Long-term legacy of phytoremediation on plant succession and soil microbial communities in petroleum-contaminated sub-Arctic soils

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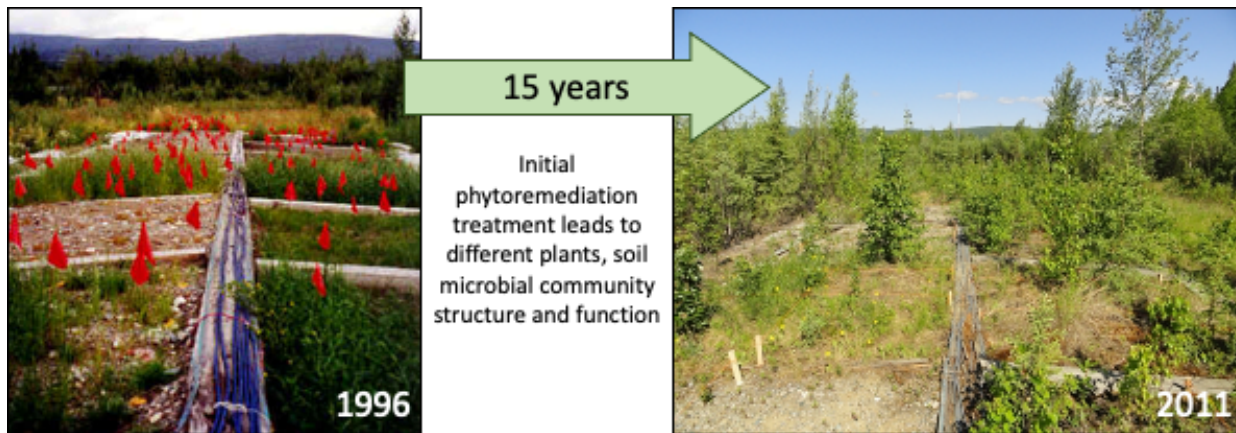
Abstract. Phytoremediation can be a cost-effective method of restoring contaminated soils using plants and associated microorganisms. Most studies follow the impacts of phytoremediation solely across the treatment period and have not explored long-term ecological effects. In 1995, a phytoremediation study was initiated near Fairbanks, Alaska, to determine how the introduction of annual grasses and/or fertilizer would influence degradation of petroleum hydrocarbons (PHCs). After one year, grass and/or fertilizer treated soils showed greater decreases in PHC concentrations compared to untreated plots. The site was then left for 15 years with no active site management. In 2011, we re-examined the site to explore the legacy of phytoremediation on contaminant disappearance, as well as plant and soil microbial ecology. We found that the recruited vegetation, along with current bulk soil microbial community structure and function were all heavily influenced by initial phytoremediation treatment. The number of diesel-degrading microorganisms (DDM) was positively correlated with percent cover of vegetation on the site, which was influenced by initial treatment. Even 15 years later, the initial use of fertilizer had significant effects on microbial biomass and microbial community structure activities. We conclude that phytoremediation treatment has long-term, legacy effects on the plant community, which, in turn, impacts microbial community structure and function. It is therefore important to consider phytoremediation strategies that not only influence site remediation rates in the short-term, but that also prime the site for restoration of vegetation across the long-term.

Graphical Abstract. Photos of study site in 1996 (left, photo credit: C.M. Reynolds) and in 2011 (right, photo credit: authors) showing changes in plant communities in the different soil types and phytoremediation treatment plots.

1 Introduction

There is a worldwide legacy of contamination from operational and abandoned mining and industrial sites, waste disposal sites, abandoned and active military installations, and other anthropogenic activities (Stow *et al.*, 2005; Pier *et al.*, 2002). In particular, petroleum hydrocarbons (PHCs) are widespread in the environment as a consequence of their broad use in society, posing serious human health hazards. In the state of Alaska, there are nearly 2,000 documented sites with contaminated soil, many of which contain petroleum (Alaska Department of Environmental Conservation Spill Prevention and Response, 2015). The harsh climate and remote location of many communities at high latitudes pose unique challenges for contaminant remediation. Lack of road access can make traditional physical remediation methods difficult, cost-prohibitive, or even impossible (Kuiper et al., 2004). Bioremediation offers a potentially affordable and feasible solution for contaminant clean-up in remote communities which are especially dependent on petroleum products for heat and transportation. Soil amendments, such as fertilizer or plants, can be used to stimulate the degradation of contaminants by indigenous soil microorganisms and is often more cost effective in remote areas (Wenzel, 2009; Aken et al., 2010; Leewis et al., 2016b). The biodegradation of organic pollutants in plant-associated soils is hypothesized to be driven by the release of secondary plant metabolites (SPMEs) (Leewis et al., 2016a; Musilova et al., 2016). This is because soil bacterial communities which are plant-associated are able to use or transform SPMEs, such as phenols, through diverse groups of broadly specific enzymes such as aromatic ring-hydroxylating dioxygenases (Zubrova et al., 2021). These enzymes have also been implicated in the transformation and degradation of organic contaminants, such as PHCs, which are structurally similar to SPMEs (Leigh et al., 2002; Singer et al., 2003; Musilova et al., 2016; Zubrova et al., 2021).

The microbial and vegetative processes associated with biologically-based remediation strategies, i.e. phytoremediation and rhizoremediation, have been a focus of study for decades (Fletcher and Hegde, 1995; Siciliano and Germida, 1998; Palmroth et al., 2002; Gerhardt et al., 2009; Macek et al., 2009; Glick, 2010). In many cases, remediation treatment results in decreases in soil contaminant concentrations and changes in the soil microbial community (Liu et al., 2015; Yergeau et al., 2015; Bell et al., 2013; Palmroth et al., 2005; Leewis et al., 2016a; Lopez-Echartea et al., 2020). Many of these studies, however, focus on the short-term effects of treatment on microbial and vegetative communities and contaminant

transformation. Monitoring the long-term effects of contamination and subsequent treatment on soils, beyond the first few

65   years of active site management or experimentation, is important to assessing the long-term impacts on ecological processes

such as plant succession and microbial community function (Susarla et al., 2002; Mukherjee et al., 2014). Few studies have

investigated the long-term effects of remediation treatments on PHC-contaminated soils; most focused on the effects of land-

farming on contaminant concentrations (Loehr and Webster, 1996; Mills et al., 2006) without determining vegetation

recruitment or effects of treatment on microbial communities. Other studies have explored the long-term effects of treatment

70   and soil chemistry on microbial communities (Sutton et al., 2013; Liang et al., 2016), yet without determining the effects of

treatment on vegetation recruitment. While these studies provide insight to the disappearance and biodegradation of legacy

PHC contamination, even these comprehensive investigations stop short of fully characterizing the *in-situ* degradation and

plant-associated changes in microbial communities that would be necessary for an integrative understanding of the effects of

different remediation strategies on site ecology. More long-term studies of ecosystem recovery are essential to achieving an

75   integrated understanding of how plant communities, remediation treatment, contaminant concentration, and microbial

populations interact as sites change over time.

In a series of studies, we have sought to further understand the long-term (over-15-year) relationships between original

phytoremediation treatment, colonizing vegetation, soil microbial communities, and the degradation of petroleum

contaminants at an experimental site in interior Alaska (Leewis et al., 2013; Reynolds et al., 1997a; Reynolds and Koenen,

80   1997). The original phytoremediation experiment by Reynolds and colleagues was designed to investigate the effects of

combinations of planting and fertilizer treatments on the remediation of soils contaminated with either crude oil or diesel fuel

(Reynolds et al., 1997a; Reynolds and Koenen, 1997; Reynolds et al., 1999). In Leewis *et al*. (2013), we described changes in

the soil chemistry, plant community composition, total petroleum hydrocarbon (TPH) concentrations, and bacterial community

as determined by a molecular fingerprinting technique (i.e., terminal restriction fragment length polymorphism - T-RFLP)

85   after 15 years without active site management. Papik et al., 2023 further explores the legacy of original treatment on the

bacterial rhizosphere and endosphere communities (Papik et al., 2023). In the current study, we build upon this previous work

and aim to i) evaluate the interactive effects of petroleum contamination and phytoremediation on site recovery over time; and

ii) identify the factors that drive vegetation community structure, soil microbial community structure and function, and TPH

loss. Here, we report additional analyses of the same soil samples from Leewis *et al*. (2013) to further characterize the microbial

community using: A) culture-based techniques to determine the ability of the soil community to use TPH as a carbon source

(MPNs), B) phospholipid fatty acids (PLFAs) to fingerprint the entire soil community (including bacteria, fungi, and protozoa),

and C) 16S rRNA gene sequencing to determine the identity of bulk soil bacterial community members. We additionally

incorporated the newly analysed plant percent cover data, and previously reported soil chemistry, petroleum hydrocarbon, and

plant community count data into statistical analyses of the microbial community.

