Reef complexity influences distribution and habitat choice of the corallivorous seastar *Culcita schmideliana* in the Maldives

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**Abstract** The cushion seastar *Culcita schmideliana* has gained major attention in the last few years because of its selective predation on juvenile corals, as well as its ability to generate large demographic assemblages, causing delays in coral recovery after large mortality events in the Republic of Maldives. However, a lack of data regarding the factors affecting its distribution and habitat selection still persists in this area. Here, we adopted a novel approach in the study of corallivorous seastar habitat selection that combined ecological and digital photogrammetry data. In this regard, we tested 3 different parameters as factors influencing seastar habitat choice in the South-East region of Faafu Atoll, Republic of Maldives, namely prey abundance, Linear Rugosity Index (LRI), and Average Slope (AS). The analysis of selectivity coefficient (Ei) of seastars for different habitat types showed a preference for reefs characterized by medium AS values (Ei = 0.268), a LRI included between 2 and 2.5 (Ei = 0.180), and a juvenile coral density ranging between 10 and 20 colonies m⁻² (Ei = 0.154). A multiple linear regression analysis showed that different AS and LRI values explained the 43.1% ($R^2 = 0.431$, $P = 0.007$) and the 48.1% ($R^2 = 0.481$, $P = 0.024$) of variance in seastars abundance, respectively, while juvenile coral densities did not significantly affect this ($R^2 = 0.132$, $P = 0.202$). These results provide new information on the distribution and behaviour of an important corallivore of Maldivian reefs, such as *C. schmideliana*.

**Keywords** Corallivory · Cushion seastar · Habitat selection · Structure from Motion

**Introduction**

Predation on scleractinian corals is a common and well described ecological phenomenon in coral reef ecosystems (Rice et al. 2019). Corallivores are a trophic guild that encompass several taxa of vertebrate and invertebrate organisms adopting different strategies for coral consumption, i.e. removing mucus, tissue, or skeleton from the cnidarian prey (Cole et al. 2008; Rotjan and Lewis 2008). Among coral predators, corallivorous seastars represent a major plague for tropical reefs, in particular because of massive population outbreaks of the crown of thorns seastar (*Acanthaster* spp.), widely described to cause extensive coral mortality and shifts in coral assemblages, with implications for the whole reef community (Pratchett 2010; Kayal et al. 2012; Baird et al. 2013). The effects of corallivorous seastars are particularly grievous considering the large number of stressors affecting coral reefs’ integrity in the last years, such as extreme climatic events and subsequent mass bleaching, or increased pollution (Carpenter et al. 2008; Patricola and Wehner 2018; Rice et al. 2019). In this context, it is important to consider the role of other coral feeding seastars, which, despite being still poorly studied, have been shown to affect reef resilience, in particular following large coral mortality events (Raj et al. 2018; Mah 2020).
Cushion seastars (*Culcita* spp.) have been reported as resident organisms of Indo-Pacific coral reefs, representing a persisting element affecting benthic community structure and integrity (Raj et al. 2018; Bruckner and Coward 2019). Despite sharing some common features with crown of thorns seastars, such as the same predation modality and similar larval ecology, *Culcita* spp. generally consume lower amounts of coral prey per year than *Acanthaster* spp., and show a lower fecundity (Glynn and Krupp 1986; Birkeland and Lucas 1990; Quinn and Kojis 2003). However, cushion seastars show several peculiar feeding habits. In particular, they exert a preferential predation towards juvenile corals of a few genera, such as *Pocillopora* and *Pavona*, directly compromising the reef community structure and composition, as well as its recovery and resilience, as previously described for Maldivian coral reefs (Bruckner and Coward 2019; Montalbetti et al. 2019a). Indeed, in the Republic of Maldives *Culcita schmideliana* has gained special attention due to its ability to generate large demographic assemblages, with seastar densities often exceeding those recorded for *Acanthaster planci* in the same area, causing delays in reef recovery (Saponari et al. 2018; Bruckner and Coward 2019; Montalbetti et al. 2019b). Although the feeding preferences and different ecological patterns of *C. schmideliana* in the Maldivian archipelago have been already explored (Bruckner and Coward 2019; Montalbetti et al. 2019a; Montalbetti et al. 2021), the biotic and abiotic factors affecting the distribution of seastars are still unknown and need further study. Among these, prey availability has been hypothesized to represent an influencing factor controlling the distribution of cushion seastars, as demonstrated for other corallivorous organisms (Brooker et al. 2013; Wolf et al. 2014; Saponari et al. 2021). In addition, it has been hypothesized that the limited mobility of the cushion seastar, due to its round-shaped unarmed body, could compromise its ability to climb on and consume large coral colonies, preventing its ability to move on complex three-dimensional coral reefs (Thomasin 1976; Hawkins 2006). However, no experimental evidence has been provided so far.

