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# The summer bacterial and archaeal community composition of the northern Barents Sea

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ARTICLE INFO

Keywords: Phytoplankton derived carbon Microbial ecology Arctic microbes Nansen Legacy Bacterial succession Microbial loop

## ABSTRACT

Climate change related alterations in the Arctic have influences on the marine ecosystems, in particular on phytoplankton bloom dynamics. Since phytoplankton blooms are the main provider of carbon sources to the microbial loop, the bacterial and archaeal community are affected by the changes as well. Warmer water and less sea ice can lead to an earlier onset of phytoplankton blooms and consequently also to changes in the bacterial and archaeal community are affected by the changes as well. Warmer water and less sea ice can lead to an earlier onset of phytoplankton blooms and consequently also to changes in the bacterial and archaeal community dynamics throughout Arctic summers. Here, we compared the bacterial and archaeal community composition during three summers (2018, 2019, and 2021) along a transect from the Barents Sea to the Arctic Ocean north of Svalbard. We used 16S rRNA gene sequencing to investigate changes in the communities in time and space. The main results showed that, *Gammaproteobacteria (Nitrincolaceae), Bacteroidia (Polaribacter),* and *Alphaproteobacteria* (SAR11 clade 1a members) dominated the bacterial and archaeal community in the surface waters but varied in abundance patterns between the years. The variations are potentially a result of different phytoplankton bloom stages and consequently differences in the availability of carbon sources. The distinctly different deep water communities were dominated by *Candidatus* Nitrosopumilus, *Marinimicrobia*, and members of the SAR324 clade in all years. The results indicate that changes in phytoplankton bloom dynamics can influence bacterial and archaeal community and thereby marine carbon cycling in surface waters, although direct links to the effects of global warming remain uncertain.

#### 1. Introduction

The Arctic warms four times faster than the global average in response to global warming (Rantanen et al., 2022). One of the transition zones between the Atlantic and the Arctic Ocean is the Barents Sea. Here, warm Atlantic water mixes with cold Artic water over a shallow shelf area. The increased warm water import from the Atlantic, a process termed "Atlantification", has lead to 50% decreased sea-ice coverage of the northern Barents Sea within 10 years from 1998 to 2008 (Årthun et al., 2012; Polyakov et al., 2017). These alterations of the Arctic environment may have profound effects on the highly dynamic and sensitive marine ecosystems (Loeng, 1991; Reigstad et al., 2002; Smedsrud et al., 2013; Wassmann et al., 2011). Seasonal investigations of the biological components of the ecosystem are rare which impedes the prediction of the consequences of global warming. The goal of the

Nansen Legacy project is to investigate the ecosystem of the Barents Sea and provide a baseline for comparative future studies and ecosystem predictions.Table 1

The first indications for major changes in the Arctic have already been shown, as early phytoplankton blooms were detected under the sea ice and the chances for the occurrence of autumn blooms have increased by 70% (Ardyna et al., 2014; Assmy et al., 2017). This is likely correlated with a prolongation of the vegetative season by 10–15 days due to loss of sea ice cover (Arrigo and van Dijken, 2015). The longer vegetative season may influence various levels of the marine food web, such as fish, copepods, phytoplankton, and also the bacterial and archaeal communities (Haug et al., 2017; Li et al., 2009; Mueter et al., 2009). However, the responses of organisms and food webs, particularly the bacterial and archaeal communities, to changes in environmental factors remain poorly known.

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https://doi.org/10.1016/j.pocean.2023.103054

Available online 30 May 2023

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Most of the marine primary production (>50%) takes place during these phytoplankton spring bloom (Sakshaug, 2004). Such spring blooms are the main source of carbon for many trophic levels in marine ecosystems, including the bacterial and archaeal communities. In the North Sea, the phytoplankton spring bloom are followed by successions of mostly heterotrophic bacteria such as Bacteroidetes, Gammaproteobacteria and Roseobacter, based on the predominant carbon sources (Teeling et al., 2016, 2012). Similar bacterial successions have been found in the polar regions, where a marine community dominated by members of the SAR11 clade (Schattenhofer et al., 2009; Thiele et al., 2012), transfered to a community with high relative abundances of Gammaproteobacteria (Oceanospirillaceae, Halomonadaceae and Alteromonadaceae), Bacteroidetes (Polaribacter) and Alphaproteobacteria (Rhodobacteraceae and Roseobacter) as a result of a phytoplankton bloom (Alonso-Sáez et al., 2008; Wietz et al., 2021; Wilson et al., 2017; Zeng et al., 2013). On the contrary, Thaumarchaeota (Nitrosopumilus maritimus) and Chloroflexi (SAR202 clade), among others, were relatively low in summer, but dominant in winter communities (Alonso-Sáez et al., 2012; Grzymski et al., 2012; Müller et al., 2018; Wilson et al., 2017). Since temperature and the quantity and quality of dissolved organic carbon compounds are major drivers of the bacterial and archaeal community, changes of the vegetative season may have significant effects on the diversity and structure of these communities and consequently the microbial food web.

