



Spatial and Seasonal Variability of Reef Bacterial Communities in the Upper Gulf of Thailand

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Reefs at Ko Samae San (S), Khao Ma Cho (K), and Ko Tao Mo (T), in the Gulf of Thailand (GoT) represent a biodiversity hotspot, and bacteria play significant roles in maintaining the health of these coral reefs and their biogeochemical cycles. Therefore, this study analyzed bacterial communities (microbiota) from healthy corals and nearby seawater and sediment, using B-RISA and 16S rRNA gene sequencing. Sampling was done in one dry and one wet season to provide an initial assessment of variation with environmental conditions. The most prevalent coral species were Porites lutea, Platygyra sinensis, Acropora humilis, and Acropora millepora. The B-RISA and the sequencing results were correlated, which increased confidence the results. The microbiota varied among corals, seawater, and sediment and between the wet and dry seasons. Percentages of bacteria with known functions varied among sample types and seasons, and their relative abundances supported previously reported essential functions, such as prevention of disease (e.g., Pseudoalteromonas, Psychrobacter, and Cobetia were more abundant on corals in the dry season). Pearson's correlations and multiple factor regressions identified dissolved oxygen (DO), temperature, salinity, and density as significant influences on the microbiota. The equations estimated the relative abundance of a community comprising 147 bacterial genera, as well as the relative abundance of Pseudomonas, Clostridium, Verrucomicrobium, and Epulopiscium ($R^2 \ge 0.721$). These results represent the first descriptions of microbiota from corals, and surrounding seawater and sediments in the upper GoT.

Keywords: diversity, microbiota, microbiome, reef, coral

INTRODUCTION

Recently, an advance in next generation sequencing has allowed metagenomics to identify the bacteria diversity (or microbiota) in a manner that yields improved characterization of assemblages because it eliminates any bias associated with culturing (Riesenfeld et al., 2004; Khitmoh et al., 2017). Microorganisms, particularly bacteria, are abundant and play major roles in the metabolic functioning of almost every biological system, including highly productive systems like coral reefs (Webster and Reusch, 2017; Kelly et al., 2018). Bacteria live independently or associated

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with corals, and have roles that include obtaining and recycling resources, such as carbon and nitrogen. Vibrionales, Pseudomonadales, Oceanospirillales, and Oscillatoriales are examples of bacterial orders reported to be coral holobionts because their symbiotic relationships support the persistence of the coral reefs (Rohwer et al., 2002; Reshef et al., 2006; Rosenberg et al., 2007). Corals provide habitat while bacteria photosynthesize, acquire, and decompose organic and inorganic nutrients, and produce antibiotics and antioxidants to boost immunity for corals and promote resistance against invasions by pathogens and environmental stress (e.g., global warming). Moreover, commensal bacteria compete with pathogens for space and nutrients (Lesser et al., 2004; Rosenberg and Loya, 2004; Rosenberg et al., 2007; Lema et al., 2012; Rädecker et al., 2015; Webster et al., 2016). Coral holobionts have been reported to be specific to sites and coral species (Rohwer et al., 2002; Koren and Rosenberg, 2006; Chen et al., 2011; Carlos et al., 2013). The presence of key bacterial symbionts also has been reported in the surrounding seawater and sediment (Austen et al., 2002; Rusch and Gaidos, 2013). This study utilized bacterial-ribosomal intergenic spacer analysis (B-RISA) and the 16S rRNA gene sequencing to characterize the symbiotic microbiota of healthy corals from sites in the upper Gulf of Thailand (GoT). The study sites are considered a global hotspot for marine animal biodiversity (Phongsuwan et al., 2013; Ramírez et al., 2017).

The coastline along the GoT is nearly 1,600 km long, spanning \sim 100 islands along the east and 150 islands along the west. Fringing reefs are found throughout this region, and the prevalent coral species are Acropora, Platygyra, and Porites (Phongsuwan et al., 2013; Latypov, 2015). Unfortunately, due to global warming, many coral populations in Thailand have suffered periodic coral bleaching (Baker et al., 2008; Chavanich et al., 2009; Hoeksema and Matthews, 2011; Phongsuwan and Chansang, 2012). This problem increases the importance of a database documenting the microbiota of corals that may help them cope with stresses. This study included characterization of microbiota for four dominant species, Porites lutea, Platygyra sinensis, Acropora humilis, and Acropora millepora, and it also compared their microbial communities with those in the surrounding seawater and sediment during one dry and one wet season.

B-RISA is an established method that uses electrophoresis to classify bacterial diversity based on characteristic intergenic DNA lengths between the 16S and 23S rRNA genes of bacterial species. Different species (or families) have different B-RISA lengths. B-RISA is simple and inexpensive, and it serves as cross-validation for estimates of species richness derived from 16S rRNA gene sequencing, but it does not identify individual taxa (Danovaro et al., 2006; Grillevet et al., 2009; Sikaroodi and Gillevet, 2012; Gobet et al., 2014; Jami et al., 2014; Dehingia et al., 2015).

In contrast, data on 16S rRNA gene sequences can be used to identify operational taxonomic units (OTUs), which supports estimates of alpha and beta diversity, correlations with environmental factors (such as temperature and salinity), regressions to predict relative abundances in different environmental conditions, and evaluation of microbial metabolic functions. Because the metabolic functions of bacterial taxa were available, the metabolic potential of the bacterial community can be estimated from its composition. Knowledge of the microbiota can be considered crucial to sustaining and rehabilitating coral reefs.

