

Spatial heterogeneity of soil quality within a Mediterranean alley cropping agroforestry system: Comparison with a monocropping system

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Esther Guillot, Isabelle Bertrand, Cornelia Rumpel, Claudia Gomez, Didier Arnal, et al.. Spatial heterogeneity of soil quality within a Mediterranean alley cropping agroforestry system: Comparison with a monocropping system. European Journal of Soil Biology, 2021, 105, pp.103330. 10.1016/j.ejsobi.2021.103330. hal-03251274

HAL Id: hal-03251274 https://hal.inrae.fr/hal-03251274

Submitted on 13 Jun2023

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1	Spatial heterogeneity of soil quality within a Mediterranean alley cropping
2	agroforestry system: comparison with a monocropping system
3	
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19	Highlights
20	- Temperate alley cropping systems induce spatial heterogeneity in soil quality
21	- Soil quality is improved until 2 m beyond the tree row in the interrow
22	- 21 years of agroforestry increased the soil quality by 20% compared with monocrop

- 23 Abstract
- 24

25 Alley cropping agroforestry systems are complex agroecosystems highlighted for their 26 positive effects on soil quality. However, the potential spatial heterogeneity of soil quality 27 created by tree rows at the plot scale has seldom been studied. The aim of this study was to 28 evaluate soil quality at the plot scale, under tree rows and along transects perpendicular to the 29 tree row and to compare alley cropping systems with monocropping systems. This study was 30 performed on an alley cropping system that combined hybrid walnut trees (21 years old) and 31 peas. Topsoil was sampled at tree rows between 1 and 2, 2 and 4 and 4 and 6.5 m from the 32 tree row in the alley cropping system, as well as in a neighbouring monocropping plot. 33 Physical, chemical and microbiological indicators of soil quality were measured. Tree row 34 implantation induced spatial heterogeneity in the chemical indicators, microbial biomass, activities and community structure at the alley cropping plot scale. Alley cropping not only 35 36 improved microbiological soil quality indicators within the tree rows but also in the interrows 37 when compared to a monocropping system. These indicators were then integrated into one 38 soil quality index (SQI) built through a statistical approach. The soil quality index was 39 calculated for the monocropping plot and for each position within the alley cropping plot. 40 After 21 years of agroforestry practice, tree rows and permanent grass cover improved the 41 SOI until 2 m in the interrow. Weighted SOIs were calculated relative to the surface area of 42 each location for the entire alley cropping plot (i.e., tree row + interrow positions) and for the 43 entire alley cropping interrow (i.e., removing the tree row surface area). The weighted SQI of 44 the entire alley cropping plot significantly increased compared with that of the monocropping 45 plot.

47 Keywords: Alley cropping system, spatial heterogeneity, soil quality index, microbial
48 activities, soil organic carbon

49 **1. Introduction**

50

Soil quality can be defined as "the continued capacity of the soil to function within ecological and land-use boundaries, to sustain productivity, to promote the quality of air and water, and to maintain plant, animal and human health" [1]. Soil quality indicators are parameters that are sensitive enough to be modified by land use or management practices [2], [3] and should provide key information concerning the composition, structure and function of the soil [4]. In addition, a combination of physical, chemical and biological indicators must be used to correctly interpret soil quality [5], [6].

58 Physical indicators are related to the "inherent quality", which is influenced by soil 59 age and past climates [7], while most chemical and all biological indicators can be associated 60 with the "dynamic quality", which is highly sensitive to land use and soil management [8]. 61 Inherent and dynamic soil quality indicators refer to changes over long and medium to short 62 terms, respectively [9]. Among physical soil quality indicators, water storage, bulk density, 63 aggregate stability and texture are the most frequently measured [2]. Chemical indicators can 64 be evaluated to characterize soil fertility and nutrient availability (e.g., soil organic carbon, 65 total N, Olsen P, CEC, pH, exchangeable ions, and N mineralization rate) [10]. Biological 66 indicators of soil quality are necessary to connect abiotic soil properties to soil functioning 67 [2], [11]. Soil microbial biomass, abundance, diversity and activity play key roles as the main 68 drivers of soil organic matter decomposition and nutrient cycling [12], [13]. Soil 69 microorganisms can be studied as soil quality indicators through their biomass (e.g., microbial 70 biomass C), activities (e.g., enzyme activities and substrate-induced respiration) and 71 community structure (e.g., PLFA and sequencing) [14]. A minimum dataset of indicators has

to be selected either through expert opinion [15] or through statistical approaches [5], [16] to assess soil quality, avoiding collinearity between indicators. Both methods allow the construction of a soil quality index (SQI), which accounts for a more or less comprehensive set of indicators [10], [17]. The use of aggregated SQIs is currently strongly developed and can improve the comparisons of land use or management effects.

77 Land management is recognized as one of the main drivers affecting soil quality 78 improvement or degradation [18]. Temperate agroforestry systems comprising a combination 79 of trees and crops are examples of agroecosystems that provide a sustainable alternative to 80 conventional cropping systems, in which plant diversity is low; hereafter, these conventional 81 cropping systems are called monocropping systems. A recent review of European agroforestry 82 practices confirmed their positive effect on ecosystem services; however, this effect was context-dependent [19]. One of the most important ecosystem services promoted by 83 84 agroforestry systems is their potential capacity for carbon (C) storage, both temporarily in tree 85 biomass through photosynthesis [20] and in the long term in soil organic matter [21]. High 86 inputs of above- and belowground litter in the tree row in alley cropping can provide higher 87 amounts of C and nutrient resources and thereby modify soil microbial communities [22].

88 Even though a global positive effect of agroforestry on soil quality has been 89 recognized [23], it should be noted that only a small proportion of studies have considered the 90 spatial heterogeneity induced by tree rows. As highlighted by Cardinael et al. [24], future 91 studies should integrate more spatial dynamics and investigate lateral spatial heterogeneities 92 instead of concentrating on only the tree row. The few publications that have considered the 93 potential spatial heterogeneity of soil quality have not allowed us to claim that agroforestry 94 systems significantly improve soil quality at the plot scale or within the interrow. For 95 instance, [25] did not detect differences between tree rows and interrows in microbial biomass 96 C (MBC) and basal respiration or [26] in bacterial diversity or [27] in enzyme activities. Mungai et al. [28] found significantly more enzyme activities in tree rows than in interrows
but no difference within cultivated interrows. The spatial extent to which tree rows and their
understorey vegetation strips affect soil quality should be investigated.

