



Soil moisture and land use are major determinants of soil microbial community composition and biomass at a regional scale in northeastern China

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Abstract. Global environmental factors impact soil microbial communities and further affect organic matter decomposition, nutrient cycling and vegetation dynamic. However, little is known about the relative contributions of climate factors, soil properties, vegetation types, land management practices and spatial structure (which serves as a proxy for underlying effects of temperature and precipitation for spatial variation) on soil microbial community composition and biomass at large spatial scales. Here, we compared soil microbial communities using phospholipid fatty acid method across 7 land use types from 23 locations at a regional scale in northeastern China (850 × 50 km). The results showed that soil moisture and land use changes were most closely related to microbial community composition and biomass at the regional scale, while soil total C content and climate effects were weaker but still significant. Factors such as spatial structure, soil texture, nutrient availability and vegetation types were not important. Higher contributions of gram-positive bacteria were found in wetter soils, whereas higher contributions of gram-negative bacteria and fungi were observed in drier soils. The contributions of gram-negative bacteria and fungi were lower in heavily disturbed soils than historically disturbed and undisturbed soils. The lowest microbial biomass appeared in the wettest and driest soils. In conclusion, dominant climate and soil properties were not the most important drivers governing microbial community composition and biomass because of inclusion of irrigated and managed practices, and thus soil moisture and land use

appear to be primary determinants of microbial community composition and biomass at the regional scale in northeastern China.

1 Introduction

The soil microbial community plays important roles in regulating organic matter decomposition, nutrient cycling, soil structural formation and even plant interactions (Wardle et al., 2004; Harris et al., 2009). At the same time, it is subjected to the influences of environmental conditions, land use and spatial structure (Yang et al., 2013). Although there is a growing body of evidence indicating that climate, soil property, vegetation, spatial structure and land use are the most important determinants of the global and regional patterns in soil microbial communities (Kreft and Jetz, 2007; Nielsen et al., 2010; Zinger et al., 2011; Pasternak et al., 2013; Tsiknia et al., 2014), how to disentangle the contributions of multiple drivers to microbial community composition and biomass remains unclear.

Regional climate factors exert major influences on distributions of microbial communities by determining temperature and soil water availability along topographic gradients (Hackl et al., 2005; Carletti et al., 2009; Brockett et al., 2012). Drenovsky et al. (2010) and Brockett et al. (2012) found that soil water availability was an important determinant of microbial community composition, and fun-

gal:bacterial biomass ratios decreased with increased soil water saturation at regional scales. In different types of natural Austrian forests, Hackl et al. (2005) showed that mean annual temperature was the major factor influencing microbial community composition in zonal forest but that soil water availability was most closely correlated with microbial community in azonal forests.

Soil property has been found to strongly correlate with soil microbial community structure and abundance at regional scales. Previous studies have reported that soil texture, organic matter content, N availability and pH exhibited dominant effects on soil microbial community composition, while climatic effects were weaker but still significant at regional scales (Šantrúcková et al., 2003; Brockett et al., 2012; Yang et al., 2013; Tsiknia et al., 2014). For example, Tsiknia et al. (2014) identified soil total organic C, pH and geographic distance as the most important determinants of microbial community abundance at the watershed scale in Greece. Moreover, plant communities differing in species composition are likely to produce litter and vary in their chemical composition, which may subsequently influence soil microbial community composition (Zhang et al., 2005a; Eskelinen et al., 2009). As a biotic driver, plants may also exert great effects on soil microbial communities by controlling allocation of belowground photosynthates (Kaiser et al., 2011).

Spatial structure (which serves as a proxy for underlying effects of temperature and precipitation for spatial variation) influences the organization of community as a functional variable in a different way to which the background in which biological and environmental factors act on community and ecosystem (Borcard et al., 1992). Recent research has shown there to be strong autocorrelations between microbial groups, and geographic distance could explain a high proportion of microbial community variation (Tsiknia et al., 2014). However, Fierer and Jackson (2006) claimed that soils with similar environmental characteristics have similar bacterial communities regardless of geographic distance at continental scales. Using spatial trend surface analysis, Drenovsky et al. (2010) also found that spatial structure did not influence microbial community composition across three biogeographical provinces in California.

At regional scales, land use change is the major reason for spatial heterogeneity. It has been shown that land use changes lead to great variation in soil microbial community composition in diverse ecosystems (Drenovsky et al., 2010), though their impacts depend on many factors, including the original vegetation that is being replaced and associated land management practices such as tillage and fallow periods, as well as related water and nutrient applications such as irrigation and fertilization (Scanlon et al., 2007; Ma et al., 2013; Yang et al., 2013; Chen et al., 2014). In one study, Drenovsky et al. (2010) reported that distinct microbial communities were associated with land use types and disturbance at the regional scale in California. Tillage influences multi-

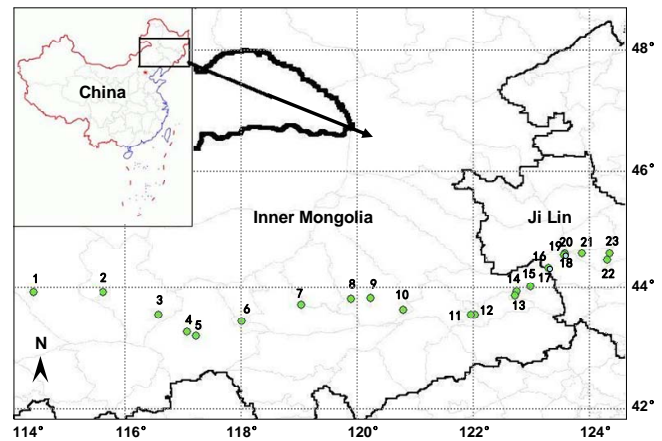


Figure 1. Sample locations (1–23; see Table 1) at a regional scale in northeastern China.

ple soil physical and chemical properties, disrupts soil fungal hyphae (Evans and Miller, 1990), and alters microbial community composition (Ingram et al., 2008; Drenovsky et al., 2010). Recently, changes in land use have occurred in temperate areas of northeastern China as a result of expansion of farmlands and grazed rangelands at the expense of natural habitats; however, little is known about soil microbial community composition to land use changes at regional scales.