MATERIALS AND METHODS

Sampling

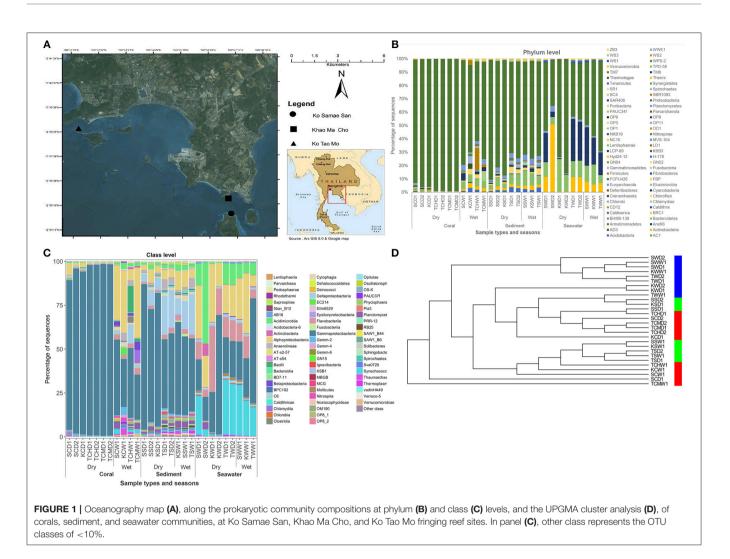
Samples were collected at S, K, and T during one dry (April-May) and one wet (September–October) seas The authors also thanked P. Sodsai and W. Poomipak for sequencing advice. on in 2014 (**Figure 1A**). The coral reefs at S, K, and T appeared healthy in both seasons. In each season, 3–10 replicate samples of seawater, sediments, and the prevalent coral species (*P. lutea*, *P. sinensis, A. humilis,* and *A. millepora*) were collected. Corals that were 5 cm in diameter were collected, along with 5 L each of sediment directly below each colony within 1 m and 50 g each of sediment directly below each colony within 1 m. All samples were placed on ice immediately, and genetic analyses were performed upon samples arrival within 7–14 days. The environment at each site was characterized by measuring the temperature, pH, salinity, density, and concentration of dissolved oxygen (DO) in the seawater.

Metagenomic Extraction

Metagenomic DNA was isolated using Power Water DNA Isolation Kit (for seawater samples) and Power Soil DNA Isolation Kit (coral and sediment samples) (MoBio, Carlsbad, CA, USA), following the manufacturer's instructions. Microbiota in seawater were captured by filtering 2.5 L using a sterile 0.22- μ vacuum filtration system (Merck Millipore, Massachusetts, USA). For coral and sediment samples, 1 g of coral was homogenized by grinding and 1 g of sediment were used. Metagenomic template in each replicate was extracted and checked for the quality and quantity of DNA by agarose gel electrophoresis and nanodrop spectrophotometry, respectively.

B-RISA to Identify Subgroups in Independent Sampling Replicates

B-RISA was performed according to established protocols, using universal primers 16S-1392F (5'-GYACACACCGCCCGT-3') and 23S-125R (5'-GGGTTBCCCCATTCRG-3') (Hewson and Fuhrman, 2004; Danovaro et al., 2006; Gobet et al., 2014; Jami et al., 2014). To prevent PCR bias, two independently PCR reactions were performed and pooled for each sample. Each 12.5 μ l PCR comprised 1× EmeraldAmp? GT PCR Master Mix (TaKaRa, Shiga, Japan), 0.3 μ M forward primer, 0.3 μ M reverse primer, and 50 ng of template. The thermocycling conditions were 95°C for 4 min, followed by 30–35 cycles of 94°C for 1 min, 55°C for 1 min, and 72°C for 2 min, and ended with 72°C for 10 min. Subgroups were identified by banding patterns following agarose gel electrophoresis. Each band was quantified as a percentage of the sum of all bands using ImageJ.



16S rRNA Gene Library Preparation and Next Generation Sequencing

Libraries of 16S rRNA genes were constructed following Caporaso et al. (2012), using universal prokaryotic primers (5'-GTGCCAGCMGCCGCGGTAA-3') and 515F 806R (5'-GGACTACHVGGGTWTCTAAT-3'), appended with 5' Illumina adapter and 3' Golay barcode sequences. Each 25-µl PCR reaction comprised 1× EmeraldAmp® GT PCR Master Mix (TaKaRa), 0.2 µM of each primer, and 75 ng of the metagenomic DNA. The thermocycling conditions were 94°C for 3 min, and 28-30 cvcles of 94°C for 45 s, 50°C for 60 s, and $72^{\circ}C$ for 90 s, followed by $72^{\circ}C$ for 10 min. Triplicate PCRs were performed and pooled for each sample to prevent stochastic bias. The amplicons were agarose gel purified using GenepHlowTM Gel Extraction Kit (Geneaid Biotech Ltd., New Taipei City, Taiwan), quantified with Picogreen (Invitrogen, Eugene, Oregon, USA), and 240 ng of each sample (or subgroup) was pooled for sequencing by Miseq300 platform (Illumina, San Diego, CA, USA), along with the sequencing primers and index sequence (Caporaso et al., 2012). Sequencing was performed at Chulalongkorn Medical Research Center, Chulalongkorn University (Bangkok, Thailand).

Quality Screening, Taxon Classifications, and Community Comparison

Raw sequences were processed according to Mothur's standard operating procedures for MiSeq (Schloss et al., 2009). Quality screening included removal of sequences that have (i) ambiguous bases, (ii) >1 mismatch base in the primer region, (iii) ≥ 1 mismatch base in the barcode, (iv) >10 homopolymers, (v) a minimum sequencing quality score of <35 over a 50bp window, (vi) a read length of <350 bases, and (vii) chimera sequences (Huse et al., 2010; Edgar et al., 2011). Sequences were aligned against Greengenes (McDonald et al., 2012) to remove contaminated sequences, such as mitochondria and chloroplast sequences that also contain 16S rRNA gene. Operational taxonomic units (OTUs) were assigned based on a naïve Bayesian method (Wang et al., 2007), with a score threshold of 80% (default parameter). Samples were normalized to an equal sampling depth corresponding to the sample size of the fewest number of sequences (42,655). Good's coverage

to estimate a coverage index, rarefaction curve, alpha diversity (Chao1 richness, Shannon diversity, and inverse Simpson index), beta diversity [thetayc dissimilarity coefficient and principal coordinates analysis (PCoA)], and unweighted pair group with arithmetic mean (UPGMA) clustering were conducted using Mothur (Schloss et al., 2009). A complete linkage clustering analysis was computed using R. Similarity profile analysis (SIMPROF) using Bray Curtis dissimilarity indices was also performed as an additional test of the significance of groupings (p < 0.05) (p < 0.05). For sample where two independent sequencings were performed based on B-RISA subgrouping, a student's *t*-test was computed to if the composition of the subgroups differed significantly (p < 0.01), and the subgroups were merged if p > 0.01.

