Microbial Communities in the East and West Fram Strait During Sea Ice Melting Season

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Climate models project that the Arctic Ocean may experience ice-free summers by the second half of this century. This may have severe repercussions on phytoplankton bloom dynamics and the associated cycling of carbon in surface waters. We currently lack baseline knowledge of the seasonal dynamics of Arctic microbial communities, which is needed in order to better estimate the effects of such changes on ecosystem functioning. Here we present a comparative study of polar summer microbial communities in the ice-free (eastern) and ice-covered (western) hydrographic regimes at the LTER HAUSGARTEN in Fram Strait, the main gateway between the Arctic and North Atlantic Oceans. Based on measured and modeled biogeochemical parameters, we tentatively identified two different ecosystem states (i.e., different phytoplankton bloom stages) in the distinct regions. Using Illumina tag-sequencing, we determined the community composition of both free-living and particle-associated bacteria as well as microbial eukaryotes in the photic layer. Despite substantial horizontal mixing by eddies in Fram Strait, pelagic microbial communities showed distinct differences between the two regimes, with a proposed early spring (pre-bloom) community in the ice-covered western regime (with higher representation of SAR11, SAR202, SAR406 and eukaryotic MALVs) and a community indicative of late summer conditions (post-bloom) in the ice-free eastern regime (with higher representation of Flavobacteria, Gammaproteobacteria and eukaryotic heterotrophs). Co-occurrence networks revealed specific taxon-taxon associations between bacterial and eukaryotic taxa in the two regions. Our results suggest that the predicted changes in sea ice cover and phytoplankton bloom dynamics will have a strong impact on bacterial community dynamics and potentially on biogeochemical cycles in this region.

Keywords: Arctic Ocean, phytoplankton bloom, microbial interactions, bacterioplankton, network analysis

INTRODUCTION

In recent decades, Arctic warming has resulted in remarkable environmental changes in the Arctic Ocean, and the region is warming much faster than the global mean rate (Dobricic et al., 2016; Sun et al., 2016). Arctic sea ice has declined by approximately 50% since the late 1950s, and its extent is shrinking at approximately 10% per decade since the late 1990s (Kwok and Rothrock, 2009;
were the first to describe changes of bacterial community microorganisms requires physical sampling. Wilson et al. (2017). However, monitoring the dynamics of heterotrophic (chl a) by satellites in ice-free ocean areas, with substantial extent, be monitored using remote sensing of chlorophyll a (Nöthig et al., 2015). Phytoplankton bloom dynamics may, to some extent, be monitored using remote sensing of chlorophyll a (chl a) by satellites in ice-free ocean areas, with substantial limits due to Arctic fog and the dark season (Perret et al., 2011). However, monitoring the dynamics of heterotrophic microorganisms requires physical sampling, Wilson et al. (2017) were the first to describe changes of bacterial community composition in the eastern Fram Strait throughout a polar year. In accordance with observations from other polar regions (Alonso-Sáez et al., 2008; Iversen and Murrell, 2011; Ghiglione and Murray, 2012; Williams et al., 2012), their results showed that the extreme seasonality of polar marine ecosystems, with ice-covered dark winter conditions and extended irradiance in summer, leads to pronounced seasonal differences in heterotrophic bacterial communities. Winter-time bacterial communities in the upper water column showed higher phylogenetic and functional diversity compared to the summertime, with increased importance of chemolithotrophic processes (e.g., Alonso-Sáez et al., 2014; Müller et al., 2018). During late spring, the increasing irradiance and decreasing sea ice cover initiate large phytoplankton blooms, which can lead to major shifts in heterotrophic bacterial community composition.

Biological interactions among microbes are important drivers of the dynamics in pelagic microbial communities (Fuhrman et al., 2015). Specific interactions between phytoplankton and heterotrophic bacteria have been documented, many of which are based on the exchange of energy sources and metabolites, including various forms of chemical signaling (Cole, 1982; Grossart et al., 2006; Grossart and Simon, 2007; Ramanan et al., 2016). Analyses of bacterial communities co-occurring with diatoms, using advanced molecular approaches, revealed complex interspecies signaling (Amin et al., 2012). While a full characterization of such interactions requires targeted experiments under laboratory conditions, molecular methods in combination with network analyses allow us to identify potential interactions directly from environmental samples (e.g., Gilbert et al., 2012; Lima-Mendez et al., 2015; Peura et al., 2015; Milici et al., 2016; Chafee et al., 2018).

One such interaction with relevance to the proportion of pelagic recycling versus carbon export is the physical association of bacteria with plankton detritus. Pelagic bacteria have different strategies to tap into the detritus pool, free-living in the water column or associated with particulate matter (Stocker, 2012). Previous studies have revealed strong differences between potential associations of free-living (FL) and particle-associated (PA) bacteria with microbial eukaryotes (Lima-Mendez et al., 2015; Milici et al., 2016). While the FL fraction is often dominated by cosmopolitan oligotrophic bacteria that rely on the availability of organic matter in the dissolved fraction (Morris et al., 2012; Giovanni et al., 2014), the PA fraction is usually represented by copiotrophic motile bacteria which colonize living or decaying microbial eukaryotes, fecal pellets, gel-like particles or other forms of particulate organic matter (Simon et al., 2002; Herndl and Reinthaler, 2013; Busch et al., 2017).

Microbial studies of the photic layer of Fram Strait have so far focused on eukaryotic plankton (Killas et al., 2013; Nöthig et al., 2015; Metfies et al., 2016), and biogeochemical recycling of detritus by bacteria (Piontek et al., 2014, 2015). Although bacteria are key players in the biogeochemical cycling of carbon and nutrients in the water column (Azam and Malfatti, 2007; Falkowski et al., 2008), very little is known about the composition and the dynamics of their communities in this region. In order to understand the impact of projected environmental changes on these communities,
it is necessary to establish a fundamental knowledge about
the biogeography and variability of microbial communities
in the Fram Strait. Using a set of measured and modeled
environmental parameters and sequence-based assessments
of microbial community composition, the objectives of the
study were: (1) to identify differences in bacterial community
composition in the two hydrographic regimes of Fram Strait in
relation to hydrographical and biogeochemical parameters; (2) to
test whether these differences are related to specific productivity
phases of the Arctic pelagic ecosystem; (3) to assess whether and
to what extent these differences are reflected in specific taxon–
taxon associations between bacterial and eukaryotic community
members.

