Soybean and cotton spermosphere soil microbiome shows dominance of soil-borne

copiotrophs

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Abstract

The spermosphere is the transient, immediate zone of soil around imbibing and germinating seeds, rich in microbial activity. It represents a habitat where there is contact between seed-associated microbes and soil microbes, but is studied less compared to other plant habitats. Previous studies on spermosphere microbiology were primarily culture-based or did not sample the spermosphere soil as initially defined in space and time. Thus, the objectives of this study were to develop an efficient strategy to collect spermosphere soils around imbibing soybean and cotton in non-sterile soil and investigate factors contributing to changes in microbial communities. The method employed sufficiently collected spermosphere soil as initially defined in space (3-10 mm soil around a seed) by constraining the soil sampled with an cork borer and confining the soil to a 12-well microtiter plate. Spermosphere prokaryote composition changed over time and depended on the crop within six hours after seeds were sown. By 12 to 18 hours, crops had unique microbial communities in spermosphere soils. Prokaryote evenness dropped following seed imbibition and the proliferation of copiotroph soil bacteria in the phyla Proteobacteria and Firmicutes. Due to their long history of plant growth promotion, prokaryote OTUs in *Bacillus*, *Paenibacillus*, *Burkholderia*, *Massilia*, *Azospirillum*, and *Pseudomonas* were notable genera enriched in the cotton and soybean spermosphere. There was no consistent evidence that fungal communities changed like prokaryotes. However, fungi and prokaryotes were hub taxa in cotton and soybean spermosphere networks. Additionally, the enriched taxa were not hubs in networks, suggesting other taxa besides the copiotrophic may be important for spermosphere communities. Overall, this study advances knowledge in the assembly of the plant microbiome early in a plant's life, which may have plant health implications in more mature plant growth stages.

Introduction

When a seed is sown, it imbibes water and releases nutrient-rich seed exudates that fuel interactions between soil and seed-associated microbes in a plant habitat called the spermosphere (Nelson 2004; Nelson 2018; Schiltz et al. 2015; Shade et al. 2017; Windstam and Nelson 2008). Seed exudates have long been recognized to stimulate microbial growth, including a direct link to facilitating pathogen growth chemotactically towards seeds (Nelson 2004; Short 1976; Slykhuis 1947; Stanghellini 1971). The spermosphere or "spermatosphere" was first described by Slykhuis (1947), who observed the inhibition of a fungal pathogen by three fungal species around a germinating seed. Verona (1958) defined the same habitat as a "zone of elevated microbial activity" around a germinating seed. Nelson (Nelson 1986, 1988; 2004; van Dijk and Nelson 2000; Windstam and Nelson 2008) more formally defined the spermosphere as "the short-lived, rapidly-changing and microbiologically dynamic five-to-ten-millimeter zone of soil around a germinating seed", which is the definition we adhere to in this study. Despite its importance for plant health outcomes, the spermosphere is less studied than other plant-associated habitats such as the rhizosphere or phyllosphere (Aziz et al. 2022; Schiltz et al. 2015; Shade et al. 2017).

Seed germination occurs in three distinct phases. Phase I is a physical process characterized by seed imbibition and fast carbon-rich exudate release into the soil hours after seeds are sown. The highest levels of exudate release are completed in as little as six hours (Lynch 1978; Nelson 2004; Simon and Raja Harun 1972). The initial phase of exudate release is followed by a plateau characterizing phase II, then radical emergence, which begins the formation of the rhizosphere, and more exudate release in phase III (Nelson 2004; Schiltz et al.

2015).

The spermosphere represents a critical zone for establishing vertically inherited seed

microbes, and horizontal interactions between soil and seed-associated microbes (Chesneau et al.

2022; Chesneau et al. 2020; Rochefort et al. 2021; Shade et al. 2017; Simonin et al. 2022). The

outcome of these interactions can affect the life or death of the plant soon after sowing seeds

(McKellar and Nelson 2003; Windstam and Nelson 2008). For example, soil-borne *Pythium* can

fully colonize and kill germinating seeds of various crop species within 12-24 hours (Hayman

1969; Nelson 1986, 1988; Stanghellini 1973).

Spermosphere pathogens still cause millions of dollars in crop loss yearly (Allen et al.

2017; Bradley et al. 2021; Mueller et al. 2020). Because of enhanced genetics and other factors,

seed ranks first or second in operating costs borne by soybean and cotton farmers each year

(USDA-ERS 2022). Additionally, trends toward earlier planting dates, increased frequency of

heavy pulsed rain events, and variable temperature conditions experienced by farmers at planting

can create soil moisture and temperature conditions that stress the germinating seed (Morris et al.

2021; Munkvold 1999; Roth et al. 2020). Conservation tillage (low or no-till) can also lead to

harboring plant pathogens on plant debris left in the field from the previous growing season

(Bockus and Shroyer 1998). Consequently, the protection of seeds from pathogens that specialize

in spermosphere colonization is vital to improved crop productivity.