Multiple Factor Regression

The relative abundances of all microbiota were correlated with measures of environmental conditions by calculating Pearson's correlation coefficients in Mothur (Schloss et al., 2009), and the results were visualized with XLSTAT-Ecology software (www. xlstat.com/en/solutions/ecology). Regression equations were derived as described elsewhere (Pal and Bhattacharya, 2013; Teng et al., 2015; French and Finch, 2016; Jia et al., 2016). Assuming the bacterial community is a dependent variable changing upon the independent environmental variables, the relative abundance of bacteria can be predicted by the environmental variables. Significant environmental factors were extracted from the correlation matrix and combined using a principal component analysis (PCA) and the Varimax Factor Rotation, which yielded the eigenvalues and scores. Relevant eigenvalues and scores were selected using default thresholds (Pal and Bhattacharya, 2013; French and Finch, 2016; Jia et al., 2016). The significant factors and their scores were combined in regression equation to estimate the dependent variable Y (represent the bacterial abundance) from the independent variables X_1 to X_i (each of water physical variables) and β_0 to β_i as coefficient constants, as follow.

$$Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \dots + \beta_i X_i$$

For each equation, an Ordinary Least Square analysis yielded an *R*-squared (R^2) value, which represents its goodness of fit (a percent that *Y* can be explained by the X_i equation). R^2 of >0.7, or the closer to 1, is considered acceptable, meaning that \geq 70% that *Y* can be explained by the X_i equation. Similarly, the relative abundances of common bacterial genera were correlated with environmental parameters, and the results were used to generate regression equations.

Potential Metabolic Functions

The potential metabolic function of the community was predicted following the established protocols (Langille et al., 2013). The functional profiles of the communities were predicted from 16S rDNA sequences using PICRUSt (Langille et al., 2013). The metabolic functions were categorized by KEGG (Kyoto Encyclopedia of genes and genomes) pathways, given level 1

representing pathways and level 2 representing categories of genes (COGs).

Availability of Supporting Data

Nucleic acid sequences in this study were deposited in an NCBI open access Sequence Read Archive database, accession number SRP095762.

RESULTS

Sampling Sites, Environmental Conditions, and Metagenomic DNA

Differences in environmental conditions among sites (temperature, pH, salinity, DO, and density) were small. Larger differences were recorded between the two seasons for pH, salinity and density (**Table 1**).

The average concentration of microbial metagenomic DNA was higher in the sediments, followed by corals. For corals, the wet season yielded about three times more metagenomic DNA than the dry season (**Table 1**). From **Table 1** and so on, sample names are abbreviated by: the first letter represents site (S, K, T); the middle is sample types (Coral, Sediment, seaWater), and for site with two coral species the next letter denotes the species of the corals; and last letter is wet (W) or dry (D) seasons; followed by a number 1 only, or 1 and 2, depending on the number of B-RISA subgroups.

B-RISA to Identify Subgroups in Independent Sampling Replicates

Replicate samples of coral, sediment, and seawater fell into 1–2 B-RISA subgroups per sample (e.g., SCD1 and SCD2 for 2 B-RISA subgroups, and SCW1 for 1 B-RISA subgroup). The B-RISA patterns shared among the same sample types were similar (**Table 2**).

16S rRNA Gene Sequences

Every B-RISA subgroup was successfully 16S rRNA gene library constructed and sequenced, according to the number of quality reads (avg. 199,107 reads per subgroup), which yielded statistically sufficient coverage of averagely 99.99 and 99.90% of the community diversity at the phylum and genus levels, respectively (**Table S1**: Good's coverage indices). Further evidence of a sufficient sequencing depth was provided by the OTU richness curves, which reached plateau (**Figure S1**). Comparing across sample types, sediments and seawater had higher species diversity than corals (**Table S1**: Shannon diversity indices), and bacteria were found to be dominant in all samples (>98% of all OTUs), whereas archaea were absent or rare (**Figure S2**). Bacteria in phylum Proteobacteria were the most numerous, followed by Cyanobacteria, Bacteroidetes, Actinobacteria, and Firmicutes (**Figure 2**).

As **Figures 1B,C** illustrated the similarity between microbiotic communities in the replicate subgroups, student's *t*-tests were computed based on the abundances of OTUs at the genus level between pairs of B-RISA subgroups, when available. A cutoff of $p \le 0.01$ was used because $p \le 0.001$ might be too stringent. Most pairs were found to be similar (p > 0.1), except TSD1 and TSD2

