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# First network analysis of interspecific associations of abyssal benthic megafauna reveals potential vulnerability of abyssal hill community



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#### ABSTRACT

The distribution of organisms is related to both environmental factors and interactions between organisms. However, such associations between organisms across an abyssal megafaunal community have not previously been investigated at landscape scale because of a lack of positional data on specimens over such scales. We quantified spatial distributions and investigated interspecific associations in benthic megafaunal communities in three contrasting habitats on the Porcupine Abyssal Plain, two on the abyssal plain and one on the flank of a modest abyssal hill (~50 m above the plain). We used a Bayesian Network Inference Algorithm approach, which considers the ecosystem as a network, facilitated by robust positioning of specimens determined through seabed photography captured with an autonomous underwater vehicle. We found non-random intraspecific distributions of most morphotypes in all areas. The organisms in two interspecific networks on the abyssal plain had high connectance and link density, while the network at the Hill site was notable in the lack of inter-dependencies and highly dependent on one morphotype - Ophiuroidea. The reduced connectance of the hill community suggests that it is operating under a different regime and potentially more vulnerable to perturbation than those on the plain. Interspecific dependencies on the abyssal plain occurred across broad taxonomic groupings, and were thought to be a result of similar relationships between pairs of organisms and the substrate, and competition for detrital resource. In addition, some intraspecific pairs changed dependency direction at different scales. Our results suggest that the scales of inter- and intraspecific aggregation will be important considerations in the design of community assessments, and in spatial planning for their conservation.

#### 1. Introduction

## 1.1. Organism distribution in abyssal seabed habitats

The spatial distribution of organisms is related to environmental factors and biotic interactions between individual organisms in a habitat, and can influence ecosystem dynamics (Massol et al., 2011). Key to understanding the susceptibility or vulnerability of an ecosystem to change is the network of these interactions and associations between taxa (May, 1977), with communities of highly connected organisms (complex networks) more resistant to change than simpler networks (Rooney and McCann, 2012). These ecological networks could consist of only trophic relationships (food webs), but may also include habitat associations and inter-specific interactions, or a combination of all or some of these (Ings et al., 2009; Olff et al., 2009). The structure of these networks is crucial for understanding ecosystem dynamics, specifically

ecosystem stability and resilience (Proulx et al., 2005) via different processes including functional and/or trophic redundancy (Lawton and Brown, 1994; Thebault and Loreau, 2005), competition trade-offs (Wang and Loreau, 2016), stabilizing feedback loops (Mitchell and Neutel, 2012) and/or functional complexity (Van Voris et al., 1980). Therefore, by finding the ecological network of a community, we can understand its dynamics (D'Alelio et al., 2016).

Associations and interactions are commonly found because organisms are rarely randomly distributed (Taylor et al., 1978). Environmental filtering results in different communities whose composition and structure depends on the environment and processes such as resource limitation (Emerson and Gillespie, 2008), and provides a spatial pattern to community structure. Spatial variation in environmental factors, such as substrate type, provides a variety of niches for organisms with different traits; trait 'lability' provides flexibility for organisms to exploit different environments. Interactions occur between

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individuals within and between taxa, with some resulting in aggregation (e.g. predation, symbiosis, high competition for common resource). Thus, the spatial heterogeneity of the deep-sea benthos likely influences the distribution of organisms in communities there.

Spatial variation in environmental factors in the deep seabed habitat has been observed to alter the distribution of megafauna at regional (10 km) to local (metre) scales. Bathymetric features, from canyons (De Leo et al., 2010) to seamounts and abyssal hills (Clark et al., 2009; Durden et al., 2015; Durden et al., 2020a) and undulating topography (Simon-Lledo et al., 2019a,b), alter the near-seabed current flow, with consequent alterations to the seabed substrate and food availability (as suspended and deposited organic matter) (Turnewitsch et al., 2013: Turnewitsch et al., 2015). Hard substrate, such as nodules (Simon-Lledo et al., 2019a,b; Vanreusel et al., 2016) garbage (Bergmann and Klages, 2012; Schlining et al., 2013), and sponge spicule patches (Laguionie-Marchais et al., 2015), provides habitat for attaching organisms, and its availability varies at local (meter) to regional scales (10 km). Deposited organic material varies at hectare scale (Morris et al., 2016) and locally, settling into traces and furrows, with distributions of opportunistic fauna related to the patchiness of the resource (e.g. foraminifera, Gooday, 1993; holothurians, Kaufmann and Smith, 1997).

The influence of organism interactions on the distribution of deepsea benthic megafauna is less well understood, with many dispersion studies limited by small sample sizes. Intraspecific aggregations have been observed at scales of 1-100 m, such as 'herds' of the holothurian Kolga hyalina (Billett and Hansen, 1982), aggregations of holothurians Peniagone sp. and Elpidia minutissima (Lauerman and Kaufmann, 1998), Paroriza prouhoi (Tyler et al., 1992b), echinoids Echinus affinis and Phormosoma (Grassle et al., 1975), xenophyophores (Lauerman et al., 1996), and ophiuroids (Lauerman and Kaufmann, 1998; Smith and Hamilton, 1983). Such aggregations were thought to facilitate the maximal use of deposited food and reproduction (Tyler et al., 1992a). Fewer studies examine interspecific interactions between abyssal megafaunal taxa. Serendipitous observations of direct associations between abyssal taxa have been made using seabed photography, such as predation on a polychaete worm by the anemone Iosactis vagabunda (Durden et al., 2015b). Photography has also captured interactions between organisms accessing the same food source, such as the hosting of the anemone Kadosactis commensalis by the holothurian Paroriza prouhoi (Bronsdon et al., 1993), augmented with data from collected specimens. Similar methods have also revealed interactions between organisms with a common method of exploiting environmental conditions, such as the attachment of anemones to glass sponge spicules (e.g. Amphianthus bathybium and Daontesia porcupina) where both organisms extend into the water column above the seabed to exploit suspended material in areas of enhanced currents. Despite these observational efforts, no studies have examined interspecific associations between organisms across a megafaunal community on abyssal plains and considered the impact on ecosystem dynamics.