This study describes the pelagic microbial communities during three Nansen Legacy expeditions that took place during the summers of 2018, 2019, and 2021, investigating the hypothesis that differences in phytoplankton bloom status or nutrient concentrations will affect the bacterial and archaeal community.

#### 2. Materials & methods

#### 2.1. Sampling and environmental parameters

Samples were taken during the RV "Kronprins Haakon" cruises 2018707 (from Aug. 6th to 23rd 2018), 2019706 (from Aug. 5th to 27th 2019), and 2021708 (from July 12th to 29th 2021) along a transect in the Barents Sea east of Svalbard (Fig. 1). During the transect in 2018, only the first 5 stations, P1-P5 were sampled, while the transects of 2019 and 2021 included stations P6 and P7. Samples were taken using a Niskin rosette sampler following the Nansen Legacy protocol (The Nansen Legacy, 2021). Water samples were collected from the surface (5-10 m), the chlorophyll maximum, or if no clear maximum was found 20 m, 200 m, and deep (10-15 m above sea bottom; SUP 1). Samples from < 75 m were then defined as surface samples, based on the similarity of the bacterial and archaeal community of surface and deep samples. For all water samples 7 L of seawater were filtered on Sterivex filters (Merck, Darmstadt, Germany) and frozen immediately and kept at -80 °C until DNA extraction. During the cruises temperature, salinity, and the concentrations of phosphate, silicate, nitrate, and nitrite were



**Fig. 1.** Map of the Nansen Legacy transect in the Barents Sea with stations marked in blue and colour codes showing the ice conditions at the time of sampling in 2018 (A), 2019 (B), and 2021 (C). Ice data Credits: The Norwegian Ice Service - MET Norway. Ice data are from 15.08.2018, 15.08.2019, and 20.07.2021. Bathymetry Credits: NOAA National Centers for Environmental Information (NCEI); International Bathymetric Chart of the Arctic Ocean (IBCAO); General Bathymetric Chart of the Oceans (GEBCO).

measured (Chierici, 2021a, 2021b; Jones et al., 2022). In addition, seawater was filtered through GFF filters using vacuum filtration, and chlorophyll *a* was extracted for 12–24 h at 4 °C in the dark from algae collected on the filters using 5 ml methanol. The Chl a concentration was measured using a Turner design fluorometer. The abundance of phytoplankton cells was acquired on all cruises (Assmy et al., 2022a, 2022b, 2022c; Kohlbach et al., 2023). The bacterial and archaeal cell concentrations were determined using flow cytometry (The Nansen Legacy, 2021). Triplicates of 1.8 ml glutaraldehyde fixed seawater were used for flow cytometry using a FACS Calibur (Becton Dickinson, Oxford, UK) flow cytometer according to Marie and co-workers (Marie et al., 1999). Stained samples were counted at a low flow rate of around 60  $\mu$ L min<sup>-1</sup> and different groups of bacteria and archaea discriminated on a biparametric plot of green florescence (530/30) vs. side scatter (SSC,488/10).

## 2.2. DNA extraction and sequencing

The DNA of the samples was extracted using the DNeasy Power Water Sterivex Kit (QIAGEN, Hilden, Germany) according to the manual. From the extracted DNA sequencing libraries were prepared to target the V4 region of the 16S rRNA genes, using primers 515F - 5'-GTGYCAGCMGCCGCGGTAA-3' and 806R - 5'-GGAC-TACNVGGGTWTCTAAT-3' (Apprill et al., 2015; Parada et al., 2016). The libraries were sequenced using Illumina MiSeq technology with paired-end reads of 2  $\times$  250 bp length at the Integrated Microbiome Resource in Halifax, Canada.