100 In the present work, we evaluated the effect of an alley cropping agroforestry system 101 on the physical, chemical and biological components of soil quality compared to a 102 neighbouring, monocropping system managed in the same way as the interrows in the alley 103 cropping system. We first hypothesized that spatial gradients of soil quality occur between the 104 tree row and middle of the interrow, with the greatest values of soil quality indicators on and 105 close to the tree row and decreasing values with increasing distance from the tree row. 106 Second, we hypothesized that the weighted soil quality of alley cropping, integrating its 107 spatial heterogeneity throughout the plot, is higher than that of the neighbouring 108 monocropping plot.

109

110 **2. Materials and methods**

111 2.1. Site description

112 The experimental site of Restinclières is located at Prades-le-Lez, 15 km north of 113 Montpellier, southern France (43°42'15 N, 3°51'41 E). This location has a subhumid 114 Mediterranean climate, with an average temperature of 15.4 °C and average annual rainfall of 115 658 mm between 2000 and 2016 (Fréjorgues weather station). The soil is a deep Fluvisol 116 (WRB, 2007) with a silty-clay texture. In February 1995, hybrid walnut trees (Juglans regia × 117 nigra cv. NG23) were planted 4 m apart along east-west rows, with an interrow distance of 13 118 m. The initial planting density was 208 trees ha⁻¹, but the plot was thinned in 2004 down to a 119 present density of 110 trees ha⁻¹. Adult walnut trees have not been pruned since 2014. The 120 tree rows (2 m wide) were covered with spontaneous herbaceous vegetation, as they have not 121 been ploughed or treated with herbicides since 1995. Each interrow was ploughed to a 20 cm

122 depth every year before the winter crop was sown and was fertilized at an approximate rate of 150 kg N ha⁻¹ yr⁻¹ (as NH₄NO₃). Most of the time since 2000, durum wheat has been the 123 124 major crop in rotation, but pea has been introduced every 4 years since 2010, and barley has 125 been introduced since 2015 [29]. Over the 2014-2016 period, durum wheat (Triticum 126 turgidum durum cv. Claudio)-barley (Hordeum vulgare cv. Augusta)-pea (Pisum sativum cv. 127 Igloo) rotation was practised on this 4.6-ha alley cropping plot and on the adjacent 1.4-ha 128 monocropping plot, which has been managed in the same way as the cultivated interrows in 129 the alley cropping since the experimental site was established. All crops were sown between 130 the end of October and the beginning of December and harvested by the end of June/early 131 July.

132

133 2.2. Soil sampling

134 Soil sampling was conducted in April 2016, when pea was growing in the 135 monocropping plot and in the interrows of the alley cropping plot. At this stage, the walnut 136 trees had no leaves yet. In the 1.4 ha monocropping plot, five sampling areas of 1 m^2 each, at least 15 m apart, were identified for soil sampling. In each 1 m² surface area, one composite 137 138 soil sample, based on 4 soil cores (8 cm in diameter) from each angle of the area, was 139 collected from the 0-15 cm layer of topsoil using a root auger. For the alley cropping plot, 140 five transects at least 40 m apart were sampled from the tree row to the middle of the interrow 141 on both the north and south sides of the tree row. Each north-south transect corresponded to 7 142 positions (35 soil samples in total) (Fig. A.1). Soils were sampled in the tree row, with the 143 centre of the $2-m^2$ sampling area located on the axis of the tree row, 1 m from the nearest tree. 144 Other sampling areas were located in the interrow between 1 and 2 m from the middle of the 145 tree row, between 2 and 4 m and between 4 and 6.5 m. Soil samples were collected from the 146 0-15 cm layer of topsoil using a root auger at each angle of a centred subsurface (representing half of the total surface area of the location) (**Fig. A.1**). Areas sampled on the north side are indicated by "_N", and those on the south side are indicated by "s". Once collected, the fresh soil samples were passed through a 2-mm sieve, and subsamples were stored at 4 °C for microbial biomass and MicroResp® analyses, which were performed within two weeks after sampling and at -20 °C prior to freeze-drying for PLFA analyses.

152

153 2.3. Physical and chemical analyses

154 Physical and chemical indicators were measured on the north side of the trees 155 (corresponding to 4 positions per transect) and in the monocropping system, i.e., for a total of 156 25 samples. The moisture content was determined for all samples after drying for 48 h at 105 157 °C. Bulk density was determined by dividing the soil dry mass by the volume of the root 158 auger. Additional analyses were performed by the national routine soil testing lab of INRAE 159 at Arras (LAS, France). Soil texture was determined after CaCO₃ dissolution, and pH was 160 determined in a water extract. The cation exchange capacity (CEC) was determined using the 161 cobalt hexamine chloride method and analysed using inductively coupled plasma-atomic 162 emission spectrometry (ICP-AES) [30]. Soil organic C (SOC) and total N were determined by 163 dry combustion. The available P content was determined colorimetrically after extraction 164 using the Olsen method [31]. Roots were collected from each soil core (15 cm height \times 8 cm 165 diameter) and pooled into one composite sample (4 cores) for each area. They were separated 166 into three size classes as follows: 0-2 mm (fine roots), 2-5 mm (medium roots) and > 5 mm 167 (large roots).