Sources	Types of samples	Sample depth	Seasons	Temp. (°C)	рН	Salinity (psu)	DO (mg/L)	Density (kg/m ³)	B-RISA pattern	•	A260/280	Metagenomic DNA concentration
Ko Samae	Coral (Porites	2	Dry	30.2	8	31.17	5.41	1022.76	1	SCD1	1.75	820
San	lutea)								2	SCD2	1.94	940
			Wet	29.9	7.45	30.72	5.37	1018.8	1	SCW1	1.79	3,020
	Sediment	2	Dry	30.2	8	31.17	5.41	1022.76	1	SSD1	1.79	6,287
									2	SSD2	1.81	6,040
			Wet	29.9	7.45	30.72	5.37	1018.8	1	SSW1	1.82	5,472
	Water	2	Dry	30.2	8	31.17	5.41	1022.76	1	SWD1	1.89	2.44
									2	SWD2	1.86	2.10
			Wet	29.9	7.45	30.72	5.37	1018.8	1	SWW1	1.87	2.25
Khao Ma Cho	Coral (Platygyra	3	Dry	30.2	8	31.17	5.41	1022.76	1	KCD1	1.73	830
	sinensis)		Wet	29.9	7.45	30.72	5.37	1018.8	1	KCW1	1.73	3,602
	Sediment	3	Dry	30.2	8	31.17	5.41	1022.76	1	KSD1	1.83	8,200
			Wet	29.9	7.45	30.72	5.37	1018.8	1	KSW1	1.80	5,960
	Water	3	Dry	30.2	8	31.17	5.41	1022.76	1	KWD1	1.82	1.27
									2	KWD2	1.97	0.83
			Wet	29.9	7.45	30.72	5.37	1018.8	1	KWW1	1.89	1.25
Ko Tao Mo	Coral (Acropora	2	Dry	30.5	8.11	32.50	5.51	1025.4	1	TCHD1	1.96	660
	humilis)								2	TCHD2	1.81	557
			Wet	29.5	7.89	31.20	5.59	1021.7	1	TCHW1	1.76	3,033
	Coral (Acropora	2	Dry	30.5	8.11	32.50	5.51	1025.4	1	TCMD1	1.81	800
	millepora)								2	TCMD2	1.83	954
			Wet	29.5	7.89	31.20	5.59	1021.7	1	TCMW1	1.87	5,320
	Sediment	2	Dry	30.5	8.11	32.50	5.51	1025.4	1	TSD1	1.82	6,750
									2	TSD2	1.84	7,353
			Wet	29.5	7.89	31.20	5.59	1021.7	1	TSW1	1.84	14,804
	Water	2	Dry	30.5	8.11	32.50	5.51	1025.4	1	TWD1	1.88	1.56
									2	TWD2	1.84	0.95
			Wet	29.5	7.89	31.20	5.59	1021.7	1	TWW1	1.79	0.17

TABLE 1 | Locations, water physical properties, and metagenomic DNA of sample collections.

*Unit of concentration is in ng/ml for seawater, and ng/g for sediment and coral.

yielded a value near the cutoff (p = 0.019). In addition, SWD1 and SWD2 subgroups that correspond to water from Ko Samae San in the dry season replicates were found to contain dissimilar communities (p = 0.005) (**Table 3**).

Community Similarity by Sample Type

Figures 1B,C illustrated that the microbiota in seawater were different from those in other types of samples across sites and seasons, and that the microbiota of corals differed between seasons but not sites. These findings remained valid when the B-RISA subgroups were merged (**Figure 3**). To further prove that the communities clustered by sample types, the results of UPGMA clustering grouped the seawater communities, separate from the coral communities that seemed to be subdivided between the dry and wet seasons (**Figure 1D**), with similar results confirmed by complete linkage clustering (p < 0.001). The complete linkage clustering demonstrated three major clusters belonging mainly to the seawater communities (**Figure S3**: right branch), the dry-season coral communities

(middle branch), and the wet-season coral communities (left branch). The p statistic when clustering by sample type was statistically robust (p < 0.001), whereas clustering by seasons was less consistent (p = 0.076) and clustering by sites was not significant (p = 0.352). The smaller the p-value corresponds to the cluster separation with significance. As additional confirmation, SIMPROF documented the same clusters (**Figure S4**). Next, how each community is related in two-dimension plot was illustrated by PCoA, given several samples of the same type were overlapped in the position (**Figure 4A**). The PCoA showed similar results regardless of whether the subgroups were separated (**Figure 4A**) or merged (**Figure 4B**).

Correlation With Environmental Variables

Pearson's correlation coefficients indicated that communities varied with differences in temperature, DO, density, salinity, and pH (p < 0.01; **Figure 5**). The communities were not significantly correlated with the total numbers of OTU or depth of the water,

	nes				Seawater								Sec	Sediment			
Site		Ko Samae San	San		Khao Ma Cho	0		Ko Tao Mo		Y	Ko Samae San	an	Khao	Khao Ma Cho		Ko Tao Mo	0
Season		Dry	Wet		Dry	Wet		Dry	Wet		Dry	Wet	Dry	Wet		Dry	Wet
B-RISA patterns	erns SWD1	01 SWD2	SWW1	KWD1	KWD2	KWW1	TWD1	TWD2	TWW1	SSD1	SSD2	SSW1	KSD1	KSW1	TSD1	TSD2	TSW1
1,5	1,517 bp																
1,2(1,200 bp																
1,0(1,000 bp (37%,	. "					(33%,										
	1,000 hn	0					1,000 (ad										
106	900 hn						$\frac{1}{2}$										
08	800 bp				(22%,												
					800 bp)												
02	700 bp				(34%, 700 bp)												
09	600 bp (37%, 650 bp)	650 bp) (46%,		(42%, 600 bp)			(31%, 650 bb)			(13%, 520 bp)	(15%, 520 bp)	(100%, 500 bp)					(33%, 600 bp)
20	500 bp			(58%, 550 bp)			(22%,550 bp)			(34%, 500 bp)	(85%, 500 bp)	-	(100%, 550 bp)	(100%, 550 bp)	(66%, 580 bp)	(100%, 580 bp)	(34%, 500 bp)
40	400 bp (26%, 400 bp)	54%, (54%, bp) 400 bp)	(81%, 400 bp)			(100%, 400 bp)	(14%, 450 bp)	(100%, 450 bp)							(34%, 450 bp)		(33%, 400 bp)
30	300 bp																
50	200 bp		1							(53%, 250 hn)							
10	100 bp																
Sample names	nes								Coral								
Site		Ko	Ko Samae San (<i>P. lutea)</i>	(P. lutea)		Khao N	Khao Ma Cho (<i>P. sinensis</i>)	sinensis)		Ko Ta	Ko Tao Mo (A. <i>humilis</i>)	umilis)		-	Ko Tao M	Ko Tao Mo (<i>A. millepora</i>)	ıra)
Season			Dry		Wet	Dry		Wet		Dry			Wet		Dry		Wet
B-RISA patterns	erns	SCD1	SCD2		SCW1	KCD1	5	KCW1	TCHD1	Б	TCHD2	Ť	TCHW1	TCMD1	5	TCMD2	TCMW1
- 	1,517 bp																
1,1	1,200 bp																
1,0	1,000 bp																
б 	900 bp								(61%, 900 bp)	(dq 0(
ω ω	800 bp								(10%, 800 bp)								
	700 bp				(44%, 700 bp)	(100%, 700 bp)	(dq OC		(29%, 700 bp)		(100%, 700 bp)	(d				(100%, 700 bp)	(0
0	600 bp		(34%, 550 bp)		(22%, 600 bp)									(100%, 600 bp)	(dq 0(
2	500 bp (10	(100% 500 bp)	(65.85%, 500 bp)	(dq 0((100%	(100%, 500 bp)				
4	400 bp			(3	(34%, 400 bp)												
ल 	300 bp																
5	200 bp																
1	100 bp																

TABLE 2 | B-RISA pattern of prokaryotic communities.