Reef structural complexity is primarily influenced by coral species abundance and composition and by the physical conditions in which corals grow (Luckhurst and Luckhurst 1978; Todd 2008), and has been largely reported to influence biodiversity and community structure of coral reefs at multiple scales (Richardson et al. 2017; Price et al. 2019). Among the wide spectrum of habitat metrics used to quantify coral reefs habitat structure (Fukunaga et al. 2019), Linear Rugosity Index (LRI) and Average Slope (AS) represent two efficient indices to estimate habitat morphological complexity, providing a useful instrument that can numerically express bottom profile heterogeneity (Risk 1972; Friedman et al. 2012, Dustan et al. 2013; Burns et al. 2016; Pascoe et al. 2021). Linear rugosity has been defined as the comparison of a straight line transect with a flexible chain draped over the reef along the same transect, while the slope indicates the maximum rate of elevational change from each cell to its neighbours in units of degrees (Risk 1972; Burrough and McDonell 1998). In previous studies, these indices have been used to correlate reef complexity with the abundance of reef organisms, such as reef fishes (Risk 1972; Knudby and LeDrew 2007; Kuffner et al. 2007; Gonzalez-Riviero et al. 2017; Nugraha et al. 2020). Furthermore, recent advance in photogrammetry techniques and the application of Structure from Motion (SfM) algorithms allowed to obtain 3-dimensional (3D) models from sequences of overlapping images acquired from multiple viewpoints (Westoby et al. 2012; Fonstad et al. 2013; Leon et al. 2015; D’Urban et al. 2020). The application of these methods allows to calculate topographic metrics of reef areas with high precision, reducing costs and time necessary for surveys and providing more reliable values of reef complexity (Figueira et al. 2015; Storlazzi et al. 2016; Urbina-Barreto et al. 2021; Anelli et al. 2017; Rossi et al. 2021; Ventura et al. 2021).

The aim of the present study was to investigate different factors possibly influencing *C. schmideliana* distribution and habitat choice on Maldivian reefs. Specifically, we tested the hypothesis that *C. schmideliana* distribution could be influenced by biotic factors, such as the availability of juvenile coral prey, and abiotic factors, such as 3D complexity. For this purpose, we designed field experiments that combined digital modelling of reef geomorphology with ecological data, in order to characterize seastar selectivity for areas displaying different values of these biotic and abiotic parameters.

**Materials and methods**

**Study area**

The logistic base of the sampling activity was the MaRHE Center on the inhabited island of Magoodhoo (3°4′49.08″N, 72°57′57.19″E), Faafu Atoll, Republic of Maldives (Fig. 1). In the South-East region of the atoll, six different sites were chosen, depending on their accessibility, their geographical location, and their geomorphological and environmental features. In particular, three sites, namely Dharamboodhoo reef, Dhigu reef, and Route 66 reef, were located on the external ocean-facing side of the atoll rim, usually subject to intense hydrodynamic conditions and characterised by steep walls, and thus classified as “outer” reefs. The other three sites, namely Freeclimbing reef, Maaga reef, and Sunny reef, were located inside the atoll rim, exhibiting typical low-energy
reef features with several different growth morphologies of corals, gentle slopes, and lagoon-patch reefs, and thus classified as "inner" reefs (Montano et al. 2012) (Fig. 1).

