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**Bacterial contributions of bio-crusts and litter crusts to nutrient cycling in the Mu Us Sandy Land**

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## 19    **ABSTRACT**

20    Desertification has become an important issue for the sustainable development of human society at  
21    global scale and has led to the changes in soil properties and vegetation cover. Biocrusts and litter  
22    crusts play roles in improving the soil microhabitat of sandy ecosystems. Soil microbial  
23    communities mediate ecosystem functions in various ecosystems, e.g., soil biogeochemical  
24    processes. However, limited information is available about how the underlying processes of  
25    bio-crusts/litter crusts restoration are driven by soil bacterial communities in sandy land. Here, we  
26    investigated the changes in soil bacteria from three groups (sandy land, bio-crusts, and litter crusts)  
27    and three soil layers (0-2 cm, 2-5 cm, 5-10 cm) with nine replicates each collected in July 2019  
28    utilized high-throughput pyrosequencing of the V4-V5 rRNA gene region. Most soil nutrients  
29    (SOM, AP, AK, and TN) and enzyme activities (BG and DHA) had differences among the three  
30    groups and three soil layers. OTU richness and diversity of bacteria were positively correlated with  
31    most soil variables. The constructed co-occurrence networks between soil variables and bacterial  
32    communities, and within bacterial communities showed that bacterial taxa had closer relationships  
33    with all soil variables in crusts than sandy land and varied among the three sand groups (sandy land,  
34    bio-crusts, and litter crusts). The result showed that the composition of bacterial community was  
35    regulated mainly by soil variables and crust types. Compared with sandy land, more predictors in  
36    nutrient cycling were found in crust types. They played major roles in nutrient cycling in desert  
37    ecosystem restoration on the basis of random forest modeling. Our findings indicate some bacterial  
38    taxa may played the predominant roles in connecting with soil variables and other bacterial taxa  
39    across crusts types, and litter crusts and bio-crusts drive the nutrient cycling by mediating the  
40    restoration of bacterial taxa in sandy ecosystems.

41

42   **Keywords:**

43   Bacterial communities; Bio-crusts; Litter crusts; Nutrient cycling; Sandy land restoration

44

45   **1. Introduction**

46       Land desertification poses a great threat to all types of ecosystems, it can damage ecosystem's  
47   basic functions and services to sustain life, causes the losses in soil nutrients, the decline in soil  
48   potential productivity, and the reduction in vegetation (D'Odorico et al., 2013). Arid and semiarid  
49   areas are among the most susceptible to land desertification, however, they covered approximately  
50   one-third of the earth's land and have been expanding rapidly, this is due to climate change and  
51   human activities, such as overcultivation, overgrazing, and urbanization (Asner et al., 2003;  
52   Sivakumar et al., 2005)(Gao et al., 2017). With the increase of the world population and  
53   deterioration of living environment, desertification has becoming one of a major issue for the in  
54   human societies at global scale(D'Odorico et al., 2013). For instance, according to The  
55   Desertification and Sanditification Sate of China, China had a desert area of 2.6 million square  
56   kilometers, and another 1.7 million square kilometers of sandy area in 2014, which covers about  
57   27.2% and 17.9% of the country's land, respectively (State Forestry Administration, 2015). The Mu  
58   Us Sandy Land, which is located in central north of China, is the region with high risks of  
59   desertification in arid Asia (Wang et al., 2017). In 1999, the Grain for Green Program was launched  
60   by Chinese government with aims to halt soil erosion and improve the ecological environment (i.e.  
61   the losses of soil fertility and the decrease of vegetation coverage). It is the largest ongoing  
62   revegetation project in China and also one of the largest conservation projects in the world. This  
63   project converted croplands into grasslands or shrubs and increased vegetation coverage from  
64   31.6 % in 1999 to 59.6 % in 2013 on the Loess Plateau (Chen et al., 2015b; Uchida et al., 2005).

65 Meanwhile, apart from the "Grain for Green" Program, several other initiatives have been carried  
66 out to restore soil fertility and alter the sand surface to control desertification in sand areas, such as  
67 mechanical sand barriers (Bo et al., 2015) and afforestation (Zeng et al., 2008). These initiatives  
68 enhanced development of bio-crusts and litter crusts on the Loess Plateau, either directly or as a  
69 result of a general improvement of environmental conditions. Better environmental conditions,  
70 including appropriate humidity and temperature, promote the development of bio-crusts and litter  
71 crusts in the Mu Us Sandy Land (Liu et al., 2019b).

72 Biological soil crusts (bio-crusts), which are composed of cyanobacteria, lichens, mosses,  
73 fungi, and other nonvascular photoautotrophs, are typical for dryland ecosystems worldwide and  
74 represent an essential functional component of the pedosphere. Bio-crusts can stabilize soil,  
75 increase soil fertility, impact hydrologic cycles, alter soil organic matter content, and provide a  
76 home for belowground organisms (Reed et al., 2019; Torres-Cruz et al., 2018). Litter crusts are  
77 defined as the cohesiveness of the soil surface shaped by litter and soil and forms a hard shell by the  
78 mixing of sand and litter organisms in the wind - water erosion crisscross zone Bio-crusts and litter  
79 crusts play crucial roles in improving microhabitat conditions, forming soil organic matter, affecting  
80 hydrological processes, and soil bacterial communities in sandy lands during restoration (Jia et al.,  
81 2018; Leloup et al., 2018; Liu et al., 2019b).

82 In this context, soil variables are most important factors to impact microbial communities, such  
83 as soil pH, soil texture and, available nutrients (Chen et al., 2015a; Fierer and Jackson, 2006). On  
84 the contrary, microbial communities are important indicators of rehabilitated ecosystems (Banning  
85 et al., 2011) and drive the Earth's biogeochemical cycles (Falkowski et al., 2008). Soil bacterial  
86 communities represent the greatest biodiversity reservoir and greatly affect ecosystem functions and  
87 services (Falkowski et al., 2008; Wagg et al., 2014). An enhanced appreciation of the connection

88 between environment and microbial ecology, in the last decade, has led to many studies focused on  
89 the distribution of soil microbial communities (Karimi et al., 2018), the influence of microbial  
90 diversity on plant community (Jiao et al., 2019) and multifunctionality of terrestrial ecosystem  
91 (Delgado-Baquerizo et al., 2016; Falkowski et al., 2008; Jiao et al., 2019; Karimi et al., 2018).  
92 However, limited information is available about the response of bacterial communities on soil  
93 variables in natural desert ecosystems. Moreover, each microbe may play a different functional role  
94 in complex microbial ecosystem (Li et al., 2019). Experimental evidence suggested that the  
95 bacterial communities at phylum level are similar in the two sample types. The relative abundance  
96 of several genera has considerably differences at the genus level (Jakobsen et al., 2019). However,  
97 the foundational role of the bacterial genera in regulating key ecosystem processes (i.e. nutrient  
98 cycling) of litter crusts and bio-crusts in the sandy ecosystem is lacking. Thus, we must expand our  
99 insight into the functions of the microorganisms, particularly bacteria genera, in the bio-crusts and  
100 litter crusts in the restoring sand ecosystem.

101 The present study aims to (1) elucidate the variations in soil quality and bacterial communities  
102 coupled with the soil quality of bio-crusts and litter crusts, (2) explore the correlations between the  
103 soil bacterial taxa and soil variables and the bacterial taxa among themselves in the bio-crusts and  
104 litter crusts networks, (3) identify the contributions of the annotated bacterial taxa to nutrient  
105 cycling during ecosystem restoration. To achieve these aims, we used high-throughput  
106 pyrosequencing of the V4-V5 rRNA gene region to compare the variations of bacterial communities  
107 in relation to soil variables, their co-occurrence networks, and their contributions to the soil  
108 functioning (nutrient cycling) in bio-crusts and litter crusts of restoring sandy ecosystems.

109

## 110 **2. Materials and methods**

111

## 112 **2.1. Study sites and sample collection**

113 The study was carried out in the eastern part of the Mu Us Sandy Land ecosystem (110°21'  
114 -110°23'E, 38°46'-38°51'N; 1080-1270 m Altitude), located in Shenmu County at the northern of  
115 Shanxi Province, China. This region, as the most arid area in Asia, is one of the largest dune areas in  
116 the north of China (Wang et al., 2017). Psammophytic shrubs and herbaceous plants are mainly  
117 dominant plant species in this study site (Jia et al., 2018). According to our previous studies,  
118 bio-crusts and litter crusts, as two major contributors, covered about 40% and 30% of the Mu Us  
119 Sandy Land, respectively (Jia et al., 2018). To compare the effects of bio-crusts and litter crusts on  
120 the surface microhabitats of the sandy land, three sites (sandy land, bio-crusts, litter crusts) with  
121 similar environmental conditions in terms of underlying subsoil, microtopography, and soil  
122 hydrology, were selected and the distance between them was above 500 m apart. Sand samples were  
123 collected in July 2019 from sampling sites covered by bio-crusts and litter crusts, respectively. After  
124 removing the litter horizon, nine replicate sites were randomly selected above 10 m apart and three  
125 sand layers (0-2 cm, 2-5 cm, 5-10 cm). Each replicate was mixed with five sand cores by a zigzag  
126 pattern (Liu et al., 2019a). In total, 81 soil samples = 3 sites (sandy land, bio-crusts, litter crusts) × 3  
127 depths (0-2 cm, 2-5 cm, 5-10 cm) × 9 replicates were obtained, and all the sand samples were taken  
128 to the laboratory on ice within 24 h. A small part of each sample (~2 g) for the DNA analysis was  
129 transported to the company (Novogene, Beijing, China) on ice. Another part was sieved (~2 mm)  
130 for the analysis of soil properties. The rest was stored at -80 °C.

