

Microbial Communities Collectively Recycle Cadavers Over One Year of Human Decomposition

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Abstract

During terrestrial vertebrate decomposition, host and environmental microbial communities work together to drive biogeochemical cycling of carbon and nutrients. These mixed communities undergo dramatic restructuring in the resulting decomposition hotspots. To reveal the succession of both the active microbial members and the metabolic pathways they use, we generated metatranscriptomes from soil samples collected over one year from below three decomposing human bodies. Soil microbes increased expression of heat shock proteins in response to decomposition products changing physiochemical conditions (*i.e.*, reduced oxygen, high salt). Increased fungal lipase expression implicated fungi as key decomposers of fat tissue. Expression of nitrogen cycling genes was phased with soil oxygen concentrations: during hypoxic soil conditions, genes catalyzing N-reducing processes (*e.g.*, hydroxylamine to nitric oxide and nitrous oxide to nitrogen gas during reduced oxygen conditions) were increased, followed by increased expression of nitrification genes once oxygen diffused back into the soil. Increased expression of bile salt hydrolases implicated a microbial source for

the high concentrations of taurine typically observed during vertebrate decomposition. Collectively, microbial gene expression profiles remained altered even after one year. Together, we show how human decomposition alters soil microbial gene expression, revealing both ephemeral and lasting effects on soil microbial communities.

Keywords: Human Decomposition, Microbial Succession, Metatranscriptomics, Soil Microbial Ecology

Introduction

Soil microbial communities are important drivers of ecosystem processes in terrestrial environments. Many soil microbes are decomposers that degrade complex organic matter and drive nutrient cycling in terrestrial ecosystems. Environmental disturbances can impact the presence and/or activity of soil microorganisms involved in these cycles, ultimately affecting nutrient availability and greenhouse gas emissions, such as CO₂ and N₂O [1, 2]. Vertebrate death and subsequent carcass deposition in terrestrial ecosystems is one disturbance resulting in the deposition of large quantities of organic C and N [3–10], along with other elements (P, K, S, *etc.*) [11], which collectively stimulate microbially-mediated biogeochemical cycling. In addition to this, changes in pH, temperature, and fluctuations in soil oxygen provide abiotic filtering further impacting microbial metabolic strategies [7–9, 11–13]. Vertebrate decomposition also results in mixing of host and environmental microbes: the animal’s microflora are flushed into the soil along with decomposition products where they further contribute to decomposition processes (*e.g.*, organic nitrogen mineralization) [14].

While C and N transformations have been documented during decomposition, the functional response of microbes and their roles in nutrient cycles remain unclear. The composition and structure of decomposition-impacted soil microbial communities have been investigated using sequencing of marker genes amplicons (*i.e.*, 16S rRNA, 18S

rRNA, ITS). This has allowed for the identification of changes in microbial biodiversity and taxonomic succession in response to vertebrate decomposition, revealing patterns that include increases in the anaerobic taxa *Firmicutes* and *Bacteroidetes* [15]. However, few studies have integrated soil biogeochemistry with microbial community composition, which can further help to describe microbial ecology in these decomposition systems. Taylor et al. (2024) [13] showed that fungal community shifts were linked to changes in soil dissolved oxygen, highlighting interactions between soil microbes and changes in the surrounding environment. While insightful for making potential connections between taxa and environment, these analyses do not inform which taxa are active members of the community, which functional pathways/genes are expressed, and how these pathways facilitate decomposition processes.

RNA sequencing (*i.e.*, metatranscriptomics) and metabolomics can be used to investigate microbial community functional succession during decomposition. They can identify how ecological functions, including C and N cycling, are impacted by decomposition events in terrestrial ecosystems. To date, applications of metatranscriptomics to vertebrate decomposition samples have been limited to internal host communities [16, 17]: Burcham et al. (2019) [16] revealed differential expression of amino acid and carbohydrate metabolism in the heart during mouse decomposition, while Ashe et al. (2021) [17] documented taxonomic shifts in gene expression of oral microbial communities during human decomposition.

We expected that the impacted soil microbial community, which includes a mix of host and environmental taxa, would also have altered gene expression profiles, given the release of decomposition byproducts into the soil during terrestrial decomposition. We previously assessed the decomposition-impacted soil metabolome [18], demonstrating a prevalence of amino acids and suggesting upregulation of organic nitrogen metabolic pathways. Additionally, DeBruyn et al. (2021) [18] showed the soil metabolome was

139 surprisingly still altered compared to starting conditions at the end of that 21-week
 140 study, suggesting long-term impacts of decomposition on soil microbial functioning.
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 142 Here, we investigated soil microbial gene expression during a one-year period of human
 143 decomposition. The overarching goal of this work was to assess the effects of vertebrate
 144 decomposition on ecosystem function by characterizing community-level shifts in soil
 145 microbial function. We hypothesized that: (i) gene expression would shift over time as
 146 resources were consumed and transformed and soil chemical and physical conditions
 147 changed due to the influx of decomposition products during soft tissue degradation
 148 [8, 9, 18]; (ii) gene expression for enzymes involved in nitrogen cycling would be altered,
 149 as changes in nitrogen pools have been previously described in decomposition soils [8];
 150 (iii) expression of genes involved in lipid metabolism would increase, as lipids from
 151 the body entered the soil during decomposition and previous studies identified lipoly-
 152 tic organisms in decomposition soils [12, 19]; (iv) microbial expression profiles in the
 153 impacted soil would not fully recover and remain altered even after a year, as previ-
 154 ous studies have shown that community composition was still altered at least a year
 155 after decomposition began [20, 21]. We analyzed metatranscriptomes of soil samples
 156 collected at six key timepoints over one year of human decomposition to determine
 157 the identity of active populations and the expression of genes and pathways relevant
 158 to the enhanced biogeochemical cycling observed in decomposition hotspots. We com-
 159 pared gene expression between decomposition timepoints and control soils that were
 160 unexposed to decomposition products to identify functions or functional pathways
 161 of interest. We show: (i) decomposition shifts soil microbial community gene expres-
 162 sion, with the effects still measurable after one year; (ii) expression of genes related to
 163 stress response are elevated in decomposition soils; (iii) expression of genes encoding
 164 triacylglycerol lipase differed between fungi (increased) and bacteria (decreased), indi-
 165 cating differential responses between bacterial and fungal decomposers; (iv) evidence
 166 for phased nitrification and denitrification, driven by changes in soil dissolved oxygen;
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(v) evidence for organic sulfur processing (taurine) via bile salt hydrolases. This direct assessment of function expands the fundamental understanding of terrestrial vertebrate decomposition, providing insight into pathways of biogeochemical cycling within these hotspots.

Results

Soil Physiochemistry

Soil chemistry was altered in response to the presence of decomposing human cadavers, with multiple parameters still impacted after one year [13]. Generally, soil pH decreased and remained low in decomposition soils. Soil electrical conductivity (EC) increased in response to decomposition, remaining elevated through approximately day 58 before gradually decreasing throughout the remainder of the study (Supplementary Fig. 1). Respiration (evolved CO₂) increased by an order of magnitude beginning at day 12, which corresponded to a reduction in soil dissolved oxygen (DO) to 29% - 48.9%. Ammonium concentrations increased 78-fold, reaching maximum concentrations between days 12 and 58. This was followed by decreased ammonium and increased nitrate concentrations at day 86, with nitrate concentrations reaching a maximum at day 168 (Supplementary Fig. 1).

Microbial gene expression in response to human decomposition

Gene expression profiles in decomposition-impacted soils shifted away from controls and day zero samples as decomposition progressed (Fig 1A). Expression was most different from controls on study days 58, 86, 168 (Supplementary Fig. 2), before starting to return toward control conditions on study day 376. After one year of decomposition, soil gene expression profiles had not returned to pre-decomposition conditions, as evidenced by their clustering away from controls and day zero samples in the MDS plot (Fig 1A).