*Culcita schmideliana* and juvenile coral abundance

The distribution of *C. schmideliana* was estimated through the belt transect method at each sampling site. Specifically, three depth ranges (0–5 m, 5–10 m, and 10–15 m) were considered for each site, and three 50 × 4 m belt transects (200 m² each) spaced at least 10–20 m apart were randomly placed at each depth zone, for a total of nine transects per site. In each transect, the total number of seastars encountered was recorded and the maximum diameter of each specimen was measured to the nearest centimeter (cm) using a ruler.

We considered as juvenile corals all those colonies with a maximum diameter lower than 10 cm, since these specimens have been demonstrated to represent the preferred coral prey of *Culcita* spp. (Glynn and Krupp 1986; Norström et al. 2007; Montalbetti et al. 2019a). In order to estimate their abundance, a minimum of eight PVC quadrat of 1 × 1 m were randomly placed within the area covered by each belt transect. Photographs of quadrats were taken using a Canon G7X Mark II camera in a Fantasea FG7XII underwater housing, and then analysed with the ImageJ software (http://rsb.info.nih.gov/ij/). Each quadrat was scaled through dimensional references present on the PVC frame, and the maximum diameter of coral colonies within the square area was measured. The total number of juvenile corals was then recorded for each quadrat and the density per m² was calculated for each depth zone and site.

Underwater photogrammetry data acquisition

The image acquisition was done following the metric tape as ground referencing, in correspondence with belt transects for each sampling site, using a Canon G7X Mark II camera in a Fantasea FG7XII underwater housing. A dive computer was used to record the depth at each end of the metric tape.

For each depth range, photographs were collected with a fixed acquisition protocol along two transects of 20 m (Fig. 2). Approximately 250 images were shot for each transect covering an area of 500/60 m². Specifically, shots were taken continuously (with a rate of 1 photo every 1/2 s) while swimming roughly two meters above the monitored areas beside two parallel paths. The camera lens

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**Fig. 1** a Satellite overview of the Maldivian archipelago. b Satellite close-up of the study area corresponding to South-Eastern region of Faafu Atoll. Sampling sites are marked in red. The position of MaRHE Center is marked in blue (Basemap sources: Esri, DigitalGlobe, GeoEye, i-cubed, USDA FSA, USGS, AEX, Getmapping, Aerogrid, IGN, IGP, swisstopo, and the GIS User Community)
was kept parallel to the substrate, and the frontal overlap between two consecutive images was always higher than 80%, while the lateral overlap was 70%. Finally, to complete the survey, the diver took additional shots with the camera lens at a 45° angle along the same paths. The overall time for image acquisition along a single transect was ~20 min.

**SfM data processing**

The software Agisoft Metashape Professional (https://www.agisoft.com/) was used to perform SfM processing on the acquired images. The photo sets of each transect were processed following a standard SfM workflow (Figueira et al. 2015; Agisoft 2018; Fallati et al. 2020) divided into four main steps. 1) Overlapping images were aligned by identifying common points, using a high accuracy setting, achieving a 3D sparse point cloud. 2) A dense point cloud was generated from the sparse cloud, with high quality and moderate filter settings. 3) A Digital Elevation Model (DEM) was obtained from the dense cloud. 4) A high-resolution orthomosaic was derived from the DEM as the final output. Moreover, in order to correctly scale the models, we used the metric tape as a known length object in the scene to create scale bars (Leon et al. 2015; Ferrari et al. 2016, 2017). Specifically, after the photos alignment, we placed on the tape four markers spaced 10 cm apart, in order to obtain DEMs and orthomosaics correctly scaled for each sampling site (Young et al. 2017). Then, the models were imported into a GIS environment for further analysis.