## 131 **2.2. Sand characters and enzyme activities**

132 Sand organic matter (SOM) were measured by potassium dichromate colorimetric method  
133 (Nelson and Sommers, 1982); total phosphorus (TP), and available phosphorus (AP) were

determined by molybdenum anti-colorimetric method (Olsen and Sommers, 1982); ; total nitrogen (TN) was determined with the Kjeldahl method (Bremner and Mulvaney, 1996); total potassium (TK), and available potassium (AK) were determined with Flame photometry as described previously (Page et al., 1982).  $\beta$ -glucosidase (BG), dehydrogenase (DHA), Urease (UA), alkaline phosphatase (ALP), and acid phosphatase (ACP) were measured by the methods described in Tabatabai, 1994 (Tabatabai, 1994; Taylor et al., 2002). Briefly, BG activity was determined as the amount of p-nitrophenol (PNP) released; ALP and ACP were determined by p-nitrophenol (PNP) released at pH 11 and 6.5, respectively. UA was measured by the determination of ammonia released. DHA was determined as the amount of the TPF released (Taylor et al., 2002). All the samples were determined using three replicates. The selected parameters reflect either resource pools (SOM, TN, TP, AP, TK) in biogeochemical cycles or important processes regulating availability of these pools (BG, DHA, UA, ALP, ACP). Such as, the important ecosystem processes related to the cycling of carbon (SOM, BG and DHA), nitrogen (TN and UA), phosphorus (AP, TP, ALP, and ACP), and potassium (AK and TK) (Jiao et al., 2019; Jing et al., 2015).

148

### 149 **2.3. DNA extraction and 16S rRNA gene sequencing**

150 Genomic DNA from each sand sample was extracted from 1g sand using the OMEGA soil  
151 DNA Kit (Omega Bio-Tek, Inc., Norcross, GA, USA) based on the manufacturer's instructions. To  
152 assess the bacterial communities, the V4–V5 region of the bacterial 16S rRNA gene was amplified  
153 with the primers: 515F (5' – GTGCCAGCMGCCGCGGTAA – 3') / 907R (5' –  
154 CCGTCAATTCCTTTGAGTTT – 3'). The purified PCR products were pooled and sequenced on  
155 the Illumina HiSeq (300~bp paired-end reads) platform (Novogene, Beijing, China).

156



## 157 2.4. Data analysis

158 The acquired sequences were processed to remove low - quality sequences using the QIIME  
159 pipeline (Caporaso et al., 2011). The remaining sequences were classified into operational  
160 taxonomic units (OTUs) with 97% sequence similarity by the UPARSE pipeline (Edgar, 2013).  
161 OTU richness, Chao1 index, Shannon index, ACE index, and Simpson index were calculated by the  
162 OTU table in QIIME (Caporaso et al., 2010). The changes in soil nutrients, enzyme activities, and  
163 bacterial communities, as well as the relative abundance of the microbial phyla among three soil  
164 groups (sandy land, bio-crusts, and litter crusts) and three soil layers (0-2 cm, 2-5 cm, and 5-10 cm),  
165 were conducted based on one-way ANOVAs with Tukey's tests by GraphPad Prism version 8.0.2  
166 (GraphPad Inc. San Diego, CA, USA). The normality of data and the equality of variance were  
167 tested. If the data did not meet normality or homogeneity, non-parametric Kruskal-Wallis analyses  
168 were used. Redundancy analysis (RDA) was performed to visualize the influence of soil variables  
169 on bacterial community composition in R package "vegan" (Oksanen et al., 2013).

170 The co-occurrence networks were constructed for bio-crusts and litter crusts based on  
171 significant correlations between bacterial genera and all of the soil nutrients (Pearson's correlation,  
172  $p.\text{thres} = 0.05$ ,  $r.\text{thres} = 0.6$ ), and among the bacterial genera themselves (Pearson's correlation,  
173  $p.\text{thres} = 0.05$ ,  $r.\text{thres} = 0.8$ ), which were visualized by R packages 'igraph' (Hartmann et al., 2015;  
174 Qian et al., 2018). In the co-occurrence networks, each node represents one bacterial genus and  
175 each edge represents a significant correlation between two nodes. A set of metrics: number of edges  
176 (Num. edges), average. degree, average. path. length, diameter, and modularity were calculated to  
177 estimate network topological features. Num. edges represent the number of edges. To identify the  
178 major statistically significant bacterial predictors for sand nutrient cycling, a Random Forest (RF)  
179 modeling was performed with the forest (5,000 trees) using the "random Forest" package (Archer,

2016). The model significance was computed by the R package “A3” (Fortmann-Roe, 2013). A total of 35 classified microbial phyla and 143 annotated genera from 6 predictors at the phylum level were selected in the Random Forest modeling. Percentage increases in the mean squared error (MSE) was used to estimate the importance of variables. All statistical analyses were performed by GraphPad Prism 8.02 or R software (v3.6.3; <https://www.r-project.org/>).

### 3. Results

#### 3.1. Variation in sand nutrients and enzyme activities

Soil nutrients and enzyme activities typically differed among three soil groups and three sand layers, most of which were significant differences. Litter crusts had significantly higher SOM, TN, AP, AK, BG, DHA, UA, and ACP than those in the sandy land and the bio-crusts in surface soil (0-2 cm and 2-5 cm); TK, SOM, TK, AK, TN, ALP, ACP were significantly higher in the bio-crusts than those in the sandy land in the surface soil (0-2 cm and 2-5 cm). SOM, AP, AK, BG, and DHA were highest in the topsoil (0-2 cm) of the bio-crusts and litter crusts. (Fig. 1). These results indicated that both litter crusts and bio-crusts improved the sand soil quality compared with sandy land, and litter crusts had more significant improvement in sandy soil quality than bio-crusts, especially, in the surface of the sand layer. The statistical information listed in Table S1.

#### 3.2. Variation in sandy bacterial community

In total, 6,865,683 high-quality sequences were classified into 14560 operational taxonomic units (OTUs) after the 97% sequence similarity cutoff across 81 sand samples (Table S2). The OTU richness in biocrust and the litter crusts were significantly greater than this in the sandy land in the surface soil (0-2 cm and 2-5 cm) (Fig. 2B and Table S3). Bacterial OTU were primarily classified into the 66 microbial phyla, and the most dominant bacterial phyla were the phyla Proteobacteria

(28.2%), Actinobacteria (23.86%), Acidobacteria (18.92%) (Fig. 2A, and Table S4). 39.57% OUT were classified into the 707 genera, the most dominant identified genera were Sphingomonas (2.4%) and Nocardioides (2.2%), respectively belonging to the Proteobacteria, and Actinobacteria phyla (Table S5). For most of the bacterial phyla and genera, relative abundance of the phyla or genera bacteria significantly different among three sand groups ( $P > 0.05$ ) (Table S4, Table S5). The Shannon index of bacterial community diversity in the litter crusts and bio-crusts were significantly higher than this in sandy land in the surface soil (0-2 cm and 2-5 cm). The bacterial communities clearly differed among three soil groups (Fig. 2E, Fig. 2F, Fig. 2H and Table S3).

211

### 212 **3.3. Co-occurrence patterns of soil bacterial communities**

213 Redundancy analysis (RDA) was performed to investigate the influence of soil variables on  
214 bacterial community among three sand groups and three sand layers. We found that soil variables  
215 well explain dynamic changes of bacterial communities at the phylum level or genus level.  
216 However, the correlations are not alike between the level of bacterial phylum and genus (Fig. 3 and  
217 Table S6). For instance, AK was the most important variable for litter crusts bacterial community at  
218 genus level, whereas, SOM, AP, TN, DHA, GB, and UA were more important at the genus than at  
219 the phylum level (Fig. 3 and Table S6). Furthermore, the co-occurrence networks were constructed  
220 for three sand groups to investigate the correlations between the bacterial genera and soil nutrients  
221 (Fig. 4A and Table S7). The network structure was distinct, the bio-crusts network had more edges  
222 (34) than sandy land network (33) and litter crusts network had more than two times (2.71) as many  
223 edges as found in bio-crusts. There is the highest average degree in the litter crusts and higher  
224 average degree in the bio-crusts network than these in the sandy land network. These showed that  
225 most soil nutrients had closer relationships with bacterial genera in the litter crusts than in the

226 bio-crusts and sandy land.

227        Given the interactions between bacterial taxa, we constructed co-occurrence networks to  
228 explore the interactions between the bacterial genera with each other in three sand groups,  
229 respectively (Fig. 4B and Table S8). Diverse topological characteristics were observed. The  
230 modularity indices were 0.8810.587, and 0.892 in the sandy land, bio-crusts, and litter crusts  
231 network, respectively. These values (The modularity indices  $> 0.4$  show that the network has a  
232 modular structure) mean these networks had a modular structure. The highest average degree in the  
233 sandy land network than these in the litter crusts and bio-crusts networks. There were 975 edges and  
234 94.26% positive correlations identified as co-occurrences in the bio-crusts and 1,787 edges and  
235 99.89% positive correlations in the litter crusts. In sum, bacterial co-occurrence patterns were  
236 distinctly different between the bio-crusts and litter crusts.

237

### 238 **3.4. The potential contributions of bacterial taxa to sand nutrient cycling**

239        The contributions of bacterial communities to sand functionings were evaluated by Random  
240 Forest (RF) modeling. We uncovered the potential major bacterial drivers of sand nutrient cycling in  
241 the crust types by RF analysis, including 35 microbial phyla. We discovered that 9 bacteria phyla  
242 were the most important predictors to nutrient cycling in the bio-crusts and litter crusts (Fig. 5A).  
243 Planctomycetes ,    Cyanobacteria ,    Armatimonadetes ,    Rokubacteria ,    Nitrospirae ,  
244 Latescibacteria ,    Deinococcus-Thermus were predictors in bio-crusts and Actinobacteria,  
245 Chloroflexi, Planctomycetes, Cyanobacteria, Armatimonadetes, Rokubacteria, Latescibacteria, and  
246 Deinococcus-Thermus were predictors in litter crusts. Furtherly, 143 annotated genera from the  
247 Actinobacteria, Planctomycetes, Chloroflexi, Armatimonadetes, Deinococcus-Thermus, Nitrospirae  
248 were selected to identify the major predictors at the genus level (Table S5). Compare to sandy land,

249 more predictors were found in the bio-crusts and litter crusts. We observed 22 and 20 predictors to  
250 nutrient cycling in bio-crusts and litter crusts, respectively (Fig. 6). *Blastococcus*, *Couchioplanes*,  
251 *Crossiella*, *Geodermatophilus*, *Actinoplanes*, *Parviterribacter*, *Marmoricola*, *Rhizocola*,  
252 *Tepidisphaera*, and *Fimbriiglobus* are the same predictors to the nutrient cycling between the  
253 bio-crusts and litter crusts. Most of the predictors were distinct between bio-crusts and litter crusts  
254 not only at the phylum level but also at the genus level and the particular bacterial consortium made  
255 important contributions to soil functionings.

