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Using high-throughput sequencing of ITS2 to describe Symbiodinium metacommunities in St. John, U.S. Virgin Islands

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Symbiotic microalgae (Symbiodinium spp.) strongly influence the performance and stresstolerance of their coral hosts, making the analysis of Symbiodinium communities in corals (and metacommunities on reefs) advantageous for many aspects of coral reef research. High-throughput sequencing of ITS2 nrDNA offers unprecedented scale in describing these communities, yet high intragenomic variability at this locus complicates the resolution of biologically meaningful diversity. Here, we demonstrate that generating operational taxonomic units by clustering ITS2 sequences at 97% similarity within, but not across, samples collapses sequence diversity that is more likely to be intragenomic, while preserving diversity that is more likely interspecific. We utilize this 'within-sample clustering' to analyze Symbiodinium from ten host taxa on shallow reefs on the north and south shores of St. John, US Virgin Islands. While Symbiodinium communities did not differ between shores, metacommunity network analysis of host-symbiont associations revealed Symbiodinium lineages occupying 'dominant' and 'background' niches, and coral hosts that are more 'flexible' or 'specific' in their associations with Symbiodinium. These methods shed new light on important questions in coral symbiosis ecology, and demonstrate how application-specific bioinformatic pipelines can improve the analysis of metabarcoding data in microbial metacommunity studies.

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ABSTRACT

- 12 Symbiotic microalgae (Symbiodinium spp.) strongly influence the performance and stress-
- 13 tolerance of their coral hosts, making the analysis of *Symbiodinium* communities in corals
- 14 (and metacommunities on reefs) advantageous for many aspects of coral reef research.
- 15 High-throughput sequencing of ITS2 nrDNA offers unprecedented scale in describing these
- 16 communities, yet high intragenomic variability at this locus complicates the resolution of
- 17 biologically meaningful diversity. Here, we demonstrate that generating operational
- 18 taxonomic units by clustering ITS2 sequences at 97% similarity within, but not across,
- 19 samples collapses sequence diversity that is more likely to be intragenomic, while
- 20 preserving diversity that is more likely interspecific. We utilize this 'within-sample
- 21 clustering' to analyze *Symbiodinium* from ten host taxa on shallow reefs on the north and
- south shores of St. John, US Virgin Islands. While *Symbiodinium* communities did not differ
- 23 between shores, metacommunity network analysis of host-symbiont associations revealed



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Symbiodinium lineages occupying 'dominant' and 'background' niches, and coral hosts that are more 'flexible' or 'specific' in their associations with Symbiodinium. These methods shed new light on important questions in coral symbiosis ecology, and demonstrate how application-specific bioinformatic pipelines can improve the analysis of metabarcoding data in microbial metacommunity studies.

INTRODUCTION

The composition of symbiotic algal communities (Symbiodinium spp.) in reef-building scleractinian and milleporine corals plays a major role in their biology and ecology, as identity and the functional performance of symbionts influences emergent properties of the holobiont, including its photobiology, energetics, growth rates, and susceptibility to stress (Rowan, 2004; Cantin et al., 2009; Jones and Berkelmans, 2011). Even slight differences in the relative abundance of different symbionts may have important functional consequences for the host (Loram et al., 2007; Cunning et al., 2015a). Moreover, variation in these communities among individuals within a host species, and within individuals over time, is an important trait modulating sensitivity of corals to environmental stress (Putnam et al., 2012) and their ability to respond in beneficial ways to changing environmental conditions by 'shuffling' or 'switching' symbionts (Baker, 2003). Therefore, an accurate and comprehensive description of Symbiodinium communities informs understanding of the ecological performance of corals. Symbiodinium identity is primarily described using genetic sequences of chloroplast (cp23S and psbA^{ncr}) and nuclear markers (18S, 28S, ITS1, and ITS2) (LaJeunesse, 2001; Santos et



45 al., 2003; Pochon et al., 2012). Together, these markers have been used to identify nine 46 major 'clades' within the genus Symbiodinium (clades A through I (Pochon and Gates, 47 2010)), which have been further divided into 'types' based on the marker used to identify 48 them. Symbiodinium species are beginning to be described based on combinations of 49 markers, including microsatellites to establish reproductive isolation (i.e., to satisfy the 50 biological species concept) (LaJeunesse et al., 2014). However, ecological surveys of 51 Symbiodinium diversity still generally rely on commonly-used marker genes, such as ITS2. 52 Consequently, high-throughput sequencing of ITS2 is being utilized to characterize 53 Symbiodinium communities with unprecedented resolution (e.g., Arif et al., 2014; Quigley et 54 al., 2014; Thomas et al., 2014; Edmunds et al., 2014; Cunning et al., 2015b; Ziegler et al., 55 2017). However, such analyses often create datasets consisting of millions of sequence 56 reads and hundreds of thousands of distinct sequence variants (Ziegler et al., 2017), which 57 places great importance on the ways in which ITS2 sequence diversity is analyzed and 58 interpreted in relation to biological diversity. 59 While ITS2 initially was investigated as a potential species-level marker in *Symbiodinium* 60 (LaJeunesse, 2001), it is now understood that this marker is not sufficiently variable to 61 distinguish all species within the genus (Finney et al., 2010; Thornhill et al., 2013). For 62 example, the 'B1' ITS2 sequence variant is shared by S. minutum, S. pseudominutum, and S. 63 antillogorgium, and potentially other species (Parkinson et al., 2015). Moreover, the 64 position of ITS2 within the tandemly-repeating ribosomal DNA array creates multiple ITS2 65 sequence variants within a single genome (Thornhill et al., 2007) that evolve through 66 concerted evolution (e.g. (Dover, 1986)). In fact, concerted evolution may mask species 67 divergence within Symbiodinium by maintaining ancestral sequence variants as numerical



