

Bacterial and fungal co-occurrence patterns in agricultural soils amended with compost and bokashi

Luo, Y.; Gonzalez Lopez, J.B.; Veelen, H.P.J. van; Sechi, V.; Heijne, A. ter; Bezemer, T.M.; Buisman, C.J.N.

Citation

Luo, Y., Gonzalez Lopez, J. B., Veelen, H. P. J. van, Sechi, V., Heijne, A. ter, Bezemer, T. M., & Buisman, C. J. N. (2022). Bacterial and fungal co-occurrence patterns in agricultural soils amended with compost and bokashi. *Soil Biology And Biochemistry*, *174*. doi:10.1016/j.soilbio.2022.108831

Version:Publisher's VersionLicense:Creative Commons CC BY 4.0 licenseDownloaded from:https://hdl.handle.net/1887/3563299

Note: To cite this publication please use the final published version (if applicable).



Contents lists available at ScienceDirect

Soil Biology and Biochemistry



journal homepage: www.elsevier.com/locate/soilbio

Bacterial and fungal co-occurrence patterns in agricultural soils amended with compost and bokashi



Yujia Luo^{a,b,*}, Juan Bautista Gonzalez Lopez^b, H. Pieter J. van Veelen^b, Valentina Sechi^b, Annemiek ter Heijne^a, T. Martijn Bezemer^{c,d}, Cees J.N. Buisman^{a,b}

^a Environmental Technology, Department of Agrotechnology and Food Sciences, Wageningen University, P.O. Box 17, 6700, AA, Wageningen, the Netherlands

^b Wetsus, European Centre of Excellence for Sustainable Water Technology, Oostergoweg 9, 8911, MA, Leeuwarden, the Netherlands

^c Above-Belowground Interactions Group, Institute of Biology, Leiden University, P.O. Box 9505, 2300, RA, Leiden, the Netherlands

^d Netherlands Institute of Ecology (NIOO-KNAW), Department of Terrestrial Ecology, Droevendaalsesteeg 10, 6708, PB, Wageningen, the Netherlands

ARTICLE INFO

Keywords: Organic amendments Co-occurrence network Inter-kingdom correlations Field experiment

ABSTRACT

The living soil harbors a significant number and diversity of bacteria and fungi, which are essential in sustaining soil ecosystem functions. Most studies focus on soil bacteria or fungi, ignoring potential interrelationships between kingdoms that coevolve and synergistically provide ecosystem functions. In a seven-year agricultural field, we explored the effects of organic amendments (OAs; i.e., compost and bokashi) on intra- and inter-kingdom co-occurrence networks of soil bacterial and fungal communities. We observed that OAs changed the co-occurrence patterns of bacteria and fungi. Distinct network modules were observed in the unamended and amended soils, and the physicochemical properties of the soil could partially explain the formation of these modules. We also found that compost and bokashi increased the proportion of positive correlations, and this could reduce competition among microorganisms for soil resources. Our study reveals that soil management with OAs affects relationships between bacterial and fungal subpopulations that physically co-exist, cooperate, and compete in non-random structured networks. It highlights that ecosystem functions may depend on inter-kingdom interactions shaped by different amendments and their applied dose.

1. Introduction

Soil ecosystem functioning is a fundamental component of productive agriculture, largely driven by a myriad of positive and negative, direct and indirect interactions among soil organisms (Wagg et al., 2019). These soil communities generally consist of diverse bacterial and fungal subpopulations that coevolve and synergistically provide ecosystem functions (Fuhrman, 2009; Bezemer et al., 2010; Haq et al., 2014). Intra- and inter-kingdom interactions cannot be understood only based on richness, diversity, and community composition (Wagg et al., 2019), but aggregated organisms that physically co-exist, cooperate and compete in non-random structures that can be conceptually projected and investigated as co-occurrence networks (Barberán et al., 2012; Lu et al., 2013; Ma et al., 2016). While biological activity in agricultural soils facilitates the availability of nutrients for crops from organic amendments (OAs), it remains unclear if common OAs alter the structure and function of soil microbial networks.

The soil microbial biomass is viewed as the "eye of the needle" in the

soil, through which organic matter must pass (Jenkinson, 1977; Sparling, 1985). In this sense, soil microbes are sensitive to the quality and distribution of soil organic pools. Studies have shown that abiotic variables induced by OAs control soil bacterial or fungal community composition (Pérez-Piqueres et al., 2006; Lu et al., 2015; Dumontet et al., 2017), which subsequently shapes the relationships between microbial taxa (Prosser et al., 2007). While several studies have shown that the addition of OAs can affect co-occurrence patterns of soil bacterial or fungal communities, these studies did not interrogate interrelationships between kingdoms (Ling et al., 2016; Schmid et al., 2018; Xue et al., 2018; Yang et al., 2009, 2020; Zhou et al., 2019).

Nonetheless, these studies revealed that the increased complexity of the networks contributed to the stability and resilience of the native soil microbiota (Yang et al., 2019). In addition, modules (a group of nodes more densely connected to each other than to nodes outside the group) in bacterial networks correlated differentially with specific soil variables after applying different OAs (Ling et al., 2016). Here, a node refers to a bacterial or fungal OTU. The connections of nodes in the network are

* Corresponding author. Environmental Technology, Wageningen University, P.O. Box 17, 6700, AA, Wageningen, the Netherlands. *E-mail addresses:* yujia.luo@wetsus.nl, yujia.luo@outlook.com (Y. Luo).

https://doi.org/10.1016/j.soilbio.2022.108831

Received 27 August 2021; Received in revised form 5 September 2022; Accepted 9 September 2022 Available online 15 September 2022 0038-0717/© 2022 The Authors. Published by Elsevier Ltd. This is an open access article under the CC BY license (http://creativecommons.org/licenses/by/4.0/). called edges, presenting positive or negative correlations between microbes. Exploring whether and how co-occurrences among bacteria and fungi in the soil are shaped by commonly applied OAs should provide a more integrated understanding of the soil microbiome and may provide insights into ecosystem functioning for optimizing agricultural soil management.