Seed and seedling pathogens are primarily managed with chemical seed coatings

containing fungicides and oomicides. However, improved knowledge on spermosphere

microbiology and ecology would support the successful inclusion of alternative strategies to

chemical seed treatments. For example, biocontrol of *Pythium* from seed-applied *Enterobacter*

cloaceae could be achieved by metabolizing long-chain fatty acids, which otherwise stimulated

the germination of *Pythium* sporangium (Kageyama and Nelson 2003; van Dijk and Nelson

2000; Windstam and Nelson 2008). Studies on the spermosphere have either been culture-based

or have more recently focused on the contribution of the indigenous seed microbiome by using

sterile or soilless growth conditions or pre-imbibed or pre-germinated seeds, which do not

sample the spatial and temporal properties of the spermosphere soil (Barret et al. 2015;

Johnston-Monje et al. 2016; Moroenyane et al. 2021; Rochefort et al. 2021). Indeed, natural

seed-associated epiphytes and endophytes compete with pathogens (Smith et al. 1999; Torres-

Cortes et al. 2019). While commendable, these studies largely ignore the influence of the initial

seed exudate release on the spermosphere soil microbes. Therefore, a mechanistic understanding

of the complex interactions in spermosphere soil will aid in novel treatments for seed and

seedling pathogens and help our understanding of plant microbiome assembly.

However, one major challenge in studying spermosphere soil using high-throughput

culture-independent techniques may be a lack of a quick and efficient method of collecting

spermosphere soil (Schiltz et al. 2015). Here, we aimed to capture changes in microbial diversity

in the spermosphere soils as soybean and cotton seeds underwent phases I and II of seed

germination (i.e., pre radical emergence). We sampled the spermosphere soil of cotton and

soybean by constraining the soil zone within wells of a 12-well plate and sampling precisely

three to six millimeters of soil around an imbibing seed with an appropriately sized cork borer,

extracted DNA, and sequenced the 16S and ITS from cotton and soybean spermosphere soil. We

hypothesized that seeds would imbibe water rapidly and follow previously established phases of

exudate release, which would alter microbial diversity and co-occurrence patterns. We also

hypothesized that spermosphere soils would be distinct based on crop species. Therefore, the

objectives of this study were twofold: 1) characterize the bacterial and fungal microbial

communities associated with cotton and soybean spermosphere soil compared to control soil and

2) determine how microbial diversity and co-occurrences change over time as a seed imbibes

water.

Materials and Methods

Soil collection and preparation. The soil used in this study was collected from a field used for

cotton, soybean, and corn rotation from Prattville Agricultural Research Unit in Prattville,

Alabama (32.42533, -86.4452), since this soil showed consistent emergence of both cotton and

soybean in preliminary experiments (*data not shown*), and was not known to contain a high

abundance of any specific seedling pathogen. Approximately three liters of soil from the top ten

centimeters were collected and transported to the lab. The soil was sieved to eliminate stones and

pebbles and air-dried for 24 hours to ensure homogeneity in water content. The soil was used

immediately after air drying. Six to seven grams (6 ml) of soil was transferred to each well of the

12-well microtiter plates (VWR American cat no.:10861-556, USA), containing three one-

millimeter holes in the bottom of all wells for drainage. Each well in the 12-well microtiter plates

measured a total volume of 6.8 ml, depth of each well of a 12-well plate was 15 mm with a width

of 23 mm. Each well containing soil was watered with 1.5 ml of sterile water (25% soil

moisture), and the water was allowed to circulate for one hour before the seeds were sown.

Nontreated Williams-82 soybean or nontreated delinted Delta Pine 1646 B2XF cotton

were used in this study and were sorted to discard discolored seeds or seeds with cracked seed

coats (Nelson 1986). The weight of individual dry seeds was recorded before use and after

imbibition to record how much water was imbibed. The initial weight of soybean seeds was

between 170 and 250 mg, and cotton seeds weighed between 60 and 110 mg. The average size of

soybeans used was between five to eight millimeters in diameter and spherical. Cotton seeds

were more oblong, three to four millimeters in diameter and ten millimeters long. Seeds were

surface-sterilized by soaking in 6% bleach solution for 10 minutes in a sterile petri-dish and

washed three times with sterile distilled water. Seeds were surface sterilized to maximize the

effect of seed exudates on the growth of microbes from the soil. Six replicate seeds were sown

into the center of individual wells, halfway into the 15 mm depth of the well, using flamed

forceps. Wells containing only soil without a cotton or soybean seed were used as a control. The

12-well microtiter plates were placed in a planting tray covered with a lid to keep the soil from

drying. Planting trays containing 12-well microtiter plates were placed inside a growth chamber

at 25°C.

Collection of spermosphere. Spermosphere soil samples were collected at 0, 6, 12, and 18 hours

after sowing. Wells containing control soil were sampled as a control, and are hereafter referred

to as control soil. Spermosphere soil and control soil samples were collected using an 11 mm

cork borer cleaned of soil with 70% ethanol and flame sterilized between samples. The 11 mm

cork borer was specifically used since the spermosphere is defined as the first 5-10 mm of soil

around a germinating seed (Nelson 2004) and allowed soil collection within this range based on

the seed sizes stated previously. Therefore, given the size of the well, the volume of soil used,

and the seed sizes, the spermosphere soil sampled consisted of three to six millimeters on either

side of a soybean seed and seven to ten millimeters above and below a soybean. Similarly, the

spermosphere soil sampled for cotton consisted of seven to eight millimeters on either side and

five to ten millimeters above and below the seed. A diagram of the sampling procedure for

soybean is shown in Figure 1. In preliminary experiments, bacterial populations in spermosphere

soils sampled with this method increased significantly by 1.15 log in soybean and about 0.8 log

in cotton compared to control soil (Supplemental Figure 1).