256

## 257 **4. Discussion**

258

### 259 **4.1 Influence of crusts on sand characters and bacterial communities**

260 Bio-crusts and litter crusts improved sand surface microhabitats, including soil properties and  
261 hydrological processes, and caused the development of soil fertility (Ferrenberg et al., 2018; Liu et  
262 al., 2019b). Our study showed that most sand nutrients, enzyme activities, and the diversity of soil  
263 bacteria communities increased markedly during the development of bio-crusts and litter crusts.  
264 This finding indicated that these crusts have a positive effect on sandy ecosystem restoration.  
265 In this study, litter crusts enhanced most sand nutrients and enzyme activities compared with  
266 bio-crusts. This result is consistent with previous study that litter crusts significantly increase soil  
267 organic matter than those in bio-crusts (Liu et al., 2019b). This condition can be apparently and  
268 partially due to many substrates for decomposition provided from the litter crusts, thereby  
269 elucidating the improvement of soil quality in sandy litter crusts. Soil quality determined the nature  
270 of vegetation series and the achievement of ecological restoration (Putten, W.H 2013). Previous

271 studies indicated that the diversity of bacteria typically increases with ecosystem restoration. The  
272 richness and diversity of bacterial communities in the litter crusts and bio-crusts were significantly  
273 higher than in the sandy land. This condition is attributed to the opportunities for the interactions of  
274 different species among themselves are provided by the improved soil quality (Liu et al., 2019a).  
275 These crusts can provide favorable environment for vegetation species formation via improving soil  
276 surface microhabitats of sandy land in the wind-water erosion crisscross region. Moreover, most  
277 soil nutrients contents and enzyme activities were greatest in topsoil, as shown in previous results  
278 (Liu et al., 2018). These differences in soil properties may have an influence on the soil  
279 microorganisms. The diversity of bacteria was higher in the deep layer (5-10 cm) than in the other  
280 layers in sandy land. This finding is inconsistent with studies that the diversity of bacterial  
281 communities commonly decreases with increasing soil depth (Jiao et al., 2018a). This result may be  
282 due to the specific environmental conditions of sandy land, including the high air temperature, low  
283 soil humidity, and abundant solar radiation in surface soil (Liu et al., 2018).

#### 284 **4.2 Influence of litter crusts and bio-crusts on co-occurrence networks**

285 Many studies have reported that the co-occurrence patterns of complex ecological interactions  
286 that form bacterial communities can demonstrate the interactions of soil variables and bacterial taxa.  
287 These patterns are generally used to evaluate the information on community interactions in natural  
288 habitats. In our co-occurrence networks, the interactions of soil variables and bacterial genera were  
289 complicated in the litter crusts network. This result may be related to the accumulation of soil  
290 nutrients and enzyme activities that contributed to the bacterial community activity. The bio-crusts  
291 and litter crusts networks were significantly different at the genus level, which may be due to their  
292 heterogeneity in response to the soil properties and different habitats.

293 Soil microbes may be related to the soil properties and among themselves through various

mechanisms (Ma et al., 2016). In this study, we found that the bacterial networks were distinct among the three sample groups. Bacterial genera had more connections with each other in the litter crusts network (num. edges = 1787) than in the bio-crusts network (num. edges = 975). More positive correlations identified as co-occurrences were found in the litter crusts (99.89%) than in the bio-crusts (94.26%). Dominant positive correlations illustrate that most bacterial genera may share similar ecological niches or synergistically operate in the litter crusts environment, which is consistent with other microbial networks (Aschenbrenner et al., 2017; Zhang et al., 2018). The bacterial taxa enriched in the litter crusts might benefit from sufficient soil nutrients, thereby enabling them to take up the leading ecological niches in the interaction network. In litter crusts network, the most important genera, including *Romboutsia*, *Paeniclostridium*, and *Mogibacterium* belong to the phyla Firmicutes. The lowest relative abundance of these genera and the most important roles were observed. Our study showed that the relative abundance of bacterial taxa is not directly related to ecosystem function, consistent with sulfate reducer *Desulfosporosinus* with the low (0.006%) abundance managed the majority of soil SO<sub>4</sub> reduction (Pester et al., 2010). It is known that members of the Firmicutes have ability to degrade cellulose in the litter. But, the highest relative abundance of the Firmicutes were observed in the sandy land probably because the much phenotypic variation of its members enables these organisms to live in various environments (Lawson et al., 1993), and many members with spore-forming ability are able to endure harsh environmental conditions (Zhuang et al., 2010). However, negative correlations (5.74%), which show co-exclusion between the two bacterial genera, were rarer than positive ones in the bio-crusts network. The number of negative links was higher than in the litter crusts network, probably due to a more competitive connection between bacterial genera in the bio-crusts. More negative correlations were found between *Gaiella* and other genera, for instance, *Gaiella* and *Hydrocarboniphaga*,

317 Microvirga, or Belnapia. Gemmatimonas, Hydrocarboniphaga, Microvirga, and Belnapia were the  
318 important genera in the bio-crust network. Compared with the sandy land, the highest relative  
319 abundance of the Gaiella and lower abundance of the Hydrocarboniphaga, Microvirga, or Belnapia  
320 were found in the bio-crusts. Gaiella is chemoorganotrophic and had the ability to utilize organic  
321 acids, amino acids, and some sugars as single carbon sources but not utilize hydrocarbons as carbon  
322 (Albuquerque et al., 2018). Hydrocarboniphaga, Microvirga, or Belnapia belong to the  
323 Proteobacteria, these members were facultative and aerobic bacteria and can utilize various organic  
324 substrates (Slezak et al., 2017). Most organisms contain or produce small amounts of hydrocarbon,  
325 such as, the fermentation of many bacteria in the soil and the decomposition of plants.  
326 Hydrocarboniphaga active in hydrocarbon degradation (Palleroni et al., 2004) and Microvirga can  
327 degrade some hydrocarbon (i.e. Tween 20, D-sorbitol, adonitol, and alpha-iso-leucine) as carbon and  
328 nitrogen sources (Veyisoglu et al., 2016). This observation may be because the accumulation of  
329 hydrocarbons has an opposite effect on the growth of Gaiella and Hydrocarboniphaga or Gaiella  
330 and Microvirga in bio-crusts. These results may indicate the preferences of specific bacteria for soil  
331 crust types and substrates.

#### 332 **4.3 Bacterial community predictors of sandy nutrient cycling**

333 Plant and microbial diversity drive terrestrial ecosystem multifunctionality  
334 (Delgado-Baquerizo et al., 2016). Recent research provides evidence that microbial communities  
335 play pivotal roles in driving soil nutrient cycling in reforested ecosystems (Jiao et al., 2018a). Our  
336 results showed that the members of the predictors varied with the crust types in sandy land.  
337 Nitrospirae was the important and unique predictor to the nutrient cycling in the bio-crusts likely  
338 due to its diverse metabolism. Most of its genera are aerobic chemolithotrophs, including nitrifiers,  
339 dissimilatory sulfate reducers, and magnetotactic forms (Garrity and Holt, 2001). Nitrospirae is an



340 extensive nitrite-oxidizing bacterial taxa and plays a major role in the soil nitrogen cycle. In our  
341 study, Nitrospirae showed the highest relative abundance in the bio-crusts, and had a slight higher  
342 content of TN and UA compared with in sandy land, which participate in nitrogen cycle. This result  
343 showed that the nitrogen-cycling bacterial group was crucial in the development of the bio-crusts,  
344 and the soil conditions in the bio-crusts were enhanced with long-term crust restoration in the sandy  
345 land. Actinobacteria and Chloroflexi are the important and unique predictors to the nutrient cycling  
346 in the litter crusts. This condition is probably because Actinobacteria, as an excellent indicator of  
347 soil biological activity, metabolize cellulose, lignin, and other complex polymers, mediate the  
348 decomposition of organic matter in ecosystems, and influence the nutrient cycling in the soil (Kirby,  
349 2005). The relative abundance of Actinobacteria accounted for the high proportion across the three  
350 groups with their capacity to colonize bare soil (Suela Silva et al., 2013). However, their decrease  
351 with the increase in soil nutrients agrees with the study that Actinobacteridae are more abundant in  
352 patches without vegetation than in shrubs (Hortal et al., 2013). Actinobacteria play a beneficial role  
353 in the soil by providing protection against abiotic stresses and enhancing plant nutrition acquisition  
354 (Shi et al., 2019). Chloroflexi can offer energy through photosynthesis, degradation of plant-derived  
355 compounds, and organic matter decomposition (Wang et al., 2018). Previous study showed that it  
356 was negatively correlated with TN and organic carbon in the litter crusts (Lozano et al., 2014) and  
357 the relative abundance of Chloroflexi decreased with the development of soil (Brown and  
358 Jumpponen, 2014). These findings are consistent with our study that showed the lowest relative  
359 abundance of Chloroflexi and the highest content of TN and organic matter, and its relative  
360 abundance is indirectly correlated with soil nutrients and enzyme activities. Actinobacteria and  
361 Chloroflexi are pivotal in predicting the cycling of sand nutrients under the litter crusts. Our  
362 experimental results reveal the distinct contributions of bacterial taxa to soil functions

363 (multi-nutrient cycling) in bio-crusts and litter crusts in sandy lands.