In this study we compare microbial community composition and biomass from 23 locations across 7 land use types (i.e., rangeland, artificial grassland, grazed rangeland, farmland, returned cropland, woodland and rice field) at a regional scale in the Northeast China Transect (NECT). The NECT is identified as a midlatitude semiarid terrestrial transect and is sensitive to climate change and disturbance, and thus provides an ideal setting to investigate distribution patterns of soil microbial community. Our work specifically aimed at disentangling the contributions of climate, soil properties, vegetation types, spatial structure and land use on microbial community composition and biomass at the regional scale. We hypothesize that climate and soil properties are the primary drivers affecting soil microbial community composition and biomass, because the climatic gradient, especially precipitation, is one of the most notable features in this region (Wang et al., 2003).

2 Materials and methods

2.1 Study locations

The field study was conducted on a regional scale (43°12′–44°36′ N, 114°34′–124°18′ E) across Jilin province and Inner Mongolia (about 850 km from east to west, and 50 km from north to south) with 23 locations in the Northeast China Transect (NECT) (Table 1, Fig. 1). The NECT was identified as a core project of International Geosphere-Biosphere Pro-

Table 1. Sample locations (1–23), coordinates of the sample location, land use types, vegetation types and number of replicates (*n*).

Location	No.	Coordinates	Land use type	Vegetation type	<i>n</i>
Baogedawula	1	43°56' N, 114°34' E	Rangeland	Desert steppe	8
Dabuxiletu	2	43°55' N, 115°44' E	Rangeland	Desert steppe	8
	2		Grazed rangeland	Desert steppe	8
Aqiwula	3	43°33' N, 116°40' E	Rangeland	Steppe	10
	3		Woodland	Wood and shrub	8
Dalainuori	4	43°16' N, 117°09' E	Rangeland	Steppe	8
Sanyi	5	43°12' N, 117°18' E	Woodland	Wood and shrub	8
Xinchengzi	6	43°27' N, 118°04' E	Rangeland	Steppe	14
	6		Returned cropland	Alfalfa	8
Xinfuzhilu	7	43°43' N, 119°04' E	Grazed rangeland	Steppe (site 1)	4
	7			Steppe (site 2)	4
Tianshan	8	43°50' N, 119°55' E	Rangeland	Steppe	8
	8		Returned cropland	Almond	16
Tianshan	9	43°50' N, 120°15' E	Rangeland	Steppe	9
	9		Returned cropland	Almond	9
Shaogen	10	43°38' N, 120°47' E	Rangeland	Steppe (site 1)	8
	10			Steppe (site 2)	8
	10		Farmland	Corn	8
Molimiao	11	43°34' N, 121°55' E	Rangeland	Steppe (site 1)	8
	11			Steppe (site 2)	8
	11		Farmland	Corn	8
Yuxin	12	43°34' N, 121°59' E	Rice field	Rice	14
Baixingtu	13	43°52' N, 122°41' E	Woodland	Wood and shrub	8
Baolongshan	14	43°56' N, 122°42' E	Rangeland	Meadow (site 1)	7
	14			Meadow (site 2)	6
	14		Farmland	Corn	8
Jiamatu	15	44°01' N, 122°56' E	Rangeland	Meadow (site 1)	8
	15			Meadow (site 2)	8
	15		Farmland	Corn	8
	15			Red bean	7
Taipingchuan	16	44°21' N, 123°14' E	Rangeland	Meadow	9
	16		Rice field	Rice	9
Yaojingzinan	17	44°21' N, 123°14' E	Woodland	Wood and shrub (site 1)	11
	17		Woodland	Wood and shrub (site 2)	10
	17		Farmland	Peanut	8
Yaojingzi	18	44°34' N, 123°29' E	Rangeland	Meadow (site 1)	8
	18			Meadow (site 2)	7
	18		Farmland	Peanut	8
	18			Mung bean	8
	18			Corn	8
Yaojingzi	19	44°35' N, 123°30' E	Rangeland	Meadow	14
Yaojingzi	20	44°34' N, 123°31' E	Artificial grassland	Meadow (site 1)	7
	20			Meadow (site 2)	8
	20		Farmland	Corn	8
Wulanaodu	21	44°36' N, 123°48' E	Rangeland	Meadow (site 1)	8
	21			Meadow (site 2)	8
	21		Farmland	Corn	7
	21		Woodland	Wood and shrub	9
Chaganhua	22	44°35' N, 124°16' E	Rangeland	Meadow (site 1)	8
Wulantuga	22	44°28' N, 124°18' E	Rangeland	Meadow (site 2)	8
	23			Meadow	8
	23		Farmland	Corn	6
	23			Peanut	6
	23		Woodland	Wood and shrub	8
Total					451

gramme (IGBP), which represents an array of regional-scale gradients on all continents that vary in major environmental variables (Koch et al., 1995). This area has a continental monsoon climate, with large seasonal temperature and precipitation gradients. Long-term (1950–2000) mean annual temperature, precipitation and radiative dry index at this spatial scale range from approximately 1.3 to 6.8 °C, 237 to 472 mm and 0.91 to 1.44, respectively. The elevation gradients range from 140 to 1309 m (<http://www.worldclim.com>; Zhang et al., 1997; Appendix A1). Mean soil total C, N and C : N varied 3.3-fold, 2.4-fold and 2.7-fold, respectively, across the region. Overall, there were 7.4-fold and 2.8-fold differences in soil water content and water holding capacity, respectively, whereas soil origin and pH differed only slightly (Appendix A1).

Spatial climatic variability, especially precipitation, is one of the most notable features of the transect. Due to the large decrease in precipitation from the east (Jilin province) to the west (Inner Mongolia), vegetation vary gradually from moist meadows in the east to typical steppes and desert steppes in the west, with farmlands, returned croplands and woodlands spread evenly across the gradient (Wang et al., 2003, 2011; Appendix A1). All farmlands were irrigated only a few times (2–3 times) during the growing season, and rice field was flood-irrigated. The regions have remarkable variations in climate, land use types and vegetation types, which make it an ideal region for studying the primary factors driving soil microbial community composition and biomass. Detailed descriptions of land use types, vegetation types and soil properties can be found here in Table 1 and Appendix A1 as well as in Zhang et al. (1997) and Ni and Zhang (2000).