68 dominants in multiple derived lineages (Thornhill et al., 2013). Together, these features of 69 ITS2 complicate the interpretation of intragenomic versus interspecific variation, and 70 preclude its use as a true species-level marker for *Symbiodinium*. Nevertheless, numerically 71 dominant intragenomic variants of ITS2 are still phylogenetically informative across the 72 genus (LaJeunesse, 2001), and resolve diversity at a functionally and ecologically important 73 level. Moreover, due to the large quantity of existing sequence data for comparative 74 analysis (e.g., (Franklin et al., 2012; Tonk et al., 2013)), and the relative ease of amplifying 75 and sequencing ITS2, it remains an essential and powerful marker for *Symbiodinium*. 76 Therefore, it is important to develop best practices in the bioinformatic analysis and 77 interpretation of ITS2 metabarcoding surveys of *Symbiodinium*. 78 Here, we describe a Symbiodinium metacommunity associated with scleractinians and a 79 *Millepora* hydrocoral in St. John, U.S. Virgin Islands, with the objectives of: 1) developing an 80 appropriate bioinformatic approach for *Symbiodinium* ITS2 metabarcoding surveys, and 2) 81 exploring network analysis and statistical applications for such datasets that can advance 82 understanding of *Symbiodinium* metacommunity ecology. In addressing the first objective, 83 we test the ability of 'within-sample clustering' (i.e. clustering sequences at 97% similarity 84 within each sample independently) to generate biologically relevant operational taxonomic 85 units (OTUs) from ITS2 metabarcoding data. Specifically, this approach addresses the fact 86 that dominant ITS2 sequences from different *Symbiodinium* species may be more similar to 87 one another than intragenomic variants within one Symbiodinium (Thornhill et al., 2013; 88 Arif et al., 2014; Parkinson et al., 2015). Therefore, clustering all sequences together 89 underestimates diversity by collapsing different species into the same OTU. Conversely, 90 treating each ITS2 sequence as a unique Symbiodinium type overestimates diversity due to



91 intragenomic variation. Within-sample clustering may better approximate true diversity by 92 exploiting key assumptions regarding the distribution of symbionts among samples, and 93 sequences among symbionts. These assumptions include: 1) that most coral colonies are 94 dominated by a single *Symbiodinium* type (LaJeunesse and Thornhill, 2011; Pettay *et al.*, 95 2011; Baums et al., 2014), and 2) that different numerically dominant ITS2 sequence 96 variants diagnose different Symbiodinium types, even when they differ by only 1 nucleotide 97 (Sampayo et al., 2009). These assumptions suggest that when closely related ITS2 98 sequences (i.e., that are >97% similar; Arif et al. (2014)) occur within the same sample, 99 they are more likely to be intragenomic variants, but when they are numerically dominant 100 in different samples, they are more likely to represent distinct Symbiodinium taxa. 101 Accordingly, clustering sequences at 97% similarity within each sample independently may 102 collapse variability that is more likely to be intragenomic (i.e., occurring within a sample), 103 while maintaining variability that is more likely to be interspecific (i.e., occurring in 104 different samples). 105 Using within-sample clustering, we analyzed Symbiodinium communities in ten host 106 species across the north and south shores of St. John. We explore ecological patterns that 107 can be revealed by large-scale metabarcoding datasets, including 1) testing for whole-108 community differences associated with north and south shore locations, 2) analyzing coral-109 Symbiodinium metacommunity association networks, and 3) quantifying the variability in 110 symbiont communities (i.e., beta diversity, or 'symbiosis flexibility') among individuals 111 within a host species. The large quantity of data, and the kinds of analytical approaches 112 facilitated by ITS2 metabarcoding, have the potential to revolutionize understanding of Symbiodinium metacommunity ecology in reef corals and other taxa harboring similar 113



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symbionts. More generally, the present study demonstrates how ecological knowledge can inform bioinformatic analyses using marker genes for which sequence diversity does not map directly to species.

MATERIALS AND METHODS

Sample collection and environmental conditions

Eighty-four tissue samples were collected from scleractinian and milleporine corals at seven sites around St. John, US Virgin Islands, USA (Fig. 1), between August 7th and 9th, 2012, as permitted by the Virgin Islands National Park (permit VIIS-2012-SCI-0017). Samples were collected on snorkel from 3-7m depth, and colonies to be sampled were selected haphazardly in upward-facing locations exposed to ambient irradiance. Tissue biopsies (~5-mm diameter and ~5-mm depth) were removed from coral colonies with metal borers, and the holes were plugged with non-toxic modeling clay. Broad-scale differences in the hydrographic regimes at points representative of the north and south shores of St. John (North: 18.368141°N, 64.711967°W; South: 18.315465°N, 64.726840°W) were evaluated from remote sensing tools to determine sea surface temperature (°C), surface chlorophyll concentrations (mg chl a m⁻³), and integrated wave exposure on a power scale (I m⁻³). These features were selected as they corresponded to anecdotal observations of differences between the water masses on both shores, and conditions known to affect coral performance, Symbiodinium biology (i.e., SST (Coles and Jokiel, 1977) and wave regime (Atkinson *et al.*, 1994)), and the supply of food resources that play important roles in coral nutrition (Houlbrèque and Ferrier-Pagès, 2009). SST and



135 chlorophyll were determined using data from the MODIS sensor on the Aqua satellite. SST 136 was evaluated from nighttime records, and chlorophyll from ocean colour, both of which 137 were obtained at 1 km spatial resolution for each month, averaged over the years 2003-138 2010, from the IMaRS website (http://imars.usf.edu/modis); if data were not available at 139 the chosen coordinates (above), values from the nearest available pixel were used. 140 Boxplots were generated from the 12 monthly climatological values obtained for each 141 response (Fig. 1). 142 Wind-driven wave exposure on a power scale for a given site is dependent on the wind 143 patterns and the configuration of the coastline, which defines the fetch. To calculate wave 144 exposure, wind speed and direction at each location were acquired from the QuickSCAT 145 (NASA) satellite scatterometer from 1999 to 2008 at 25 km spatial resolution (Hoffman 146 and Leidner, 2005). Coastline data were obtained from the Global Self-consistent, Hierarchical, High-resolution, Shoreline (GSHHS v 2.2) database, which provides global 147 coastline at 1:250,000 scale (Wessel and Smith, 1996). From these data, wave exposure 148 149 was calculated using the methods based on wave theory (after Chollett et al. 2012) for 32 150 fetch directions (equally distributed through 360°). Total wave exposure (summed over all 151 directions) was calculated in R using the packages maptools (Bivand and Lewin-Koh, 152 2016), raster (Hijmans, 2016), rgeos (Bivand and Rundel, 2016), and sp (Bivand et al., 153 2013). 154 ITS2 sequencing and bioinformatics 155 Coral tissue samples were preserved in ~500 µL Guanidinium buffer (50% w/v 156 guanidinium isothiocyanate; 50 mM Tris pH 7.6; 10 µM EDTA; 4.2% w/v sarkosyl; 2.1%



