Expression of macromolecular organic nitrogen degrading enzymes identifies potential

mediators of soil organic N availability to an annual grass

Running title: Microbial protease expression in rhizosphere

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Abstract

Nitrogen (N) commonly limits terrestrial plant growth partly because most soil-N is present as

macromolecular organic compounds and not directly available to plants. Soil microbes degrade

these large N-containing substrates to gradually release plant-available inorganic-N throughout

the growing season, potentially meeting plant demand. Knowing which microbes are responsible

for release of organic N, as well as their spatiotemporal patterns of activity, can enable microbial

management strategies that increase plant access to soil-N and reduce dependency on fertilizer-

N. We used time-resolved metatranscriptomes to follow taxonomy-resolved differential

expression of N-depolymerizing enzymes.

Taxonomic groups show adaptations based on extracellular proteases to specialized habitats in

soil characterized by presence (Betaproteobacteria) or absence (Thaumarcheota) of live roots and

root detritus (Deltaproteobacteria and Fungi). A similar increase of eukaryotic chitinases near

root detritus hints at predation of fungi. Others demonstrate temporal patterns such as increasing

expression over time, implying increased competitiveness with substrate depletion (Chloroflexi).

Phylotypes from the same genus can have different potential benefits to the plant based on

protease expression (e.g., *Janthinobacterium*), which should be considered when designing

bioaugmentation. We identify one *Janthinobacterium* phylotype and two Burkholderiales that

may be candidates for bioaugmentation near young roots and a *Rhizobacter* which could benefit

mature roots.

Keywords

Rhizosphere, microbiome, protease, chitinase, functional guilds, metatranscriptome, gene

expression, litter

Introduction

Plants are commonly limited by nitrogen (N) in temperate soils since their access to the largest soil N pools is constrained by the activity of microbes responsible for ammonification of macromolecular nitrogen and N_2 fixation [1–3]. To a large degree, plants depend on microbial degradation of macromolecular forms of N such as proteins and chitin [4–6] and can potentially increase N availability by releasing exudates that stimulate microbial turnover of organic N pools [7–10]. Depolymerization of high molecular weight detrital organic N is a primary rate-limiting step in soil N mineralization [11, 12] and depends on the activity of extracellular enzymes such as lysozyme, protease, chitinase, nuclease and urease. The resulting N monomers are taken up by microorganisms, and roughly 30% of the amino-acid carbon is respired, leading to excretion of excess ammonium which can benefit the plant [13–16]. While the bulk-scale activity of extracellular N degrading enzymes in soil is established, the spatial and temporal dynamics of gene expression underpinning N mineralization are unknown. Additionally, soil microorganisms with the genomic capacity for organic N degradation are diverse [17], and it is unclear whether there is niche partitioning amongst them [18].

Patterns and controls of soil N depolymerization and mineralization may be highly dependent on the habitat (e.g., rhizosphere, detritusphere), since availability of organic-N substrates varies in both time and space, as does the prevalence of fungal, faunal, and bacterial degraders, which can enable or limit N-cycling. In the rhizosphere, organic N is available as amino acids, nucleotides, niacin, and choline (derived from plant exudates [10]), lignoproteins and aromatics (from sloughed off cells), nucleic acids, and microbial cell wall amino sugar polymers (N-acetylglucosamine, N-acetylmuramic acid) derived from the bloom of cells that develops as roots grow [18]. Prior work indicates that root exudates can increase degradation of

64   soil organic matter by up to 380% [19]. In contrast, the detritusphere has a relatively higher

65   proportion of aromatic N and lignoproteins. Bulk protease enzyme activity has been shown to be

66   enhanced by both litter addition [20] and root exudates [21].