2.2 Soil samplings

Four hundred fifty-one soil samples from 23 locations including 7 land use types were collected along the NECT in 12–18 July 2012. Between 6 and 16 soil core samples were collected randomly per site (100 × 100 m) for determination of soil microbial communities (Table 1).

The samples were taken with a cylindrical soil sampler (5 cm inner diameter, 15 cm length) for the 0–15 cm layer, and then immediately preserved at 4 °C in a cooler for transport to the laboratory within 1 week of collection. The fresh samples were processed using a 2 mm sieve and manually cleaned of any visible plant tissues. Two subsamples of each sample were obtained: one was air-dried for routine soil analyses and the other was stored at –70 °C for phospholipid fatty acids analysis.

2.3 Soil microbial community analysis

Phospholipid fatty acids (PLFAs) were extracted and quantified from 8.0 g (dry weight equivalent) of soils using a procedure described by Bossio and Scow (1998). The separation and identification of extracted PLFAs were carried out

according to the standard protocol of the Sherlock Microbial Identification System V_{4.5} (MIDI) and a gas chromatograph (Agilent 6850, USA). “A : B ω C” represents the number of carbons in the compound: the number of double bonds in the carbon chain, followed by double-bond location from the methyl (ω) end of the molecule (Bossio and Scow, 1998). *Cis* and *trans* conformations are indicated by the suffixes c and t. The prefixes a and i indicate *anteiso* and *iso* branching. 10Me specifies a methyl group on the 10th carbon from the carboxyl end of the molecule, OH indicates a hydroxyl group, and cy indicates cyclopropane fatty acids. In addition, the fatty acids “sum” indicates that imperfect peak separation occurs and refers to two or more fatty acids that have the same retention time (Drenovsky et al., 2004).

Thirty-one fatty acids were included in the analyses. (1) Branched fatty acids indicative of gram-positive bacteria: a13:0, i14:0, i15:0, i16:0, i17:0 and a17:0; (2) monounsaturated fatty acids indicative of gram-negative bacteria: 16:1 ω 7c, 17:1 ω 8c, 18:1 ω 5c, 18:1 ω 9t, 17:0cy and 19:0cy (Frostegård et al., 1993, 1996); (3) saturated fatty acid (common in soil microorganisms): 14:0, 15:0, 16:0, 17:0, 18:0 and 20:0; (4) two fatty acids (18:2 ω 6c, 18:1 ω 9c) were chosen to represent the fungi (Frostegård et al., 2011); and (5) actinomycetes was represented by 10Me 17:0 fatty acid. The fatty acids 14:2 ω 6c and 14:1 ω 8c were found in only three samples and were thus excluded from the data set. The ratio of 17:0cy (17cy) to 16:1 ω 7c (precursor) was used as an indicator of physiological stress (Knivett and Cullen, 1965). The viable microbial biomass was calculated by summing concentration of all fatty acids detected in each soil samples (White et al., 1979). Total percentages of fatty acid identified for each microbial group were calculated to represent their relative contributions to the total microbial biomass. The fungal : bacterial fatty acid (gram-positive + gram-negative bacteria) ratio was also included in the data analysis (Frostegård et al., 1996).

2.4 Soil property analyses

Soil inorganic N (NH₄⁺-N + NO₃⁻-N) was extracted with 2 M KCl solution, and the extractant was determined using a flow injection autoanalyzer (FIAstar 5000, Denmark). Soil pH was measured at a soil : water ratio of 1 : 2.5 with a pH electrode (PHS 29, China). Soil total C and N content were measured using an elemental analyzer (Elementar analyser vario Max CN, Germany). Soil texture was determined by the optical size analyzer (Mastersizer 2000, England). Gravimetric soil water content was measured by oven-drying samples at 105 °C for 24 h. Soil water holding capacity was measured by means of the Wilcox method (Wilcox, 1962).

2.5 Statistical analyses

Unconstrained ordination (correspondence analysis: CA) was used to compare soil microbial communities among

samples ($n = 451$) using the Canoco for Windows 4.5 package (Ithaca, NY, USA). CA is an indirect gradient analysis method which can provide the basic overview of soil samples and maximize the correlation between fatty acids and samples (Lepš and Smilauer, 2003). Constrained ordination (canonical correspondence analysis: CCA) was used to represent the relationships among environmental factors (habitat, land management, spatial structure), sample patterns and fatty acids distributions (Lepš and Smilauer, 2003). Qualitative factors were coded for the program using a set of “dummy factors”. That is, if a sample has a particular value of the factor, then the corresponding dummy factor has the value 1.0 and the other dummy factors have a value of 0.0 for the same sample.

In order to separate the effects of environmental factors on microbial communities, the variation partitioning procedure with CCA was used in the analysis (Borcard et al., 1992). The environmental factors were divided into three groups: (1) habitat (mean annual temperature, mean annual precipitation, radiative dry index, elevation, soil texture, pH class, soil N availability, soil C and N content, soil C : N, water holding capacity), (2) land management (tillage, grazing, historical tillage, flooding), and (3) spatial structure (x , y , xy , x^2 , y^2 , x^2y , xy^2 , x^3 , y^3). The third group consisted of nine terms, in which latitudinal (x) and longitudinal (y) coordinates were used to calculate a cubic trend surface. Spatial trend surface analysis is one of the quantitative ecological methods used to study the relation between spatial structure and species abundance distribution in community (Legendre, 1990). The variation partitioning procedure decomposed the total variability into eight parts: individual effect of habitat (X_1), land management (X_2), spatial structure (X_3), combined effects of habitat and land management (X_4), combined effects of land management and spatial structure (X_5), combined effects of habitat and spatial structure (X_6), combined effects of the three groups of environmental factors (X_7), and residual variation (X_8). A complete explanation of these partitioning analyses can be found in Lepš and Smilauer (2003).

