

ORIGINAL ARTICLE

Organic carbon transformations in high-Arctic peat soils: key functions and microorganisms

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A substantial part of the Earths' soil organic carbon (SOC) is stored in Arctic permafrost peatlands, which represent large potential sources for increased emissions of the greenhouse gases CH₄ and CO₂ in a warming climate. The microbial communities and their genetic repertoire involved in the breakdown and mineralisation of SOC in these soils are, however, poorly understood. In this study, we applied a combined metagenomic and metatranscriptomic approach on two Arctic peat soils to investigate the identity and the gene pool of the microbiota driving the SOC degradation in the seasonally thawed active layers. A large and diverse set of genes encoding plant polymer-degrading enzymes was found, comparable to microbiotas from temperate and subtropical soils. This indicates that the metabolic potential for SOC degradation in Arctic peat is not different from that of other climatic zones. The majority of these genes were assigned to three bacterial phyla, Actinobacteria, Verrucomicrobia and Bacteroidetes. Anaerobic metabolic pathways and the fraction of methanogenic archaea increased with peat depth, evident for a gradual transition from aerobic to anaerobic lifestyles. A population of CH₄-oxidising bacteria closely related to *Methylobacter tundripaludum* was the dominating active group of methanotrophs. Based on the in-depth characterisation of the microbes and their genes, we conclude that these Arctic peat soils will turn into CO₂ sources owing to increased active layer depth and prolonged growing season. However, the extent of future CH₄ emissions will critically depend on the response of the methanotrophic bacteria.

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Introduction

Recent estimates point out that 277 Pg of soil organic carbon (SOC) are stored in Arctic peatlands (Tarnocai *et al.*, 2009), which corresponds to 1/3 of the CO₂ in the atmosphere. These peat soils have acted as carbon sinks since the Holocene (Post *et al.*, 1982; Harden *et al.*, 1992). In contrast, they are substantial sources of methane (CH₄), releasing ~35 Tg per year, ~6% of the global CH₄ emissions (Cao *et al.*, 1996). CH₄ is a more potent greenhouse gas than carbon dioxide (CO₂), having 25 times the global warming potential of CO₂ on a 100-year timescale (Forster *et al.*, 2007). Arctic and especially high-Arctic regions are already exposed to, and predicted to experience, a strong temperature increase until the end of the century (4–8 °C higher

annual surface air temperatures) (IPCC, 2007), which is expected to lead to expanded frost-free vegetation periods and increased active layer depths in permafrost soils (Tarnocai *et al.*, 2009).

The major factors in SOC degradation in peatlands, like in mineral soils, are microorganisms of the bacterial, archaeal and eukaryotic (fungi) domains of life, participating in a cascade of aerobic and anaerobic degradation steps, eventually resulting in the emission of CH₄ and CO₂. Major parts of the peat carbon are plant polymers such as cellulose and hemicellulose. The degradation of these to oligomeric and monomeric sugars is considered one of the most important steps in anaerobic degradation of SOC (Kotsyurbenko, 2005) and is catalysed by a diverse set of hydrolytic extracellular enzymes produced by microorganisms. Further steps in the anaerobic degradation are carried out through anaerobic respiration (like denitrification), fermentative and methanogenic pathways, while microbial methane oxidation constitutes the biological filter for methane emissions from peat. Phenolic substances accumulate to high concentrations in the anaerobic, water-logged layers of peat soils, partly owing to low activity of phenol oxidases,

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which require oxygen for function. The inhibiting effect of phenolic substances has been suggested as a major factor for the low SOC degradation rates in peat soils (Fenner and Freeman, 2011).

Several studies have targeted the microorganisms involved in CH₄ formation (for example, Ganzert *et al.*, 2007; Kotsyurbenko *et al.*, 2007; Metje and Frenzel, 2007; Høj *et al.*, 2008) and oxidation (for example, Warttinen *et al.*, 2003; Liebner *et al.*, 2009; Martineau *et al.*, 2010; Graef *et al.*, 2011) in Arctic soils. Broader studies have also been conducted, for example, on the biogeography of soil bacterial communities in different climatic zones (Chu *et al.*, 2010), the dynamics of Arctic soil microbial communities in relation to the composition of plant communities (Zak and Kling, 2006) and the availability of SOC (Waldrop *et al.*, 2010 and Coolen *et al.*, 2011). In recent years, metagenomics and metatranscriptomics have developed into powerful tools in microbial ecology that enable non-targeted studies of the genetic potential, gene expression and the composition of soil microbial communities (for example, Tringe *et al.*, 2005 and Urich *et al.*, 2008). Until now only two metagenomic studies on Arctic soils have been conducted, comparing the community in the permafrost with its overlying active layer (Yergeau *et al.*, 2010) and investigating the response of a permafrost microbial community to thaw (Mackelprang *et al.*, 2011).

It is in the active layers, where the SOC degradation in Arctic peat soils occurs, that more knowledge about the processes and organisms involved is required to understand these and to predict the magnitude of CH₄ and CO₂ emissions from Arctic peat in a warmer climate. We have in this study applied, for the first time, a combined metagenomic and metatranscriptomic approach on high-Arctic peat soils to obtain on-site information about the genomic potential for SOC transformation and to identify the active microorganisms driving these processes. We compared the Arctic peat metagenomes with metagenomes from temperate soils and proposed a model of SOC degradation in Arctic peat, including anaerobic respiratory and fermentative pathways, and methanogenic and methanotrophic microorganisms. The data allow some predictions about future CH₄ and CO₂ emissions from Arctic peat soils.

Materials and methods

Study sites and sampling

We have investigated the active layers of two high-Arctic fens on Svalbard (Norway). Solvatn (N78°55.550 E11°56.611) and Knudsenheia (N78°56.544 E11°49.055) were sampled in August 2009 at the peak of the growing season (Supplementary Figure S1 and S2). Both sites are located in the vicinity of the research settlement Ny-Ålesund. Solvatn is situated on a marine terrace right next to the settlement, while Knudsenheia is located ~5 km northwest of Ny-Ålesund. For both sites,

Knudsenheia and Solvatn, two biological replicates were prepared. Three peat blocks (20 × 30 × 20 cm) were cut from each of the replicate sites, S1 and S2 (Solvatn) and K1 and K2 (Knudsenheia; see Supplementary Figures S1–S3), and transported to the laboratory within 2 h after sampling. The active layer depths was ~40 cm at both sites. The peat blocks were subsequently stored for 2–3 h at 4 °C until further processing. The three blocks were separated into upper (more oxic) and lower (more anoxic) layers, and the corresponding layers, within each replicate site, were pooled to account for heterogeneity of each replicate site (Supplementary Figure S3). The layer separation was based on visual distinction of the layers based on its colour and structure. The lower layers were processed under nitrogen atmosphere to avoid extensive oxygen contamination. After pooling, samples were immediately shock frozen in liquid nitrogen. The whole procedure took ~5–10 min. Soil pH was measured using the KCl extraction method on fresh peat samples. Water content was estimated by drying peat over night at 120 °C and reweighing the samples. To estimate organic matter contents, the dried samples were combusted at 450 °C overnight. The samples were reweighed, and the ash weight was calculated and added to the weight of organic matter. Concentrations of volatile fatty acids and ethanol were determined by high pressure liquid chromatography analysis of pore water (Metje and Frenzel, 2005) extracted from soil samples before freezing. Results were integrated using peak simple version 393, and 1 mM standards.