## 2.3. Sequence analyses

The raw sequences are available at the European Nucleotide Archive under the project number PRJEB57292. For the generation of Amplicon Sequence Variants (ASVs), the DADA2 pipeline was used in R (Callahan et al., 2016). Within the pipeline, primers were removed, the quality of the sequences checked, and a static trim with 250 bp and 220 bp was conducted. The resulting reads were dereplicated and used to generate ASVs. Subsequently, a merging step of the complementary reads was performed, chimeras removed, and the taxonomy of the ASVs assigned using a trained Silva database, based on the Silva release SSU Ref NR v138 (Quast et al., 2013). Thereafter, samples with < 10,000 sequences, and ASVs of mitochondria, chloroplasts, eukaryotes, and  $< 1 \times 10^{-5}$  % of relative abundance were removed. Using the remaining reads, an approximate Maximum Likelihood tree was calculated using FastTree2 (Price et al., 2010) based on sequence alignments using mafft aligner (Katoh et al., 2002). The samples were then analyzed using weighted Unifrac distances for PERMANOVAs and Redundancy analyses (RDA). Correlations of the most abundant taxa with environmental variables were tested using the Pearson correlation method. The analyses were done using R and the "vegan", "tidyverse", "phyloseq", "ggplot2", "forcats", "patchwork", "scales", "microbiomeSeq", and "ape" packages (Lin Pedersen, 2020; Mazerolle, 2020; McMurdie and Holmes, 2013; Oksanen et al., 2020; Paradis and Schliep, 2019; R core team, 2021; Ssekagiri et al., 2017; Wickham, 2020; Wickham et al., 2019; Wickham and Seidel, 2020).

## 3. Results

#### 3.1. Environmental variables and cell numbers

Temperatures ranged from 5.7 °C in surface waters to 1.8 °C and were higher in 2018 than in the other years (SUP 1). On ice free stations, the temperature was higher at the surface and decreased with depth, while the ice covered stations P6 and P7 showed negative temperatures in the first 40–60 m,which increased towards 200 m, andthen decreased again with depth (SUP 1). Salinities ranged around 34.9 at stations P1, P6, and P7 and around 34.5 at the other stations (SUP 1). Generally, the

salinity increased after the first 5–40 m, an effect that was strong in ice covered stations, as a signature of ice melt (SUP 1). Similarly, the PO<sub>4</sub> concentrations increased with depth, reaching  $\sim 0.8 - 1.0 \,\mu\text{mol} \,l^{-1}$  for the deepest samples (SUP 1). The NO<sub>2</sub> concentrations were low ranging between 0.01 and 0.35  $\mu\text{mol} \,l^{-1}$ , while NO<sub>3</sub> concentration were low in surface waters, but increased with depth (SUP 1). The SiO<sub>4</sub> concentrations also increased with depth ranging from  $> 0.7 \,\mu\text{mol} \,l^{-1}$  in surface waters up to 13.1  $\mu\text{mol} \,l^{-1}$  (P7, 2021) at depth (SUP 1). Bacterial and archaeal cell abundance decreased from the surface to depth (SUP 1). Overall, cell numbers were lowest in 2018 and varied between stations in 2019 and 2021. At P2 - P5, and P7, cell numbers were higher in 2019, as compared to 2021, and showed the highest overall numbers at P5 with 2.3 × 10<sup>6</sup> cells ml<sup>-1</sup> (SUP 1). While most stations showed distinct maxima in 2019, in 2021 maxima were only found at station P1 (8.7 × 10<sup>5</sup> cells ml<sup>-1</sup>) and P6 (1.2 × 10<sup>6</sup> cells ml<sup>-1</sup>; SUP 1).