2 Methods

2.1 Legacy Field Study Background and Site Description

In 1995, a phytoremediation field study was initiated by the Army Corps of Engineers Cold Regions Research and

Engineering Laboratory (ACE CRREL) located in Fairbanks, Alaska (Reynolds et al., 1997b, 1999; Reynolds and Koenen,

1997). The study sought to compare the effects of different phytoremediation treatments on two soil and contamination types:

A) crude oil-contaminated soils collected from a gravel pad, or B) diesel-contaminated soils collected during the removal of

an underground storage tank. The experiment was initiated at the ACE CRREL Farmers Loop Facility, where soils were placed

in piles in adjacent lined areas. Each of the separate soil piles was then divided into individual plots (Fig. 1) which represented

different phytoremediation strategies. The original phytoremediation treatment plots were; two plots not planted or fertilized

(c1 & c2), one plot spread with fertilizer (f), one plot planted with annual ryegrass (*Lolium multiflorum*; p1), one plot planted

with a mixture of annual ryegrass and Arctared fescue (*Festuca rubra*; 1:1 mixture, p2), one plot fertilized and planted with

annual ryegrass (p1f), and one plot fertilized and planted with both ryegrass and fescue (p2f; Fig. 1C). Fertilizer was applied

to soil surfaces by hand at approximately 620 g/m^2 of N, P, and K using a commercially available fertilizer (granular 20-20-

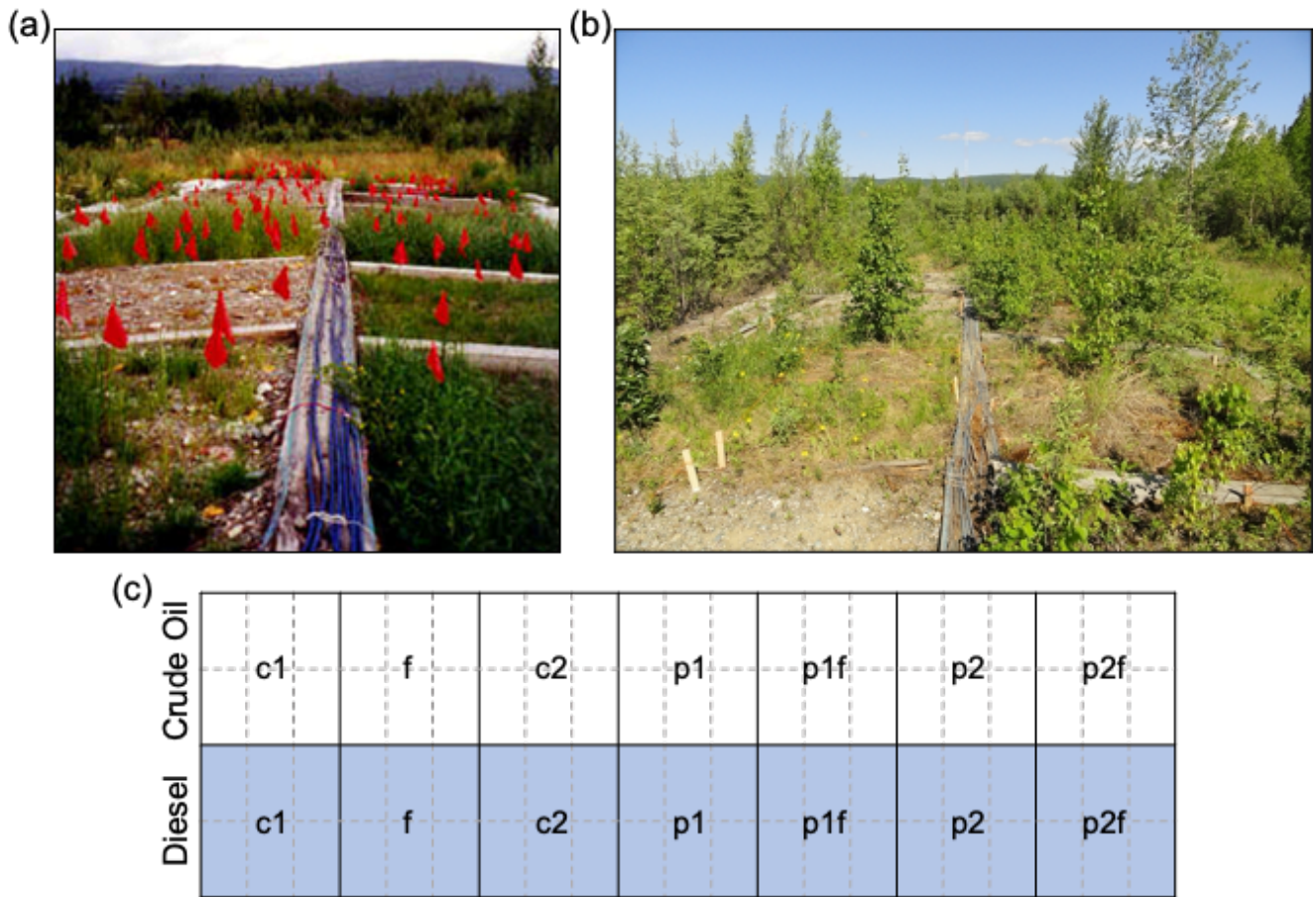
10). Seeds were spread by hand at approximately 10.8 g/m^2.

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Figure 1. Overview of Site. Photos of study site in (A) 1996 (photo credit: C.M. Reynolds) and (B) in 2011 showing plant colonization of the different soil types and phytoremediation treatment plots. C) Overview of the original phytoremediation treatment plots (outlined in thick black lines). Half of the plots are soils from a gravel pad originally contaminated with crude oil (white background) and half are soils from near a leaking diesel fuel storage tank contaminated (grey background). The original treatments applied in 1995 were as follows: no treatment (c1, c2),

115 **planted with annual ryegrass (p1), a mix of annual ryegrass and Arctared fescue (p2), treated with fertilizer (f), and/or no added nutrients (no "f" indicated). For this follow-up study, original plots were each subdivided into six subsections (outlined in dashed grey lines) to allow for pseudo-replication.**

2.2 Sample Collection and Vegetation Characterization

In June 2011, we initiated a follow-up study of the initial phytoremediation experiment to determine the continuing influence of original treatment on current TPH concentrations, microbial community presence and abundance, and plant community composition. Due to the lack of replication in the original study design, we divided each of the original treatment plots into six 1 x 1.5 m^2 "subsections" (Fig. 1C). From each subsection, we collected soil samples into both sterile plastic zip-top bags for microbial analyses and a glass jar with Teflon lined lids for TPH analysis. Samples were sieved through a sterile 2.5 mm sieve and stored at A) -80 °C for molecular and TPH analysis or, B) at 4 °C for culture-based microbial assays (i.e., most-probable number analyses - MPN) which took place within one week of collection. Due to the number of subsamples and to reduce the overall costs of analyses, we conducted analyses on the following number of subsections per plot: A) culturable diesel degrading microorganisms (DDM) – six subsections, B) PLFAs – four subsections, C) bacterial community structure and TPH – three subsections. Reported values are averages with standard deviations.

At the time of soil sampling, we also photographed each individual subsection to estimate percent cover of vegetation. Vegetation coverage was classified into four "types": grasses (including live and dead), forbs (herbaceous flowering plants; for example, fireweed, dandelions, vetch, *Oxytropus*, etc.), trees, and bare ground (including crusts and moss cover). Each photo was overlain with a 5 x 5 grid, and if an individual block within the grid was covered more than 50% with a vegetation type, then that block was counted as positive for that vegetation type (Vanha-Majamaa et al., 2000; Chen et al., 2010).