364       Increasing attentions in manipulating host-microbiome interactions by adding bacteria in a  
365 range of systems should focus on a fine scale to analyze the relationships between the microbial  
366 populations and soil functioning under natural conditions. However, most studies have focused on  
367 the microbial indicator at the phylum level or class level of bacteria. In our study, we aimed to  
368 determine the predictors of nutrient cycling at the genus level. Our results indicated that  
369 approximately 31% of predictors (10) were the same between the bio-crusts and litter crusts at  
370 genus level, and they were more connected with other bacterial genera in the co-occurrence  
371 networks. Hence, they may affect the soil ecosystem functioning by contributing to nutrient cycling  
372 in the crust types of ecosystem restoration. The unique bacterial predictors' groups in the bio-crusts  
373 or litter crusts with these bacterial consortiums can play important roles in nutrient cycling in  
374 different habitats. This finding is consistent with the study of distinct microbial communities that  
375 can exhibit distinct responses in different habitats (Wagner et al., 2016). The number of bacterial  
376 predictors were more in the crusts than in the sandy land. This result suggested the importance of  
377 soil bacterial communities in impacting ecosystem functioning (multiple nutrient cycling) during  
378 the development of the bio-crusts and litter crusts in sandy land. Our results showed particular  
379 bacterial consortium play important roles in predicting soil nutrient cycling in sandy ecosystem  
380 restoration. In a microbial ecosystem, the identification of the key microbial populations is often  
381 associated with the occurrence and abundance of species in local habitat (Mei et al., 2016).  
382 Therefore, the important drivers at the genus level with litter crusts and bio-crusts contribute to the  
383 applications of the key microbial driver in ecosystem restoration. These observations indicate  
384 bacteria participate in the biogeochemical cycling of multi-nutrients in the litter crusts and  
385 bio-crusts and the importance of investigating distinct responses contributed to sand nutrient cycling

386 in the sandy ecosystem restoration. These studies enrich our knowledge on crusts and bacterial  
387 communities in restoring sandy land.

## 388 **5. Conclusions**

389 Soil bacterial communities represent important variables for predicting nutrients cycling of  
390 restoration trajectories, thereby affecting belowground ecological restoration. In this study, the  
391 bacterial communities showed increased diversity and varied composition and structure in the crust  
392 types compared with the sandy land. The diversity and OUT richness were positively correlated  
393 with soil nutrients (except TP) in surface soil (0-2 cm and 2-5 cm). Litter crusts network had closer  
394 relationships between the soil bacterial taxa and soil nutrients and more positive correlations among  
395 themselves than in the bio-crusts network. The bacterial drivers play the most important roles in  
396 mediating sand nutrient cycling in the crust types of ecosystem restoration. These findings increase  
397 our understanding of the complex interactions between bacterial communities and crust types  
398 during the ecosystem recovery. The distinct response strategies of particular bacterial groups at the  
399 genus level can be important for the comprehensive understanding of the belowground microcosms  
400 with litter crusts and bio-crusts in the surface sand. Our study provides a new perspective that the  
401 exploration of the specific functions of particular bacterial consortiums in nutrient cycling is crucial  
402 to their applications in pivotal ecosystem functioning. Future work should be conducted to isolate  
403 the most important drivers of the bacterial taxa. Bacterial inoculants may promote soil bacterial  
404 functioning in nutrient cycling and may be potentially implemented as an approach for increasing  
405 soil fertility in degraded lands or agricultural lands.

## 407 **Supporting Information**

408 Appendix A: Table S1, S2, S3, S4, S5, S6, S7 and S8

409

## 410 **Declaration of Competing Interest**

411 The authors declare no competing financial interests.

412

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420

## 421 **Authors' contributions**

422 G.L.W conceived the idea and designed the study; X.L., Y.L. and L.Z. performed the experiment

423 and collected the data; X.L. analyzed the data; X.L. and G.L.W led the writing of the manuscript

424 with the help of R.Y. All authors contributed critically to the draft and gave final approval for

425 publication.

426

## 427 **References**

428 Albuquerque, L., Rainey, F.A., Costa, M.S., 2018. Gaiella. Bergey's Manual of Systematics  
429 of Archaea and Bacteria, 1-7. <https://doi.org/10.1002/9781118960608.gbm01469>.

430 Archer, E., 2016. rfPermute: Estimate Permutation p-Values for Random Forest Importance Metrics.  
431 <https://CRAN.R-project.org/package=rfPermute>.

432 Aschenbrenner, I.A., Cernava, T., Erlacher, A., Berg, G., Grube, M., 2017. Differential sharing and  
433 distinct co-occurrence networks among spatially close bacterial microbiota of bark, mosses  
434 and lichens. Molecular Ecology, 26, 2826-2838. <https://doi.org/10.1111/mec.14070>.

Asner, G.P., Archer, S., Flinthughes, R., Ansley, R.J., Wessman, C.A., 2003. Net changes in regional woody vegetation cover and carbon storage in Texas drylands, 1937-1999. *Global Change Biology*, 9, 316-335. <https://doi.org/310.1046/j.1365-2486.2003.00594.x>.

Banning, N.C. et al., 2011. Soil microbial community successional patterns during forest ecosystem restoration. *Applied and Environmental Microbiology*, 77, 6158-6164. <https://doi.org/10.1128/AEM.00764-11>.

Biswas, S., Shivaprakash, M.K., 2019. Comparative studies of concomitant release of secondary and micronutrients by Potassium Solubilizing Bacteria (KSB) from different minerals. *International Journal of Pure and Applied Bioscience.*, 7, 341-345. <http://dx.doi.org/10.18782/2320-7051.7699>.

Bo, T.L., Ma, P., Zheng, X.J., 2015. Numerical study on the effect of semi-buried straw checkerboard sand barriers belt on the wind speed. *Aeolian Research*, 16, 101-107. <http://dx.doi.org/10.1016/j.aeolia.2014.10.002>.

Bremner, J.M., Mulvaney, C.S., 1996. Nitrogen-total. In: Page, A.L., Ed., *Methods of soil analysis, Part 2. Chemical and Microbiological Properties*. American Society of Agronomy and Soil Science Society of America, Madison, 72, 595-624.

Brown, S.P., Jumpponen, A., 2014. Contrasting primary successional trajectories of fungi and bacteria in retreating glacier soils. *Molecular Ecology*, 23, 481-497. <https://doi.org/10.1111/mec.12487>.

Burke, D.J., Weintraub, M.N., Hewins, C.R., Kalisz, S., 2011. Relationship between soil enzyme activities, nutrient cycling and soil fungal communities in a northern hardwood forest. *Soil Biology and Biochemistry.*, 43, 795-803. <https://doi.org/10.1016/j.soilbio.2010.12.014>.

Caporaso, J.G. et al., 2010. QIIME allows analysis of high-throughput community sequencing data. *Nature Methods*, 7, 335-336. <https://doi.org/10.1038/nmeth.f.303>.

Caporaso, J.G. et al., 2011. Global patterns of 16S rRNA diversity at a depth of millions of sequences per sample. *Proceedings of the National Academy of Sciences of the United States of American Society for Microbiology*, 108 Suppl 1, 4516-4522. <https://doi.org/10.1073/pnas.1000080107>

Chen, F.S., Zeng, D.H., Zhou, B., Singh, A.N., Fan, Z.P., 2006. Seasonal variation in soil nitrogen availability under Mongolian pine plantations at the Keerqin Sand Lands, China. *Journal of Arid Environments.*, 67, 226-239. <https://doi.org/10.1016/j.jaridenv.2006.02.017>.

Chen, J. et al., 2015a. Impact of soil composition and electrochemistry on corrosion of rock-cut slope nets along railway lines in China. *Scientific Reports.*, 5, 14939. <https://doi.org/10.1038/srep14939>.

Chen, Q.L. et al., 2020. Rare microbial taxa as the major drivers of ecosystem multifunctionality in long-term fertilized soils. *Soil Biology and Biochemistry*, 141, 107686. <https://doi.org/107610.101016/j.soilbio.102019.107686>.

Chen, Y.P. et al., 2015b. Balancing green and grain trade. *Nature Geoscience*, 8, 739-741. <https://doi.org/10.1038/ngeo2544>.

Corstanje, R., Reddy, K.R., Prenger, J.P., Newman, S., Ogram, A.V., 2007. Soil microbial eco-physiological response to nutrient enrichment in a sub-tropical wetland. *Ecological Indicators*, 7, 277-289. <https://doi.org/210.1016/j.ecolind.2006.1002.1002>.

D'Odorico, P., Bhattachan, A., Davis, K.F., Ravi, S., Runyan, C.W., 2013. Global desertification: Drivers and feedbacks. *Advances in Water Resources*, 51, 326-344. <https://doi.org/10.1016/j.advwatres.2012.01.013>.

480 Delgado-Baquerizo, M. et al., 2018. A global atlas of the dominant bacteria found in soil. *Science*,  
481 359, 320–325. <https://doi.org/310.1126/science.aap9516>.

482 Delgado-Baquerizo, M. et al., 2016. Microbial diversity drives multifunctionality in terrestrial  
483 ecosystems. *Nature Communications*, 7, 10541. <https://doi.org/10.1038/ncomms10541>.

484 Edgar, R.C., 2013. UPARSE: highly accurate OTU sequences from microbial amplicon reads.  
485 *Nature Methods*, 10, 996-998. <https://doi.org/10.1038/nmeth.2604>.

486 Falkowski, P.G., Fenchel, T., Delong, E.F., 2008. The microbial engines that drive Earth's  
487 biogeochemical cycles. *Science*, 320, 1034-1039. <https://doi.org/10.1126/science.1153213>.

488 Ferrenberg, S., Faist, A.M., Howell, A., Reed, S.C., 2018. Biocrusts enhance soil fertility and  
489 *Bromus tectorum* growth, and interact with warming to influence germination. *Plant and*  
490 *Soil.*, 429, 77-90. <https://doi.org/10.1007/s11104-017-3525-1>.