Chl a concentrations were low in 2018, never exceeding  $0.5 \,\mu g \, l^{-1}$ . In 2019, the Chl a concentrations were higher and distinct chlorophyll maxima were present at all stations (SUP 1). The highest Chl a concentrations were measured in 2021 (6.2  $\mu$ g l<sup>-1</sup> at P5) except P7 (Table 1; SUP 1). However, the highest integrated abundance of phytoplankton was found in 2019, where all stations exceeded  $1 \times 10^{10}$  cells per m<sup>-3</sup>, with a maximum of 8.8  $\times$  10<sup>10</sup> cells per m<sup>-3</sup> at station P6 (Table 1; Assmy et al., 2022b; Kohlbach et al., 2023). The phytoplankton was dominated by flagellates in 2019, especially Heterosigma sp. at stations P5 - P7 (Table 1; Kohlbach et al., 2023). The abundance was lower in 2021 with a maximum of  $4.5 \times 10^{10}$  cells per m<sup>-3</sup> at P1 (Table 1; Assmy et al., 2022c). This peak was constituted by Dinobryon spp. Cells, while stations P4-P6 showed more diatom abundance and a peak of Strombidium sp. ciliates at P6 (Table 1; Kohlbach et al., 2023). Phytoplankton abundance was lowest in 2018, where only station P5 exceeded an abundance of  $1.0 \times 10^{10}$  cells per m<sup>-3</sup> dominated by *Chysochromutina sp.* (Table 1; Assmy et al., 2022a; Kohlbach et al., 2023). In 2018 all stations were ice free, while in 2019 and 2021 stations P5 to P7 were covered with ice (Table 1). Station P4 was mostly ice covered in 2019 and station P3 and P4 became ice free just before sampling in 2021 (Table 1). Stations P1 and P2, as well as station P3 in 2018 and 2019, were ice free for at least a month before sampling (Table 1.

#### Table 1

Table with ice coverage, ice free time, maximum Chl a concentration and integrated phytoplankton cell abundance over the first 90 m (\* first 30 m, # first 60 m) per station per year (Assmy et al., 2022a, 2022c, 2022b; Kohlbach et al., 2023; Assmy pers. comm.).

| Year | Station | Ice<br>cover<br>[%] | Ice free<br>time<br>[days] | Chl a conc.<br>maximum [µg<br>l <sup>-1</sup> ] | Integrated<br>phytoplankton<br>abundance [cells<br>m <sup>-3</sup> ] |
|------|---------|---------------------|----------------------------|-------------------------------------------------|----------------------------------------------------------------------|
| 2018 | P1      | 0                   | 220                        | 0.42                                            | $5.1 	imes 10^9$                                                     |
|      | P2      | 0                   | 86                         | 0.38                                            | $6.0 	imes 10^9$                                                     |
|      | P3      | 0                   | 80                         | 0.26                                            | $1.7	imes10^9$                                                       |
|      | P4      | 0                   | 76                         | 0.06                                            | $9.6 	imes 10^8$                                                     |
|      | P5      | 0                   | 76                         | 0.24                                            | $1.2 	imes 10^{10}$                                                  |
| 2019 | P1      | 0                   | 89                         | 1.22                                            | $2.9 	imes 10^{10}$                                                  |
|      | P2      | 0                   | 43                         | 1.24                                            | $3.3	imes10^{10}$ #                                                  |
|      | P3      | 0                   | 44                         | 0.80                                            | $2.6	imes10^{10}$                                                    |
|      | P4      | 82                  | 0                          | 1.37                                            | $3.0 	imes 10^{10}$ *                                                |
|      | P5      | 95                  | 0                          | 2.57                                            | $6.9	imes10^{10}$                                                    |
|      | P6      | 93                  | 0                          | 1.29                                            | $8.8\times10^{10}$                                                   |
|      | P7      | 95                  | 0                          | 1.74                                            | $1.8	imes10^{10}$                                                    |
| 2021 | P1      | 0                   | 195                        | 3.49                                            | $4.5 	imes 10^{10}$                                                  |
|      | P2      | 0                   | 39                         | 2.17                                            | $9.8	imes10^9$                                                       |
|      | P3      | 0                   | 1                          | 1.53                                            | $3.5	imes10^9$                                                       |
|      | P4      | 0                   | 2                          | 6.08                                            | $1.0	imes 10^{10}$                                                   |
|      | P5      | 98                  | 0                          | 6.19                                            | $1.4	imes 10^{10}$                                                   |
|      | P6      | 93                  | 0                          | 2.72                                            | $1.4	imes10^{10}$                                                    |
|      | P7      | 100                 | 0                          | 0.23                                            | $4.6 	imes 10^9$                                                     |

\* = Integrated over the first 30 m.

 $^{\#}$  = Integrated over the first 60 m.