231 Some correlations were observed between gene expression shifts and soil physiochemi-
 232 cal data at decomposition timepoints. Canonical correspondence analysis (CCA) was
 233 used to constrain gene expression data with soil physiochemical data (Fig 1B). CCA1
 234 and CCA2 explained 29.2% and 18.1% of the variance in gene expression, respectively.
 235 Transcript profiles at day 12 were associated with an increase in soil carbon to nitrogen
 236 ratio (C:N). Gene expression profiles at days 58 to 86 were positively correlated with
 237 increased soil temperature, EC, and evolved CO₂, while study day 168 was associated
 238 with elevated levels of soil NO₃. Further, Permutational Analysis of Variance (PER-
 239 MANOVA) revealed that internal accumulated degree hours (ADH) ($p = 0.001$), soil
 240 temperature ($p = 0.039$), pH ($p = 0.033$), and EC ($p = 0.031$) significantly explained
 241 some of the variation in gene expression profiles ($p < 0.05$). No other soil chemical
 242 variables were significant at $\alpha = 0.05$ (Supplementary Table 1).
 243
 244 Overall, decomposition changed soil gene expression profiles over the one-year study
 245 relative to control soils. Differential expression analysis between decomposition and
 246 control soils identified 7,047 down-regulated and 38,425 up-regulated genes. Gene
 247 transcripts that were associated with control soils belonged to a wide variety of clus-
 248 ters of orthologous genes (COG) functional categories. Specifically, the top 20 genes
 249 whose expression was higher in control soils belonged to ten unique COG categories,
 250 including signal transduction mechanisms, transcription, and those of unknown func-
 251 tion. In contrast, the top 20 genes whose expression was higher in decomposition soils
 252 only fell into four COG categories (Supplementary Fig. 3 A): 1) post-translational
 253 modification, protein turnover, and chaperones; 2) energy production and conversion;
 254 3) cell motility; and 4) carbohydrate transport and metabolism. The most common
 255 COG category represented in decomposition soils (80% of the top 20 genes) was post-
 256 translational modification, protein turnover, and chaperones. Within this category,
 257 several heat shock stress response genes were identified, including clpB, dnaK, groL2,
 258 SSA2, HSP82, and clpB (Supplementary Table 2). Further investigation of these genes
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over time shows that their expression increased, typically reaching maximum transcript levels around study days 58 and 86 (Fig 2). This corresponded to elevated soil temperatures below decomposing bodies between study days 12-80, with soil temperatures increasing to approximately 43°C [13], as well as maximum soil EC and minimum dissolved oxygen measurements between days 12 and 58 (Supplementary Fig. 1).

Taxonomy associated with top differentially expressed gene transcripts also differed between control and decomposition soils. The top 40 most significant differentially expressed gene transcripts in decomposition soils were associated with Fungi, *Actinobacteria*, and *Xanthomonadales*, while gene transcripts in controls were associated with *Acidobacteria*, *Cyanobacteria*, *Proteobacteria* (α , δ , γ), and *Planctomycetes* (Supplementary Fig. 3 B). The greatest number of differentially expressed genes relative to control samples was observed at day 86, where we saw 145,460 and 124,883 up- and down-regulated genes, respectively.

Temporal gene expression shows shifted decomposer functions

Differential expression analysis between sequential study days revealed which genes were altered during decomposition. The top ten transcripts that changed in representation (increased/decreased), determined by the lowest p-values from differential expression analysis, are reported in Supplementary Table 3 and Fig 3.

Expression of genes annotated with the COG categories cell wall/membrane/envelope biogenesis, inorganic ion transport and metabolism, and carbohydrate transport and metabolism increased proportionally from day 0 to 12. In contrast, expression of secondary metabolite biosynthesis, transport, and catabolism genes decreased during this period (Fig 3A). Transcripts from *Bacilli* and *Clostridia* increased, while transcripts from *Actinobacteria* decreased between study days zero and 12 (Fig 3).

323 Between days 12 and 58, 90% of the top 10 upregulated genes were associated with
 324 the translation, ribosomal structure and biogenesis COG and all were taxonomically
 325 associated with *Betaproteobacteria* (Fig 3A,B). Many of these genes were annotated
 326 as ribosomal protein large (RPL), involved in ribosomal binding. Genes across multi-
 327 ple COG categories with taxonomic associations to *Bacilli* and *Clostridia* decreased
 328 between study days 12 and 58, six of which were transcripts that previously increased
 329 between days zero and 12 (Fig 3B, Supplementary Table 3).
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 331 Multiple transcripts associated with the energy production and conversion COG, as
 332 well as transcripts annotated as inorganic transport and metabolism, and translation,
 333 ribosomal structure and biogenesis, increased between days 58 and 86 (Fig 3A). Two of
 334 the upregulated energy and production and conservation transcripts were associated
 335 with cytochrome c oxidase subunits in *Betaproteobacteria*, while another was anno-
 336 tated as *hao*, encoding the enzyme hydroxylamine dehydrogenase which is involved in
 337 conversion of hydroxylamine to nitrite during nitrification (Supplementary Table 3).
 338 Further investigation into hydroxylamine dehydrogenase showed a significant increase
 339 in *hao* transcripts at day 86 followed by subsequent decreases at days 168 and 376
 340 ($F = 4.183$; $p = 0.02$). This increase corresponded to decreased soil ammonium lev-
 341 els and subsequent accumulation of nitrate (Supplementary Fig. 1). Half of the 10
 342 most downregulated genes between days 58 and 86 were not assigned to a COG (*i.e.*,
 343 unclassified) or were of unknown function.
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 345 Differential expression comparing study days 86 with 168 and 168 with 376 identified
 346 genes across a variety of functional categories, with many unclassified in the COG
 347 database or with unknown function (Fig 3A). Expression of carbohydrate transport
 348 and metabolism genes associated with *Bacilli* decreased between day 168 and 376.
 349 *Acidobacteria* transcripts increased in decomposition-impacted soils between study
 350 day 168 and 376, but were not associated with any single COG category (Fig 3B).
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Organic carbon metabolism

We expected to observe increased expression of lipid metabolizing genes during active and advanced decomposition as microbes degraded lipids deposited in the soil [19]. Therefore, we investigated changes in triacylglycerol lipase (enzyme commission number: 3.1.1.3) gene transcription in our soils. Generally, lipase transcripts decreased as decomposition progressed (HLM $F = 6.564$, $p < 0.001$), however we also observed a significant interaction between study day and taxonomic annotation ($F = 8.786$; $p < 0.001$). Specifically, lipase gene transcripts annotated as bacteria decreased with decomposition time ($F = 10.392$; $p = 0.001$), while fungal lipase transcripts increased, reaching a maximum at study day 58 ($F = 4.509$; $p = 0.015$) (Fig 4).

Nitrogen- and sulfur compound transformations

Expression of nitrogen cycling genes was impacted in response to human decomposition. Due to the detection of hydroxylamine oxidoreductase (*hao*) transcripts in our differential expression analysis, and our hypotheses predicting changes to nitrogen transformation processes, the expression of genes encoding common enzymes involved in nitrogen cycling (*nifH*, *nirB*, *nirK*, *norB*, *nosZ*, *nrfA*, *nxrA*, and *amoA*) were assessed using their enzyme commission numbers (Fig 5A,B). *nifH*, encoding a subunit of nitrogenase which is involved in nitrogen fixation, displayed little to no changes in gene expression between control and decomposition soils. Transcripts for two genes encoding enzymes contributing to the last two steps of denitrification, *norB* (nitric oxide reductase) and *nosZ* (nitrous oxide reductase), increased between study days 12 and 86, and decreased at study day 168 before increasing again at day 376. In contrast, expression of genes encoding nitrate reductase, *narG*, and NO-forming nitrite reductase, *nirK*, remained low until day 376 when transcripts for both genes increased. As noted above, expression of *hao*, encoding hydroxylamine dehydrogenase, increased at study day 86 before decreasing at remaining timepoints (Fig 3A, Fig 5B). Expression

of *amoA*, encoding a subunit of ammonia monooxygenase, and *nxrA*, encoding a subunit of nitrite oxidoreductase, which are involved in nitrification, changed in response to decomposition. *amoA* transcripts initially decreased at day 12, remaining reduced until study day 376. Similarly, abundance of transcripts encoding enzymes involved in dissimilatory nitrate reduction, *nirB*, and *nrfA*, was low for the first 168 days, with *nrfA* expression increasing at day 376 (Fig 5B).