Stepwise multiple linear analyses were used to determine the relationships of soil microbial community composition and biomass or the contribution of each microbial group with environmental factors. Differences among the sites in soil microbial biomass and contribution of each microbial group were tested using one-way ANOVA. Data management and statistical analyses were performed using SPSS 17.0 software (SPSS, Chicago, IL, USA).

3 Results

3.1 Variation in soil microbial communities

The first axis of CA ordination explained 27.5 % of the variation in microbial community composition, mainly reflecting soil moisture gradients and land disturbance intensity

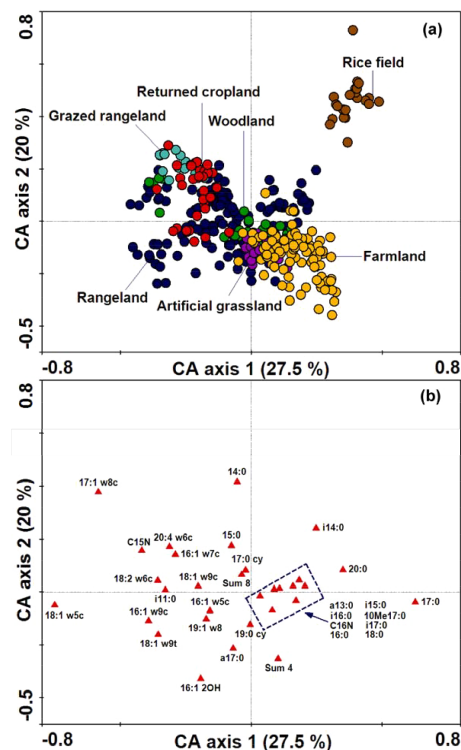


Figure 2. Ordination plots of correspondence analysis (CA) of all samples and fatty acids. **(a)** Ordination plot of 451 samples scores across 7 land use types (rangeland, artificial grassland, grazed rangeland, farmland, returned cropland, woodland and rice field). **(b)** Ordination plot of 31 fatty acids scores. The fatty acids scores are near the points for samples in which they occur with the highest relative contributions.

(Fig. 2a, b). Wetter soils (e.g., rice field, moisture rangeland) and heavily disturbed soils (e.g., farmland) with more branched fatty acids (gram-positive bacteria: $a13:0$, $i14:0$, $i15:0$, $i16:0$, $i17:0$) and saturated fatty acids ($14:0$, $15:0$, $16:0$, $17:0$, $18:0$, $20:0$) were positioned along the right side of the first axis. Drier soils and lightly and historically disturbed soils (e.g., dry rangeland, grazed rangeland, returned cropland) with more fungal ($18:2\omega6c$, $18:1\omega9c$) and monounsaturated fatty acids (gram-negative bacteria: $16:1\omega7c$, $16:1\omega9c$, $17:1\omega8c$, $18:1\omega5c$, $18:1\omega9t$) were plotted along the left side of the first axis.

The second axis of CA ordination described 20 % of the variation in the composition, mainly associated with management practices and spatial variation. In heavily disturbed habitats, the positions of flood-irrigated rice field and farmland were separated along the second axis (Fig. 2a).

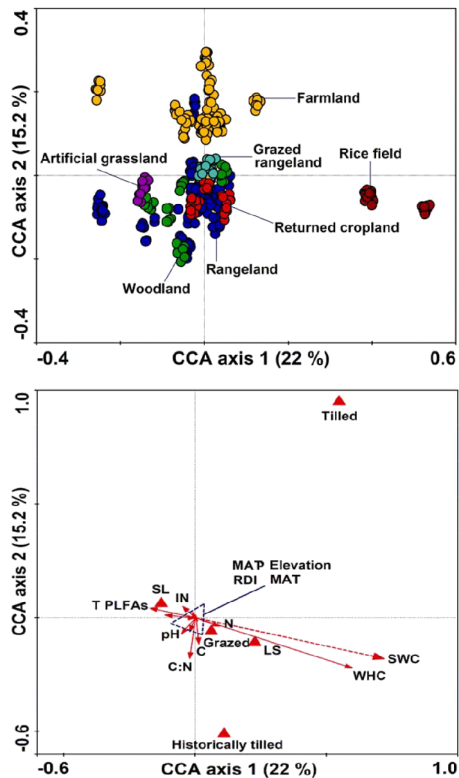


Figure 3. Ordination plots of canonical correspondence analysis (CCA) of all samples and environmental factors. (a) Ordination plot of 451 samples scores across 7 land use types. (b) Ordination plot of habitat and management factors scores in which spatial structure was run as a covariate. Mean annual temperature (MAT), mean annual precipitation (MAP), radiative dry index (RDI), elevation, soil water content (SWC, including natural precipitation and managed inputs), soil inorganic N (IN), soil total C and N (C, N), soil C : N, total (T) PLFAs, water holding capacity (WHC) and soil pH were quantitative environmental factors, and soil texture (loamy sand, LS; sandy loam, SL), land management practices (tilled, historically tilled, grazed) were qualitative (nominal) environmental factors. Quantitative factors were plotted as vectors, and qualitative factors were plotted as centroids.