Spermosphere soil containing the seed inside the core within the cork borer was

transferred into sterile envelopes, and 0.25 ml was immediately transferred to 2 ml disruptor

tubes (Omega Bio-Tek E.Z.N.A Soil DNA; Norcross GA), then stored at -80ºC until DNA

extraction. The remaining soil clinging to the seed was washed off, the seed blotted dry of excess

water, and the weight of the seed was recorded after sample collection and compared to the

initial individual seed weight to determine the water imbibed by each seed.

DNA extraction, amplification, and sequencing. The total DNA was extracted from the

spermosphere and control soils following the manufacturer's recommendation. Amplification

and library construction of 16S or ITS rDNA was performed with a three-step Polymerase Chain

Reaction (PCR) (Lundberg et al. 2013; Longley et al. 2020). Briefly, the 16S region of the

ribosomal DNA (rDNA) was amplified using the forward and reverse primers 515F and 806R

(Caporaso et al. 2011). Amplification of ITS used the primers ITS1F and ITS4. Following the

amplification of the respective rDNA regions, the amplicons were linked to variants of

frameshift primers, and then a 10 bp barcode was added for sample identification. Library

negative controls consisted of DNA extraction without soil and no template PCR water controls.

The ZymoBIOMICS microbial community DNA standard (Zymo Research, Irvine, CA) was

used as a positive control mock community. A fungal synthetic mock community was used as a

positive control for fungi (Palmer et al. 2018). DNA amplification was confirmed with gel

electrophoresis, and successfully amplified libraries were normalized using SequalPrep™

Normalization Plate Kit (Thermo Fisher, USA). Normalized amplicons were then pooled and

concentrated 20:1 using the 50K Dalton Millipore filters (Sigma-Aldrich, USA). The pooled

library was cleaned using AMPure XP beads at a ratio of 0.7X (Beckman Coulter, USA).

Cleaned amplicon pools were verified by gel electrophoresis, quantified using a Qubit

fluorometer (Thermo Fisher, USA), and sequenced on an Illumina MiSeq 2x300 bp using the v3

600 cycles kit at SeqCenter LLC (Pittsburgh, PA). Primers and cycling parameters to construct

libraries were the same as in Noel (2022).

Read processing. The quality of demultiplexed reads was assessed using the FastQC, and primer

sequences were removed using cutadapt 4.0 (Martin 2011). Prokaryote 16S V4 sequences were

merged using VSEARCH 2.21.1 (Rognes et al. 2016). Only forward fungal ITS1 reads were

used since reverse reads did not overlap. Fungal reads were trimmed to remove the conserved

18S regions. Reads were then truncated to equal length (fungi 200 bp, prokaryote 300 bp) and

quality filtered using VSEARCH 2.21.1 with an expected error threshold of 1.0. Singletons were

removed and reads *de novo* clustered based on 97% identity into prokaryote OTUs (pOTUs) or

fungal OTUs (fOTUs) using USEARCH v11.0.667, which includes a chimera detection and

removal step (Edgar 2010; Edgar et al. 2011). The resulting pOTUs were aligned using MAFFT

v7.505 (Nakamura et al. 2018), and a phylogenetic tree was estimated using FastTree v2.1.20

(Price et al. 2010), then midpoint rooted with FigTree v1.4.4 (Rambaut 2018). Taxonomy was

assigned to resulting pOTUs using the SINTAX algorithm (Edgar 2016) against the SILVA

138.1 database (Quast et al. 2013). Fungal taxonomy was assigned using the ribosomal database

project's Naïve Bayesian Classifier algorithm against the UNITE fungal ITS database version

9.0 (Nilsson et al. 2019).

Data analysis. Data were primarily analyzed using phyloseq v.1.34.0 (McMurdie and Holmes

2013) and vegan v2.5-7 (Oksanen et al. 2022) of the statistical software R v.4.2.2. All plots were

generated using the data visualization package 'ggplot2 v.3.3.5' (Wickham 2016). Contaminant

OTUs detected in the negative controls were removed with decontam v1.10.0 (Davis et al. 2018).

Samples with less than 10,000 were discarded. Fungal samples with less than 1000 reads were

discarded due to low sequencing coverage.

Richness, Pielou's evenness (Pielou 1966), and Faith's Phylogenetic Diversity (Faith

1992) were used to determine within sample diversity differences in diversity using Kruskal-

Wallis one-way analysis of variance. Read counts were then normalized using the cumulative

sum scaling with metagenomeSeq v1.32.0 (Paulson et al. 2013) and subjected to principal

coordinate analysis based on Bray-Curtis distances for fungi and prokaryotes or Weighted

Unifrac distances for prokaryotes only. This analysis was followed by a Permutational Analysis

of Variance (PERMANOVA) implemented with the 'adonis2' function to determine the

differences in centroids of the prokaryote or fungal communities across time points and soil

versus spermosphere. Differences in multivariate dispersion were also evaluated using the

'betadisper' function.

Differential abundance analysis was conducted with Analysis of Compositions of

Microbiomes with Bias Correction version 2 (ANCOM-BC2) (Lin and Peddada 2020).

