Changes in Bacterial and Fungal Soil Communities in Long-Term Organic Cropping Systems

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Abstract: Long-term organic farming aims to reduce synthetic fertilizer and pesticide use in order to sustainably produce and improve soil quality. To do this, there is a need for more information about the soil microbial community, which plays a key role in a sustainable agriculture. In this paper, we assessed the long-term effects of two organic and one conventional cropping systems on the soil microbial community structure using high-throughput sequencing analysis, as well as the link between these communities and the changes in the soil properties and crop yield. The results showed that the crop yield was similar among the three cropping systems. The microbial community changed according to cropping system. Organic cultivation with manure compost and compost tea (Org_C) showed a change in the bacterial community associated with improved soil carbon and nutrient content. A linear discriminant analysis effect size showed different bacteria and fungi as key microorganisms for each of the three different cropping systems, for conventional systems (Conv), different microorganisms such as Nesterenkonia, Gallibacter, Gramella, Limnobacter, Pseudoalteromonas, Pantoe, and Sporobolomyces were associated with pesticides, while for Org_C and organic cultivation with manure (Org_M), other types of microorganisms were associated with organic amendments with different functions, which, in some cases, reduce soil borne pathogens. However, further investigations such as functional approaches or network analyses are needed to better understand the mechanisms behind this behavior.

Keywords: compost; high-throughput sequencing; sheep manure; soil properties; crop yield; organic farming; microbial community structure

1. Introduction

Currently, one third of agricultural soils worldwide are moderately or highly degraded, thus affecting production [1]. Chemical fertilizers, pesticides, and herbicides are commonly used to maintain soil fertility and crop production in conventional farming [2], and this has generated an increase in greenhouse gas emissions and soil degradation, as well as a decrease in soil biodiversity [3]. In this context, organic farming is of growing interest. Organic farming includes the use of organic fertilizers, such as compost, manure, or green manure, and places an emphasis on techniques like rotation with companion
plants or intercropping, and pest and disease control by natural methods, avoiding synthetic chemical compounds, preserving the environment, and providing human beings with high-nutrition crops that are free of chemicals [4]. In the European Union, almost 180 million hectares are dedicated to agriculture, of which around 13 million were dedicated to organic farming in 2018. Spain is the EU country with the largest area devoted to organic farming [5], with more than 2.2 million hectares, accounting for 16.7% of the total farmed land in the country.

Some studies have reported that organic crop production is between 10–30% lower than in conventional farming, with differences depending on the crop species, growing conditions, and management practices [6]. Nonetheless, recent studies have highlighted that long-term organic management can contribute crop yields similar to those found in conventional farming, once the system has been stabilized, after initial years with a reduction in crop yields [7,8]. For this reason, more information about the relative variability of organic systems compared with conventional ones is necessary in order to reinforce the use of organic farming.

Soil microorganisms play an important role in ecosystem processes, such as carbon cycling, decomposition, nutrient cycling, and soil aggregate formation [9], and determining this relationship is very complex [10]. Understanding how these microorganisms respond to organic matter, inorganic fertilizers, and soil management can help farmers to improve soil health for crop production [10–12]. Approaches like high-throughput analysis of bacterial and fungal communities can show taxonomic shifts, shaping the patterns of the ecological interactions that regulate the structure, function, and resilience of soil microbial communities under organic farming compared with conventional farming.

However, there are no conclusive results regarding this. Some studies have reported that after long-term organic farming, the microbial diversity, soil sustainability, and beneficial microorganisms involved in plant health were higher than in conventional farming [13–15]. However, Bell et al. [16] and Krishnaraj and Sabale [17] found that organic cropping systems showed no significant differences, or showed an even lower microbial diversity, compared with conventional cropping. So, we found a gap of knowledge that needs more results from long-term field experiments, such as the one presented here, to know how soil microbial communities change across different organic farming types.

The aim of this paper was to investigate the potential differences between two organic systems (organic cultivation with manure compost and compost tea (Org_C) and organic cultivation with manure (Org_M) with a conventional system (Conv). For this purpose, in the three studied agricultural systems, the soil bacterial and fungal communities were studied through high-throughput sequencing analysis, focusing and studying their relationship using physico-chemical soil properties and crop production. We hypothesized that (a) organic systems would not have significant differences on crop production compared with the conventional system, as once an organic system is stabilized, soil functionality is improved, contributing to a high nutrient availability and soil health; (b) both organic systems (Org_C and Org_M) would promote changes in the microbial structure and abundance compared with the conventional system; and (c) that changes promoted by cropping system can be either beneficial or detrimental to plants, thus influencing soil stability and quality.

2. Materials and Methods
2.1. Experiment Description and Sampling

The study site was located in the Campo de Cartagena, an agrarian region of southeastern Spain. The soil was a Haplic Calcisol (Loamic, Hypercalcic) IUSS [18]. The area has a mean annual temperature of 17.5 °C, a mean annual precipitation of 280 mm, and an annual potential evapotranspiration of 1300 mm. The site has been under vegetable cultivation since the early 1990s, using drip fertigation, rotation, and multiple cropping. Three cropping systems were selected for this study, where the following five random
plots (~1 ha) were set up for each cropping system: (1) a conventional system using a yearly addition of sheep manure as an organic amendment, inorganic fertilizer for fertigation, and pesticides (Conv); (2) an organic system using a yearly addition of compost, amino acids to provide N, compost tea to provide organic compounds and nutrients for fertigation, no pesticides, and the use of cover crops of oat (*Avena sativa*) and vetch (*Vicia sativa*) between cropping seasons (Org_C); and (3) an organic system using a yearly addition of sheep manure, amino acids as fertigation to provide N, and no pesticides (Org_M). The three cropping systems are described in detail in Table 1. One composite sample derived from ten subsamples (0–10 cm depth) was collected using an auger from each plot in February 2018 after the harvest of a leaf cabbage crop (*Brassica oleracea var. sabellica*).