Nucleic acids extraction, reverse transcription and 454 sequencing

From each layer of the biological duplicates (Supplementary Figure S3), six or more parallel extractions of nucleic acids were performed using a modified version of the Griffith's protocol (Urich *et al.*, 2008). In samples for metatranscriptomic analysis, DNA was digested using the RQ1 DNase treatment (Promega, Madison, WI, USA), followed by RNA purification using the MEGAclean kit (Ambion, Austin, TX, USA). Samples were subjected to reverse transcription using the Superscript II double-strand complementary DNA synthesis kit (Invitrogen, Carlsbad, CA, USA), following the manufacturer's protocol, with the exception that both first- and second-strand synthesis was carried out for 4 h. RNA template addition was in the range of 500–1000 ng. Concentrations of double-stranded complementary DNA were estimated by SYBR Green I (Invitrogen) assay (Leininger *et al.*, 2006). Nucleic acids for metagenomic analysis were subjected to RNase A treatment for 20 min at 37 °C, followed by phenol:chloroform:isoamylalcohol extraction and chloroform:isoamylalcohol extraction. DNA concentrations were estimated using a NanoDrop spectrophotometer (Thermo Fisher Scientific, Madison, WI, USA). Roche 454 GS FLX Titanium sequencing

(454 Life Sciences, Branford, CT, USA) was carried out at the CEES at the University of Oslo.

Bioinformatic analysis

Metatranscriptomic and metagenomic sequences were first filtered using LUCY (Chou and Holmes, 2001), removing short (<150 bp) and low-quality sequences (>0.2% error probability). Metatranscriptomic sequences stemming from ribosomal RNA (rRNA) and putative messenger RNAs (mRNAs) were separated by comparing all sequences against a combined database of small and large subunits of rRNA (SSU and LSU rRNA) using BLASTN (Altschul *et al.*, 1997) and MEGAN (Huson *et al.*, 2007; Urich *et al.*, 2008). Sequences with a bit score <50 were assigned as putative mRNA tags. Metagenomic and putative mRNA tags in the metatranscriptomes were subjected to the 454 Replicate filter (Gomez-Alvarez *et al.*, 2009) for the removal of artificially replicated sequences. The filter was applied to remove exact duplicate sequences from the metatranscriptomes (settings: sequence identity cutoff: 1.0, length difference requirement: 1.0 and number of beginning base pairs to check: 3). For the metagenomes, the default settings were applied.

SSU ribo-tags were taxonomically assigned by MEGAN analysis of a BLASTN file against a SSU rRNA reference database (parameters: min. bit score 150, min. support 1, top percent 10; 50 best blast hits) (Urich *et al.*, 2008; Lanzen *et al.*, 2011). SSU ribo-tags assigned to the order Methylococcales were assembled into ribo-contigs using CAP3 (Huang and Madan, 1999), using two subsequent rounds of assembly with (1) a minimum overlap of 150 bp with a minimum similarity threshold of 99% and mismatch and gap scores of -130 and 150, and (2) minimum overlap of 150 bp and minimum 97% similarity threshold, respectively. Contigs were uploaded in ARB (<http://www.arb-home.de/>), and contigs longer than 1300 bp together with selected reference sequences were used to generate a maximum likelihood tree with the RAxML (Stamatakis *et al.*, 2005) algorithm implemented in ARB using default settings.

The metagenome sequences and the mRNA fraction of the metatranscriptome were functionally annotated for initial screening using the metagenomics (MG)-RAST server with maximum *e*-value for a significant match set to $1e-4$ (Meyer *et al.*, 2008). The sequences were also taxonomically binned by MEGAN analysis (parameters: min. bit score 50, min. support 1, top percent 2; 50 best blast hits) of BLASTX files against the RefSeq protein database (*e*-value < $1e-4$). Screening for specific functional genes and transcripts were carried out using custom reference databases generated from the UniProtKB/Swiss-Prot or UniProtKB/TrEMBL databases. BLASTX searches were carried out with an *e*-value threshold of $1e-10$. All sequences below threshold were selected as query sequences for a BLASTX search with the same parameters against

the RefSeq database. Only those that were assigned the same function in the second round were considered positives.

In the metagenomes, the genes encoding carbohydrate-active enzymes were annotated. The metagenomic reads were translated into all six frames, each frame into separate ORFs (open reading frames), avoiding any '*' characters marking stop codons in a resulting ORF. All ORFs equal to 40 amino acids or larger were screened for assignable conserved protein domains. Glycoside hydrolases (GH), proteins with a carbohydrate-binding motif, carbohydrate esterases, polysaccharide lyases and other carbohydrate-active enzymes were inspected by reference HMMs (Hidden Markov Models) using HMMER tools (<http://hmmer.janelia.org/>) with the PFAM database HMMs (PFAM release 25, <http://pfam.janelia.org>). All database hits with *e*-values below a threshold of 10^{-4} were counted. For the GH44 family, no PFAM HMM is available. Therefore, representative sequences were selected from the CAZy website (<http://www.cazy.org>), the sequence regions corresponding to the family were determined and used to create a HMM for the GH44 family. The screening was performed on the HPC computer STALLO at the University of Tromsø (<http://docs.notur.no/uit>). A selection of sequences was also used as query sequences in BLASTP searches against the RefSeq database of the NCBI. Within the chosen threshold of the HMMER, all identified carbohydrate-active enzymes gave either a corresponding hit or no hit in the RefSeq database. For the taxonomic assignment of gene and transcript sequences, the corresponding BLAST outputs were uploaded in MEGAN (parameters: min. bit score 35, min. support 1, top percent 10; 50 best blast hits).

Statistical analysis

Significant differences between the frequencies of conserved protein domains in ORFs of different soils were evaluated statistically by using the R package (R_Development_Core_Team, 2009) using the χ^2 contingency table test. The contingency table holds the frequency counts of hits and non-hits for a certain PFAM domain category of two different soils. The total frequency count is given by all hits found for any domain in the PFAM database. In cases where the frequencies are too low to meet the rules of the test, the probabilities of observing a sample statistic as high as the test statistic were calculated by Monte Carlo simulations with 100 000 replicates.

Data deposition

The sequence data generated in this study was deposited in the Sequence Read Archive of NCBI and are accessible through accession number SRP014474.