168

169 2.3.1. Density and particle size fractionation

170 Density fractionation was performed on the north side of the trees and in the 171 monocropping system, i.e., for a total of 25 soil samples. The fractionation procedure was

derived from Roscoe et al. [32]. Briefly, 25 g of soil was shaken with 100 ml of a 1.6 Mg m⁻³ 172 dense sodium polytungstate solution (NaPT) and then centrifuged for 30 min at 6800 g. The 173 suspended material was termed the free light fraction (f-LF ≤ 1.6 Mg m⁻³) and was separated 174 from the supernatant by filtration using a 0.7 µm glass fibre filter. The material recovered on 175 176 the filter paper was washed with distilled water to remove residual NaPT. New NaPT (density 177 1.6 Mg m⁻³) was added to the remaining material. The new suspension was subjected to 178 ultrasonic dispersion (25 J ml) for 3 min. Afterwards, the suspension was centrifuged at 6800 179 g for 30 min, and the supernatant was filtered to recover the occluded light fraction (o-LF \leq 1.6 Mg m⁻³) (procedure see above). The remaining material was then separated into two 180 181 particle-size fractions by wet sieving: 50-2000 μ m (sand) and < 50 μ m (silt and clay). Before 182 C and N measurements, each fraction was decarbonated by acid fumigation using 12 M HCl 183 in a desiccator for 6 h [33]. After removing calcium carbonate, C and N contents in the total 184 soil, the f-LF o-LF and 50-2000 μ m and < 50 μ m fractions were analysed by dry combustion 185 with an elemental analyser (Elemental Analyser Vario Pyro Cube). This density fractionation 186 procedure allowed the separation of free soil organic matter (f-LF), soil organic matter associated with minerals (o-LF and 50-2000 µm) and physically protected soil organic matter 187 188 (< 50 µm).

189

190 2.3.2. Non-cellulosic neutral sugar determination

191 Rhamnose, fucose, mannose, galactose, glucose, ribose, arabinose and xylose were 192 quantified in the total soil, f-LF, o-LF and $< 50 \ \mu m$ fractions following the protocol of 193 Rumpel and Dignac [34] modified through the addition of 0.9 ml of 2 M EDTA after 194 hydrolysis with trifluoroacetic acid (TFA) [35]. Briefly, 500 mg for total soil, between 60 and 195 70 mg for f-LF and o-LF and 600 mg for the $< 50 \ \mu m$ size fraction were added to 10 ml of 196 TFA and hydrolysed at 105 °C for 4 h. After hydrolysis, myoinositol was added as an internal

197 standard. Samples were filtered through a glass fibre filter (0.7 µm), and TFA was eliminated 198 by evaporation before derivatization. Aldoses were reduced to their corresponding alditols 199 after the addition of 1 ml of NaBH₄ dissolved in dimethyl sulfoxide (DMSO). Acetylation 200 was performed by adding 2 ml of acetic anhydride and 2 ml of glacial acetic acid using 201 methylimidazole (2 ml) as the catalyst. The reaction was stopped after 10 min by adding 7 ml 202 ice-cold deionized water. The derivatised sugar monomers were extracted with 1 ml of 203 dichloromethane. The analyses were performed using a gas chromatograph (HP 6890 GC-204 FID) equipped with an SGE BPX-70 column (60 m \times 0.32 mm internal diameter, 0.25 μ m 205 film thickness). The gas chromatography oven temperature program was 200 °C to 250 °C at 8 °C min⁻¹ and isothermal at 250 °C for 15 min with helium as the carrier gas. 206

The sum of extracted monosaccharides from plant-derived hemicellulose and microbial products [36] is hereafter termed "sugars". The proportion of microorganismderived sugars in relation to plant-derived sugars can be estimated from the ratio of hexose:pentose sugars: (galactose + mannose):(arabinose + xylose), hereafter called GM:AX [37]. GM:AX ratios < 0.5 and > 2 are representative of carbohydrates predominantly derived from plants and microorganisms, respectively [37].

213

214 2.4. Soil microbiological indicators

The study plot had an east-west tree row orientation, and we wanted to test whether this orientation could modify soil microbiological parameters, which are sensitive to changes in climatic conditions at the local scale. Accordingly, we measured microbiological parameters on both the north and south sides of the tree row., i.e., for a total of 40 samples.

219 Microbial biomass C (MBC), N (MBN) and P (MBP) contents were quantified using 220 the chloroform fumigation-extraction technique [38]. Briefly, 10 g (for MBC and MBN) or 2 221 g (for MBP) of equivalent dry soil was exposed to chloroform vapor for 24 h and then 222 extracted with 40 ml of 0.025 M K₂SO₄ (for MBC and MBN) or with 40 ml 0.5 M NaHCO₃ 223 (for MBP), shaken for 45 min, centrifuged (10 min at 2683 g) and then filtered through 0.22-224 µm PTFE filters. Soil organic C and N concentrations in the extracts were measured by a 225 TOC/TON analyser (OI-Analytical, Aurora 1030, College Station, USA). Inorganic P 226 concentrations in the extracts were quantified colorimetrically using the malachite green 227 method [39]. Microbial biomass C, N, and P (MBC, MBN and MBP) contents were 228 calculated from the difference between the chloroform-fumigated and non-fumigated samples. 229 We applied a conversion factor of 0.45 for MBC [40], 0.54 for MBN [41] and a conversion factor of 0.4 for MBP [40]. Data are expressed in mg C, N or P kg soil⁻¹. The total C of 230 231 nonfumigated samples was used to represent dissolved organic carbon (DOC).

232 Basal and substrate-induced respiration was measured using the MicroRespTM 233 approach [42]. The soil water holding capacity (WHC) was determined using Richard's 234 membrane press at pF 2.5 [43]. Sieved soils were adjusted to 40% of their WHC and were 235 preincubated for 7 days at a temperature of 23 °C \pm 2 °C in the dark following the protocol of 236 Bérard et al. [44]. We tested three different soluble C substrates: glucose, trehalose and 237 alanine [44]. In a 96-deep-well microplate, 350 mg of soil sample was distributed, and then 25 μ l of each substrate was dispensed in each of the wells at 10 μ g substrate mg⁻¹ soil. The 96-238 239 deep-well microplate was then sealed with a CO₂-trap gel and incubated for 6 h in dark 240 conditions at 23 ± 2 °C. The optical density of each well of the CO₂-trap gel was measured on 241 a fluorometric microplate reader (Victor 3, Perkin Elmer) at 570 nm.