**Structural complexity evaluation**

The orthomosaics and DEMs were imported in ArcMap 10.8 (ESRI) and analyzed using the 3D Analyst Tools to extrapolate LRI s and AS along the transects (Burns et al., 2019; Magel et al. 2019; Fallati et al. 2020). First, a virtual transect of 20 m was laid along with the metric tape (clearly visible on screen thanks to the ultra-high resolution of the model, i.e. 0.02 cm) on the central part of each processed orthomosaic and DEM. We used the function “Interpolate shape” to add the z-values, associated with the grids of the DEM, to the linear transect. Then, using the function “Add surface information” we calculated the 3D distance of the line. The LRI was calculated as the distance of a transect that considered changes in vertical surfaces (3D), divided by the linear distance of a transect (2D):

\[
LRI = \frac{3D\text{Length}}{2D\text{Length}}
\]

Values of LRI close to 0 indicate a lower three-dimensionality, while larger values indicate a higher degree of morphological complexity.
In addition, the Average Slope (AS) along the transects was estimated using “Functional surface algorithms” included in the 3D Analyst Tools. AS was obtained by weighing each slope, calculated at each segment along the line, by its 3D length, and then determining the average. The output slope values were calculated as ‘percent slope,’ which is the rise divided by run (tangent θ) multiplied by 100 (Burrough and McDonell 1998; Burns et al. 2015, 2016; Fallati et al. 2020). A small AS value reflects flat terrain, while a large AS value infers a steep terrain.

Data analysis

Data normality was verified through Shapiro–Wilk test. For purpose of comparison, seastars densities have been reported as seastars ha\(^{-1}\) (Bruckner and Coward 2019; Montalbetti et al. 2019b). Two-way ANOVA followed by Tukey’s post-hoc test was performed to evaluate significant differences in seastar abundance in the different depth zones, and in different reef types (outer vs inner). The AS values obtained in different sites were grouped into three categories: “low” (< 1000), “medium” (1000–2000), and “high” (> 2000). These were calculated considering the maximum and the minimum slope recorded, and symmetrically dividing this range into three bins. Similarly, LRI values were grouped in three categories (< 2, 2–2.5, and > 2.5). In order to evaluate significant differences of seastar size between different AS categories, one-way ANOVA followed by Tukey’s post-hoc test was performed. Kruskal–Wallis test followed by multiple pairwise comparisons was used to test significant differences in seastar size among different LRI categories, since data did not meet the assumption of normality (Zar 1999).

Seastar habitat preferences for different AS and LRI categories and different coral juvenile densities (classified in four density categories: < 5, 5–10, 10–20, > 20 juveniles m\(^{-2}\)) were tested through Van der Ploeg and Scavia selectivity coefficient (Ei), following Lechowicz (1982). This coefficient is defined for a group i as:

\[
E_i = \frac{W_i - \bar{W}}{W_i + \bar{W}}
\]

where Wi represents the value of Chesson’s α, and n represents the number of habitat types (Lechowicz 1982). Chesson’s α value (Wi) is defined as:

\[
W_i = \frac{\sum ri}{P_i}
\]

where ri represents the frequency of a habitat category (AS, LRI, and juvenile density) in the environment, and P represents the frequency of the same habitat category in which the organism of interest is found (Chesson 1978). Values of selectivity coefficient range between -1 and 1, with -1 meaning complete avoidance of a habitat category, and 1 meaning exclusive preference for a habitat type (Van der Ploeg and Scavia 1979).

A multiple linear regression analysis was performed to examine the relationship between the abundance of seastars and the AS, LRI, and juvenile coral density. All statistical analysis were performed using the SPSS software version 27 (IBM, New York), and data are presented as mean ± standard error (SE), unless otherwise stated.