491 Fierer, N., Jackson, R.B., 2006. From the cover: The diversity and biogeography of soil bacterial  
492 communities. *Proceedings of the National Academy of Sciences of the United States of*  
493 *America*, 103, <https://doi.org/10.1073/pnas.0507535103>.

494 Fortmann-Roe, S., 2013. Accurate, Adaptable, and Accessible Error Metrics for Predictive. R  
495 package version 0.9.2.

496 Gao, L.Q. et al., 2017. Biological soil crusts decrease erodibility by modifying inherent soil  
497 properties on the Loess Plateau, China. *Soil Biology and Biochemistry.*, 105, 49-58.  
498 <http://dx.doi.org/10.1016/j.soilbio.2016.11.009>.

499 Garrity, G.M., Holt, J.G., 2001. Phylum BVIII. Nitrospirae phy. nov.. In: Boone D.R., Castenholz  
500 R.W., Garrity G.M. (eds) *Bergey's Manual® of Systematic Bacteriology*. Springer, New  
501 York, NY., 451-464. [https://doi.org/10.1007/978-0-387-21609-6\\_25](https://doi.org/10.1007/978-0-387-21609-6_25).

502 Hartmann, M., Frey, B., Mayer, J., Mader, P., Widmer, F., 2015. Distinct soil microbial diversity  
503 under long-term organic and conventional farming. *The ISME Journal*, 9, 1177-1194.  
504 <https://doi.org/10.1038/ismej.2014.210>.

505 Hortal, S. et al., 2013. Soil microbial community under a nurse-plant species changes in  
506 composition, biomass and activity as the nurse grows. *Soil Biology and Biochemistry*, 64,  
507 139-146. <http://dx.doi.org/10.1016/j.soilbio.2013.04.018>.

508 Jakobsen, A.M. et al., 2019. Bacterial community analysis for investigating bacterial transfer from  
509 tonsils to the pig carcass. *International Journal of Food Microbiology*, 295, 8-18.  
510 <https://doi.org/10.1016/j.ijfoodmicro.2019.02.003>.

511 Jia, C. et al., 2018. Formation of litter crusts and its multifunctional ecological effects in a desert  
512 ecosystem. *Ecosphere*, 9, e02196. <https://doi.org/10.1002/ecs2.02196>.

513 Jia, F.F., Lu, R.j., Gao, S.Y., Li, J.F., Liu, X.K., 2015. Holocene aeolian activities in the southeastern  
514 Mu Us Desert, China. *Aeolian Research*, 19, 267-274.  
515 <http://dx.doi.org/10.1016/j.aeolia.2015.01.002>.

516 Jiao, S. et al., 2018a. Soil microbiomes with distinct assemblies through vertical soil profiles drive  
517 the cycling of multiple nutrients in reforested ecosystems. *Microbiome*, 6, 146.  
518 <https://doi.org/10.1186/s40168-018-04526-40160>.

519 Jiao, S. et al., 2018b. Plant growth and oil contamination alter the diversity and composition of  
520 bacterial communities in agricultural soils across China. *Land Degradation and*  
521 *Development*, 29, 1660-1671. <https://doi.org/10.1002/ldr.2932>.

522 Jiao, S. et al., 2019. Temporal dynamics of soil bacterial communities and multifunctionality are  
523 more sensitive to introduced plants than to microbial additions in a multicontaminated soil.  
524 *Land Degradation and Development*, 30, 852-865. <https://doi.org/10.1002/ldr.3272>.

Jing, X. et al., 2015. The links between ecosystem multifunctionality and above- and belowground biodiversity are mediated by climate. *Nature Communications*, 6, 8159. <https://doi.org/8110.1038/ncomms9159>.

Karimi, B. et al., 2018. Biogeography of soil bacteria and archaea across France. *Science Advances*, 4, eaat1808. <https://doi.org/1810.1126/sciadv.aat1808>

Kirby, R., 2005. Actinomycetes and Lignin Degradation. *Advances in Applied Microbiology*, 58, 125-168. [https://doi.org/10.1016/S0065-2164\(05\)58004-3](https://doi.org/10.1016/S0065-2164(05)58004-3).

Lawson, P.A., Llop-Perez, p., Hutson, R.A., Hippe, H., Collins, M.D., 1993. Towards a phylogeny of the clostridia based on 16S rRNA sequences. *FEMS Microbiolog*, 87-92. <https://doi.org/10.1111/j.1574-6968.1993.tb06493.x>.

Leloup, J. et al., 2018. Unravelling the effects of plant species diversity and aboveground litter input on soil bacterial communities. *Geoderma*, 317, 1-7. <https://doi.org/10.1016/j.geoderma.2017.12.018>.

Li, L.Y. et al., 2019. An in vitro model maintaining taxon-specific functional activities of the gut microbiome. *Nature Communications*, 10, 1-11. <https://doi.org/10.1101/616656>.

Li, X.R., Jia, R.L., Chen, Y.W., Huang, L., Zhang, P., 2011. Association of ant nests with successional stages of biological soil crusts in the Tengger Desert, Northern China. *Applied Soil Ecology*, 47, 59-66. <https://doi.org/10.1016/j.apsoil.2010.10.010>.

Li, Y.Q. et al., 2012. Mongolian pine plantations enhance soil physico-chemical properties and carbon and nitrogen capacities in semi-arid degraded sandy land in China. *Applied Soil Ecology*, 56, 1-9. <https://doi.org/10.1016/j.apsoil.2012.1001.1007>.

Liu, X.K. et al., 2018. Evolution of Peatlands in the Mu Us Desert, northern China, since the last deglaciation. *Journal of Geophysical Research: Earth Surface*, 123, 252-261. <https://doi.org/10.1002/2017jf004413>.

Liu, Y. et al., 2019a. Temporal and spatial succession and dynamics of soil fungal communities in restored grassland on the Loess Plateau in China. *Land Degradation and Development*, 30, 1273-1287. <https://doi.org/10.1002/ldr.3289>.

Liu, Y., Cui, Z., Huang, Z., Miao, H.T., Wu, G.L., 2019b. The influence of litter crusts on soil properties and hydrological processes in a sandy ecosystem. *Hydrology and Earth System Sciences*, 23, 2481-2490. <https://doi.org/10.5194/hess-23-2481-2019>.

Lozano, Y.M., Hortal, S., Armas, C., Pugnaire, F.I., 2014. Interactions among soil, plants, and microorganisms drive secondary succession in a dry environment. *Soil Biology and Biochemistry*, 78, 298-306. <https://doi.org/10.1016/j.soilbio.2014.08.007>.

Ma, B. et al., 2016. Geographic patterns of co-occurrence network topological features for soil microbiota at continental scale in eastern China. *The ISME Journal*, 10, 1891-1901. <https://doi.org/10.1038/ismej.2015.261>.

Marasco, R. et al., 2018. Rhizosheath microbial community assembly of sympatric desert spargrasses is independent of the plant host. *Microbiome*, 6, 215. <https://doi.org/10.1186/s40168-018-0597-y>.

Mei, R., Narihiro, T., Nobu, M.K., Kuroda, K., Liu, W.T., 2016. Evaluating digestion efficiency in full-scale anaerobic digesters by identifying active microbial populations through the lens of microbial activity. *Scientific Reports*, 6, 34090. <https://doi.org/10.1038/srep34090>.

Nelson, D.W., Sommers, L.E., 1982. Total Carbon, Organic Carbon, and Organic Matter. In Page A L. *Methods of Soil Analysis. Part 2. Chemical and Microbiological Properties*. American Society of Agronomy and Soil Science Society of America, Madison, 9, 539-577.



Oksanen, J. et al., 2013. Package 'vegan'. Community Ecology Package. 2, 1-295.  
<https://CRAN.R-project.org/package=vegan>.

Olsen, S.R., Sommers, L.E., 1982. Phosphorus. In Page A L, Miller R H, Keeney D R. Methods of Soil Analysis. Part. 2. American Society of Agronomy and Soil Science Society of America, Madison, 403-430.

Page, A.L., Miller, R.H., Keeney, D.R., 1982. Methods of Soil Analysis. 2nd eds. Part 2. Chemical and Microbiological Properties. . American Society of Agronomy and Soil Science Society of American, Madison, 539-579.

Palleroni, N.J., Port, A.M., Chang, H.K., Zylstra, G.J., 2004. Hydrocarboniphaga effusa gen. nov., sp. nov., a novel member of the gamma-Proteobacteria active in alkane and aromatic hydrocarbon degradation. International Journal of Systematic and Evolutionary Microbiology, 54, 1203-1207. <https://doi.org/10.1099/ijs.0.03016-0>.

Pester, M., Bittner, N., Deevong, P., Wagner, M., Loy, A., 2010. A 'rare biosphere' microorganism contributes to sulfate reduction in a peatland. The ISME Journal, 4, 1591-1602. <https://doi.org/10.1038/ismej.2010.75>.

Qian, X. et al., 2018. Shifts in community composition and co-occurrence patterns of phyllosphere fungi inhabiting Mussaenda shikokiana along an elevation gradient. PeerJ, 6, e5767. <https://doi.org/10.7717/peerj.5767>.

Ravi, S., D.Breshears, D., E.Huxman, T., D'Odorico, P., 2010. Land degradation in drylands: Interactions among hydrologic–aeolian erosion and vegetation dynamics. Geomorphology, 116, 236-245. <https://doi.org/10.1016/j.geomorph.2009.1011.1023>.

Reed, S.C., Delgado-Baquerizo, M., Ferrenberg, S., 2019. Biocrust science and global change. New Phytologist, 3, 223. <https://doi.org/10.1111/nph.15992>

Shi, S.H. et al., 2019. Response of microbial communities and enzyme activities to amendments in saline-alkaline soils. Applied Soil Ecology, 135, 16-24. <https://doi.org/10.1016/j.apsoil.2018.11.003>.