2.3 Phospholipid Fatty Acid (PLFA) Assay

PLFA extractions and analyses were performed following the methods of (Wu et al., 2009) with minor modifications as follows. Each soil sample (ca. 4.8 g) was freeze-dried and the weight after freeze-drying was recorded for data analysis. All drying steps were performed with N$_2$ at a temperature below 40 °C. Phospholipids were extracted using 8 mL methanol (HPLC grade, Fischer Scientific, Pittsburgh, PA). Fatty acid methyl esters were extracted using hexane (3 x 1 mL) and then were dried under N$_2$ (HLPC-grade hexane, EM, (Germany). Prior to GC analysis, 80 µL of the internal standard, methyl nonadecanoate (0.737 nmol/µL; Sigma-Aldrich, Switzerland), was added and the samples were dried completely. The samples were then dissolved in 80 µL hexane and analyzed on an Agilent 6890 N Gas Chromatograph, using an Agilent 19091B-102 (25.0m x 200µm x 0.33µm) capillary column, and MIDI peak identification software using an internal standard (Version, 6.1; MIDI

145   Inc., Newark, DE). PLFA biomarkers and biomass were assigned to microbial groupings as outlined in Table S1 (Song et al.,

2008; Kaiser et al., 2010; Ngosong et al., 2012; Stella et al., 2015; Oburger et al., 2016; Polivkova et al., 2018).

2.4 Enumeration of Diesel-Degrading Microorganisms

    The abundance of culturable aerobic DDM was determined by a 96-well plate most-probable-number (MPN) method

adapted from (Haines et al., 1996), following the modifications detailed in Leewis *et al.*, (2016a). Soils were stored at 4 °C

150   and processed within a week of collection. Briefly, microorganisms were extracted from triplicate soil samples in a solution

of 1% w/v sodium pyrophosphate (Fisher Scientific, USA) with 3 g of sterile glass beads by shaking horizontally for 1 h,

followed by 30 min of settling. This suspension was used to create a 10-fold serial dilution in Bushnell-Haaas medium for

each sample in triplicate, along with positive and negative controls. After inoculation, filter-sterilized diesel fuel oil #2 was

added to each test well as the sole-carbon source. Plates were then incubated in the dark at room temperature for 14 days, after

155   which we added a sterile indicator dye (3 g·L^{-1} p-iodonitrotetrazolium; MP Biomedicals, USA) to each well to indicate the

presence of active respiring microorganisms. After 48 hours, wells were scored according to the presence of a red or pink

colour and MPN values calculated using previously published MPN tables (Man, 1983).

2.5 Amplicon Preparation and Sequencing

    Soil DNA was extracted from a 0.5 g soil subsample using the FastDNA SPIN kit for soil (MP Biomedicals, Ohio,

160   USA) following the manufacturer's instructions. DNA was eluted into 50 µL of water and stored at -20 °C until analysis. DNA

concentrations were evaluated by measuring absorbance at 260 and 280 nm using a NanoDrop ND-1000 Spectrophotometer

(Thermo Fischer Scientific, USA). Region V4-V5 of the 16S rRNA gene were amplified using primers 563-577F, 5'-

AYTGGGYDTAAAGNG-3', and 926-909R, 5'-CCGTCAATTCMTTTRAGT-3' (Uhlik et al., 2012; Cole et al., 2009). Each

of the primers was synthesized with sequencing adaptors (454 Sequencing Application Brief No. 001-2009, Roche), the

165   forward primer was also modified with different barcodes (454 Sequencing Technical Bulletin No. 005-2009, Roche) to allow

for multiple samples to be pooled for sequencing. The PCR mixture (final volume, 25 µl) contained 1 µl each primer (10 µM),

0.5 µl dNTP mix (10 mM), 2.5 µl FastStart 10 X Buffer #2, 0.25 µl FastStart HiFi Polymerase (5 U/µl), and 18.75 µl molecular

biology grade water (Roche). The following thermal cycling scheme was used: initial denaturation at 95 °C for 3 min and 30

cycles of denaturation at 95 °C for 30 sec, annealing for 1 min at 55 °C, and extension at 72 °C for 1.5 min, followed by a final extension period at 72 °C for 10 min. Each PCR product was obtained in three parallel reactions, the resulting preparations were mixed, purified using Pure-Link PCR purification kit (Invitrogen, USA) and pooled for downstream sequencing. Roche 454 GS FLX Titanium sequencing (454 Life Sciences, USA) was performed on pooled reactions at the Utah State University Center for Integrated Biosystems, Logan, Utah, USA. It should be noted that although older technologies, such as 454-pyrosequencing, are known to suffer from drawbacks such as limited read depth and higher error rates in repeated complex sequences (e.g. homopolymer regions), they can still provide a comparable and comprehensive overview of environmental communities and biodegradation processes as assessed by other sequencing technologies, including Illumia MiSeq (Lauber et al., 2009; Bell et al., 2011; Luo et al., 2012; Goordial et al., 2016; Akaçin et al., 2022).

2.6 Sequence Analyses

Raw pyrosequencing data (*.sff files) were processed using the mothur software package version 1.45.3 and following the standardized operating procedure (Schloss et al., 2011), accessed July 2021, with minor modifications as described previously (Uhlik et al., 2012). Sequences with > 3% sequence similarity were grouped into the same operational taxonomic units (OTUs). These were then classified by mothur-implemented RDP reference files (trainset18_062020; accessed July 2021) (Wang et al., 2007; Cole et al., 2014) and sequences associated with mitochondria, chloroplasts, archaea, eukaryote, and unknown Kingdoms were removed. Data were then rarefied to the minimum library size of 3537 sequences. Pyrosequencing reads were deposited in NCBI Short Read Archive under bioproject number PRJNA950456. All coding for bioinformatics analyses is available at github.com/mcleewis/FarmersLoop.

2.7 Statistical Analyses

All statistical analyses were performed in R version 3.6.1 (R Core, 2018). To determine difference in the means of variables, we used a Kruskal-Wallace (KW) rank sum test with original treatment as the factors. We then used a pairwise Wilcox test with no assumption of equal variance and corrected for multiple comparisons using a Benjamini–Hochberg (BH) correction to further test differences between individual original treatments. Kendall's tau rank correlation tests (τ) were used to examine relationships between independent variables.

Alpha and beta diversity metrics were calculated using the phyloseq package (McMurdie and Holmes, 2013) on rarefied bacterial sequence data and tested using the vegan package (Oksanen et al., 2017). The influence of treatment and soil properties on bacterial communities was visualized using non-metric multidimensional scaling (NMDS, *metaMDS()*) with vector fitting of associated soil and plant data (*envfit()*) and tested using a permutational multivariate analysis of variance (PERMANOVA, *adonis()*). For multivariate analyses, PFLA data were not aggregated as in Table S1, but treated as individual biomarkers. Strongly autocorrelated soil chemical data was removed from the vector fitting prior to analysis (*cor()*, Pearson's r > 0.70), P-values were corrected using the *p.adjust* function (*method = "BH"*). The relationship between microbial community structure and measured soil properties was assessed using a Mantel test (*mantel()*). Multi-response permutation procedure (MRPP, *mrpp()*) was used to test the influence of original treatment or soil type on microbial community structure. Multivariate testing of the microbial community was conducted using the Bray-Curtis dissimilarity measure (*distance = "bray"*) and significance quantified using permutations tests (*permutations = 999*). Major OTUs representing more than 2% of the total communities were depicted in a heat map using the *heatmap2* function. Differential sequence analsyis was conducted using the DESeq2 (Love et al., 2014) R package, with p-values adjusted for multiple comparison with Benjamini–Hochberg (BH) correction. All coding for statistical analyses is available at github.com/mcleewis/FarmersLoop.

3. Results

3.1 Prior Findings on TPH, Soil Properties, and Plant Communities

The results detailing soil chemistry and TPH concentrations are previously described in detail in (Leewis et al., 2013); we briefly summarize those findings here (Table S2, S3). The initial TPH concentrations measured in crude oil-contaminated soils was 6,070 ppm, and 8,350 ppm in diesel-contaminated soils (Reynolds et al., 1997b, a). In 1996, after almost a year of treatment, crude oil-contaminated soils were 27% - 47% lower than initially measured concentrations. In diesel-contaminated soils, soil TPH concentrations were 65% - 87% lower than initial measurements (Table S2).

After 238 days of treatment, TPH concentrations in soils treated with plants or fertilizer (p1, f, p2) and soils without any amendment (c1, c2) decreased relative to the initial (day 0) TPH measurements (Table S2). Soils treated with both plants and fertilizer (p1f, p2f) had significantly lower TPH concentrations than the untreated soils (c1, c2) after 238 days of treatment

(Reynolds et al., 1997a; Reynolds and Koenen, 1997; Reynolds et al., 1999). After completion of the initial phytoremediation project in 1996, the site was not actively managed for approximately 15 years, with no nutrient or re-seeding amendments made. As previously reported, we found that increased TPH disappearance was associated with increased numbers of native trees and shrubs such as willow (*Salix* spp.), birch (*B. neoalaskana*), white spruce (*S. glauca*), and balsam poplar (*Populus balsamifera*), regardless of the original phytoremediation treatment (Leewis et al., 2013).