RESULTS

Phytoplankton Bloom Dynamics Across
the Fram Strait

Based on previously defined physical characteristics of the two
main currents of Fram Strait (Rudels et al., 2013), we identified
two origins of our sample sets: (1) the eastern Fram Strait with
warmer and more saline Atlantic Water of the WSC; (2) the
western Fram Strait with colder and less saline Polar Water of the
EGC (Figure 1). The two regions had distinct sea ice conditions
at the time of sampling, with an ice-covered regime in EGC and
an ice-free regime in WSC (Figure 1). Furthermore, measured chl
a concentrations showed higher concentrations in the WSC, and
chla was present down to water depths of more than 100 m in
this region (Figure 1B).

In WSC all measured inorganic nutrients (silicate – SiO$_3$,
nitrate – NO$_3$, and phosphate – PO$_4$) showed lower concentrations near the surface compared to deeper
water layers below the pycnocline (roughly below 50 m). Contrary, in EGC there were only small differences in nutrient
concentrations throughout all measured depths. In addition,
while measurements of SiO$_3$ and PO$_4$ concentrations in
deep water layers were similar between the regions, NO$_3$
concentrations were lower in EGC (Supplementary Figure 1).
The depth of the water column pycnocline represents the
mixed layer depth during the last winter (Rudels et al., 1996).
Generally only the nutrients above the pycnocline within the
photic zone (upper ~50 m) are consumed by phytoplankton.
Therefore, the calculated differences in nutrient concentrations
($\Delta$) below and above the seasonal pycnocline provide a proxy
estimation for phytoplankton productivity in the different
regions, since the beginning of the seasonal bloom (Table 1).
The estimated productivity based on the stoichiometry of
consumed nutrients (see Material and Methods), as well as the
integrated chl a and phytoplankton carbon biomass all showed
higher values in WSC. Furthermore, based on a ratio 1:1 of
NO$_3$:SiO$_3$ we estimated that the contribution of diatoms to
the total productivity was roughly 30% in both regions. However,
biomass estimates of diatoms showed a much larger fraction of
the total phytoplankton biomass in EGC at the time of sampling
(Table 1).
The values represent the mean and the standard deviation for each parameter and productivity between the EGC and WSC regions.

<table>
<thead>
<tr>
<th>Nutrient consumption (μmol L⁻¹)</th>
<th>EGC</th>
<th>WSC</th>
</tr>
</thead>
<tbody>
<tr>
<td>ΔNO₃</td>
<td>3.2 ± 1.2 (8)</td>
<td>6.4 ± 0.9 (5)</td>
</tr>
<tr>
<td>ΔSiO₂</td>
<td>1 ± 0.9 (5)</td>
<td>2.4 ± 0.9 (5)</td>
</tr>
<tr>
<td>ΔPO₄</td>
<td>0.07 ± 0.03 (4)</td>
<td>0.3 ± 0.2 (5)</td>
</tr>
</tbody>
</table>

**Phytoplankton community**

<table>
<thead>
<tr>
<th>Parameter</th>
<th>EGC</th>
<th>WSC</th>
</tr>
</thead>
<tbody>
<tr>
<td>Integrated chl. a conc. [μg g⁻¹</td>
<td>8.1 ± 4.8 (6)</td>
<td>43.3 ± 26.4 (5)</td>
</tr>
<tr>
<td>Estimated productivity since winter [g C m⁻² yr⁻¹]</td>
<td>250 ± 98 (6)</td>
<td>509 ± 73 (5)</td>
</tr>
<tr>
<td>Estimated diatom contribution to productivity [%]</td>
<td>32</td>
<td>37</td>
</tr>
<tr>
<td>Phytoplankton carbon estimate in chl a max.* [μg g⁻¹</td>
<td>0.76 (4)</td>
<td>5.6 (3)</td>
</tr>
<tr>
<td>Diatom carbon estimate in chl a max.* [μg g⁻¹</td>
<td>0.28 (4)</td>
<td>0.49 (3)</td>
</tr>
<tr>
<td>Estimated diatom contribution to phytoplankton biomass in chl a max.* [%]</td>
<td>36</td>
<td>8</td>
</tr>
</tbody>
</table>

The values represent the mean and the standard deviation for each parameter and the number in parentheses represents the number of stations. Negative values in nutrient consumption were excluded from the mean calculation. *Phytoplankton carbon measurements were calculated from microscopy counts of the different phytoplankton groups and previously published in Engels et al. (2017).

To verify that these differences in biogeochemical parameters represent different ecosystem states, we estimate surface chl a dynamics of the biogeochemical model FESOM-REcoM2, set to the studied dates, to estimate the phytoplankton bloom stages in the two regions. Because of the lack of chl a remote sensing measurements for the ice-covered regions, we could only use the ice-free region for calibration (Supplementary Figure 9). In the model, a strong relationship between the estimates of chl a and the shifting sea ice edge was observed (Figure 2). In the beginning of June, surface chl a concentrations were elevated in the whole ice-free area of WSC, while remaining very low in the ice-covered EGC (Figures 2A–C). In the second half of June 2014, with the ice thinning and the sea ice edge shifting westward, an increase in surface chl a concentrations was observed also in EGC (Figures 2D–F).

**Differences in Microbial Community Composition Between the Eastern and Western Regions of the Fram Strait**

Using Illumina 16S rRNA amplicon sequencing of the V3-V4 hypervariable region, we obtained a final dataset of 2,462,994 reads (amplicons) in 63 samples, which were assigned to 7,167 OTUs associated with 406 bacterial taxonomic lineages. The OTUs which were taxonomically assigned to chloroplasts or mitochondria were excluded from further analysis. The rarefaction curves did not reach a plateau in any of the samples, and on average the samples covered 60% of the bacterial community richness (Supplementary Table 1 and Supplementary Figure 2A). However, coverage-based rarefaction estimations (i.e., Good’s estimator), revealed a sample completeness higher than 98% in all samples (Supplementary Figure 2B; Chao and Jost, 2012; Chao et al., 2014). This suggests that although additional OTUs could be expected with additional sequencing, our sequencing depth was satisfactory to represent most of the diversity within the bacterial communities.

Comparison of bacterial community composition between the different regions and fractions was conducted based on the presence/absence of an OTU (Figure 3). A total of 974 OTUs (13% of the total OTUs) were shared throughout the entire dataset, and represented more than 75% of all sequences. Especially the FL communities of both regions were similar (Figure 4). Hence, differences between the bacterial communities mainly resulted from variations in the proportional abundance of these taxa.