Sivakumar, M.V.K., Das, H.P., Brunini, O., 2005. Impacts of present and future climate variability and change on agriculture and forestry in the arid and semi-arid tropics. Climatic Change., 70, 31-72. <https://doi.org/10.1007/s10584-005-5937-9>.

Slezak, R., Grzelak, J., Krzystek, L., Ledakowicz, S., 2017. The effect of initial organic load of the kitchen waste on the production of VFA and H<sub>2</sub> in dark fermentation. Waste Manag, 68, 610-617. <https://doi.org/10.1016/j.wasman.2017.06.024>.

State Forestry Administration, P.R.C., 2015. The desertification and sandification state of China.

Suela Silva, M., Naves Sales, A., Teixeira Magalhaes-Guedes, K., Ribeiro Dias, D., Schwan, R.F., 2013. Brazilian cerrado soil Actinobacteria ecology. BioMed Research International, 2013, 503805. <http://dx.doi.org/10.1155/2013/503805>.

Tabatabai, M.A., 1994. Soil Enzymes. In R. W. Weaver, J. S. Angle, & P. S. Botttomley (Eds.), Methods of Soil Analysis: Microbiological and Biochemical Properties. Soil Science Society of America, Madison, 775-833. <https://doi.org/10.2136/sssabookser5.2.c37>

Taylor, J.P., Wilson, B., Mills, M.S., Burns, R.G., 2002. Comparison of microbial numbers and enzymatic activities in surface soils and subsoils using various techniques. Soil Biology and Biochemistry, 34, 387-401. [https://doi.org/10.1016/S0038-0717\(01\)00199-7](https://doi.org/10.1016/S0038-0717(01)00199-7).

Torres-Cruz, T.J. et al., 2018. Species-specific nitrogenase activity in lichen-dominated biological soil crusts from the Colorado Plateau, USA. Plant and Soil, 429, 113-125. <https://doi.org/10.1007/s11104-018-3580-2>.



615 Uchida, E., Xu, J., Rozelle, S., 2005. Grain for green: cost-effectiveness and sustainability of  
616 China's conservation set-aside program. *Land Economics*, 81, 247–264.  
617 <https://doi.org/10.3368/le.81.2.247>.

618 Veyisoglu, A. et al., 2016. *Microvirga makkahensis* sp. nov., and *Microvirga arabica* sp. nov.,  
619 isolated from sandy arid soil. *Antonie Van Leeuwenhoek*, 109, 287–296.  
620 <https://doi.org/10.1007/s10482-015-0631-z>.

621 Wagg, C., Bender, S.F., Widmer, F., van der Heijden, M.G., 2014. Soil biodiversity and soil  
622 community composition determine ecosystem multifunctionality. *Proceedings of the*  
623 *National Academy of Sciences of the United States of American Society for Microbiology*,  
624 111, 5266–5270. <https://doi.org/10.1073/pnas.1320054111>.

625 Wagner, M.R. et al., 2016. Host genotype and age shape the leaf and root microbiomes of a wild  
626 perennial plant. *Nature Communications*, 7, 12151. <https://doi.org/10.1038/ncomms12151>.

627 Wang, C. et al., 2018. Bacterial communities and their predicted functions explain the sediment  
628 nitrogen changes along with submerged macrophyte restoration. *Microbial Ecology*, 76,  
629 625–636. <https://doi.org/10.1007/s00248-018-1166-4>.

630 Wang, X.M. et al., 2017. Key driving forces of desertification in the Mu Us Desert, China.  
631 *Scientific Reports*, 7, 3933. <https://doi.org/10.1038/s41598-017-04363-8>.

632 Xu, M.P. et al., 2020. Dynamics of bacterial community in litter and soil along a chronosequence of  
633 *Robinia pseudoacacia* plantations. *Science of The Total Environment*, 703, 135613.  
634 <https://doi.org/10.1016/j.scitotenv.2019.135613>.

635 Zeng, D.H., Hu, Y.L., Chang, S.X., Fan, Z.P., 2008. Land cover change effects on soil chemical and  
636 biological properties after planting Mongolian pine (*Pinus sylvestris* var. *mongolica*) in  
637 sandy lands in Keerqin, northeastern China. *Plant and Soil*, 317, 121–133.  
638 <https://doi.org/10.1007/s11104-008-9793-z>.

639 Zhang, B.G., Zhang, J., Liu, Y., Shi, P., Wei, G.H., 2018. Co-occurrence patterns of soybean  
640 rhizosphere microbiome at a continental scale. *Soil Biology and Biochemistry*, 118, 178–186.  
641 <https://doi.org/10.1016/j.soilbio.2017.12.011>.

642 Zhuang, X.L., Han, Z., Bai, Z.H., Zhuang, G.Q., Shim, H., 2010. Progress in decontamination by  
643 halophilic microorganisms in saline wastewater and soil. *Environmental Pollution*, 158,  
644 1119–1126. <https://doi.org/10.1016/j.envpol.2010.01.007>.

645

646 **Figure legends**

647

648 **Figure 1.** Variation in soil properties and enzyme activities (n=9) among three soil groups (sandy  
649 land, biocrusts, and litter crusts) and three soil layers (0-2 cm, 2-5 cm, and 5-10 cm). Different  
650 lowercase letters indicate significant difference among three soil group in the same soil layer ( $p <$   
651 0.05), and different uppercase letters indicate significant difference among three soil layers in the  
652 same soil group ( $p < 0.05$ ). Error bars indicate standard deviation.

653

654 **Figure 2.** Variation in the microbial communities among three sand groups (sandy land, biocrusts,  
655 and litter crusts). (A) Relative abundances of the microbial taxa annotated (>1% of total community)  
656 at the phylum level; Difference in OTU richness (B), Chao1 index (C), Shannon index (D) of the  
657 microbial community (n=9) among three soil groups (sandy land, biocrusts, and litter crusts) and  
658 three soil layers (0-2 cm, 2-5 cm, and 5-10 cm). Difference in OTU richness (E), Chao1 index (F),  
659 Shannon index (H) of the microbial community (n=27) among three soil groups. Color of blue, red,  
660 green represent sandy land, biocrusts, and litter crusts, respectively. Different lowercase letters  
661 indicate significant difference among three soil group in the same soil layer ( $p < 0.05$ ), and different  
662 uppercase letters indicate significant difference among three soil layers in the same soil group ( $p <$   
663 0.05). Error bars indicate standard deviation.

664 **Figure 3.** Redundancy analysis (RDA) for identifying the influence of soil nutrients and enzyme  
665 activities on bacterial community composition at the phylum level (A) or the genus level (B) .  
666 Dashed ellipses represent nine treatments; Arrows represent the soil variables associated with  
667 bacterial community composition.

668

669 **Figure 4.**

670 Co-occurrence networks of soil bacterial communities in the sandy land, biocrusts, and litter crusts.  
671 The color of nodes represent bacterial genera and soil variables (red nodes represent soil variables,  
672 the other color nodes represent bacterial genera). (A) The correlations between the soil variables  
673 and bacterial taxa. (B) The correlations among the bacterial taxa themselves. Red edges represent  
674 positive correlation, blue edges represent the negative correlation.  
675 SOM, sand organic matter; TK, total potassium; AK, available potassium; TP, total phosphorus; AP, available  
676 phosphorus; TN, total nitrogen; UA, urease activity; GB,  $\beta$ -glucosidase activity; DHA, dehydrogenase activity;  
677 ALP, alkaline phosphatase activity; ACP, acid phosphatase activity.

678

679 **Figure 5.** Random forest (RF) shows all annotated microbial drivers at the phyla level for sand  
680 nutrient cycling in sandy land, biocrusts, and litter crusts, respectively. MSE is the mean square  
681 error. MSE% values represent the importance of these predictors. Higher MSE% values mean more  
682 important predictors. The significance of the model was estimated by the R package “A3”. \*,  $P <$   
683 0.05; \*\*,  $P < 0.01$ .

684

685 **Figure 6.** Random forest (RF) shows all potential drivers (MSE% values  $> 5\%$ ) of the phyla  
686 Actinobacteria, Planctomycetes, Chloroflexi, Armatimonadetes, Deinococcus-Thermus, Nitrospirae  
687 at the genus level for sand nutrient cycling in the sandy land, biocrusts and litter crusts, respectively.  
688 Color of orange, red and green represent the sandy land, biocrusts and litter crusts, respectively.  
689 MSE is the mean square error. MSE% values represent the importance of these predictors. Higher  
690 MSE% values mean more important predictors. The significance of the model was estimated by the  
691 R package “A3”. \*,  $P < 0.05$ ; \*\*,  $P < 0.01$ .