Results

Habitat characterisation

Thirty-six ultra-high-resolution 3D models were generated from the 7946 images acquired in correspondence of the six sampling sites. The average DEMs cell size was 1.2 ± 0.5 mm, while the orthomosaics had a pixel resolution of 0.4 ± 0.07 mm. The average LRI of the whole investigated area was 2.02 ± 0.10, with the maximum complexity reached in Dharamboodhoo reef (2.48 ± 0.33), which also showed the highest AS value (2275.42 ± 110.14), and the minimum 3-dimensionality recorded in Freeclimbing reef (1.46 ± 0.04), where the lowest AS value was also obtained (697.75 ± 295.82) (Table 1).

**Culcita schmideliana** and juvenile coral abundance

In the whole study area, an average density of 17.88 ± 3.74 coral juveniles m\(^{-2}\) was found (Table 1). Notably, the site with maximum and minimum juvenile density were again Dharamboodhoo reef (26.14 ± 8.80 juveniles m\(^{-2}\)) and Freeclimbing reef (3.07 ± 1.63 juveniles m\(^{-2}\)), respectively (Table 1).

A total number of 201 seastars were found in the study area, with an average density of 186.1 ± 22.34 seastars ha\(^{-1}\) (Table 1). Specifically, the site with the highest seastar density was Sunny reef (283.32 ± 75.15 seastars ha\(^{-1}\)), while the lowest density was recorded in Freeclimbing reef (133.33 ± 28.87 seastars ha\(^{-1}\)). No significant differences were found in the number of *C. schmideliana* at different depths or in outer and inner sites and the combination of both factors (Two-way ANOVA, P > 0.05).

Size of seastars in different habitats

The size of seastars was significantly different in relation to AS values and LRIs of the investigated sites (Fig. 3). In particular, *C. schmideliana* diameter was significantly larger in sites showing LRI value lower than 2 (Kruskal–Wallis test P = 0.033) (Fig. 3a), and medium AS values (One-way ANOVA, F = 12.95 P < 0.0001) (Fig. 3b).
Three main factors and their influence on *Culcita schmideliana* choice of habitat were tested through the selectivity coefficient ($E_i$) (Fig. 4). Seastars exhibited a general preference for habitat with medium AS values ($E_i = 0.268$), a marked avoidance for habitat with low AS values ($E_i = -0.353$), and a slight avoidance of areas with high AS values ($E_i = -0.118$) (Fig. 4a).

Considering seastar preferences for habitats with different degrees of seafloor morphological complexity, the preferred habitat of *C. schmideliana* was characterized by a LRI value included between 2 and 2.5, with a $E_i = 0.180$ (Fig. 4a).

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### Culcita schmideliana habitat preferences

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Considering seastar preferences for habitats with different degrees of seafloor morphological complexity, the preferred habitat of *C. schmideliana* was characterized by a LRI value included between 2 and 2.5, with a $E_i = 0.180$ (Fig. 4a). Areas with a LRI larger than 2.5, thus meaning those with the highest complexity, were avoided by seastars and gave an $E_i$ value of -0.213 (Fig. 4b). Moreover, areas showing the lowest three-dimensionality, considered as those with a LRI lower than 2, gave an $E_i$ close to 0 ($E_i = -0.025$), indicating a level of selectivity coefficient close to random (Fig. 4b).

Concerning juvenile corals density, the preferred habitats of *C. schmideliana* were those characterized by a density of juveniles $m^{-2}$ included between 10 and 20 ($E_i = 0.154$), while areas with the lowest juvenile density were avoided ($E_i = -0.269$) (Fig. 4c). Areas with more than 20 juveniles $m^{-2}$, and areas with a juvenile density between 5 and 10 $m^{-2}$, showed a slight preference and avoidance, respectively, even if close to random, with values of $E_i = 0.079$ for the first, and $E_i = -0.059$ for the latter (Fig. 4c).