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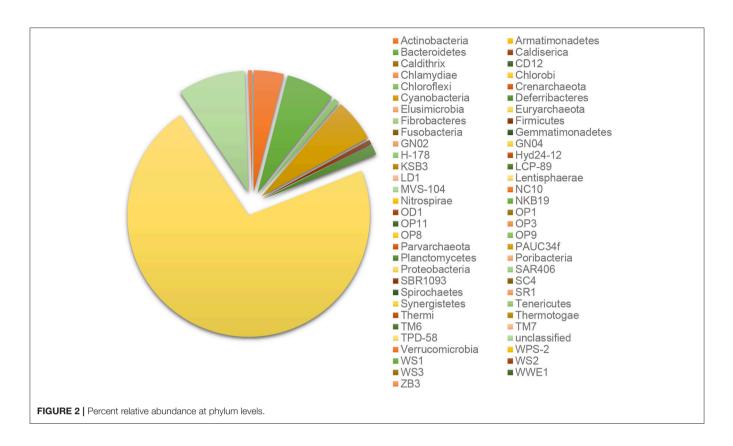


TABLE 3 | Determination of significant differences between B-RISA subgroups by student's *t*-test for prokaryotic communities.

B-RISA subgroups	Mean	Standard deviation	N	99% Confidence interval	p-value
SCD1	293.737	5, 187.253	4,500	-760.858,	0.120
SCD2	580.254	11,205.559	4,500	187.823	
SSD1	266.389	4,346.213	4,500	-413.435,	0.192
SSD2	405.359	5,670.889	4,500	135.495	
SWD1	501.237	6,443.038	4,500	22.260,	0.005*
SWD2	206.726	2,952.271	4,500	566.764	
KWD1	323.531	4,987.981	4,500	-453.610,	0.341
KWD2	445.898	7,033.751	4,500	208.875	
TCHD1	223.282	4,308.613	4,500	-690.846,	0.103
TCHD2	489.861	10, 142.263	4,500	155.294	
TCMD1	138.739	3,019.033	4,500	-261.50,	0.279
TCMD2	231.526	4,920.541	4,500	75.928	
TSD1	194.192	2,521.912	4,500	-430.609,	0.019
TSD2	399.400	5,298.835	4,500	20.194	
TWD1	147.658	1,860.419	4,500	-92.617,	0.876
TWD2	141.675	1,768.336	4,500	104.583	

*Represents p < 0.01.

which highlighted the reliability of the analysis because numbers of OTUs should not affect the assemblage composition when sequencing depths were sufficient. Furthermore, all samples were collected in 2–3 m of water, so this study did not have sufficient scope to evaluate the influence of depth.

Multiple Regression Analysis for Predictive Community and Individual Bacterial Genus Abundances From Environmental Variables

The following equation that predicts the total relative abundance of 147 OTUs (Y147; **Table S2**) yielded an R_2 of 0.999.

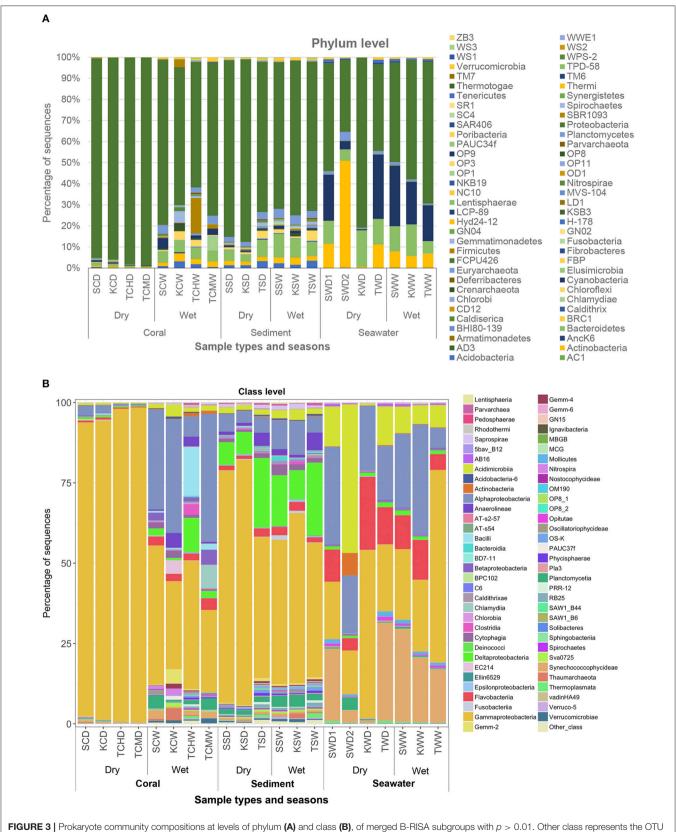
 $Y_{147} = 338.907 + 0.018X_1 + 10.572X_2 - 5.781X_3 + 66.6X_4 - 0.82X_5$

by X_1 is sample depth (unit m), X_2 is seawater temperature (unit °C), X_3 is salinity (unit psu), X_4 is DO (unit mgL⁻¹), X_5 is density (unit kgm⁻³).

All 146 of 147 OTUs in the Y_{147} were found for the TCHW1 community, and thus the 146 bacterial OTUs of this community could be reliably predicted.