In order to further investigate the differences in community composition between the different regions, we performed differential abundance tests for all shared OTUs from both the FL and PA fractions using ‘DESeq2’. The OTU which had a fold change of absolute value higher than 1 and an adjusted p-value < 0.05 was defined as ‘differentially abundant OTU’ – daOTU. Furthermore, using ‘GAGE’ we tested for the enrichment of bacterial groups at a lower taxonomic resolution, i.e., that of bacterial families. Only bacterial families in which all OTUs were enriched in only one region and showed statistical significance (adjusted p-value < 0.05), were considered to be enriched.

A total of 757 (10% of all OTUs) and 869 (12% of all OTUs) daOTU were identified in the FL and PA fractions, respectively (Supplementary Figure 3). For both fractions, the EGC region was represented by a higher proportion of daOTU compared to the WSC (60 and 65% for FL and PA, respectively), as well as by a higher number of sequence-enriched bacterial families (Figure 5). The WSC was characterized, in both fractions, by few significantly enriched families in various taxonomic groups, such as Alphaproteobacteria (Rhodobacteraceae) and Gammaproteobacteria (Piscirickettsiaceae, Portiococaceae). Furthermore, Flavobacteria (Cryomorphaceae) and Gammaproteobacteria (OM182 clade) were significantly enriched in the FL fraction of WSC. Enriched taxa in the EGC were distributed across a broader taxonomic range, with large differences also between the fractions. In the FL fraction the significantly enriched families were associated with the poorly classified Chloroflexi (SAR202), Marinimicrobiia (SAR406) and Deltoproteobacteria (SAR324, Bdellovibrionaceae), as well as members of Alphaproteobacteria (SAR11, Rhodospirillaceae) and Gammaproteobacteria (Colwelliaceae, Pseudoalteromonadaceae and JTB255). In the PA fraction significantly enriched families were associated mainly with Deltoproteobacteria (Bdellovibrionaceae, Bradydromadaceae, Oligoflexaceae, NB1–j) and Gammaproteobacteria (Pseudoalteromonadaceae, Shewanellaceae and JTB255).

A similar workflow was applied to investigate microbial eukaryotic communities. Using Illumina 18S rRNA amplicon sequencing of the V4 hypervariable region, we obtained a final dataset of 2,396,433 reads (amplicons) in 33 samples,
which were assigned to 4,419 OTUs associated with 173 eukaryotic taxonomic lineages. The eukaryotic OTUs which were taxonomically assigned to metazoa were excluded from further analysis. Rarefaction curves did not reach a plateau in any of the samples, and on average the samples covered 75% of the eukaryotic community richness (Supplementary Table 1 and Supplementary Figure 2C). Nevertheless, coverage-based rarefaction estimations (i.e., Good’s estimator), revealed a sample completeness higher than 98% in all samples (Supplementary Figure 2D; Chao and Jost, 2012; Chao et al., 2014). This suggests that although additional OTUs could be expected with additional sequencing, our sequencing depth was satisfactory to represent most of the diversity within the eukaryotic communities.

A corresponding OTU presence/absence analysis between eukaryotic communities in each region revealed that 2,502 OTUs (56% of the total OTUs) were shared between the regions (Supplementary Figure 4), comprising more than 80% of the sequences in all eukaryotic samples (Figure 6). Hence, the relatively high proportion of region-specific OTUs showed very low relative sequence abundances. Furthermore, the taxonomic groups Syndiniales, Dinophyceae (dinoflagellates) and Diatomea showed larger number of daOTU in EGC (Supplementary Figure 5). In the WSC on the other hand, the largest taxonomic group (in terms of number of daOTU) was the heterotrophic Thecofilosea (Cercozoa).

Environmental Drivers of Microbial Communities in the Fram Strait

Bacterial cell densities and production estimates based on leucine incorporation showed statistically significant differences between the two regions (t-test, p < 0.001; Figures 7A,B and Supplementary Table 2). The results showed almost one order of magnitude higher bacterial cell densities in WSC compared to EGC, as well as higher ratios between high nucleic acid (HNA) and low nucleic acid (LNA) cells. Total bacterial productivity was higher in the WSC compared to the EGC region, while cell specific productivity (total productivity divided by cell concentration) did not show significant difference between the regions. Moreover, a principal coordinate analysis (PCoA) of bacterial community composition revealed significant differences between samples according to their geographic origin, in addition to clear differences in the community structure of FL and PA fractions (Figures 7C,D). Samples from different depths showed no clear clustering. The separation of samples according to their bacterial community structure was confirmed using a permutational multivariate analysis of variance. Similar
differences between the regions were observed for the microbial eukaryotic community, with higher phytoplankton estimated biomass in the WSC (Table 1), and community composition clustering according to regions, although to a lesser extent than bacterial communities (Supplementary Figures 6A,B).

To compare the explanatory power of a range of environmental variables in structuring bacterial communities, we performed redundancy analysis (RDA) and constrained the ordination by the following environmental parameters: temperature, salinity, chl a, and consumed nutrients (ΔNO₃, ΔSiO₃, and ΔPO₄). Due to the different environmental conditions in EGC and WSC regions, we selected these parameters to account for the combined effect of the different water masses (temperature and salinity) and different ecosystem states (chl a and nutrients). The analysis was performed separately for FL and PA bacterial communities, as the fractions may be influenced by different environmental factors (Figures 8A,B). In accordance with the PCoA ordination (Figure 7C), both FL and PA fractions exhibited a strong separation of bacterial communities between EGC and WSC (mainly along RDA axis 1, which explained roughly 80% of the variance). Using a stepwise model selection test (‘ordistep’ algorithm in ‘vegan’ package), we identified that temperature, salinity and chl a were the strongest explanatory variables in the FL fraction, explaining 66% of the total variance. Community variation in the PA fraction was mainly explained by temperature, salinity, chl a and consumed nitrate (ΔNO₃), which explained 63% of the total variance. A similar stepwise model selection test for the microbial eukaryotic community revealed that community variation was mainly explained by temperature, salinity, consumed silicate (ΔSiO₃) and nitrate (ΔNO₃), adding up to 38% of the total explained variance (Supplementary Figure 6C).