3.2 TPH and Soil Chemistry

In 2011, soil TPH concentrations were 80 - 95% lower than the concentrations measured in 1996 (Table S2). Concentrations of remaining TPH were higher in crude oil-contaminated soils (ca. 631 – 760 ppm) compared to diesel-contaminated soils (ca. 319 – 430 ppm). Differences in lingering TPH concentrations may be associated with the original soil contaminant, as crude oil contains a more recalcitrant compounds than refined diesel fuels, or the soil quality itself, as the gravelly crude oil-contaminated soils were more nutrient poor compared to diesel-contaminated soils (Table S3). However, after 15 years of treatment no one treatment resulted in significantly more reduction of TPH concentrations ($P > 0.05$; Table S2).

Unfortunately, the initial 1995-1996 investigation did not include measurements of soil chemistry or texture; those data are only available for soils collected in 2011 (Table S3). Overall, diesel-contaminated soils were significantly higher in total carbon, NO_3-N, and K, with finer soil texture than crude oil-contaminated soils (KW $P < 0.01$). Soils which had been originally fertilized remained significantly higher in total P and K compared to unfertilized soils (KW $P < 0.001$), regardless of contaminant type.

3.3 Long-term Responses of Plant Community to Phytoremediation Treatment

After 15 years with no active site management, the plant community within each treatment plot changed significantly from the first establishment of the site in 1995, and none of the originally planted grasses were present at the time of sample collection (Leewis et al., 2013). Plant communities differed according to soil type (MRPP A = 0.175, *P* = 0.001): in the finer-textured diesel-contaminated soils, there were more, and larger, woody plants (e.g., *Salix* spp., *Populus* spp., *Betula neoalaskana*, and *Picea glauca*) present when compared to the crude oil-contaminated soils (Table S4, Fig. S1). The crude oil

plots, with soils that originated from a gravel pad, were more heavily colonized by non-native plants from the family *Fabaceae* such as *Taraxacum officinale* (dandelion), *Vicia cracca* (common vetch), and *Trifolium hydridum* (clover). Native plants including *Ledum decumbens* (Labrador Tea), *Epilobium angustifolium* (fireweed), along with willow (*Salix* spp.), spruce (*Picea* spp.), and poplar seedlings (*Populus* spp.) were present throughout the site, regardless of soil type (i.e., crude oil-contaminated gravelly soils or diesel-contaminated fine-textured soils). Further exploration found that original phytoremediation treatment explained 63% of the variation in the current plant community structure (PERMANOVA CO: P = 0.001, R^2 = 0.630, F = 3.978; DE: P = 0.001, R^2 = 0.635, F = 4.077). In both soil types, plots originally fertilized had greater coverage with grasses and less bare ground than other plots, on average 53% and 38% of area covered by grasses in crude oil and diesel plots respectively (Fig. 2, S1). Crude oil-contaminated soils had more bare ground than diesel-contaminated soils (39% and 13% respectively). In crude oil-contaminated soils, originally untreated plots had an average of 44% bare ground, and those initially planted 79% bare ground. Plots not originally fertilized had more coverage of trees compared to other plots, on average 16% in crude oil and 66% with tree coverage in diesel contaminated plots not originally fertilized (Fig. 2, S2).
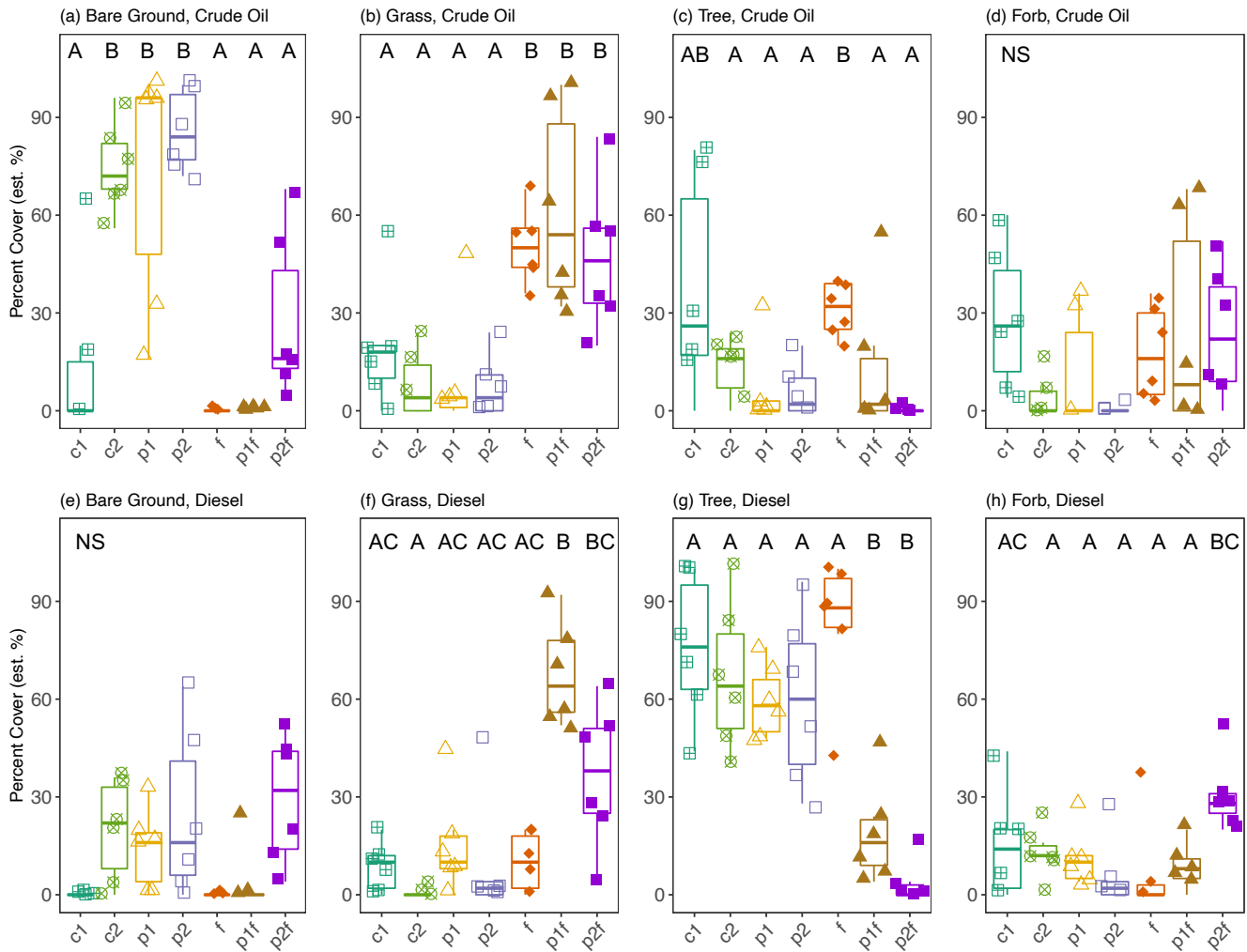
Figure 2. Estimated percent cover of vegetation groups in either Crude Oil (dark grey) or Diesel (light grey) plots. Measurements were based on visual estimates in each of the six sub-plots. The values shown are means with 95% confidence intervals (N = 6), note y-axes have different scales. Treatments indicated are as follows: no treatment (c1, c2), planted with annual ryegrass (p1), a mix of annual ryegrass and Arctared fescue (p2), treated with fertilizer (f), and/or no added nutrients (no "f" indicated). Significant differences in percent cover are indicated by different letters, "NS" indicates no significant differences found.

3.4 Quantity of Diesel-Degrading Microorganisms is associated with TPH, Vegetation

To assess the ability of the soil community to use petroleum hydrocarbons (PHC) as a carbon source, we measured the MPN of culturable DDMs across all the treatment plots (Fig. 3A). Overall, there were significantly more DDM in the crude oil-contaminated, gravely soil (average 5.57 ± 0.71 log MPN / g soil) than diesel-contaminated soils (4.31 ± 0.77 log MPN / g soil; $P < 0.05$ CO vs DE; Fig. 3A). We also found differences in the amount of DDM associated with the original phytoremediation treatment: in the crude oil soils, an initially untreated plot (6.34 ± 0.24 log MPN / g soil) and the fertilizer-only plot (6.43 ± 0.61 log MPN / g soil) had the highest numbers of DDM ($P < 0.04$, Kruskal-Wallis). In diesel soils, one initially untreated plot had significantly higher numbers of DDM (5.35 ± 0.81 log MPN / g soil) than the other treated plots ($P < 0.04$, Kruskal-Wallis).