# 1.2. Bayesian network inference – A tool for examining organism associations

One approach to understanding ecosystem dynamics is to consider the ecosystem as a network (e.g. Mitchell and Neutel, 2012). Ecological networks can be computed using discrete Bayesian Network Inference Algorithms (BNIAs) with morphotypes considered "nodes" and dependencies between them (where the abundance of one morphotype depends upon that of another) described as "edges" which link correlating morphotypes together (see Section 2.3) (Heckerman et al., 1995). BNIAs can infer network structures and non-linear interactions, and have been used extensively to reveal gene regulatory networks (Yu et al., 2002, 2004), neural information flow networks and ecological networks (Smith et al., 2006, Milns et al., 2010) and more recently palaeontological communities (Mitchell and Butterfield, 2018). The structure produced by the BNIA reflects the associations caused by co-

localizations rather than a specific interaction. For example, a negative dependency between two morphotypes could correspond to competitive exclusion or exclusive niches and positive dependencies could reflect trophic interactions, habitat associations or facilitations. While it is not possible to infer the underlying processes resulting in dependencies or connectance from our BNI analyses alone, the most likely processes can sometimes be inferred using biological observations.

By using BNIA, direct dependencies between morphotypes can be detected, minimising auto-correlation between two variables. For example, if A depends on B which depends on C, there could be a correlation between A and C. However, this correlation would not represent an interaction or association between A and C, merely the two correlations between A and B and B and C, so the BNIA would report the edges A to B and B to C, and not report A to C. The BNIA approach enables only the realised dependencies to be found, ensuring only actual interactions and associations between morphotypes are found.

#### 1.3. Aim

We quantify spatial distributions and investigate interspecific associations in benthic megafaunal communities on an abyssal plain using the BNIA approach. The approach is facilitated by robust positioning of specimens determined through seabed photography captured with an autonomous underwater vehicle, covering an extensive seabed area (Morris et al., 2014). We investigate the benthic invertebrate megafaunal communities at hectare scale in three areas at the well-studied Porcupine Abyssal Plain (PAP) Sustained Observatory (PAP-SO), a longterm time series site at 4850 m water depth in the northeast Atlantic (Hartman et al., 2012), two on the abyssal plain and one on the flank of an abyssal hill (mean water depth 50 m above the abyssal plain). The sediment conditions and benthic community on the hill were recently found to be significantly and substantially different to those in the areas on the plain; sediments in this area were coarser but organic carbon content was lower, and total numerical and biomass density, diversity, and suspension feeder density were higher on the hills than on the abyssal plain (Durden et al., 2020a). Differences between the two areas on the abyssal plain were more subtle; differences in sediment particle size were not detected, but the compositions of the two communities were significantly different, related to differences in the densities of the two most common morphotypes (Ophiuroidea and the anemone Iosactis vagabunda), and in terms of biomass related to differences in contributions from large holothurians. These three areas provide contrasting habitats in which to examine abyssal ecosystem dynamics in terms of intra- and interspecific associations of benthic megafauna. We consider these associations in terms of ecosystem dynamics, including organism interaction with the seabed substrate and feeding modes.

#### 2. Methods

## 2.1. Megafauna in seabed photographs

Individual megafaunal specimens (>1 cm in size) were located and identified as one of 75 morphotypes in randomized downward-facing seabed images captured by an autonomous underwater vehicle. The seabed images were captured along transects spaced at 100 m intervals forming a 'fine scale grid' pattern in each of three locations: PAP Central (the location of sampling for the 30-year time series), on the flank of an abyssal hill ('Hill' grid) and on the abyssal plain to the northwest of this hill ('North Plain' grid), as described in Durden et al. (2020a). Full details of image capture and processing are provided in Morris et al. (2014); in brief, image processing involved some colour correction, the removal of overlap between adjacent images and combining  $\sim$  10 successive images into a 'tile' (seabed area  $\sim$  14 m²), and georeferencing the location of each tile based on the location of the vehicle (latitude/longitude) during image capture. A total of 3374 tiles (representing 44255 m² seabed) across the three grids were used in this

analysis. Some morphotypes consisted of groups of species (i.e. genus, family- or order-level identification), because species-level identification from images is not possible or practicable (e.g. Ophiuroidea, Elpidiidae spp., Porifera).

Locations of identified megafaunal specimens were derived from the respective tile locations, then transformed to a two-dimensional flat surface using the R package geosphere (Hijmans, 2019). The area of each 1 km  $\times$  1 km fine-scale grid was split into cells, so that each contained the same length of photographic transect (i.e. the same sampling intensity); cell sizes were 100 m  $\times$  100 m (1 ha). The positions of the megafauna within the tiles was at the sub-10 m scale, so negligible within each hectare cell. Morphotype densities were calculated within each cell, so that the associations found describe interactions and associations that occur between the cells; these density data were used in the computation of both intra-specific distributions and the inter-specific association networks.