Results and discussion

The active layers of the two permafrost fens at Solvatn and Knudsenheia were slightly acidic

(pH5–6), had a water content of ~70–90%, an organic matter content of 40–90% (Supplementary Table S1) and were characterised by a moss cover dominated by *Calliergon richardsonii* (Solheim *et al.*, 1996; Warttinen *et al.*, 2003; Høj *et al.*, 2005). The mosses were interspersed by grasses (*Dupontia pelligera*), which were heavily suppressed by grazing Barnacle geese. High concentrations of acetate, lactate and ethanol indicated anaerobic metabolisms through fermentative and acetogenic pathways, even in the top layers (Supplementary Table S1). Remarkably high yields of nucleic acids (DNA and RNA) per gram dry peat were obtained from all samples and layers (Supplementary Table S2), which suggests a high microbial abundance and activity at the peak of the growing season. The yields, especially of RNA, decreased with depth, indicative of decreasing microbial activity. Replicated metatranscriptome data sets were obtained from all upper layers (Supplementary Figure S3) whereas the reverse transcription was severely inhibited in the lower layers, presumably owing to phenolic substances (Supplementary Table S2). Metagenomic libraries were obtained from all layers of one duplicate from Solvatn and Knudsenheia (K1 and S2, Supplementary Figure S3). 454 Titanium pyrosequencing of the metatranscriptomes and metagenomes resulted in 120 000–190 000 sequences per library, with one exception (S2b complementary DNA, Supplementary Table S3). The number of rRNA sequences in these libraries ranged from 100 000 to 150 000 while the number of putative mRNA sequences ranged from 7000 to 8500.

Community structure of active microorganisms

The community composition of bacteria, archaea and eukaryotes, as determined from the small subunit rRNA fragments (SSU ribo-tags) of the metatranscriptomes (up to 74 000 ribo-tags), revealed that the biological replicates of both sites had a similar composition of taxa at domain, phylum and class levels of resolution (Supplementary Figure S4). Ribo-tags are considered a measure of living biomass, as they stem from ribosomes, most likely indicative of active organisms (Urich *et al.*, 2008; Urich and Schleper, 2011). Bacteria comprised most ribo-tags; ~70%–80% in all upper layers and ~84% in the lower, anoxic layer of Solvatn, S2b (Figure 1). Eukaryotic ribo-tags comprised ~15%–30%, whereas the fraction of Archaea was surprisingly small in all the upper layers (~0.01%–0.13%), but higher in S2b (~1%). The majority of archaeal sequences stemmed from methanogenic archaea. Similar low relative abundances of Archaea have been reported from Alaskan permafrost soils (for example, Waldrop *et al.*, 2010; Wilhelm *et al.*, 2011). The bacterial community composition in both peat soils was very similar, with most ribo-tags stemming from eight phyla, most

notably the Proteobacteria (~37%–45% of bacterial ribo-tags), Actinobacteria (~15%), Planctomycetes (~9%–14%), Verrucomicrobia (~9%–11%), Acidobacteria (~8%–10%) and the Chloroflexi (~7%–9%). Within the Proteobacteria, Deltaproteobacteria (mainly *Myxococcales*) were the most abundant class (~19%–21% of bacterial ribo-tags). Alphaproteobacteria made up ~8%, Betaproteobacteria comprised ~5% while Gammaproteobacteria represented on average 3.2% of the total community. The major eukaryotic taxon was the Protist kingdom Alveolata, putative grazers of bacteria, which represented ~17%–20% of eukaryotic ribo-tags in both layers of Solvatn and 6%–8% in the top layer of Knudsenheia (Figure 1). Other abundant Protists were the Amoebozoa, Rhizaria (both predators of bacteria) and Stramenopiles. Interestingly, the Protists constituted the major eukaryotic group, exceeding even the fraction of fungal (~2%–8%) and metazoan (~11%–13%) ribo-tags (Figure 1). This is in contrast to temperate and boreal mineral soils, where the fungi and metazoa comprise the majority of either eukaryotic ribo-tags (Urich *et al.*, 2008) or biomass (Schaefer, 1990; Berg and Bengtsson, 2007) and the Protists are much less abundant. Our data indicate that the bacterial energy channel might be more important than the fungal energy channel in these peat soils. Metazoan ribo-tags originated from typical micro- and meso-faunal groups of higher trophic levels, including Nematoda, Tardigrada, Rotifera and Arthropoda. Ascomycota, and to a minor extent Basidiomycota, dominated the small fungal community. At Solvatn, a change in abundance with depth was observed for several taxa (Figure 1). While Alphaproteobacteria, Betaproteobacteria and Chloroflexi were present in larger fractions in the lower layer, Planctomycetes and Verrucomicrobia decreased with depth. Most notably, however, was the increase in the fraction of Archaea and the decrease of Protists and fungi with depth. The ribo-tags originated mostly from actively transcribing organisms, indicated by a taxonomic binning of the mRNA fraction, which showed similar community profiles for the bacterial taxa (Figure 2a).

There are few studies addressing Arctic peat soils to compare our metatranscriptomic data with. 16S rRNA gene clone library, and metagenomic studies of permafrost mineral soils have shown that the Actinobacteria, Proteobacteria and Acidobacteria dominate these environments (Hansen *et al.*, 2007; Liebner *et al.*, 2008; Steven *et al.*, 2008; Chu *et al.*, 2010; Yergeau *et al.*, 2010; Wilhelm *et al.*, 2011), while clone library, T-RFLP and FISH studies of a sub-Arctic acidic sphagnum peat report a dominance of Alphaproteobacteria, Betaproteobacteria, Acidobacteria, Verrucomicrobia and Planctomycetes (Dedysh *et al.*, 2006; Pankratov *et al.*, 2011). A recent study of permafrost peat showed that the active layer communities were dominated by Actinobacteria, Proteobacteria and Chloroflexi, while the permafrost microbiota also contained large