157 v/v β-mercaptoethanol) and shipped to the Hawaii Institute of Marine Biology (HIMB). 158 Genomic DNA was extracted from each coral tissue sample following a guanidinium-based 159 extraction protocol (Cunning et al., 2015b), and sent to Research and Testing Laboratory 160 (Lubbock, TX) for sequencing of ITS2 amplicons ('itsD' and 'its2rev2' primers from Stat et 161 al. (2009)) on the Illumina MiSeq platform with 2x300 paired-end read chemistry. 162 Paired reads from each sample (provided in .fastq format by Research and Testing 163 Laboratory) were merged using illumina-utils software (Eren et al., 2013b) with an 164 enforced Q30-check and an overlap ≥ 150 bp with ≤ 3 mismatches required to retain a 165 sequence. Chimeric sequences were removed using usearch61 (Edgar, 2010) implemented 166 in QIIME (Caporaso et al., 2010). Primers were trimmed using cutadapt (Martin, 2011) 167 allowing 3 mismatches, and only sequences with both forward and reverse primer matches 168 and length ≥ 250 bp after trimming were retained. Subsequently, three different clustering 169 approaches, each based on the uclust algorithm (Edgar, 2010) and implemented in QIIME, 170 were used to group ITS2 sequences into operational taxonomic units (OTUs): 1) clustering 171 at 100% identity, 2) clustering at 97% identity across samples (i.e., sequences from all 172 samples clustered together), and 3) clustering at 97% identity within samples (i.e., 173 sequences from each sample clustered independently). For each 97% cluster, the most 174 abundant sequence variant was chosen as the representative sequence, and within-sample 175 clusters were merged across samples if their representative sequences were 100% 176 identical. After removing singleton clusters, representative sequences for each OTU were 177 assigned taxonomy by searching a custom reference database of Symbiodinium ITS2 178 sequences using the Needleman-Wunsch global alignment algorithm implemented in the 179 Biostrings package (Pagès et al., 2016) in R (R Core Team, 2014). Each OTU was then



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assigned a name corresponding to the reference sequence(s) with the highest alignment score; if the match was <100%, the assignment was given a unique superscript. If the match was <90%, the sequence was blasted to the NCBI nt database and omitted if the top hit did not contain "Symbiodinium". The reference database comprised *Symbiodinium* ITS2 sequences downloaded directly from NCBI. These sequences included those used in previous reference databases (Cunning et al., 2015b) supplemented with additional sequences of Symbiodinium from Caribbean corals (Finney et al., 2010; Green et al., 2014). Reference sequences were separated and aligned by clade using muscle (Edgar, 2004), and trimmed to equal length using the osmart-trim command from oligotyping software (Eren et al., 2013a), before being reconcatenated into a single fasta file. The bioinformatic pipeline used here was implemented using a series of bash scripts which can be found in the data archive accompanying this paper (http://github.com/jrcunning/STJ2012). These scripts are provided along with all raw data and a Makefile which can be executed to fully reproduce the present analysis.

Symbiodinium data analysis

OTU tables, sample metadata, and taxonomic data were imported into R using the phyloseq package (McMurdie and Holmes, 2013) to facilitate downstream analyses. OTU tables were filtered to remove any OTU that was not observed at least 10 times, or any sample with < 200 sequences, and counts were transformed to relative abundance. Permutational analysis of variance (PERMANOVA) was used to test for differences in *Symbiodinium*



202 species. 203 Network analysis and visualization was performed in R using the igraph package (Csárdi 204 and Nepusz, 2006). Networks were created featuring 'dominant' (>50% relative abundance) and 'abundant' (>1% relative abundance) OTUs, with weighted edges 205 206 proportional to the number of individuals within a species in which a symbiont OTU 207 occurred. A 'background' symbionts network was also created with unweighted edges 208 defined based on the presence of a symbiont at <1% relative abundance in at least one 209 individual within the species. To simplify visualization of the background symbiont 210 network, clade D OTUs were merged into a single node under the assumption that clade D in the Caribbean comprises a single species (Symbiodinium trenchii; (LaJeunesse et al., 211 212 2014)), and symbiont nodes connected to ≤ 2 coral species were removed from the 213 network. All network layouts were constructed based on the Fruchterman-Reingold algorithm (Fruchterman and Reingold, 1991). 214 Beta diversity (sensu (Anderson *et al.*, 2006)) was calculated as the multivariate dispersion 215 216 of samples within a coral species. Principal coordinate analysis of Bray-Curtis 217 dissimilarities of square-root transformed OTU counts was used to calculate average 218 distance-to-centroid values for each species, which were then compared statistically by a 219 permutation test. This analysis was implemented using the betadisper function in the 220 vegan package (Oksanen et al., 2016), based on (Anderson et al., 2006).

community composition between the north and south shores of St. John within each coral



221 **RESULTS**

222	Sample collection and environmental data
223	A total of 84 coral samples representing ten host species were collected at two sites on the
224	north shore and five sites on the south shore of St. John (Fig. 1). These samples generated
225	1,490,813 sequences after merging paired reads, removing chimeric sequences, and
226	trimming primers. After clustering, OTUs with <10 sequences and samples with <200
227	sequences were excluded, leaving 80 samples for downstream analysis.
228	The environmental conditions broadly characterizing the north and south shores are
229	presented in Fig. 1. From 2003 to 2010, the north shore of St. John was characterized by
230	slightly lower sea surface temperatures, higher chlorophyll a concentrations, and lower
231	wave exposure, relative to the south shore.
232	Comparison of clustering approaches
232233	Comparison of clustering approaches The number of OTUs resolved, as well as the number of sequences per OTU and per sample,
233	The number of OTUs resolved, as well as the number of sequences per OTU and per sample,
233234	The number of OTUs resolved, as well as the number of sequences per OTU and per sample, depended on the clustering approach (Table 1 and Fig. S1). More OTUs were resolved as
233234235	The number of OTUs resolved, as well as the number of sequences per OTU and per sample, depended on the clustering approach (Table 1 and Fig. S1). More OTUs were resolved as the clustering resolution increased (i.e., 97% OTUs across samples $< 97\%$ OTUs within
233234235236	The number of OTUs resolved, as well as the number of sequences per OTU and per sample, depended on the clustering approach (Table 1 and Fig. S1). More OTUs were resolved as the clustering resolution increased (i.e., 97% OTUs across samples < 97% OTUs within samples < 100% OTUs), with fewer reads per OTU (Table 1). The number of sequences per
233234235236237	The number of OTUs resolved, as well as the number of sequences per OTU and per sample, depended on the clustering approach (Table 1 and Fig. S1). More OTUs were resolved as the clustering resolution increased (i.e., 97% OTUs across samples $< 97\%$ OTUs within samples $< 100\%$ OTUs), with fewer reads per OTU (Table 1). The number of sequences per sample was less for the 100% OTU approach because more OTUs were filtered out of the
233234235236237238	The number of OTUs resolved, as well as the number of sequences per OTU and per sample, depended on the clustering approach (Table 1 and Fig. S1). More OTUs were resolved as the clustering resolution increased (i.e., 97% OTUs across samples $< 97\%$ OTUs within samples $< 100\%$ OTUs), with fewer reads per OTU (Table 1). The number of sequences per sample was less for the 100% OTU approach because more OTUs were filtered out of the dataset by the minimum threshold count of 10 reads per OTU.