#### 2.2. Analysis of non-random intra-specific distributions

In order to investigate the variation of intra-specific distributions in each photographic grid, chi-squared analyses of the counts per cell were performed in R (R Core Team, 2017). In this analysis, detectability of non-randomness is a function of abundance and strength of the pattern. That is, strong correlations can be detected with low abundances, and high abundances provide the power to find weaker correlations. Significance was reported at  $\alpha=5\%$ . In order to check that relative abundance was not driving the non-random intra-specific distributions we performed a Kruskal-Wallis test between the abundances and the chi-squared p-values.

#### 2.3. Bayesian network analysis of inter-specific associations

The Bayesian network (BN) consists of the best-fitting set of edges between nodes (i.e. the variables) which are found through a heuristic search mechanism Banjo (Smith et al., 2006) because finding the bestfitting BN is computationally intractable (Chickering, 1996). Assessment of a given BN is performed by calculating the Bayesian Scoring Metric, which calculates the probability that the network encodes of the statistical dependences of the observed data (Milns et al., 2010 and references within Appendix B). For ecological networks, "greedy" searches are an optimal technique for determining the network structure (Milns et al., 2010). Within Banjo greedy searches generate a random network, score how well the network fits the observed data and then randomly add edges to the network. If the score increases, then the edge is kept and another randomly added, and if the score decreases then it is removed. This process is repeated until there are no possible edge additions that increase the score and so corresponds to a local maxima within the search space. Because this best-fit BN is only a local maxima, the search needs to be repeated multiple times to ensure the global maxima is found. See Milns et al. (2010) and references within Appendix B for more details of the scoring and searching algorithms for ecological networks.

Bayesian network inference was performed in Banjo (Smith et al., 2006), with data preparation for Banjo (grouping and discretization) and statistical analysis in R (R Core Team, 2019). Further analysis of Banjo outputs, when required, used the functional language Haskell (Jones, 2003); scripts are available on Github (https://github.com/egmitchell/bootstrap).

#### 2.3.1. Data preparation – Discretization and contingency test filtering

The BNIA used requires discrete data, which ensures that correlations are not over-fitted to data noise and only the relative densities of each morphotype are important (Milns et al., 2010; Yu et al., 2004). Morphotype abundances by cell were split into three bins: zero counts, and counts below and above the median for the morphotype by site. For ecological species abundance datasets, three different bins have been

deemed to be a good balance between maintaining information in the dataset (achieved with many bins) and increasing statistical power (achieved with fewer bins; Yu, 2005).

In order to avoid Type I errors introduced by high numbers of morphotypes with zero counts, we excluded morphotypes which were found in fewer than 33% of the cells. Generally, the result is that low density (less common) morphotypes are excluded; however, this method of exclusion could potentially mean that a morphotype with high abundance in a very limited area is excluded from analyses. This latter type of exclusion happened in the Hill grid, where two taxa with sufficient overall abundance to be included (*Actinauge abyssorum* and *Peniagone* spp.), were excluded from the network analyses because of the aggregation of these taxa over a very limited area. While there may be subtle inter-specific correlations that we are unable to detect, it is unlikely that these less common morphotypes would have a significant influence on community dynamics.

To further guard against Type I errors, we also used a method of contingency test filtering that removed from consideration an edge between two variables whose joint distribution showed no evidence of deviation from the distribution expected from their combined marginal distributions (chi-squared tests, p>0.25; Milns et al., 2010). This threshold was used to ensure no chance of removing truly dependent dependencies, so that only artefacts, such as those found between high zero counts, were removed from consideration. These links were provided to the BNIA to exclude from consideration.

#### 2.3.2. Bayesian network inference

The BNIA software used was Banjo v2.0.0, a publicly available Java based algorithm (Bernard and Hartemink, 2006; Smith et al., 2006). For details of the algorithm please see Smith et al. (2006) and Milns et al. (2010). The discretized data was input into Banjo which then generated a random network based on the input variables. A 'greedy search' was repeated 10 million times for each set of input data and the most probable network was then output. The maximum number of edges leading to a node was set to 3 to balance predictive power with overfitting limit artefacts (Milns et al., 2010; Yu, 2005).

To minimise bias from outliers, we bootstrapped at 95% level (Magurran, 2013) by randomly-selecting 95% of the total number of grids cells for each subsample and then finding the subsample network using Banjo. For each edge calculated, the probability of occurrence was calculated, and the resultant distributions analysed find the number of Gaussian sub-distributions using normal mixture models (Fraley et al., 2012). This probability distribution was bimodal for each dataset, which suggests that there were two distributions of edges, those with low probability of occurrence, and those highly probable edges. The final network for each area was taken to be those edges which were highly probable. The threshold for being labelled 'highly probable' depended on the network (as determined by the normal mixture modelling analyses): 55% for PAP Central, 49% for Hill and 52% for North Plain. The magnitude of the occurrence rate is indicated in the network by the width of the line depicting the edge.

The direction of the edge between nodes in the network indicates which node (morphotype) has a dependency on the other node (morphotype); this direction is indicated in the network by an arrowhead. For each edge, the directionality was taken to be the direction which occurred in the majority of bootstrapped networks. Where there was no majority (directional edges have a probability between 0.4 and 0.6) the edge was said to have bi-directionality, or indicated a mutual dependency; these are shown without arrows).

