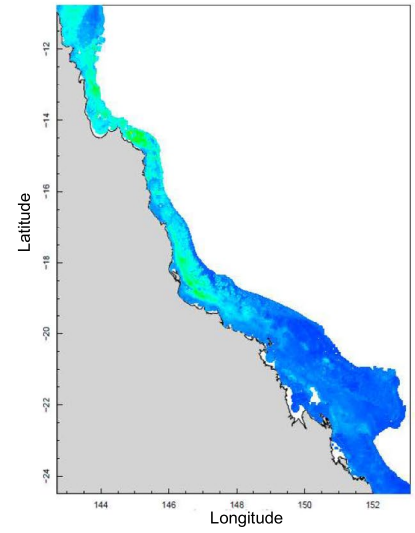
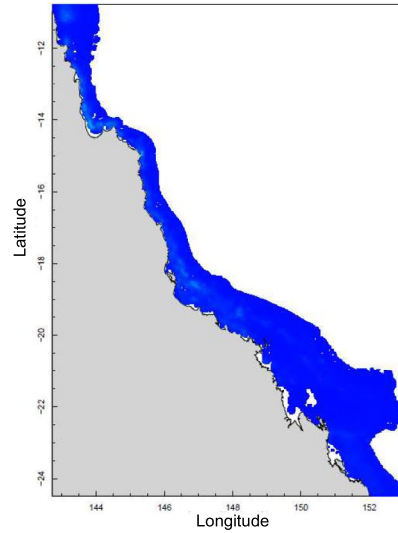
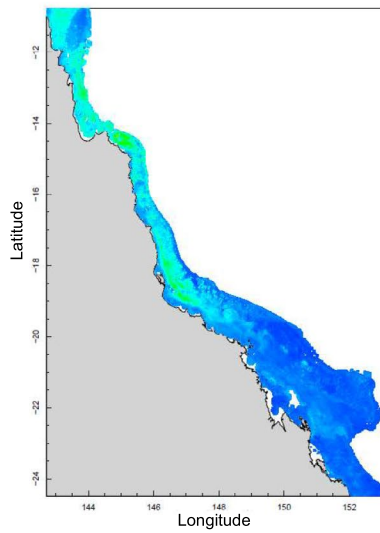
Overall, our study presents the most comprehensive and extensive quantitative synthesis of information regarding the status of benthos invertebrate communities in multiple regions worldwide. We highlight the importance of quantifying benthos status for environmental risk assessments in comparison to simpler spatial overlap

## Distribution

## Impact

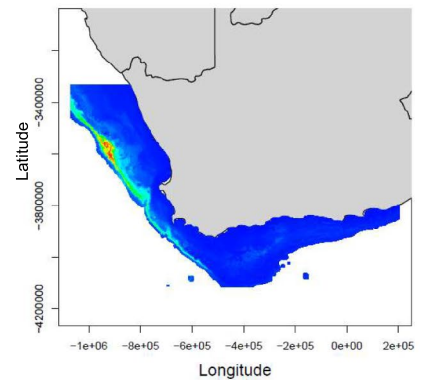
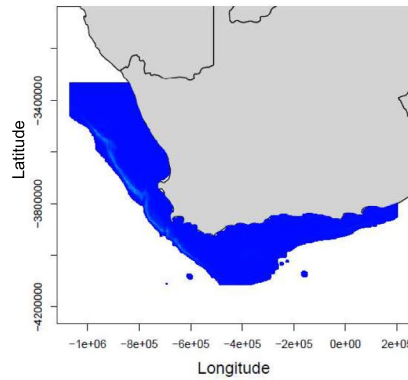
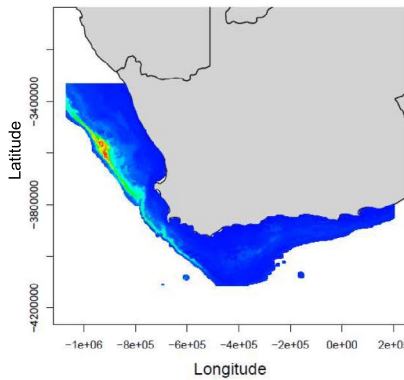
## Status

(a)