67   Soil N pools and by proxy, N mineralization activity, change with time [22] in part due to

the succession of both exudate quality and microbial communities [10, 23]. At the bulk soil level,

activity of extracellular proteases is thought to drive soil N cycling [12], but what controls this

activity is not always clear [24]. Subtle distinctions caused by time and soil habitat are likely

masked because most studies are conducted on whole soils, which contain a mixture of

rhizosphere, detritusphere and bulk regions. In grassland soils, the rhizosphere is a particularly

critical hotspot for microbial biomass and activity due to rapid recycling of root exudation and

debris, and functions as a quasi-digestive system decomposing molecules inaccessible to plants.

However, root exudation, which serves as a carbon source for rhizosphere bacteria, declines near

older root sections [25]. Therefore, it has been hypothesized that bacteria in a mature rhizosphere

environment may be forced to target less labile, higher C:N sources of organic N such as plant

litter [26]. Indeed, it has been shown that the mature *Avena* sp. rhizosphere has higher rates of

gross nitrogen mineralization and ammonia consumption compared to the young rhizosphere

[28].

81   To determine spatial and temporal patterns and identify which groups of microbes were

primarily responsible for decomposition of protein and chitin, we assessed the expression of

genes that code for enzymes degrading macromolecular organic N near actively growing and

decomposing roots of a common annual grass *Avena fatua* (wild oat grass). We used wild oat as

a potential source for putative beneficial microbes to domestic oat, as wild oat in our system does

not experience fertilization and therefore needs to acquire N from soil organic matter via its

microbiome. We analyzed 48 metatranscriptomes collected over three weeks of active root

growth, using a well-developed plant mesocosm approach[23, 26–28] with a fully-characterized

annual grassland soil from Northern California. A different subset of these metatranscriptomes

was previously used to analyze expression of genes coding for degradation of carbohydrates, in

an effort to determine the effect of roots and litter on carbon cycling [28]. The previous study

provided important context that allowed us to compare trends in carbon and nitrogen cycling.

The primary goal of the analysis presented here was to discover the roles of indigenous

rhizosphere microorganisms in macromolecular nitrogen degradation by extracellular enzymes

throughout root growth and aging. This identification is the first step towards *in-situ* inoculation.

Inoculation of seeds with plant growth promoting bacteria has shown promise in reducing the

need for fertilizers[29] and improving plant biomass[30] as well as rhizosphere available

nitrogen[31].

Materials and methods

Experimental design, sample collection and sequencing

The experimental design, sample collection and sequence data processing are described

in detail in [28]. Briefly, common wild oat *Avena fatua* was grown for six weeks in rhizobox

microcosms containing soil from the Hopland Research and Extension Center (HREC) in

northern California, a Bearwallow–Hellman loam (pH 5.6, 2% total C, include 7th approx. name

here) packed at field bulk density. Roots were then allowed to grow into a sidecar soil region

with a transparent wall, where the root growth timeline was marked at 3 days, 6 days, 12 days

and 22 days. In half of the sidecars, the soil was amended with dried *A. fatua* root detritus (C:N =

13.4) chopped to 1 mm. Bulk soil bags, inaccessible to roots, were placed in each sidecar; a bulk

soil treatment amended with root detritus was also included. At each timepoint, three replicate

microcosms were destructively harvested for paired rhizosphere and bulk soil, both detritus

amended and unamended, yielding a total of 48 samples representing four treatments at four

timepoints: rhizosphere, rhizosphere + litter, bulk, and bulk + litter (sup. fig. S1).

Harvested soil (1 g) was placed immediately in Lifeguard Soil Preservation Reagent

(MoBio) and processed according to the company protocol. Roots and supernatant were

removed, and the soil was stored in -80°C. DNA and RNA were co-extracted using a phenol-

chloroform procedure [32, 33] and separated with an AllPrep kit (Qiagen). RNA was DNase

treated (TURBO DNase, Thermo-Fisher Scientific), depleted in ribosomal RNA (Ribo-Zero

rRNA Removal Kit, Epicentre) and reverse transcribed into cDNA. Metatranscriptomes were

sequenced for 48 samples on an Illumina HiSeq 2000 2x150 (TruSeq SBS v3) at the Joint

Genome Institute (JGI), in Walnut Creek CA.