Significantly different OTUs were detected based on Holm-Bonferonii corrected p-values. Then,

microbial co-occurrence networks with prokaryotes and fungi were constructed using SpiecEasi

v1.1.2 (Kurtz et al. 2015) and compared between soybean spermosphere soil, cotton

spermosphere soils, and control soil using NetCoMi v.1.1.0 (Peschel et al. 2021). For network

construction, spermosphere soil samples and soil samples without seeds at 12 and 18 hours were

filtered to a common set of taxa with a relative abundance above 0.001% and occupancy above

90%. Co-occurrence association matrices were estimated using the Meinshausen and Bühlmann

algorithm with the 'nlambda' set to 100, sampled 100 times, and the 'lambda.min.ratio' set to 10^{-1}

. All resulting networks contained stability values of 0.048 or above, close to the 0.05 StARS

algorithm stability target. Association matrices for spermosphere soils or control soil were

compared using the 'netAnalyse' function from NetCoMi. Hub taxa were identified based on

eigenvector centrality values above the 95% quantile of a fitted log-normal distribution.

Comparison of the hub taxa composition was based on the Jaccard similarity index.

The Data files and scripts used for this analysis are available on GitHub

(https://github.com/Noel-Lab-Auburn/SpermosphereMicrobiome2022). Raw sequence reads

were deposited to the sequence read archive with the accession number PRJNA925866.

Results

Sequencing outputs. Mock OTUs for the fungi and prokaryotes made up 99.9% of the

composition of the positive controls, indicating minimal cross-contamination. Nine prokaryotic

OTUs were filtered after detection in negative control samples, resulting in 2,090,814 16S V4

reads of 8088 OTUs across 71 samples with a median read depth of 29,237 reads per sample.

Nineteen fungal OTUs were detected in negative controls and taken out, resulting in 2,534,301

ITS1 reads with 1904 fungal OTUs across 69 samples and a median read depth of 37,933 reads

per sample. Rarefaction curves indicate that much of the diversity was adequately captured

(Supplemental Figure 2).

Prokaryote community dominance correlates with water imbibition. The individual

measurement of seed weight for soybean and cotton seeds before and after spermosphere

collection indicate that water was imbibed from the surrounding soil (Fig. 2a). Overall, soybean

seeds imbibed more (250-300 mg) than cotton seeds (50-80 mg) and both seeds increased in seed

weight within the first six hours indicating imbibition within this timeframe, then a plateau after

six hours.

We observed a reduction in prokaryote evenness (or increase in dominance) over time in

spermosphere soils but not in control soil (Fig. 2b). At time-point 0, there was no significant

difference in the evenness of prokaryote taxa ($P = 0.18$), as expected. At 6 hours, we observed a

reduction in the evenness of prokaryote taxa in soybean spermosphere soils, compared to cotton

spermosphere ($P = 0.004$) or control soil ($P = 0.002$). At 12 and 18 hours, both cotton and

soybean spermosphere samples had significantly reduced evenness compared to control soil and

each other ($P \leq 0.009$). Prokaryote evenness was significantly negatively correlated with water

imbibition, meaning that as seeds imbibed water and released exudates, prokaryote communities

became more dominant (cotton r = -0.58, $P < 0.001$; soybean r = -0.57, $P = 0.004$) (Fig. 2c).

However, the crop did not alter prokaryote richness or phylogenetic diversity compared to

control soil. Prokaryote richness and phylogenetic diversity dropped significantly over time

regardless of habitat ($P < 0.001$). Additionally, there was no consistent evidence that fungal

richness or evenness was altered in spermosphere samples compared to control soil. Still, a few

samples of soybean spermosphere soil and control soil dropped in evenness after 18 hours

(Supplemental Figure 3), becoming more dominant in one fOTU2 *Fusarium* (Supplemental

Figure 4).

Spermosphere prokaryote composition depends on the crop. Spermosphere soils had

different prokaryote community compositions than control soil. A visualization of the change in

the most abundant prokaryote composition over time is shown in Figure 3. Prokaryote

communities were driven by habitat ($P < 0.001$), time since sowing ($P < 0.001$), and interaction

of both factors ($P < 0.001$) (Supplemental Table 1). The interaction prompted a closer look into

the differences observed between crops by splitting the data by time-point (Figure 4; Table 1). At

0 hours, no significant difference in prokaryote communities existed between bulk soil compared

to soybean and cotton spermosphere, as expected (Bray-Curtis, $P = 0.452$; Weighted Unifrac, P

$= 0.192$). However, as early as 6 hours, we observed significant differences between control soil

and spermosphere soil samples (Bray-Curtis, $P < 0.001$; Weighted Unifrac $P < 0.001$).

Differences were extended through 12 hours (Bray-Curtis, $P < 0.001$; Weighted Unifrac $P <$

0.001) and 18 hours (Bray-Curtis, $P < 0.001$; Weighted Unifrac $P < 0.001$), where it was clear

the spermosphere formed unique community compositions within soybean or cotton. Further,

differences in multivariate dispersions were not observed supporting true differences in centroids

rather than group dispersions (Table 1). This same trend was not observed with fungi. Time did

alter fungal community composition ($P = 0.01$), but there was no evidence that soybean or cotton

altered fungal community composition compared to control soils ($P = 0.09$).

Enriched prokaryotes in the spermosphere have unique and shared taxa among crops.

Differential abundance analysis determined sets of pOTUs significantly enriched in the

spermosphere of cotton and soybean compared to control soil (Figure 5a; Supplemental Table 2.