We next examined the relationship between DDM and other factors on the site at the time of collection, including TPH concentrations and current vegetation. There was a weak trend between counts of DDM and TPH concentrations in both soil types in which increased counts associated with decreased TPH concentrations, although this relationship was not significant in either crude oil or diesel-contaminated soils (Fig. S2A, S2B). In both crude oil and diesel soils, there was a positive relationship between DDM and the percent of vegetation coverage in all plots (i.e. no bare ground), although this relationship was only significant in crude oil-contaminated soils (CO: Kendall's $\tau = 0.3512$, $P = 0.0025$; DE: Kendall's $\tau = -0.0367$, $P = 0.7623$; Fig. S3A, 3B). We further explored this plant-contaminant relationship and found a significant positive relationship between DDM and the percent cover of trees in the crude oil plots (Kendall's $\tau = 0.3550$, $P = 0.0026$; Fig. S2C), but not in the diesel plots (Fig. S2D). In crude oil plots there was also significant relationship between the amount of bare ground and lower DDM (Kendall's $\tau = -0.3512$, $P = 0.0025$), but no similar relationship was found in the diesel-contaminated plots (Kendall's $\tau = 0.0367$, $P = 0.7623$). There was no statistically significant relationship found between percent coverage of grasses or forbs and DDM in either contamination type ($P > 0.06$).

(a) DDM

(b) Total Microbial Biomass
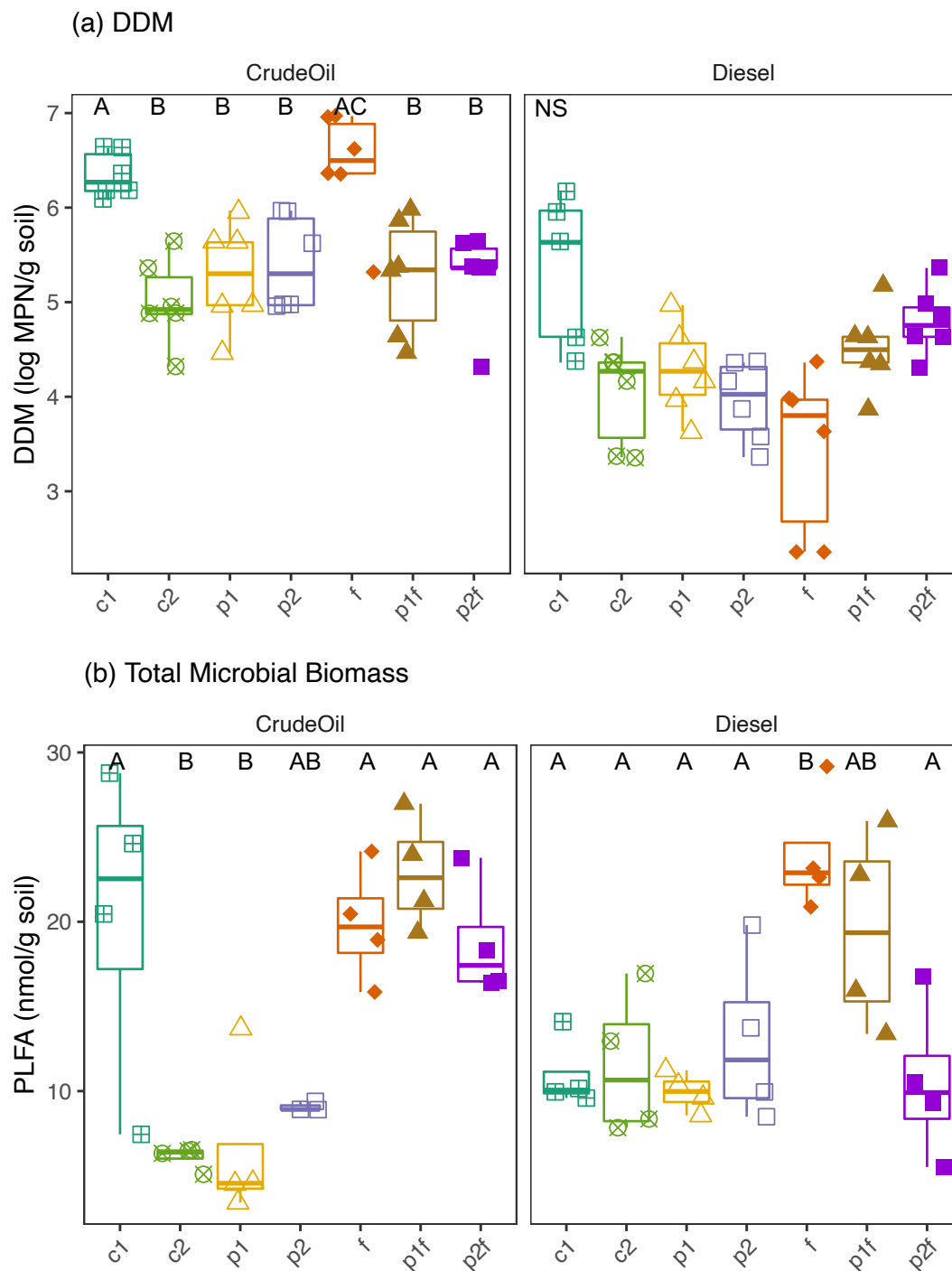
Figure 3. Measures of A) Most probable number (MPN) of diesel degrading microorganisms or B) total microbial biomass as measured by PLFA biomarkers in either Crude Oil (dark grey) or Diesel (light grey) soils. Measurements

were based on 10 triplicate 1 g soil samples. The values shown are means with 95% confidence intervals (N = 6). Treatments indicated are as follows: no treatment (c1, c2), planted with annual ryegrass (p), a mix of annual ryegrass and Arctared fescue (p2), treated with fertilizer (f), and/or no added nutrients (no "f" indicated). Significant differences in percent cover are indicated by different letters, "NS" indicates no significant differences found.

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3.5 Soil Microbial Communities are Influenced by Treatment and Vegetation

Microbial biomass was significantly influenced by both the original phytoremediation treatment and soil type (i.e. crude oil-contaminated soils or diesel-contaminated soils), as indicated by analysis of PLFAs. Total microbial biomass was

300   generally higher in plots which had been fertilized compared to plots that had not received original fertilization (Fig. 3, $P < 0.05$, Kruskal-Wallis), but there was no clear trend between grouped PLFA biomarkers and initial treatment (Fig. S4). However, we did find a significant association between total microbial biomass and estimated total percent cover of vegetation in all plots (Fig. S3; CO: Kendall's $\tau = 0.6101$, $P < 0.0001$; DE: Kendall's $\tau = 0.3012$, $P = 0.0374$). We further explored the influence of original treatment and soil properties on microbial community structure as defined by PLFA biomarkers using

305   PERMANOVA. Most variation in the composition of PFLA biomarkers was explained by original phytoremediation treatment (CO contaminated soils: 82%, DE contaminated soils 61%) and by the presence of fertilizer in that treatment (CO contaminated soils: 36%, DE contaminated soils 24%; Table 1). In diesel soils, the concentration of TPH and NO_3^- explained ca. 20% variation in the composition of PLFA biomarkers (Table 1). In crude oil soils, other measured soil properties, such as percent C, pH, NO_3^-, K, P, and CEC, also significantly explained variation in the composition of PLFA biomarkers (Table 1). In crude

310   oil-contaminated soils coverage of grasses, amount of bare ground, and (marginally) coverage of forbs were significantly associated with variation in the composition of PLFA biomarkers (Table 1).

Table 1. Association of soil microbial community structure, as measured by PLFA biomarkers or 16S rRNA gene sequencing, with factors such as original treatment type, presence of fertilizer in original treatment, amount of vegetation present on the site at the time of sample collection, or soil nutrient and structure data. P values adjusted for multiple comparisons using BH.