**Associations Between Bacteria and Eukaryotic Microbes – Based on Co-occurrence Networks**

Two separate co-occurrence networks were constructed to examine potential associations between free-living bacteria and microbial eukaryotes ('FL network') and between particle-associated bacteria and microbial eukaryotes ('PA network') at the
chl a max. depth. In the FL network 85% of potential associations were positive, in the sense that sequence-richer taxa of bacteria were associated with sequence-richer taxa of eukaryotes. The PA network consisted of a larger number of total potential associations, but only 71% of them were positive (Supplementary Table 3). An overview of both positive and negative associations (Figure 9) revealed two taxonomic groups that showed highest numbers of associations in both fractions together, the eukaryotic order Syndiniales (Alveolata) and the bacterial order Flavobacteriales (Flavobacteriia). In addition, high number of potential associations was associated with Gammaproteobacteria, such as Alteromonadales, and Oceanospirillales (Figures 9A,C). Among the microbial eukaryotes, two groups showed relatively high numbers of associations: Diatomea and Dinophyceae (Dinoflagellata; Figures 9B,D).

In order to identify regionally specific associations of microbial eukaryotes with bacterial taxa, we generated for each fraction a sub-network of positive associations between eukaryotic OTUs and previously identified bacterial daOTU for the EGC and WSC, respectively (Supplementary Figure 3).
FIGURE 5 | Enriched bacterial families between the regions. Taxonomic enrichment analysis was performed separately on the FL (A) and the PA (B) fractions, and only statistically significant taxa were included in data representation (adjusted p-value < 0.05). The x-axis represents the log2 fold change in sequence abundance. Enrichment in the EGC region is represented in the blue area while enrichment in the WSC region is represented in the red area. The color code represents taxonomic classes and each point represents the log2 fold change of each taxonomic family. The number associated with each symbol represents the number of OTUs in the family.
The sub-network topologies showed different patterns in the FL and the PA networks. Overall, the FL network consisted of 159 nodes of daOTU, out of a total 363 bacterial OTUs in the network (81 daOTU in EGC and 78 in WSC). In the PA network there were 226 nodes of daOTU, out of a total 363 bacterial OTUs in the network (197 daOTU in EGC and 30 daOTU in WSC). Subsequently, the sub-networks were clustered into metanodes, each incorporating OTUs of a specific taxonomic group (Figure 10). The clustered sub-networks of both fractions revealed strong differences between the regions, with larger number of taxon-taxon associations in the EGC. The strongest associations, based on the number of connecting edges, in all sub-networks, were related to co-occurrences of Syndiniales (Alveolata) with various bacterial orders such as Flavobacteriales and Oceanospirillales.

DISCUSSION

Pelagic Ecosystem State – in situ and in silico Observations

In our study we investigated the summer dynamics of pelagic bacterial communities from the photic zone of Fram Strait (top 60 m). Using measurements of physical and biogeochemical...
parameters, combined with sea ice coverage, we separated the Strait into two main pelagic ecosystem regions (Figure 1). These different regions were directly related to the distinct current systems in the Strait; one transporting Atlantic Water to the Arctic Ocean (WSC) and the other one exporting Polar Water and sea ice (EGC; Beszczynska-Möller et al., 2011). These distinct current systems differed not only in physical characteristics of the water (temperature and salinity) but also in their nutrient concentrations (Table 1 and Supplementary Figure 1). The different geochemical and sea ice conditions potentially affect biological processes in these distinct regions (e.g., nutrient and light limitation of the phytoplankton bloom). We thus used a combination of measured and modeled biogeochemical variables to further investigate the ecosystem states in the two regions.

The high phytoplankton biomass and production estimates (Table 1), as well as elevated bacterial cell densities in the WSC compared to the EGC (Figures 7A,B), are likely related to the decaying phytoplankton bloom (Pinhassi and Hagström, 2000; Riemann et al., 2000; Alonso-Sáez et al., 2008; Buchan et al., 2014). Further evidence for such a relationship has been detected by a previous study in Fram Strait, which showed correlations of bacterial activity with concentrations of amino acids and carbohydrates in the water (Piontek et al., 2014). In the WSC region maximum integrated chl a values during seasonal blooms reach up to 100 mg/m$^3$ (Nöthig et al., 2015). Thus based on the chl a concentrations, the fully depleted nutrients above the pycnocline and the low pCO$_2$ (Table 1 and Supplementary Figure 7), we conclude that we had sampled a post-phytoplankton bloom situation. In the EGC, the low nutrient depletion in surface waters, the low chl a concentration and the high pCO$_2$ rather suggest a pre-phytoplankton bloom stage. Moreover, the stoichiometry-based estimate of new production in both regions was in a comparable range to previous estimates of Nöthig et al. (2015) in Fram Strait as well as to estimates in other regions of the Arctic Ocean (Arrigo et al., 2008; Wassmann et al., 2010; Boetius et al., 2013). The generally high ratio between NO$_3$ and PO$_4$ concentrations in the EGC indicate a Pacific origin of the sampled Polar Water (Wilson and
Wallace, 1990), and PO₄ may be one of the limiting factors for the development of a phytoplankton bloom in this region, at the time of the sampling (Taylor et al., 1992).

In order to test whether the biogeochemical differences between the sampled regimes represent different ecosystem states, or simply represent hydrographical differences between Polar Water and Atlantic Water, we used surface chlorophyll a dynamics obtained from the coupled FESOM-REcoM2 model (Figure 2 and Supplementary Material; Schourup-Kristensen et al., 2014). In June 2014, when the sea ice cover, hydrographical and nutrient conditions fit well with observations (Supplementary Figures 8, 9), the annual dynamics produced by the model showed an increase in surface chl a concentration in EGC in the second half of June, associated with the seasonal thinning of the sea ice in the region (Leu et al., 2011; Nöthig et al., 2015). Moreover, in the WSC the model showed a decline in surface chl a concentration throughout the month. In summary, our observations and the model results support the hypothesis that during the time of sampling early phytoplankton bloom conditions prevailed in the ice-covered EGC (first half of June), and that the phytoplankton bloom of the ice-free WSC was already in decline (second half of June).