3.2 Relationship between microbial communities and environmental factors

Soil microbial community composition across seven land use types at the regional scale was distinguished by environmental factors with the CCA ordination (Fig. 3a, b). The first axis explained 22% of the variation in microbial community composition, mainly associated with water regime (i.e., soil water availability) and water holding capacity. The second axis described 15.2% of the variation, primarily related to management intensity (tillage > historical tillage or grazing). Climate factors (mean annual precipitation and temperature, radiative dry index, elevation) did not show strong relationships with distribution of microbial communities. Factors such as soil texture (sandy loam), soil inorganic N con-

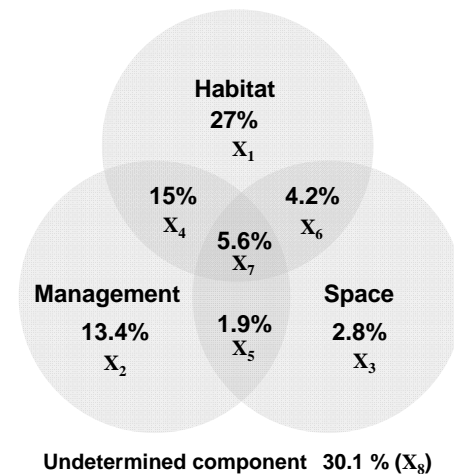


Figure 4. Variation partitioning procedure of microbial community composition, explained by habitat (mean annual temperature and precipitation, radiative dry index, elevation, soil texture, pH, soil C and N content, soil C : N, inorganic N, total PLFAs, water holding capacity), land management (tilled, historically tilled, grazed, flooded) and spatial structure (x , y , xy , x^2 , y^2 , x^2y , xy^2 , x^3 , y^3 : the nine terms in which latitudinal (x) and longitudinal (y) coordinates were used to calculate a cubic trend surface) factors.

tent and pH plotted near the origin, and thus would not be the major drivers of microbial community composition (Fig. 3b).

3.3 Variation partitioning

Forward selection of the three groups of environmental factors with CCA suggested that the soil microbial community composition was significantly related to the habitat (X_1) (mean annual precipitation and temperature, radiative dry index, elevation, soil texture, pH, soil nutrient content, water holding capacity) and land management (X_2) (tillage, grazing, historical tillage, flooding). The variation partitioning procedure showed that total explained variation in microbial community composition was 69.9% ($X_1+X_2+X_3+X_4+X_5+X_6+X_7$) and undetermined variation in it was 30.1% (X_8) (Fig. 4). The largest unique fraction in the explained variation was the effect of habitat (X_1 : 27%), which had a strong overlap with land management (X_4 : 15%). In addition, the land management effect was also considerable (X_2 : 13.4%), whereas the unique effect of spatial structure (X_3 : 2.8%) was very small and statistically not significant.

Table 2. Results of stepwise multiple regression analyses. Independent variables: soil moisture (%), soil total carbon content (C, %), mean annual precipitation (MAP), radiative dry index (RDI), soil water holding capacity (WHC). Dependent variables: soil microbial community composition (SMCC), soil total PLFAs (i.e., microbial biomass, TPLFAs; nmol g^{-1}), percentages of branched PLFAs (gram-positive bacteria) (BP, %), monounsaturated PLFAs (gram-negative bacteria) (MP, %), saturated PLFAs (common in microorganism) (SP, %), fungal PLFAs (F, %), bacterial PLFAs (B, %) and fungal : bacterial PLFAs (F : B). Negative values of parameter estimates refer to negative relationships between the examined dependent variables and the independent variables.

	Variable entered	Parameter estimate	Partial r^2	Probability
SMCC	Soil moisture	–	0.31	0.000
	Tillage	–	0.23	0.000
TPLFAs	Soil moisture	6.794	0.11	0.000
	Soil total C	0.607	0.11	0.000
	RDI	–26.893	0.10	0.000
BP	Soil moisture	0.262	0.57	0.000
	Tillage	1.783	0.06	0.000
MP	Soil moisture	–0.105	0.57	0.000
	Tillage	–3.800	0.17	0.000
SP	Soil moisture	0.329	0.49	0.000
	RDI	–3.796	0.09	0.000
F	RDI	7.074	0.57	0.000
	Tillage	–1.580	0.14	0.000
	Soil moisture	–0.042	0.06	0.000
B	MAP	–0.044	0.20	0.000
	Soil total C	1.218	0.07	0.000
	WHC	0.158	0.06	0.000
	Tillage	1.514	0.05	0.001
F : B	RDI	0.142	0.42	0.000
	Tillage	–0.033	0.12	0.000
	Soil moisture	–0.002	0.11	0.000

3.4 Soil microbial biomass and contributions of microbial group

Soil microbial biomass (i.e., total PLFAs) varied 2.4-fold across all the land use types in this region ($P < 0.05$, one-way ANOVA; Fig. 5a). The highest value appeared in one of the rangelands (ca. 35 nmol g^{-1}), and the lowest value appeared in rice field (ca. 16 nmol g^{-1}). Total PLFAs in artificial grassland, grazed rangeland, farmland and returned cropland had intermediate values.

The contributions of each microbial group across seven land use types varied significantly, except that of actinomycetes. Higher contributions of gram-positive bacteria were found in wetter soils, whereas higher contributions of gram-negative bacteria and fungi were observed in drier soils. The contributions of gram-negative bacteria and fungi were lower in heavily disturbed soils than historically disturbed and undisturbed soils ($P < 0.05$, one-way ANOVA; Fig. 5a–f). Similar to the variation in fungi, the highest fungal : bacterial PLFAs (ca. 0.35) appeared in one of the rangelands, and the lowest value occurred in rice field (ca. 0.15) (Fig. 5g). Sur-

prisingly, the 17cy : precursor ratio (used as an indicator of the anaerobic stress) across seven land use types fluctuated randomly at this regional scale (Fig. 5h).

Stepwise multiple regression analysis demonstrated that 54 % of the variation in microbial community composition could be explained by soil moisture and tillage. Soil moisture, soil total C content and radiative dry index together accounted for 32 % of the spatial variation in total microbial biomass. Soil moisture alone contributed to 57 % of the variation in the contributions of both branched and monounsaturated PLFAs. In this region, radiative dry index, soil moisture and tillage together accounted for 77 and 65 % of the variation in the contribution of fungal PLFAs and fungal : bacterial PLFAs. Of the spatial variability in the contribution of bacterial PLFAs, 38 % could be attributed to the combination of precipitation, soil total C content, water holding capacity and tillage (Table 2).