The following equations (**Table S2**) also yielded $R_2 \ge 0.7$, and they predicted the relative abundance of *Pseudomonas* ($Y_{Pseu.}$), *Clostridium* ($Y_{Clos.}$), *Verrucomicrobium* ($Y_{Verr.}$), and *Epulopiscium* ($Y_{Epul.}$).

$$\begin{split} Y_{Pseu.} &= 173.237 - 0.07X_1 + 4.307X_2 - 2.334X_3 \\ &\quad + 27.776X_4 - 0.372X_5, \quad R^2 = 0.908 \\ Y_{Clos.} &= 45.851 - 0.002X_1 + 1.359X_2 - 0.746X_3 \\ &\quad + 8.551X_4 - 0.107X_5, \quad R^2 = 0.982 \end{split}$$



classes of <10%.

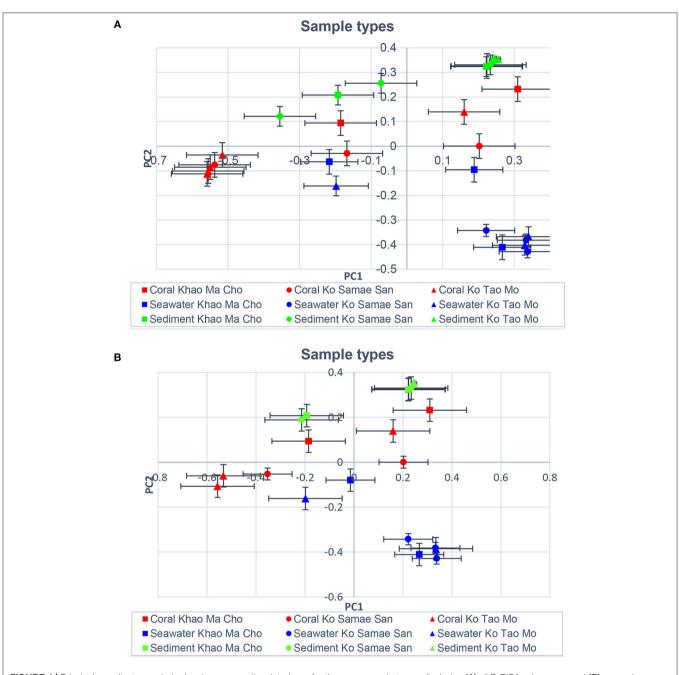


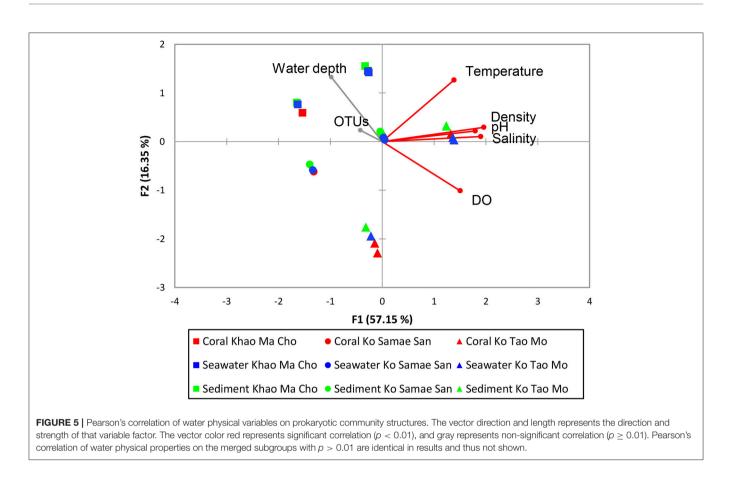
FIGURE 4 | Principal coordinates analysis showing community relatedness for the same sample types, displaying (A) all B-RISA subgroups, and (B) merged subgroups with p > 0.01.

 $Y_{Verr.} = 11.323 - 0.007X_1 + 0.19X_2 - 0.092X_3$ +1.046X_4 - 0.019X_5, $R^2 = 0.721$ $Y_{Epul.} = 5.849 + 0.003X_1 + 0.184X_2 - 0.102X_3$ +1.168X_4 - 0.014X_5, $R^2 = 0.946$

by X_1 is sample depth (unit m), X_2 is seawater temperature (unit °C), X_3 is salinity (unit psu), X_4 is DO (unit mgL⁻¹), X_5 is density (unit kgm⁻³).

Distributions Across Sample Types

Figure 6 illustrated the frequencies of bacterial orders previously reported as important for corals: Vibrionales, Pseudomonadales, Oceanospirillales, Flavobacteriales, Acidimicrobiales, Synechococcales, Desulfobacterales, and



Oscillatoriales. These groups dominated in samples of corals and sediments. Vibrionales (mostly members of genus *Pseudoalteromonas* accounted 96.3% of Vibronales), Pseudomonadales (*Psychrobacter* 91.26%), and Oceanospirillales (*Cobetia* 75.86% and *Endozoicomonas* 0.01%) were abundant on corals (55–75%) and in the sediments (12–19%) during the dry season. In contrast, in the wet season, sediments contained 86% Oscillatoriales (cyanobacteria in the genus *Lyngbya*) and 79% Desulfobacterales (sulfate-reducing bacteria), which contrasted with their 49 and 36% contributions to abundances in the dry season (**Figure 6**).

In seawater, Flavobacteriales, Acidimicrobiales, and Synechococcales were prevalent. The prior two genera were more abundant in the dry season (47–58%) than during the wet season (25–35%), while Synechococcales showed the opposite trend (dry season 37% vs. wet season 58%; **Figure 6**).