**Functional and Regional Differences in Microbial Communities Across the Fram Strait**

Both WSC and EGC regions exhibited a large number of OTUs, which were unique to one of the regions (Figure 3). However, these OTUs represented only a small proportion of the total sequence abundance of the bacterial community, and consisted of taxa, which were previously identified as rare bacterial community members in the Arctic Ocean (Galand et al., 2009). The vast majority of the sequence proportion was related to...
OTUs which were shared between the regions and fractions (Figure 4). Moreover, bacterial community variations in the FL and PA fractions were explained by the same environmental parameters, suggesting that both fractions are subject to similar environmental drivers (Hanson et al., 2012). Hence, we hypothesized that community variation was mostly driven by environmental factors such as bloom stage, selecting for different sequence proportions of shared OTUs. It is important to note that size-fractionated filtration may lead to different observations compared to bulk filtration (Padilla et al., 2015). In this study we did not observe a clogging of filters, but cannot exclude effects on FL and PA fractions.

In order to investigate differences in the relative contributions of the shared OTUs to the communities in WSC and EGC, we identified differentially abundant OTUs (daOTU) in both the FL and PA fractions (Supplementary Figure 3). Flavobacteria and Gammaproteobacteria were the two main heterotrophic bacterial taxa which showed high numbers of daOTU and numerous enriched taxa in both fractions (Figure 5 and Supplementary Figure 3). For both fractions combined, the WSC consisted of almost twice the number of flavobacterial daOTU compared to EGC (176 and 107 daOTU, respectively), suggesting an enrichment of this taxonomic group by post-bloom conditions in this region. Flavobacteria specialize on targeting complex organic biopolymers and were previously described to respond to phytoplankton blooms in high latitudes (Simon et al., 1999; Teeling et al., 2012; Williams et al., 2013; Chafee et al., 2018). Moreover, Cryomorphaceae, a significantly enriched flavobacterial family in the FL fraction of WSC (Figure 5), was previously identified as one of the main taxa responding to
FIGURE 10 | Sub-networks of the FL and PA fractions in the chl. a max. of EGC and WSC regions. Taxonomic groups were clustered into metanodes. Colors of the nodes represent the different taxonomic domains, while the size is proportional to the number of OTUs grouped together in the metanode. The edges represent taxon-taxon positive (co-occurrence) associations, and the width of the edges represents the number of associations between the metanodes.

Additionally, in both fractions, there was a large number of daOTU and several significantly enriched families related to Gammaproteobacteria (Figure 5 and Supplementary Figure 3). These opportunistic copiotrophs, which have previously been described from both FL and PA fractions, are highly diverse and specialized in adapting to a wide range of carbon sources, also responding to different stages of phytoplankton blooms (Alonso-Sáez et al., 2008; Teeling et al., 2012; Nikrad et al., 2014). Interestingly, the genus Balneatrix (Oceanospirillales) which was previously identified to strongly correlate with phytoplankton bloom presence in the North Sea (Wemheuer et al., 2014), accounted for 30 daOTU in the WSC and only 5 daOTU in the EGC, which may be linked to the different phytoplankton bloom conditions in the region. Furthermore, the order Pseudoalteromonadales which consisted of several significantly enriched families in both fractions in EGC (Figure 5), contains several psychrophilic genera which were previously found in sea ice (Bowman et al., 1997; Brown, 2001; Brinkmeyer et al., 2003; Eric Collins et al., 2010; Yu et al., 2015), and their enrichment in the EGC may thus be partly a result of their release from ice-associated communities. An interesting observation was provided by two outlier samples. Although they originated from the WSC, the proximity of station 1W and HG9 to the sea ice edge (Figure 1), potentially resulted in bacterial communities more similar to stations from the EGC (Figure 7C). This may indicate that the effect of the seasonal phytoplankton bloom extends into the zone where both water masses mix, e.g., by eddies (Wekerle et al., 2017). Several cryptic taxonomic groups, such as Chloroflexi (SAR202), Marinimicrobia (SAR406) and various members of Deltaproteobacteria, were significantly enriched in EGC (Figure 5), and also consisted of a large number of unique OTUs.
in this region. These enriched taxonomic groups in the ice-covered EGC were previously reported from surface waters in the western Svalbard region (WSC) during the Arctic winter (Wilson et al., 2017). Therefore, our results support and strengthen the hypothesis of Wilson et al. (2017) that bacterial community dynamics in Fram Strait are to a large extent affected by seasonal variability (e.g., availability of light under changing sea ice conditions), rather than hydrographic differences between water masses.

Enriched eukaryotic taxa differed strongly between the EGC and WSC regions (Supplementary Figure 5), with the taxonomic groups being consistent with previously reported seasonal dynamics in the Arctic Ocean (Lovejoy, 2014). In the EGC region all enriched taxa were related to previously identified, dominant members of pelagic Arctic winter communities (e.g., Syndiniales; Guillou et al., 2008; Jepcott et al., 2016; Marquardt et al., 2016). Two different taxonomic groups of phytoplankton were enriched in the WSC: the class of green algae Prasinophyceae abundant photosynthetic organisms in late summer-autumn seasons in the Arctic (Lovejoy et al., 2007; Vader et al., 2015; Marquardt et al., 2016; Metfies et al., 2016; Joli et al., 2017). Furthermore, several heterotrophic eukaryotic taxa (e.g., Thecofilosea) were enriched in the WSC. These organisms are mainly grazers and depend on the presence of phytoplankton and bacteria (Monier et al., 2013); their higher representation may thus be linked to the declining phytoplankton bloom in the WSC. Microbial eukaryotic community composition clearly differed between the two regions (Supplementary Figure 6C). Interestingly, stations 10 and 8.5 W showed some similarity to the WSC region, which may be related to a coastal phytoplankton bloom east of Greenland (Supplementary Figure 8). However, overall our observations of the microbial eukaryotic community further support our classification of early bloom conditions in the EGC and late bloom conditions in WSC.

**Co-occurrence Networks Reveal Potential Candidates for Cross-Domain Interactions**

Numerous studies have described shifts in bacterial community composition during phytoplankton blooms (Teeling et al., 2012; Wemheuer et al., 2014; Chafee et al., 2018), but very little is known about specific biotic interactions between bacteria and phytoplankton during blooms (Töpper et al., 2010; Amin et al., 2012; Hartmann et al., 2013; Lima-Mendez et al., 2015). Our results revealed an enrichment of specific bacterial taxa in the different regions, which we suggest to be related to the seasonal development of the phytoplankton bloom. Using network co-occurrence analyses (Faust and Raes, 2012), we therefore tested whether these enriched taxa exhibit potential associations with eukaryotic microbes in the chl a max. communities.