The microbial community structure was assessed using the phospholipid fatty acid (PLFA) method described by [45]. All analyses of PLFAs were performed by Microbial iD, Inc. (Newark, USA). Briefly, lipids were extracted from 5 g of freeze-dried soil by using a modified Bligh-Dyer extraction with 19 ml of extractant. Lipids were separated on a solidphase extraction column, and phospholipids were eluted with 5 ml of methanol. After 247 evaporation, phospholipids were transesterified to fatty acid methyl esters, extracted in 4 ml of 248 hexan, evaporated again and then analysed using a gas chromatograph (Agilent 6890 249 Technologies, Wilmington, USA) [45]. PLFA peaks were identified using the MIDI PLFAD1 250 calibration mix and naming table (MIDI, Inc., Newark, USA). Individual PLFA markers were 251 used to determine the number of peaks identified per position (tree row, $1-2 \text{ m}_{N/S}$, $2-4 \text{ m}_{N/S}$, 252 4-6.5 m_{N/S} for alley cropping and monocropping). Gram-positive (GP) bacteria were 253 identified by 13:0 iso, 14:0 iso, 15:1 iso w6c, 15:1 anteiso w9c, 15:0 iso, 15:0 anteiso, 16:0 254 iso, 16:0 anteiso, 17:1 iso w9c, 17:1 anteiso w9c, 17:1 anteiso w7c, 17:0 iso, 17:0 anteiso, 255 18:0 iso, and 20:0 iso; gram-negative (GN) bacteria were identified by 13:f1 w5c, 13:1 w4c, 256 14:1 w5c, 15:1 w6c, 16:1 w9c, 16:1 w7c, 17:1 w8c, 17:0 cyclo w7c, 18:1 w7c, 18:1 w5c, 19:1 257 w8c, 19:0 cyclo w7c, 20:1 w9c, 20:1 w6c, 21:1 w8c, 21:1 w3c and 22:1 w3c (reference table 258 from MIDI, Inc., Newark, DE). The peaks associated with 12:0, 14:0, 15:0, 16:0, 17:0, 18:0, 259 20:0, 22:0, 23:0 and 24:0 were considered general indicators [46], and they were not 260 integrated in analyses and were classified as "non-identified peaks" in Table A.1. The GP:GN 261 ratio was calculated based on the previously described peaks. A relative measure of the 262 fungi/bacteria (F:B) ratio was calculated by dividing the fungal PLFA marker: 18:2 w6c, 18:1 263 w9c [46], [47], [48] and 16:1 w5c, the last one representing the arbuscular mycorrhizal fungi 264 (AMF) by the sum of GN and GP bacteria. Actinobacteria were identified by 16:0 10-methyl, 265 17:1 w7c 10-methyl, 17:0 10-methyl, 18:1 w7c 10-methyl, 18:0 10-methyl and 19:1 w7c 10-266 methyl. Other unknown peaks were associated with the "others" category and were 267 considered only in the calculation of the relative abundance of each PLFA group. All PLFA 268 markers are listed in **Table A.1**.

269

270 2.5. Calculation of the soil quality index (SQI)

271 We followed the method proposed by Obriot et al. [10] to build an aggregated index of 272 soil quality. This method is based on 4 steps and starts with a dataset from different soil 273 variables and leads to a unitless single score. The first step (step 1, [10]) consisted of the 274 inventory of soil parameters. The second step (step 2, [10]) aimed to produce a minimum 275 dataset (MDS) for each category. Only those parameters that significantly discriminated the 276 various locations within the agroforestry plot (mixed model analysis, P < 0.05) were considered relevant [10]. Then, a Pearson correlation analysis was performed to identify 277 278 correlated parameters (> 0.8) and to avoid redundant information in the aggregated index. For 279 the third step (step 3, [10]), each indicator was normalized between 0 and 1 using the "more is 280 better", "less is better" or "optimum" response curves. For each parameter, the type of 281 response curve was determined according to the literature [10], [49], [50]. For all parameters, 282 we used the "more is better" response curves except for bulk density and Olsen P, for which the "optimum" response curve was chosen [51], [52]. Selected indicators from the MDS were 283 284 then computed in a principal component analysis (PCA), and the contribution of each 285 indicator to the dimensions of the PCA was calculated. Finally, the SQI was calculated 286 according to Equation 1 and Equation 2 (step 4, [10]), with a weighting of the transformed 287 variables using the PCA eigenvectors and the percentages of total variability explained by 288 each principal component:

289

- 290 <u>Equation 1</u>: $W_i = \sum_{j=1}^{p} \lambda_j x f_j$
- 291
- 292

Equation 2: SQI =
$$\sum_{i=1}^{n}$$
 Si x Wi

293

where fj = relative percentage of total variability attributed to each principal component, λj = sum of squared coordinates on each eigenvector, Si = normalized indicator scores and Wi = weighted factors. One SQI was calculated for each position within the alley cropping plot, and one for the entire alley cropping (hereafter called weighted alley cropping), weighting each surface area on the north side for the SQI. Another SQI was calculated for the alley cropping interrow excluding the tree row area in the weighted calculation (hereafter called weighted interrow). The last SQI was calculated for the monocropping system.

- 301
- 302
- 303 2.6. Statistical analyses

Mixed models with distance to the tree row as a fixed effect and transect number as a random effect were created for all soil parameters. Normal distribution of residuals and homogeneity of variance were tested by using the Shapiro and Bartlett tests. When necessary, data were Box-Cox transformed (MASS R package). If significant, a Tukey post hoc test was used for pairwise multiple comparisons (P < 0.05). Differences between monocropping and each position from the spatial gradient were tested by a one sample t-test (P < 0.05) considering the monocropping system as a reference.

311 A Pearson correlation matrix was calculated to assess relationships between 312 environmental parameters and the abundance of soil microbial functional groups. Redundancy 313 analysis (RDA) was performed to assess the relationship between the soil microbial 314 community composition, i.e., relative abundance of different microbial groups and 315 environmental parameters. Most discriminating variables were selected using a forward 316 procedure, and significance was also tested (100,000 permutations) (vegan R package). All 317 statistical analyses were performed with R software v.3.2.3 (R development Core Team, 318 2015).

319

320 **3. Results**

321 *3.1. Physical and chemical soil quality*

The soil was calcareous with more than 50% calcium carbonate, which explained the alkaline pH (**Table 1**). Levels of Olsen P were systematically low regardless of the position in the alley cropping or monocropping plots. The tree row showed significantly higher values than the interrow positions for the following variables: soil moisture, SOC, DOC, N_{tot}, mineral N, Olsen P, CEC and exchangeable cations.