The influence score (IS) can be used to gauge the type and strength of the interaction between two nodes. If the IS = 1, this corresponds with a positive correlation: that is, a high density of morphotype 1 corresponds to a high density of morphotype 2, and the dependency arrow would point from morphotype 2 to morphotype 1. An IS of -1 corresponds to a negative correlation: a high density of morphotype 1 corresponds to a low density of morphotype 2. An IS = 0 does not mean

Table 1 Total megafaunal morphotypes for the areas studied, morphotypes included in the analysis (those that occurred in > 33% of grid cells of 100 m  $\times$  100 m) and the number of connected morphotypes are those included in the Bayesian Network.

Study area	Total morphotypes	Morphotypes included in analysis	Number of connected morphotypes	Number of 100 m $\times$ 100 m cells
PAP Central	62	14	13	100
Hill	68	19	15	100
North Plain	59	8	7	80

there is no correlation between the two nodes, but rather that the correlation is non-monotonic so that the interaction is positive at low densities and negative at high densities (or vice versa). The mean IS for each edge was calculated for each site.

#### 3. Results

#### 3.1. Intraspecific spatial distributions

In the grids on the abyssal plain, 14 and 23% of all morphotypes were included in the analysis (at PAP Central and North Plain, respectively), while 28% of morphotypes in the Hill grid were included (Table 1). As these communities are dominated by few morphotypes (which were included), the included morphotypes accounted for the vast majority of specimens (92.6%; 92.5% and 85.3%, respectively). Seven included morphotypes were common to all grids; of these, three were mobile deposit feeders, one was a mobile predator/scavenger and three were sessile suspension feeders.

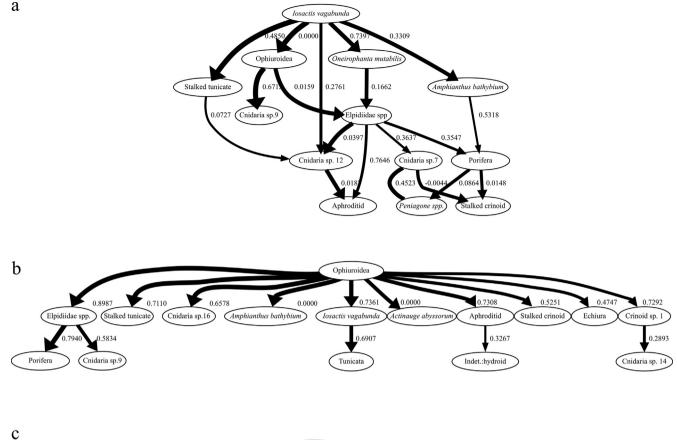
Total megafaunal density across all three study areas was non-random (Kolmogorov-Smirnov tests all p < 0.0001). The distribution of all fourteen morphotypes tested in the PAP Central fine grid were non-random (Table 2), while five of eight morphotypes in the North Plain grid were non-randomly distributed, and fifteen of nineteen morphotypes on the Hill randomly distributed. The three most common morphotypes, *Iosactis vagabunda*, Elpidiidae spp. and Ophiuroidea, were non-randomly distributed in all grids (Fig. 1). Morphotypes included in analyses of PAP Central and Hill grids with non-random distributions included Stalked tunicate, *Oneirophanta mutabilis* and Stalked crinoid. Cnidaria sp.9, Aphroditid and Porifera were also found to have non-random distributions at these sites, but random distributions in the North Plain grid. *Amphianthus bathybium* and Indeterminate (Indet.):



Fig. 1. Example specimens of notable morphotypes included in the interspecific distribution and intraspecific network analysis, clockwise from top left: Elpidiidae spp. (a morphotype that includes *Amperima rosea*, *Kolga hyalina* and *Elipidia minutissima*), burrowing anemone *Iosactis vagabunda*, *Amphianthus bathybium* attached to the stalk of Porifera, and Ophiuroidea. Elpidiidae spp. (a morphotype that includes Amperima rosea, Kolga hyalina and Elipidia minutissima, but not Peniagone spp.).

Table 2
Results of Chi-squared single-morphotype analysis of included benthic megafauna in  $100 \text{ m} \times 100 \text{ m}$  quadrats at each of the grids, ordered by number of specimens (n). Where p-value < 0.05 the morphotype was interpreted to be non-randomly distributed. Degrees of freedom are 99 for PAP Central; 99 for Hill and 79 for North Plain. Indet is short for Indeterminate.