242 assigned the same dominant OTU to samples dominated by different sequence variants, 243 while 97% within-sample clustering assigned distinct dominant OTUs to these samples. 244 Because the outcomes of the latter approach are more consistent with the current 245 understanding of ITS2 sequence diversity as it links to *Symbiodinium* biology and ecology 246 (see Discussion), the remainder of the results is presented using the 97% within-sample 247 clustering approach. Additional detailed comparisons of the outcomes of each clustering 248 approach for each coral species are available at http://github.com/jrcunning/STJ2012. 249 Symbiodinium community composition 250 Within the set of coral samples, clade B had the highest relative abundance (46.6%), followed by clade C (41.3%), A (10.5%), D (1.5%), and G (0.1%). The number of OTUs 251 within each clade followed a similar pattern, with the highest number in clade B (66), 252 253 followed by clade C (25), A (6), D (5), and G (4). The distribution of these clades within each 254 sample is shown in Fig. 3. 255 Differences in Symbiodinium between shores 256 No significant differences in *Symbiodinium* community composition between the north and 257 south shores were detected for any host species (Table 2). However, qualitative differences 258 were apparent: *Porites furcata* was dominated by either clade A or clade C *Symbiodinium* on 259 the south shore, but only by clade C on the north shore (difference between shores 260 PERMANOVA: p=0.056). Siderastrea siderea was dominated by either clade C or clade D on 261 the south shore, but only by clade C on the north shore (difference between shores 262 PERMANOVA: p=0.071).



Network analysis of Symbiodinium metacommunity

264	Patterns of association between hosts and Symbiodinium were analyzed using networks
265	focusing on "abundant" (>1% relative abundance; Fig. 4), "dominant" (>50% relative
266	abundance; Fig. 5A), and "background" (<1% relative abundance; Fig. 5B) symbionts in
267	each host species, based on 97% within-sample clustering. (Networks for individuals
268	within each species are available at http://github.com/jrcunning/STJ2012). In the network
269	for abundant symbionts (Fig. 4), 37 OTUs were observed. P. strigosa associated with the
270	greatest number of OTUs (n=17), followed by $F. fragum$ (n=5), $M. alcicornis$ (n=5), and $D.$
271	cylindrus (n=4); these primarily comprised OTUs closely related to B1 and B19. In contrast,
272	only a single OTU occurred at $>1\%$ relative abundance in both <i>P. astreoides</i> (A4) and <i>M.</i>
273	cavernosa (C3). All coral species hosted at least one "abundant" Symbiodinium OTU that
274	was also abundant in at least one other coral species, except for <i>S. radians</i> . This species was
275	dominated by C46, but also contained Symbiodinium closely related to B1 and B19, as well
276	as B5.
277	In the network for dominant symbionts, 15 different Symbiodinium OTUs were included,
278	defined as those occurring at >50% relative abundance within a host (Fig. 5A). A single
279	OTU identical to <i>Symbiodinium</i> B1 was the most prevalent dominant symbiont, dominating
280	all D. cylindrus samples and many P. strigosa (4 of 9), O. annularis (2 of 4), M. alcicornis (6 of
281	10), and <i>F. fragum</i> (8 of 9) samples. Other closely-related clade B OTUs (similar to B1, B1d,
282	and B19) occasionally dominated <i>P. strigosa, M. alcicornis,</i> and <i>F. fragum,</i> while <i>O. annularis</i>
283	was dominated just as often by Symbiodinium C7. After B1, the next most prevalent
284	dominant symbionts were <i>Symbiodinium</i> C3, which dominated all <i>M. cavernosa</i> and most <i>S.</i>



285 siderea (7 of 10), and Symbiodinium A4, which dominated all P. astreoides and 3 of 10 P. 286 furcata. P. furcata and S. siderea were each occasionally dominated by other symbiont 287 OTUs, including C45 or C45a (*P. furcata*), and C1 or D1 (*S. siderea*). All coral species were 288 dominated by at least one Symbiodinium OTU that also dominated at least one other coral 289 species, except for *S. radians*, which was exclusively dominated by C46. 290 There were 6 symbionts that occupied a 'background' niche, defined as <1% relative 291 abundance in three or more coral species (Fig. 5B). A member of Symbiodinium clade D 292 (which has been described as S. trenchii (LaJeunesse et al., 2014)) was found at background 293 levels in the greatest number of host species (7), followed by A4 and C3 (both in 6 taxa). 294 C31, C46, and B1 were each detected at background levels in 4 host species. 295 Symbiotic flexibility of host species 296 Symbiont communities from each individual host were ordinated in multivariate space, and 297 the dispersion of individuals within each host species (i.e., average distance to centroid) 298 was calculated as a metric of symbiosis 'flexibility', with higher values indicating hosts with 299 greater variability in symbiont community composition among individuals, and lower 300 values indicating hosts with greater uniformity. This metric revealed significant differences 301 in symbiotic flexibility among host species (Fig. 6). The highest distance to centroid was 302 found in *P. strigosa*, *P. furcata*, *M. alcicornis*, *S. siderea*, and *M. annularis*, and the lowest was 303 found in M. cavernosa. Flexibility was also low in D. cylindrus, S. radians, and P. astreoides, 304 and intermediate in F. fragum.