Ninety-four percent of the enriched taxa belonged to Proteobacteria (57%, n = 27) and

Firmicutes (36%, n = 17). The remaining three belonged to the Actinobacteria. Within the

Proteobacteria, the enriched taxa were spread across ten prokaryote families, with the most

enriched taxa in the Oxalobacteraceae (41%, n = 11). The majority of the enriched Proteobacteria

were unidentified at the genus level (n = 12) but included *Massilia* (n = 3), *Noviherbaspirillum*

(n = 2), *Burkholderia*/*Paraburkholdaria* (n = 2), *Aquabacterium* (n = 1), *Pseudomonas* (n = 2),

Cupriavidus (n = 1), *Pantoea* (n = 1), *Paucimonas* (n = 1), *Rubellimicrobium* (n = 1), and

Azospirillum (n = 1) (Figure 5b). Within the Firmicutes, all but one pOTU belonged to the

Bacilli class with the genera *Paenibacillus* (n = 8), *Bacillus* (n = 4), *Brevibacillus* (n = 1),

Exiguobacterium (n = 1), and *Tumebacillus* (n = 1). Many enriched taxa were shared between

cotton and soybean (n = 18), indicating that similar taxa take advantage of releasing exudates

from seeds (Figure 5b). All the enriched pOTUs were present in control soil samples meaning it

was unlikely they originated from the seed but were present in the soil and proliferated upon

exudate release from the seeds.

Cotton and soybean spermosphere networks are more connected and have distinct

microbial hub taxa. Cotton and soybean spermosphere networks were compared to each other

and to the control soil to determine if they contained different topologies, different sets of

network hubs, and the centrality of spermosphere-enriched taxa. Overall, network topology

parameters were similar between networks except for the number of separate components. In

other words, the spermosphere soil networks mainly consisted of one more prominent

component and fewer disconnected sub-networks than the control soil network (Figure 6). For

example, the control soil network contained 30 components and 80 nodes within the largest

component. Soybean and cotton spermosphere soil networks had more nodes within the largest

component (cotton = 136, soybean = 121). However, control soil and a slightly higher positive

edge percentage (61% without a seed, 58% cotton spermosphere, 52% soybean spermosphere)

(Supplemental Table 3).

324   Comparison between central nodes was significantly different, indicating that the hub

taxa were different between networks given the same sets of taxa used to construct the networks

(Table 2). Hub taxa for cotton consisted of six prokaryote OTUs and two fungal OTUs. Cotton

prokaryote hubs consisted of three Archaea OTUs in the Nitrososphaeraceae family (pOTU1009,

pOTU19, and pOTU12), two OTUs from the Gaiellales (pOTU46, pOTU119), and pOTU29

Massilia (Figure 6a; Table 3). Fungal cotton hubs were fOTU56 *Fusarium chlamydosporum*, and

a yeast fOTU36 *Hannaella oryzae* (Figure 6a; Table 3), whereas the soybean network had three

fungal hub taxa: fOTU64 *Helicoma*, fOTU10 *Bartalinia pondoensis*, and an unidentified Fungus

fOTU36 (Figure 6b; Table 3). Prokaryote hub taxa in the soybean spermosphere network

contained pOTU11 *Sphingomonas*, pOTU132 Nocardioides, pOTU1559 Chloroflexi TK10,

pOTU36 Angustibacter, and pOTU349 *Methylobacterium*/*Methylorubrum* (Figure 6b; Table 3).

The network from the control soil contained only fungal hubs, different than the identities of

spermosphere fungal hubs except for fOTU36 *Hannaella oryzae*. Spermosphere-enriched taxa

included in the network analysis were not hub taxa indicating that although enriched in a

spermosphere, other microbial taxa besides the enriched taxa play an essential role in

maintaining spermosphere network structure (Figure 6).

Discussion

342   To our knowledge, this is the first study to use culture-independent sequencing to study

soybean and cotton spermosphere soil microbiomes during the first phases of seed germination.

The advancement that allowed this was the method that constrained non-sterile soil to wells

within a 12-well plate and sampled around an imbibing seed with a cork borer. The technique

enabled the precise and efficient collection of spermosphere soils as defined in space and time

(Nelson 2004), which we believe this represents a more realistic spermosphere habitat. The focus

on spermosphere soil in the first phases of seed germination differs from other studies that pre-

imbibe or pre-germinate seeds in axenic conditions. We hypothesized and observed a rapid

increase in water imbibition followed by a plateau characterizing phase I and phase II of seed

germination. Prokaryote community structure changed in as little as six hours for soybean and

twelve hours for cotton. We did observe that crops had unique prokaryote community structures

in the spermosphere that were distinct from the control soil, typified by differences in network

hub taxa and network topologies. The differing hub taxa demonstrate that others besides the

enriched taxa are integral to each crop's spermosphere community structure. However, despite

the differences in composition and hub taxa, among the most important observations was the

commonality in the enriched copiotrophic taxa with a long history of benefiting plant growth,

such as *Bacillus*, *Paenibacillus*, *Burkholderia*, *Massilia*, *Azospirillum*, and *Pseudomonas*.

In this study, we further define the development of the spermosphere of cotton and

soybean at six to twelve hours after sowing, which aligns with previous studies of increased

spore germination and full colonization of cotton seeds by *Pythium ultimum* twelve hours after

sowing (Nelson 1986, 1988). We observed an increase in water imbibed by both cotton and

soybean seeds in the first six hours, which is consistent with previous reports that documented

increased water imbibition and exudation within the first few hours after sowing (Simon and

Raja Harun 1972). Imbibition ceased representing phase II of germination, indicating the

saturation of nutrient reserves and synthesis of products required for the extending radicle

(Nelson 2004).