PAP Central				Hill				North Plain			
Morphotypes	n	$\chi^2$	p value	Morphotypes	n	$\chi^2$	p value	Morphotypes	n	$\chi^2$	p value
I. vagabunda	12012	689.81	< 0.0001	Ophiuroidea	3619	2436.17	< 0.0001	I. vagabunda	1029	455.81	< 0.0001
Ophiuroidea	2672	252.41	< 0.0001	Elpidiidae spp.	2090	1411.19	< 0.0001	Ophiuroidea	878	274.70	< 0.0001
Elpidiidae spp.	1944	312.3	< 0.0001	I. vagabunda	932	620.51	< 0.0001	Elpidiidae spp.	264	336.75	0.001
Cnidaria sp.9	706	185.88	< 0.0001	Porifera	490	401.36	< 0.0001	Cnidaria sp.9	79	96.50	0.1759
Aphroditid	581	168.18	< 0.0001	Cnidaria sp.16	331	329.24	< 0.0001	Aphroditid	69	60.52	0.1212
Stalked tunicate	517	231.06	< 0.0001	Crinoid sp.1	318	353.94	< 0.0001	Stalked tunicate	50	161.41	< 0.0001
Cnidaria sp.12	330	110.19	0.4158	Aphroditid	287	268.04	< 0.0001	Porifera	33	74.88	0.7910
Amphianthus bathybium	284	134.24	0.0213	Stalked tunicate	280	393.67	< 0.0001	Amphianthus bathybium	31	76.88	0.9066
Oneirophanta mutabilis	279	133.84	0.0226	Cnidaria sp.9	253	314.49	< 0.0001				
Porifera	267	120.05	0.1475	Tunicata	229	191.2	< 0.0001				
Cnidarian sp.7	200	114.97	0.2603	Stalked crinoid	116	176.45	< 0.0001				
Peniagone spp.	180	125.81	0.0715	Echiura	94	143.88	0.0044				
Indet.: hydroid	156	122.13	0.1148	Amphianthus bathybium	79	181.76	< 0.0001				
Stalked crinoid	134	95.47	0.8364	Daontesia porcupina	78	118.49	0.1770				
				Oneirophanta mutabilis	76	158.77	0.0003				
				Indet.: hydroid	51	147.18	0.0024				
				Molpadiodemas villosus	50	152.3	0.0009				
				Cnidarian sp. 14	41	114.3	0.2790				
				Pseudostichopus aemulatus	41	110.72	0.3959				



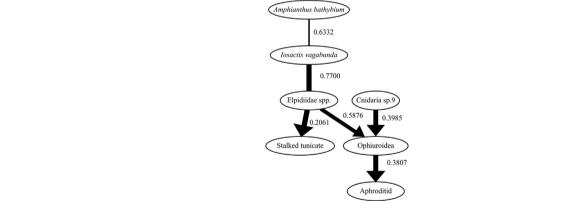


Fig. 2. Bayesian networks of associations between benthic megafauna observed in seabed photographs captured at three habitats at the Porcupine Abyssal Plain, (a) PAP Central, (b) Hill and (c) North Plain. Dependencies between morphotypes are indicated by the lines connecting the two morphotypes, the width of which indicates the occurrence rate in the bootstrap analyses (wider lines indicate higher occurrence). Arrows indicate nonmutual dependence between two taxa; mutual dependencies are indicated as lines without arrows. Mean interaction strengths of the correlations are indicated; positive interaction strengths indicating aggregation, negative interaction strengths indicating segregation, and zero indicating different aggregation and segregation behaviours at different densities. For more details, see 'Methods'. Morphotypes having no dependencies with other morphotypes are not shown (PAP central: Indet.: hydroid; Hill: *Pseudostichopus aemulatus, Oneirophanta mutabilis, Molpadiodemas villosus, Daontesia porcupine*; North Plain: Porifera).

hydroid were distributed non-randomly at PAP Central, but randomly in the Hill grid.

#### 3.2. Bayesian networks of interspecific associations

On the abyssal plain (PAP Central and North Plain grids), all but one of the included morphotypes in each grid were connected to others (92% and 88%, respectively), while in the Hill grid 77% of included morphotypes were connected (Table 1). These results were not reflections of relative abundance – we performed a Kruskal-Wallis test to compare the total abundances for each taxa to the p-values of the intra-

specific chi-squared tests (from Table 2) between the spatial quadrats and found that there was not a significant correlation (p > 0.05).

Our Bayesian network inference found that all three areas had multiple dependencies between morphotypes (Fig. 2). The majority of the dependencies were positive (Table 3), with only positive dependencies for North Plain grid (6 of 6), 13 positive dependencies for Hill grid (of 15) and 17 positive dependencies (of 19) for PAP Central. There was only one negative dependency, in PAP Central. Where the interaction strength (IS) was zero, this indicates different behaviour at different densities, of which there were three in PAP Central and one in Hill grid. The highest mean IS was in the Hill grid network (0.6267) and

network, excluding the zero IS values. Connectance is the number of realised interactions from the total possible number (May, 1973), while Link Density is the mean number of edges per node; both computed including the of interspecific Bayesian network analysis of megafaunal morphotypes in each of the three photographic grids at the Porcupine Abyssal Plain. The 'mean IS' is the mean interaction strength of the connected

unconnected 1	nodes. Co	nnectance2 and Li	nk Density 2 are simila	ır metrics, but include on	ıly the nodes with	hin the netwo	rk. Max chair	ı <i>lengt</i> h is the la	rgest number of morphoty	unconnected nodes. Connectance2 and Link Density 2 are similar metrics, but include only the nodes within the network. Max chain length is the largest number of morphotypes linked by a linear chain of dependencies.	in of dependencies.
	mean IS	mean IS Positive dependencies	Negative dependencies	Non-monotonic dependecies	Connectance	Connectance Connectance2 Link Densi	2 Link Density	Link Density 2	Link Density Max dependency chain length	Mean dependency chain length	Number of chains
PAP Central 0.2989	0.2989	17	1	1	0.1124	0.2436	1.4615	1.3571	5	4.27	11
Hill	0.6267	13	0	2	0.0586	0.1250	0.9375	0.70704	2	2.00	4
North Plain	0.4960	9	0	0	0.1224	0.2857	0.8571	0.7500	5	4.00	3

the lowest in PAP Central (0.2989), with a mean IS for North Plain grid of 0.4960. Most dependencies were directional, with no mutual dependencies on Hill grid, two out of six mutual dependencies on North Plain grid and one mutual dependency on PAP Central.