Diversity of Metabolic Function Potentials

The metabolic potentials predicted from the community compositions indicated the KEGG pathway of "metabolism" (avg. 49.42%) was the most abundant, followed by "genetic information processing" (16.3%), "unclassified" (14.47%), and "environmental information processing" (13.75%; **Figure 7A**). The functional classification from the KEGG pathways into the COGs (**Figure 7B**) indicated a prevalence of "membrane transport" (i.e., transports large molecules such as polysaccharides, proteins, and siderophores; 11.43%), "amino

acid metabolism" (10.47%), and "carbohydrate metabolism" (i.e., sugar alcohols and central carbohydrates; 9.8%) (Figure 7B). These functions are related to important processes in coral reefs, including metabolism of carbohydrates, amino acids, and lipids (Tout et al., 2014). Functions such as DNA replication and repair, transcription, translation, energy metabolism, or the incorporation of key cofactors and vitamins are also important for coping with abiotic pressures, such as parasites, heavy metal contamination, ultraviolet radiation, exposure to reactive oxygen, etc. (Tout et al., 2014; Ainsworth et al., 2015; Badhai et al., 2016). COGs involved in cellular processes and signaling (i.e., chemotaxis, two-component system response), membrane transport, signal transduction, and cell motility (Figure 7B), might facilitate cellular interactions and metabolic exchanges between the bacterial community and the coral host (Tout et al., 2014; Ainsworth et al., 2015).

DISCUSSION

Coral reefs, also known as the tropical rainforest of the sea, are a crucial source of the Earth's food supplies (bioproductivity) and biodiversity. Coral microbiota plays a critical role in the preservation of healthy corals and coral reefs. While reefs in the GoT are hotspots of diversity (Ramírez et al., 2017), knowledge, and data for their microbiota are lacking. This study represented the first culture-independent microbiota databases of the upper

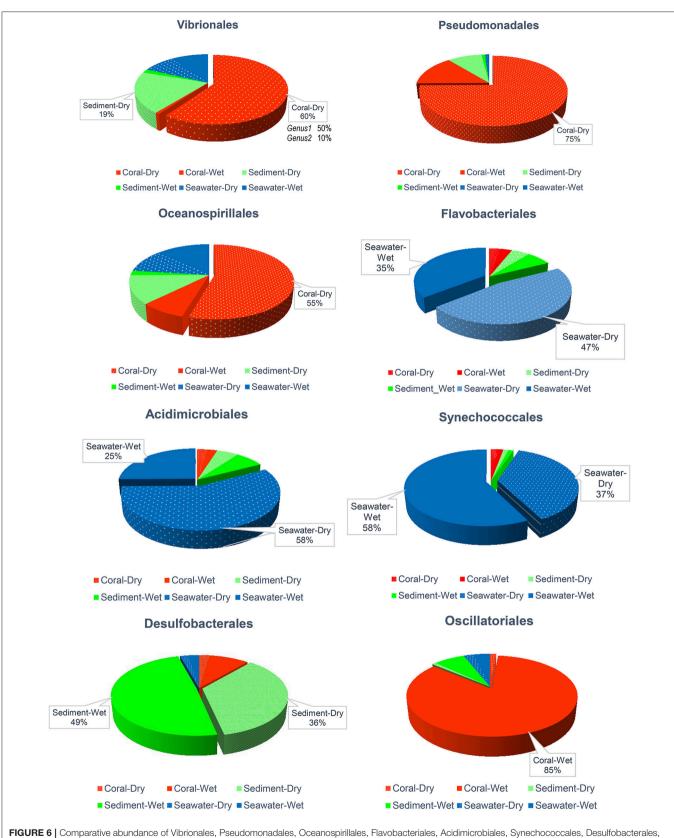
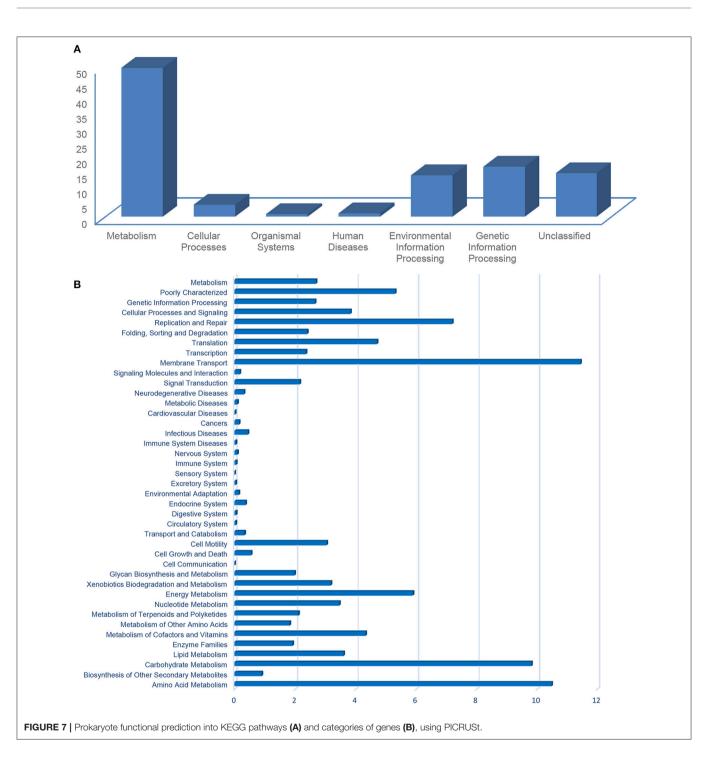


FIGURE 6 Comparative abundance of Vibronales, Pseudomonadales, Oceanospirillales, Flavobacteriales, Acidimicrobiales, Syne and Oscillatoriales orders across sample types and seasons. Samples with \leq 18% were omitted written percentages.



GoT coral sites, that characterized differences in assemblages from corals, seawater and sediment in one wet and one dry season.

Comparisons of microbial metagenomes among coral, seawater, and sediment identified seawater as having the lowest concentration of metagenomic material, which is consistent with the dilute and pristine nature of the seawater. For corals, higher metagenomic concentrations were documented during the wet season, suggesting higher microbial abundance and activity during this season. Overall, bacteria dominated the prokaryotic communities, with <1% archaea. This result supports previous reports of 0–1% of archaea in healthy coral systems (Wegley et al., 2007; Biers et al., 2009; Gilbert et al., 2012; Somboonna et al., 2014). Archaea were reported in polluted environments, such as those contaminated by heavy metals and oil (Biers et al., 2009; Somboonna et al., 2012). Analyzing the community diversity among the different samples, seasons, and sites showed that the communities in seawater clustered together, the communities in

sediments showed some similarity with the coral communities, and the communities on corals separated according to season. This ecotype-specific microbiota underlie the coral-associated community. The similar communities among the three sites, which share similar environmental condition and healthy corals, indicates the generality of these bacterial associations.