The largest C content was systematically found in the smallest fraction, i.e., $< 50 \ \mu m$ fraction, contributing to more than 50% of the total soil C (**Table 2**). The highest C content was found in the tree rows regardless of the fraction, e.g., within the 50-2000 μm fraction, 3.6 g C kg⁻¹ soil was measured in the tree rows versus 2.0 - 2.3 g C kg⁻¹ soil within the interrows and 1.5 g C kg⁻¹ soil in the monocropping plots. The C:N ratios of f-LF and o-LF were significantly lower under the tree row (18.7 and 16.5, respectively) than under the middle of the interrow (22.8 and 18.7, respectively) (**Table A.2**).

The highest sugar-C concentration was also found in the < 50 μ m fraction (1.87 - 2.36 mg sugar C g⁻¹ soil) and did not significantly differ among positions in the alley cropping plot or compared with the monocropping plot (**Table 3**). The sugar-C concentration in f-LF under the tree row was significantly (two-fold) higher than that in the middle of the interrow or in the monocropping plot. In the same fraction, the proportion of microorganism-derived sugars in relation to plant-derived sugars (GM:AX ratio) decreased from 0.85 in the tree row down to 0.67 in 4-6.5 m_N and was significantly higher than that in the monocropping plot (0.62).

In the tree row, the total root biomass was more than 2-fold higher than anywhere within the interrow or compared with the monocropping plot (**Fig. A.2**). Significant differences were noticeable for fine- and medium-class roots.

We observed a soil MBC close to 2-fold higher in the tree row (427 mg C kg⁻¹) than in the middle of the interrow (219 mg C kg⁻¹ soil) and in the monocropping plot (199 mg C kg⁻¹ soil) (**Fig. 1**). The same difference was observed for MBN and MBP. Compared with monocropping, we systematically found a significant increase in MBC, MBN and MBP in the tree row and in 1-2 m_{N/S}.

The MB-C:N ratio significantly increased with increasing distance from the tree row, from 6.0 in the tree row to 7.4 in 4-6.5 m_N (**Table A.3**). The MB-C:P ratio was significantly lower under the tree row (24.5) than in 4-6.5 m_N (41.4). Values of MB-C:N and values of MB-C:P were similar in the monocropping plot and close to those in the tree rows.

Basal respiration was significantly increased only in the 1-2 m_s position compared with all others (**Table 4**). Compared with monocropping, glucose-induced respiration was significantly enhanced at all positions in the alley cropping plot except in 4-6.5 m_N. Alanineand trehalose-induced respiration was significantly higher in the tree row and close to it in the alley cropping plot than in the monocropping plot. Alanine-induced respiration was more than 2-fold higher under the tree row than in the middle of the interrow. The metabolic quotient (qCO₂) was similar across all positions (*P* > 0.05, **Table 4**).

362 The PLFA diversity, i.e., the number of detected peaks, decreased with increasing 363 distance from the tree row (Fig. A.3). Actinobacteria abundance was significantly highest in the tree row (2.58 μ g PLFAs g⁻¹ soil) and similar for the interrow positions and for the 364 monocropping plot (Fig. 2a). For GP and GN bacteria, AMF and other fungi, soils under the 365 366 tree row and in 1-2 m_s showed significantly higher abundance than those in the middle of the 367 interrow or in the monocropping plot (Fig. 2a, b). The F:B ratio did not show significant 368 differences between positions within the alley cropping plot or compared with the monocropping plot: all values were approximately 0.30 ± 0.1 (data not shown). The GP:GN 369

370 ratio was significantly lower in and close to the tree row compared with the interrow or
371 monocropping plot (Fig. 2a).

372

373 *3.3 Soil quality index*

Eight indicators, i.e., bulk density, SOC, N_{tot}, CEC, Olsen P, basal respiration, and glucose- and alanine-induced respiration ultimately contributed to the construction of the SQI (**Fig. 3**).

377 We observed a gradual decrease in the SQI with increasing distance from the tree row (0.76 to 378 0.53). The monocropping plot had an SQI similar (0.52) to that found in the middle of the 379 interrow. The contribution of each individual indicator to SQI construction was approximately 380 12%, except Olsen P, which contributed less than 3%. Individual contributions varied 381 according to the position in the alley cropping plot following a same spatial pattern that was 382 similar to the SQI. The weighted SQI of the entire alley cropping plot (0.64) was significantly 383 20% higher than that of the monocropping plot (0.52), while an increase of 10% of the 384 weighted SQI for the alley cropping interrow (0.57) was not significantly different from that 385 of the monocropping system (P = 0.08) (Fig. 3).

386

387 *3.4. Relationships between environmental parameters and microbial community structure*

388 To understand the link between soil environmental parameters and soil microbial 389 community structure, we i) calculated a Pearson correlation matrix between different soil 390 characteristics and diversity, and total PLFA biomass and abundance of each microbial group 391 (**Table A.4**) and ii) performed RDA to identify the determinants of the relative abundance of 392 PLFA markers (**Fig. 4**).

The two first axes explained 45 and 10% of the total variation, respectively (**Fig. 4**). The first axis clearly distinguished the relative abundance of actinobacteria and other GP bacteria, which were positively correlated with silt content, and the relative abundance of
AMF, GN bacteria and fungi. The AMF and GN bacteria were positively correlated with
sugar content and GM:AX ratio in the f-FL fraction, and fungi were positively correlated with
sugar content in the o-LF fraction and negatively correlated with clay content.

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- 401
- 402 **4. Discussion**

403 4.1. Soil quality index in agroforestry and monocropping plots

404 In the present work, the selected indicators for the minimum dataset were in line with 405 expert opinion [2], [53], [54]. Bulk density and SOC content represent important weighted 406 indicators in soil quality assessment for their role in water infiltration, storage and supply, as 407 well as CEC, N and P contents for nutrient storage and supply [11]. Among the selected 408 biological indicators, basal respiration is considered a physiological trait of microbial 409 communities, and glucose-induced respiration is assumed to be proportional to active 410 microbial biomass [44]. In this study, we focused on the microbial component of soil 411 functioning, although we are aware that other biological indicators, such as those related to 412 meso- or macro-fauna, could have been helpful for soil quality assessment [9], [55].