The most complex area was PAP Central, exhibiting high connectance and link density (Table 3), highest dependency chain length of 11 morphotypes. The Hill grid was notable in the lack of inter-dependencies (Fig. 2, Table 3), with half the connectance of the grids on the abyssal plain. Both the PAP Central and North Plain grids had similar mean chain length, maximum length of dependency chain of 5 and connectance, but the North Plain grid differed in having fewer chains and lower link density than PAP Central grid.

#### 4. Discussion

#### 4.1. Contrasting interspecific network structures in contrasting habitats

We found substantial differences in the interspecific associations of the benthic megafauna in the three sites studied. The networks of fauna in both grids on the abyssal plain are highly connected, while connectance in the network at the Hill is low. The dependency on a single morphotype (Ophiuroidea) within the Hill grid suggests that this network has little redundancy, so changes to the density of Ophiuroidea may spur substantial variation in the dynamics of this community. Connectance correlates with increased ecosystem stability and robustness independently of species richness (Dunne et al., 2002; Gardner and Ashby, 1970), through the increase of stabilizing processes such as of stabilizing feedback loops of (Mitchell and Neutel, 2012) so that the significantly lower connectance of the Hill site suggests that it is less robust to perturbations that the plain sites.

The starkly different network structure at the Hill grid suggests different community dynamics from the abyssal plain community. Megafaunal communities at the Hill and plain sites have different community structures (Durden et al., 2020a), which are likely related to increased hard substrate availability and coarser sediments on the hill, a result of hydrodynamic conditions (Turnewitsch et al., 2015). In addition, the difference in the abundance and biomass k-dominance plots (a measure of the successional state of a community; Clarke, 1990; Warwick and Clarke, 1994) was similar for the two grids on the abyssal plain, but significantly different on the hill, suggesting that the community there was in a different successional state, or more recently disturbed (Durden et al., 2020a). The difference in successional state may be related to the differences in connectance and dependence found in the networks; specifically, the lower connectance of the Hill network and the suggestion that it is less robust to perturbations than the communities on the abyssal plain may be related to recent disturbance there.

It is also likely that the communities of the Hill and abyssal plains are subject to contrasting organizing forces with respect to available detrital resource. Deposit-feeding communities generally exist in trophically stable environments, which may be resource limited, while suspension-feeding communities exist in trophically more variable and resource-independent environments (Levinton and Kelaher, 2004). These generalities appear to apply to the PAP-SO megabenthos, where the community on abyssal hills is known to be more trophically diverse than on the plain (Durden et al., 2015; Durden et al., 2020a), and nearseabed detrital inputs in these two habitats have been estimated to be different in both quantity and its partitioning in the community (Durden et al., 2017). Furthermore, both top-down and bottom-up processes influence patch size in deposit feeding communities, such as the abyssal plain, with some local-scale patchiness likely introduced by localized inputs of detritus. These localized inputs of detritus may in turn influence the distribution of organisms with less mobility, while highly mobile deposit feeders alter the patchiness through grazing at distances greater than the existing patch size (Levinton and Kelaher, 2004). Unfortunately, detrital patch sizes at the scales examined in this

study have not yet been established at PAP-SO.

#### 4.1.1. PAP Central grid

The relatively high degree of connectance between morphotypes in the PAP Central grid involves connections between mobile and sessile organisms, and also between morphotypes that may feed on similar detrital fractions. The three most abundant megafaunal organisms on the abyssal plain, Iosactis vagabunda, Elpidiidae spp. and Ophiuroidea (Durden et al., 2015; Durden et al., 2020a), are major components of the network; at least one of these three morphotypes is involved in 12 of the 19 connections in this network. The base morphotype in most chains (with IS value > 0.3) is a sessile organism (Porifera, Cnidaria sp.7. Cnidaria sp.9 or Stalked tunicate), likely attached to hard substrate such as clinker or deposited ice-rafted dropstones (as previously observed in images; Durden et al., 2015), to which one of these three mobile organisms is dependent. This result suggests an aggregation of mobile morphotypes around the hard substrate and/or hard substrateattaching fauna, a behaviour observed for similar benthic megafauna around Antarctic dropstones (Chickering, 1996; Ziegler et al., 2017). An intermediary is Amphianthus bathybium (Fig. 1), which is itself associated with Porifera as it attaches to siliceous sponge spicules (Riemann-Zürneck, 1987). Sponges and hard substrate-attaching fauna provide three-dimensional structures in the deep sea, with which mobile fauna have been observed to be associated (Beaulieu, 2001; Buhl-Mortensen et al., 2010; Lacharité and Metaxas, 2017).

The mobile morphotypes in the network, including the holothurian Oneirophanta mutabilis, are surface deposit feeders, and thus individuals are likely co-located in order to feed on patches of deposited detritus. Some mobile surface deposit feeders alter their feeding behaviour related to the availability of detritus (Durden et al., 2020b), or movements to stay proximate to an area of detritus (Kaufmann and Smith, 1997). Environmental filtering related to detritus availability is likely important; aggregation suggests high competition between organisms for the shared detrital resource with low availability. The holothurians Elpidiidae spp. and Oneirophanta mutabilis differ in terms of trophic level based on δ<sup>15</sup>N values (Iken et al., 2001), the latter grazing approximately 5 times faster (Durden et al., 2019). The trophic levels of specimens of several Ophiuroidea are similar to each of Elpidiidae spp. and O. mutabilis (Iken et al., 2001). The hemisessile Iosactis vagabunda moves at a slower speed, moving burrow locations approximately every 20 days (Durden et al., 2015b) to exploit deposited detritus around the burrow. Its trophic position is considerably higher, likely because of its occasional predation; I. vagabunda occupies a similar trophic position to the Stalked tunicate based on  $\delta^{15}N$  values (Iken et al., 2001), which may explain their association in the network.