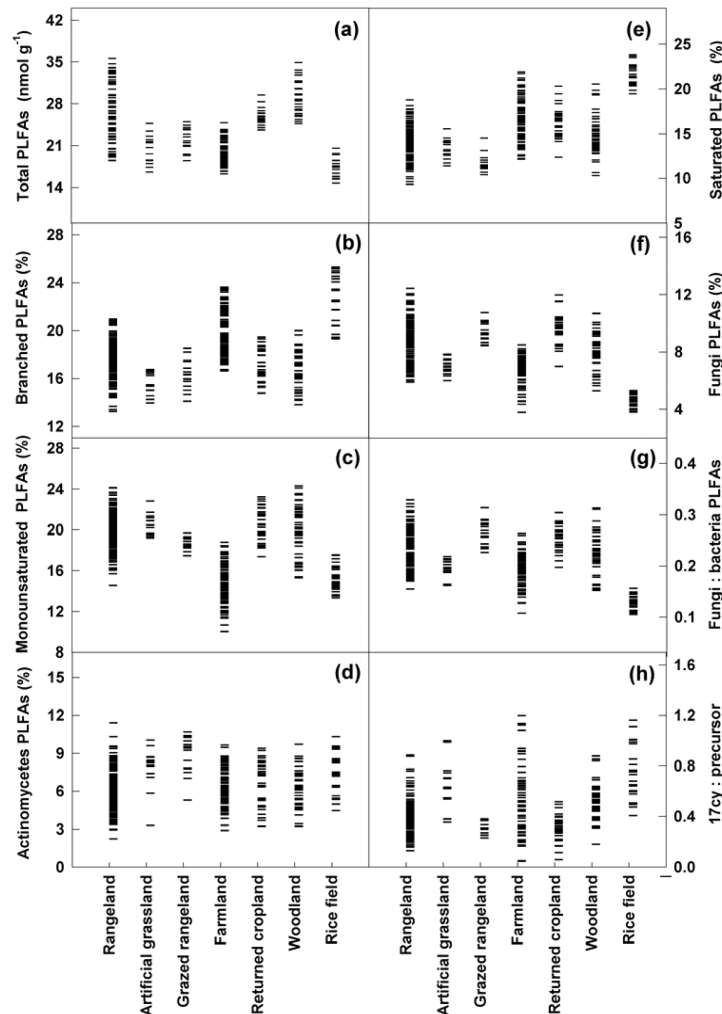


Figure 5. Soil microbial biomass (i.e., total PLFAs), percentages of branched PLFAs (gram-positive bacteria), monounsaturated PLFAs (gram-negative bacteria), actinomycetes (10Me), saturated PLFAs (i.e., common in microorganism), fungi (F), fungal : bacterial PLFAs (F : B) and 17cy : precursor across seven land use types at a regional scale in northeastern China.

4 Discussion

In the exploration of the primary drivers regulating distributions of soil microbial communities and disentangling relative contributions of multiple environmental factors (e.g., climate, soil texture, pH, soil organic matter content, vegetation type), land management practices and spatial structure on microbial community composition and biomass are important challenges in microbial ecology. In this study, soil moisture is one of the main controls on microbial communities across seven land use types at the regional scale, explaining 31 % of the variation in microbial community composition (Fig. 4; Table 2). Drier soils were more enriched in gram-negative bacteria and fungi, whereas wetter soils were more enriched in gram-positive bacteria (Fig. 5). These findings are in agreement with previous observations (Rinklebe and Langer, 2006; Entry et al., 2008; Clark et al., 2009; Dren-

ovsky et al., 2010; Ma et al., 2014). Drought stress likely facilitates increased survival of fungi, because soil fungi rely on more aerobic conditions and are more tolerant of drought due to their filamentous nature (Zhang et al., 2005a). Aerobic filamentous fungi have variable hyphal networks that can relocate water and nutrient resource by cytoplasm translocation (Klein and Paschke, 2004). As soils become water-saturated, soil oxygen levels are reduced, creating an environment favorable for facultative and obligate anaerobic bacteria (Drenovsky et al., 2004).

It has been proposed that the ratio of cyclopropane fatty acids to its precursor can be used to indicate the levels of anaerobic and nutritional stress, because oxygen depletion could trigger conversion of monoenoic fatty acids to cyclopropane fatty acid products (Kieft et al., 1997; Drenovsky et al., 2010). For instance, Drenovsky et al. (2010) reported that cyclopropane fatty acid : precursor ratios (17cy : (16:1 ω 7c);

19cy:(18:1 ω 7c)) were significantly high under conditions of low O₂ concentration and high temperature. However, whether cyclopropane fatty acid is representative of aerobic conditions is debatable. Bossio and Scow (1998) found that the cyclopropane fatty acids were insensitive to water availability across a large-scale precipitation gradient in California. Similarly, our result also show that the 17cy: precursor responded modestly to high water availability in this region (Fig. 5h), whereas we do not know for sure what limits the formation of cyclopropane. This insensitivity to anaerobic conditions in the soils contrasts with its widespread use as an anaerobic marker. These findings suggest that ratios of cyclopropane fatty acids to its precursor are not generally useful as taxonomic indicators of respiratory type at regional scales.

Distinct microbial community composition and biomass are associated with land disturbance levels and management practices at the regional scale in northeastern China. Continuously farming of land is widely occurring in various biomes across the world. Repeated tillage heavily disturbs soil physical properties and decreases soil bulk density and water-retaining capacity (Bescansa et al., 2006). This frequent disturbance in soil properties during tillage (and associated fertilization) could rapidly alter microbial community composition due to the different competitive ability of specific microbial groups. The groups with the capacity of rapid adaptation to the frequently changing soil environment (e.g., bacteria) could take advantage of new resources in disturbed habitats (Cookson et al., 2008; Sun et al., 2011). Consistent with other large-scale studies, conventional tillage soils had higher proportions of gram-positive bacteria and lower proportions of fungi in this study (Fig. 2b) (Galvez et al., 2001; Zhang et al., 2005a). The ability of gram-positive bacteria to sporulate may allow them to withstand tillage or other anthropogenic disturbance. In contrast, fungi are sensitive to disturbance and their hyphae density would decrease significantly in response to tillage (Drenovsky et al., 2010).