| Cultivation System | Conv | Org_C | Org_M |
|--------------------|------|-------|-------|
| Geographical coordinates | 37°48′18.5″ N, 0°51′49.2″ W | 37°51′39.3″ N, 0°54′03.3″ W | 37°49′30.2″ N, 0°52′28.4″ W |
| Crop 2017–2018 season | *Brassica oleracea var. sabellica* | | |
| Harvest | Manual on 20–25 February 2018. Crop residues were incorporated in the soil | | |
| Crops grown in previous years | *Lactuca sativa/Brassica oleracea var. Italica* (2015/2016) | *Aptium graveolens/Cucumis melo* (2016/2017) |
| | *Foeniculum vulgare/Cucurbita moschata* (2012/2013) | | |
| | *Apium graveolens/Cucumis melo* (2014/2015) | | |
| | *Lactuca sativa/Brassica oleracea var. Italica* (2015/2016) | | |
| | *Brassica oleracea var. Italica/Capsicum annum* (2013/2014) | | |
| Organic amendments (amount per year) | 15,000 kg ha⁻¹ sheep manure | 10,000 kg ha⁻¹ sheep compost; compost tea * | 15,000 kg ha⁻¹ sheep |
| Fertilizers (amount per year) | 15 kg ha⁻¹ ENTEC solub 21 (ammonium sulfate with inhibition of nitrification); 10 L ha⁻¹ phosphoric acid; 15 kg ha⁻¹ calcium nitrate; 10 L ha⁻¹ nitric acid | 10 L ha⁻¹ EcoZen NPK 2-2-7 (aminoacids); 10 L ha⁻¹ Sunfol veg agri 12% (aminoacids) | |
| Pesticides | Linuron; Indoxacarb 30%; Cypermethrin; Lambda chialotrin 10%; Imidacloprid 20%; Spinosad; Azadirachtin 3.2%; Emamectin 0.85%; Clortalonil 50%; Difenoconazol 25%; Azoxystrobin 2.5%; Propamocarb 52%; Ciflufenamid | No application of chemical pesticides | |

Conv—conventional system; Org_C—organic cultivation with sheep manure compost and compost tea; Org_M—organic cultivation with sheep manure. * The compost tea was made on each farm by steeping mature compost in water for 24 h.

The characteristics of the sheep manure, compost, and tea compost are shown in Table S1. Sheep manure and its compost were chosen because of their proximity to the experimental plot, thereby reducing the environmental and economic impact of their transport to the experimental site. The amount of compost added annually in the Org_C treatment before each crop cycle was lower than the amount of manure added before each cycle for the Conv or Org_M treatments; because soluble organic compounds were continuously added during the crop cycle by fertigation with the compost tea in Org_C, the organic matter was not added all at once but rather over an extended period of time (Table 1).
Soil samples were separated into two aliquots. One was kept at ambient temperature for physicochemical analyses and the other was stored in a cool box with ice for a molecular analysis. The samples were taken to the lab immediately. The soil for the molecular analysis was sieved at <2 mm and stored at −20 °C. The soil was air-dried for one week for the physicochemical analyses and sieved at <2 mm. The soil cores were taken using steel cylinders to determine the soil bulk density [19].

2.2. Soil Properties and Crop Yield

The actual field soil moisture content (FMa) was gravimetrically determined according to De Angelis [20]. The cation exchange capacity (CEC) and exchangeable Ca²⁺, Mg²⁺, K⁺, and Na⁺ were determined using BaCl2 as the exchangeable cation, following ISO (international standard method) 13536 [21]. The soil water content at wilting point (SWW) and soil water content at field capacity (SWFC) were calculated using the retention curve method [22], in which moist samples were dried by raising the air pressure in an extractor with a porous ceramic plate [23]. The soil pH and electrical conductivity (EC) were measured in deionized water (1:5 w/v). The total organic carbon (TOC), inorganic carbon (IC), and total nitrogen (TN) were determined using an elemental CHNS-O analyzer (EA-1108, Carlo Erba, Barcelona, Spain), and the CaCO₃ content was calculated from the IC. The particulate organic carbon (POC), defined as a fresh or decomposing organic material, was measured according to Cambardella and Elliot [24]; in long-term experiments, POC can be used as an early indicator of soil organic matter (SOM), corresponding to the functional pool of organic matter stabilized by specific mechanisms [25]. Soil NH₄⁺ was extracted with 2M KCl in a 1:10 soil/extractant ratio, and was calorimetrically measured [26,27]. Soil NO₃⁻ was extracted with deionized water in a 1:10 soil:extractant ratio and was measured by ion chromatography (Metrohm 861). The available P (P) was measured using the Olsen method [28]. The available Fe, (Fe) Mn (Mn), Cu (Cu), and Zn (Zn) were extracted by chelation using DTPA (1:2 w/v) [29,30]. The available B (B) was extracted with deionized water (1:5 w/v) at 50 °C [31]. The available nutrients were measured using ICP-MS (7500CE, Agilent, Santa Clara, CA, USA) and gas chromatography triple quadrupole mass spectrometry (GC-MS; Agilent 7890B GC coupled to an Agilent 7010B MS system, Santa Clara, CA, USA).