Similar to several other studies, we observed the soil microbes respond to seed exudates

and dominated the spermosphere microbiomes (Buyer et al. 1999; Hardoim et al. 2012; Ofek et

al. 2011). We observed changes in phylogenetic dissimilarity between crops, and since

phylogenetically similar species are more likely to share ecological characteristics and functional

traits (Morrissey et al. 2016), it may be expected that the spermosphere communities in our study

changed in a functional capacity as in Buyer et al. (1999). However, we observed varying

spermosphere prokaryote composition in different plant species, which may highlight the

importance of sample collection at the initial stages of seed germination and imbibition rather

than at later hours potentially after radicle emergence. Additionally, as in Buyer et al. (1999),

future studies should also include other soils with inherently different communities to understand

the contribution of different soil microbial pools to forming the spermosphere.

The difference between crops may have also been due to differences in the amount of

water imbibed. We noticed that soybean imbibed more than cotton seeds, likely due to seed size

(Soldan et al. 2021; Vančura 1972). Vančura and Hanzlíková (1972) demonstrated increased

quantities of seed exudates as seed size increased. Different varieties of common bean have been

shown to differ in the amount of seed exudates, with larger seeded varieties releasing more

exudates (Kato et al. 1997). Thus, we speculate that the greater and faster turnover in microbial

communities of the soybean spermosphere compared to cotton may be due to the larger size of

soybean seeds and increased exudation, which potentially supported a larger habitat for the

microbes to occupy. It also leaves the question if microbial communities would have converged

on similar compositions if a later sampling point was included.

Regardless, as a result of water imbibition and seed exudation, we observed a change in

dominance in the spermosphere microbiome over time with both crops. Ota et al. (1991) showed

specific nitrogen-fixing bacteria had increased dominance in the spermosphere of cocklebur

seeds but not in soil. Upon revealing enriched taxa in soybean and cotton spermosphere soils, we

found some commonalities. Importantly, Bacilli was enriched in both crop's spermosphere soils.

Since these *Bacilli,* including *Tumebacillus*, *Paenibacillus,* and *Bacillus,* have historically been

associated with plant growth promotion and disease protection and have commercial potential,

they were notable. It indicates their ability to utilize seed exudates quickly for growth. Seed

exudates have been reported to induce chemotaxis, seed colonization, and biofilm formation of

B. amyloliquefaciens (*velezensis*) by enhancing active cell division (Yaryura et al. 2008).

Paenibacillus polymyxa isolated from wheat and peanut rhizosphere increased the survival of

Arabidopsis thaliana in the presence of the oomycete pathogen *Pythium aphanidermatum* when

applied as root treatment (Timmusk et al. 2009). Identifying these enriched taxa is important for

prioritizing future work on a mechanistic understanding of the spermosphere microbial

interactions that will improve the development of efficacious biologically based disease solutions

(Nelson 2004; Weller 2007).

In terms of seed versus soil origin, there were OTUs with a low relative abundance and

low occupancy that only occurred in cotton or soybean spermosphere samples and were absent

from the soil. However, we hesitate to conclude they originated from the seed without directly

identifying seed epiphytes and endophytes since it was impossible to know if the unique

microbes were seed associates colonizing the spermosphere or if rare members of the soil only

present spermosphere samples by chance. Furthermore, surface disinfecting seeds used in this

study likely reduced the number of seed epiphytes that would colonize the spermosphere. The

implications of surface disinfecting seeds have been argued elsewhere (Barret et al. 2015;

Schiltz et al. 2015; Shade et al. 2017). Another limitation of our approach that limited our ability

to identify seed-associated microbes may be the use of OTUs rather than amplicon sequence

variants (ASVs) or zero radius OTUs (zOTUs). A finer clustering method may be better suited to

studying the transmission of seed-associated microbes into the spermosphere since genotypes

originating from the seed may be grouped within 97% OTUs originating from the soil. We

recognize microbes originating from the seed can colonize seedlings and other plant organs,

which can alter plant health (Bintarti et al. 2022b; Bintarti et al. 2022a; Chesneau et al. 2022;

Chesneau et al. 2020; Johnston-Monje et al. 2016; Mitter et al. 2017; Rochefort et al. 2021;

Shade et al. 2017; Simonin et al. 2022). For example, it was recently demonstrated that crop

flowers sprayed with a beneficial bacterium can colonize endosperm and transmit to germinating

seeds (Mitter et al. 2017). While the importance of seed-associated microbes on plant health is

evident, little is known about seed endophytes and interactions with horizontally acquired soil

organisms, which tend to contribute a large portion of the microbial diversity to the seedling

microbiome (Buyer et al. 1999; Nelson 2018; Rochefort et al. 2021).

In terms of the microbial networks, we observed different hubs and different topologies

given the same set of taxa used for network construction. While fungal diversity was not altered

in this study, fungal OTUs were identified as hubs, potentially demonstrating meaningful

interactions with in a spermosphere soil. Of most interest was the yeast *Hannella* since these

fungi are commonly found in soils, the phyllosphere, and as part of the core seed and

phyllosphere microbiome (Noel et al. 2022; Simonin et al. 2022; Yurkov 2018). *Dioszegia*, in

the same family as *Hannaella*, was identified as a network hub in the phyllosphere (Agler et al.