In this study, we report a seven-year field experiment in which we investigated how two OAs, compost, and bokashi, applied at high and low doses, affected long-term soil bacterial and fungal intra- and interkingdom relationships using co-occurrence network analysis. Compost and bokashi are obtained from aerobic composting and fermentation, respectively. Moreover, bokashi is prepared with the addition of commercial microbial inoculum, which contains five main groups of microorganisms, including lactic acid bacteria, photosynthetic bacteria, veasts, actinobacteria, and fermenting fungi (Shin et al., 2017). After a seven-year annual application of compost and bokashi, we expect that soil microorganisms co-occur in a non-random way rather than by chance in the compost and bokashi amended soils, with a higher level of non-random networks than the unamended soil. In addition, compost and bokashi are very different regarding the stability of organic substrates and nutrient status. The composting process leads to the removal of labile organic fractions and the concentration of more recalcitrant fractions, stabilizing organic matter with low nutrient content and slow-release fertility (Neher et al., 2015). On the other hand, the C content in bokashi can be more readily metabolized than that in compost due to less efficient mineralization during fermentation than in composting. Therefore, we expect that adding compost and bokashi will shape the physicochemical properties of soil, and this will result in distinct microbial modules. We also expect that adding compost and bokashi will affect the intra- and inter-kingdom associations since compost and bokashi provide organic substrates and nutrients. Antagonistic and mutualistic relationships have been reported between specific bacteria and fungi but depend on the local conditions (Nazir et al., 2012; Fritsche et al., 2014). For instance, fungi can serve as potential providers of nutrients and energy for bacteria in carbon- and nutrient-limited soils (Haq et al., 2014). Therefore, we expect that in soils amended with compost and bokashi, the inter-kingdom associations will be reduced compared with the unamended soil and that a higher OA application dose will lead to a lower percentage of inter-kingdom relations.

2. Materials and methods

2.1. Field experiment description

The field experiment was initiated in autumn 2013 by Agriton (Noordwolde, The Netherlands) at the experimental farm Ebelsheerd of Stichting Proefboerderijen Noordelijke Akkerbouw (SPNA, Nieuw Beerta, The Netherlands). The farm field soil consists of 46% clay, 37% silt, and 12% sand with a pH of 8.2. The soil contains 8.8% organic matter. The area is characterized as strong swelling and shrinking soils with high soil compaction (Bronswijk and Evers-Vermeer, 1990; Roelf-sema, 2010). Compost and bokashi were applied yearly before winter wheat (*Triticum aestivum*), with low and high application doses (Table 1). Compost and bokashi were prepared from the same amount of grass that underwent composting and anaerobic fermentation, respectively. The commercial microbial inoculum (EM-1[®], Agriton, The Netherlands) was used to start the fermentation process when preparing Bokashi. All soils received two levels of artificial N fertilization. Table 1 shows the arrangement of the field experiment.

Soil cores (0–15 cm deep, 3 cm diameter) were collected from each plot in October 2020. Soil samples were transported to the laboratory on ice, and each soil sample was divided into three subsamples afterward. The first group of subsamples was stored at 4 °C to measure water content (WC) and organic matter content (OM). The second group of subsamples was dried at 65 °C for three days until a constant weight was

Table 1

Description of the field experiment. Ten sub-treatments were clustered into three treatments: unamended soil (control soil), compost amended soil, and bokashi amended soil. Compost and bokashi had two application doses, and two levels of chemical fertilization were applied as well. Each sub-treatment had three replicates, resulting in 30 plots in total. The plots were randomly distributed. NA: not applicable. F- and F+ represent a low and a high application dose of N fertilizer, respectively.

	Tr	eatment	Organic amendment type	Organic amendment dose (ton/ha)	N Fertilizer dose (kg/ ha)
1	Control.		NA	NA	50
	F-	Unamended			
2	Control.	soil (control)	NA	NA	100
	F+				
3	CL.F-		Compost	3.8	50
4	CL.F+	Compost	Compost	3.8	100
5	CH.F-	amended soil	Compost	9.5	50
6	CH.F+		Compost	9.5	100
7	BL.F-		Bokashi	9.7	50
8	BL.F+	Bokashi	Bokashi	9.7	100
9	BH.F-	amended soil	Bokashi	24.25	50
10	BH.F+		Bokashi	24.25	100

reached for measuring pH, electrical conductivity (EC), water extractable nutrients, total nutrients, and dissolved organic carbon (DOC). The third group of subsamples was stored at -20 °C for DNA extraction.

2.2. Physicochemical analysis of soil samples

WC was measured after drying soil samples in a forced-air oven at 105 °C for 8 h. Afterward, samples were burned at 550 °C for 2 h to quantify OM. Soil pH and EC were measured using a Mettler Toledo SevenExcellence™ in a 1:10 soil/MilliQ water suspension (w/v) after 2 h shaking at 25 °C. Water extractable nutrients, including NO₃, NO₂, PO₄³⁻, NH₄⁺, and K⁺, were measured in a 1:10 soil/MilliQ water suspension (w/v). The suspension was centrifuged at 3750 g after 2 h shaking at 25 °C. The supernatant was filtered through a 0.45 μ m membrane filter and then analyzed by ion chromatography (Metrohm Compact IC 761). Before analyzing total nutrients, soil samples were ground into fine particles and mixed homogeneously. Total carbon (TC) and total nitrogen (TN) were determined with an elemental analyzer (Interscience FlashSmart CHNSO). Total phosphorus (TP) and total potassium (TK) were analyzed by inductive coupled plasma optical emission spectrometry (Perkin Elmer Optima 5300 DV) after microwave digestion with acids (Milestone Ethos Easy SK-15). DOC was extracted and prepared the same way as water extractable nutrients, then analyzed by a TOC analyzer (Shimadzu TOC-L). The effects of the OAs and chemical fertilization on soil physicochemical properties were tested by ANOVA (alpha = 0.05), followed by pair-wise comparison (TukeyHSD, family-wise error rate 5%). The assumptions of normality and homogeneity of variances were assessed for each ANOVA model.