### 3.2. Bacterial and archaeal community composition

The sequencing resulted in 11,713,482 unmerged input sequences and 4,301,652 merged sequences after all quality controls. PERMA-NOVA analyses of the bacterial and archaeal communities at all depths showed no significant differences between the samples, the stations, or the different years. The only significant differences was found between the surface and mesopelagic samples (p = 0.007; n = 67). Therefore, surface and deep samples were analysed separately. These analyses showed significant differences between 2018 and 2019/2021 in the surface (p = 0.007; n = 29), as well as at depth (p = 0.035; n = 21). Redundancy analyses (RDA) of the surface and mesopelagic waters showed significant correlations of the surface communities with temperature (Pr(>F) = 0.046), salinity (Pr(>F) = 0.007), and ice-free days (Pr(>F) = 0.018; Fig. 3). This separated most of the samples taken in 2018 and stations P1 and P2 from stations P3-P7 in 2019 and 2021. Significant correlations with the mesopelagic communities were found for station, depth, phosphate concentrations, cell abundance (Pr(>F) <0.01), and salinity (Pr(>F) = 0.038; Fig. 2) thus separating stations P6 and P7 from stations P1 - P5.

The bacterial and archaeal community in all samples was dominated by *Alphaproteobacteria*, *Gammaproteobacteria*, *Bacteroidia*, and *Verrucomicrobia*, which made up > 99% of the surface community (Fig. 3). In the mesopelagic, these classes were still most abundant, but *Thermoplasmata*, members of the SAR324 clade, and specifically *Nitrosphaeria* increased to ~ 32% of the community, thus distinguishing deep samples from the surface water at all stations (Fig. 3).

During 2018, *Gammaproteobacteria* were most abundant with ~ 37% relative abundance at surface and depth, and maxima at stations P2 and P5, where relative abundances of 54.1% (P2; 180 m) and 44.6% (P5; 155 m) were found (Fig. 3). In 2019 and 2021 the relative abundance of *Gammaproteobacteria* was lower with total means of ~ 30% relative abundance (Fig. 4). Still, in 2021 at station P1 at 25 m, the highest *Gammaproteobacteria* abundance was found with 63.6% relative abundance (Fig. 3).

Among the *Gammaproteobacteria*, *Nitrincolaceae* were the most abundant with  $18.6 \pm 9.0\%$ ,  $8.4 \pm 6.2\%$ , and  $12.0 \pm 10.8\%$  in 2018, 2019, and 2021, where a maximum of 45.4% was reached (Fig. 4). *Pseudohongiella*, *Methylophagaceae*, and members of the SAR92, SAR86, and OM60(NOR5) clades were also found in higher relative abundance in surface waters as compared to deep waters throughout the sample set (Fig. 4). On the contrary, members of the SUP05 cluster were found in higher relative abundance at depth (Fig. 4).

Alphaproteobacteria were the second most abundant class in 2018 in surface waters with 33.0  $\pm$  9.7% relative abundance in the surface and 21.1  $\pm$  3.7% at depth (Fig. 3). In 2019 they constituted for 23.5  $\pm$  14.7% total relative abundance and in 2021 with 21.0  $\pm$  13.0% relative abundance in surface waters (Fig. 3). The *Alphaproteobacteria* were dominated by members of the SAR11 clade Ia with 19.3  $\pm$  9.5% in 2018, 29.5  $\pm$  17.6% in 2019, and 13.8  $\pm$  10.7% in 2021 in the surface, decreasing to 11.6  $\pm$  3.5%, 7.2  $\pm$  4.6%, and 10.3  $\pm$  6.0% in deep waters (Fig. 4). This was complemented by members of the SAR11 clade (Fig. 4).