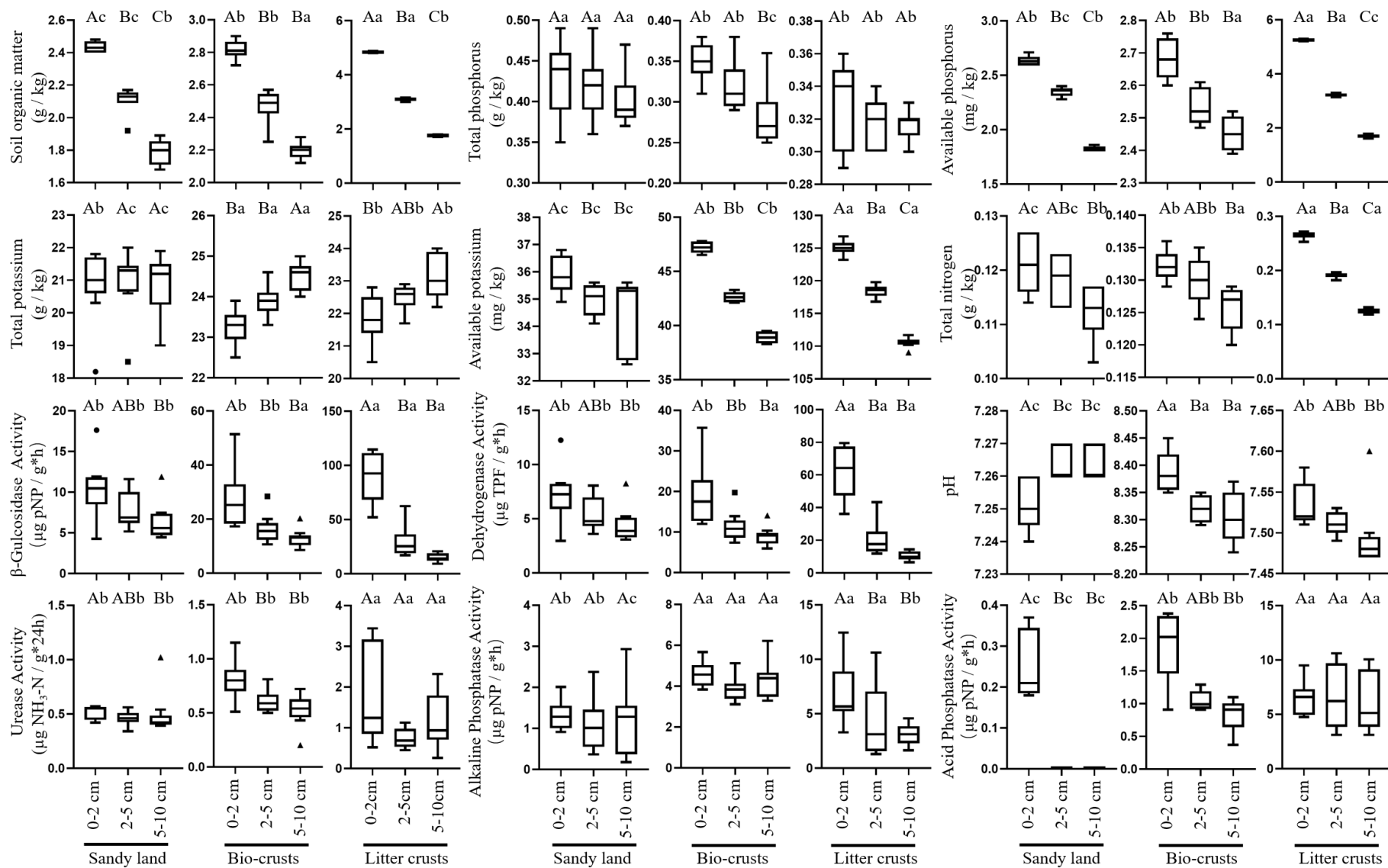


Figure 1

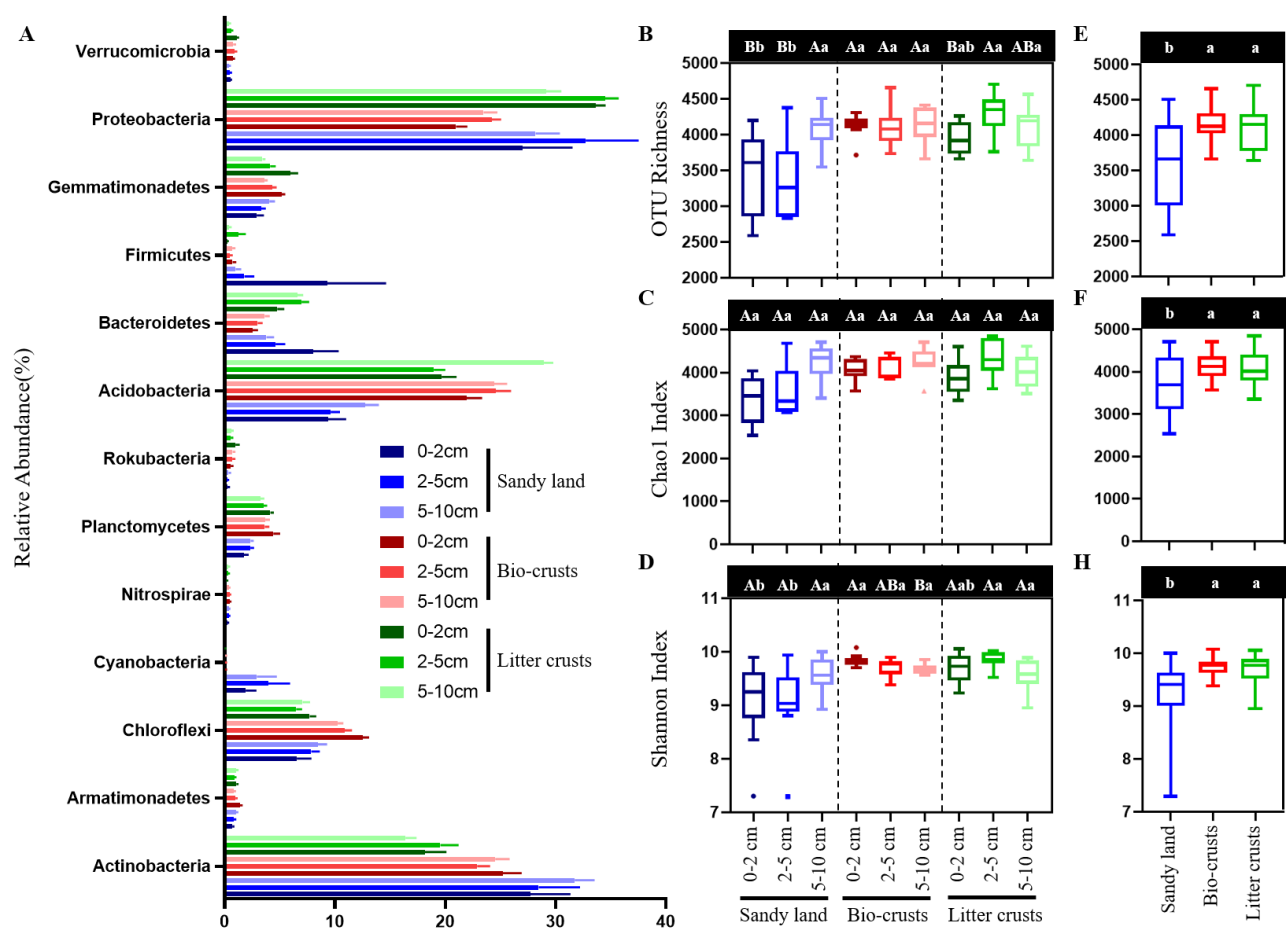


Figure 2

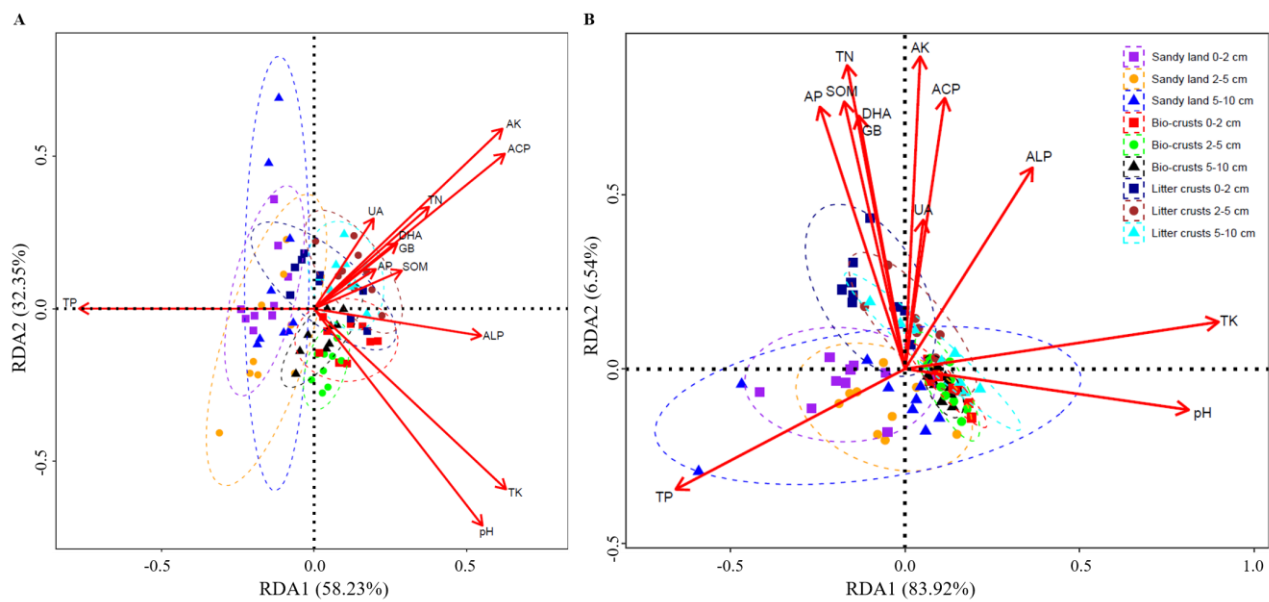


Figure 3