2.3. Microbial community analysis of soil

DNA was extracted from 0.25 g of soil using the DNeasy Power Soil Kit (Qiagen) following the manufacturer's protocol. DNA concentration was quantified by Quantus (Promega), and purity was checked by a Nanodrop spectrophotometer (Thermo Scientific NanoDrop 1000 spectrophotometer) with OD_{260} and OD_{280} . Extracted DNA samples were stored at -20 °C. DNA samples were normalized to 20 ng/µL for library preparation and were sequenced (MrDNA, TX, USA) on a MiSeq (Illumina). Libraries for bacteria were constructed using primers 338F (ACTCCTACGGGAGGCAGCAG) (Fierer et al., 2005) and 806R (GGAC-TACHVGGGTWTCTAAT) (Caporaso et al., 2011). Libraries for fungi were constructed using primers ITS1F (CTTGGTCATTTA-GAGGAAGTAA) and ITS2R (GCTGCGTTCTTCATCGATGC) (Adams

et al., 2013). The raw sequence data can be accessed at the European Nucleotide Archive (ENA) at EMBL-EBI under accession number PRJEB46153 (https://www.ebi.ac.uk/ena/browser/view/PRJEB46153). Raw sequence data were analyzed by QIIME2 (version 2019.10) following the methodology in our previous publication (Luo et al., 2022). The downstream analyses of bacterial and fungal communities were performed in RStudio (R version 4.2.1) using the phyloseq package (McMurdie and Holmes, 2013) and the vegan package (Oksanen et al., 2020).

2.4. Co-occurrence network analysis

Bacterial sequence data were first quality trimmed and clustered (de novo clustering) into OTUs with a 90% identity threshold using VSEARCH (Rognes et al., 2016) in QIIME2. The 90% identity threshold reduces the size of the OTU table significantly compared with the more standard "species-level" cutoff level of 97% identity, reducing the network complexity and greatly facilitating visualization and analysis of the data (Konstantinidis and Tiedje, 2007; Barberán et al., 2012). Fungal sequence data were quality trimmed and clustered (closed-reference clustering) into OTUs using a 97% identify threshold. After filtering (> 0.01% abundance, > 4 samples prevalence) and clustering, we detected 275 bacterial OTUs (1,636,362 reads) and 270 fungal OTUs (1,591,059 reads) across all soil samples. Bacterial and fungal sequence data were first rarefied at the same depth (29,073) and then concatenated into a single OTU table. Core OTUs for network analysis were defined by retaining OTUs detected in at least five out of six replicate samples of a treatment. Core OTUs comprised > 88.2% of the total relative abundance (Table 2). A total of 181 OTUs were shared across all treatments. and the following OTU counts were retained per treatment: control (n =244 OTUs), CL (n = 255), CH (n = 260), BL (n = 260), BH (n = 267). Proportions of OTUs shared by and unique to the five treatments are presented as supplementary information (Fig. S1).

To determine if OA treatments differentially affected bacterial and fungal co-occurrences, we used robust correlation analysis that accounts for compositionality and sparsity using SparCC (Sparse Correlations for Compositional data) (Friedman and Alm, 2012), implemented with the SpiecEasi package (Kurtz et al., 2015). The correlation between two OTUs was considered robust if the correlation coefficient > |0.6| and P < 0.01 (Barberán et al., 2012). Non-random co-occurrence patterns were evaluated based on checkerboard scores (C-score) using a null model approach that assumes no interactions between microbial taxa (Stone and Roberts, 1990). C-scores were calculated using the vegan package (Oksanen et al., 2020). A higher C-score denotes a higher level of non-randomness or network structure. To assess network topologies, we calculated the average degree, average clustering coefficient,

Table 2

Key properties of networks. Control: unamended soil; CL: soil amended with a low dose of compost (3.8 ton/ha); CH: soil amended with a high dose of compost (9.5 ton/ha); BL: soil amended with a low dose of bokashi (9.7 ton/ha); BH: soil amended with a high dose of bokashi (24.25 ton/ha).

	Control	CL	CH	BL	BH		
No. of original nodes	244	255	260	260	267		
total relative abundance	93.5%	96.8%	95.4%	95.9%	88.2%		
After selection of robust correlation							
No. of nodes	216	242	237	251	249		
No. of edges	413	691	693	1104	741		
positive edge/negative edge	0.39	0.63	1.24	0.54	0.67		
edge density	0.02	0.02	0.02	0.04	0.02		
C-score	13.8	29.9	30.0	66.4	31.7		
average degree	3.8	5.7	5.8	8.8	6.0		
average clustering coefficient	0.3	0.4	0.5	0.4	0.4		
average path distance	5.5	5.0	5.9	3.8	4.9		
modularity	0.77	0.70	0.72	0.53	0.65		
No. of modules	16	11	11	7	11		
major modules (degree $>$ 5)	9	9	7	7	9		

average path distance, and modularity. All calculations were performed using the igraph packages (Csardi and Nepusz, 2006). Networks were explored and visualized using Gephi (version 0.9.2) (Bastian et al., 2009). Mantel tests were performed to estimate correlations between the major network modules (considering OTUs in each module and their relative abundance) and the physicochemical properties of the soil using the ecodist package (Goslee and Urban, 2007).