Expression of macromolecular N degrading genes identified in denovo assembled

metatranscriptomes

Raw reads were quality-trimmed (Q20) and rRNA and tRNA reads were removed.

Library size, evenness, richness and Shannon diversity were comparable between experimental

groups with a mean library size of 43 M paired end reads after filtering. In contrast to the single

approach used by Nuccio et al. [28], we also *denovo* assembled quality-controlled

metatranscriptomic reads into contigs within each sample. Contigs smaller than 200bp were

discarded and the remaining contigs from all samples were clustered at 99% identity with cd-hit-

est keeping the longest sequence as the cluster representative [34]. Open reading frames (ORFs)

were predicted using Prodigal [35]. Extracellular protease ORFs were identified by reciprocal

BLAST to extracellular peptidases from the MEROPS database [17, 36]. We acknowledge that

this method may generate a conservative estimate of the number of proteases detected. ORFs

were assigned a peptidase group (serine-, metallo-, cysteine-peptidase and others) by their best

reciprocal BLAST hit. Taxonomy was determined by best BLAST hit to the NCBI non-

redundant database (NR, access date July 29th 2019). Chitinase ORFs were identified using six

chitinase hidden Markov models (HMMs) from the Kyoto Encyclopedia of Genes and Genomes

(KEGG): K01183, K03791, K13381, K17523, K17524 and K17525. Only the first three were

detected in our dataset. Lysozyme ORFs were identified using the HMM for KEGG orthology

K07273, extracellular nuclease with K01150 (codes for Dns gene, undetected) and K07004, and

urease subunit A, B and C with K01428, K01429 and K01430 respectively. Reads were then

mapped back to ORFs requiring minimum identity 95% and 75% breadth using bbmap [37].

Read counts were normalized using DESeq2 [38]. Heat maps of normalized counts were

generated in R using the heatmap2 function in gplots [39]. Normalized transcript counts per gene

per time point were compared between groups (rhizosphere, litter and litter-amended

rhizosphere) using ANOVA and Tukey post-hoc test. Boxplots were generated in ggplot2 [40].

Expression of extracellular protease genes from a curated collection of Hopland-soil genomes

The quality-controlled reads were mapped using BBsplit [37] requiring 80% identity

against a dereplicated reference database of 282 HREC soil genomes including isolates [10],

single amplified genomes (SAGs) [28], metagenomic assembled genomes (MAGs) (NCBI

PRJNA517182) and stable isotope probing MAGs (SIP-MAGs) [41]. On average, 12.3% (range

6.2-31.9%) of the reads per library mapped unambiguously to genomes from the reference

database. This additional approach was taken to investigate gene expression within the context of

a genome and to search for overlap in guild membership between extracellular protease defined

guilds and previously defined CAZy guilds [28]. Three MAGs that were classified as unknown

or domain bacteria in previous studies were assigned taxonomy using GTDB-Tk version 1.3.0

[42, 43]. Verification of the taxonomic assignment of MAG Burkholderiales_62_29 was done

using GToTree [44] with complete reference genomes of Betaproteobacteria from RefSeq (Feb

28, 2020).

Open reading frames (ORFs) were predicted in all genomes using prodigal [35] and

annotated using KEGG [45] and ggKbase (http://ggkbase.berkeley.edu). Extracellular proteases,

which do not have hidden Markov models (HMMs) capable of separating them from intracellular

proteases, were identified by gene nucleotide identity of at least 90% and coverage of at least

60% to *de novo* assembled ORFs of extracellular protease from the metatranscriptomes. Gene

counts were identified using Rsubread featureCounts [46].

Differential expression of chitinases in the genome collection was already performed in a

previous publication on expression of carbohydrate active enzymes (CAZy) and was therefore

not repeated here [28].