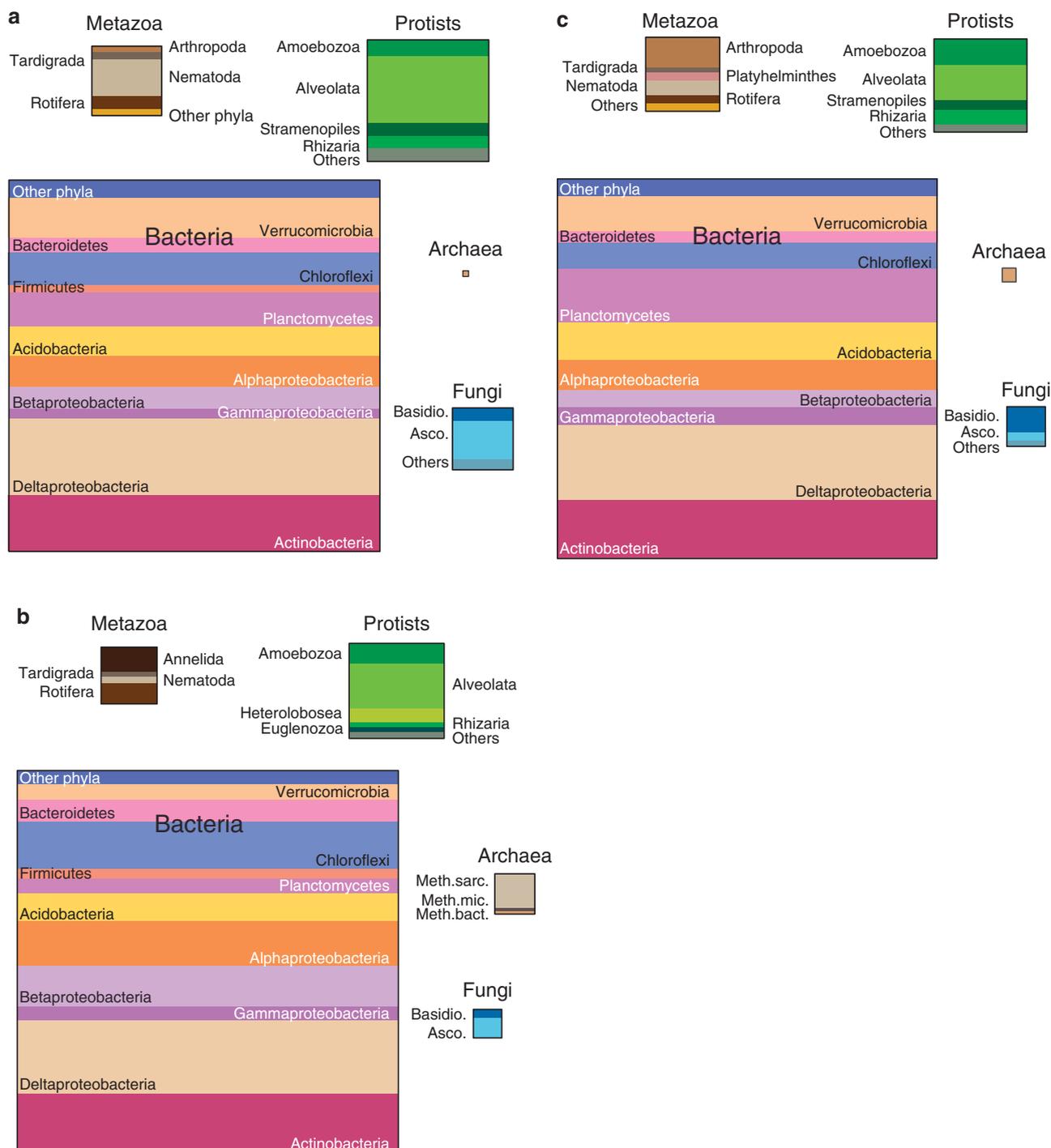


Figure 1 Three-domain community profile of the microbiota in Svalbard peatlands. The figures are based on the ribo-tags fraction of the metatranscriptome. The size of the boxes is proportional to the fraction of ribo-tags of the respective taxa. (a) Top layer of Solvatn peat generated from two biological replicate data sets (S1a and S2a). (b) Lower layer of Solvatn generated from one data set (S2b). (c) Top layer of Knudsenheia peat generated from two biological replicate data sets (K1a and K2a). Asco, Ascomycota; Basidio, Basidiomycota; Meth.bac., *Methanobacteriales*; Meth.mic., *Methanomicrobiales*; Meth.sarc., *Methanosarcinales*.

populations of Bacteroidetes and Firmicutes (Mackelprang *et al.*, 2011). Interestingly, taxa such as the Verrucomicrobia and Planctomycetes, which were among the most abundant taxa in the Svalbard peat soils as well as the acidic Sphagnum peat, were not reported as major taxa in the studies on mineral soils.

Genomic potential for degradation of plant polymers
Plant polymer and phenolic compound degradation are key processes of SOC decomposition in peat soils. We analysed the genomic potential for these processes, by screening the metagenomes for genes encoding conserved protein family (PFAM) domains

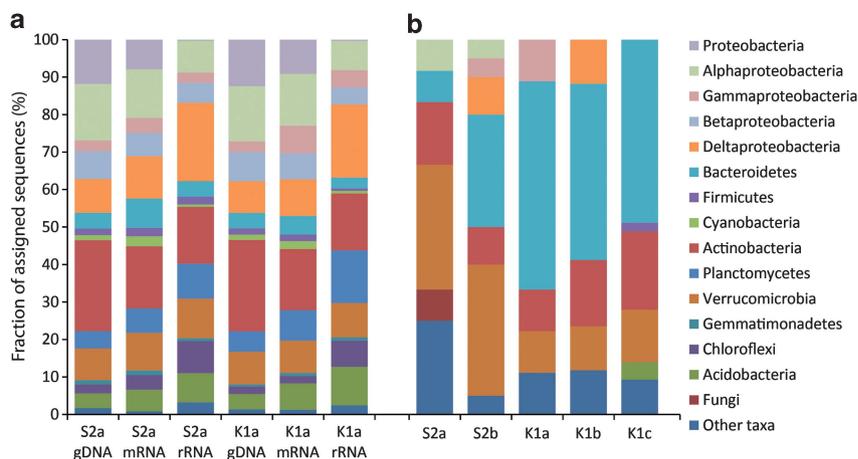


Figure 2 Taxonomic assignment of metatranscriptomic and metagenomic sequences. The community structure is displayed at phylum level resolution with Proteobacteria split into classes. Sequences assigned to Proteobacteria refer to sequences that could not be assigned to class level resolution. S2 and K1 indicates where the samples were collected, in Solvatn and Knudsenheia respectively, while a, b and c indicates the depth of the sample from top and down (Supplementary Figure S3). (a) Taxonomic assignment of ribo-tags, mRNA and metagenomic DNA (gDNA) to the domain bacteria. (b) Taxonomic assignment of the metagenomic sequences encoding polysaccharide-degrading enzymes (cellulases, endohemicellulases and debranching enzymes). All sequences assigned to these categories were pooled together and taxonomically binned using MEGAN (see Materials and methods for details).

of relevant enzymes, that catalyse the hydrolysis of abundant plant polymers such as cellulose and hemicelluloses, and the degradation of phenolic compounds (for example, lignin) (Table 1). Despite the rather uniform plant cover, dominated by the moss *C. richardsonii*, we identified a large variety of CAZY (carbohydrate-active enzyme) families, including 76 GH families, 36 carbohydrate-binding module, 3 phenol oxidase, 3 polysaccharide lyase and 5 carbohydrate esterase families (Supplementary Table S4). GH5 were the most abundant among the families of cellulases. Genes encoding endohemicellulases (especially GH families 10 and 26 that target xylan), debranching enzymes and phenol oxidases involved in the degradation of phenolic compounds were also abundant (Table 1).