Expression of genes involved in metabolism of nitrogen and sulfur-containing compounds were also impacted by human decomposition. Specifically, four of the top ten genes whose expression decreased at day 12 were related to taurine metabolism, with their annotations associated with *tauD*, encoding taurine dioxygenase. (Supplementary Table 3). Further investigation into *tauD* showed that mean expression of these genes decreased steadily over one year, beginning at day 12 (Fig 6B); however, *tauD* expression in response to human decomposition was variable across taxonomic associations. Most *tauD* transcripts were associated with *Gammaproteobacteria*, *Actinobacteria*, *Betaproteobacteria*, *Alphaproteobacteria*, and fungi. While a majority of the *tauD* gene queries displayed reduced expression over time, expression of fungal-associated and a few *Betaproteobacteria*-associated *tauD* genes increased at day 58 (Supplementary Fig. 4). Sources of taurine in the human body include taurine absorbed from the diet and taurine produced from anaerobic microbial deconjugation of bile salts via bile salt hydrolase (BSH) enzymes [22]. Therefore, we examined transcripts encoding BSH enzymes in decomposition soils. Expression of these genes was elevated at days 12, 58, and 86 before converging toward pre-decomposition levels at days 168 and 376 (Fig 6A). Hierarchical liner mixed effects (HLM) models showed that both *tauD* (HLM $F = 7.356$, $p = 0.002$) and BSH ($F = 13.768$, $p < 0.001$) gene expression was significantly different over time (Fig 6A,B).

Discussion

The goal of this study was to assess microbial gene expression in soils responding to human decomposition. Metatranscriptomics were applied to soil samples collected over one year from below three decomposing human bodies. From this, we found that microbial gene expression reproducibly shifted over time. Additionally, we showed that gene expression profiles had not recovered to pre-decomposition conditions after one year. Comparison of control and decomposition expression profiles revealed that heat-shock proteins were elevated in response to decomposition. We also described expression patterns between decomposition timepoints, noting changes in functional gene categories at certain timepoints, in particular with respect to lipid, nitrogen and sulfur metabolism.

Decomposition impacted soil community gene expression for at least one year

Gene expression profiles remained altered after one year of decomposition. It is unclear if soil microbial communities, in terms of gene expression profiles, had reached a new steady state as a result of decomposition, or if they would eventually return to pre-decomposition conditions. The soil pH, EC, NH_4^+ , NO_3^- , and total nitrogen (TN) exhibited differences (although not statistically significant) in these soils following a year of decomposition, however bacterial and fungal community structures, as assessed by rRNA amplicon libraries, were still altered [13]. This indicates that decomposition can continue to structure microbial communities and impact their function for extended periods of time. While nutrient pools and communities both demonstrate less rapid change at later time points in the study, there is no evidence suggesting an arrival at a steady-state post-disturbance microbial community within our study. In some studies, human decomposition can result in elevated carbon and nutrients (organic nitrogen, ammonium, nitrate, and phosphate) for longer than a year [3], suggesting

decomposition events have long-lasting effects on the local ecosystem. Together, this has implications for terrestrial ecosystem processing (*e.g.*, nutrient cycling, emission of greenhouse gasses, etc.), as we show that decomposition alters functional metabolism pathways within soil microbial communities. It is clear that extended sample collections beyond a single year are needed to address how long microbial communities are affected, and whether there is a return to the original state or some new altered community condition.

Bacteria, fungi, and archaea were all represented by expressed genes throughout decomposition, suggesting that members of all three domains have the potential to contribute to decomposition processes and nutrient cycling. While a majority of annotated transcripts were identified as bacterial, fungal transcripts were the second most abundant group. Fungal transcripts made up almost half (*e.g.*, seven of the top fifteen) of the significantly differentially expressed genes associated with decomposition-impacted soils. Additionally, with respect to expression shifts between decomposition timepoints, fungal transcripts were among the topmost upregulated genes at study day 86. This is not surprising as fungi are key decomposers, involved in the degradation of organic matter in terrestrial ecosystems [23]. It was interesting to see an increase in certain fungal transcripts, such as lipase, at study days 58 and 86 when soil oxygen began to recover. We would expect lipids to enter the soil as tissues are broken down during decomposition, so we were surprised to see bacterial lipase genes decrease during decomposition. This suggests that microbial activity in decomposition soils may be constrained by the changing chemical environment, potentially altered oxygen levels in the case of bacterial lipase gene expression. Prior work with these same soils showed that soil oxygen concentration was a key driver of changes in both bacterial and fungal community composition [13].

Increased stress response during decomposition	553
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Soil microbial communities expressed stress response genes in response to human	555
decomposition. Differential expression analysis identified increased expression of mul-	556
tipte heat shock proteins associated with the taxa <i>Xanthomonadales</i> , <i>Actinobacteria</i> ,	557
and fungi. Upon further investigation, expression of these genes increased through day	558
58 and remained high for the remainder of the year. Soil temperature was elevated rel-	559
ative to controls between study days 8 and 80, with maximum temperatures $>40^{\circ}\text{C}$,	560
while soil electrical conductivity increased up to $663\ \mu\text{S cm}^{-1}$ (16X higher than back-	561
ground) through day 58 before slowly decreasing through the remainder of the study.	562
Soil electrical conductivity correlates with ionic strength and can be an indicator	563
of increased salinity [24]. During vertebrate decomposition, elevated conductivity in	564
impacted soils is attributable to sodium (Na), potassium (K), and ammonium (NH_4)	565
[8–11, 13]. As a result, we would expect these microbes to be experiencing both heat	566
and osmotic stress during this period. Prior work has observed increased heat shock	567
gene expression during salt stress in paddy soils [25] and the presence of both heat	568
and osmotic stress genes in desert soils along a salt gradient [26], suggesting saline	569
conditions can alter the expression of heat and/or osmotic stress genes. In our study	570
we observed the stress response within soil microbial communities was stimulated dur-	571
ing human decomposition. At this time, however, it is unclear if expression of these	572
genes is in response to heat stress alone, or in combination with osmotic stress.	573
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Increased expression of fungal lipase genes during	580
decomposition	581
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Human fat tissue contains lipids that are broken down during decomposition. There-	583
fore, we assessed expression of triacylglycerol lipase genes in decomposition soils. Our	584
results show that expression of triacylglycerol lipase genes was altered in response	585
to decomposition, and these shifts differed between bacterial and fungal transcripts.	586
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Specifically, bacterial triacylglycerol lipase transcripts decreased in response to decomposition, while fungal triacylglycerol lipase transcripts increased. Further, expression of these genes corresponded to changes in relative abundance of the fungal classes *Saccharomycetes*, *Sordariomycetes*, and *Eurotiomycetes* [13]. These fungi have been previously associated with decomposition soils [27, 28] and are known to contain triacylglycerol lipase genes in their genomes [29, 30], suggesting that they play a role in lipid degradation in decomposition soils.

Our observation of an overall decrease in triacylglycerol lipase transcripts contrasts with previous work by Howard et al. (2010) [19], who observed increased copies of Group 1 lipase genes via qPCR during swine (*Sus scrofa*) decomposition. Fatty acid composition differs between human and pig tissue [31], potentially altering the lipid profile available for microbes, leading to differences in decomposition products within the soil [18]. These products can then directly or indirectly alter community composition and/or activity of functional proteins via substrate availability or the chemical environment. Further, in a side-by-side comparison of human and pig decomposition at the same location, soil pH increased under pigs, but decreased under humans [18]. Altered pH and soil chemistry could result in different functions and gene expression in decomposition-impacted soils. Many triacylglycerol lipases have a pH optimum that is neutral to basic [32–34], so cells may be decreasing expression under acidic conditions in human decomposition soils. Availability of lipid species and changes to pH may select for taxa that favor these substrates/pH conditions; for example, Mason et al. (2022) [12] showed a relationship between the abundance of the fungal taxa *Saccharomycetes* and antemortem body mass index (BMI) due to relative proportions of fat and muscle tissue.