Given the strong effects of heavy disturbance on soil microbial communities, it is interesting to find that microbial community composition in lightly and historically disturbed soils (i.e., grazed rangelands, returned croplands) was similar to that in undisturbed soils. These results are supported by observations in other studies (Bardgett and McAlister, 1999; Ingram et al., 2008; Sun et al. 2011). Ingram et al. (2008) reported that long-term light grazing showed no effect on soil organic C content and microbial community composition based on concentrations of PLFA biomarkers in a mixed-grass ecosystem. As the disturbance ceased, microbial biomass increased, probably because more time and resources were available for specific microbial groups that have a slower growth rate (e.g., fungi) (Zhang et al., 2005b). However, Buckley and Schmidt (2003) reported that microbial community composition did not differ significantly between conventionally cultivated fields and fields that had been abandoned from cultivation for 9 years. A possible explanation for this result is that long-term sustainable tillage decreases

soil C content; thus the recovery of soil organic matter to pre-agricultural levels may require decades or even centuries.

Many previous studies have demonstrated that vegetation types, soil properties and spatial structure can influence soil microbial community function and abundance through providing suitable habitats and food sources (Kourtev et al., 2003; Šantrůcková et al., 2003; Han et al., 2007; Chen et al., 2014), whereas our findings of microbial community composition were not related to these factors across this region. In the current study, soils were sampled in different vegetation types and soil nutrient content, but the microbial community composition was very similar at the same geographical location in natural habitats (e.g., meadow vs. wood and shrub; data not shown) (Fig. 5). Similar trends were observed in heavily disturbed habitats: the distributions of microbial communities depended on land disturbance levels and practices rather than agricultural plant species. For example, the farmland land use types (e.g., corn, peanut, mung bean, red bean) in the same location clustered together in CCA ordination despite the different plant species that they represented (Figs. 2, 3, 5). These results are consistent with the study of Drenovsky et al. (2010), who reported that microbial community composition was more strongly influenced by disturbance than by agricultural plant species in California.

Habitat factors and land management triggered complex interactive effects on soil microbial community composition at the regional scale in northeastern China, as the value of shared variance fraction was 15 % without considering the variation explained by all three components (Fig. 4). This was similar to the findings of Drenovsky et al. (2010) that environmental factors caused significant interactions on microbial community composition at large spatial and temporal scales. The significant shared effects in our study could be attributed to the strong effects of land disturbance (e.g., flooding, irrigation, tillage) on soil properties that then affect microbial communities. The findings suggest that land management could partly control soil environmental effects on microbial community composition and biomass at regional scales.

Inconsistent with the hypothesis, soil moisture and land use were the most important factors driving microbial community composition and biomass at the regional scale in northeastern China. In this study, soil moisture was determined not only by natural precipitation but also by managed inputs; thus the effect of precipitation was weaker but still significant. In addition, factors such as spatial structure, soil texture, pH and vegetation types did not have significant relationships with microbial community composition and biomass. These findings will improve predictions of the ecological processes and consequences of ecosystems under global changes.

Appendix A

Table A1. Sample locations (1–23; see Fig. 1), land use types, land management practices, vegetation types, climatic indices and soil properties. MAP, mean annual precipitation (mm); MAT, mean annual temperature (°C); RDI, radiative dry index; ELE, elevation (m); TC, soil total C (%); TN, soil total N (%); SWC, soil water content (%); WHC, water holding capacity; IN, soil inorganic N content (mg kg⁻¹); SL, sandy loam; LS, loamy sand.