We wanted to know if the microbiotas of our moss-dominated, carbon-accumulating high-Arctic peat soils contain a different genetic potential for the degradation of plant polymers and recalcitrant phenolic compounds than the microbiota of soils from other climate zones, having a vascular plant cover. Therefore, we compared our metagenomic data sets with metagenomes from temperate grasslands and farmlands, subtropic rain forest soils and Arctic permafrost mineral soils (see Supplementary Table S5 for details about the soil data sets), focussing on the PFAMs relevant for plant polymers and phenolic compound degradation. The functional potential was similar in all soils, and genes for the same PFAMs dominated the metagenomes (Supplementary Table S6). A χ^2 test did not detect statistically significant differences between the metagenomes from arctic peat and the other metagenomes (Supplementary Table S7), regardless of the variations in geography or plant coverage.

The chemical composition of moss cell walls is generally considered to be simpler than the one from

vascular plants; mosses have a different and less complex hemicellulose composition, and do not contain lignin, but a simpler and less abundant phenolic compound called lignan (Pena *et al.*, 2008; Sarkar *et al.*, 2009; Popper *et al.*, 2011), with the exception of *Sphagnum* species containing an abundant lignin-like polymer (Ligrone *et al.*, 2008). It was therefore surprising that the microbiotas harboured such a similar genomic potential. This result might to some extent be explained by methodological constraints. For instance, the PFAMs for phenolic compound degradation do not distinguish between lignin and lignan as substrate. Also, the enzymes themselves might have a broad substrate specificity. Finally, the presence of many similar types of polymers in both mosses and vascular plants should be reflected in an overall similar genetic repertoire for degradation in the respective soil microbiotas. It might well be that the actual composition of plant polymers is only reflected in the gene expression patterns of the microbiota and not in its genetic potential. However, the rather small number of mRNAs in the metatranscriptome prevented us from studying this.

Another major constraint for SOC degradation, besides the short growing season, is the accumulation of phenolic compounds in the water-logged, predominantly anaerobic peat soils, which inhibits enzymatic activity (Fenner and Freeman, 2011). Expression of phenol oxidases was detected in the top layer of Solvatn. However, the genetic potential for phenol oxidases was observed in all peat layers, indicating that a degradation of these inhibitory compounds can occur if the peat soils get oxygenated, possibly through more pronounced drought-rwetting cycles at higher temperatures, like it has recently been shown for temperate peatlands (Fenner and Freeman, 2011).

Table 1 PFAMs that target plant polymers, identified in metagenomes from high-Arctic peatlands of Svalbard

	<i>S2a</i>	<i>S2b</i>	<i>K1a</i>	<i>K1b</i>	<i>K1c</i>
<i>Cellulases</i>					
GH5	4 (2.3)	4 (1.8)	4 (2.1)	8 (3.9)	18 (5.8)
GH6	2 (1.2)	0 (0)	1 (0.5)	1 (0.5)	2 (0.7)
GH7	0 (0)	1 (0.5)	0 (0)	1 (0.5)	0 (0)
GH9	1 (0.5)	2 (0.9)	4 (2.1)	1 (0.5)	6 (2.0)
GH44	2 (1.2)	2 (0.9)	3 (1.6)	2 (1.0)	0 (0)
GH45	1 (0.6)	0 (0)	0 (0)	0 (0)	1 (0.3)
GH48	0 (0)	0 (0)	0 (0)	0 (0)	0 (0)
Sum	10 (5.9)	9 (4.1)	12 (6.3)	13 (6.3)	27 (8.8)
<i>Debranching enzymes</i>					
GH51	5 (2.9)	8 (3.6)	3 (1.6)	5 (2.4)	13 (4.2)
GH54	1 (0.6)	3 (1.4)	0 (0)	0 (0)	0 (0)
GH62	0 (0)	0 (0)	1 (0.5)	0 (0)	0 (0)
GH67	2 (1.2)	1 (0.5)	0 (0)	0 (0)	2 (0.7)
GH78	1 (0.6)	2 (0.9)	2 (1.1)	10 (4.9)	15 (4.9)
Sum	9 (5.3)	14 (6.4)	6 (3.1)	15 (7.3)	30 (9.7)
<i>Endohemicellulases</i>					
GH8	0 (0)	0 (0.0)	1 (0.5)	0 (0)	2 (0.7)
GH10	2 (1.2)	6 (2.7)	1 (0.5)	5 (2.4)	7 (2.3)
GH11	0 (0.0)	0 (0.0)	0 (0.0)	0 (0.0)	1 (0.3)
GH12	0 (0.0)	1 (0.5)	0 (0.0)	0 (0)	0 (0)
GH26	4 (2.3)	4 (1.8)	0 (0)	1 (0.5)	12 (3.9)
GH28	0 (0.0)	1 (0.5)	4 (2.1)	2 (1.0)	8 (2.6)
GH53	0 (0.0)	3 (1.4)	3 (1.6)	2 (1.0)	4 (1.3)
Sum	6 (3.5)	15 (6.8)	9 (4.7)	10 (4.9)	34 (11.0)
<i>Oligosaccharide-degrading enzymes</i>					
GH1	9 (5.3)	10 (4.6)	6 (3.1)	14 (6.8)	20 (6.5)
GH2	5 (2.9)	0 (0.0)	4 (2.1)	4 (1.9)	15 (4.9)
GH3	5 (2.9)	9 (4.1)	6 (3.1)	9 (4.4)	18 (5.8)
GH29	5 (2.9)	4 (1.8)	4 (2.1)	11 (5.3)	13 (4.2)
GH35	2 (1.2)	0 (0.0)	0 (0.0)	0 (0.0)	3 (1.0)
GH38	0 (0.0)	0 (0.0)	1 (0.5)	3 (1.5)	4 (1.3)
GH39	2 (1.2)	5 (2.3)	1 (0.5)	2 (1.0)	7 (2.3)
GH42	3 (1.8)	2 (0.9)	2 (1.1)	6 (2.9)	11 (3.6)
GH43	8 (4.7)	7 (3.2)	5 (2.6)	8 (3.9)	10 (3.2)
GH52	0 (0.0)	0 (0.0)	0 (0.0)	2 (1.0)	0 (0.0)
Sum	39 (23)	37 (17)	29 (15)	59 (29)	101 (33)
<i>Phenolic compound-degrading enzymes</i>					
Laccase	1 (0.6)	8 (3.6)	7 (3.7)	11 (5.3)	12 (3.9)
Dioxygenase	2 (1.2)	8 (3.6)	3 (1.6)	2 (1.0)	5 (1.6)
Peroxidase	4 (2.3)	5 (2.3)	3 (1.6)	8 (3.9)	1 (0.3)
Sum	7 (4.1)	21 (9.6)	13 (6.8)	21 (10.2)	18 (5.8)
Tot. ass. seq.	17 067	21 980	19 104	20 581	30 834

The metagenomic profile is divided into five categories according to the functional role of the protein families in plant polymer degradation, a format previously applied in Allgaier *et al.* (2010) and Pope *et al.* (2010). Numbers reflect the number of significant hits (see Materials and methods for details). Shown in brackets are the numbers in $\%_{\text{total}}$ of the total number of sequences assigned to a PFAM. Tot. ass. seq., the number of sequences that were assigned to a PFAM (See Materials and methods for details). See Supplementary Table S4 for PFAM accession numbers and description of protein families.