Both FL and PA networks consisted of a large number of edges (Figure 9 and Supplementary Table 3), which may indicate potential ecological interactions between taxa. Among the bacterial taxa in the FL network, a large number of associations was related to the typically free-living SAR11 clade (Giovannoni, 2017). In the PA network, on the other hand, large number of associations were related to typical particle-associated Gammaproteobacteria, such as Alteromonadales (Crespo et al., 2013; Fontanez et al., 2015). In both fractions, Flavobacteria and Syndiniales outnumbered all other taxonomic orders in terms of the number of associations. These observations are in line with a previous report from the global plankton interactome study conducted as part of the global Tara Oceans expedition (Lima-Mendez et al., 2015), which did, however, not cover the Arctic Ocean.

Roughly 30–40% of bacterial nodes in the networks consisted of daOTU associated with one or more eukaryotic taxa. Interestingly, “regional” (WSC vs. EGC) sub-networks displayed strong differences between both regimes in the PA fraction, with a much higher number of associations in the EGC (Figure 10). Little is known about the lifestyle and physiology of many of the organisms identified in the networks, especially for the bacterial fraction, and the translation of observed associations into biological traits is thus extremely limited (Ramanan et al., 2016). Furthermore, in many cases the association may represent a common response of taxonomic groups to environmental conditions, rather than direct interaction between them (Weiss et al., 2016). Nevertheless, the observed cross-domain associations showed clear differences between the regions with different phytoplankton bloom conditions, resulting in the enrichment of specific bacterial taxa and the development of distinct ecological networks. It has been previously proposed that shifts in the timing and composition of phytoplankton blooms, as well as temporal mismatches with grazers resulting in an altered food web, are among the main impacts of climate change in the Arctic (Soltwedel et al., 2016; Engel et al., 2017). Our observations of specific associations between eukaryotes and bacteria in the plankton suggest that such ecological shifts may be accompanied by substantial changes in the microbial community structure.

**CONCLUSION**

Our study revealed strong differences in pelagic microbial community activity and structure in the photic layers of the ice-free eastern (WSC) and ice-covered western (EGC) Fram Strait during summer 2014. Measured and modeled biogeochemical parameters suggested distinct ecosystem states in the two regions, namely different stages of the summer phytoplankton bloom, as a result of differences in sea ice cover and irradiance. Although it is challenging to conclusively decouple effects of water masses, seasonally driven biogeochemistry and biotic associations, our study shows that differences in bacterial communities between the regions could be explained by environmental parameters associated with phytoplankton bloom dynamics. This includes a strong increase in bacterial cell densities and activity in response to a declining phytoplankton bloom in the WSC, with an enrichment of phytoplankton bloom associated bacterial taxa commonly known to degrade phytoplankton products, such as Flavobacteria. In contrast, the EGC region showed high relative sequence proportions of bacterial taxa that have been associated with Arctic winter conditions (e.g., SAR202 clade, Marinimicrobia and Deltaproteobacteria).
Moreover, co-occurrence networks provided evidence for a high variety of potential interactions between bacteria and microbial eukaryotes in the early bloom conditions, and their possible specialization with the advancement of the phytoplankton bloom. In times of a rapidly changing Arctic Ocean, our results highlight the potential impact of future ice-free summers on the structure and function of Arctic Ocean pelagic microbial communities. Additional sampling throughout the year will help to better resolve seasonally driven microbial community dynamics and contrast them to long-term shifts.

**MATERIALS AND METHODS**

**Field Sampling**

Samples were collected in Fram Strait during the Polarstern expedition PS85 (June 6th – July 3rd 2014) from the eastern Greenland shelf to the west coast of Spitsbergen (Supplementary Table 1 and Figure 1). Sampling was carried out with 12 L Niskin bottles mounted on a CTD rosette (Sea-Bird Electronics Inc. SBE 911 plus probe) equipped with double temperature and conductivity sensors, a pressure sensor, altimeter, chlorophyll fluorometer, and transmissometer. The chlorophyll maximum depth (chl a max) was determined based on chl a fluorescence during the downcast, while the water samples were collected during the upcast. Along the transect samples were collected from surface water (5–10 m), the chl a max (10–30 m) and below the chl a max (30–60 m, Supplementary Table 1). Hydrographic data of the seawater including temperature and salinity were retrieved from the PANGAEA database (Rabe et al., 2014). Water masses were identified based on their hydrographic characteristics, according to Rudels et al. (2013).

**Sampling for Bacterial Communities**

For assessing bacterial community composition, 2 L of water were filtered through successive membrane filters of 3 μm (Whatman Nucleopore, 47 mm polycarbonate), and 0.22 μm (Millipore Sterivex filters) using a peristaltic pump (Masterflex; Cole Parmer). All samples were stored at −20°C until DNA isolation.

**Sampling for Eukaryotic Microbial Communities**

For assessing eukaryotic community composition, 2 L subsamples were taken in PVC bottles from the Niskin water samplers. Eukaryotic microbial cells were collected by sequential filtration using a Millipore Sterifil filtration system (Millipore, United States). Each water sample was filtered through three different mesh sizes (10, 3, and 0.4 μm) on 45 mm diameter Isopore Membrane Filters at 200 mbar. All samples were stored at −20°C until DNA isolation.

**DNA Isolation and Amplicon Sequencing Bacteria**

Genomic bacterial DNA was isolated from the 3 μm and the 0.22 μm filter membranes to analyze the particle-associated (PA) and the free-living (FL) community, respectively, in a combined chemical and mechanical procedure using the PowerWater DNA Isolation Kit (MO BIO Laboratories, Inc., Carlsbad, CA, United States). Prior to DNA isolation the sterivex cartridges of the 0.22 μm membranes were cracked open in order to place the filters in the kit-supplied bead beating tubes. The isolation was continued according to the manufacturer’s instructions, and DNA was stored at −20°C. Library preparation was performed according to the standard instructions of the 16S Metagenomic Sequencing Library Preparation protocol (Illumina, Inc., San Diego, CA, United States). The hypervariable V3–V4 region of the bacterial 16S rRNA gene was amplified using bacterial primers S-D-Bact-0341-b-S-17 (5′-CCTACGGGNGGCWGCAG-3′) and S-D-Bact-0785-a-A-21 (5′-GACTACHVGGGTATCTAATCC-3′; Klindworth et al., 2013). Sequences were obtained on the Illumina MiSeq platform in a 2 × 300 bp paired-end run (CebiTeC Bielefeld, Germany), following the standard instructions of the 16S Metagenomic Sequencing Library Preparation protocol (Illumina, Inc., San Diego, CA, United States).