DISCUSSION

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Bioinformatic analysis of ITS2 diversity

The structure of *Symbiodinium* communities that was detected in host species was impacted by the bioinformatic approach used to analyze ITS2 sequence data. Whether sequences were clustered into operation taxonomic units (OTUs) collectively across the entire dataset, or independently within each sample, affected the number and taxonomic assignment of OTUs. Clustering sequences across samples reduced the total number of OTUs by 98% relative to the number of OTUs detected with no clustering (Table 1). Withinsample clustering resolved intermediate diversity, with 13% more OTUs than acrosssample clustering. Comparing taxonomic assignments of the OTUs generated by these approaches demonstrates that across-sample clustering often assigns the same dominant OTU to samples with different dominant sequence variants (Fig. 2), while within-sample clustering assigns them distinct OTUs that reflect the dominant variant in the sample. This occurs because OTU identity is determined by the most abundant member sequence, and when members occur across many samples (i.e., during across-sample clustering), each member's abundance reflects the number (and sequencing depth) of samples in the dataset that contain it. Therefore, a particular sequence variant from a more extensively-sampled and/or deeply-sequenced host may determine the identity of an OTU that includes closelyrelated but different sequences in other samples, even though they do not actually contain the given sequence. Such merging of OTUs across samples is undesirable, since even single nucleotide differences in a sample's dominant ITS2 variant may indicate different Symbiodinium species (Sampayo et al., 2009).



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Therefore, even though clustering at 97% similarity may be needed to collapse intragenomic variation within a single Symbiodinium (Arif et al., 2014), when such clustering is applied across a dataset potentially comprised of many closely related Symbiodinium, it is likely to also collapse interspecific variation and underestimate diversity. This limitation was also encountered by Arif et al. (2014) when clustering closely-related clade C sequences across samples from multiple hosts. Despite being comprised mostly of C41 variants, ~3,000 ITS2 sequences from Acropora hemprichii were subsumed into a single OTU that was named C1 after the dominant variant among the ~7,000 sequences from *Pocillopora verrucosa* with which they were clustered. In this case (and as shown in Fig. 2), clustering sequences across samples leads to the conclusion that samples with different dominant sequence variants are dominated by the same *Symbiodinium* OTU. This outcome occurs because the identity of OTUs in any given sample may be determined by other samples in the dataset; indeed, using this approach, the same sample and sequence assemblage may receive different OTU assignments if it were part of a different set of samples. Meanwhile, clustering sequences within each sample independently assigns OTUs that reflect only the assemblage of sequences within that sample, and therefore does not depend on the presence or abundance of sequences from other samples included in the analysis. The outcome of within-sample clustering of ITS2 sequences as applied herein better reflects patterns of *Symbiodinium* diversity and ecology that have been revealed by more variable markers (psbAncr) and microsatellites (Finney et al., 2010; Thornhill et al., 2013); namely, that clades B and C in the Caribbean comprise numerous different Symbiodinium species that tend to associate with different coral host taxa. Indeed, only



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within-sample clustering assigned different dominant clade C OTUs to samples of O. annularis and S. siderea (Fig. 2), which have been differentiated based on other markers (Thornhill *et al.*, 2013). The within-sample clustering approach should also reflect patterns that would be observed by sequencing dominant ITS2 bands from denaturing gradient gel electrophoresis (DGGE; the most commonly use method of describing *Symbioinium* diversity prior to metabarcoding), since both methods rely on numerically dominant sequence variants to assign taxonomy. Furthermore, metabarcoding overcomes the primary limitations of the DGGE method by providing more quantitative data and sensitive detection of low abundance taxa (Quigley et al., 2014). Therefore, we recommend a withinsample clustering approach for metabarcoding studies where many different Symbiodinium types are likely to be encountered. Importantly, the assumptions underlying a within-sample clustering approach -- that samples typically contain one *Symbiodinium* species that can be diagnosed by its most abundant intragenomic ITS2 variant -- will not always be met. Indeed, multiple Symbiodinium frequently co-occur in single coral colonies (Silverstein et al., 2012), which undermines support for assuming that variation within a sample is intragenomic. However, co-occurring Symbiodinium in many cases are members of different clades, whose ITS2 sequences are divergent enough to be resolved separately by 97% clustering. Thus, only when very closely related types co-occur in a sample (e.g., Sampayo et al., 2007; Wham and LaJeunesse, 2016) is this approach likely to fail. Other problematic cases may occur when multiple intragenomic sequence variants are co-dominant (i.e., comprise nearly equal proportions of the rDNA array), such that slight differences in their relative abundance may lead to different OTU assignments in samples with otherwise highly similar sequence



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assemblages. These challenges may be overcome by basing taxonomic assignment on more complex criteria than just the most abundant sequence variant, such as multiple codominant variants, or more evolutionarily derived variants. While no treatment of ITS2 data can provide species-level descriptions of *Symbiodinium* communities (since the marker itself does not provide this resolution), within-sample clustering nevertheless accommodates known evolutionary complexities in the ITS2 locus to provide community descriptions that better reflect species-level diversity than either across-sample clustering or no clustering at all. When treated with an appropriate bioinformatic approach, ITS2 metabarcoding can provide a comprehensive and quantitative analysis of Symbiodinium diversity, and can be easily applied across host species to rapidly survey the *Symbiodinium* metacommunity on a reef-wide scale. Symbiodinium metacommunity ecology in St. John In the ten host species sampled, the most prevalent *Symbiodinium* were members of clades B and C, which is consistent with previous analyses of Symbiodinium diversity on shallow reefs in this region (LaJeunesse, 2004; Correa et al., 2009; Finney et al., 2010). Less frequent associations between hosts and members of clades A and D were observed, although clade A dominated *P. astreoides* and some *P. furcata*, and clade D dominated one *S.* siderea. Finally, clade G was observed at low relative abundances (<5%) in some coral colonies, which likely would not have been detected without high-throughput sequencing. While the environmental conditions differed between the north and south shores of St. John (e.g., wave exposure, temperature, and chlorophyll a), there were no statistically significant differences in symbiont community composition in the corals sampled on either



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shore. This may partly be due to small sample sizes, since two species showed a trend for differences between shores: *S. siderea* and *P. furcata* were more frequently dominated by 397 *Symbiodinium* in clades D and A (respectively) on the south shore, compared to the north 398 shore (Fig. 2). Symbiodinium in clades D and A are typically associated with warm and 399 variable temperatures (e.g., Baker et al., 2004; Oliver and Palumbi, 2011) and shallow 400 habitats with high light intensities (LaJeunesse, 2002). Therefore, it is consistent with expectations that they were more prevalent on the south shore, where temperatures were 402 slightly higher, and chlorophyll levels were lower, suggesting greater light penetration into 403 the water column. However, the lack of differentiation between shores in the symbiont 404 communities of most host species suggests that environmental differences at this scale are 405 not major drivers of Symbiodinium community structure. Network analysis of associations between *Symbiodinium* types and coral species revealed metacommunity-level patterns in *Symbiodinium* ecology. First, ITS2 type B1 was 408 "abundant" (i.e. >1% relative abundance in a sample) in seven of ten host species (Fig. 4), 409 and "dominant" (i.e. >50%) in five of ten (Fig. 5), suggesting it is a generalist symbiont. 410 However, analysis of other markers such as chloroplast and microsatellite loci has revealed that Symbiodinium ITS2 type B1 in the Caribbean is comprised of multiple lineages that show high host species fidelity (Santos et al., 2004; Finney et al., 2010; Parkinson et al., 413 2015). Likewise, the apparent generalists *Symbiodinium* C3 (which dominated colonies of *S.* 414 siderea and M. cavernosa) and Symbiodinium A4 (which dominated P. astreoides and P. furcata, Fig. 5) may also be comprised of multiple host-specialized lineages (Thornhill et al., 415 416 2013), which could be revealed using higher resolution genetic markers.