The negative interaction between Cnidarian sp.7 and Stalked crinoid indicates spatial separation of these suspension-feeding morphotypes. This negative interaction is likely due to niche filtering since suitable hydrodynamics may be facilitating their suspension feeding. It may be simply related to substrate texture, with Cnidarian sp.7 utilising soft substrate and Stalked crinoid requiring hard substrate.

#### 4.1.2. Hill fine-scale grid

The network of the Hill grid is characterised by the major role of Ophiuroidea, which was the most abundant morphotype at the Hill, and more abundant there than at the sites on the abyssal plain (Table 2). Ophiuroidea at the Hill were more strongly associated with Elpidiidae spp. and *Iosactis vagabunda* that at PAP Central. These latter two morphotypes were less abundant on the Hill than on the abyssal plain, but were each connected in the network to sessile morphotypes (Porifera, Cnidaria sp.9 and Tunicata), as on the plain. Ophiuroidea at the Hill was also associated directly and individually with 5 other sessile suspension feeding morphotypes and three other mobile ones, some of which it was associated via Elpidiidae spp. at PAP Central (e.g. Stalked crinoid, Aphroditid). One major chain was unbroken from the PAP Central network: from Ophiuroidea to Elpidiidae spp. to Porifera. The

'Ophiuroidea' morphotype consists of all ophiuroids (including at least *Ophiocten hastatum* and *Ophiomusium lymani*), and interspecific differences in feeding habits are likely – four different morphotypes studied by Drazen et al. (2008) and Iken et al. (2001) had different  $\delta^{15}$ N values. Thus, many of the associations with Ophiuroidea may be related to its feeding modes, including surface deposit feeding, or predation/scavenging, in similarity with the polychaete Aphroditid. Such flexibility of feeding mode may be advantageous in the hill environment, where proportionally less organic matter is deposited on the seafloor, as some suspended material is likely removed by suspension feeders before deposition (Durden et al., 2017).

#### 4.1.3. North plain grid

The morphotypes in the North Plain grid were highly connected, as in the other abyssal plain grid, despite the network being less complex. As in the PAP Central grid, strong associations between the mobile deposit feeding Iosactis vagabunda, Elpidiidae spp. and Ophiuroidea are likely related to these organisms aggregating around patches of detritus. Similarly, the network analysis suggests that these mobile morphotypes are found to aggregate with sessile organisms (in this case, Amphianthus bathybium and Stalked tunicate Fig. 2). I. vagabunda was found to have mutual dependencies with A. bathybium and Elpidiidae spp., in contrast to the relationships of these two pairs at PAP Central. At PAP Central, I. vagabunda is dependent on A. bathybium and on Elpidiidae spp., but the latter relationship is via dependencies on Ophiuroidea and Oneirophanta mutabilis (which is not present at the North Plain). Thus, the dependency of Elpidiidae spp. on Ophiuroidea at the North Plain is the reverse of the dependency between these morphotypes at PAP Central. These differences in the network of the North Plain from that of PAP Central suggest that the relative decrease in the density of I. vagabunda and the increases in the densities of Ophiuroidea and Elpidiidae spp. alter the relationships between these morphotypes and potentially the community function there.

#### 4.2. Contrasting interspecific distributions

Despite most common morphotypes having non-random distributions across both abyssal hill and plain habitats, some morphotypes had different interspecific distributions between these habitats. The anemone Amphianthus bathybium was non-randomly distributed at PAP Central, but randomly distributed in the other two grids (Table 2, Fig. 1). Porifera, its host, was similarly non-randomly distributed at PAP Central, and also in the Hill grid, but randomly distributed at the North Plain. Again, where the distributions were determined to be random, this randomness may be a true signal or the abundances of these morphotypes and the weakness of the signal may be too low to detect any patterns. Note though that because there is no significant correlation with abundance and non-random spatial behaviour (p = 0.4748) any non-random signal for low abundance morphotype would have to be weak to not be detected. Similarly, the lack of detection of any dependency between these two morphotypes at the Hill and North Plain grids, in contrast to those found at PAP Central, may also be related to relative low densities of one or both morphotypes at these sites. However, both live and dead Porifera provide substrate to which A. bathybium attaches (Riemann-Zürneck, 1987), and the distribution of such stalks was found to be random at another abyssal site (Beaulieu, 2001), complicating the analysis of this morphotype association pair.

#### 4.3. Spatial considerations

Organisms may aggregate or associate differently in different areas and at different scales, a situation which occurred at the abyssal plain sites for some organism pairs. At PAP Central, the dependency direction between Cnidarian sp.7 with Peniagone spp., a minor pair in terms of abundance, was scale-dependent. By contrast, at the North Plain site,

both dependencies of the dominant *Iosactis vagabunda* (with *Amphianthus bathybium* and with Elpidiidae spp.) changed orientation with scale. This suggests that at some scale, Elpidiidae spp. may be highly dependent on others, with the network being highly centred on this organism.