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Factor	Crude Oil PLFA			Crude Oil 16S rRNA			Diesel PLFA			Diesel 16S rRNA		
	R2	P (adj)	F	R2	P (adj)	F	R2	P (adj)	F	R2	P (adj)	F
Original Phytoremediation Treatment	0.82	**0.0037**	10.46	0.64	**0.0037**	4.14	0.61	**0.01792**	3.59	0.58	**0.004**	3.23
Fertilization	0.36	**0.0059**	10.48	0.30	**0.0037**	8.07	0.24	**0.0328**	6.07	0.22	**0.0037**	5.31
Tree (% cover)	0.12	0.1472	2.67	0.10	*0.0747*	2.1634	0.05	0.3264	1.10	0.11	**0.0160**	2.46
Grass (% cover)	0.32	**0.008**	8.88	0.24	**0.0053**	5.9332	0.02	0.6330	0.39	0.12	**0.0213**	2.51
Forb (% cover)	0.20	*0.057*	4.70	0.11	*0.0747*	2.3772	0.08	0.2400	1.69	0.07	0.2240	1.39
Bare Ground (% cover)	0.62	**0.0053**	31.64	0.30	**0.0053**	8.0259	0.09	0.2400	1.80	0.06	0.3264	1.15
TPH	0.07	0.2987	1.36	0.05	0.3837	1.08	0.21	**0.0470**	4.98	0.11	**0.0215**	2.33
pH	0.57	**0.0037**	25.07	0.31	**0.0037**	8.57	0.19	*0.0645*	4.39	0.19	**0.004**	4.45
NO3-	0.35	**0.0059**	10.56	0.16	**0.016**	3.63	0.27	**0.01704**	7.05	0.07	0.179	1.46
P	0.28	**0.026**	7.55	0.25	**0.0037**	6.39	0.08	0.26437	1.63	0.20	**0.006**	4.77
K	0.53	**0.0059**	21.46	0.31	**0.0037**	8.34	0.02	0.10095	0.39	0.16	**0.004**	3.51
CEC	0.51	**0.0037**	20.00	0.22	**0.0037**	5.35	0.11	0.168	2.43	0.11	**0.026**	2.24
%C	0.52	**0.0037**	20.98	0.25	**0.0037**	6.40	0.12	0.12821	2.64	0.07	0.176	1.46
Sand	0.07	0.3007	1.33	0.04	0.6623	0.77	0.02	0.748	0.32	0.10	**0.037**	2.11
Silt	0.04	0.4947	0.78	0.04	0.7249	0.69	0.02	0.70158	0.41	0.12	**0.017**	2.56
Clay	0.04	0.4995	0.78	0.11	*0.064*	2.31	0.09	0.192	1.98	0.11	**0.008**	2.31

Ordination of individual PLFA biomarkers indicated that soil community structure was most strongly associated with soil type (Fig. S5A), and when split by soil type there was also an influence of original phytoremediation treatment (Fig. 4A, 4B). We tested the strength of these associations and found the relationship between original treatment and PLFA-derived biomarkers was stronger in the crude oil-contaminated soils (MRPP A = 0.542, P = 0.001) than the diesel-contaminated soils (MRPP A = 0.268, P = 0.005). A vector-fitting method used to interpret the ordinations using measured soil chemical properties indicated that in the crude oil soils several factors were significantly associated with the distribution of all PLFA biomarkers in ordination space, particularly in plots that were initially fertilized which were associated with increase soil nutrients (e.g., P, K) and other metrics of soil health (e.g., near neutral pH, CEC, and %C; Fig. 4A). In diesel-contaminated soils however, only NO_3^- was associated with the distribution of PLFA biomarkers in ordination space (Fig. 4B).
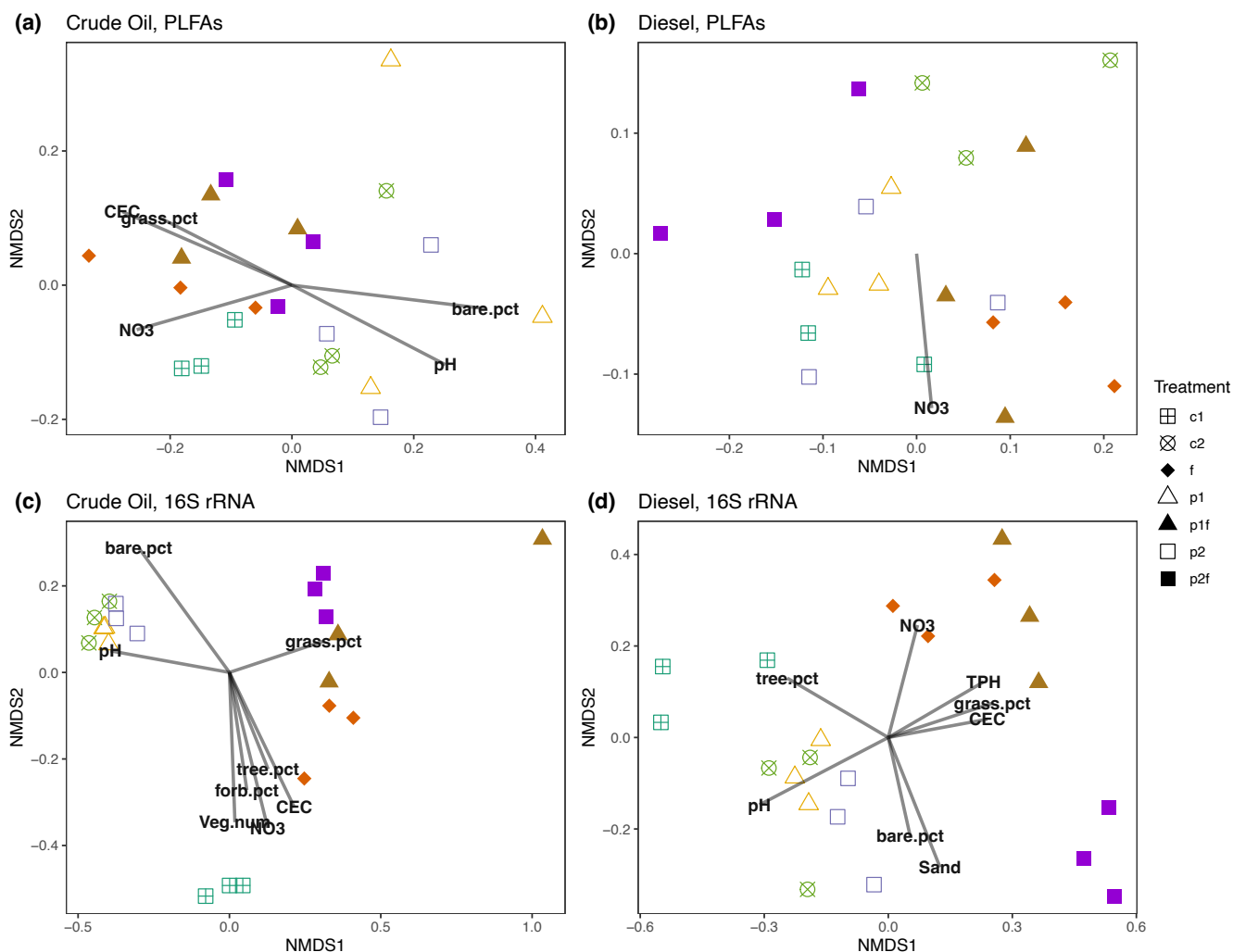
Figure 4. Non-metric multidimensional scaling ordination analysis (NMDS) of soil PLFAs (A & B) or 16S rRNA genes (C & D) from crude oil (A, C) or diesel (B, D) contaminated soils, with subsequent fitting of environmental vectors onto the ordination (P < 0.05). The vectors are as follows: NO3 = nitrate, CEC = cation exchange coefficient, Veg.num =total vegetation counts (excluding trees), Tree.num = total number of trees, TPH = total petroleum hydrocarbons, Sand = soil texture (i.e., %sand, % silt, % clay). Solid symbols indicate treatments originally fertilized. Treatments indicated are as follows: no treatment (c1, c2), planted with annual ryegrass (p1), a mix of annual ryegrass and Arctared fescue (p2), treated with fertilizer (f), and/or no added nutrients (no "f" indicated).

Analysis of bacterial community structure using 16S rRNA gene sequencing revealed the strong influence of soil type (diesel-contaminated silty soil vs crude oil-contaminated gravelly soil) on the soil community (Fig. S5B). Alpha diversity of

the bacterial community as measured by the Shannon Index was higher in the crude oil soils when compared to the diesel-contaminated soils (Fig. S6; P = 0.0005, Kruskal-Wallace). When separated by soil type, bacterial communities were significantly associated original phytoremediation treatment; plots that were originally fertilized grouped separately from other treatments (\CO MRPP A = 0.214, P = 0.013; DE MRPP A = 0.356, P = 0.002; Fig. 4C, 4D). Vector fitting analysis indicated that pH, soil chemical properties (e.g., NO_3^-, CEC, nutrients) heavily influenced the bacterial community structure in both soil types (Fig. 4). In both soil types, the number of trees present on each plot was associated with bacterial community structure in the originally unplanted or unfertilized plots.