Evidence for phased denitrification and nitrification

The human body is a concentrated source of nitrogen that is released into the surrounding soil during decomposition. Expression of common marker genes for nitrogen cycling was altered in decomposition soil and suggested nitrogen transformations during human decomposition are driven by soil oxygen concentrations with hydroxylamine as an important intermediate. We observed low or reduced expression of the nitrification genes *nrrA* and *amoA* between days 12 and 86, during a period when oxygen was reduced to 39% - 85%. This was concomitant with an accumulation of ammonium, which reached a maximum on day 12, and low nitrate indicating that nitrification was inhibited. This period of reduced soil oxygen constraining nitrification was also described in a decomposition experiment with beaver (*Castor canadensis*) carcasses Keenan et al. (2018) [8].

We observed increased gene expression for the enzyme hydroxylamine dehydrogenase (HAO) at day 86 while oxygen was reduced (~85%). This corresponded to simultaneous increases in expression of genes encoding subunits of nitric oxide reductase (*norB*) and nitrous oxide reductase (*nosZ*). Traditionally HAO has been thought to process hydroxylamine to nitrite during nitrification, while nitric oxide reductase and nitrous oxide reductase are enzymes involved in the last two steps of denitrification converting nitric oxide (NO) to dinitrogen gas (N₂). However, recent work suggested hydroxylamine can be converted to nitric oxide (NO), and can interact with multiple phases of the nitrogen cycle [35]. Even though *amoA* expression was shown to decrease during reduced oxygen conditions, *amoA* transcripts were still present and likely able to convert ammonium to hydroxylamine as soil oxygen was not completely depleted during decomposition. Additionally, a previous study reported that the growth of the ammonia oxidizing bacteria *Nitrosomonas europaea* under anoxic conditions lead to accumulation of hydroxylamine in a chemostat bioreactor [36], suggesting anaerobic

ammonium oxidation (anammox) may also be occurring in decomposition soils. However, we did not observe increases in *nirK* expression, which might suggest conversion of nitrite to NO for use in the anammox pathway. NO produced via HAO activity may be used for anammox in these soils; however, the role of hydroxylamine as an intermediate in anammox is still debated [35]. Therefore, our current hypothesis is that hydroxylamine accumulates under anaerobic conditions during decomposition, which can then be converted to NO by HAO. This NO would then be present for anaerobic denitrifying bacteria to convert to nitrous oxide (N₂O) by nitric oxide reductase and finally to N₂ by nitrous oxide reductase. Keenan et al. (2018) [8] noted a brief increase in N₂O emissions during beaver carcass decomposition, which suggests denitrification was occurring during this phase of reduced soil oxygen concentrations.

As soils fully reoxygenated by day 168, we observed increased expression of genes encoding enzymes involved in aerobic nitrification, *amoA* and *nxrR*. Nitrification is an oxygen-dependent process which would convert accumulated ammonium to nitrate; the increase in nitrate concentrations may then serve as a substrate for denitrification. We observed increased expression of marker genes encoding all four enzymes in the complete dissimilatory denitrification pathway (*narG*, *nirK*, *norB*, and *nosZ*) at day 376. Increased expression of nitrification and denitrification marker genes is consistent with the accumulation of nitrite, nitrate, and N₂O after oxygen is reintroduced to soils described in Keenan et al. (2018) [3, 8]. Together, gene expression patterns in our study provide further insight into nitrogen transformations in during vertebrate decomposition, suggesting an important role of hydroxylamine.

Increased expression of bile salt hydrolases

Sulfur is present in various organic molecules, including taurine, a sulfur- and nitrogen-rich compound involved in bile acid formation [22]. Taurine in the human body can be absorbed from the diet or synthesized in the liver [37]. However, taurine is also

produced as a byproduct of the deconjugation of bile salts via bile salt hydrolases (BSHs) present in the anaerobic gut taxa *Lactobacillus* and *Clostridium* [22]. We observed increased expression of genes encoding BSH enzymes between days 12 and 86. Given that increased expression of BSH genes corresponded to the beginning of active decomposition, when decomposition products were observed to enter the soil, and the period of reduced dissolved oxygen in our study, it is likely that taurine accumulation is the result of BSH enzyme activity by anaerobic microorganisms. While we did not measure taurine concentrations in the present study, our results correspond to previous decomposition studies that report accumulation of taurine in various organs and body regions [38–40] and soils [18, 41] during decomposition via metabolomics, and increased relative abundance of *Clostridium* and *Lactobacillus* within the body [42–44] and in decomposition soils [20] via DNA sequencing methods, including in these soils [13].

Taurine can be metabolised through desulfurization via the α -ketoglutarate-dependent enzyme taurine dioxygenase (TauD). Specifically, this enzyme, encoded by the gene *tauD*, converts 2-oxoglutarate and taurine to produce aminoacetaldehyde, succinate, sulfite, and CO₂ [45]. Succinate and sulfite from this reaction can then be used for the citric acid cycle and sulfur metabolism, respectively. Given increased BSH expression in our study and reported taurine accumulation in others, we would expect taurine to be present for microbial metabolism by TauD. However, we observed a general decrease in *tauD* expression between days 12 through 376. This trend was driven by reduced expression of *tauD* transcripts associated with *Proteobacteria*, *Gammaproteobacteria*, and *Actinobacteria* whose relative abundance have been shown to remain consistent or increase during human decomposition [20], suggesting that *tauD* expression is downregulated under decomposition conditions. However, we noted that expression of *tauD* genes associated with fungi and a few *Betaproteobacteria* displayed increased representation at day 58, corresponding to increased expression of

783 bile salt hydrolases (BSH) between days 12 and 86. The reduction in *tauD* expres-
784 sion may be due to increased sulfur availability. We did not measure sulfur species
785 in this experiment; however, others have observed increased sulfur concentrations in
786 decomposition-impacted soils [3, 7, 11]. Thus, sulfur scavenging pathways such as tau-
787 rine desulfurization by TauD [46], whose genes are expressed under sulfur-limiting
788 conditions, likely display reduced expression under sulfur replete conditions. Addition-
789 ally, taurine may be processed through other pathways. For example, taurine can be
790 deaminated by taurine dehydrogenase to produce sulfite and acetyl-CoA for carbon
791 metabolism [45, 47]. Overall, our results suggest that human decomposition has poten-
792 tial impacts on soil sulfur biogeochemistry through deposition of inorganic (sulfate)
793 and organic (sulfur-containing amino acids) sulfur compounds.
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802 Conclusion

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805 This study investigated soil microbial gene expression during human decomposition.
806 Metatranscriptomic analysis of soils from three human individuals shows that decom-
807 position impacted microbial community gene expression profiles, exhibiting functional
808 shifts over one year. This included altered expression of genes involved in lipid, N
809 and S metabolism as microbes processed the nutrient-rich tissues of the human body.
810 Additionally, we noted that functionality within decomposition-impacted soils was
811 still affected after one year and had not returned to starting or background condi-
812 tions. Together, these results show that vertebrate decomposition has lasting impacts
813 on local soil ecosystems, including soil microbial communities. These results have
814 important implications for understanding biogeochemical changes due to vertebrate
815 mortality events in terrestrial ecosystems.
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Materials and Methods

Study design

In February 2018, three deceased male human subjects (hereafter, “donors”) were placed supine on the soil surface at the University of Tennessee Anthropology Research Facility (ARF) and allowed to decompose. Located in Knoxville, TN (35° 56’ 28” N, 83° 56’ 25” W) the ARF is a roughly 2-acre outdoor facility dedicated to studying human decomposition [48]. The soils at the ARF are comprised of the Loyston-Talbott-Rock outcrop (LtD) and Coghill-Corryton (CcD) complexes. LtD soils are a silty clay loam and channery clay overlaying lithic bedrock, while CcD soils are comprised of clay from weathered quartz limestone [13, 48]. A site that had not been previously exposed to decomposition was used for this study.