The multiple linear regression analysis allowed us to define which of the factors considered exerted the highest influence on the cushion seastar abundance in the study area. A significant regression equation was found for AS values ($F_{(3,28)} = 2.8, R^2 = 0.431, P = 0.007$) (Fig. 5a), and for linear rugosity ($F_{(3,28)} = 2.8, R^2 = 0.481, P = 0.024$).

### Table 1  Average values ($\pm$ SE) of seastar densities, coral juvenile densities, AS values, and LRIs found in each sampling site

| Site          | Coral Cover % | Seastars/ha $\pm$ SE | Coral recruits/m2 $\pm$ SE | LRI $\pm$ SE | Slope Value $\pm$ SE |
|---------------|---------------|-----------------------|----------------------------|--------------|-----------------------|
| Dharamboodhoo | 53.45         | 172.22 ± 30.93        | 26.14 ± 8.80               | 2.8 ± 0.33   | 2275.42 ± 110.4       |
| Dhigu Reef    | 27.37         | 211.11 ± 33.79        | 15.66 ± 5.8                | 2.05 ± 0.3   | 1419.99 ± 163.95      |
| Freeclimbing  | 19.8          | 133.33 ± 28.87        | 3.7 ± 1.63                 | 1.46 ± 0.04  | 697.5 ± 295.82        |
| Maaga         | 10.01         | 144.5 ± 30.93         | 12.61 ± 7.25               | 1.87 ± 0.20  | 1632.5 ± 259.3        |
| Route 66     | 24.68         | 172.18 ± 5.56         | 25.73 ± 11.98              | 2.11 ± 0.14  | 1451.35 ± 123.60      |
| Sunny Reef    | 35.38         | 283.32 ± 75.15        | 24.08 ± 6.77               | 2.16 ± 0.17  | 2266.88 ± 319.28      |
| Total         | 28.46 ± 6.6   | 186.1 ± 22.34         | 17.88 ± 3.74               | 2.2 ± 0.10   | 1623.96 ± 242.70      |

**Fig. 3** Average diameter ($\pm$ SE) of seastars found in areas a with different values of LRI and b with different AS values. Kruskal–Wallis test followed by multiple pairwise comparisons was used to test significant differences in seastars diameter in areas with different LRI ($P < 0.033$). One-way ANOVA followed by Tukey's post-hoc test was used to test significant differences in diameters of seastars found in areas with different AS values ($P < 0.0001$). Significant differences are marked with * in proximity of each relative column.
Fig. 4  Values of Van der Ploeg and Scavia Selectivity coefficient (-1 = complete avoidance; 0 = random choice; +1 = exclusive preference) of seastars for areas with a different AS values b different LRIs and c different densities of coral juveniles.

(Fig. 5b), showing that the 43.1% and the 48.1% of the variance in seastar abundance was explained by differences in AS values and LRI, respectively. The regression equation to assess the variation in seastar abundance in relation to the variation of juvenile coral densities resulted to be not significant (F(3,28) = 2.8, R² = 0.132, P = 0.202) (Fig. 5c).

Fig. 5 Multiple linear regression analysis displayed as partial regression plots. Relationships between seastar density and a AS values (P = 0.007) b LRI (P = 0.024) and c juvenile corals m⁻² (P = 0.202). All data relative to the three independent variables have been log-transformed in the partial regression plots. Continuous lines indicate partial regression of each independent variable.

Discussion

Among the wide range of disturbances that have affected the integrity of Maldivian reefs in the last decades, including mass bleaching events (Tkachenko 2015; Perry and Morgan 2017), coral diseases (Seveso et al. 2012, 2015; Montano et al. 2015, 2016), and land reclamation (Fallati et al. 2017), corallivorous outbreaks represent one of the greatest threats, and C. schmideliana has certainly played a fundamental role (Saponari et al. 2014; Saponari et al. 2018; Montalbetti et al. 2019a; Caragnano et al. 2021). However, the factors driving the distribution and the habitat choice of this organism still remain unclear. In this study, we analyzed the role of different geomorphological parameters in the habitat selection of C. schmideliana, through the use of novel photogrammetry...
techniques that allowed us to accurately reconstruct the reef topology. In addition, prey availability has also been investigated as a possible influencing factor.