The leaf cabbage yield (kg ha⁻¹) was calculated based on the weight of all marketable plants (suitable for sale according to size) in each plot.

2.3. DNA Extraction, PCR, and Sequencing

Soil DNA was extracted from 1 g of soil (wet weight) using the DNeasy Power Soil Kit (Qiagen, Hilden, Germany), following the manufacturer’s protocol, with the modifications described by Taskin et al. [33]. The quantity and quality of DNA extracts were quantified using a Qubit 3.0 Fluorometer (Invitrogen, Thermo Fisher Scientific, Waltham, MA, USA) and a NanoDrop 2000 fluorospectrometer (Thermo Fisher Scientific, Waltham, MA, USA).

The bacterial community was determined through the next-generation-sequencing of bacterial 16S hypervariable regions using the Ion Torrent™ Personal Genome Machine™ (PGM, London, UK) System. Bacterial 16S regions were amplified using an Ion 16S™ Metagenomics Kit (Thermo Fisher Scientific, Waltham, MA, USA) with two different degenerate primer sets to amplify regions V2–4–8 and V3–6, V7–9. The amplified 16S amplicons were then processed using an Ion Xpress™ Plus Fragment Library Kit in combination with an Ion Xpress™ Barcode Adapter 1-96 Kit (Thermo Fisher Scientific, Waltham, MA, USA). All of the purification processes between incubation and the amplification reactions of library preparation were processed using DynaMag™-2 magnetic racks.
(Thermo Fisher Scientific, Waltham, MA, USA) and an AMPure XP Purification Kit (Beckman Coulter, Brea, CA, USA). Library preparation and barcoding were followed by the determination of the size and concentration of the final libraries using an Agilent 2100 Bioanalyzer system and the Agilent High Sensitivity DNA kit (Agilent, Santa Clara, CA). Sequencing templates were prepared using an Ion One Touch 2 System and an Ion PGM™ Hi-Q™ View OT2 Kit (Thermo Fisher Scientific, Waltham, MA, USA). The sequencing reaction was performed using Ion Torrent PGM (Thermo Fisher Scientific, Waltham, MA, USA) with an Ion PGMTM Hi-QTM View Sequencing Kit (Thermo Fisher Scientific, Waltham, MA, USA).

Fungal libraries were prepared using a protocol based on the method proposed by Smith and Peay [34], with some modifications. Sequencing libraries were produced by PCR amplification using the primer pairs ITS1–ITS2 tailed with the Illumina adapters (Illumina, San Diego, CA, USA). The reverse primers were barcoded using the 12-base Golay barcodes [35]. DNA extraction had a minimum concentration of 10 ng/µL with a least 200 ng provided. The PCR amplifications were conducted in a final volume of 30 µL containing 3 µL of buffer 10×, 0.7 µL of each primer (10 mM), 0.6 µL of 10 mM dNTP, 2 µL of template DNA (10 ng/µL), 21.98 µL of PCR-grade water, and 0.12 µL of Invitrogen Platinum Taq DNA polymerase High Fidelity (Cat N° 11304-011), using the following conditions: 3 min initial denaturation at 95 °C, followed by 35 cycles of denaturation at 95 °C for 45 s, annealing at 50 °C for 1 min, extension at 72 °C for 1 min, and a final extension at 72 °C for 10 min. The PCR products were cleaned up from primers using Agencourt AMPure XP beads (Beckman Coulter, Indianapolis, IN, USA), following the manufacturer’s instructions. The PCR products were checked on a Bioanalyzer DNA 1000 kit (Agilent, Santa Clara, CA, USA) to verify the size. Amplicons were quantified with Qubit using the dsDNA HS Assay kit (Thermo Fisher Scientific, Waltham, MA, USA). Amplicon libraries were sequenced on the Illumina MiSeq machine (Illumina, San Diego, CA, USA), together with a 10% PhiX (Illumina, San Diego, CA, USA), control library to generate 300 bp paired end reads.

2.4. Sequencing Data Processing

For bacterial raw sequences, the barcodes and primers were trimmed according to the BaseCaller software (ThermoFisher, Waltham, US). The sequences were denoised with ACACIA v 1.53 [36], and low quality sequences were discarded using the Quantitative Insights into Microbial Ecology (QIIME) pipeline v 1.9.1 [37] from the Microbiome Helper Virtual Box v 2.3 [38]. Briefly, bacterial sequences with a Q < 25 were removed and the retained sequences were then assigned to Operational Taxonomic Units (OTUs) based on 97% similarity with the SILVA reference database 128 after filtering chimeras using VSEARCH v 2.4.3 [39] with the ribosomal database project (RDP_trainset16_022016.fa) [40]. Low confidence OTUs were removed. To correct the sampling effect, the number of sequences was established at 19,840.

Fungal raw reads were trimmed for adapters and low quality reads using the Trimomatic v 0.38 program [41], setting the quality cutoff to 20 in 24 bp sliding windows. Trimmed reads were assembled using the paired-end read merger (PEAR) program v 0.9.10 [42]. Chimeras were removed using VSEARCH v 2.4.3 from the QIIME pipeline v 1.9.1 using the UCHIME reference dataset (uchime_sh_refs_dynamic_original_985_03.07.2014.fasta). OTUs were assigned based on 97% similarity using the open reference OTU picking protocol implemented in the QIIME toolkit v 1.9.1. Taxonomy was assigned using the UNITE database (sh_taxonomy_qiime_ver7_dynamic_20.11.2016.txt)[43]. Low abundance OTUs (OTUs with less than three reads) were removed.