Our aim was to aggregate information obtained using different parameters into one integrative soil quality index based on several soil functions: water, C and nutrient storage and cycling and sustainability of biological activities. In the present work, we initially had a large number of indicators (30 in total), and several methods for SQI calculation offered to us. Comparing three different SQI constructions, i.e., additive SQI, weighted additive SQI and statistically modelled SQI, [56] showed that the last one was more efficient in time and cost. The method for index calculation proposed by [10] allowed us to objectively select the most relevant indicators to assess changes in physical, chemical and biological soil quality in the studied field plots. Without this statistical approach, we could have overestimated the positive effect of tree rows on soil quality because of the co-variability of the measured indicators. However, this method considers as relevant for the first step of minimum dataset selection only parameters that significantly changed statistically according to the treatments, which could be considered a subjective choice.

This is the first time, to the our knowledge, that soil quality was assessed in temperate alley cropping through the use of an integrative and weighted SQI. We showed a strong positive effect from the 21-year-old trees and accompanying permanent grass cover in the tree rows of the studied alley cropping system on soil physicochemical indicators compared with the interrow and monocropping systems; this effect was restricted to the tree rows (**Table 1**). The increase in SOC content in the tree rows led to an increase in those soil properties that are usually related to it, such as CEC and exchangeable cation contents [10] (**Table 1**).

433 We showed that the SQI was significantly improved in an alley cropping system in 434 comparison with a monocropping system, considering the spatial heterogeneity induced by 435 these systems. This indicates here that the positive effect from the trees and herbaceous cover 436 on the weighted entire alley cropping SQI was significant beyond the tree rows. The tree 437 rows, which represent approximately 13% of the field plot surface area, yielded a significant 438 beneficial effect on SQI over approximately 20% compared with monocropping (Fig. 3). 439 These findings should be put in perspective, as the results depend on one site and one 440 sampling date. Additional sub-annual sampling would be necessary to reject a possible 441 seasonal effect. Even though not significant (P = 0.08), there was a 10% increase in the SQI 442 weighted interrow (i.e., 1-2 m + 2-4 m + 4-6.5 m) compared with the neighbouring 443 monocropping plot. Repeating these indicator measurements in time and increasing the444 number of replicates would allow us to see if this trend could be significant or not.

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446 *4.2.* What explains the higher SOC content in the alley cropping system?

447 In our study, the topsoil of tree rows showed SOC contents 20-30% higher than that of 448 the interrow and 50% greater than that of the monocropping plot (**Table 2**), which is in line 449 with former results at the same study site [57]. Within and close to tree rows, belowground 450 litter, as aboveground litter can also be enhanced, as tree roots have been shown to colonize 451 the topsoil at the same agroforestry site until 1.5 m in the interrow [58]. In addition, a 452 substantial portion of roots can originate from (i) herbaceous cover root systems, which can 453 colonize the topsoil close to the tree row, as shown by [59] in younger walnut tree alley 454 cropping, or ii) weed colonization with low dispersal abilities, originating from herbaceous 455 cover [60]. Perennial plants can indeed allocate up to 10-15% more C belowground than 456 crops [61], [62]. The absence of tillage in the tree rows likely favoured some accumulation of 457 both C and N [63].

The novelty of our study is that it provides greater insight into the pools of C found in the topsoil at various locations in the alley cropping system than in the monocropping system. We demonstrate that the additional C found in the topsoil comes from fractions with a higher C:N ratio (> 16) (**Table A.2**), which is more similar to that of plant debris than that of smaller size fractions with a C:N ratio < 10, i.e., close to the C:N ratio of soil microbial communities and soil organic matter [64], [65]. These coarser fractions can be more easily mineralized by microorganisms, suggesting a less stable C pool.

The increased GM:AX ratio in the f-LF fraction from the middle of the interrow to the tree row suggests that a higher proportion of microbial-derived carbohydrates was present in the tree row (**Table 3**). This may have induced higher and faster microbial turnover based on 468 more organic matter recycling through microorganisms and could indicate that the production 469 of organic compounds is easier to stabilize [66], [67]. The RDA confirmed that the increased 470 GM:AX ratio favours more fast-growing microorganisms, i.e., microbial communities with 471 higher GN bacteria proportions (**Fig. 4**).

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4.3. Spatial heterogeneity of microbial soil properties in the alley cropping system

474 Our work demonstrated that an agroforestry system such as the one we studied here 475 may considerably impact soil microbial communities through its biomass, activity and 476 structure, resulting in the creation of within-plot spatial heterogeneity.

477 Higher microbial biomass and activities in and close to the tree row may represent a 478 higher potential capacity to decompose organic matter and improve nutrient cycling. 479 MicroResp® appeared not to be the most appropriate method to evaluate the activity of 480 microorganisms in the carbonate-rich soil; however, as carbonate content was similar 481 everywhere, it allowed us to compare positions. Our qCO₂ results indicate a similar 482 physiological activity of the microbial biomass regardless of the position within the 483 agroforestry plot [68]. This result was surprising because of the different tillage management 484 practices between the tree row and interrow [69]. However, [70] showed in a meta-analysis 485 that similar qCO_2 values were observed for tilled and no-tilled plots in long-term (10 years) 486 plots. This might indicate that qCO₂ is not sensitive to soil tillage or that microorganisms from 487 the tree row are, at the time of our sampling, as efficient in using C as those in the interrow or 488 in the monocropping system.

In tree rows, the MBC/SOC ratio was significantly higher than that in the interrow or in the monocropping plot (i.e., 23 versus 17, data not shown). This indicates that the MBC increase in tree rows was not merely due to the increase in SOC but also to some properties of the soil organic matter (e.g., greater C availability) within tree rows that favour thedevelopment of an abundant soil microbial biomass [71].