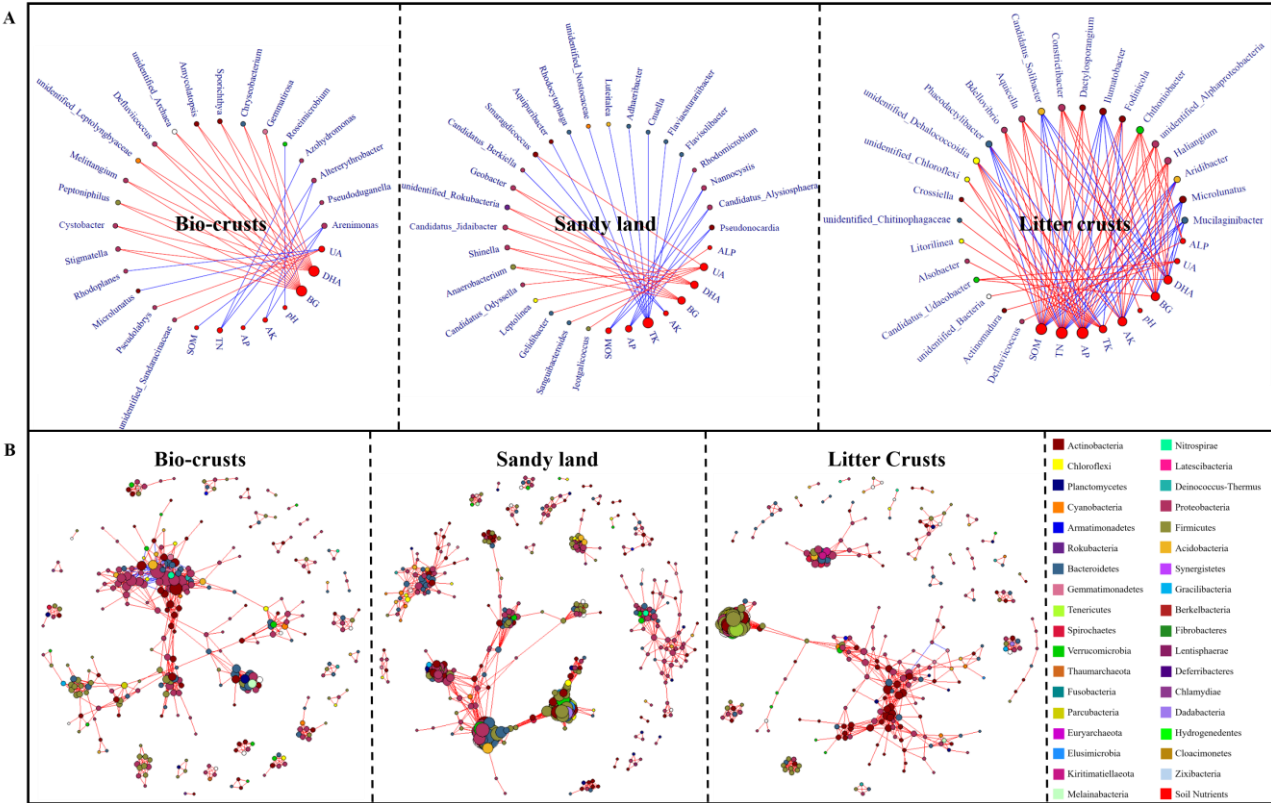


Figure 4

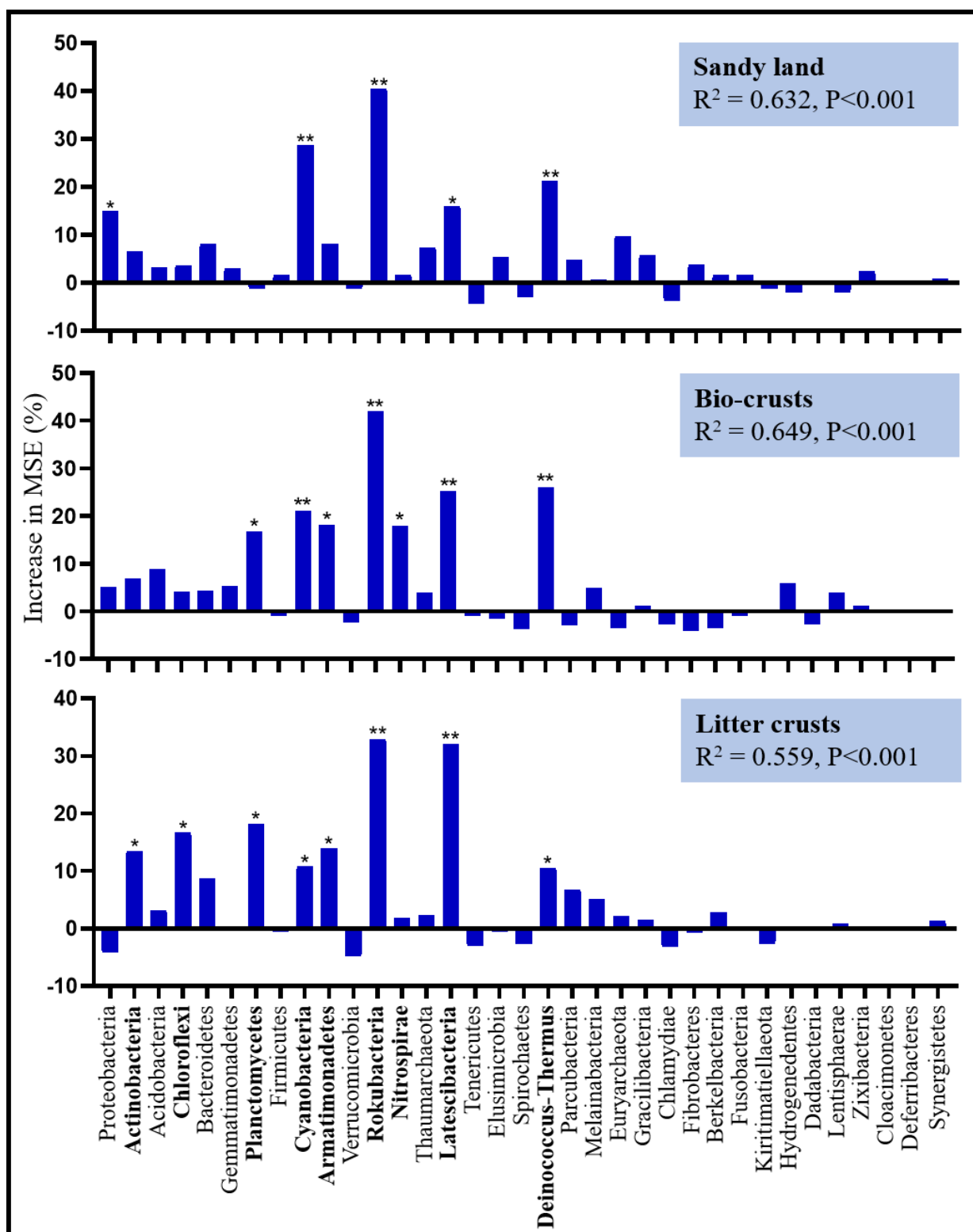


Figure 5



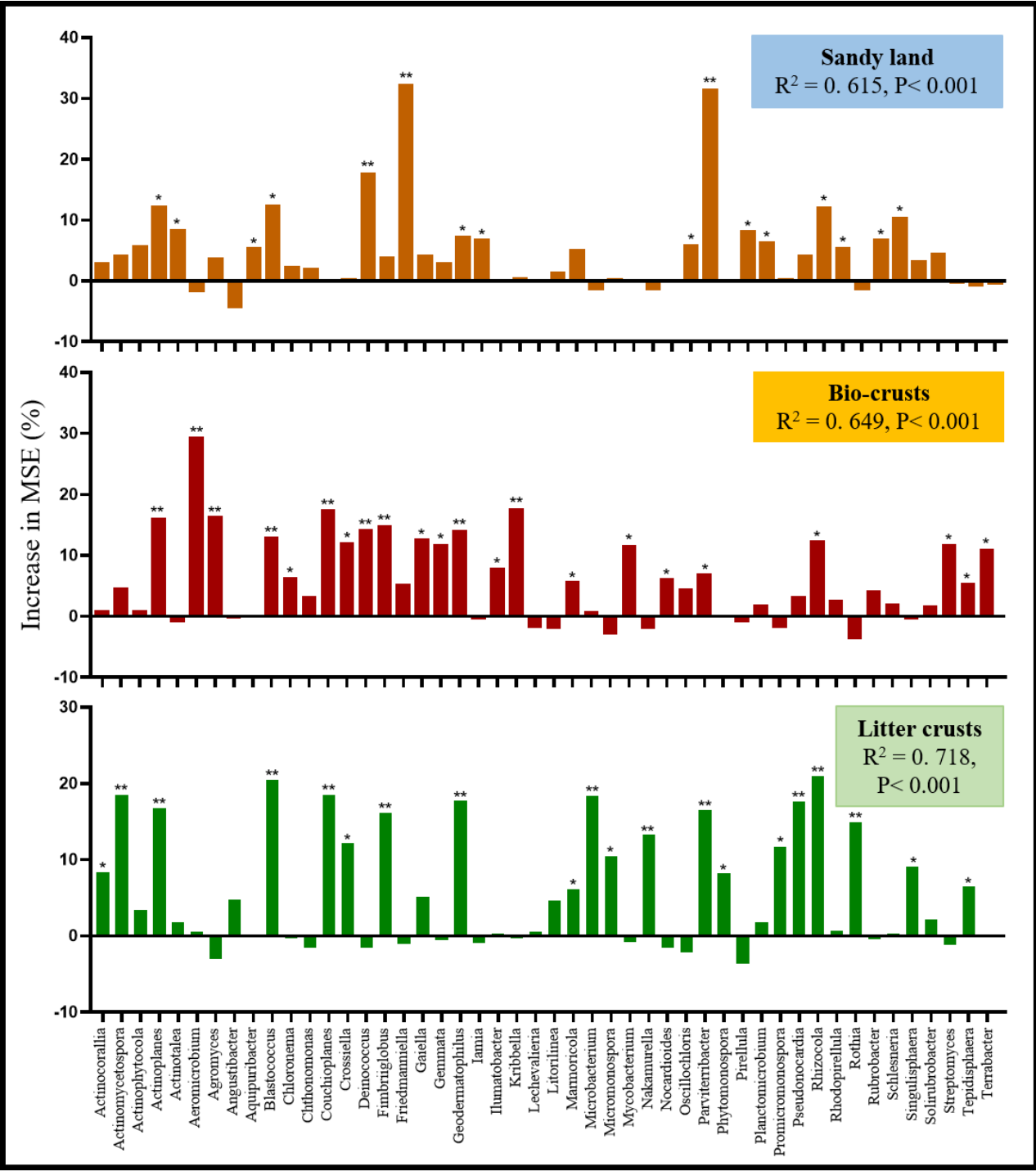


Figure 6