The sequences were uploaded to the European Nucleotide Archive (ENA) with the study accession code PRJEB38121.

2.5. Fungal Pathogen Detection by qPCR
Real-time PCR was performed to quantify the number of ITS copies in the soil DNA using a 7500 fast real-time PCR system (Applied Biosystems, Foster City, CA, USA). The reaction mixtures (15 µL) contained a final concentration of 1× TaqMan Universal Master Mix II no UNG (Applied Biosystems, Foster City, CA, USA), 0.3 µm of each primer, 0.1 µm of TaqMan probe, 0.1 mg mL\(^{-1}\) of bovine serum albumin (BSA), 3 µL of DNA template, and nuclease-free water. The PCR program consisted of an initial denaturation step at 95 °C for 10 min, followed by 40 cycles at 95 °C for 10 s, and at 60 °C for 40 s, as well as a final step at 50 °C for 2 min. Three real-time PCRs were carried out for each DNA sample. The amplification results were analyzed with 7500 Fast Real-time PCR software v.2.0 (Applied Biosystems, Foster City, CA, USA). Fungal pathogen detection and quantification were performed using the Vegalert qPCR quantitative kits for curcubits (\textit{Alternaria} spp. (ALT), \textit{Rhizoctonia solani} (RSO), and \textit{Fusarium oxysporum} (FOX); Microgaia Biotech S.L, Murcia, Spain).

2.6. Statistical Analysis

The normality and homogeneity of variance assumptions were evaluated using Shapiro-Wilk and Levene’s tests. For a mean comparison between the cropping systems, one-way analysis of variance (ANOVA) was performed, followed by Tukey’s honestly significant difference (HSD) post hoc test if the effects were significant. Where conditions for homoscedasticity were not met, we used Welch’s test followed by the “pairwise.t.test” function, with P adjusted by the Bonferroni–Holm method for multiple comparisons [44]. Non-parametric Kruskal-Wallis tests were used when normality assumptions were not fulfilled. When such test statistics were significant, Kruskal-Wallis multiple comparison Z-value tests were performed using the “dunnTest” function, with P adjusted by the Benjamini–Hochberg method in the FSA package v 0.8.3 [45]. Principal component analysis (PCA) was conducted as an unsupervised learning dimension reduction technique to visualize the cohesion and separation of the three cultivation systems. From the outcomes reported by the FactoMineR package v 1.42 [46], a PCA biplot was generated using the factoextra package v 1.0.5 [47] to assess the contribution of each parameter to the component loading.

For both bacterial and fungal communities, the rarefaction curves and the Chaol and Shannon diversity indexes were calculated using the R v 1.1.453 packages of iNEXT v 2.0.19 [48] and vegan [49]. The effects of the cultivation systems on such indexes were evaluated by one-way ANOVA. Significant differences were tested by Tukey’s HSD test. Violin plots were generated to show the distributional shape of each index across all soil samples grouped according to the cropping system.

A similarity percentages (SIMPER) analysis was conducted using the “simper” function of the vegan package v 2.5.6 to identify the parameters that most contributed to the pairwise differences between the cropping systems at a phylum level. PERMANOVA was conducted to test the differences among the cropping systems if the homogeneity of variance assumption was met, and an analysis of similarities (ANOSIM) was carried out if not.

A linear discriminant analysis (LDA) effect size (LEfSe; Galaxy community hub https://huttenhower.sph.harvard.edu/galaxy/ accessed on November 20, 2020) under the default parameters was implemented to identify differentially abundant groups among the three cropping systems [50,51].

In order to visualize and test whether the microbial community structures (OTUs) of the three cropping systems were distinct, a principal coordinate analysis (PCoA) and permutational multivariate analysis of variance (PERMANOVA) were conducted based on the Bray-Curtis distance, using the “betadisper” and “adonis” functions in the vegan package with 999 permutations. The soil microbial community composition was ordinated applying non-metric multidimensional scaling (NMDS) with the Bray-Curtis dissimilarity matrices using the “metaMDS” function in the vegan package. During the
NMDS analysis, the relationships between the soil properties and soil microbial community were assessed using the “envfit” function available in the vegan package.

3. Results

3.1. Effects of Different Cropping Systems on Soil Physico-Chemical Properties, Soil Pathogens, and Crop Yield

The univariate analysis showed that Org_M had a significantly higher pH (8.70) than Conv (8.39), with no significant differences with Org_C (8.47; Table 2). The TOC, TN, and NH₄⁺ contents were significantly higher in Org_C than in the other two systems, while the NO₃⁻ content was significantly higher in Conv than in Org_C. POC showed significantly higher values in Org_M than in Org_C. Fe, Mn, and B showed significantly higher values in Org_C and Conv than in Org_M. Mg was significantly higher in Org_C than in Org_M, and K was significantly higher in Org_C than in Conv (Table 2). The Conv system showed the significantly highest amount of total pesticides (TP).