Vibrionales, members of genus Vibrio spp., can be pathogenic and cause disease in dry seasons; however, the dominant members at our study sites, particularly on the coral A. humilis, were Pseudoalteromonas spp. Pseudoalteromonas spp. are not pathogenic, and they were reported to have antibacterial activity on A. millepora and Fungia granulose (Kooperman et al., 2007; Littman et al., 2009). This finding suggests an increasing antibacterial activity during the dry season, when outbreaks of bacterial disease occur most often. Similarly, the predominant orders Pseudomonadales (genus Psychrobacter) and Oceanospiralles (Cobetia and Endozoicomonas) likely benefit the corals. Psychrobacter supports the metabolism of complex carbohydrates, amino acids, and lipids, and were reported in abundance in the mucus of Acropora and Porites (McKew et al., 2012; Badhai et al., 2016). For members of Oceanospiralles, interactions with corals remain unclear, although they have been reported as relatively abundant in tropical reefs. Some studies suggested their function might involve sulfur and nitrogen cycling, as well as produce hydrolytic enzymes to assist the corals in acquiring nutrients (Goffredi et al., 2005; Robertson et al., 2016). On the other hand, some studies have suggested their associations with coral diseases such as white plaque disease (Cárdenas et al., 2012). In the wet season, the microbiota of corals was dominated by Lyngbya and Trichodesmium of Oscillatoriales. Lyngbya is a benthic cyanobacteria that plays an important role as a producer of nutrients and bioactive compounds (Paul et al., 2005; Cruz-Rivera and Paul, 2006, 2007; Sharp et al., 2009). Trichodesmium is fundamental for nitrogen fixation and was reported to play a significant role in the food chain of the Belizean barrier reef and elsewhere (Villareal, 1995), including the GoT (Cheevaporn and Menasveta, 2003). Finding Desulfobacteraceae in the sediment was expected, as these bacteria are known to recycle organic sulfates (Gobet et al., 2012). For seawater, Flavobacteriales (phylum Bacteroidetes) and Acidimicrobiles (phylum Actinobacteria) that were reported as common at our sites had also been reported elsewhere (Lauro et al., 2009; Mizuno et al., 2015). These bacteria can scavenge organic debris for carbon and energy sources, supporting their growth and survival under nutrient limiting conditions in seawater. These species are believed to support cycling of the food network (Kelly et al., 2014). In total, the species denoted as coral microbiota from sites in the upper GoT include many potentially beneficial bacteria that have been reported as essential previously.

Pearson's correlations determined that temperature, DO, density, salinity, and pH impact the structure of bacterial communities. Multiple regressions with the acceptable R²-values allowed the prediction of the combined relative abundance of 147 OTUs, and also the relative abundances of *Pseudomonas*, *Clostridium*, *Verrucomicrobium*, and *Epulopiscium*, from the environmental conditions. The ability to predict the spatial and temporal variability of microbial communities of coral reefs

and their associations with key ecological processes can provide substantial benefits, in particular predicting responses to global warming and anthropogenic pollution. Indeed, our equations were validated by replacing the environmental parameters (*Xi*) from the TCHW1 data in **Table 1**, and the microbial abundance between the predicted equation vs. the sequencing result were compared: Y_{147} yielded 4.95% (the sequencing yielded 5.10%), Y_{Pseu} yielded 2.53% (the sequencing yielded 2.36%), and Y_{Clos} yielded 1.14% (the sequencing yielded 0.70%), for examples. Although the difference between the sequencing and the prediction was small, the predictive equations are limited to certain species and sites. More intense sampling at more sites with more diverse environments, may allow a more complete characterization of reef microbiota.

In conclusion, these findings represent the first cultureindependent characterization of microbiota on corals, in seawater and in sediments in the upper GoT. The database serves as a reference for other sites with similar characteristics. Bioinformatics has contributed to the understanding of the spatial and seasonal variability of reef microbiota, and its association with coral species, and different environmental characteristics. The multiple factor regression analysis provided is an example of the benefits of the microbiota databases. Additional studies of microbiota that sample multiple times in different seasons and at more sites with corals in various states of health are ongoing, and the results should help us decipher the core microbiota that are indispensable for maintaining the health of corals.

AUTHOR CONTRIBUTIONS

DB conducted molecular biology experiments and performed data analysis. AW and PK helped in analysing the data. VV conceived the study. SC provided samples and conceived the study. NS conceived the study, coordinated the experiments, analysed the data, and wrote the manuscript. All authors read and approved the final manuscript.

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SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fmars. 2018.00441/full#supplementary-material

Figure S1 | Estimated taxonomic richness (rarefaction curve) at levels of phylum (A), class (B), order (C), family (D), and genus (E) of prokaryotic communities.

Figure S2 | Percent relative abundance of prokaryotes categorized into bacteria and archaea.

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Figure S3 | Complete linkage cluster analysis of corals, sediment and seawater communities, at Ko Samae San, Khao Ma Cho, and Ko Tao Mo fringing reef sites.

Figure S4 | A test of significance for profiles clustering by SIMPROF. (A) All B-RISA subgroups and (B) merged subgroups with p > 0.01. Significant clustering (p < 0.05) is indicated with shared color branches.

Table S1 | Estimates of sample coverage and alpha diversity indices at phylum and genus levels of 16S rDNA profiles.

Table S2 | List of 147 genus level OTUs predicted by $Y_{147} = 338.907 + 0.018 X_1 + 10.572 X_2 - 5.781 X_3 + 66.6 X_4 - 0.82 X_5.$

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Conflict of Interest Statement: The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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