The abundance of seastars in the investigated area was not affected by the site exposure, as a comparable number of individuals was observed in inner and outer reefs. Furthermore, seastar distribution resulted to be relatively homogeneous at different depth ranges, similar to a previous study in which seastar density did not change significantly within the first 30 m of depth (Montalbetti et al. 2019a). Although the cushion seastar has been occasionally reported at depths of up to 50–60 m, it seems to prefer shallowest areas with high coral cover, probably due to the larger availability of coral prey (Thomassin 1976; Glynn 1986). Moreover, since larval settlement is influenced by sea bottom substrate, it could also represent an influencing factor of adult seastar distribution in different reef habitats and depths (Ebert 1983). Indeed, it has been shown that settlement of Acanthaster cf. solaris larvae in relatively shallow waters could forewarn seastar population eruptions and tend to increase with the proportion of coral rubble, while decreasing with the proportion of live hard coral (Wilmes et al. 2020). Since very little is known about C. schmideliana larval settlement and its success on different substrates, the possibility that early life stage population of seastars can influence adult distribution needs to be further explored.

As long as the ability to move on complex reefs is a key component of the behaviour of corallivorous echinoderms, our results about seastar size suggested that C. schmideliana mobility could be influenced by body anatomy. Indeed, seastars found in areas with a low morphological complexity were significantly larger compared to those found in more complex areas, suggesting that larger specimens could be prevented from moving within complex reef frames. Culcita schmideliana is a round-shaped un-armed seastar, and these peculiar anatomical features have been hypothesized to represent a strong influencing factor in its distribution and habitat choice (Quinn and Kojis 2003; Montalbetti et al. 2019a). Limited mobility caused by anatomical features could also explain the lower average size of seastars found in areas with elevated AS values. The largest specimens were found in areas with medium AS values, and smaller seastars were found in reefs with low degree of inclination. These results strengthen the hypothesis that body size could represent a key element of C. schmideliana habitat choice. Echinoderms have a strict connection with reef structure due to different ethological aspects, such as diurnal cryptic behaviour that is often reported for coral predator seastars (Dumont et al. 2007; Pratchett et al. 2017). In particular, crown of thorns often showed a radical behavioral change between diurnal and nocturnal conditions, being hidden within reef crevices and cavities during the day and getting out for preying during night (Rivera-Posada et al. 2014; Burn et al. 2020). Diurnal crypticity has been reported also for cushion seastars and other corallivore invertebrates (Wolf et al. 2014; Enochs and Glynn 2017). In this context, we hypothesized that seastar anatomical features and the need of hiding during daylight hours could be correlated with the choice of habitats with different degrees of steepness and 3-dimensionality, although this aspect should be further investigated.