Table 2. Effects of the three cropping systems on soil properties.

| Soil Properties | Conv                 | Org_C                 | Org_M                 |
|-----------------|----------------------|-----------------------|-----------------------|
| pH              | 8.39 ± 0.17 b        | 8.47 ± 0.14 ab        | 8.70 ± 0.10 a         |
| EC (dS m⁻¹)     | 0.54 ± 0.15          | 0.52 ± 0.13           | 0.38 ± 0.04           |
| TOC (g kg⁻¹)    | 11.49 ± 0.28 ab      | 15.64 ± 3.37 a        | 9.01 ± 3.49 b         |
| TN (g kg⁻¹)     | 1.13 ± 0.19 b        | 1.59 ± 0.34 a         | 0.93 ± 0.24 b         |
| POC (g kg⁻¹)    | 2.67 ± 0.72 ab       | 2.20 ± 0.55 b         | 4.03 ± 1.40 a         |
| NH₄⁺ (mg kg⁻¹)  | 0.10 ± 0.23 b        | 1.33 ± 0.15 b         | 0.00 ± 0.00 a         |
| NO₃⁻ (mg kg⁻¹)  | 53.04 ± 28.27 a      | 11.86 ± 7.10 ab       | 27.00 ± 13.57 b       |
| Bulk density (kg dm⁻³) | 1.24 ± 0.06          | 1.27 ± 0.07           | 1.34 ± 0.09           |
| SWW (cm³ cm⁻³)  | 0.12 ± 0.01          | 0.12 ± 0.02           | 0.12 ± 0.03           |
| SWFC (cm³ cm⁻³) | 0.22 ± 0.01          | 0.25 ± 0.03           | 0.22 ± 0.02           |
| CEC (cmol kg⁻¹) | 14.82 ± 0.86         | 17.47 ± 4.13          | 12.76 ± 2.20          |
| CaCO₃ (%)       | 44.65 ± 2.71         | 45.54 ± 7.57          | 47.03 ± 1.92          |
| FMA (cm³ cm⁻³)  | 0.17 ± 0.03          | 0.20 ± 0.06           | 0.19 ± 0.03           |
| Ca (cmol kg⁻¹)  | 8.44 ± 0.83          | 10.03 ± 2.40          | 7.19 ± 1.49           |
| Mg (cmol kg⁻¹)  | 3.54 ± 0.11 ab       | 4.39 ± 1.09 a         | 3.13 ± 0.54 b         |
| K (cmol kg⁻¹)   | 0.62 ± 0.15 b        | 0.85 ± 0.17 a         | 0.78 ± 0.06 ab        |
| Na (cmol kg⁻¹)  | 2.12 ± 0.32          | 2.19 ± 0.86           | 1.64 ± 0.23           |
| P (mg kg⁻¹)     | 20.15 ± 5.24         | 14.65 ± 7.71          | 14.33 ± 7.48          |
| Cu (mg kg⁻¹)    | 2.17 ± 0.74          | 3.17 ± 0.81           | 2.19 ± 0.75           |
| Zn (mg kg⁻¹)    | 4.75 ± 2.99          | 5.46 ± 1.65           | 4.48 ± 0.91           |
| Fe (mg kg⁻¹)    | 6.19 ± 2.91 a        | 6.97 ± 2.69 b         | 2.99 ± 1.24 a         |
| Mn (mg kg⁻¹)    | 9.47 ± 1.03 a        | 7.91 ± 2.37 a         | 4.66 ± 0.36 b         |
| B (mg kg⁻¹)     | 1.68 ± 0.11 a        | 1.94 ± 0.28 a         | 1.31 ± 0.17 b         |
| TP (ng g⁻¹)     | 232.00 ± 146.54 a    | 13.36 ± 9.11 b        | 6.08 ± 5.21 b         |
| ALT (log copy ITS g⁻¹ soil) | 4.24 ± 0.32 a      | 4.33 ± 0.49 a         | 2.05 ± 1.88 b         |
| RSO (log copy ITS g⁻¹ soil) | 0.67 ± 1.51      | 2.72 ± 1.61           | 2.02 ± 1.86           |
| FOX (log copy ITS g⁻¹ soil)    | 3.08 ± 1.73 a        | 3.30 ± 0.44 a         | 0.63 ± 1.41 ab        |

EC—electrical conductivity; TOC—total organic carbon; TN—total nitrogen; POC—particulate organic carbon; SWW—soil wilting point; SWFC—field capacity; CEC—cation exchange capacity; FMA—actual field soil moisture; Available (Ca, Mg, K, Na, P, Cu, Zn, Fe, Mn, and B); TP—total pesticides; ALT—Alternaria spp.; RSO—Rhizoctonia solani; FOX—Fusarium oxysporum. Conv—conventional system; Org_C—organic cultivation with sheep manure compost and compost tea; Org_M—organic cultivation with sheep manure. Values (mean ± standard deviation n = 5) followed by different lower letters correspond to significant differences between cultivation systems (Tukey’s test or pairwise t-test by groups); (ns)
non-significant differences between cultivation systems. (-) the test does not proceed; significant levels: *** \( p < 0.001; ** \( p < 0.01; * \ p < 0.05.

The abundance of *Alternaria* spp. (ALT) was significantly the lowest in Org_M, while *Fusarium oxysporum* (FOX) was significantly lower in Org_M than in Conv and Org_C; there were no significant differences in *Rhizoctonia solani* (RSO) among the cropping systems (Table 2).

The obtained biplot reflected the differences among the three cropping systems (Figure 1). Org_C was associated with TOC, TN, B, Mg, and NH_4^+, contrary to Org_M, which was negatively correlated with POC (Figure 1). The Conv system was positively correlated with the total pesticide content.