We aimed to identify the microorganisms possessing the genes encoding these hydrolytic enzymes by using taxonomic binning with the MEGAN software (See Materials and methods for details). However, this type of taxonomic binning is problematic, owing to the unequal representation of reference genomes for many taxa in the public databases (Urlich *et al.*, 2008) and horizontal gene transfer. We therefore compared the community profiles from SSU ribo-tags with the taxonomically assigned metagenomic and metatranscriptomic sequences from the peat top layers using MEGAN (Figure 2a).

These profiles were similar for many taxa, suggesting that genes and transcripts can be used as taxonomic markers with some level of confidence. The genes assigned to the three major categories of polysaccharide degradation (cellulases, endohemicellulases and debranching enzymes; see Table 1) were taxonomically annotated. The bacterial phyla Bacteroidetes, Actinobacteria and Verrucomicrobia possessed the majority (>70%) of these genes (Figure 2b). The profiles of the two peat soils differed, with the Knudsenheia metagenomes having a higher fraction of genes from Bacteroidetes and

Table 2 Metagenomic and metatranscriptomic reads assigned to enzyme classes that catalyse key steps in anaerobic respiration and fermentation, and fermentative pathways

Pathway/function	E.C.	Gene	Metagenomes					Metatranscriptomes	
			S2a	S2b	K1a	K1b	K1c	S1a + S2a	K1a + K2a
Nitrate reduction	1.9.6.1	<i>napA</i>	1	6	7	4	1	0	0
Nitrate reduction	1.7.99.4	<i>narG</i>	10	6	14	19	28	0	0
Nitrite reduction	1.7.2.1	<i>nirK</i>	1	1	2	3	3	0	0
Nitrite reduction	1.7.2.1	<i>nirS</i>	0	1	0	2	1	0	2
Nitric oxide reduction	1.7.99.7	<i>norB</i>	1	4	3	2	4	0	1
Nitrous oxide reduction	1.7.99.6	<i>nosZ</i>	3	3	1	2	1	0	0
Sulphate reduction	1.8.99.3	<i>dsrA</i>	0	1	1	4	11	0	0
Fermentation: [FeFe] hydrogenase	1.12.7.2	<i>hydA</i>	2	9	0	7	8	0	0
Homoacetogenesis	6.3.4.3	<i>fthfs</i>	6	12	7	10	23	0	0
Methanogenesis	2.8.4.1	<i>mcrA</i>	0	0	1	0	2	0	0
Acetoin, butanediol metabolism			55	92	60	85	133	8	8
Acetone, butanol, ethanol synthesis			243	298	319	280	346	17	17
Acetyl-CoA fermentation to butyrate			310	362	406	329	407	16	22
Butanol biosynthesis			235	293	315	263	308	9	19
Fermentations: lactate			52	76	56	74	104	2	3
Fermentations: mixed acid			107	157	121	131	190	5	8
Tot. ass. sequences			69 176	63 527	85 664	54 889	72 166	6315	6205

First column shows function or main pathway, second column shows the enzyme class entry of the ExPASy bioinformatics resource, while third column shows the gene abbreviation. Merged biological replicates for the two sites Solvatn (S1a + S2a) and Knudsenheia (K1a + K2a) are listed under metatranscriptomes. Total ass. sequences, the number of sequences assigned a function by MEGAN analysis. Sequences assigned to pathways were annotated using (MG)-RAST (see Materials and methods for details).

Actinobacteria, while Solvatn had a higher fraction of genes assigned to Verrucomicrobia. Surprisingly, few genes were taxonomically assigned to fungi (Figure 2b). This was in accordance with the low abundance of fungal, and especially basidiomycotal ribo-tags (Figure 1). The latter have been reported as particularly potent decomposers of SOC (Waldrop *et al.*, 2010). Low fungal abundances in permafrost soils and the influence of plant community composition on fungal population size have been shown before (Zak and Kling, 2006; Yergeau *et al.*, 2010; Waldrop *et al.*, 2010).

Anaerobic respiration and fermentation

We investigated the genomic potential for and the expression of key enzymes in anaerobic respiration and fermentation, processes which drive the anaerobic decomposition of SOC. Especially the genes encoding dissimilatory nitrate reductases were abundant (Table 2). This suggests that denitrification is important, which is supported by the high nitrate concentrations detected (70 and 84 µg per gram soil in Solvatn and Knudsenheia, respectively; Alves, 2011). An important role of denitrification has also been suggested by recent metagenomic studies of permafrost affected soils (Yergeau *et al.*, 2010; Mackelprang *et al.*, 2011). Despite the high abundance of denitrification genes, only *nirS* (nitrite reductase) and *norB* (nitric oxide reductase) transcripts were identified (Table 2); however, the small size of the mRNA pools in the metatranscriptomic data sets prevented further statements. Also sulphate reduction might have a role, as genes of the key enzyme DSR were found and their abundance

increased with soil depth in Knudsenheia (Table 2). The majority of the identified anaerobic respiration genes were assigned to taxa within Actinobacteria and Proteobacteria (Supplementary Figure S5), although the genes were broadly distributed among several different taxa, as previously described for denitrifiers (Philippot and Hallin, 2005) and sulphate reducers (Wagner *et al.*, 2005). Genes and transcripts of several fermentative pathways were detected (Table 2), explaining the high concentration of fermentation products in the peat soils (Supplementary Table S1). Likewise, genes encoding the terminal hydrogenase of H₂-evolving fermentations (*hydA*) were found. Their abundance increased with depth, as did the abundance of formyltetrahydrofolate synthetase genes (*fhs*), which encodes the key enzyme of the Acetyl-CoA pathway of homoacetogenesis. The majority of *hydA* and *fhs* genes were assigned to Firmicutes and Actinobacteria, suggesting an important role of taxa within these phyla in the anaerobic degradation of SOC in the Svalbard peat (Supplementary Figure S6). Taken together, these findings indicate a gradual transition from aerobic to anaerobic degradation of SOC with depth.