We further examined the relationship between soil and phytoremediation treatment factors with bacterial community structure using a PERMANOVA test. In both soil types, original treatment type explained more than half of the variation in bacterial community structure (CO 64%, DE 58%) and the presence of fertilizer in the original treatment explained approximately a quarter of the variation (CO 30%, DE 22%). Measures of soil health, such as near-neutral pH, nutrient concentrations (P, K), CEC, and soil texture, were also associated with the variation in bacterial community structure in both crude oil- and diesel-contaminated soils (Table 1). The plant communities present also helped explain variation in bacterial community structure. In diesel-contaminated soils, percent coverage of trees and grasses was associated with bacterial community structure (Table 1). In crude oil-contaminated soils, coverage of grasses, amount of bare ground, and (marginally) coverage of forbs were significantly associated with bacterial community structure (Table 1). We then used a Mantel test to further explore these relationships between the species composition of plant communities and the bacterial community structure. There was a strong relationship between the bacterial community and the vegetation present on the site in both crude oil (Mantel r = 0.2147, P = 0.013) and diesel-contaminated soils (Mantel r = 0.3568, P = 0.002).

To determine differences among bacterial communities, we constructed a heatmap of families which were at least 2% abundant in more than one sample (Fig. S7). The family *Bradyrhizobiaceae* dominated the bacterial community in diesel-contaminated soils (27.62%) and was also abundant in the crude oil soils (8.89%). The family *Acidobacteria Gp6* was also abundant across all the treatments and contaminant types, with a higher relative abundance in crude oil soils (11.60%) than in diesel soils (8.36%). Members of the families *Gemmatimonadaceae* (crude oil: 4.20%, diesel: 7.63%), *Rhodospirillaceae* (crude oil: 1.73%, diesel: 2.66%), *Xanthomonadaceae* (crude oil: 1.36%, diesel: 1.08%), and *Caulobacteraceae* (crude oil:

365 1.40%, diesel: 1.69%), among others, were present in all investigated soils (Fig. S7). At all sites, there was also a large

percentage of bacteria unclassified at the family level (diesel, 9.81%; crude oil, 13.29%), suggesting that they may be novel

sequences of uncharacterized bacteria (Fig. S7). To further explore the potential influence of original treatment on soil

communities, we conducted differential sequence analysis (Fig. 5) to identify individual bacterial OTUs significantly

associated with planting, fertilization, or a combination of planting and fertilization. Separately, for both crude oil- and diesel-

370 contaminated soils, we conducted three pairwise comparisons among the following treatments: a) initially fertilized (f) versus

planted (p1 and p2), b) initially planted and fertilized (p1f and p2f) versus planted (p1 and p2), and initially fertilized (f) versus

planted and fertilized (p1f and p2f). In general, similar taxa were identified in both the heatmap and differential abundance

analysis, with many unclassified OTUs identified as differentially abundant, but also members of the *Bradyrhizobiaceae* and

Acidobacteria families.

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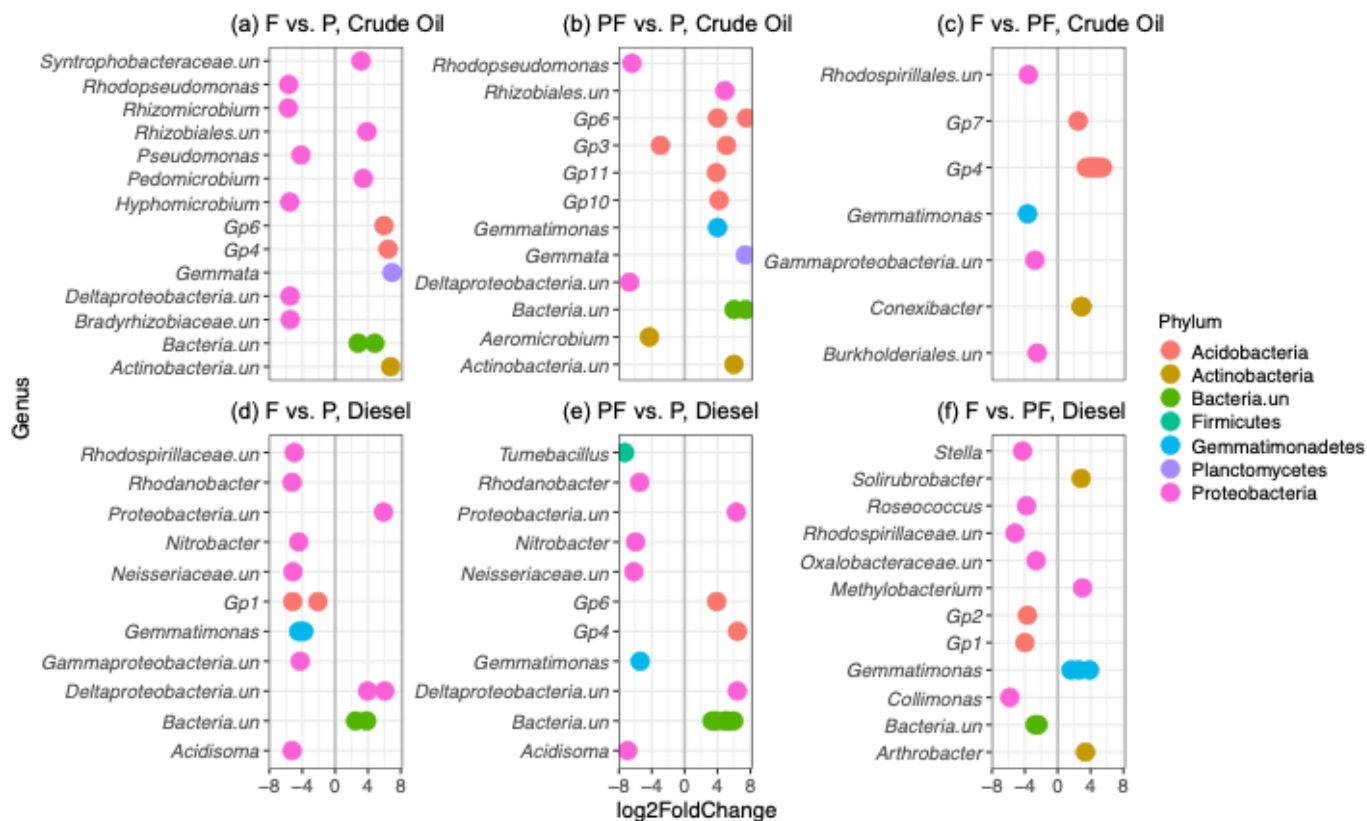
Figure 5. Differential sequence analysis showing significantly enriched (P$_{adj}$ < 0.001) bacterial OTUs from crude oil (a-c) and diesel (d-f) contaminated soils. Pairwise comparisons include initially fertilized (F) versus planted with one or

two grasses (P), initially planted and fertilized (PF) versus planted only (P), and initially fertilized (F) versus planted and fertilized. Negative log$_2$ fold change (left of vertical grey line) values represent OTUs significantly enriched in treatment listed first, while positive log$_2$ fold change (right of vertical grey line) represents OTUs significantly enriched in the second listed initial treatment. Only the top 15 differentially abundance OTUs with the highest log$_2$ fold change are represented. Multiple observations per row represent OTUs belonging to the same genus. Colours represent phyla, genera ending with ".un" indicate unclassified taxa.

4 Discussion

Here, we investigated the long-term influence of phytoremediation on recruited plant communities, contaminant disappearance, and soil microbial community structure and capacity for PHC degradation, 15 years after phytoremediation treatment activities had stopped. We found that phytoremediation treatment had deterministic effects on the recruited vegetation (Fig. 2, S1) and associated bulk soil microbial communities (Table 1, Fig. 3, 4, S5). Together with the measurements of soil microbial community structure and PHC degradation capacity, these data provide a more holistic, long-term view of the impact of treatment on plants and soil microbial communities at this site.