The decomposition field experiment is fully described in Taylor et al. (2024) [13]. Briefly, experiments were conducted in a block design, where each block consisted of one decomposition site and one control site [13]. In total three blocks, *i.e.*, three donors paired with three respective control sites, were included in the study. Each control site was chosen in a manner to ensure their location was uphill and roughly 2 m away from decomposition sites [13]. Donor internal temperatures were recorded by probes located in the abdomen, while ambient air temperatures were monitored via sensors located roughly 50 cm above the soil surface. Soil temperature and salinity were measured with sensors placed directly underneath each individual (Decagon Devices, GS3) [13]. Donors ranged from 65 to 86 in age and were within 1 kg of each other with regard to weight (90.7 to 91.6 kg); donor BMI varied between 27.7 to 29.6 [13].

875 Sampling and physiochemistry

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877 Decomposition of all subjects was observed for one year. During the one-year study
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879 period, soils were sampled at 20 timepoints chosen to correspond with morpho-
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881 logical stages of decomposition as described by Payne (1965) [49]. Once advanced
882 decay was reached, soils were collected at intervals of 350 accumulated degree days
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884 (ADD), calculated using ambient air temperatures, up to one year. All soil cores
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886 were taken using a 1.9 cm (3/4 inch) diameter soil auger to a depth of 16 cm. Soils
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888 were divided into two depth fractions: 0-1 cm (interface) and 1-16 cm (core) for the
889 analyses reported in Taylor et al. (2024) [13]; the entire 0 to 16 cm core was used
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891 for this current study. Decomposition soils were taken from directly beneath the
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893 cadavers, taking care to not re-sample the same location more than once. At the
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895 time of sampling, soil dissolved oxygen was measured in triplicate using an Orion
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897 Star™ A329 pH/ISE/Conductivity/Dissolved Oxygen portable multiparameter meter
(ThermoFisher) [13].

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900 A subset of 6 study timepoints were chosen for metatranscriptomics analysis. Study
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902 days 0, 12, 58, 86, 168, and 376 were chosen as they represented distinct biogeochemi-
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904 cal phases during decomposition (Supplementary Fig. 5). Study day 0 was chosen as a
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906 baseline sample prior to cadaver placement. Study day 12 was the start of active decom-
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908 position, during the initial bloom when soil microbial activity was rapidly increasing
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910 resulting in the onset of hypoxia: soil ammonium reached maximum concentrations
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912 and soil oxygen was at minimum (approximately 39%). Study day 58 was during a
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914 climax period of sustained elevated microbial activity, characterized by high ammo-
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916 nium concentrations, hypoxia, maximum soil temperatures and minimum soil pH [13].
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918 Study day 86 represented a period of declining microbial activity, declining hypoxia,
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920 and increasing nitrate concentrations. Study day 168 was chosen as nitrate was at its
maximum and soil dissolved oxygen had recovered to 99%. Finally, day 376 was cho-
sen to represent the end of the study, 1 year since cadaver placement. Each study day

was represented by four soil samples for RNA extraction: one pooled control sample which was a mix of the three control locations, plus one sample from each of the three donors, yielding a total of 24 samples for this study.

Soil samples were transported to the University of Tennessee (Knoxville, TN) and processed within 24 hours of collection. Soils were homogenized by hand to remove insect larvae, roots, rocks, and other debris (> 2 mm). A subset of soils were used to measure pH, electrical conductivity (EC), and evolved CO_2 as described in Taylor (2024). Soil nitrogen species (NH_4^+ , NO_3^-) and total carbon (TC) and nitrogen (TN) were measured in all soil samples as described in [13]. Reported values for soil physiochemistry represent the full 16 cm core; estimated by summing interface and core values reported by Taylor et. al, (2024) [13] in 1:16 and 15:16 ratios, respectively. Control values reported here are means of the three experimental controls that were unimpacted by decomposition.

Roughly 10 g of soil was reserved for nucleic acid extraction, placed in a 4 oz. Whirl-Pak™ bag (Nasco), flash frozen in liquid nitrogen, and stored at -80°C until further analysis. Bacterial and fungal community composition was assessed via amplicon sequencing of the 16S rRNA gene and ITS2 region as described in Taylor et al. (2024).

RNA Extraction and Sequencing

RNA was extracted from 2 g of soil using Qiagen's RNeasy® PowerSoil® Total RNA kit. Manufacturer's instructions were followed with a few modifications. Soils became saline during decomposition; therefore, we followed the manufacturer's suggestion and incubated all extracts at -20°C following addition of solution SR4 (step 9) to decrease salt precipitation. All RNA samples were resuspended in 40 μl of Solution SR7. RNA concentrations were assessed fluorometrically using the Qubit® RNA HS assay (catalog no. Q32852) with 1 μl of RNA. DNA contamination was removed by DNase

967 treating RNA extracts twice using Qiagen’s DNase Max® kit in 50 µl reactions. RNA
 968 concentrations were remeasured after DNase treatment. PCR with V4 16S rRNA gene
 969 primers [50, 51] was conducted using RNA extracts as the template to confirm removal
 970 of all DNA prior to sequencing. RNA aliquots were shipped to HudsonAlpha Dis-
 971 covery (Huntsville, AL) for library preparation and RNA sequencing. Dual-indexed
 972 libraries were prepared using the Illumina® Stranded Total RNA prep with riboso-
 973 mal RNA depletion via ligation with Ribo-Zero Plus. Libraries were then pooled and
 974 sequenced on Illumina’s NovaSeq 6000 v4 platform, resulting in demultiplexed fastq
 975 files for each sample.
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983 Bioinformatics

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 985 Illumina sequencing of the 24 libraries yielded a total of 5,073,476,730 reads, or
 986 2,536,738,365 paired reads, with a mean of 105,697,432 paired reads per sample. Read
 987 quality control (QC) was conducted in KBase [52] using Trimmomatic [53]. Paired
 988 fastq files were imported to KBase through Globus. Poor quality reads were removed
 989 (4.7% of all reads), and adapters trimmed via Trimmomatic (v0.36) using default set-
 990 tings and the TruSeq3-PE-2 adapter file, resulting in 4,834,123,062 total reads. After
 991 QC check with FastQC, trimmed libraries were exported as fastq files from KBase
 992 through Globus. Remaining ribosomal RNA was filtered using bbmap (maxindel =
 993 20, minid = 0.93) from the Joint Genome Institute’s (JGI) bbtools suite [54]. Fil-
 994 tering of ribosomal RNA further removed 7.3% of reads, leaving 4,479,804,360 reads
 995 for assembly. Following this step, all non-ribosomal reads from all 24 samples were
 996 merged into one file. Reads were then co-assembled into contigs using the de novo
 997 assembler MEGAHIT (v1.2.9) [55] (-12 -k-min 23, -k-max 123, -k-step 10).
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for a meta-sample (-p meta -f gff). After co-assembly, a total of 6,257,674 gene calls were identified by Prodigal. Next, predicated genes were functionally and taxonomically annotated using eggNOG mapper (v2.1.6) using basic settings to perform a diamond blastp search [58]. From this, 1,048,573 proteins were annotated by eggNOG-mapper (16.7%). Most of the annotated proteins were taxonomically annotated as bacteria (91.3%), followed by eukaryotes (7.6 %), and archaea (0.81 %). Of the 7.6% of eukaryotic proteins, 64.4% (4.9% of all proteins) were annotated as fungi. For this study, genes of interest included all bacterial, archaeal, and fungal proteins, therefore all non-fungal eukaryotic proteins (32,004) were removed prior to downstream analysis. Transcript counts for all genes of interest were obtained by mapping reads from each respective sample to genes of interest obtained from co-assembly using QIAGEN CLC Genomics Workbench 20.0 (<https://digitalinsights.qiagen.com/>). The percent of reads mapped to genes of interest ranged from 21% to 38% between samples, with an average of 31% reads mapped. Gene counts were then combined in a single file and used for downstream analyses in R.

Differential Expression

Transcript counts from all samples were combined in a single workable data file and imported into R for differential expression analysis using the R packages edgeR [59] and limma [60] following a modified pipeline by Phipson et al. (2020) [61]. The transcript count table was imported into R and converted to a DGElist object. Genes without sufficient counts for statistical analysis were removed to increase power using the edgeR function filterByExpr(), using study day as the comparison group.