Methane cycling

The fraction of methanogenic archaea increased with depth (Figure 1 and Supplementary Figure S6), reflecting the presumably more anoxic conditions and higher concentrations of fermentation intermediates (Supplementary Table S1). Three orders of methanogenic archaea were found, *Methanobacteriales*, *Methanomicrobiales* and *Methanosarcinales*,

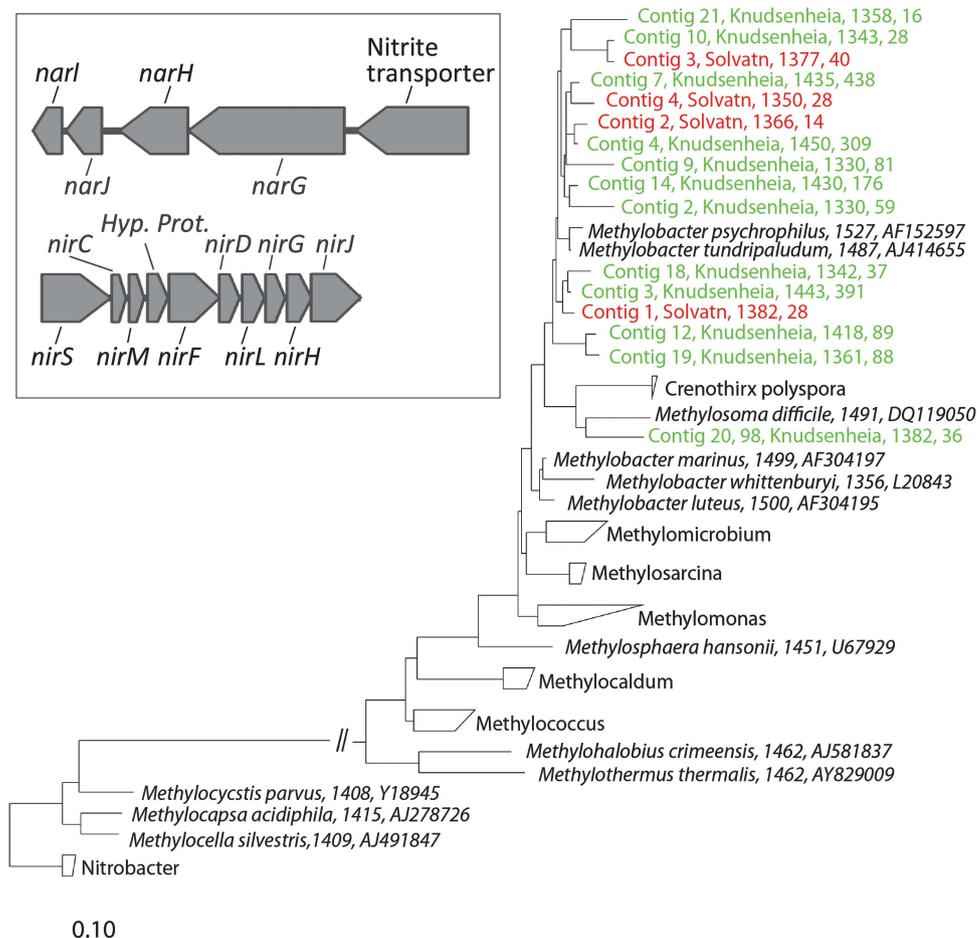


Figure 3 *M. tundripaludum* in Svalbard peat soils. The phylogenetic tree shows assembled SSU rRNA contigs of type I methanotrophs. Most of the nearly full-length ribo-contigs (14 out of 16) are >97% identical to *M. tundripaludum*. Ribo-contig description includes the following: contig id, site (Knudsenheia-green, Solvatn-red), contig length and the number of single ribo-tags that went into the assembly. The reference sequence description (black) includes sequence length and accession number. The length of the bar indicates 0.10 changes per nucleotide. The tree was constructed using the ARB software (See Materials and methods for details). The insert shows the nitrate reductase (open reading frames (ORFs) 3936–3940) and nitrite reductase operons (ORFs 3936–3940) identified in the genome of *M. tundripaludum*.

Table 3 mRNAs assigned to *M. tundripaludum* in the Knudsenheia metatranscriptome

Annotation	AC	Function	No. of reads
<i>pmoC</i>	ZP_07656564.1	Methane oxidation	8
<i>pmoA</i>	ZP_07656563.1	Methane oxidation	7
<i>pmoB</i>	ZP_07656562.1	Methane oxidation	6
CsbD family protein	ZP_07656622.1	Stress response	2
MgtE intracellular region	ZP_07654467.1	Magnesium transport	2
Transport-associated protein	ZP_07656600.1	Transport	2
Methanol dehydrogenase	ZP_07652313.1	Methanol oxidation	1
6-Phospho 3-hex.	ZP_07653633.1	RuMP pathway step 2	1
Succinate dehydrogenase	ZP_07652392.1	Electron transport Complex II	1
NADH dehydrogenase	ZP_07653777.1	Electron transport Complex I	1
NADH:ubiquinone oxidoreductase	ZP_07655325.1	Electron transport Complex I	1
Elec. trans. prot. SenC	ZP_07652162.1	Biogenesis of respiratory systems	1
Nitrogenase iron protein	ZP_07654124.1	Nitrogen fixation	1

Abbreviations: elec. trans. prot. SenC, electron transport protein SenC; mRNA, messenger RNA; RuMP, ribulose monophosphate. Functionally annotated transcripts related to energy, carbon and nitrogen metabolism are shown. mRNA sequences were taxonomically assigned to *M. tundripaludum* by MEGAN (see Materials and methods for details).