Statistical analyses

All features and their abundance (represented by metatranscriptomic read counts

normalized to sequencing depth) were analyzed with DESeq2 [38] requiring p-value < 0.05

(adjusted for multiple comparisons). Ordination and visualization were conducted in R using

ggplot2 [40] and vegan [47]; PERMANOVA (vegan function adonis) was used to detect

significant treatment factors affecting expression of nitrogen depolymerization genes (protease

expression ~ location * treatment * time). We define 'guilds' as groups of taxa with similar gene

upregulation patterns of extracellular proteases in both time and soil habitat. Guilds were

assigned by hierarchical clustering based on differential expression of extracellular proteases

compared to unamended bulk soil, generating four functional guilds: "Rhizosphere",

"Detritusphere", "Aging root", and "Low response". Effects of the four experimental treatments

(bulk, rhizosphere, litter and litter-amended rhizosphere) on protease gene expression were

assessed by ANOVA (p adjusted for multiple comparisons), N=3 per time point.

Data availability

The R code used in this work is publicly available at

https://github.com/ellasiera/Protease_ISME_2022. All data used in this publication, including

raw reads and genome collection, is publicly available. Metagenomes assembled MAGs can be

found in NCBI PRJNA517182, stable isotope probing MAGs in http://ggkbase.berkeley.edu/,

single amplified genomes in IMG under study name Mediterranean Grassland Soil Metagenome

and single amplified genomes in IMG, see sup. table S2 in Nuccio et al., 2020 for accession

numbers [28]. Raw reads can be found in JGI IMG. For JGI ID's (accession numbers) see

supplemental table S1 in Nuccio et al. 2020 [28].

Results

Expression of enzymes targeting macromolecular N

Extracellular enzymes depolymerizing macromolecular N in soil mainly target proteins, cell wall

components, and nucleic acids[48]. Multiple genes for these functions were identified in this

analysis: extracellular nuclease (Xds), urease (UreABC), lysozyme targeting peptidoglycan,

chitinase (*chit1*) targeting fungal cell walls and insect exoskeleton and extracellular proteases.

However, expression of extracellular proteases was an order of magnitude higher than the other

extracellular N degrading enzymes (sup. fig. S2), and like chitinases, was affected by the

presence of litter and roots. In contrast, extracellular nuclease (Xds), urease (UreABC) and

lysozyme were not influenced by either living or decaying roots, thus, we chose to focus on

extracellular proteases and chitinases.

Normalized transcript abundance for 4948 extracellular protease genes from bacteria

(4846) and fungi (102) was significantly affected by time (3-way PERMANOVA, F=2.8,

p=0.038), litter amendment (F=118.9, p<0.001) as well as interactions between time:treatment

(F=26.9, p<0.001), time:location (F=7.7, p<0.001) and location:treatment (F=15.5, p<0.001).

Expression in unamended soil (no litter) generally increased slightly over time (fig. 1A), whereas

litter-amended soil had initially high expression that then decreased and leveled off over time

(fig. 1B).

The 73 distinct chitinase transcripts were expressed at a substantially lower level than the

extracellular proteases. The expression of the *chit1* gene increased over time in litter-amended

soils and was higher than in the unamended treatments (3-way PERMANOVA, p<0.001) (fig.

1C). Expression of fungal chitinases was lower than bacterial chitinases, and both were generally

lower than eukaryotic chitinases (fig. 1D). Transcripts for chitinase KEGG orthologs *CHI3L1_2*,

CHI3L1_4 and *chiA* were not detected at all and the putative chitinase K03791 had low

expression and no significant effects with time or treatment (data not shown). Chitinase CHID1

was detected but was most closely related to plants and therefore disregarded. A companion

study analyzed differential expression of chitinases within our curated genome collection and

while only one MAG was found to upregulate chitinase, it did so in the presence of litter [28].

Figure 1: Aggregated expression of extracellular proteases in bulk and rhizosphere soils from common wild oat *Avena fatua* microcosms, grown without root litter amendment (A), or with litter amendment (B); expression of chitinase gene *chit1* (C) and mean counts of chitinases at the domain level (D) for unamended bulk soil (B), litter-amended bulk soil (BL), unamended rhizosphere (R) and litter-amended rhizosphere (R L). Boxplots represent 75% of the data, whiskers denote 90% of the data and dots (e.g., panel A bulk litter 6 days) represent outliers. Note that the scale of the Y axis varies between panels. Statistical significance of linear regression compared to 3 days within the same location and treatment group: * p value <0.05, ** p value <0.01, *** p value < 0.001.