Raw counts were then log2 normalized and gene expression profiles compared via multidimensional scaling (MDS) and hierarchical clustering. Multidimensional scaling (MDS) was conducted using plotMDS() from the limma package to assess differences between samples. MDS values were extracted from the MDS object, and the first two

1059 dimensions plotted using ggplot2 [62]. We also assessed the relationship between gene
 1060 expression profiles and changes in the soil environment using canonical correspondence
 1061 analysis (CCA). Environmental variables of interest included decomposition time in
 1062 accumulated degree hours (ADH) based on ambient temperatures, ADH based on
 1063 internal gut temperatures, ADH based on soil temperatures, gravimetric moisture,
 1064 pH, electrical conductivity (EC), dissolved oxygen (DO), CO₂ (μmol gdw⁻¹), NH₄ (mg
 1065 gdw⁻¹), NO₃ (mg gdw⁻¹), N %, C %, and CN ratio. First, permutational multivariate
 1066 analysis of variance (PERMANOVA) with adonis() (vegan v2.6.7) [63] was used to
 1067 identify significant soil parameters. Then the vegan functions cca() and scores() were
 1068 applied to run the CCA and extract scores, respectively. Scores for the first two
 1069 dimension were plotted using ggplot2, with loadings extracted from the CCA biplot.
 1070 For differential expression analysis, raw filtered reads were normalized using edgeR's
 1071 trimmed mean of M values (TMM) normalization using the function calcNormFac-
 1072 tors(). TMM normalized reads were then log2 transformed using limma's voom() and
 1073 differential expression assessed. Empirical Bayes shrinkage was used correct to p-
 1074 values for false discovery rates. The topmost up and down regulated genes for each
 1075 comparison, determined by log2 fold change and adjusted p-values, were then reported.
 1076 Expression of certain genes were assessed after performing transcripts per million
 1077 (TPM) normalization and statistical analyses with a combination of analysis of vari-
 1078 ance (ANOVA) and post-hoc Tukey tests. ANOVA across all timepoints were applied
 1079 to hierarchical linear mixed effects models to account for repeated sampling within
 1080 each donor block.

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1098 Data availability

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1100 Raw RNA sequence files from the Illumina Novaseq are available at the National Cen-
 1101 ter for Biotechnology Information's (NCBI) Sequence Read Archive (SRA) as a part
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of [BioProject PRJNA1066312](#) under BioSample accession numbers SAMN45195141-
 SAMN45195164. Additional datasets supporting the conclusions of this article are
 available on [GitHub](#).

Code availability

The code used for analysis and to generate figures are available on [GitHub](#).

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Figures

Figure 1: Microbial gene expression profiles are altered during human decomposition. Multidimensional scaling (MDS) shows gene expression within soils changed as decomposition progressed (A). Canonical correspondence analysis (CCA) shows that environmental variables explained 47.3% of the variation in gene expression profiles (B). Variables in bold red type significantly ($p < 0.05$) explained some of the variation in gene expression profiles as assessed by Permutational Analysis of Variance (PERMANOVA). In both panels decomposition phase of soils (Supplementary Fig. 5) are denoted by symbol shape, while color represents study day. In B, soil physiochemical variable loadings are represented by arrows: Gravimetric water content (GWC), electrical conductivity (EC), pH (pH), dissolved oxygen (DO), respiration (evolved CO_2 $\mu\text{mol gdw}^{-1}$), ammonium (NH_4), and nitrate (NO_3) concentrations (mg gdw^{-1}), percent carbon (%C), percent nitrogen (%N), carbon:nitrogen ratio (C:N), ambient temperature (T_A), and soil temperature (T_S).

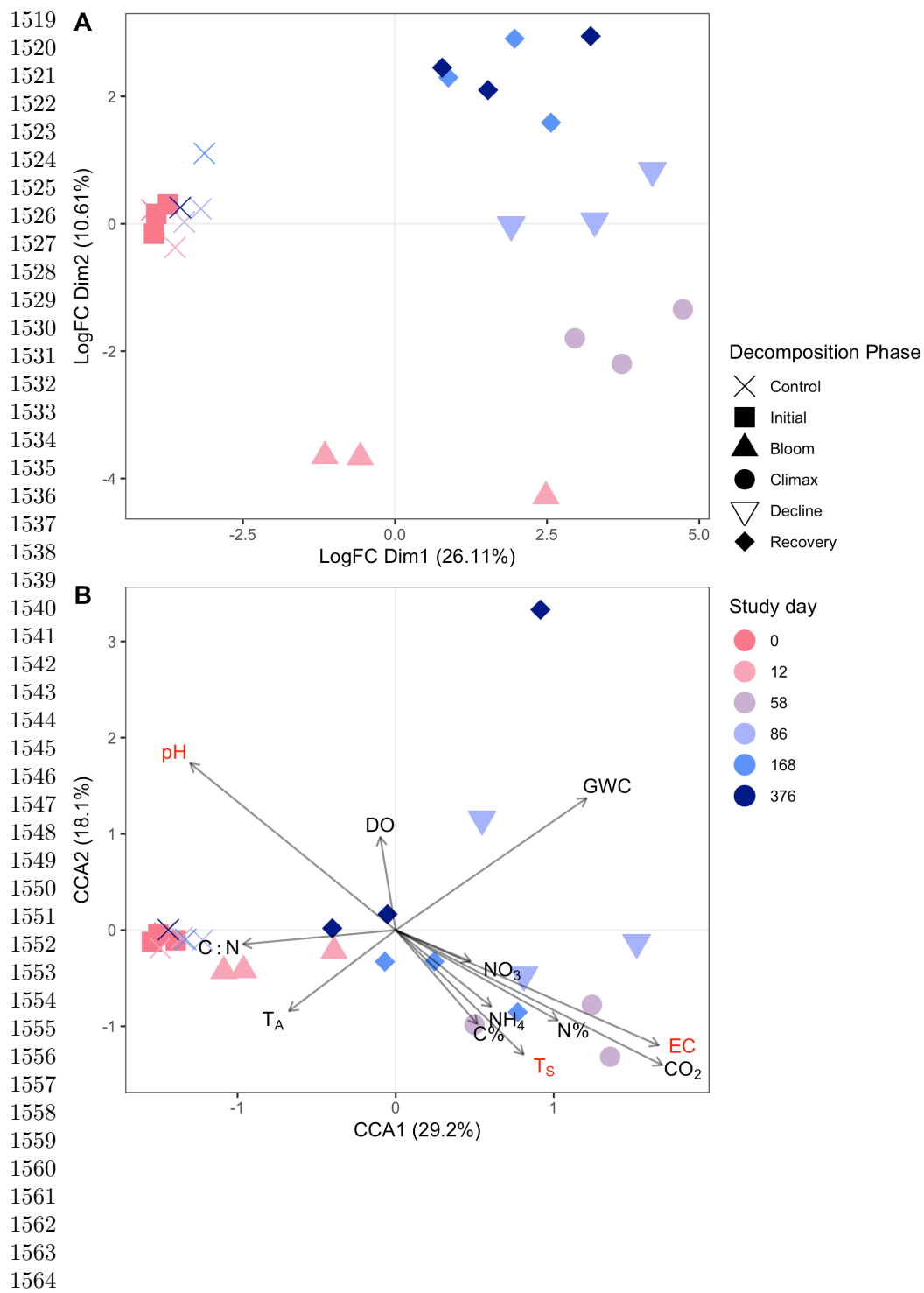


Figure 2: Mean normalized log2 expression of heat shock proteins identified by differential expression analysis comparing decomposition and control soils. Each panel represents a single heat shock gene, labeled with gene names, identified via Prodigal. Symbol color denotes if the sample is a control (CON, green), or one of three individuals: SP1 (orange), SP2 (purple), or SP3 (pink). Error bars are standard error of individual query genes in the top 20 transcripts associated with decomposition soils.