417 In addition to hosting apparent generalist ITS2 lineages, most coral species sampled in St. John (except M. cavernosa and P. astreoides) hosted abundant Symbiodinium OTUs that 418 419 were not abundant in any other corals, and are thus apparently more host-specific. Unique 420 OTUs in the B1-radiation were abundant in *M. alcicornis* and *D. cylindrus*, while others from 421 both the B1- and B19-radiations (see LaJeunesse, 2004) were abundant in F. fragum, S. 422 radians, and P. strigosa. Several clade C OTUs similarly were only abundant in one species: 423 C1144 in P. strigosa, C7 in O. annularis, C1 in S. siderea, C1c/C45 and C45a in P. furcata, and 424 C46 in *S. radians*. Interestingly, *S. radians* was the only species in which none of the 425 abundant symbionts occurred in any other host, which may reflect the unique ecology of S. 426 radians as the only study species that typically forms small, encrusting colonies as adults. 427 To reveal symbionts that consistently occupied a 'background' niche, we identified OTUs 428 present at <1% relative abundance in samples of three or more host species. This 429 distribution suggests that these symbionts can occupy a range of hosts, but are unlikely to 430 be dominant symbionts, at least under prevailing environmental conditions. The most prevalent 'background' symbiont in St. John was a member of clade D (Symbiodinium 431 432 trenchii (LaJeunesse et al., 2014)), which proliferates within hosts during and after thermal 433 stress (Thornhill et al., 2006; LaJeunesse et al., 2009; Silverstein et al., 2015), or in marginal 434 environments (LaJeunesse et al., 2010). Other symbionts occupying a background niche in 435 St. John, such as Symbiodinium A4 and C3, may similarly become dominant under different 436 sets of environmental conditions. These background symbionts, therefore, may perform an 437 important functional role within the metacommunity by broadening the fundamental niche 438 that a host may occupy (sensu Bruno et al. (2003)), thereby increasing its resilience to environmental change (Correa and Baker, 2011). 439



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Indeed, the ability of corals to associate with different symbionts may be an important trait that mediates their sensitivity to stress (Putnam et al., 2012), and their ability to change symbionts over time in response to environmental change (Baker, 2003). While flexibility in symbiosis ecology has been quantified previously for individual corals (i.e., based on the diversity of symbionts co-occurring within a single host colony (Putnam et al., 2012)), here we quantify flexibility at the host species level based on the variability of symbiont community structure among multiple colonies (Fig. 6), a metric of beta diversity that can be statistically compared among host species (Anderson et al., 2006). This metric revealed that *M. cavernosa*, *P. astreoides*, *S. radians*, and *D. cylindrus* had the lowest symbiotic flexibility, meaning that all sampled individuals had similar symbiont community structure. In these host species, symbiont communities may be more constrained by host biology and/or less responsive to environmental variation. On the other hand, P. strigosa, P. furcata, M. alcicornis, and S. siderea had high flexibility, meaning that sampled individuals displayed more divergent symbiont communities. Community structure in these hosts may be more responsive to variability in the environment (e.g., Kennedy et al., 2016), or subject to greater stochasticity. While future work should investigate whether the scope for symbiont community change over time within individuals is linked to variability among individuals, we suggest that the latter represents a useful metric of symbiosis flexibility that can be easily quantified using metabarcoding data.

Conclusions

ITS2 metabarcoding and within-sample OTU clustering represents a powerful approach to quantitatively and comprehensively describe *Symbiodinium* metacommunity composition



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on coral reefs. The scale and resolution of datasets generated in this way facilitate new analytical applications that can address critical topics in coral symbiosis ecology, including changes in dominant and background symbionts across environmental gradients and over time, and the role of metacommunity processes in shaping *Symbiodinium* communities. Describing these trends has the potential to greatly advance understanding of coral responses to environmental change. 468 **Acknowledgements** Funding for this work was provided in part by the Long Term Research in Environmental Biology program of the US National Science Foundation (NSF) (DEB 13-50146). RC was supported by a NSF Postdoctoral Research Fellowship in Biology (NSF-PRFB 1400787). Remote sensing observations of sea surface temperature and phytoplankton pigment concentration were processed at the Institute for Marine Remote Sensing at the College of Marine Science, University of South Florida, in St. Petersburg, Florida, and provided courtesy of Frank Müller-Karger and Iliana Chollett. References Anderson MJ, Ellingsen KE, McArdle BH. (2006). Multivariate dispersion as a measure of beta diversity. *Ecology Letters* **9**: 683–693. Arif C, Daniels C, Bayer T, Banguera-Hinestroza E, Barbrook AC, Howe CJ et al. (2014). Assessing Symbiodinium diversity in scleractinian corals via next-generation sequencingbased genotyping of the ITS2 rDNA region. *Molecular Ecology* 23: 4418–4433.