The sizes of the photographic grids and the cells determined the maximum and minimum of spatial scales at which interspecific distributions and intraspecific associations would be detected; the number of grid cells was a balance between having sufficient available data per cell, and sufficient cells for analysis. Thus, our analyses would not detect within-cell (< 100 m) patchiness and aggregations, nor nonrandom behaviour which occurs over the kilometre scales of the fine-scale transects. Patchiness and aggregations that occur between 100 m and 1 km could be detected. These limitations, in combination with the observed densities and community structure at this scale, determined the included morphotypes and detectable network interactions.

Only a small fraction (13-28%) of the morphotypes at each site were sufficiently abundant to be included in the intraspecific dispersion and interspecific network analyses. This was related to the density dominance of the megafaunal assemblages by a few organisms at this scale, particularly in the two grids on the abyssal plain (Durden et al., 2020a). The included morphotypes are generally small to medium in size, particularly in the two abyssal plain grids; the larger morphotypes, such as Psychropotes longicauda, Paroriza prouhoi, Benthodytes spp. and Benthothuria sp., are present at low densities (Durden et al., 2020a), precluding their inclusion in the analysis. Thus, their distributions remain unknown and the networks represent only a portion of the potential morphotype associations within each assemblage. Two large holothurians, Pseudostichopus aemulatus and Molpadiodemas villosus, were included in the analysis of the Hill grid, but were not found to have non-random distributions or associations with other morphotypes there. As these omitted morphotypes are generally large, mobile fauna, likely with larger ranges given their fast locomotion (Durden et al., 2019), they may have non-random distributions or associations at spatial scales larger than that of this study.

The exclusion of low density morphotypes in combination with the spatial scale of the study resulted in some known associations of megafauna at the PAP not being identified. For example, pairing in the large holothurian *Paroriza prouhoi*, and non-random co-occurrence between epi- and basibionts, such as *P. prouhoi* acting as host for the anemone *Kadosactis commensalis* (Bronsdon et al., 1994; Bronsdon et al., 1993), or the anemone *Actinauge abyssorum* residing on siliceous sponge fibres and polychaete tubes (Riemann-Zürneck, 1986). These associations involve direct interactions, at scales of less than 1 m, and thus would not have been detected even if the large holothurians (including *P. prouhoi*) were included in the analysis. The distributions and associations of morphotypes that exist at low densities at this scale, and those that aggregate at small spatial scales, and their impacts on community structure, could be assessed with more high density seabed photography.

# 4.4. Temporal considerations

This analysis provides a snapshot of the community as photographed in summer 2012. However, megabenthic communities in abyssal habitats are known to vary in activity and composition at interannual and seasonal time scales (Billett et al., 2010; Kuhnz et al., 2014; Durden et al., 2020b), with likely alterations to the patterns of intraspecific and interspecific aggregation. Interannual differences in the structure of the community has been previously found to alter its function in terms of carbon flow at this site (Durden et al., 2017) and another abyssal plain site (Dunlop et al., 2016). Densities of Amperima rosea, a small holothurian and component of Elpidiidae spp., and Ophiuroidea have fluctuated significantly interannually at the PAP-SO site (PAP Central here); at the time of this survey, Elpidiidae spp. occurred at densities in between the boom and bust densities found

previously (Billett et al., 2010; Billett et al., 2001). As one of the most common megafaunal morphotypes, significant fluctuations in its density could alter interspecific aggregations in the community. Seasonal reproduction in deep-sea echinoderms (Tyler et al., 1982) may involve periodic intraspecific aggregations of holothurians (e.g. Tyler et al., 1992b). Seasonal detrital inputs may induce aggregations of organisms exploiting the deposited detritus, since many abyssal deposit feeders are selective feeders (FitzGeorge-Balfour et al., 2010; Ginger et al., 2001). Patch selectivity in deposit feeding holothurians (Uthicke and Karez, 1999) is exemplified by the loop or run-and-mill behaviour of Oneirophanta mutabilis, a taxon presumed to feed on patches of detritus (Kaufmann and Smith, 1997). The periodic input of detritus also induces seasonality in the seabed surface activity of some burrowing organisms (Durden et al., 2020b). Thus, the community structure and inter- and intraspecific aggregations likely also vary seasonally.

#### 5. Conclusions

We found a complex network of interspecific associations in two abyssal plain communities, and a contrasting simple network on a modest abyssal hill (~50 m above the plain), where the community was highly dependent on a single morphotype. The significant differences between these networks suggest that the communities are operating under contrasting regimes, likely with different environmental filtering (related to hard substrate and organic matter availability) or different organizing forces (related to differing faunal function, such as through flexible feeding modes). These results also suggest the vulnerability of the megafaunal community on the abyssal hill to perturbations, as opposed to the relative robustness of the abyssal plain communities. Thus, disturbances to the sedimentary environment, such as from sediment deposition as a result of deep-sea mining (Jones et al., 2017), or alterations to the deposition of detrital material, which is likely to result from climate change (e.g. Jones et al., 2014), may impact the structure and dynamics of communities in these two habitats differently. Furthermore, the spatial scales of heterogeneity and aggregation for organisms of interest will be important to the robust design of any assessment of megafaunal communities in abyssal benthic habitats, including baseline environmental assessments (e.g. for Environmental Impact Assessments; Clark et al., 2017; Durden et al., 2018) and assessments of ecosystem change (e.g. environmental monitoring), or spatial planning for their conservation (e.g. Jones et al., 2018). These results also demonstrate the importance of considering the interactions of communities as a whole when assessing ecosystem vulnerabilities.

## **Declaration of Competing Interest**

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

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