The most distinct differences between surface and deep were found in the *Bacteroidia*, which showed a relative abundance of  $27.4 \pm 10.6\%$ ,  $36.2 \pm 20.5\%$ , and  $44.3 \pm 19.7\%$  at the surface in 2018, 2019, and 2021, as compared to  $7.8 \pm 2.8\%$ ,  $8.9 \pm 7.5\%$ , and  $7.2 \pm 4.6\%$  relative abundance at depth (Fig. 3). The high variations in surface water in 2019 resulted from abundances as high as 80.0% and 64.8% at stations P6 and P5 (Fig. 3). The relative abundance of different genera of *Bacteroidia* showed more variability with dominant genera being in high abundance at the surface of specific stations and generally in lower abundance at depth. *Polaribacter* was most abundant with 52.0% relative abundance at station P1 in 2021 and even 72.0% relative abundance at station P6 in 2019, dominated by the most abundant ASV in the dataset (57.3% relative abundance) (Fig. 4). *Cryomorphaceae* were found most



**Fig. 2.** Ordination plot of a redundancy analyses of all years separated by surface and mesopelagic. Environmental variables were abbreviated and units were omitted for readability (Ice = ice cover, Phytoplankton = Integrated phytoplankton cell abundance (90 m), Cells = Bacterial and archaeal cell abundance, Temp. = Temperature). Chl a concentration, Ice cover and ice free days were omitted in the RDA of the mesopelagic samples.

abundant at station P4 in 2018 and *Formosa* were highest at station P1 in 2019, both showing distinct fluctuations between stations in all years (Fig. 4). The fluctuations of the other dominant members, namely *Ulvibacter* and members of the NS5 clade, were not as distinct. Still *Ulvibacter* showed 13.8  $\pm$  1.2% relative abundance in the surface waters of stations P4 and P5 in 2021 (Fig. 4).

Verrucomicrobia were the next abundant class in the surface  $\sim 2.5\%$ relative abundance in all years, increasing to  $\sim 4.2\%$  relative abundance at depth (Fig. 3). Lentimonas showed increased relative abundance in the surface waters, reaching a maximum of 12.0% in 2021 (Fig. 4). The archaeal Nitrosphaeria and Thermoplasmata were found in low relative abundance in the surface in all years, with the exceptions of three samples where they accounted for 2.1% and 0.2% at station P2 (2021), 11.5% and 1.7% at station P7 (2021), and 15.3% and 0.7% at station P3 (2019) (Fig. 3). At depth, these two groups were increased in all years with relative abundance up to 23.8  $\pm$  3.3% (2019) and 7.8  $\pm$  2.0% (2021; Fig. 3). Cand. Nitrosopumilus and members of the Marine Group II were the most abundant Archaea in deep waters with 8.8  $\pm$  4.5% and 11.6  $\pm$  4.5%, 20.1  $\pm$  6.0% and 5.1  $\pm$  2.6%, and 19.1  $\pm$  6.8% and 6.5  $\pm$ 2.2% relative abundance in the consecutive years, both being nearly absent in surface waters of all years (Fig. 4). Similarly, members of the SAR324 clade and Marinimicrobia were almost exclusively found in deep waters with 4.1  $\pm$  2.7%, 3.7  $\pm$  2.4%, and 4.5  $\pm$  2.3%, as well as 3.5  $\pm$ 2.5%, 3.1  $\pm$  2.2%, and 3.5  $\pm$  2.3% relative abundance in the different years (Fig. 4).

The most abundant taxa were used for correlation analyses with the environmental variables, resulting in a distinct pattern of "deep" taxa with the members of the SAR324 clade, the SAR11 clade II, the SUP05 cluster, Marine Group II archaea, *Cand.* Nitrosopumilus, and *Marinimicrobia* all correlated with depth and the correspondingly higher nutrient concentrations (Fig. 5). On the contrary, most taxa with higher relative abundance in the surface, such as *Polaribacter, Flavobacteriaceae*, *Ulvibacter, Cryomorphaceae*, *Nitrincolaceae*, the NS5 marine group, and SAR92 were negatively correlated to these variables, but positively correlated to Chl a (Fig. 5).

#### 4. Discussion

The overall community structure, dominated by Gammaproteobacteria, Alphaproteobacteria, and Bacteroidia in the surface waters and by Nitrosphaeria, Thermoplasmata, Marinimicrobia, and members of the SAR324 clade in the deep waters, resemble previous reports of Arctic marine communities (de Sousa et al., 2019; Müller et al., 2018; Wilson et al., 2017). The distinct differences between the surface and the deep water communities were confirmed by Pearson correlations of the most abundant taxa, showing that surface taxa are positively correlated to Chl a and negatively correlated to nutrients and vice versa for deep taxa. Most Bacteroidia and Gammaproteobacteria show positive correlations with Chl a and negative correlations to nutrients and salinity, showing that Chl a, as compared to nutrients, is an important factor for the proliferation of the surface community. In deep waters, the nutrient concentrations and salinity are elevated and correlate with the most important taxa in these samples. The differences of communities in the deepest samples of stations P6 and especially P7 to other deep communities, are potentially due to the greater depth of > 500 m and the consequently higher abundance of "deep" taxa, such as Marinimicrobia, members of the SAR324 and SUP05 clades. The bottom depths of the other stations were between  $\sim 332$  m (P4) and  $\sim 160$  m (P5), potentially allowing for some mixing of the surface and the deep community, while this decreases with depth.