**Eukaryotic Microbes**

Genomic eukaryotic DNA was isolated from the 10, 3, and 0.4 μm filter membranes using the NucleoSpin Plant Kit (Machery-Nagel, Germany), following the manufacturer protocol. The resulting DNA-extracts were stored at −20°C. DNA concentrations were determined using the Quantus Fluorometer (Promega, United States) according to the manufacturer’s protocol, and equal volumes of the isolated genomic DNA from the three different filter fractions were pulled together. Library preparation was performed according to the standard instructions of the 16S Metagenomic Sequencing Library Preparation protocol (Illumina, Inc., San Diego, CA, United States). The hypervariable V4 region of the eukaryotic 18S rRNA gene was amplified using 528F (5′-GCGTAAATCCCAGCTTCAA-3′) and 964iR (5′- ACTTTGATTGTTGAGT-3′) primers. All PCRs had a final volume of 25 μL and contained 12.5 μL of KAPA HIFI Mix (Kapa Biosystems, Roche, Germany), 5 μL of each primer 1 μmol L−1 and 2.5 μL DNA-template ~5 ng. The DNA-template was a mix of equal volumes of genomic DNA isolated from the three different filter fractions, i.e., 10, 3, and 0.4 μm. PCR amplification was performed in a thermal cycler (Eppendorf, Germany) with an initial denaturation (95°C, 3 min) followed by 25 cycles of denaturation (95°C, 30 s), annealing (55°C, 30 s), and extension (72°C, 30 s) with a single final extension (72°C, 5 min). The PCR products were purified from an agarose gel 1% w/v with the AMPure XP PCR purification kit (Beckman Coulter, Ing., United States) according to the manufacturer’s protocol. Subsequent to purification DNA concentrations in the samples were determined using the Quantus Fluorometer (Promega, United States). Subsequently, indices and sequencing adapters of the Nextera XT Index Kit (Illumina, United States) were attached in the course of the Index PCR. All PCRs had a final volume of 50 μL and contained 25 μL of KAPA HIFI Mix (Kapa Biosystems, Roche, Germany), 5 μL of each Nextera XT Index Primer 1 μmol L−1, 5 μL DNA-template ~5 ng and 10 μL PCR grade water. PCR amplification was performed in a thermal cycler (Eppendorf,
Germany) with an initial denaturation (95°C, 3 min) followed by 8 cycles of denaturation (95°C, 30 s), annealing (55°C, 30 s), and extension (72°C, 30 s) with a single final extension (72°C, 5 min). Prior to quantification of the amplification products with the Quanta Fluorometer (Promega, United States) for sequencing the final library was cleaned up using the AMPure XP PCR purification kit (Beckman Coulter, Ing., United States). Sequences were obtained on the Illumina MiSeq platform in a 2 × 300 bp paired-end run (AWI Bremerhaven, Germany), following the standard instructions of the 16S Metagenomic Sequencing Library Preparation protocol (Illumina, Inc., San Diego, CA, United States).

Bioinformatics and Statistical Analyses
Both bacterial and eukaryotic libraries were subject to similar bioinformatic pipelines. The raw paired-end reads were primer-trimmed using ‘cutadapt’ (Martin, 2011), quality trimmed using ‘trimmomatic’ with a sliding window of four bases and a minimum average quality of 15 (v0.32; Bolger et al., 2014). The reads were merged using PEAR (v0.9.5; Zhang et al., 2014), and all merged reads below 350 bp or above 450 bp were removed from the dataset. Clustering into OTUs was done with the ‘swarm’ algorithm using default parameters (v2.0; Mahé et al., 2015). Chimeric sequences were identified and removed using ‘uchime’ function in VSEARCH (v1.9.7; Rognes et al., 2015). Chimeric sequences were identified and removed using ‘uchime’ function in VSEARCH (v1.9.7; Rognes et al., 2016). One representative sequence per OTU was taxonomically classified using ‘SINA’ (SILVA Incremental Aligner; v1.2.11; Silva reference database release 128; Quast et al., 2013) at a minimum alignment similarity of 0.9, and a last common ancestor consensus of 0.7 (Fruesse et al., 2012). The OTUs which were not taxonomically assigned to Bacteria/Eukarya or occurred with only a single sequence in the whole dataset (‘singleton’). The OTUs which were not taxonomically assigned to Bacteria/Eukarya or occurred with only a single sequence in the whole dataset (‘singleton’). The OTUs which were not taxonomically assigned to Bacteria/Eukarya or occurred with only a single sequence in the whole dataset (‘singleton’). The OTUs which were not taxonomically assigned to Bacteria/Eukarya or occurred with only a single sequence in the whole dataset (‘singleton’).

All statistical analyses were conducted using R (v3.4.1)\(^1\) in RStudio (v1.0.153; RStudio Team, 2015). Sample data matrices were managed using the R package ‘phyloseq’ (v1.20.0; McMurdie and Holmes, 2013) and plots were generated using the R package ‘ggplot2’ (v2.2.1; Gómez-Rubio, 2017). A prevalence threshold (i.e., in how many samples did a taxon appear at least once) of 5% was applied to the OTU table prior to downstream analysis following (Callahan et al., 2016). All alpha diversity parameters and curves were obtained using R package ‘iNEXT’ (v2.0.12; Hsieh et al., 2018). The rarefaction curves for each sample were generated based on 40 equally spaced rarefied sample sizes with 100 iterations.

Principal coordinate analysis was conducted on variance stabilized OTU abundance matrices (McMurdie and Holmes, 2014). The significance of the clustering was tested using the ‘ADONIS’ function in the R package ‘vegan’ (v2.4-5; Oksanen, 2017). The fold-change in abundance of each OTU between the regions was calculated using the R package ‘DEseq2’ (v1.16.1; Love et al., 2014). The method applies a generalized exact binomial test on variance stabilized OTU abundance. The taxonomic enrichment test was performed using the generally applicable gene-set enrichment method in the R package ‘GAGE’ (v2.26.3; Luo et al., 2009). The results were filtered by significance, after correction for multiple-testing according to Benjamini and Hochberg (1995) with an adjusted p-value < 0.05. The shared OTUs calculations and visualization were conducted using R packages ‘UpSetR’ (v1.3.3; Conway et al., 2017) and ‘VennDiagram’ (v1.6.18; Chen and Boutros, 2011).