Morphological complexity has been reported as an ecological factor correlated with fish assemblage biodiversity and biomass in different tropical regions, such as the Caribbean Sea and Indian Ocean (Harborne et al. 2012; Darling et al. 2017). Furthermore, reef 3-dimensionality has been addressed as a strong predictor of abundance, biomass, diversity, and trophic structure of diurnal, non-cryptic reef fishes (Darling et al. 2017). However, how these morphological factors can influence the distribution of mobile invertebrates is a far less examined question (Stella et al. 2011; Graham and Nash 2013). Our results suggest that seastars tend to avoid habitats with high values of linear rugosity, highlighting once again that the anatomical shape of seastars could play a crucial role in habitat selection. The same pattern of distribution has previously been observed for other reef invertebrates such as sea urchins, which seem to aggregate more frequently and more abundantly on reefs with moderate rugosity due to their scarce ability to climb on complex coralline structures (Weil et al. 2005; Dumont et al. 2007). On the contrary, for smaller benthic organisms, such as penaeid shrimps, an increase in substrate rugosity and complexity coincided with an increased differentiation of their spatial distribution due to a larger number of micro-habitats suitable for prey, thus allowing the coexistence of several prawn species, each exploiting different microhabitats and patches (Gribble et al. 2004, 2007). In the case of C. schmideliana, the preferential choice of habitats with a medium value of rugosity could result from the sum of different factors, such as body anatomy, behavioral traits, and larval settlement. In addition, selectivity coefficient values indicated a well-defined pattern of avoidance by seastars of reef areas with low and high AS values, while habitats displaying medium AS values were preferred. The role of reef inclination in determining the distribution patterns of benthic organisms has been largely ignored so far, but our data suggest that C. schmideliana could selectively prefer areas with medium AS values. We hypothesize that this selection may be related to the settlement of juvenile seastars. Juveniles recruitment patterns have been shown to influence adult distribution in several echinoderm species, and are controlled by different factors, such as availability of food,
sedimentation rates, and reef complexity (Glynn and Krupp 1986; Birkeland and Lucas 1990; Klumpp and McKinnon 1992; Quinn and Kojis 2003).

Finally, prey availability has widely been reported to influence corallivorous seastar distribution and movement on coral reefs (Burn et al. 2020; Ling et al. 2020). In the Maldives, C. schmideliana was observed preying preferentially on small coral colonies (diameter < 10 cm) of Pocillopora spp. and Pavana spp., although it was able to consume a large number of different coral genera (Montalbetti et al. 2019). Here, seastar preference for reef areas with a number of favourite preys larger than 10 m⁻² was observed, confirming previous findings. Coral juvenile quantification through photoquadrats analysis resulted to be an effective and quick method to investigate large portions of reef. Possible errors related to this methodology, in particular considering a qualitative identification of juvenile corals, should be taken into account in future studies, as described in Burgess et al. (2010). In this work, a coral identification at the genus level was not performed, but in future research a possible correlation between seastar densities and abundance of different coral genera should be explored.

However, taken together, our results indicate that both the morphological factors considered, namely LRIs and AS values, resulted to be more effective in explaining variance in C. schmideliana densities, compared to prey availability. Therefore, the substrate complexity and inclination mostly affected the habitat choice of C. schmideliana, rather than food abundance. It is important to emphasize that this kind of hierarchical approach has been rarely applied to seastar habitat choice studies. Previously, it was observed that juveniles of two temperate water seastar species, Pisaster ochraceus and Evasterias troschelii, chose their habitat primarily depending on bottom complexity in order to escape predators, and once they attained a size unsuitable for predators, they select their habitat mainly on the base of prey availability (Rogers and Elliot 2013). However, the information about prey predation on C. schmideliana are still scarce, whereas it is known that Acanthaster spp. are preyed by different organisms at different life stages (Cowan et al. 2017; Balu et al. 2021). It is reasonable to hypothesize a similar predatory pressure on Culcita spp., at least at the larval stage, considering that these two corallivorous seastars share many ecological similarities (Glynn and Krupp 1986). Therefore, we cannot exclude that predation could represent an influencing factor of habitat choice by C. schmideliana, and further investigations are needed to clarify this point.

In conclusion, this study provided new insights about the distribution and the habitat choice of an important corallivore of the Maldivian archipelago, partially confirming previous untested hypotheses, and adding new information that could be useful for a better understanding of the behaviour of this understudied organism. In particular, we found that body anatomy could represent a driver of habitat choice, which, combined with other behavioral traits of the seastar, could influence the distribution of these organisms. Furthermore, despite preferring areas with larger availability of food, C. schmideliana selects its habitat primarily on the base of the morphological features of the reef, like 3-dimensional complexity and steepness. Our results provide new important information about the ecology of corallivore organisms that should be considered in developing programmes for the monitoring and prevention of the impacts of this plague in Maldivian coral reefs.

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Declarations

Conflict of 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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