3. Results

3.1. Soil physicochemical properties seven years after amendment application

In general, the control soil and the CL soil had more similar physicochemical properties than the other treatments. Specifically, the control soil and the CL soil had lower concentrations of DOC and TC and higher concentrations of TP and TK than the CH, BL, and BH soils (Fig. 1). TN and other water extractable nutrients (i.e., AN, phosphate, and potassium) were generally higher in the amended soils than the control soils. OAs had no significant effects on soil pH, EC, WC, and OM (supplementary information (Table S1; Table S2)). The fertilizer dose had no significant impact on the physicochemical properties, except phosphate (Table S1; Table S2). Therefore, we assessed the impact of OAs and doses on the soil physicochemical variables without distinguishing between the two chemical fertilizer doses (Fig. 1).

3.2. Effects of OAs on co-occurrence networks of soil microorganisms

Chemical fertilization did not influence the richness and composition of the soil microbial community (Fig. S2; Table S3; Table S4). Therefore, we assessed the effects of OAs and doses on soil microbial networks and co-occurrence patterns without distinguishing between the two chemical fertilizer doses.

Robust correlation analysis revealed that OA amended soils retained more nodes (OTUs) and edges than control soil (Fig. 2 and Table 2). Higher C-scores were observed in the amended soils, indicating that the addition of either compost or bokashi increased the network structure of the soil microbial community. The control soil community network had a C-score of 13.8 (p < 0.001), which also indicated a non-random structure. The addition of a low dose of bokashi (BL) resulted in the highest C-score when all treatments were compared. The microbial taxa (nodes) in the amended soils were densely connected into seemingly more tightly knit groups than the nodes in the control soil, as indicated by higher average degrees and higher average clustering coefficients.

In all treatments, soil microbial networks were structured by modules (modularity index > 0.5; Table 2; Fig. 3). The modularity was highest in control soil, which also had more modules than any of the treatments (Table 2). The number of major modules (degree > 5) across all treatments was similar; however, the size (number of nodes) of major modules in the amended soils was generally larger than in the control. This was also reflected by the average degree and average clustering coefficient.

In addition, the formation of major modules in treatment-specific networks was correlated differentially with the soil physicochemical parameters (Fig. 4A). EC, DOC, TC, TN, AN, and phosphate were the main soil physicochemical measures that significantly correlated to treatment-specific modules, but few significant correlations (P < 0.05) were observed. The addition of compost and bokashi increased the number of bacterial-fungal correlations compared with the unamended soil, particularly in the BL treatment (Fig. 4B). This suggests that more bacterial and fungal taxa in the amended soils co-occurred and potentially interacted more frequently than in control soil. The addition of compost and bokashi increased the proportion of positive correlations while decreased the proportion of negative correlations among taxa (Table 2 and Fig. 4B). Notably, the addition of compost and bokashi decreased the proportions of negative correlations within kingdoms (i.



Fig. 1. Effects of compost and bokashi on the physicochemical properties of soil seven years after application. Box plots are shown representing the median and quartiles of the data. Experimental replication n = 6 (without distinguishing the two levels of chemical fertilization). DOC: dissolved organic carbon; TC: total carbon; TN: total nitrogen; TP: total phosphorus; TK: total potassium; AN: water extractable nitrogen, the sum of NO₃⁻, NO₂⁻, and NH₄⁺. Control: unamended soil; CL: soil amended with a low dose of compost (3.8 ton/ha); CH: soil amended with a high dose of compost (9.5 ton/ha); BL: soil amended with a low dose of bokashi (9.7 ton/ha); BH: soil amended with a high dose of bokashi (24.25 ton/ha).

e., bacteria-bacteria, fungi-fungi) compared to the unamended soil (Fig. 4B). However, increased negative correlations across kingdoms (i. e., bacteria-fungi) were observed in the amended soils, especially in the BL treatment. Common taxa of bacteria and fungi were observed in major modules (Fig. 4C). Taxa within Ascomycota fungi were highly prevalent in the networks across all treatments. Other fungal taxa (including Basidiomycota, Mortierellomycota, Glomeromycota, and Chytridiomycota) were also observed in major modules. Diverse bacterial phyla were also observed in major modules: Chloroflexi, Proteobacteria, Actinobacteria, Firmicutes, Bacteroidetes, Verrucomicrobia, Acidobacteria, and Planctomycetes were dominant across all treatments.

4. Discussion

4.1. Non-random co-occurrence patterns induced by OAs

Non-random community assembly may be a general characteristic across all life domains (Horner-Devine et al., 2007; Barberán et al., 2012). Our study, which shows significant non-random co-occurrence patterns in the soil with and without the application of OAs, supports this general notion. Our observation that the OA treatment increases the complexity of networks in native soil is in line with previous studies (Ling et al., 2016; Yang et al., 2019; Zhou et al., 2019). Our data support that the microbial community in the amended soils was more organized and potentially had a larger operational community that was functionally interrelated than that in control because of a higher C-score, more edges, and more densely correlated nodes in the modules.