2016), and the closely related yeast *Bullera* has been a network hub of the soybean phyllosphere

(Longley et al. 2020). These yeasts are generally non-pathogenic, but their ecological role is

poorly understood (Gouka et al. 2022). The prokaryote hubs were also intriguing because cotton

contained several Nitrososphaeraceae pOTUs, which likely are involved with ammonia-

oxidizdation in soils (Reyes et al. 2020). Cotton spermosphere hubs also had a *Massilia* pOTU.

Massilia is known for below-ground associations and the ability to solubilize phosphate (Silva et

al. 2017), but has also been found as a hub in above-ground plant tissues (Longley et al. 2020).

For soybean, *Sphingomonas* and *Methylobacterium*/*Methylorubrum* pOTUs were notable

network hubs since these genera have been demonstrated to be abundant in the phyllosphere and

core seed microbiome and produce plant growth-promoting hormones and UVA-absorbing

compounds (Kwak et al. 2014; Yoshida et al. 2017). The difference in hub taxa between crops

demonstrates that soybean and cotton construct unique microbial communities early in life,

which may have plant health consequences at or beyond the spermosphere stage.

However, spermosphere-enriched pOTUs were not identified as network hubs; instead,

they were located more peripheral in the networks, indicating they may be copiotrophs

responding quickly to the availability of carbon-rich exudates from the seeds (Torres-Cortes et

al. 2018). Spermosphere networks were more connected with larger components than the soil

network. Increased soil network complexity was associated with increased microbiome function

(Wagg et al. 2019). Therefore, it may be hypothesized that seed exudates help stimulate

associations between organisms or sub-communities and form more connected or stable

communities. However, further research is needed to determine how topological features of

networks are associated with plant health and why hub taxa connect to other taxa and help

assemble plant microbiomes.

The technique used in this study enabled quick and efficient collection of spermosphere

soil within phase I and II of seed germination and showed the enrichment of beneficial

copiotrophic taxa. However, these copiotrophic taxa were not central to microbial networks. This

technique could easily be applied to other sequencing methods like metagenomics or

metatranscriptomics for a better understanding of spermosphere soil microbiome functions.

Coupled with sequencing the seed microbiomes will be powerful to study interactions between

seed and seedling pathogens, chemical or biological seed treatments, and interactions with

pathogens in the spermosphere – thereby improving knowledge of spermosphere ecology, which

will lead to improved understanding of the plant microbiome.

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Table and Figures

Table 1. Permutational Analysis of Variance (PERMANOVA) and dispersion for prokaryote communities separated by time post sowing based on Bray-Curtis and weighted Uni-frac distances

Table 2. Comparison of centrality network parameters

Table 3. Hub taxa identified for each habitat

Supplemental Table 1. Permutational Analysis of Variance (PERMANOVA) for prkaryote and fungal communities

Supplemental Table 2. Differentially abundant bacterial taxa in soybean and cotton spermosphere soils

Supplemental Table 3. Topological properties of networks

Figure 1. Diagram depicting the sampling technique used for sampling spermosphere soils.
The spermosphere is defined as the 5-10 mm of soil directly surrounding a seed. (a) Photo demonstrating the sampling technique of spermosphere soil contained within an 11 mm cork borer. (b) Sampling with an 11 mm cork borer inside the confining space within wells of a 12-well plate allowed direct and controlled sampling of the spermosphere around single seeds.

Figure 2. Prokaryote spermosphere evenness follows water imbibition. (a) Water imbibition
over time for cotton and soybean seeds (n = 6). (b) Prokaryote evenness over time in control soil,
soybean spermosphere soil, or cotton spermosphere soil. Soybean had significantly lower
evenness (higher dominance) after 6, 12, and 18 hours compared to control soil. Cotton followed
the same trend but was significantly less even after 12 and 18 hours. Comparisons were made
with Wilcox ranked sign test ($\alpha = 0.05$; n = 6) (c) Spearman correlation between prokaryote
evenness and water imbibition.

Figure 3. Composition of the most abundant prokaryote OTUs changes over time. Relative
abundance of the top twenty most abundant prokaryote OTUs shifts over time within soybean
spermosphere soil, cotton spermosphere soil, or control soil.

Figure 4. Prokaryote spermosphere composition changes over time and crop. (a-d) Principal
coordinate analysis based on Weighted Unifrac distances. (e-h) Principal coordinate analysis
based on Bray-Curtis distances. Reported significance values above each plot are the result of a
permutational test of variance ($\alpha = 0.05$; n = 6). Accompanying dispersion tests are shown in
Table 1.

**Figure 5. Differential abundance of prokaryote OTUs shows enrichment of specific taxa
within the spermosphere.** (a) Points represent individual OTUS. Positive values on the x-axis
indicate the OTU was numerically more abundant in a soybean spermosphere compared to soil
without a soybean seed. Similarly, positive y-axis values indicate the OTU was numerically
more abundant in a cotton spermosphere compared to soil without a cotton seed. Colored points

are pOTUs detected as significantly different in soybean or cotton. Grey circle points are non-

significant. Point shape indicates significance in one crop or both. Points are colored by the

prokaryote order. Significance was determined within the ANCOM-BC2 algorithm with a Holm-

Bonferroni correction (α-0.05). (b) Composition of significantly enriched pOTUs colored by

genera.

Figure 6. Spermosphere soils have different topological properties and different sets of hub

taxa compared to control soil. Networks for (a) cotton spermosphere soil and (b) soybean

spermosphere soil and (c) control soil were constructed with the same set of taxa. Nodes with

different shape and colors indicate the prokaryote or fungal class. Less transparent nodes are

significant hubs based on eigenvector centrality above the 95% quantile. Nodes with red outlines

were significantly enriched in a spermosphere as detected in Fig.4.