Structural groups of extracellular proteases

Structural groups of extracellular proteases can be soil-specific and pH dependent [49].

The main groups of extracellular proteases found in our assembled metatranscriptomes were

serine-, metallo- and cysteine-proteases (2679, 1949 and 496, respectively out of 5295 variants

clustered at 99% identity). Aggregated expression patterns of all variants in each group reflected

the same order (serine>metallo>cysteine). Expression of serine- was consistently higher than

metallo-protease across all treatments (ANOVA F=2392, p=0) (fig. 2, sup. table S1). Expression

of serine-proteases was also significantly greater in the presence of litter compared to no litter

(diff=2337, p<0.001), but root litter amendment did not affect expression of metallo- or cysteine-

proteases, and there was no significant effect of time or location on these structural groups (sup.

table S1).

Figure 2: Aggregated normalized expression of extracellular protease structural groups over time in bulk, rhizosphere and root litter-amended soils. Normalized expression per replicate is plotted as dots on top of the bars. Experimental groups shown per time point are (left to right): unamended bulk soil (B), litter-amended bulk soil (BL), unamended rhizosphere (R) and litter-amended rhizosphere (RL). Extracellular protease groups are (left to right): serine-, metallo- and cysteine-protease.

Taxonomy of extracellular proteases

De novo assembled ORFs of extracellular proteases were taxonomically assigned by

BLASTP best hit against the NCBI non-redundant database (NR). Since extracellular proteases

are not marker genes, we considered taxonomic assignments only at the order level or higher,

with the exception of *Rhizobacter*, for which amino acid percent identity values were extremely

high (84±8.4% amino acid identity).

In the rhizosphere, Betaproteobacteria proteases were significantly upregulated

(ANOVA, $p<0.05$) (fig. 3A). Of 547 variants of proteases from this class, 442 were assigned as

Burkholderiales. In contrast, at most timepoints, proteases of Cyanobacteria and Thaumarchaeota

were significantly upregulated in bulk soil compared to the rhizosphere (fig. 3B; sup. table S2).

In multiple other taxonomic groups, proteases were significantly upregulated only in the

presence of litter: Fungi (fig. 3C), class Deltaproteobacteria (fig. 3D), as well as highly

represented phyla Bacterioidetes and Verrucomicrobia (sup. fig. S2; sup. table S2) and classes

Chitinophagia and Gammaproteobacteria (sup. table S3). Interestingly, protease expression

declined with time for the predatory bacteria Myxococcales (fig. 3E), Bdellovibrionales and

Cytophagia, while clades such as Chloroflexi (fig. 3F) and Actinobacteria had increased protease

expression at the final sampling point. While the Actinobacteria and Acidobacteria had a high

number of variants and high expression of proteases, we did not detect a significant effect of

either time or soil habitat/ amendment. Normalized protease expression data are summarized by

phylum (sup. fig. S3), class (sup. fig. S4) and order (sup. fig. S5), and ANOVA F and p values

are in sup. table S2, sup. table S3 and sup. table S4 respectively.

Figure 3: Expression of extracellular proteases over time for select taxonomic groups: (A) Betaprotebacteria (n=547) were upregulated in the rhizosphere, (B) Thaumarchaeota (n=112) were downregulated in the rhizosphere, (C) Fungi (n=99) were upregulated in the presence of litter, (D) Detaproteobacteria (n=228) were upregulated in the presence of litter, (E) Myxococcales (n=147) were upregulated in the presence of litter and decreased over time and (F) Chloroflexi (n=153) increased in the last time point. The legend in panel F applies to all panels.

Note that as expression levels varied over taxonomic groups and to emphasize patterns, the y

axes do not have the same scale.