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Figure and Table Legends

- 661 Figure 1. Sampling of Symbiodinium communities from ten host species (embedded
- histogram) on the north and south shores of St. John, US Virgin Islands, and the broad-scale
- 663 physical conditions associated with these two environments. Mapped points along each
- shore indicate locations from which biological samples were collected, while 'N' and 'S'
- indicate locations from which environmental data representative of the north and south
- shores were obtained. Boxplots show the sea-surface temperatures (top), chlorophyll a
- concentrations (middle), and wave exposure (bottom) recorded at these two locations.
- 668 Figure 2. Comparison of dominant OTUs assigned by 100%, 97% across-sample, and 97%
- 669 within-sample clustering approaches. A) The composition of unique sequence variants (i.e.,
- 670 100% OTUs) in each sample, sorted by relative abundance, and filled by unique colors
- 671 corresponding to unique sequence variants. The dominant OTUs within each sample
- assigned by B) 97% across-sample clustering and C) 97% within-sample clustering are
- shown by rectangles below each bar, with fill colors matching the unique sequence variants
- 674 presented in (A) to indicate the representative sequence of the assigned OTU. Colored
- 675 rectangles that span multiple bars indicate that the corresponding samples were assigned
- 676 the same dominant OTU. Rectangles are annotated with the taxonomic assignment of the
- 677 OTU (see Methods).
- 678 Figure 3. Symbiodinium community composition of each sample. Samples are plotted as
- 679 horizontal bars, sorted by species and shore of collection (north vs. south). Individual
- 680 segments of each bar represent 97% within-sample OTUs, colored by Symbiodinium clade
- 681 identity.
- Figure 4. Network analysis of abundant Symbiodinium taxa in each coral species.
- 683 Symbiodinium OTUs (circular nodes) are connected to each coral species (square nodes) in
- which they ever occurred at >1% relative abundance within an individual; thickness of
- 685 edges (i.e., links between coral and symbiont nodes) is relative to the proportion of
- 686 individuals within the coral species in which the Symbiodinium OTU was present at >1%.
- 687 Symbiont nodes are colored by clade identity, and sized proportionally to the number of
- 688 coral species in which they were found.
- 689 Figure 5. Network analysis of dominant (A) and background (B) Symbiodinium taxa in each
- 690 coral species (see Materials and Methods for details on network construction). Symbiont
- 691 nodes are colored by clade identity, sized proportionally to the number of coral species to
- 692 which they are connected.
- 693 Figure 6. Symbiotic flexibility in coral species quantified as multivariate dispersion of
- 694 Symbiodinium community composition (mean distance to centroid) ± SE. Host taxa that
- do not share a letter are significantly different (p < 0.05).
- 696 Table 1. Summary statistics for each clustering approach.
- Table 2. Mean overall, within-shore ('within'), and between-shore ('between') Bray-Curtis
- 698 dissimilarities of the Symbiodinium communities in each host species, and PERMANOVA
- 699 tests (partial R² and p-values) for a difference between shores. Between-shore tests were
- 700 not possible for O. annularis and P. astreoides since they were only sampled from one
- 701 shore.



Figure 1(on next page)

Sampling locations and environmental conditions

Sampling of *Symbiodinium* communities from ten host species (embedded histogram) on the north and south shores of St. John, US Virgin Islands, and the broad-scale physical conditions associated with these two environments. Mapped points along each shore indicate locations from which biological samples were collected, while 'N' and 'S' indicate locations from which environmental data representative of the north and south shores were obtained. Boxplots show the sea-surface temperatures (top), chlorophyll a concentrations (middle), and wave exposure (bottom) recorded at these two locations.

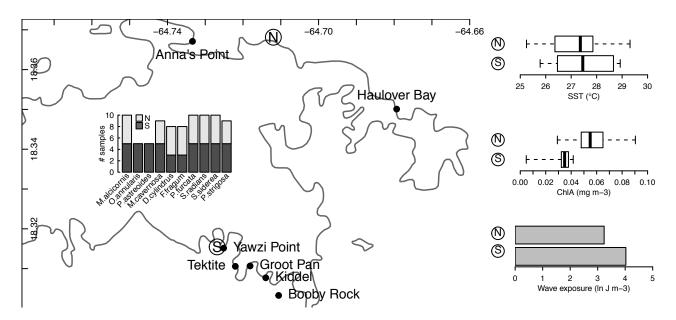


Figure 1: Figure 1. Sampling of *Symbiodinium* communities from ten host species (embedded histogram) on the north and south shores of St. John, US Virgin Islands, and the broad-scale physical conditions associated with these two environments. Mapped points along each shore indicate locations from which biological samples were collected, while 'N' and 'S' indicate locations from which environmental data representative of the north and south shores were obtained. Boxplots show the sea-surface temperatures (top), chlorophyll a concentrations (middle), and wave exposure (bottom) recorded at these two locations.

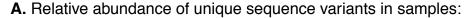


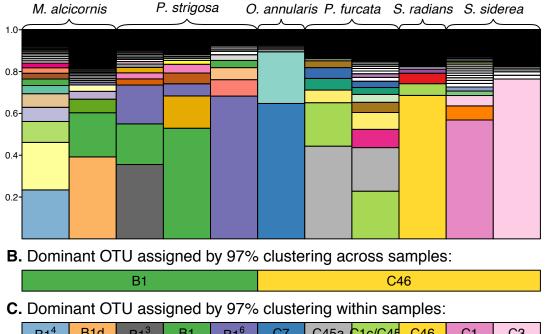
Figure 2(on next page)

OTU assignments by different bioinformatic approaches

Comparison of dominant OTUs assigned by 100%, 97% across-sample, and 97% within-sample clustering approaches. A) The composition of unique sequence variants (i.e., 100% OTUs) in each sample, sorted by relative abundance, and filled by unique colors corresponding to unique sequence variants. The dominant OTUs within each sample assigned by B) 97% across-sample clustering and C) 97% within-sample clustering are shown by rectangles below each bar, with fill colors matching the unique sequence variants presented in (A) to indicate the representative sequence of the assigned OTU. Colored rectangles that span multiple bars indicate that the corresponding samples were assigned the same dominant OTU. Rectangles are annotated with the taxonomic assignment of the OTU (see Methods).







B1 ⁴	B1d	B1 ³	B1	B1 ⁶	C7	C45a C1c/C45	C46	C1	C3
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Figure 2: Figure 2. Comparison of dominant OTUs assigned by 100%, 97% across-sample, and 97% within-sample clustering approaches. A) The composition of unique sequence variants (i.e., 100% OTUs) in each sample, sorted by relative abundance, and filled by unique colors corresponding to unique sequence variants. The dominant OTUs within each sample assigned by B) 97% across-sample clustering and C) 97% within-sample clustering are shown by rectangles below each bar, with fill colors matching the unique sequence variants presented in (A) to indicate the representative sequence of the assigned OTU. Colored rectangles that span multiple bars indicate that the corresponding samples were assigned the same dominant OTU. Rectangles are annotated with the taxonomic assignment of the OTU (see Methods).



Figure 3(on next page)

Symbiodinium community composition of each sample

Samples are plotted as horizontal bars, sorted by species and shore of collection (north vs. south). Individual segments of each bar represent 97% within-sample OTUs, colored by *Symbiodinium* clade identity.