Long after active phytoremediation ended, variation in plant communities across the site was most strongly shaped by the original treatment, although the study design did not allow us to fully determine the cause-effect nature of this relationship (Table 1, Fig. 2, Fig. S1). In particular, the initial use of fertilizer was strongly associated with more vegetation coverage, whereas plots without fertilizer application had more exposed bare ground. While the direct effects of fertilizer application may have been relatively transient, on the order of months, we did find that soil initially fertilized continued to have higher total P and K than other soils (Table S3). It is possible that the original fertilization acted to allow seeds to colonize and establish in treated plots thereby priming plant succession. Many early successional plant species such as forbs, grasses and deciduous trees produce litter that is of higher quality and therefore decomposes more rapidly than secondary successional species such as coniferous trees (Allison and Treseder, 2011). The subsequent turnover of this nutrient-rich plant biomass may also have played a role in setting plant communities on different successional trajectories through longer-term enhancement of available nitrogen (Clark et al., 2007; Cleland and Harpole, 2010). Indeed, other studies in early successional boreal forests have shown that fertilization can shift the plant community toward species with higher-quality litter and increased litter decay rates compared to unfertilized plots (Ruess et al., 1996; Allison et al., 2010).

The continuing influence of the original phytoremediation treatment was also evident in the soil microbial community (Table 1, Fig. 3, 4, S5). In particular, the initial use of fertilizer had sustained effects on the bulk soil microbial community, accounting for differences in both total microbial biomass and bacterial community structure (Fig. 3, 5). The total microbial communities as measured by PLFAs and 16S rRNA amplicon sequencing were structured according to the original treatment and the use of fertilizer. This is in agreement with several long-term studies which have found that initial fertilization in unmanaged systems can lead to sustained differences in microbial communities (Talbot and Treseder, 2012; Geisseler and Scow, 2014; Lupwayi et al., 2021). These treatment-scale effects can extend beyond the bulk-soil community into the endosphere of the plants themselves. In a sister study, we also found that the choice of initial phytoremediation strategy drove the succession of endophytic bacteria associated with colonizing vegetation (Papik et al., 2023).

Soil physiochemical properties such as pH and texture are known to influence microbial communities (Sutton et al., 2013; Leewis et al., 2022). The two soils used in our study differed substantially including different original contaminant, origin, texture, pH, and nutrient status (Table S2, S3). In our previous study of this site, (Leewis et al. 2013), we reported that these dissimilarities led to distinctions in the plants recruited on the two soils, with more primary successional and invasive plants present on the coarse-textured crude oil-contaminated soils. Here, we see how the soil physiochemical effects also influenced the microbial communities and their response to the initial treatments (Fig. 3, 4, S2, S3, S5). Total soil microbial communities and the bacterial community were all more strongly associated with initial phytoremediation treatment and fertilization in the crude oil-contaminated soils than those in the diesel-contaminated soils (Table 1). Coarse-textured soils tend to have less water and nutrient retention, conditions which are less hospitable for plant and microbial growth. Treatment such as fertilization and planting can lead to major changes in soil physiochemical properties, such as increasing soil organic matter content and nutrient retention, which can, in turn, impact the soil community (Reeve et al., 2010; Blanchet et al., 2016). Therefore, the lingering effect of treatment would be more evident in soils that started out as harsher for microbial and plant growth.

Original phytoremediation treatment strategy influenced the vegetation recruited, which in turn influenced microbial community structure and microbial biomass. While it is well known that both plants and fertilizer can cause shifts in soil microbial community structure, here it is almost impossible to tease apart their separate contributions to the microbial

community because the vegetation on each plot is driven by the initial phytoremediation treatment. Our results indicate that the percent coverage of vegetation, rather than the individual plant species themselves, was an important factor in driving soil community structure. Similarly (Brown and Jumpponen, 2014) found that the influence of vegetation composition on bacterial community structure was minor compared to the presence or absence of plants. We also found higher microbial biomass in plots with more vegetation cover which may be a consequence of initial fertilization. In addition to shifting plant recruitment patterns, fertilization can increase plant biomass, which in turn can increase litter deposition and below-ground organic carbon deposition (Geisseler and Scow, 2014), and these plant inputs to soil may be a particularly important influence in degraded soil (Knelman et al., 2012).

Microbial community function is also known to be influenced by the presence and diversity of plants (Leewis et al., 2016a; Musilova et al., 2016; Schmid et al., 2021). In our previous investigation, we found an inverse relationship between increasing counts of woody vegetation and decreasing TPH concentrations (Leewis et al., 2013). Here we show increasing DDM with more tree coverage in crude oil-contaminated soils and a trend towards increasing DDM associated with decreasing TPH concentrations (Fig. S2). Several of the enriched OTUs across all treatments belonged to the phyla *Proteobacteria* and *Actinobacteria*, members of which have been previously shown to degrade diesel in Arctic soils (Ferrera-Rodríguez et al., 2012). Additionally, taxa belonging to several *Acidobacterial* subgroups were found in high relative abundance in all plots (Fig. S7), and OTUs belonging to several *Acidobacterial* subgroups were also found to be differentially abundant in tested treatments (Fig. 5). Members of the phylum *Acidobacteria* have been previously found in association with willow trees (*Salix* sp.) in PHC contaminated soils in the far North (Bell et al., 2014). Together, this may indicate that microbial biodegradation of PHCs is an on-going process, despite very low TPH concentrations, and that the microbial functional potential is stimulated by, or at least associated with, the current vegetation on site. It has been hypothesized that plants with higher amounts of some secondary plant metabolites, such as aromatics, may lead to increased biodegradation potential in contaminated soils (Fletcher and Hegde, 1995; Singer et al., 2004; Slater et al., 2011). At higher latitudes, trees have increased concentrations of secondary metabolites when compared to similar tree species growing at lower latitudes (Stark et al., 2008), this includes many species identified on the investigated sites such as willow (*Salix* spp.; Table S4). Our data indicate that the presence and coverage of

trees, all of which found on the site were native Alaskan species (Leewis et al., 2013), are directly related to increased degradation potential and TPH disappearance in both crude oil- and diesel-contaminated soils.

455   Phytoremediation is a common practice for restoration of sites contaminated with organic pollutants, with most studies pursuing results over the short- to mid-term range (< 5 years (Palmroth et al., 2002; White et al., 2006; Siciliano et al., 2003; Elshamy et al., 2019). With phytoremediation treatment, contaminant concentrations have been shown to decrease below regulatory clean-up limits, with similar percentage losses from initial TPH concentrations as to those presented here, within a short- to mid-term time range in high-latitude soils (Robichaud et al., 2019; Lopez-Echartea et al., 2020). Even without active

460   treatment, there can be significant decreases in original contaminant concentrations which, over long timeframes, may result in a site reaching regulatory clean-up limits (Leewis et al., 2013; Robichaud et al., 2019; Lopez-Echartea et al., 2020). However, accelerating the rate of contaminant disappearance and therefore site remediation, is essential to increasing the speed of ecosystem recovery and minimizing environmental impacts while still maintaining reasonable costs. Here we show that minimal initial investment of plants and fertilizer can increase the initial TPH disappearance rates, while also priming the site

465   for long-term volunteer vegetation cover that supports continued TPH disappearance and ultimately remediation of the site. While the original grasses planted were associated with significant decreases in soil TPH concentrations, they required annual replanting. Our results suggest that initial planting with local plants, which are better adapted to prevailing environmental conditions, may provide a low-cost method to increase long-term phytoremediation potential (Slater et al., 2011; Robichaud et al., 2019).

5 Conclusions

        In conclusion, our study adds to the growing body of knowledge that phytoremediation treatment appears to be an effective means of reducing petroleum hydrocarbon concentrations in soil and stimulating indigenous TPH-degrading microorganisms, and that the effects of original treatment can be seen long after active site management has ceased. The data also indicate the importance of carefully choosing the initial phytoremediation treatments, as these conditions will leave a

475   lasting legacy on TPH transformation, vegetation recruitment, and soil microbial communities. More research is needed to

identify local native vegetation which may be optimal for initiating plant establishment and microbial succession and accelerating *in situ* remediation processes, particularly in sub-Arctic regions.

Data availability

Raw sequence data are available for download from the NCBI Short Read Archive under bioproject number PRJNA950456. All coding for analyses is available at github.com/mcleewis/FarmersLoop.

Author Contributions

MCL analysed the data & wrote the first draft of the manuscript, with support from OU. MCL and MBL designed the experiment. MCL, MBL, and CK conducted experiments. MBL provided equipment, supplies, and feedback. All authors provided revisions to the manuscript.

Conflicts of Interest

The authors declare that they have no conflict of interest.

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