Interestingly, in our study, a decrease in the proportion of negative correlations was found in the amended soils when compared to the control soil, suggesting that amelioration of soil with compost and bokashi may decrease competition among microorganisms for resources (Blagodatskaya and Kuzyakov, 2008; Mau et al., 2015). Other studies also reported that adding OAs or nutrients reduced the number and strength of negative correlations between microorganisms (Banerjee et al., 2016; Yang et al., 2019, 2020). The addition of OAs introduced

extra C and nutrients to the soil. This may reduce competition and increase the number of multiple trophic levels or resource cascades in the food web, which could improve energy use efficiency (Coyte et al., 2015; Banerjee et al., 2016). However, theoretical studies predict that compartmentalization and the presence of negative interactions will increase the stability of networks under disturbances, as negative links stabilize co-oscillation in communities (Coyte et al., 2015; Kuiper et al., 2015). In addition, a recent experimental study demonstrated that negative interactions are more common than reciprocal positive interactions in soil bacterial communities. Negative interactions can be desirable by aiding in resistance against invasive species and maintaining the stability of the community (Palmer and Foster, 2022). Even though the proportion of negative correlations was reduced in the OA amended soils, the absolute number of negative correlations was higher than in the control soils. Therefore, the addition of compost or bokashi may contribute to the stability of the microbial community by enhancing resistance or resilience to disturbance.

4.2. Distinct modules observed in the amended soils

In all treatments, we observed high modularity indices, indicating that the microbial community was structured into modules consisting of groups of bacteria and fungi that formed strong associations. A modularity index > 0.4 suggests that the network can be well-divided into modules or clusters with a strong modular structure (Newman, 2006). These findings support our hypothesis that the addition of different doses of compost and bokashi will result in distinctive microbial modules in the co-occurrence networks. In addition, the size (number of nodes) of major modules was larger in the amended soils than in the control, particularly in the BL soil. This also suggests that the addition of OAs may have changed the function of the soil microbial community with larger operational communities that could be functionally correlated.

Interestingly, only a few major modules significantly correlated with soil variables (i.e., EC, DOC, TC, TN, AN, and phosphate). Our results



Fig. 2. Network visualization of co-occurring bacteria and fungi per treatment. Control: unamended soil; CL: soil amended with a low dose of compost (3.8 ton/ha); CH: soil amended with a high dose of compost (9.5 ton/ha); BL: soil amended with a low dose of bokashi (9.7 ton/ha); BH: soil amended with a high dose of bokashi (24.25 ton/ha).

suggest that the physicochemical properties of the soil might not be the primary factors driving the formation of the network modules. We speculate that the heavy clay soil in this study field, characterized by strong compaction (Roelfsema, 2010), may weaken the influence of the physicochemical properties of the soil on the microbial network. Alternatively, there may be unknown factors (e.g., specific and multiple interactions with protists, Archaea, and other soil organisms) that are more important than the physicochemical properties of the soil in determining the soil microbial structure (Fuhrman, 2009; Deveau et al., 2018). Further studies are suggested to examine the influence of OAs on microbial networks in different soil types (e.g., sand and clay) and how the underground food web shapes microbial networks. Nevertheless, the microbial network might be susceptible to changes in EC, DOC, TC, TN, AN, and phosphate, which provide necessary nutrients for the growth, activity, and metabolism of soil microorganisms.

4.3. Organic amendments changed intra- and inter-kingdom correlations

Bacteria and fungi operate in the same environment, where complex interactions occur between these kingdoms, including predation, parasitism, competition, commensalism, and mutualism. Bacteria associated with soil fungi, including saprotrophic and mycorrhizal ones, can benefit from nutritional interactions (Haq et al., 2014). Nutritional interactions imply that bacteria feed on fungi or on their released products (De Boer et al., 2005; Leveau and Preston, 2008) or vice versa (Hildebrandt et al., 2002, 2006). Such nutritional interactions often imply commensalism or true mutualism (Nazir et al., 2012). We expected that adding OAs would weaken nutritional interactions between bacteria and fungi due to the provisioning of additional nutrients; thus, fewer bacterial-fungal correlations would be detected in the amended soils than in the unamended soil. However, this was not in line with our observation. We observed that the addition of compost and bokashi increased bacterial-fungal



Fig. 3. Major modules (degree > 5) highlighted in networks per treatment. Control: unamended soil; CL: soil amended with a low dose of compost (3.8 ton/ha); CH: soil amended with a high dose of compost (9.5 ton/ha); BL: soil amended with a low dose of bokashi (9.7 ton/ha); BH: soil amended with a high dose of bokashi (24.25 ton/ha).

correlations, mainly increasing their negative correlations compared to control, while positive correlations of bacteria-bacteria and fungi-fungi were increased after applying OAs. We hypothesize that the provisioning of nutrients (TC and TN) and the availability of nutrients (DOC, AN, and phosphate) by OAs may alleviate resource competition within kingdoms but stimulate competition across kingdoms. It is unclear what drives this. Even though efforts have been taken to uncover mechanisms driving bacterial-fungal interactions; challenges remain since the outcome of these interactions are the combined results of the physical associations (biofilm, free cell), molecular communication between (micro)organisms, and the local soil environment (Deveau et al., 2018). Further research should identify essential parameters in the soil system that can drive bacterial-fungal interactions.

We emphasize that the intra and inter-kingdom co-occurrence patterns are correlations between bacterial and fungal taxa, which indicates potential positive, negative, or neutral interactions but cannot uncover causal mechanisms structuring the communities (Barberán et al., 2012; Banerjee et al., 2016). Moreover, correlation analyses inherently imply technical challenges when analyzing microbial associations in soil, such as how to pre-process compositional abundance data (Faust, 2021). In addition, the performance of the statistical methods (including simple, direct correlation measures like Pearson and Spearman correlations) for inferring microbial association does not account for data properties that may yield spurious associations (Weiss et al., 2016; Hirano and Takemoto, 2019; Faust, 2021). Spearman correlation is still commonly applied to infer microbial networks from non-independent relative abundance data, but negative correlations are often overrepresented regardless of the true correlations underlying absolute abundances (Friedman and Alm, 2012). Methodological choices for network analyses need careful consideration of the data characteristics. Particularly, correlation techniques should be considered that account for sparsity and compositionality when necessary, such as SparCC (Friedman and