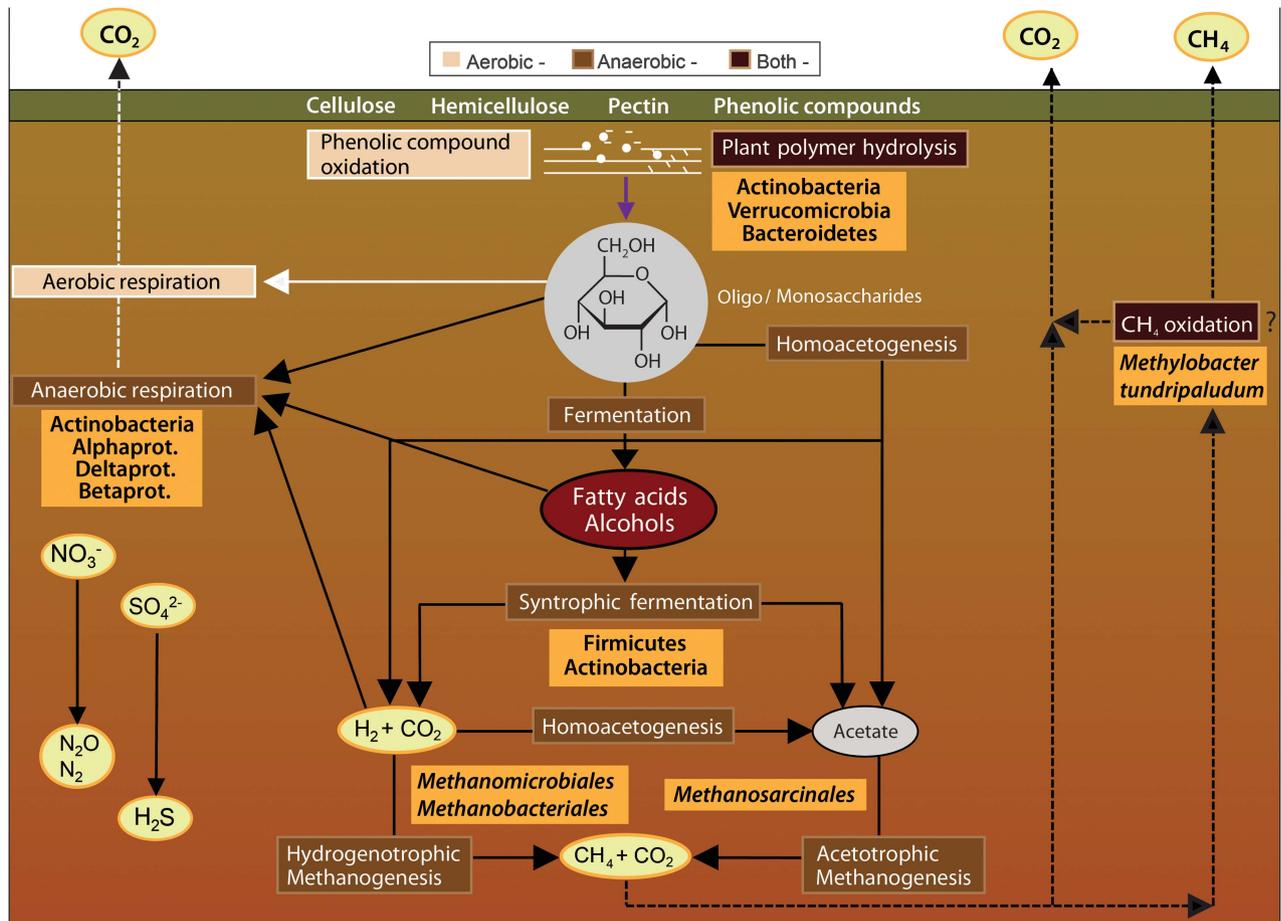


Figure 4 Schematic overview of the main degradation pathways of plant polymers in the high-Arctic peatlands of Svalbard. The pathways are divided into three categories; aerobic (beige), anaerobic (light brown) and processes occurring under both conditions (dark brown). Key microbial taxa for the different degradation steps are presented (orange boxes). The figure is adapted from Figure 1 in (Conrad, 1999).

of which the first two are hydrogenotrophic, producing CH₄ from H₂ and CO₂, whereas the *Methanosarcinales* are metabolically more versatile, carrying out hydrogenotrophic, acetoclastic and methylotrophic methanogenesis. These findings are similar to previous studies of methanogens from Solvatn (Høj *et al.*, 2005; Høj *et al.*, 2006). In Solvatn, the increased abundance of acetotrophic methanogens with depth correlated with a decrease in the concentration of acetate, indicating that the acetotrophic *Methanosarcinales* acted as the sink for acetate. In general, the high concentrations of fermentation products like ethanol, especially in the lower layers of Knudsenheia, indicated a low efficiency in the terminal fermentation steps, possibly owing to low activity of, or a lack of, established interactions between fermentative syntrophic bacteria and their methanogenic counterparts.

Type I methanotrophs of the order *Methylococcales* were abundant in both peat soils (0.1%–2.4% of ribo-tags). We did not detect active type II methanotrophs, anaerobic methanotrophic archaea (ANME), the recently described anaerobic methanotroph *Candidatus Methyloirabilis oxyfera* of the

NC10 candidate phylum (Ettwig *et al.*, 2010) or the Verrucomicrobium *Methylacidiphilium* (Islam *et al.*, 2008). We assembled long, sometimes full-length, SSU rRNAs from the ribo-tags of the type I methanotrophs to determine the phylogenetic position (See Materials and methods (Urich *et al.*, 2008; Radax *et al.*, 2012)). The vast majority of the assembled ribo-contigs was closely related to *M. tundripaludum* (Figure 3), an arctic methanotroph originally isolated from Solvatn (Wartiainen *et al.*, 2006). We could taxonomically assign 104 mRNA tags to genes encoded in the genome of *M. tundripaludum* (Svenning *et al.*, 2011). Although this is a small number, it allowed a superficial view into the gene expression and metabolism of *M. tundripaludum in situ*. Remarkably, as much as 33% of the transcripts encoded subunits of the key enzyme particulate methane monooxygenase (Table 3). This highlights the importance of particulate methane monooxygenase for *M. tundripaludum* cells, as this enzyme catalyses the initial step of both the carbon and energy metabolism through the oxidation of methane to methanol. Transcripts of the methanol dehydrogenase that catalyses the

second step of complete oxidation of methane to CO₂ and the ribulose monophosphate pathway for carbon assimilation, as well as transcripts encoding enzymes of the aerobic respiratory chain, were also identified (Table 3). Recently, *M. tundripaludum* was shown to be the dominating methanotroph in a SIP-RNA experiment with Solvatn peat (Graef *et al.*, 2011). It was also shown to be the dominant methanotroph in two soils from the Siberian and the Canadian Arctic (Liebner *et al.*, 2009; Martineau *et al.*, 2010). Although these are few studies, they point to the importance of methanotrophs closely related to *M. tundripaludum* having a key role in the biological CH₄ filter in Arctic soils. The seemingly ecological advantage of *M. tundripaludum* in these often water-logged Arctic soils might be owing to its N₂-fixation capability (Wartiainen *et al.*, 2006; Svenning *et al.*, 2011), as well as to the presence of nitrate reductase (*nar*) and nitrite reductase (*nir*) operons in its genome (Figure 3) (Svenning *et al.*, 2011). It is tempting to speculate that *M. tundripaludum* could employ the NAR and NIR enzymes in denitrifying anaerobic methane oxidation, utilising a pathway similar to that of *Candidatus M. oxyfera* (Ettwig *et al.*, 2010). Alternatively, it could have the ability to grow anaerobically on simple organic compounds using denitrification. Experiments are underway to test these hypotheses.

Conclusion

We have described the metabolic potential and activity of microorganisms involved in SOC degradation in high-Arctic peat soils (Figure 4). This first combined metagenomic and metatranscriptomic study on any soil (to our knowledge), revealed taxa that are important to specific processes such as hydrolysis of plant polymers (for example, Verrucomicrobia and Bacteroidetes), fermentations (for example, Firmicutes), methanogenesis and methanotrophy. Actinobacteria seemed to be particularly important, having a metabolic potential for carrying out several of the key steps in SOC degradation (Figure 4). Based on their relative abundance, fungi appeared to not have an important role. However, as warming will lead to an extended growing season, this might result in a change of the plant covers from mosses to vascular plants and in more oxygenated active layers, where fungi might become more important for SOC decomposition. Drier and more oxygenated active layers will then likely lead to increased peat decomposition, owing to lowered concentrations of phenolic substances, caused by increased activity of aerobic microorganisms that synthesise phenol oxidases (Fenner and Freeman, 2011). These combined effects might increase the rate of SOC degradation above that of carbon sequestration, and turn these peat soils into CO₂ sources. Increased active layer depth and higher soil temperatures will likely result in bigger habitat size

for anaerobic bacteria and methanogenic archaea, in the cases where the peat soils remain water-saturated. Here, CH₄ production likely will increase, although this might well be balanced by methanotrophic bacteria, of which *M. tundripaludum* seems to be particularly important. However, to which extent such key populations can adapt to the changing conditions remains a crucial question.

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