Functional guilds

To define functional guilds with a population-centric analysis, we mapped transcriptome reads to

a collection of 282 genomes and metagenome-assembled genomes generated from the same soil

and site[28]. Read counts were used to determine differential expression compared to

unamended bulk soil. Each genome contained multiple genes coding for extracellular proteases.

Hierarchical clustering of the mean differential expression of extracellular protease genes within

each genome was used to define four functional guilds: "rhizosphere", "detritusphere", "aging

root" and "low response" (fig. 4). These guilds include bacteria that upregulate expression of

extracellular proteases in the presence of live roots (rhizosphere), dead roots (detritusphere), live

roots that are beginning to senesce (aging root) or at low but significant levels without a specific

pattern (low response)[28]. Similarly, hierarchical clustering based on differential expression of

the most highly upregulated *de novo* assembled ORFs compared to unamended bulk soil

revealed clear rhizosphere and detritusphere guilds (sup. fig. S6). There was some overlap

between guilds identified here and guilds previously identified based on expression of CAZy

genes[28], but the degree of overlap varied by guild (sup. fig. S7).

Within the aging root guild, we noticed one member, representing an aggregated

population transcriptome, that increased differential expression of extracellular proteases at the

last time point in the presence of both roots and litter more than would have been predicted by

the sum of the treatments alone. As this pattern differed from the rest of the guild, we identified

this member as a sub-guild labeled "late synergist" due to its expression pattern. This MAG,

Burkholderiales_62_29, was identified by two independent phylogenomic analyses as

Rhizobacter. Burkholderiales_62_29 has 16 different extracellular protease genes, all of which

had similar significant upregulation patterns. While the sub-guild here contained only a single

Rhizobacter MAG, the implication of mapping transcriptomic reads at 80% identity is that each

MAG represents a taxonomic "cloud" of at least genus-level diversity. A 16S-rRNA survey of

the same samples revealed 10 operational taxonomic units (OTUs) of order Burkholderiales

which, like *Rhizobacter*, are not assigned a family [28]. Additionally, this sub-guild may include

more taxa for which we have no MAGs and therefore could not be included in this analysis.

Figure 4: Functional guilds defined by hierarchical clustering of extracellular protease

differential gene expression during a 22-day *Avena fatua* microcosm experiment. Each row

represents the mean differential expression of extracellular protease genes mapped to a reference

genome. Treatments are (left to right): litter-amended bulk soil, litter-amended rhizosphere and

unamended rhizosphere. Time points are indicated at the bottom in days. We note that a genome

may contain more than one protease gene and that reads were mapped at 80% identity, therefore

each genome also represents closely related taxa. Differential expression values per gene that

were not statistically significant were converted to zero (0) before averaging.

Discussion

Our current understanding of macromolecular nitrogen depolymerization by microbes in

soil is based almost entirely on measurement of bulk mineralization rates and genomic surveys.

Linking rates and genetic potential is challenging, as rates cannot be broken down by taxonomy

and gene presence represents a blueprint which may or may not be acted upon. For example, we

show here that of the organic N-degrading enzymes detected in this study, only protease and

chitinase are differentially expressed near roots. However, this change in expression was not

reflected in changes in the overall microbial community composition [28]. Based on genomic

potentials previously reported by Nguyen et al.[17], we hypothesized that expression of

structural groups of proteases would be dominated by metalloproteases[17] but expression of

serineproteases proved to be significantly higher across experimental groups in this study.

Therefore, to identify bacterial candidates that may be beneficial to the plant by improving

nitrogen accessibility, analysis of transcripts which provide both taxonomic and metabolic

information is crucial.

Chitinase expression, while two orders of magnitude lower than that of proteases, was

significantly higher and increased over time in litter-amended samples. This effect could imply

that the litter attracted saprotrophic fungi and possibly arthropods, the cell walls of which contain

chitin [50]. Indeed, expression of fungal proteases was significantly higher in the presence of

litter. As most of the chitinases were most closely related to eukaryotic enzymes, it is likely that

the chitin-containing hyphae were preyed upon by soil fauna.