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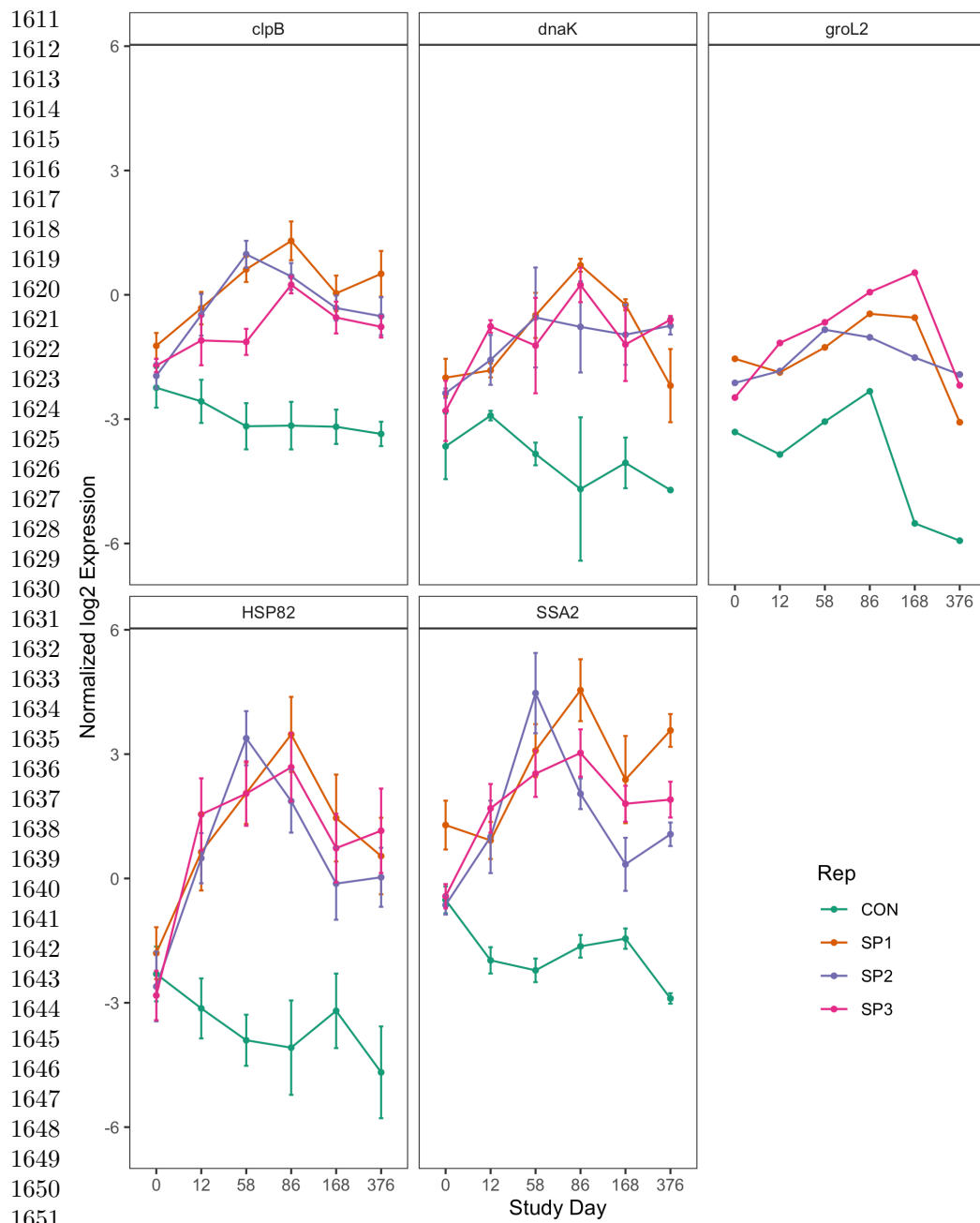
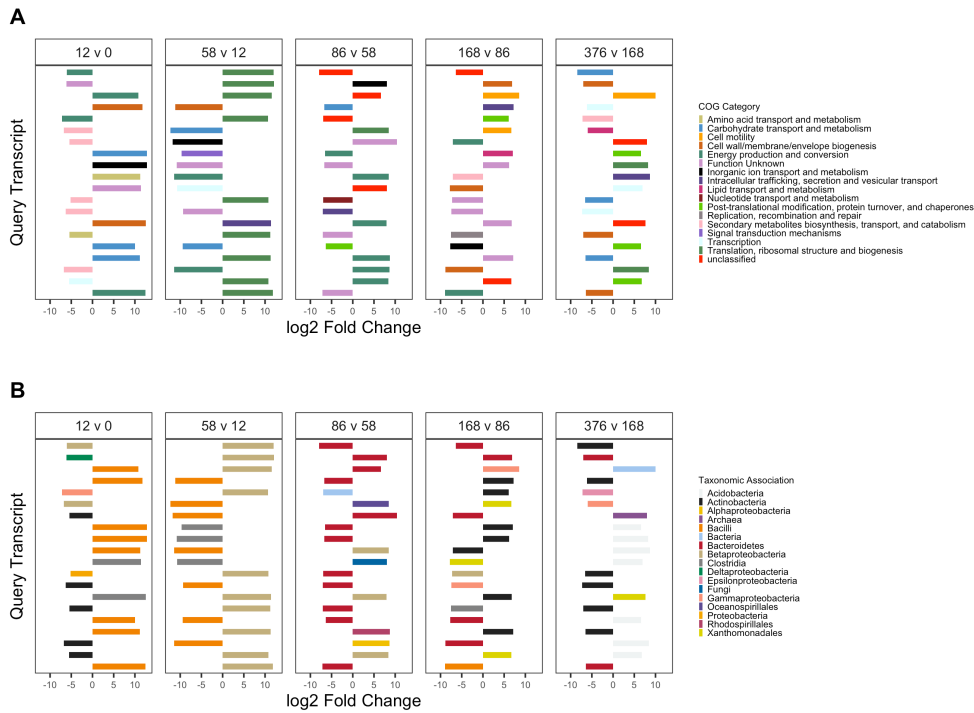


Figure 3: Top twenty up- and down-regulated genes in decomposition soils comparing sequential study days (0, 12, 58, 86, 168, 376) colored by COG functional category (A) and taxonomic annotation (B). Positive values denote increased expression compared to the preceding timepoint, while negative values denote a decrease.