Supplemental Figure 1. Preliminary experiment conducted to demonstrate the effectiveness of

the sampling procedure shown in Figure 1. Bacterial populations within spermosphere were

greater in soybean or cotton spermosphere soils compared to control soil.

Supplemental Figure 2. Sequencing outputs for prokaryotes (16S) and fungi (ITS). (a)

Composition and taxonomic output of the mock community samples. (b) rarefaction curves for

prokaryotes and fungi. Dashed lines indicate the median read depth. (c) contaminant filtering

based on OTU prevalence in negative control samples. (d) Histogram of read depth per sample.

773   **Supplemental Figure 3**. Within sample diversity measurements for soybean or cotton

774   spermosphere samples compared to control soil. (a) prokaryote richness significantly dropped

775   over time ($P < 0.001$) consistently across habitats (spermosphere or soil without seed). (b)

776   Similarly, Faith's phylogenetic distance also followed a similar pattern. No consistent

777   differences were observed over time or between habitats for fungal (c) richness or (d) evenness.

778
779   **Supplemental Figure 4**. Fungal composition of the top 20 most abundant Fungal OTUs in

780   cotton spermosphere soil, soybean spermosphere soil, or control soil.

Cotton spermosphere

N components = 9
N components excluding isolated nodes = 6
N nodes in largest component = 136

Soybean spermosphere

N components = 20
N components excluding isolated nodes = 8
N nodes in largest component = 121

Soil without a seed

N components = 30
N components excluding isolated nodes = 21
N nodes in largest component = 80

- Gammaproteobacteria
- Alphaproteobacteria
- Nitrospiria
- Myxococcia
- bacteriap25
- Verrucomicrobiae
- Methylomirabilia
- Entotheonellia
- Vicinamibacteria

- Acidobacteriae
- Chloroflexia
- TK10
- Actinobacteria
- Thermoleophilia
- Bacilli
- Gemmatimonadetes
- Nitrososphaeria
- Blastocatellia

- Dothideomycetes
- Sordariomycetes
- Leotiomycetes
- unidentified Fungi
- Mortierellomycetes
- Tremellomycetes
- Eurotiomycetes

— Negative co-association
— Positive co-association

a Spermosphere = 5-10 mm soil around seed

b Cork borer = 11 mm

15 mm

23 mm

7-10 mm

3-6 mm

5-8 mm

Soil

Table 3. Hub taxa identified for each habitat

Crop	OTU	Taxonomy
Cotton		
spermosphere soil	pOTU46	Gaiellales
	pOTU119	Gaiellales
	pOTU29	*Massilia*
	pOTU1009	Nitrososphaeraceae
	pOTU19	Nitrososphaeraceae
	pOTU12	Nitrososphaeraceae
	fOTU36	*Hannaella oryzae*
	fOTU56	*Fusarium chlamydosporum*
Soybean		
spermsphere soil	pOTU1559	Chloroflexi TK10
	pOTU132	Nocardioides
	pOTU36	Angustibacter
	pOTU11	*Sphingomonas*
	pOTU349	*Methylobacterium* /*Methylorubrum*
	fOTU64	*Helicoma*
	fOTU36	Fungi
	fOTU10	*Bartalinia pondoensis*
Soil without a seed	fOTU1	*Stagonosporopsis oculi-hominis*
	fOTU13	Teichosporaceae
	fOTU15	Cucurbitariaceae
	fOTU17	*Alternaria tenuissima*
	fOTU22	Neopestalotiopsis
	fOTU32	Pleosporales
	fOTU36	*Hannaella oryzae*
	fOTU4	*Cladosporium cladosporioides*

Table 2. Comparison of centrality network parameters

Centrality network parameter	Cotton vs. Soybean		Cotton vs. Control soil		Soybean vs. Control soil	
	Jaccard index[a]	P	Jaccard index[a]	P	Jaccard index[a]	P
degree	0.034	< 0.001	0.128	0.001	0.044	< 0.001
betweenness centrality	0.154	< 0.001	0.156	0.001	0.154	< 0.001
closeness centrality	0.152	< 0.001	0.246	0.092	0.206	0.019
eigenvector centrality	0.169	0.003	0.188	0.007	0.206	0.019
hub taxa	0.000	0.002	0.067	0.019	0.000	0.001

[a]Jaccard index of 1 indicates perfect agreement between soybean or cotton spermosphere networks

Table 1. Permutational Analysis of Variance (PERMANOVA) and
dispersion for prokaryote communities separated by time post sowing
based on Bray-Curtis and weighted Uni-frac distances

Dissimilarity Metric	Hours post sowing	df	PERMANOVA[a]			Dispersion	
			R^2	F	P	F	P
Bray-Curtis	0	2	0.113	0.959	0.437	0.937	0.407
	6	2	0.273	2.815	<0.001	0.539	0.606
	12	2	0.424	5.517	<0.001	0.828	0.533
	18	2	0.417	4.648	<0.001	0.092	0.911
Weighted Unifrac	0	2	0.142	1.238	0.192	1.898	0.190
	6	2	0.491	7.245	<0.001	2.538	0.066
	12	2	0.609	11.675	<0.001	1.186	0.376
	18	2	0.696	14.863	<0.001	1.401	0.304

[a]Permanova and dispersion tests were performed on 999 permutations