494 Compared with the monocropping system, the microbial biomass and substrate-495 induced respiration were more significantly higher on the south side of the tree row than on 496 the north side in the alley cropping plot (Fig. 1, Table 4). This result underlines the potential 497 legacy effect of microclimate created in this alley cropping agroforestry system. Dufour et al. 498 [72] showed in the same plot that photosynthetically active radiation can be reduced by 38%499 at 3 m on the north side when the tree canopy is well developed (end of June). Our soil 500 sampling was performed in April before walnut tree budburst began. At this stage, they 501 showed that the reduction in incident light was negligible (10%) compared with that in June 502 (50%). This suggests that the slight difference observed between the north and south sides of 503 the tree rows was due to a legacy effect of the 21-year-old practice.

We found that trees and the herbaceous grass community (dominated by weeds, *Bromus* and *Torilis* species) in the tree rows increased PLFA diversity and biomass. These two parameters were highly correlated with pH (negatively), soil moisture, DOC and total N, as observed by [48] and [22]. In our case, an additional correlation with root biomass highlighted its potential key role in shaping microbial community dynamics (**Table A.4**).

509 In the present study, we observed that tree rows favour a higher proportion of GN 510 bacteria than GP bacteria, leading to a lower GP:GN ratio. Environments rich in C and 511 available N are known to promote soil GN bacteria [73], which have high affinity for recent 512 plant-derived C compounds [74]. The RDA also suggested a correlation between the sugar 513 content in f-LF and the GN bacterial proportion (Fig. 4). In contrast, GP bacteria have been 514 shown to be less affected by low C availability [73]. Actinobacteria and other GP bacteria are 515 known to be able to feed on more complex polymers, such as older organic matter [48]. This 516 could explain their higher proportion in the middle of the interrow and in the monocropping 517 system. Variation in the GP:GN ratio suggests that microorganisms could be more 518 copiotrophic in the tree row than in the interrow of the alley cropping system as the GP:GN 519 ratio increases with decreasing C availability in soil [75] (**Fig. 2a**).

Moreover, high-recalcitrant litter inputs, such as walnut leaves, i.e., litter known to be rich in tannins and lignin [76], could have favoured the fungal pathway within or close to the tree row, while we did not observe this. We hypothesize that the herbaceous cover under the tree row adding labile litter and rhizodeposits buffers the recalcitrant effect of walnut litter. These results raise the question of the importance of considering herbaceous cover in future alley cropping studies.

526

527 **5.** Conclusion

528 Through the design of our sampling strategy, we evaluated changes in soil quality 529 according to the distance from the tree rows in the interrow of an alley cropping system and 530 compared them with a neighbouring monocropping system. The use of an integrative soil 531 quality index related to water and nutrient storage and to the support of biological activity 532 simplified the comparison of the various positions within the alley cropping plot and with the 533 monocropping plot. After 21 years, soil quality was significantly improved in the tree rows 534 compared with the interrow positions and monocropping. The tree rows and its permanent 535 herbaceous cover enriched the soil in organic matter through an increased input of above- and 536 belowground litters and possibly through enhanced rhizodeposition. The SOI was 537 significantly greater under tree rows and beyond until a 2-m distance from the rows, which 538 was clear evidence of spatial heterogeneity within the cultivated interrows. For the first time, 539 we showed that by considering the SQI of the weighted entire plot of the alley cropping 540 system that soil quality was significantly improved compared with that of the monocropping 541 system.

543 Acknowledgments

545	This work was supported by "La Fondation de France" and the French National
546	Program EC2CO Biohefect, France. The authors want to thank Annette Bérard (INRAE) at
547	UMR EMMAH in Avignon and Josiane Abadie (INRAE) at UMR Eco&Sols in Montpellier
548	for their help during MicroResp [™] analyses, Delphine Mézière and Lydie Dufour (INRAE) at
549	UMR SYSTEM in Montpellier for their knowledge of the study site and their advice on the
550	sampling location and Paula Fernandes (CIRAD) at UR HORTSYS for her help with the
551	PLFA.
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Figure 1



Figure 2a



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Figure 2b



(b)

Figure 3



Figure 4



Parameter	Tree row	1-2 m _N	2-4 m _N	4-6.5 m _N	MC
Clay (< 2 µm) (g kg ⁻¹ soil)	178 ± 9 ª	173 ± 24 ª	186 ± 7 ª	184 ± 13 ª	177 ± 10
Silt (2-50 µm) (g kg ⁻¹ soil)	157 ± 24 ª *	165 ± 43 ª *	157 ± 23 ª *	158 ± 17 ^a *	208 ± 16
Sand (50-2000 µm) (g kg ⁻¹ soil)	109 ± 18 ^{a *}	101 ± 22 ª	105 ± 25 ª	103 ± 22 ª	76 ± 20
Bulk density (g cm ⁻³)	1.24 ± 0.1 ^a *	1.41 ± 0.1 ^a	1.23 ± 0.1 ^a *	1.30 ± 0.1 ^a	1.43 ± 0.1
CaCO3 (g kg ⁻¹ soil)	539 ± 11 ª	545 ± 22 ª	540 ± 26 ª	539 ± 18 ^a	527 ± 20
Moisture	14.3 ± 1.8 ° *	13.3 ± 1.1 ^{bc} •	11.8 ± 0.7 ^a	11.9 ± 0.9 ^b	11.4 ± 0.6
pH _{water}	8.40 ± 0.05 ^a *	8.52 ± 0.04 ^b *	8.55 ± 0.03 ^b	8.57 ± 0.05 ^b	8.58 ± 0.03
SOC (g kg ⁻¹ soil)	19.0 ± 3.5 ^b *	16.4 ± 4.5 ^{ab}	14.6 ± 3.0 ª	13.8 ± 2.6 ª	11.8 ± 1.0
DOC (mg kg ⁻¹ soil)	34.5 ± 5.6 ^b *	23.4 ± 2.8 ª	17.4 ± 4.9 ^a	18.1 ± 3.2 ª	20.3 ± 2.6
N tot (g kg ⁻¹ soil)	1.8 ± 0.2 ^{c*}	1.4 ± 0.1 ^b	1.3 ± 0.1 ^{ab}	1.2 ± 0.1 ^a	1.1 ± 0.1
C:N	10.9 ± 1.6 ^a	11.4 ± 2.0 ^a	11.2 ± 2.3 ª	11.5 ± 3.9 ª	10.2 ± 3.1
Mineral N (mg kg ⁻¹ soil)	31 ± 5 ^b *	16 ± 4 ^{a *}	11 ± 2 ª	11 ± 2 ª	11 ± 2
Olsen P (mg kg ⁻¹ soil)	8.3 ± 2.1 ^b	5.8 ± 1.2 ^a	5.2 ± 1.6 ª	5.6 ± 2.3 ^a	5.8 ± 1.3
CEC (cmol + kg-1 soil)	$14.7 \pm 1.9 {}^{b*}$	13.0 ± 1.6 ^a	12.5 ± 1.4 ^a	12.4 ± 1.3 ª	11.8 ± 0.4
Exch. Ca (cmol + kg-1 soil)	$14.7 \pm 1.6 b^{*}$	13.3 ± 1.5 ª	13.1 ± 1.1 ^a	12.8 ±1.1 ^a	12.5 ± 0.5
Exch. Mg (cmol ⁺ kg ⁻¹ soil)	0.73 ± 0.12 ^b *	0.47 ± 0.10 ^a *	0.39 ± 0.09 ^a *	0.38 ± 0.07 ^a *	0.29 ± 0.02
Exch. K (cmol + kg ⁻¹ soil)	0.51 ± 0.08 ^b *	0.42 ± 0.08 ^{ab *}	0.36 ± 0.08 ^a	0.35 ± 0.07 ^a	0.31 ± 0.02