Fig. 4. (A) Correlation between major modules and the physicochemical properties of the soil based on the Mantel test. Significant correlations (P < 0.05) were marked with "X"; (B) Frequency of intra- and inter-kingdom correlations in the networks. The proposition of negative and positive correlation within each type of correlation is indicated by different color intensities: the dark shade of the color represents negative correlations, and the light shade represents positive correlations; (C) Frequency of nodes (bacteria and fungi) at phylum level in major modules detected in the networks. Control: unamended soil; CL: soil amended with a low dose of compost (3.8 ton/ha); CH: soil amended with a high dose of compost (9.5 ton/ha); BL: soil amended with a low dose of bokashi (24.25 ton/ha). (For interpretation of the references to color in this figure legend, the reader is referred to the Web version of this article.)

Alm, 2012). Nonetheless, our results highlight the importance of considering prokaryotic and eukaryotic components of the soil microbial community, which is often neglected when microbial communities are investigated as compartmentalized groups (De Menezes et al., 2015).

5. Conclusion

Compost and bokashi changed the intra- and inter-kingdom cooccurrence patterns of microorganisms in soils. Soil microbial communities appeared to have a higher level of non-randomness in the networks after applying compost and bokashi. This was particularly so when bokashi was applied at a low dose. OAs drove distinct network modules in terms of module compositions and the size of major modules. Our results suggest that network modules correlated with soil recourses (i.e., EC, DOC, TC, TN, AN, and phosphate). We also observed that OAs decreased the proportion of negative correlations, reducing potential competition among microorganisms for resources and increasing energy use efficiency. The addition of OAs increased bacterial-fungal correlations, particularly their negative correlations. Further studies should be conducted to identify key factors driving the changes in bacterial-fungal interactions after OAs application. In addition, it is essential to have experimental evidence validating the hypotheses generated by microbial networks (e.g., how abiotic properties of soil shape biotic correlations) since network analyses do not uncover causal relationships. The links between networks and soil ecological functions (e.g., plant growth,

C sequestration, and soil aggregation) also merit further research.

Declaration of competing interest

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

Data availability

Data will be made available on request.

Acknowledgments

This work was performed in the cooperation framework of Wetsus, European Centre of Excellence for Sustainable Water Technology (www. wetsus.eu). Wetsus is co-funded by the Dutch Ministry of Economic Affairs and Climate Policy, the European Union Regional Development Fund, the City of Leeuwarden, the Province of Fryslân, the Northern Netherlands Provinces, and the Netherlands Organisation for Scientific Research. The authors would also like to thank the members of the research soil theme (Agriton, Mulder Agro, Netherlands Institute of Ecology (NIOO-KNAW), Koninklijke Oosterhof Holman, Waterketen Onderzoek Noord (WON), and Waterschap Zuiderzeeland) for the fruitful discussions and financial support. The authors are grateful to the editor and reviewers for their time and valuable comments.

Appendix A. Supplementary data

Supplementary data to this article can be found online at https://doi.org/10.1016/j.soilbio.2022.108831.