![Figure 1. Principal component analysis (PCA) of the soil properties. The two first principal components are shown on the x- and y-axes, respectively. PCA scores represent soil samples, with colors indicating the corresponding cultivation systems. Ellipses represent 70% confidence intervals around the barycenters for the samples classified by each cultivation system. TOC—total organic carbon; TN—total nitrogen; POC—particulate organic carbon; Mg—exchangeable Mg; Fe—bioavailable Fe; Mn—bioavailable Mn; B—bioavailable B; TP—total pesticides; ALT—*Alternaria* spp.; FOX—*Fusarium oxysporum*; Conv—conventional system; Org_M—organic cultivation system with sheep manure; Org_C—organic cultivation system with sheep manure compost and compost tea (n = 5, per cropping system).](image)

No significant differences were found for the cabbage yield among the three cropping systems (Figure 2).
Figure 2. The crop yield (kg ha$^{-1}$) of the three cropping systems. Error bars represent mean ± standard deviation (SD; $n = 5$). No significant differences were found among the three cultivation systems (Kruskal–Wallis test, $p > 0.05$). Conv—conventional system; Org_M—organic cultivation system with sheep manure; Org_C—organic cultivation system with sheep manure compost and compost tea.

3.2. Effects of Different Cropping Systems on Soil Microbial Diversity

A total of 592,250 16S sequences (clustered into 18,533 OTUs) for bacteria and 1,186,964 ITS sequences (clustered into 611 OTUs) for fungi were obtained from all of the soil samples. Rarefaction curves showed a coverage value of 0.84–0.88 for bacteria and 0.99 for fungi (Figure S1). Different microbial indices were calculated for the microbial communities (Figure 3 and Table S2). The Chao1 index showed no significant differences in the bacterial and fungal communities among the three cropping systems (Figure 3 and Table S2). The Shannon index showed no differences for the bacterial community, but did show significant differences for fungi, which had significantly higher values in Conv and Org_M than in Org_C (Figure 3 and Table S2).
Figure 3. Violin plots displaying the diversity indexes in the three cultivation systems. The distributional features of the data are depicted by the kernel density trace overlaid on the descriptive statistics (median and whisker range from 25% to 75%) represented by a boxplot. Chao1 for (A) bacterial and (B) fungal communities, and Shannon index for (C) bacterial and fungal (D) communities. Conv—conventional system; Org_M—organic cultivation system with sheep manure; Org_C—organic cultivation system with sheep manure compost and compost tea (n = 5, per cropping system).

The PCoAs of the bacterial (Figure 4A) and fungal (Figure 4B) microbial communities showed significant differences among the different cropping systems, which were confirmed by PERMANOVA (F = 1.792, P = 0.006; F = 7.649, and P = 0.001 respectively).

Figure 4. Principal coordinate analysis (PCoA) of the (A) bacterial and (B) fungal community structures between the three cropping systems. Different colors and shapes represent the different points of the group—Conv is indicated by red circles, Org_C by green triangles, and Org_M by blue crosses. PCoA displays the group centroids and dispersions. Conv—conventional system; Org_C—organic cultivation with manure compost and tea compost; Org_M—organic cultivation with manure (n = 5, per cropping system).

3.3. Effects of Different Cropping Systems on Bacterial and Fungal Community Composition

The main bacterial and fungal taxa found in the different cropping systems are shown in Figures 5–7. The most dominant phyla under the different cropping systems...
were Proteobacteria (42% in average), Actinobacteria (16%), Bacteroidetes (12%), and Acidobacteria (7%; Figure 5A and Table S3). No significant differences were found among the three cropping systems when using Bray–Curtis dissimilarity at a phylum level (PERMANOVA: F = 0.821; P = 0.562). SIMPER pairwise comparisons showed that the Proteobacteria, Gemmatimonadetes, Bacteroidetes, and Actinobacteria accounted for 75% of the overall dissimilarities between Conv and Org_M or Org_C.

![Figure 5](image)

**Figure 5.** Relative abundance (>1%) of (A) bacterial and (B) fungal phylum of the three cropping systems. Bar values are mean ± SD (n = 5). Conv—conventional system; Org_C—organic cultivation with manure compost and tea compost; Org_M—organic cultivation with manure.

The most abundant fungal phylum was Basidiomycota (27% on average), followed by Olpidiomykota (25%), Ascomycota (15%), Glomeromycota (15%), and Mortierellomycota (13%; Figure 5B and Table S4). Significant differences were found between the three cropping systems (PERMANOVA: F = 4.83, P = 0.003). The highest relative abundance of Basidiomycota and Glomeromycota was observed in Org_M, while Olpidiomykota showed the highest abundance in Org_C. The highest abundance for Ascomycota and Mortierellomycota was found for the Conv system, followed by Org_M and Org_C (Table S4). SIMPER showed that Glomeromycota, Basidiomycota, and Ascomycota accounted for 73% of the dissimilarities between Conv and Org_M, while Olpidiomykota, Basidiomycota, and Ascomycota accounted for 74% of the dissimilarities between Conv and Org_C.

LEfSe was conducted to identify the taxa that display significant differences among the three cropping systems. For bacteria, the Conv system had fifteen differential taxa, while Org_C had seven and Org_M had eleven (Figures 6A and 7A). At the genus level, *Nesterenkonia* (Actinobacteria); *Galbibacter* and *Gramella* (Bacteroidetes); and *Limnobacter*, *Pseudoalteromonas*, and *Pantoe* (Proteobacteria) were the most differential taxa in the Conv system. The genomic features in the organic systems identified the genera *Aciditerrimonas* and *Isoptericola* (Actinobacteria), *Nibribacter* (Bacteroidete), *Haliangium* (Proteobacteria), and *Terrimicrobium* (Verrucomicrobia) as important taxonomic contributors for Org_M, and the genera *Dinghuibacter* and *Turibacter* (Bacteroidete), *Piscibacillus* and *Planifilium* (Firmicutes), and *Roseibacillus* (Verrucomicrobia) as important taxonomic contributors for Org_C (Figures 6A and 7A). For fungi, the Conv system showed two differential taxa; among the organic systems, Org_C showed two differential taxa and Org_M seventeen
The genus *Sporobolomyces* (Basidiomycota) was the only genomic feature in the Conv system. For the organic systems, we found the genera *Wallemiales* (Basidiomycota), *Funneliformis* (Glomerales), *Melanoleuca* (Basidiomycota), and *Alternaria* and *Paraphaeosphaera* (Ascomycota) in Org_M, and the genus *Candida* (Ascomycota) in Org_C.