The high relative abundance of *Alphaproteobacteria*, consisting mainly of members of several SAR11 clades, is common for oligotrophic marine environments (Morris et al., 2002; Schattenhofer et al., 2009; Thiele et al., 2012; Wilson et al., 2017). On the contrary, the high relative abundance of *Gammaproteobacteria*, dominated by *Nitrincolaceae* (formerly *Oceanospirillaceae*) and *Bacteroidia*, dominated by *Polaribacter*, indicate waters with elevated concentrations of phytoplankton derived carbon sources (Gomez-Pereira et al., 2010; Puddu et al., 2003; Simon et al., 1999; Teeling et al., 2016, 2012; Thiele et al., 2012). *Nitrincolaceae* were found in Antarctic phytoplankton blooms and Arctic sea ice algal blooms, and are known for high diversity in carbon utilization metabolisms and rapid in response to increased carbon availability (Liu et al., 2020; Mönnich et al., 2020; Mori et al., 2019; Park

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2018

Rel. abundance [%]

Rel. abundance [%]

100

5 m -

Chl. a max.

200 m -

Bottom -

P1

P2 P3 P4 P5 5 m-30 m -5 m-200 m -5 m -Bottom -Bottom -200 m -5 m-Bottom -30 m -Bottom -Chl. a max. Chl. a max. P4 P5 P6 P7 **P**3

10 m -

Chl. a max.













Fig. 3. Relative abundance of the 15 most abundant classes covering ~ 99% of all classes detected using 16S rRNA gene sequencing. The remaining classes are summed as "Other classes". Backgrounds of station names mark the ice coverage from open water (dark grey) to lose pack ice (grey) and dense pack ice (white) cover.

et al., 2020; Thiele et al., 2022). Polaribacter has frequently been found in correlation with phytoplankton blooms in polar, temperate, and tropical regions (Gomez-Pereira et al., 2010; Teeling et al., 2016; Thiele et al., 2015, 2012; Xing et al., 2015). These two taxa mark a prominent difference between the three years. While Nitrincolaceae are  $\sim 20$  - 45% more abundant in 2018 surface waters, Polaribacter is  $~^{\sim}70\%$  more abundant in 2019 and  $\sim$  80% more abundant in 2021 in the surface. As

Bacteroidia, such as Polaribacter, Lentimonas, and Ulvibacter, have been found in high abundance in correlation with phytoplankton blooms due to their multitude of carbon degradation pathways and carbohydrateactive enzyme content (Gomez-Pereira et al., 2010; Reintjes et al., 2020, 2019; Teeling et al., 2016, 2012), this explains the correlation of the most abundant Bacteroidia with Chl a. These differences may point to an earlier postbloom stage of the community in 2019/2021 than in



Fig. 4. Relative abundance of the most abundant taxa per class (color coded) for surface (•) and deep (•) waters of the stations P1 to P5 (2018) and P1 to P7 (2019/2021). Backgrounds of station names mark the ice coverage from open water (dark grey) to lose pack ice (grey) and dense pack ice (white) cover.



Fig. 5. Positive (red) and negative (blue) Pearson-correlations of the environmental variables and the most abundant taxa by year with significant correlations marked by \* (\* = p-value  $\leq 0.05$ , \*\* = p-value  $\leq 0.01$ , \*\*\* = p-value  $\leq 0.001$ ; p-values are Benjamini Hochberg adjusted).