References

- Adams, R.I., Miletto, M., Taylor, J.W., Bruns, T.D., 2013. Dispersal in microbes: fungi in indoor air are dominated by outdoor air and show dispersal limitation at short distances. The ISME Journal 7, 1262–1273.
- Banerjee, S., Kirkby, C.A., Schmutter, D., Bissett, A., Kirkegaard, J.A., Richardson, A.E., 2016. Network analysis reveals functional redundancy and keystone taxa amongst bacterial and fungal communities during organic matter decomposition in an arable soil. Soil Biology and Biochemistry 97, 188–198.
- Barberán, A., Bates, S.T., Casamayor, E.O., Fierer, N., 2012. Using network analysis to explore co-occurrence patterns in soil microbial communities. The ISME Journal 6, 343–351.
- Bastian, M., Heymann, S., Jacomy, M., 2009. Gephi: an open source software for exploring and manipulating networks. International AAAI Conference on Weblogs and Social Media 8, 361–362.
- Bezemer, T.M., Fountain, M.T., Barea, J.M., Christensen, S., Dekker, S.C., Duyts, H., Van Hal, R., Harvey, J.A., Hedlund, K., Maraun, M., Mikola, J., Mladenov, A.G., Robin, C., De Ruiter, P.C., Scheu, S., Setälä, H., Šmilauer, P., Van Der Putten, W.H., 2010. Divergent composition but similar function of soil food webs of individual plants: plant species and community effects. Ecology 91, 3027–3036.
- Blagodatskaya, E., Kuzyakov, Y., 2008. Mechanisms of real and apparent priming effects and their dependence on soil microbial biomass and community structure: critical review. Biology and Fertility of Soils 45, 115–131.
- Bronswijk, J.J.B., Evers-Vermeer, J.J., 1990. Shrinkage of Dutch clay soil aggregates. Netherlands Journal of Agricultural Science 38, 175–194.
- Caporaso, J.G., Lauber, C.L., Walters, W.A., Berg-Lyons, D., Lozupone, C.A., Turnbaugh, P.J., Fierer, N., Knight, R., 2011. Global patterns of 16S rRNA diversity at a depth of millions of sequences per sample. Proceedings of the National Academy of Sciences of the United States of America 108, 4516–4522.
- Coyte, K.Z., Schluter, J., Foster, K.R., 2015. The ecology of the microbiome: networks, competition, and stability. Science 350, 663–666.
- Csardi, G., Nepusz, T., 2006. The igraph software package for complex network research. InterJournal, Complex Systems 1695, 1–9.
- De Boer, W., Folman, L.B., Summerbell, R.C., Boddy, L., 2005. Living in a fungal world: impact of fungi on soil bacterial niche development. FEMS Microbiology Reviews 29, 795–811.
- De Menezes, A.B., Prendergast-Miller, M.T., Richardson, A.E., Toscas, P., Farrell, M., Macdonald, L.M., Baker, G., Wark, T., Thrall, P.H., 2015. Network analysis reveals that bacteria and fungi form modules that correlate independently with soil parameters. Environmental Microbiology 17, 2677–2689.
- Deveau, A., Bonito, G., Uehling, J., Paoletti, M., Becker, M., Bindschedler, S., Hacquard, S., Hervé, V., Labbé, J., Lastovetsky, O.A., Mieszkin, S., Millet, L.J., Vajna, B., Junier, P., Bonfante, P., Krom, B.P., Olsson, S., van Elsas, J.D., Wick, L.Y., 2018. Bacterial-fungal interactions: ecology, mechanisms and challenges. FEMS Microbiology Reviews 42, 335–352.
- Dumontet, S., Cavoski, I., Ricciuti, P., Mondelli, D., Jarrar, M., Pasquale, V., Crecchio, C., 2017. Metabolic and genetic patterns of soil microbial communities in response to different amendments under organic farming system. Geoderma 296, 79–85.
- Faust, K., 2021. Open challenges for microbial network construction and analysis. The ISME Journal 15, 3111–3118.
- Fierer, N., Jackson, J.A., Vilgalys, R., Jackson, R.B., 2005. Assessment of soil microbial community structure by use of taxon-specific quantitative PCR assays. Applied and Environmental Microbiology 71, 4117–4120.
- Friedman, J., Alm, E.J., 2012. Inferring correlation networks from genomic survey data. PLoS Computational Biology 8, 1–11.
- Fritsche, K., Van den Berg, M., De Boer, W., Van Beek, T.A., Raaijmakers, J.M., Van Veen, J.A., Leveau, J.H.J., 2014. Biosynthetic genes and activity spectrum of antifungal polyynes from Collimonas fungivorans Ter331. Environmental Microbiology 16, 1334–1345.
- Fuhrman, J.A., 2009. Microbial community structure and its functional implications. Nature 459, 193–199.
- Goslee, S.C., Urban, D.L., 2007. The ecodist package for dissimilarity-based analysis of ecological data. Journal of Statistical Software 22, 1–19.
- Haq, I.U., Zhang, M., Yang, P., Van Elsas, J.D., 2014. The interactions of bacteria with fungi in soil: emerging concepts. Advances in Applied Microbiology 89, 185–215.
- Hildebrandt, U., Janetta, K., Bothe, H., 2002. Towards growth of arbuscular mycorrhizal fungi independent of a plant host. Applied and Environmental Microbiology 68, 1919–1924.
- Hildebrandt, U., Ouziad, F., Marner, F.J., Bothe, H., 2006. The bacterium *Paenibacillus validus* stimulates growth of the arbuscular mycorrhizal fungus *Glomus intraradices* up to the formation of fertile spores. FEMS Microbiology Letters 254, 258–267.
- Hirano, H., Takemoto, K., 2019. Difficulty in inferring microbial community structure based on co-occurrence network approaches. BMC Bioinformatics 20, 1–14.
- Horner-Devine, M.C., Silver, J.M., Leibold, M.A., Bohannan, B.J.M., Colwell, R.K., Fuhrman, J.A., Green, J.L., Kuske, C.R., Martiny, J.B.H., Muyzer, G., Øvreås, L.,

Reysenbach, A.L., Smith, V.H., 2007. A comparison of taxon co-occurrence patterns for macro- and microorganisms. Ecology 88, 1345–1353.

Jenkinson, D.S., 1977. The soil biomass. New Zealand Soil News 25, 213–218.