**Figure 6.** Taxonomic cladogram obtained from the LEfSe of (A) 16S rDNA and (B) ITS. Taxa of microorganisms are highlighted by colored circles and shaded areas (Conv, Org_C, and Org_M are shown in red, green, and blue, respectively). Each circle represents a taxa and each circle’s diameter reflects the abundance of that taxa in the community. Conv—conventional system; Org_C—organic cultivation with manure compost and tea compost; Org_M—organic cultivation with manure (n = 5, per cropping system).
Figure 7. Linear discriminant analysis effect size (LEfSe) analysis showing (A) bacterial and (B) fungal microbiota changes between the three cropping systems. Conv—conventional system; Org_M—organic cultivation with manure; Org_C—organic cultivation with manure compost and tea compost (n = 5, per cropping system).
3.4. The Relationship between the Microbial Community and Soil Properties

A nonmetric multidimensional scaling (NMDS) was assayed for establishing the relationship between bacterial and fungal communities and the significant soil parameters. For bacteria, TOC, TN, NH4+, Mg, and B were the soil properties that showed a significant effect on the bacterial community composition (Table S5). However, for fungi, no significant correlation was found (Table S6).

A Spearman correlation analysis between the soil properties and bacterial and fungal genera showed a significant correlation between *Nesterenkonia* and total pesticide (0.68 **); *Gramella* and total pesticide (0.52 *); *Pseudoalteromonas* and *Pantoe* with total pesticide (0.56 *); *Sporobolomyces* and total pesticide (0.80); *Planifilium*, *Dinghuibacter*, *Turicibacter*, *Piscibacillus*, *Melanoleuca*, *Filobasidium*, and *Candida* showed a high correlation with NH4+ (0.60 *, 0.66 **, 0.59 *, 0.74 **, −0.77 **, −0.76 **, and 0.70 **, respectively); *Aciditerrimonas*, *Roseibacillus*, and *Nibribacter* showed a high correlation with TOC (0.82 ***, 0.74 **, and −0.65**; respectively) and *Aciditerrimonas* with total nitrogen (0.64 **); *Terrimicrobium* with NH4+ (−0.58 *); and *Wallemia* with both pathogens *Alternaria* spp (−0.77 **) and *Fusarium oxysporum* (−0.56 **) and with NH4+ (−0.66 *).

4. Discussion

Organic and conventional farming are nowadays defined based on differences regarding fertilization, plant cultivation, and soil management. Our three cropping systems have the same soil type and climate, are located in close proximity, and have undergone the same crop rotation. Therefore, it is likely that the differences observed in soil properties and microbial community are probably due to crop systems. The results highlight that once organic systems are stabilized over time, production can be as high as that of conventional systems if the soil and crops are effectively managed and nutrient availability is ensured [52,53]. Differences among the cropping systems related to soil properties indicated that the higher organic carbon content in Org_C may have contributed to the slightly lower pH in soil compared with Org_M, owing to the greater presence of organic acids [54,55]. The addition of compost instead of manure may therefore provide positive effects in basic soils, contributing to a decrease in pH and making nutrients more available. Meanwhile, the highest pH showed by Org_M can be explained by the buffering capacity from bicarbonates and organic acids in the manure, with no addition of nitric acid, as used in Conv to decrease the pH [56]. On the other hand, the higher NO3− levels observed in the Conv system were probably due to intensive chemical fertilizer and pesticide use, and the addition of nitric acid in fertigation [57,58].

It is interesting to highlight that soil parameters such as TOC and some nutrients like TN, Mg, and B were associated with changes in the soil bacterial community, as observed by Yang et al. [59] and Zhang et al. [60]. It is well known that compost enhances carbon and nitrogen in soils, thus changing the microbial communities [61]. In addition, Vera et al. [62] observed that the boron content in the soil was a determinant property explaining the changes in the bacterial community in agricultural soil. According to Vera et al. [63], organic matter had a key influence on the potential microbial action with high boron doses, and our findings could support that influence, as some key microorganisms from soil organic cropping system had a high correlation with boron.

Although the bacterial community structure changed in response to the different cropping systems, this was not associated with significant variations in alpha diversity [64]. It was contrary to the lower values expected in conventional systems due to the adverse effects of agrochemicals [65–67]. Moreover, a significantly higher fungal diversity was observed in the Conv system, probably due to the increase in nutrient availability, as previously observed by Geisseler and Scow [68] and Leff et al. [69], and the breakage of fungal hyphae by tillage [70]. Legacy effects of cropping systems occur in specific microbial groups and cannot be resolved by determining the diversity of the entire microbial community, as shifts in some groups might be compensated for shifts in others [71].
Most importantly, our results showed that there are different effects from the various cropping systems on the soil microbial community structure. Indeed, the impact of the cropping system is considered to be stronger than any potential spatiotemporal variations [72]. The effect of the different cropping systems was not reflected in the dominant bacterial phyla, contrary to that observed by Lupatini et al. [71] or Moreno-Espíndola et al. [73]. Hartman et al. [72] observed that differences in the bacterial community between organically and conventionally managed soils under integrated fertilization were smaller. Contrary to the bacterial community, the fungal microbial composition did change significantly among the three cropping systems. The relative abundance of Ascomycota and Mortierellomycota markedly increased in response to the fertilizer treatments. Ascomycota and Mortierellomycota rapidly metabolize rhizodeposited organic matter in rhizosphere soil, so their abundances are stimulated by nutrient substance [74].