1703 **Figure 4: Mean transcript abundance, in transcripts per million (TPM),**
1704 **of all bacterial (A) and fungal (B) triacylglycerol lipase (EC 3.1.1.3) genes**
1705 **over time.** Abundance of both bacterial ($p = 0.001$) and fungal ($p = 0.015$) lipase
1706 transcripts change significantly over time. Black lines (A, B) report mean and stan-
1707 dard deviation of TPM from three individuals (black line), while gold stars denote
1708 mean TPM in control soils. Letters are the result of post-hoc Tukey tests between
1709 decomposition timepoints. In B, bars show the relative abundance of the fungal
1710 classes *Saccharomycetes*, *Sordariomycetes*, and *Eurotiomycetes*, reported in Taylor et
1711 al. (2024).
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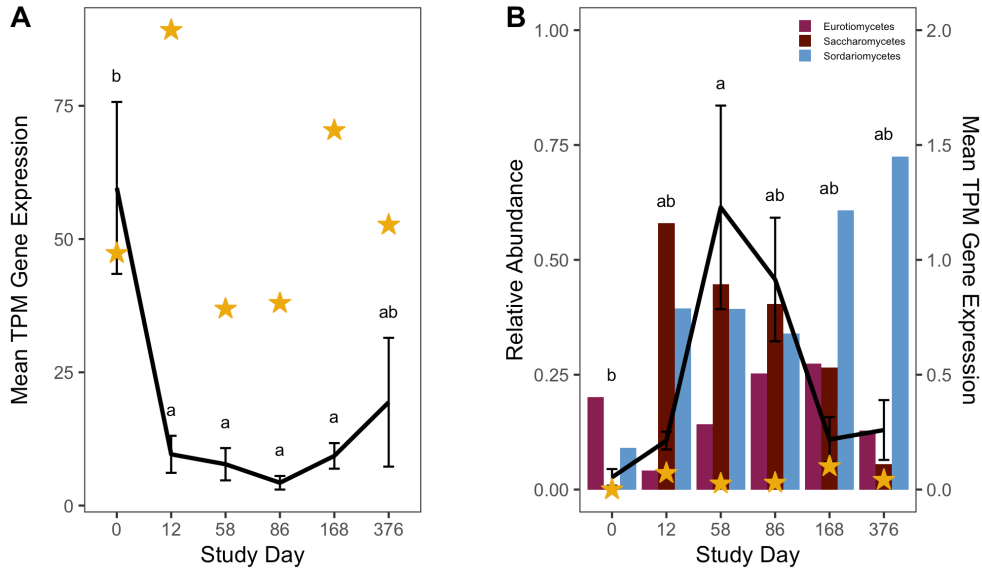
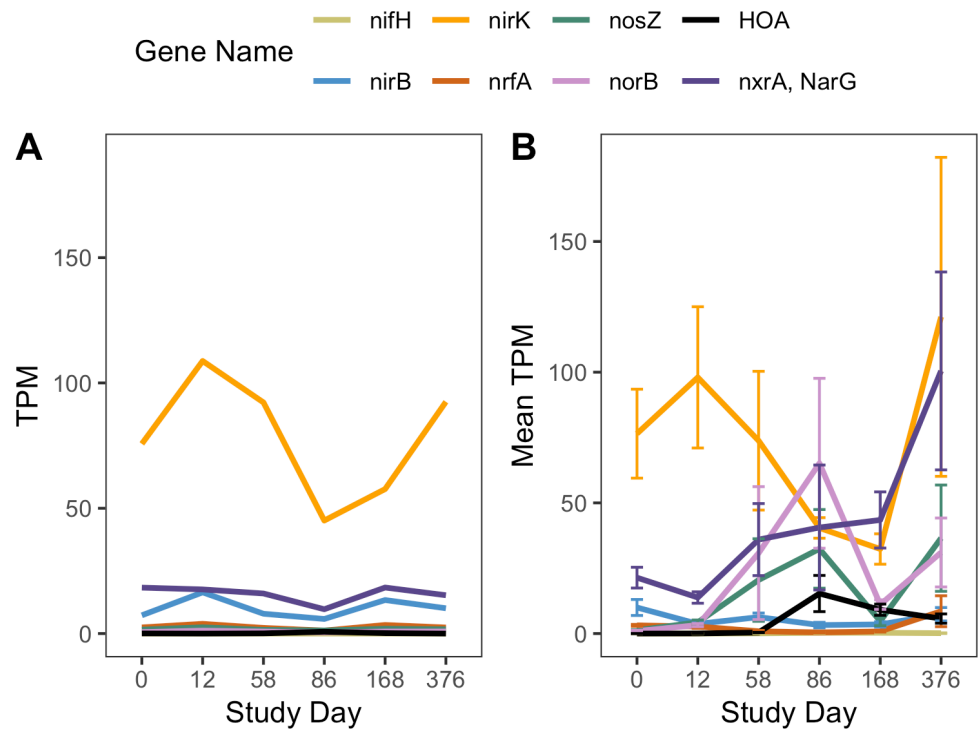
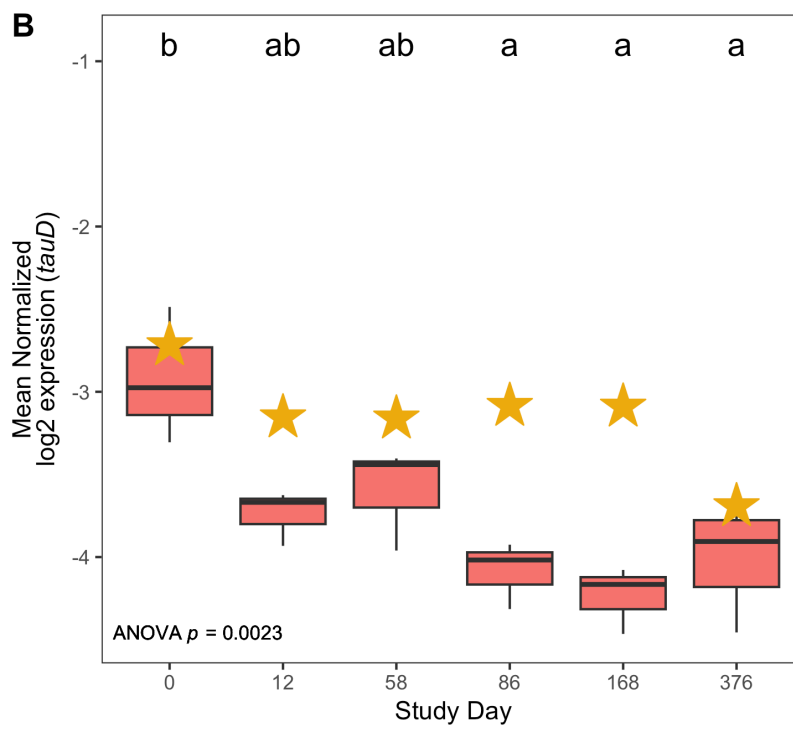
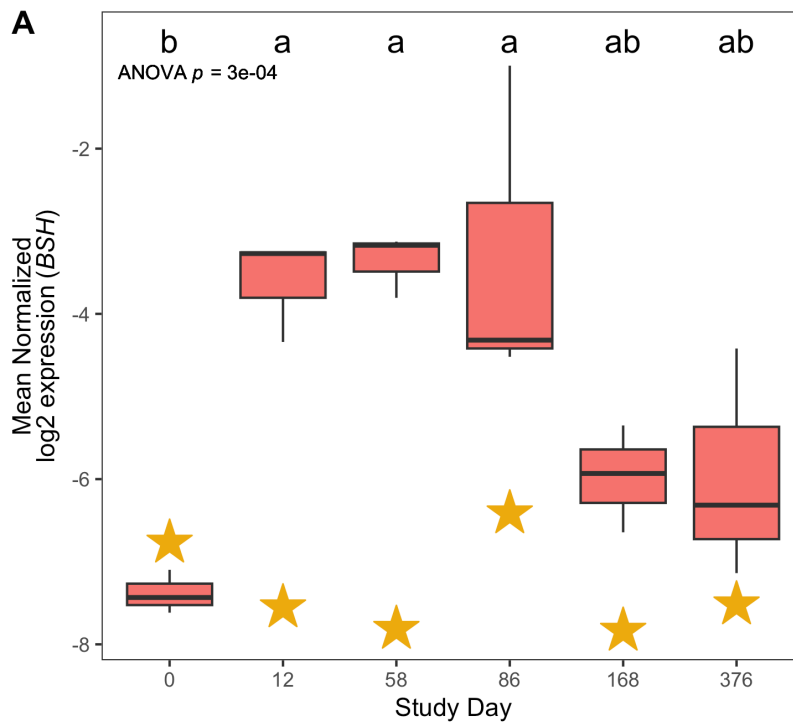


Figure 5: Mean gene expression, in transcripts per million (TPM), of commonly used marker genes for enzymes involved in nitrogen cycling over time in controls (A) and decomposition (B) soils. Data in B represent mean and standard deviation of TPM from three individuals.



1795 **Figure 6: Mean bile salt hydrolase, BSH, (A) and *tauD*, taurine dioxy-**
1796 **genase, (B) log2 normalized expression in controls (gold stars) and**
1797 **decomposition (boxplots) soils.** Boxplots display the 25th and 75th quartiles
1798 and median log2 normalized values between all three individuals at each timepoint.
1799 ANOVA p-value is the result of a hierarchical linear mixed effects model accounting
1800 for repeated measures of each donor block, while letters denote the results of *post-hoc*
1801 Tukey test.

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1887 Acknowledgements

1888

1889 We would like to thank the Forensic Anthropology Center at the University of
1890

1891 Tennessee-Knoxville for their help in setting up field experiments. We would like to
1892

1893 thank Mary Davis for her help in managing the field site and helping to obtain donors

1894 for this work. This research was funded by a National Institute of Justice Award
1895

1896 (DOJ-NIJ-2017-R2-CX-0008) to LST and JMD.

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