Co-occurrence Network Analysis
The network analysis was conducted separately using the chl a max. FL and PA bacterial communities. The cross-domain co-occurrence networks between bacteria and eukaryotes were constructed using CoNet (v1.1.1beta; Faust and Raes, 2016), as described in Lima-Mendez et al. (2015). The measure-specific p-values were merged using Brown’s method (Brown, 1975) and correction for multiple-testing was performed according to Benjamini and Hochberg (1995). Edges with an adjusted p-value above 0.05 were discarded. The constructed networks were further analyzed and visualized using the R package ‘igraph’ (v1.1.2; Csardi and Nepusz, 2006).

Calculation of Consumed Inorganic Nutrients
The raw nutrient concentration measurements were retrieved from PANGAEA (Graeve and Ludwigowski, 2017). The nutrient consumption (Δ) at each station was calculated by subtracting the mean value of all collected measurements above 50 m from the mean value of all collected measurements between 50 and 100 m (below the seasonal pycnocline). The integrated chlorophyll a and inorganic nutrient values were calculated according to (Boss and Behrenfeld, 2010). The productivity estimates were calculated using the Redfield ratio 106 C: 16 N: 1 P, and for diatom contribution the ratio of 1:1 N:Si was assumed (see Supplementary Material).

Bacterial Abundance and Productivity
Bacterial abundance was determined by flow cytometry (FACSCalibur, Becton Dickinson). Samples were fixed with glutaraldehyde at 1% final concentration and stored at −20°C. Prior to analysis, samples were stained with the fluorescent dye SybrGreen I (Invitrogen) that binds to DNA. Bacterial cell numbers were estimated after visual inspection and manual gating of the bacterial population in the cytoparameter of side scatter vs. green fluorescence. Fluorescent latex beads (Polyscience, Becton Dickinson) were used to normalize the counted events to volume (Gasol and Del Giorgio, 2000).

The incorporation of 3H-leucine (specific activity 100 Ci mmol−1) was determined to estimate bacterial production (BP).
The radiotracer was added at a saturating final concentration of 20 nmol L$^{-1}$ before three replicate samples were incubated for 4–6 h in the dark close to in situ temperature at 0–2°C. Incubations were stopped by the addition of trichloroacetic acid (TCA) at a final concentration of 5%. Samples were then processed by the centrifugation method according to Smith and Azam (1992). Briefly, samples were centrifuged at 14,000 $\times$ g to obtain a cell pellet that was washed twice with 5% TCA. Incorporation into the TCA-insoluble fraction was measured by liquid scintillation counting after resuspension of the cell pellet in scintillation cocktail (Ultima Gold AB, Perkin Elmer).

**Chlorophyll a Measurements**

The concentration of chl a was determined from 0.5 to 2 L of seawater filtered onto glass fiber filters (Whatman GF/F) under low vacuum pressure (<200 mbar); the filters were stored at $-20^\circ$C before analysis. Pigments were extracted with 10 ml of 90% acetone. The filters were treated with an ultrasonic device in an ice bath for less than a minute, and then further extracted in the refrigerator for 2 h. Subsequently they were centrifuged for 10 min at 5000 rpm at 4°C prior to measurement. The concentration was determined fluorometrically (Turner Designs), together with total phaeophytin concentration after acidification (HCl, 0.1 N) based on methods described in Edler (1979) and Evans (1980), respectively. The standard deviation of replicate test samples was <10%.

**The Biogeochemical Model FESOM-REcoM2**

To estimate biological productivity in areas and time periods that were not covered by sampling, we used the biogeochemical model REcoM2 coupled to the Finite Element sea ice Ocean Model (FESOM; Schourup-Kristensen et al., 2014). The model runs in a global setup and describes the ocean, sea ice and marine biogeochemistry, thus making it possible for us to estimate the phytoplankton bloom development stage in both the western, ice-covered part of Fram Strait and the eastern ice-free part (see Supplementary Information).

**Data Accession Numbers and Analyses Repository**

Data are accessible via the Data Publisher for Earth & Environmental Science PANGAEA (www.pangaea.de): chlorophyll a measurements - doi: 10.1594/PANGAEA.887840; bacterial counts and productivity - doi: 10.1594/PANGAEA.887881. Raw paired-end sequence, primer-trimmed reads were deposited in the European Nucleotide Archive (ENA; Silvester et al., 2018) under an umbrella project number PRJEB28027, or PRJEB26163 for Bacteria and PRJEB26288 for Microbial eukaryotes. The data were archived using the brokerage service of the German Federation for Biological Data (GFBio; Diepenbroek et al., 2014). Scripts for processing data can be accessed at https://github.com/edfadeev/Bact-comm-PS85.

**AUTHOR CONTRIBUTIONS**

EF, CB, IS, and AB designed and conducted the study. IS and KM provided sequence data for the study. AE and JP provided the cell counts and bacterial productivity data. E-MN conducted the biogeochemical measurements. VS-K provided the modeled chl a estimates. EF analyzed the data and wrote the manuscript with guidance from CB, AB, and IS. All authors contributed to the final version of the manuscript.

**FUNDING**

This project has received funding from the European Research Council (ERC) under the European Union’s Seventh Framework Program (FP7/2007-2013) research project ABYSS (Grant Agreement No. 294757) to AB. Additional funding came from the Helmholtz Association, specifically for the FRAM infrastructure, and from the Max Planck Society. This publication is Eprint ID 46861 of the Alfred Wegener Institute Helmholtz Center for Polar and Marine Research, Bremerhaven, Germany.

**ACKNOWLEDGMENTS**

We thank the captain and crew of RV Polarstern expedition PS85, as well as the chief scientist Ingo Schewe. We also thank Halina Tegetmeyer and Theresa Hargesheimer for bacterial sample processing and DNA extraction and sequencing. Laura Hehemann for producing the GIS representation of Fram Strait and Wilken-Jon von Appen for assisting with analysis of water column characteristics. This work was conducted in the framework of the HGF Infrastructure Program FRAM of the Alfred-Wegener-Institute Helmholtz Center for Polar and Marine Research.

**SUPPLEMENTARY MATERIAL**

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fmars.2018.00429/full#supplementary-material


**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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