2018. While samples in 2018 and 2019 were taken in August, samples of 2021 are from July, which could result in differences of the ecosystem state with respect to the state of the phytoplankton spring bloom. Therefore, we assume an earlier state of the bloom for 2021 than for 2019 and 2018. This is congruent with higher Chl a concentrations in 2021 than in the other years, potentially indicating the onset of a phytoplankton bloom. The lower Chl a concentrations, despite higher phytoplankton cell abundance in 2019 might indicate lower photosynthetic activity, thus indicating a later bloom stage. Similarly, Chl a concentrations and phytoplankton abundance are lower in 2018, indicating a late post bloom stage. These differences in the years could be aggravated by changes in the light environment and potentially earlier onset of the phytoplankton bloom due to differences in sea-ice cover (Dalpadado et al., 2020). This is supported by the significance of the ice free days, temperature, and salinity for the surface layer, where temperature and salinity were lower on ice covered stations, and in the separation of the samples from P1, P2, and all 2018 stations from samples from stations P3 - P7 in 2019/2021 in the RDA analyses. Lower temperatures and salinity are correlated to ice cover and might hence indirectly reflect the influence of ice cover on the bacterial and archaeal community. This influence is most likely a secondary effect. Sea ice cover reduces the light environment and hence leads to later onsets of phytoplankton blooms. These phytoplankton blooms then trigger a succession of the bacterial and archaeal community based on the availability of carbon sources. Therefore, the effect is secondary for the bacterial and archaeal community and direct correlations require larger investigations.

The bacterial and archaeal communities of stations P6 and P7 were distinctively different to the other stations sampled. The surface of station P7 was marked by the highest abundance of members of the SAR11

Clade in both years, thus indicating a rather oligotrophic state suggesting a pre-bloom or the onset of a phytoplankton bloom at these stations (Morris et al., 2002; West et al., 2008). Station P6 in 2019 was marked by an extreme dominance of *Polaribacter*, indicating the creation of favorable conditions for this taxon, potentially due to the ongoing *Heterosigma sp.* bloom. On the contrary, *Polaribacter* was relatively low on both stations in 2021. While Chl a values were low at P7, the Chl a concentrations were high at station P6. Here, the phytoplankton community was strongly dominated by *Strombidium* species (Kohlbach et al., 2023), which might not provide the preferred niche for *Polaribacter* or have exerted grazing pressure on the fast growing and hence larger *Polaribacter* cells, thus decimating their numbers (Chen et al., 2020).

The transport of Atlantic waters into the Arctic on the Svalbard Branch could potentially affect stations P1, P6, and P7 (Lind and Ingvaldsen, 2012; Loeng, 1991). In addition, the temperature profiles of stations P2 - P4 showed nearly opposite patterns than P6 and P7, which could be an indicator for different water bodies. However, a direct correlation of water bodies on the bacterial and archaeal community could not be proven and seems minor in the light of the "everything is everywhere" hypothesis (O'Malley, 2008). Secondary effects, like the drift of phytoplankton blooms and therefore carbon sources, towards these stations, may still affect the bacterial and archaeal community in the surface.

The dataset collected during three expeditions within the Nansen Legacy project showed differences of the bacterial and archaeal communities in surface waters of the Barents Sea based on the occurrence of phytoplankton and Chl a abundance, which can resemble different stages of the phytoplankton bloom and consequently carbon source availabilities. These community differences are mainly due to changes in the abundance of *Nitrincolaceae* and several *Bacteroidia* taxa, most prominently *Polaribacter*. The deep water communities were more stable and dominated by *Cand*. Nitrosopumilus, *Marinimicrobia*, and members of the SAR324 clade. Although indicating that higher average water temperatures, and less sea ice, could lead to earlier onsets of phytoplankton spring blooms, the consequences of these factors for the succession of the bacterial and archaeal community and the carbon dynamics within the microbial loop remain unclear. Over the short period of three summers, potential changes of the community dynamics might be masked by the high variability of the Arctic ecosystem. Hence, the long-term effects of global warming on the dynamics of the bacterial and archaeal communities, the microbial loop or the microbiology of the pelagic ecosystems of the Barents Sea require longer series of investigation.

#### **Declaration of Competing Interest**

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

#### Data availability

Data will be made available on request.

## Acknowledgement

This work was funded by the Research Council of Norway through the project The Nansen Legacy (RCN # 276730). We would like to thank the captain and crews of the RV Kronprins Haakon, as well as the leaders of the RF3 of The Nansen Legacy.

#### Appendix A. Supplementary data

Supplementary data to this article can be found online at https://doi.org/10.1016/j.pocean.2023.103054.

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