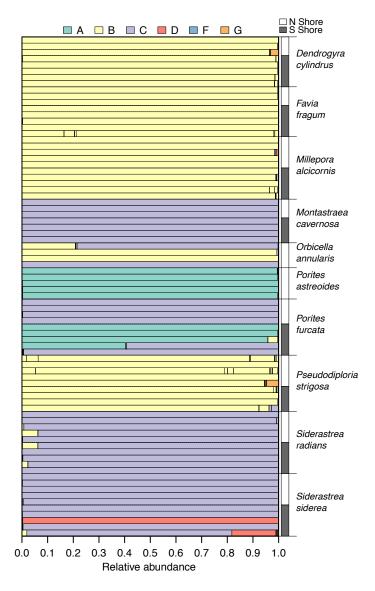


Figure 3: Figure 3. Symbiodinium community composition of each sample. Samples are plotted as horizontal bars, sorted by species and shore of collection (north vs. south). Individual segments of each bar represent 97% within-sample OTUs, colored by Symbiodinium clade identity.



Figure 4(on next page)

Network analysis of abundant Symbiodinium taxa in each coral species.

Symbiodinium OTUs (circular nodes) are connected to each coral species (square nodes) in which they ever occurred at >1% relative abundance within an individual; thickness of edges (i.e., links between coral and symbiont nodes) is relative to the proportion of individuals within the coral species in which the *Symbiodinium* OTU was present at >1%. Symbiont nodes are colored by clade identity, and sized proportionally to the number of coral species in which they were found.

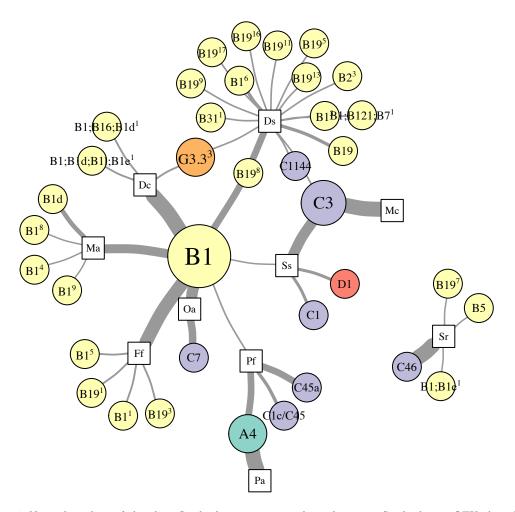


Figure 4: Figure 4. Network analysis of abundant Symbiodinium taxa in each coral species. Symbiodinium OTUs (circular nodes) are connected to each coral species (square nodes) in which they ever occurred at >1% relative abundance within an individual; thickness of edges (i.e., links between coral and symbiont nodes) is relative to the proportion of individuals within the coral species in which the Symbiodinium OTU was present at >1%. Symbiont nodes are colored by clade identity, and sized proportionally to the number of coral species in which they were found.



Figure 5(on next page)

Network analysis of dominant and background symbionts

Network analysis of dominant (A) and background (B) *Symbiodinium* taxa in each coral species (see Materials and Methods for details on network construction). Symbiont nodes are colored by clade identity, sized proportionally to the number of coral species to which they are connected.



Figure 5: Figure 5: Network analysis of dominant (A) and background (B) Symbiodinium taxa in each coral species (see Materials and Methods for details on network construction). Symbiont nodes are colored by clade identity, sized proportionally to the number of coral species to which they are connected.



Figure 6(on next page)

Symbiotic flexibility (beta diversity) of host species.

Symbiotic flexibility in coral species quantified as multivariate dispersion of Symbiodinium community composition (mean distance to centroid) \pm SE. Host taxa that do not share a letter are significantly different (p < 0.05).

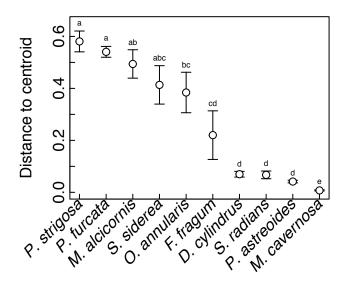


Figure 6: Figure 6: Symbiotic flexibility in coral species quantified as multivariate dispersion of Symbiodinium community composition (mean distance to centroid) \pm SE. Host taxa that do not share a letter are significantly different (p < 0.05).



Table 1(on next page)

Summary statistics for each clustering approach.



Table 1. Summary statistics for each clustering approach. $97\% \hspace{1.5cm} 97\%$

	, , , ,	<i>> 1 , 0</i>	
	OTUs(across	OTUs(within	
	samples)	samples)	100% OTUs
Number of	94	106	4718
OTUs			
Range of	10 - 742671	10 - 472752	10 - 171212
OTU counts			
Range of	706 - 169884	707 - 169890	485 - 97003
reads per			
sample			
Geometric	13141 */ 2	13137 */ 2	8141 */ 2
mean (*/			
GSD) reads			
per sample			



Table 2(on next page)

Differences in Symbiodinium communities within and between shores

Mean overall, within-shore ('within'), and between-shore ('between') Bray-Curtis dissimilarities of the *Symbiodinium* communities in each host species, and PERMANOVA tests (partial R² and p-values) for a difference between shores. Between-shore tests were not possible for *O. annularis* and *P. astreoides* since they were only sampled from one shore.



Table 2. Mean overall, within-shore ('within'), and between-shore ('between') Bray-Curtis dissimilarities of the Symbiodinium communities in each host species, and PERMANOVA tests (partial R2 and p-values) for a difference between shores. Between-shore tests were not possible for O. annularis and P. astreoides since they were only sampled from one shore.

J 1	n	overall	within	between	R2	p
Millepora	10	0.604	0.653	0.564	0.037	0.809
alcicornis						
Orbicella	4	0.632	0.632	-	-	-
annularis						
Porites	5	0.004	0.004	-	-	-
astreoides						
Montastraea	7	0	0.001	0	0.118	1
cavernosa						
Dendrogyra	8	0.017	0.016	0.019	0.188	0.268
cylindrus						
Favia fragum	8	0.251	0.309	0.202	0.086	0.75
Porites	9	0.703	0.555	0.821	0.345	0.056
furcata						
Siderastrea	10	0.03	0.031	0.029	0.064	0.953
radians						
Siderastrea	10	0.512	0.541	0.489	0.061	0.467
siderea						
Pseudodiplori	9	0.789	0.71	0.851	0.242	0.079
a strigosa						