Location	No.	Land use type	Management practices	Vegetation type	MAP	MAT	RDI	ELE	pH	Soil texture	C	N	C:N	SWC	WHC	IN
Baogedawula	1	Rangeland	Undisturbed	Desert steppe	237	1.7	1.44	1092	7.7	LS	0.67	0.12	5.32	3	12	2.05
Dabuxiletu	2	Rangeland	Undisturbed	Desert steppe	276	1.4	1.37	1158	7.8	LS	0.79	0.1	7.94	5	15	2.47
	2	Grazed rangeland	Grazed	Desert steppe	276	1.4	1.37	1158	7.9	LS	0.81	0.11	7.31	5	17	3.30
Aqiwula	3	Rangeland	Undisturbed	Steppe	340	1.3	1.33	1239	8.8	SL	1.45	0.15	9.78	7	17	3.46
	3	Woodland	Undisturbed	Wood and shrub	340	1.3	1.33	1239	7.8	SL	0.7	0.15	4.48	9	20	3.32
Dalainuori	4	Rangeland	Undisturbed	Steppe	385	1.3	1.21	1309	8.1	LS	0.84	0.14	7.67	8	18	3.77
Sanyi	5	Woodland	Undisturbed	Wood and shrub	380	2.3	1.21	1173	8	SL	1.11	0.14	7.94	9	22	7.70
Xinchengzi	6	Rangeland	Undisturbed	Steppe	397	3.5	1.23	919	7.7	LS	1.52	0.15	10.07	10	22	4.08
	6	Returned cropland	Historically tilled	Alfalfa	397	3.5	1.23	919	7.7	SL	0.9	0.1	9.96	9	23	7.79
Xinfuzhulu	7	Grazed rangeland	Grazed	Steppe (site 1)	386	5.8	1.18	735	8.4	LS	0.97	0.11	8.95	8	25	5.87
	7	Grazed rangeland	Grazed	Steppe (site 2)	386	5.8	1.18	735	8.3	LS	0.99	0.12	8.05	8	25	4.84
Tianshan	8	Rangeland	Undisturbed	Steppe	386	5.8	1.18	513	8.3	LS	1.66	0.19	8.48	8	23	6.14
	8	Returned cropland	Historically tilled	Almond	386	5.8	1.18	513	8.2	SL	0.9	0.1	8.71	10	25	13.08
Tianshan	9	Rangeland	Undisturbed	Steppe	388	5.8	1.18	413	8.2	LS	1.63	0.19	8.36	9	22	5.24
	9	Returned cropland	Historically tilled	Almond	388	5.8	1.18	413	8.2	SL	1.81	0.17	10.78	10	24	7.34
Shaogen	10	Rangeland	Undisturbed	Steppe (site 1)	385	6.8	1.12	270	8	LS	0.85	0.11	7.66	12	25	5.14
	10	Rangeland	Undisturbed	Steppe (site 2)	385	6.8	1.12	270	8.2	LS	1	0.11	9.36	11	25	4.58
	10	Farmland	Tilled	Corn	385	6.8	1.12	270	8.6	LS	0.9	0.11	8.08	11	24	20.80
Molimiao	11	Rangeland	Undisturbed	Steppe (site 1)	399	6.3	1.05	179	8.4	SL	1.05	0.12	8.85	12	25	7.52
	11	Rangeland	Undisturbed	Steppe (site 2)	399	6.3	1.05	179	8.4	SL	1.1	0.15	7.30	13	25	6.65
	11	Farmland	Tilled	Corn	399	6.3	1.05	179	8.4	SL	1	0.11	9.13	10	25	6.34
Yuxin	12	Rice field	Periodically flooded	Rice	397	6.3	1.02	211	7.8	SL	1.23	0.15	8.23	32	32	5.23
Baixingtu	13	Woodland	Undisturbed	Wood and shrub	414	6	1.02	159	7.7	SL	0.97	0.12	8.08	13	28	8.85
Baolongshan	14	Rangeland	Undisturbed	Meadow (site 1)	415	6	1	156	7.9	SL	1.3	0.13	9.02	13	26	8.45
	14	Rangeland	Undisturbed	Meadow (site 2)	415	6	1	156	7.8	SL	1.34	0.15	8.43	13	27	7.62
	14	Farmland	Tilled	Corn	415	6	1	156	7.7	SL	1.3	0.11	11.92	12	27	6.24
Jiamatu	15	Rangeland	Undisturbed	Meadow (site 1)	422	6	1	149	8.2	SL	1.73	0.17	10.20	14	27	6.08
	15	Rangeland	Undisturbed	Meadow (site 2)	422	6	1	149	8.3	SL	1.77	0.18	10.07	13	28	6.22
	15	Farmland	Tilled	Corn	422	6	1	149	8.2	SL	1.22	0.17	7.19	11	25	10.34
	15	Farmland	Tilled	Red bean	422	6	1	149	8.4	SL	1	0.17	5.56	10	25	18.35
Taipingchuan	16	Rangeland	Undisturbed	Meadow	428	5.6	0.97	150	8.6	LS	1.02	0.13	8.07	18	31	7.37
	16	Rice field	Periodically flooded	Rice	428	5.6	0.97	150	8.3	SL	1.18	0.12	9.83	35	35	8.93
Yaojingzinan	17	Woodland	Undisturbed	Wood and shrub (site 1)	435	5.4	0.97	150	7.9	SL	0.98	0.13	7.27	14	29	5.78
	17	Woodland	Undisturbed	Wood and shrub (site 2)	435	5.4	0.97	150	7.9	SL	1.16	0.16	7.27	13	28	5.78
	17	Farmland	Tilled	Peanut	435	5.4	0.97	150	7.5	LS	0.9	0.15	5.97	10	30	3.23
Yaojingzi	18	Rangeland	Undisturbed	Meadow (site 1)	435	5.4	0.97	159	7.8	SL	1.16	0.16	7.19	17	30	4.47
	18	Rangeland	Undisturbed	Meadow (site 2)	435	5.4	0.97	159	7.7	SL	0.82	0.11	9.43	18	30	5.25
	18	Farmland	Tilled	Peanut	435	5.4	0.97	159	7.5	LS	1.03	0.13	7.96	17	30	4.75
	18	Farmland	Tilled	Mung bean	435	5.4	0.97	159	7.6	SL	1.17	0.15	7.73	17	31	5.75
	18	Farmland	Tilled	Corn	435	5.4	0.97	159	7.8	SL	1	0.12	8.69	20	32	5.95
Yaojingzi	19	Rangeland	Undisturbed	Meadow	434	5.4	0.97	165	8.4	SL	2.21	0.23	9.66	23	34	8.38
Yaojingzi	20	Artificial grassland	Tilled	Meadow (site 1)	433	5.4	0.97	140	8.1	SL	1.85	0.19	9.91	14	33	6.44
	20	Artificial grassland	Tilled	Meadow (site 2)	433	5.4	0.97	140	8.1	SL	1.9	0.19	9.98	12	33	5.62
	20	Farmland	Tilled	Corn	433	5.4	0.97	140	8.1	SL	0.92	0.1	9.23	18	32	8.23
Wulanaodu	21	Rangeland	Undisturbed	Meadow (site 1)	442	5.3	0.93	152	8.1	SL	1.25	0.16	7.89	22	33	4.23
	21	Rangeland	Undisturbed	Meadow (site 2)	442	5.3	0.93	152	8.1	SL	1.3	0.16	8.03	19	34	4.87
	21	Farmland	Tilled	Corn	442	5.3	0.93	152	8.2	SL	1.74	0.24	7.02	20	32	4.12
	21	Woodland	Undisturbed	Wood and shrub	442	5.3	0.93	152	7.5	SL	1.87	0.23	8.11	20	34	6.55
Chaganhua	22	Rangeland	Undisturbed	Meadow (site 1)	467	5.1	0.93	202	8.5	LS	1.54	0.2	7.67	24	36	4.32
	22	Rangeland	Undisturbed	Meadow (site 2)	467	5.1	0.93	202	8.4	LS	1.42	0.19	7.44	22	36	5.01
Wulantuga	23	Rangeland	Undisturbed	Meadow	472	5.1	0.91	291	8.5	SL	2.16	0.2	10.63	23	34	4.85
	23	Farmland	Tilled	Corn	472	5.1	0.91	291	8.2	SL	1.73	0.24	7.36	22	33	7.75
	23	Farmland	Tilled	Peanut	472	5.1	0.91	291	7.9	SL	1.72	0.23	7.76	22	32	3.52
	23	Woodland	Undisturbed	Wood and shrub	472	5.1	0.91	291	7.8	SL	1.63	0.19	8.75	18	35	7.39

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