Fraction	Tree row	1-2 m _N	2-4 m _N	4-6.5 m _N	MC
f-LF (g C kg ⁻¹ soil)	3.6 ± 1.3 ^b *	2.1 ± 0.9 ^a	1.5 ± 0.5 ^a	1.3 ± 0.5 ª	1.2 ± 0.3
o-LF (g C kg ⁻¹ soil)	$2.2 \pm 2.2 $ ^b *	1.7 ± 1.7 ^{ab *}	1.4 ± 1.4 ª	1.2 ± 1.2 ª	1.0 ± 1.0
50-2000 µm (g C kg ⁻¹ soil)	3.6 ± 0.9 ^b	2.3 ± 0.3 ^{ab}	2.0 ± 0.2 a	2.0 ± 0.1 ^a	1.5 ± 0.1
<50µm (g C kg ⁻¹ soil)	$12.0 \pm 2.1 $ ^{b*}	9.0 ± 1.8 ª	8.6 ± 1.9 ª	7.9 ± 1.8 ª	7.8 ± 1.0

Fraction	Tree row	1-2 m _N	2-4 m _N	4-6.5 m _N	MC
Total (mg sugar C g ⁻¹ soil)	4.4 ± 0.2 ^b	4.1 ± 0.1 ^a	4.0 ± 0.2^{a}	4.0 ± 0.2^{a}	4.0 ± 0.3
f-LF (mg sugar C g ⁻¹ soil)	0.57 ± 0.19 ^b *	0.35 ± 0.11 ª	0.32 ± 0.07 ^a	0.28 ± 0.04 ^a	0.29 ± 0.08
o-LF (mg sugar C g ⁻¹ soil)	0.38 ± 0.05 ^a *	0.31 ± 0.07 ^a *	0.29 ± 0.20 ª	0.22 ± 0.09 ^a	0.16 ± 0.06
$< 50 \mu m$ (mg sugar C g ⁻¹ soil)	2.26 ± 0.4 ^a	2.15 ± 0.2 ª	$1.87 \pm 0.7 a$	2.10 ± 0.2 ^a	2.36 ± 0.1
GM:AX Total	0.95 ± 0.06 ^a	0.96 ± 0.04 ª	1.00 ± 0.04 a	0.99 ± 0.04 ª	0.97 ± 0.05
GM:AX f-LF	0.85 ± 0.15 ^b *	0.73 ± 0.09 ^{ab}	0.71 ± 0.09 ab	0.67 ± 0.09 ª	0.62 ± 0.08
GM:AX o-LF	0.95 ± 0.03 ª	0.91 ± 0.07 ª	0.97 ± 0.04 ^{a *}	0.95 ± 0.04 ª	0.90 ± 0.03
$GM:AX < 50 \mu m$	1.05 ± 0.01 a	1.03 ± 0.02 a	1.04 ± 0.02 a	1.02 ± 0.02 a	1.03 ± 0.01

Substrate	Tree row	1-2m _N	1-2m _s	2-4m _N	2-4m _s	4-6.5m _N	4-6.5m _s	MC
Water	0.57 ± 0.09 ^{ab}	0.59 ± 0.20 ^{ab}	0.72 ± 0.25 ^b	0.45 ± 0.12 ^{ab}	0.38 ± 0.11 ª	0.36 ± 0.10 ª	0.35 ± 0.06 ª	0.43 ± 0.16
Glucose	2.65 ± 0.47 ^b *	2.12 ± 0.50 ^{ab} *	2.49 ± 0.50 ^b *	1.97 ± 0.41 ^{ab *}	1.98 ± 0.31 ^{ab *}	1.72 ± 0.41 ª	1.81 ± 0.30 ^{ab} *	1.30 ± 0.21
Trehalose	1.99 ± 0.57 ª *	1.80 ± 0.61 ^{a *}	1.98 ± 0.51 ^a *	1.39 ± 0.35 ª	1.44 ± 0.22 ^a *	1.17 ± 0.52 ª	1.11 ± 0.14 ª	1.05 ± 0.19
Alanine	1.50 ± 0.56 ° *	1.03 ± 0.61 abc	1.44 ± 0.50 bc *	0.95 ± 0.20 ^{abc}	0.99 ± 0.34 abc	0.62 ± 0.16 ª	0.72 ± 0.17 ab	0.73 ± 0.12
qCO ₂	0.22 ± 0.02 ^a	0.28 ± 0.05 ^a	0.29 ± 0.08 ª	0.24 ± 0.09 ª	0.19 ± 0.05 ª	0.21 ± 0.03 ª	0.20 ± 0.03 ª	0.33 ± 0.11