- Konstantinidis, K.T., Tiedje, J.M., 2007. Prokaryotic taxonomy and phylogeny in the genomic era: advancements and challenges ahead. Current Opinion in Microbiology 10, 504–509.
- Kuiper, J.J., Van Altena, C., De Ruiter, P.C., Van Gerven, L.P.A., Janse, J.H., Mooij, W.M., 2015. Food-web stability signals critical transitions in temperate shallow lakes. Nature Communications 6, 1–7.
- Kurtz, Z.D., Müller, C.L., Miraldi, E.R., Littman, D.R., Blaser, M.J., Bonneau, R.A., 2015. Sparse and compositionally robust inference of microbial ecological networks. PLoS Computational Biology 11, 1–25.
- Leveau, J.H.J., Preston, G.M., 2008. Bacterial mycophagy: definition and diagnosis of a unique bacterial-fungal interaction. New Phytologist 177, 859–876.
- Ling, N., Zhu, C., Xue, C., Chen, H., Duan, Y., Peng, C., Guo, S., Shen, Q., 2016. Insight into how organic amendments can shape the soil microbiome in long-term field experiments as revealed by network analysis. Soil Biology and Biochemistry 99, 137–149.
- Lu, H., Lashari, M.S., Liu, X., Ji, H., Li, L., Zheng, J., Kibue, G.W., Joseph, S., Pan, G., 2015. Changes in soil microbial community structure and enzyme activity with amendment of biochar-manure compost and pyroligneous solution in a saline soil from Central China. European Journal of Soil Biology 70, 67–76.
- Lu, L., Yin, S., Liu, X., Zhang, W., Gu, T., Shen, Q., Qiu, H., 2013. Fungal networks in yield-invigorating and -debilitating soils induced by prolonged potato monoculture. Soil Biology and Biochemistry 65, 186–194.
- Luo, Y., van Veelen, H.P.J., Chen, S., Sechi, V., ter Heijne, A., Veeken, A., Buisman, C.J. N., Bezemer, T.M., 2022. Effects of sterilization and maturity of compost on soil bacterial and fungal communities and wheat growth. Geoderma 409, 115598.
- Ma, B., Wang, H., Dsouza, M., Lou, J., He, Y., Dai, Z., Brookes, P.C., Xu, J., Gilbert, J.A., 2016. Geographic patterns of co-occurrence network topological features for soil microbiota at continental scale in eastern China. The ISME Journal 10, 1891–1901.
- Mau, R.L., Liu, C.M., Aziz, M., Schwartz, E., Dijkstra, P., Marks, J.C., Price, L.B., Keim, P., Hungate, B.A., 2015. Linking soil bacterial biodiversity and soil carbon stability. The ISME Journal 9, 1477–1480.
- McMurdie, P.J., Holmes, S., 2013. Phyloseq: an R package for reproducible interactive analysis and graphics of microbiome census data. PLoS One 8, e61217.
- Nazir, R., Zhang, M., de Boer, W., van Elsas, J.D., 2012. The capacity to comigrate with Lyophyllum sp. strain Karsten through different soils is spread among several phylogenetic groups within the genus Burkholderia. Soil Biology and Biochemistry 50, 221–233.
- Neher, D.A., Weicht, T.R., Bates, S.T., Leff, J.W., Fierer, N., 2015. Changes in bacterial and fungal communities across compost recipes, preparation methods, and composting times. Biological Treatment of Solid Waste: Enhancing Sustainability 8, 119–144.
- Newman, M.E.J., 2006. Modularity and community structure in networks. In: Proceedings of the National Academy of Sciences of the United States of America, vol. 103, pp. 8577–8582.
- Oksanen, J., Blanchet, F., Friendly, M., Kindt, R., Legendre, P., McGlinn, D., Minchin, P., O'Hara, R., Simpson, G., Solymos, P., Stevens, M., Szoecs, E., Wagner, H., 2020. Vegan: community ecology package. R Package Version 2, 5–7.
- Palmer, J.D., Foster, K.R., 2022. Bacterial species rarely work together. Science 376, 581–582.
- Pérez-Piqueres, A., Edel-Hermann, V., Alabouvette, C., Steinberg, C., 2006. Response of soil microbial communities to compost amendments. Soil Biology and Biochemistry 38, 460–470.
- Prosser, J.I., Bohannan, B.J.M., Curtis, T.P., Ellis, R.J., Firestone, M.K., Freckleton, R.P., Green, J.L., Green, L.E., Killham, K., Lennon, J.J., Osborn, A.M., Solan, M., Gast, C.J. van der, Young, J.P.W., 2007. The role of ecological theory. Nature Reviews Microbiology 5, 384–392.
- Roelfsema, E., 2010. Influence of the Shrink and Swell Cycle on Soil Physical Properties of Loam and Clay Soils in the Netherlands. Master Thesis. Wageningen University & Research. https://edepot.wur.nl/160236.
- Rognes, T., Flouri, T., Nichols, B., Quince, C., Mahé, F., 2016. VSEARCH: a versatile open source tool for metagenomics. PeerJ 4, e2584.
- Schmid, C.A.O., Schröder, P., Armbruster, M., Schloter, M., 2018. Organic amendments in a long-term field trial—consequences for the bulk soil bacterial community as revealed by network analysis. Microbial Ecology 76, 226–239.
- Shin, K., van Diepen, G., Blok, W., van Bruggen, A.H.C., 2017. Variability of effective micro-organisms (EM) in bokashi and soil and effects on soil-borne plant pathogens. Crop Protection 99, 168–176.
- Sparling, G.P., 1985. The soil bomass. Soil Organic Matter and Biological Activity 25, 223–262.
- Stone, L., Roberts, A., 1990. The checkerboard score and species distributions. Oecologia 85, 74–79.
- Wagg, C., Schlaeppi, K., Banerjee, S., Kuramae, E.E., van der Heijden, M.G.A., 2019. Fungal-bacterial diversity and microbiome complexity predict ecosystem functioning. Nature Communications 10, 1–10.
- Weiss, S., Van Treuren, W., Lozupone, C., Faust, K., Friedman, J., Deng, Y., Xia, L.C., Xu, Z.Z., Ursell, L., Alm, E.J., Birmingham, A., Cram, J.A., Fuhrman, J.A., Raes, J., Sun, F., Zhou, J., Knight, R., 2016. Correlation detection strategies in microbial data sets vary widely in sensitivity and precision. The ISME Journal 10, 1669–1681.
- Xue, C., Ryan Penton, C., Zhu, C., Chen, H., Duan, Y., Peng, C., Guo, S., Ling, N., Shen, Q., 2018. Alterations in soil fungal community composition and network assemblage structure by different long-term fertilization regimes are correlated to the soil ionome. Biology and Fertility of Soils 54, 95–106.

Yang, W., Jing, X., Guan, Y., Zhai, C., Wang, T., Shi, D., Sun, W., Gu, S., 2019. Response of fungal communities and co-occurrence network patterns to compost amendment in black soil of northeast China. Frontiers in Microbiology 10, 1562.

Yang, W., Yang, Z., Guan, Y., Zhai, C., Shi, D., Chen, J., Wang, T., Gu, S., 2020. Dosedependent effect of compost amendment on soil bacterial community composition and co-occurrence network patterns in soybean agroecosystem. Archives of

Agronomy and Soil Science 66, 1027–1041.
Zhou, Z., Gao, T., Zhu, Q., Yan, T., Li, D., Xue, J., Wu, Y., 2019. Increases in bacterial community network complexity induced by biochar-based fertilizer amendments to karst calcareous soil. Geoderma 337, 691–700.