We found that in unamended soil, expression of extracellular proteases was higher in the

rhizosphere compared to bulk soil, potentially resulting from input of carbon from root exudates

creating a higher demand for nitrogen as well as microbial competition over inorganic nitrogen

with the plant. Similarly, DeAngelis et al. showed that protease specific activity was significantly

higher in the young rhizosphere compared to bulk soil, but activity (based on enzyme assays)

was not significantly different based on root age [27]. In litter-amended soil, expression of

extracellular protease was highest regardless of root presence at 3 days, suggesting that litter-

added carbon overwhelms the effect of root exudates at this early point in time. At 12 and 22

days there was no difference between litter-amended and unamended soil in either rhizosphere or

bulk soil, suggesting that the effect of litter had waned, likely due to substrate degradation. In

support, expression of carbohydrate active enzymes after litter addition generally decreased over

time [28].

Comparing membership between guilds previously defined by CAZy expression [28] and

guilds defined here by hierarchical clustering of extracellular protease expression revealed guild-

dependent trends. In the rhizosphere guilds, comprised primarily of Betaproteobacteria known to

be commonly associated with rhizosphere soil[23, 51, 52], we found high similarity in

membership, whereas in the detritusphere guilds there was low similarity. This implies that in the

rhizosphere, the same bacteria break down complex carbon and complex nitrogen, whereas in

bulk soil these processes are performed by different taxa. The functional guilds identified here

exhibited a pattern of temporal succession throughout plant maturation, where expression of

protein degrading enzyme genes in the rhizosphere could be influenced by the host plant through

control of exudation, or alternatively, may be a response to competition with the plant for labile

nitrogen. Moreover, Nuccio et al. [28], analyzing a different subset of this experimental data,

showed that microbial community composition (by 16S-rRNA) in each experimental group

changed very little compared to gene expression over the course of this experiment [28].

Therefore, the changing expression patterns over time that we observed are likely related to

changes in environmental conditions or cues, such as macromolecular N availability and

inorganic N availability, as opposed to wholesale shifts in community composition.

Augmentation of soil or seeds with bacteria that specialize in macromolecular organic N

degradation could be used to reduce the use of fertilizers. Ideally, bioaugmentation should use

bacteria endemic to the specific plant and soil in order to provide strong competitors[53].

Different patterns of expression of extracellular proteases by various taxonomic groups indicate

that potential beneficial partners for the plant change throughout root growth. Hence, functional

guild characterization can be useful to decide which taxa should be selected for bioaugmentation,

with rhizosphere-, detritusphere- and aging root-specific taxa as candidates. Ideally, organisms

from different guilds could be combined into consortia that should effectively contribute more

than a single taxon to the nitrogen economy of the plant through mineralization of nitrogen

coinciding with plant demand over time.

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Competing interests: The authors declare that they have no competing interests

Funding

This research was supported by the U.S. Department of Energy Office of Science, Office of

Biological and Environmental Research Genomic Science program under Awards DE-

SC0020163 and DE-SC0016247 to M.K.F at UC Berkeley and awards SCW1589 and SCW1678

to J.P-R. at Lawrence Livermore National Laboratory. Sequencing was conducted as part of

Community Sequencing Awards 1487 to JPR and 1472 to MKF. Work conducted at LLNL was

contributed under the auspices of the US Department of Energy under Contract DE-AC52-

07NA27344.

Authors' contributions: ETS performed the data analysis, data interpretation and drafted the

manuscript. EEN, JPR and MKF conceived and designed the original rhizosphere-detritusphere

study, EEN helped to interpret the data and revise the manuscript. JPR and MKF both interpreted

the data and substantially revised the manuscript. All authors approved the submitted version and

agreed both to be personally accountable for the author's own contributions and to ensure that

questions related to the accuracy or integrity of any part of the work, even ones in which the

author was not personally involved, are appropriately investigated, resolved, and the resolution

documented in the literature.