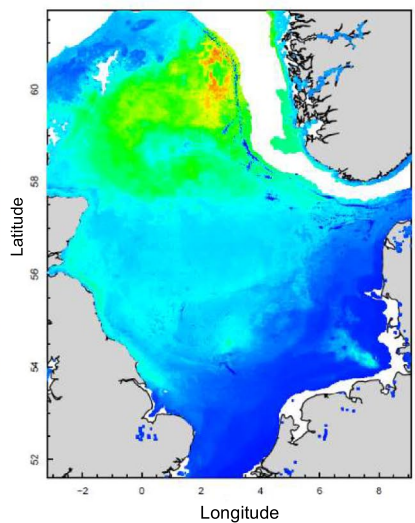
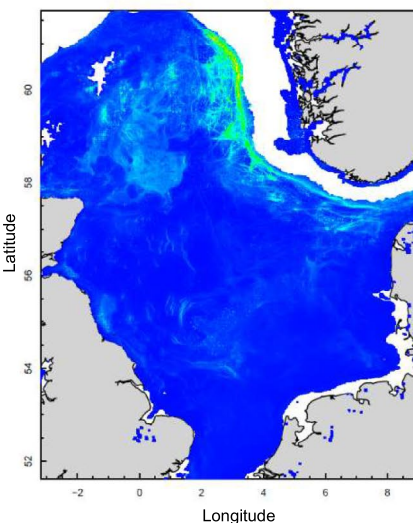
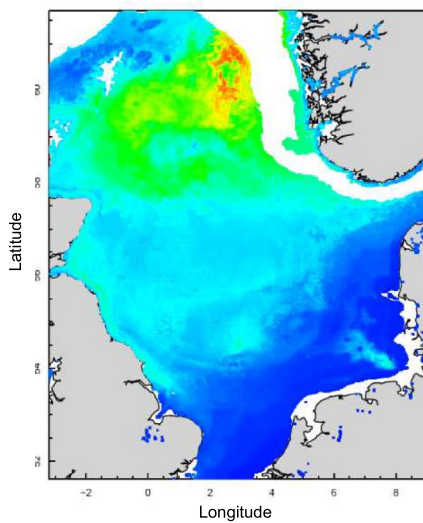
0.00000 0.00010 0.00020  
Predicted abundance (kg/ha)

(b)



0.00000 0.00010 0.00020  
Predicted abundance (kg/ha)

(c)



0e+00 1e-05 2e-05 3e-05 4e-05  
Predicted abundance (kg/ha)



**FIGURE 5** Three case-study examples of benthos groups a) North Sea bivalve group (infauna) (trawl SAR exposure 174.64%, benthos status 0.92), b) Benguela/Agulhas South African asteroidean group (trawl SAR exposure 129.32%, benthos status 0.99), c) Great Barrier Reef malacostraca group (trawl SAR exposure 15.19%, benthos status 0.98). For each region showing (left to right) the predicted abundance distribution of the benthos group, distribution of impacted abundance, and predicted benthos status distribution. Figure appears in colour in the online version only

only approaches. Our results demonstrate that, while there is a broad relationship between trawl SAR exposures and benthos status, exposure alone is not sufficient to account for benthos status or for implementing risk assessments and management decisions at regional or local scales, where adequate benthos distribution and sensitivity data (trawl impact and recovery) are available. Our study encompasses multiple regions across the globe where trawling occurs at a range of intensities and extents. However, other regions where trawl intensity is known to be higher, such as the Mediterranean Sea and South East Asia (FAO 2014; Amoroso et al., 2018; Suuronen et al., 2020), could not be included due to lack of available benthos survey data. For such regions where data (benthic or otherwise) are limited, are of poor quality (e.g. low resolution) or their acquisition is difficult, we may need to rely on coarser methods of estimating trawl risks. For example, using the broader patterns observed by spatial overlap studies, trawl exposure measures, maximum sustainable yield reference points (Fmsy), habitat status assessments (Pitcher et al., in review) or regional SARs (ratio of total swept area trawled annually to total area of region; Amoroso et al., 2018). Ideally, more benthos surveys in heavily trawled regions are needed and integrated approaches where multiple stakeholders (e.g. governmental, academic, industrial) contribute to marine benthic monitoring (Barrio-Froján et al., 2016) may offer a possible solution for better quantifying the state of the seabed in trawled areas of the world's oceans.

Findings from this study, and broader application of the approaches used in this study, will enable environmental managers to identify which regions and taxa are at greatest risk of unsustainable trawling regimes. Ideally, these assessments will need to be coupled with reference points and thresholds that indicate risk (e.g. Lambert et al., 2017). For example, is a regional benthos status of 0.95 acceptable to stakeholders and the wider community? What are the cascading effects of such a status on the wider marine ecosystem? Reference points for benthic invertebrates are undeveloped and will require further research to determine them, which will likely be specific to a given region (Couce, Engelhard, & Schratzberger, 2019; Lambert et al., 2017). However, the specificity of the status information provides useful quantitative guidance for implementing management measures to mitigate the impacts (McConnaughey et al., 2020). We suggest that such topics need to be the focus of future research to support the growing commitment for countries around the globe to implement ecosystem-based management (EBM) principles and practices, and to manage fisheries in a manner that is sustainable for marine ecosystems.

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## DATA AVAILABILITY STATEMENT

The underlying data used in this paper are available at <https://trawlingpractices.wordpress.com/datasets/>. All other data needed to repeat the analyses in the paper are presented in the paper or the supporting information, or published in cited articles and reports.

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## SUPPORTING INFORMATION

Additional supporting information may be found online in the Supporting Information section.

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