Hence, fertilizer management may result in suitable circumstances for phyla that obtain sufficient levels of C, N, and P from the top soil [75]. Basidiomycota and Glomeromycota had the greatest dissimilarity in Org_M compared with Conv. Previous studies showed that Basidiomycota was increased in soils with manure as it provides an appropriate environment for Basidiomycota [76]. However, chemical fertilizer can cause the loss of that environment and can hinder the development of this phyla [76]. A high abundance of Glomeromycota in Org_M can form arbuscular mycorrhiza with plants and absorb nutrients directly, particularly P uptake through the plant roots [77], promoting plant growth, and enhance plant resistance to various pathogens [78], so it could be possible that its high abundance is related to the lower abundance of both pathogens, *Fusarium oxysporum* and *Alternaria* spp., and the nearly non-existent abundance of Olpidymycota (*Olpidium*). Contrarily, Olpidymycota was more abundant in Org_C, in which no Glomeromycota was found. However, the possible infection of plants by some pathogens species of Olpidymycota (*Olpidium*) in subsequent crops could be determined to cause disease on some of them, along with certain environmental factors [79]. Furthermore, Carini et al. [80] discovered that up to 50% of the microbial nucleic acid sequences in environmental samples could correspond to dead and inactive biomass. So, quantification by RNA would provide a more meaningful assessment of cellular viability and plant infection [79].

The presence of several differentially taxa among cropping systems provides information on soil microbiota responses to different agricultural management practices [81]. According to the LEfSe analysis, the long-term application of pesticides in the Conv system favored the greater existence of microorganisms associated with the pesticides, such as *Nesterenkonia*, *Galbibacter*, *Gramella*, *Linnembacter*, *Pseudoalteromonas*, *Pantoe*, and *Sporobolomyces*. Agrochemicals have the potential to inhibit or eliminate certain groups of microbes and select members adapted to or able to grow under conventional farming practices [82]. *Pantoea* comprises many versatile species with different functions, like the degradation of herbicides and other toxic compounds [83]. Some *Pseudoalteromonas* strains can produce bioactive compounds [84], and Sporobolomyces is a yeast capable of pesticide degradation [60]. Under Org_C, *Turicibacter*, *Dinghiubacter*, *Planifilium*, *Roseibacillus*, *Piscibacillus*, and *Candida* were more abundant and were positively correlated with ammonium, indicating that the presence of ammonium could favor the presence of these genera. Previous studies also reported that *Turicibacter* are abundant in soil with manure application and a high total organic carbon [85]. Regarding to *Candida*, it is important to point out that it is a saprotrophic fungi that sometimes come from compost, and may cause an opportunistic hazard to human health [86]. Org_M-related genera were *Aciditerrimonas*, *Isopericola*, *Nibrabacter*, *Haliangium*, *Terrimicrobium*, *Funnelliformis*, *Wallenia*, *Melanooleuca*, and *Filobasidium*. *Aciditerrimonas* is related to ferrous-ferric redox [87], and showed a correlation with TOC and NH₄⁺. *Haliangium* is a genus in which some species are producers of haliangicins, known as antifungal compounds [88], or *Wallemiales*, which have been observed in organic cultivations in other studies [7], and have shown a strong negative correlation with *Alternaria* spp and *Fusarium*.
Oxysporum. This suggests that an increase in this genus could favor a decrease in some phytopathogens [89], as we observed in this cropping system.

5. Practical Implications of This Study

In this study, the impacts of two types of long-term organic farming and one conventional farming system on the soil chemical properties and microbial communities are highlighted. This study highly recommends organic farming, principally the one where compost was used as an organic amendment (Org_C), where no differences were observed on crop yield with the other systems and it had a stable system with a high total organic carbon and nutrients, as well as a change in the bacterial and fungal communities. In addition to the environmental and human benefits of this type of farming system, further studies should further examine the possible mechanisms behind microbial community changes related to specific biogeochemical cycles, as well as the functional approach and inter-connection between microbial communities.

6. Conclusions

This work showed that properly long-term organic systems can contribute to maintaining high-yielding and stable crops when compared to the conventional system. In particular, it is noticeable that Org_C showed a change in bacterial community associated with an improvement in the soil carbon and nutrient content that was not found with Org_M. Analyses showed that fungal communities were more sensitive to cropping systems than bacteria because of the changes on a phylum level, while bacteria changes were more apparent on a specific taxonomy level. In addition, the LEfSe analysis revealed different microorganisms associated with each of the studied cropping systems. In particular, for the Conv system, the analysis demonstrated the presence of microorganisms associated with pesticide, while Org_C and Org_M microorganisms were associated with organic matter decomposition with different putative functions that, in some cases, could reduce soil borne pathogens.

Supplementary Materials: The following are available online at www.mdpi.com/2077-0472/11/5/445/s1, Figure S1: Rarefaction curves for the bacterial and fungal soil sequences, Table S1: Characteristics of organic amedments, Table S2: Diversity index for the bacterial and fungal community, Table S3: Relative abundance of most abundant bacterial phyla, Table S4: Relative abundance of most abundant fungal phyla, Table S5: NMDS results between bacterial community and the significant properties, Table S6: NMDS results between fungal community and the significant properties.

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